

Jonathan S. Golan

The Linear Algebra a Beginning Graduate Student Ought to Know

Second Edition



Springer

THE LINEAR ALGEBRA A BEGINNING GRADUATE STUDENT
OUGHT TO KNOW

The Linear Algebra a Beginning Graduate Student Ought to Know

Second Edition

by

JONATHAN S. GOLAN

University of Haifa, Israel



A C.I.P. Catalogue record for this book is available from the Library of Congress.

ISBN-10 1-4020-5494-7 (PB)
ISBN-13 978-1-4020-5494-5 (PB)
ISBN-10 1-4020-5495-5 (e-book)
ISBN-13 978-1-4020-5495-2 (e-book)

Published by Springer,
P.O. Box 17, 3300 AA Dordrecht, The Netherlands.

www.springer.com

Printed on acid-free paper

All Rights Reserved
© 2007 Springer

No part of this work may be reproduced, stored in a retrieval system, or transmitted in any form or by any means, electronic, mechanical, photocopying, microfilming, recording or otherwise, without written permission from the Publisher, with the exception of any material supplied specifically for the purpose of being entered and executed on a computer system, for exclusive use by the purchaser of the work.

To my grandsons: Shachar, Eitan, and Sarel

תן לחכם ויחכם עוד

(משלי, פרק ט"ו)

Contents

1	Notation and terminology	1
2	Fields	5
3	Vector spaces over a field	17
4	Algebras over a field	33
5	Linear independence and dimension	49
6	Linear transformations	79
7	The endomorphism algebra of a vector space	99
8	Representation of linear transformations by matrices	117
9	The algebra of square matrices	131
10	Systems of linear equations	169
11	Determinants	199
12	Eigenvalues and eigenvectors	229
13	Krylov subspaces	267

14 The dual space	285
15 Inner product spaces	299
16 Orthogonality	325
17 Selfadjoint Endomorphisms	349
18 Unitary and Normal endomorphisms	369
19 Moore-Penrose pseudoinverses	389
20 Bilinear transformations and forms	399
A Summary of Notation	423
Index	427

For whom is this book written?

*Crow's Law: Do not think what you want to think until you know what you ought to know.*¹

Linear algebra is a living, active branch of mathematical research which is central to almost all other areas of mathematics and which has important applications in all branches of the physical and social sciences and in engineering. However, in recent years the content of linear algebra courses required to complete an undergraduate degree in mathematics – and even more so in other areas – at all but the most dedicated universities, has been depleted to the extent that it falls far short of what is in fact needed for graduate study and research or for real-world application. This is true not only in the areas of theoretical work but also in the areas of computational matrix theory, which are becoming more and more important to the working researcher as personal computers become a common and powerful tool. Students are not only less able to formulate or even follow mathematical proofs, they are also less able to understand the underlying mathematics of the numerical algorithms they must use. The resulting knowledge gap has led to frustration and recrimination on the part of both students and faculty alike, with each silently – and sometimes not so silently – blaming the other for the resulting state of affairs. This book is written with the intention of bridging that gap. It was designed to be used in one or more of several possible ways:

- (1) As a self-study guide;
- (2) As a textbook for a course in advanced linear algebra, either at the upper-class undergraduate level or at the first-year graduate level; or
- (3) As a reference book.

It is also designed to be used to prepare for the linear algebra portion of prelim exams or PhD qualifying exams.

This volume is self-contained to the extent that it does not assume any previous knowledge of formal linear algebra, though the reader is assumed to have been exposed, at least informally, to some basic ideas or techniques, such as matrix manipulation and the solution of a small system of linear equations. It does, however, assume a seriousness of purpose, considerable

¹This law, attributed to John Crow of King's College, London, is quoted by R. V. Jones in his book *Most Secret War*.

motivation, and modicum of mathematical sophistication on the part of the reader.

The book also contains a large number of exercises, many of which are quite challenging, which I have come across or thought up in over thirty years of teaching. Many of these exercises have appeared in print before, in such journals as *American Mathematical Monthly*, *College Mathematics Journal*, *Mathematical Gazette*, or *Mathematics Magazine*, in various mathematics competitions or circulated problem collections, or even on the internet. Some were donated to me by colleagues and even students, and some originated in files of old exams at various universities which I have visited in the course of my career. Since, over the years, I did not keep track of their sources, all I can do is offer a collective acknowledgement to all those to whom it is due. Good problem formulators, like the God of the abbot of Cîteaux, know their own. Deliberately, difficult exercises are not marked with an asterisk or other symbol. Solving exercises is an integral part of learning mathematics and the reader is definitely expected to do so, especially when the book is used for self-study.

Solving a problem using theoretical mathematics is often very different from solving it computationally, and so strong emphasis is placed on the interplay of theoretical and computational results. Real-life implementation of theoretical results is perpetually plagued by errors: errors in modelling, errors in data acquisition and recording, and errors in the computational process itself due to roundoff and truncation. There are further constraints imposed by limitations in time and memory available for computation. Thus the most elegant theoretical solution to a problem may not lead to the most efficient or useful method of solution in practice. While no reference is made to particular computer software, the concurrent use of a personal computer equipped symbolic-manipulation software such as MAPLE, MATHEMATICA, MATLAB or MUPAD is definitely advised.

In order to show the “human face” of mathematics, the book also includes a large number of thumbnail photographs of researchers who have contributed to the development of the material presented in this volume.

Acknowledgements. Most of the first edition this book was written while the I was a visitor at the University of Iowa in Iowa City and at the University of California in Berkeley. I would like to thank both institutions for providing the facilities and, more importantly, the mathematical atmosphere which allowed me to concentrate on writing. This edition was extensively revised after I retired from teaching at the University of Haifa in April, 2004.

I have talked to many students and faculty members about my plans for this book and have obtained valuable insights from them. In particular, I would like to acknowledge the aid of the following colleagues and students who were kind enough to read the preliminary versions of this book and

offer their comments and corrections: Prof. Daniel Anderson (University of Iowa), Prof. Adi Ben-Israel (Rutgers University), Prof. Robert Cacioppo (Truman State University), Prof. Joseph Felsenstein (University of Washington), Prof. Ryan Skip Garibaldi (Emory University), Mr. George Kirkup (University of California, Berkeley), Prof. Earl Taft (Rutgers University), Mr. Gil Varnik (University of Haifa).

Photo credits. The photograph of Dr. Shmuel Winograd is used with the kind permission of the Department of Computer Science of the City University of Hong Kong. The photographs of Prof. Ben-Israel, Prof. Blass, Prof. Kublanovskaya, and Prof. Strassen are used with their respective kind permissions. The photograph of Prof. Greville is used with the kind permission of Mrs. Greville. The photograph of Prof. Rutishauser is used with the kind permission of Prof. Walter Gander. The photograph of Prof. V. N. Faddeeva is used with the kind permission of Dr. Vera Simonova. The photograph of Prof. Zorn is used with the kind permission of his son, Jens Zorn. The photograph of J. W. Givens was taken from a group photograph of the participants at the 1964 Gatlinburg Conference on Numerical Algebra. All other photographs are taken from the MacTutor History of Mathematics Archive website (<http://www-history.mcs.st-andrews.ac.uk/history/index.html>), the portrait gallery of mathematicians at the Trucsmatheux website (<http://trucsmaths.free.fr/>), or similar websites. To the best knowledge of the managers of those sites, and to the best of my knowledge, they are in the public domain.

1

Notation and terminology

Sets will be denoted by braces, $\{ \}$, between which we will either enumerate the elements of the set or give a rule for determining whether something is an element of the set or not, as in $\{x \mid p(x)\}$, which is read “the set of all x such that $p(x)$ ”. If a is an element of a set A we write $a \in A$; if it is not an element of A , we write $a \notin A$. When one enumerates the elements of a set, the order is not important. Thus $\{1, 2, 3, 4\}$ and $\{4, 1, 3, 2\}$ both denote the same set. However, we often do wish to impose an order on sets the elements of which we enumerate. Rather than introduce new and cumbersome notation to handle this, we will make the convention that when we enumerate the elements of a finite or countably-infinite set, we will assume an implied order, reading from left to right. Thus, the implied order on the set $\{1, 2, 3, \dots\}$ is indeed the usual one. The empty set, namely the set having no elements, is denoted by \emptyset . Sometimes we will use the word “collection” as a synonym for “set”, generally to avoid talking about “sets of sets”.

A finite or countably-infinite selection of elements of a set A is a **list**. Members of a list are assumed to be in a definite order, given by their indices or by the implied order of reading from left to right. Lists are usually written without brackets: a_1, \dots, a_n , though, in certain contexts, it will be more convenient to write them as ordered n -tuples (a_1, \dots, a_n) . Note that the elements of a list need not be distinct: $3, 1, 4, 1, 5, 9$ is a list of six positive integers, the second and fourth elements of which are equal to 1. A countably-infinite list of elements of a set A is also often called

a **sequence** of elements of A . The set of all distinct members of a list is called the **underlying subset** of the list.

If A and B are sets, then their **union** $A \cup B$ is the set of all elements that belong to either A or B , and their **intersection** $A \cap B$ is the set of all elements belonging both to A and to B . More generally, if $\{A_i \mid i \in \Omega\}$ is a (possibly-infinite) collection of sets, then $\bigcup_{i \in \Omega} A_i$ is the set of all elements that belong to at least one of the A_i and $\bigcap_{i \in \Omega} A_i$ is the set of all elements that belong to all of the A_i . If A and B are sets, then the **difference set** $A \setminus B$ is the set of all elements of A which do not belong to B .

A **function** f from a nonempty set A to a nonempty set B is a rule which assigns to each element a of A a unique element $f(a)$ of B . The set A is called the **domain** of the function and the set B is called the **range** of the function. To denote that f is a function from A to B , we write $f : A \rightarrow B$. To denote that an element b of B is assigned to an element a of A by f , we write $f : a \mapsto b$. (Note the different form of the arrow!) This notation is particularly helpful in the case that the function f is defined by a formula. Thus, for example, if f is a function from the set of integers to the set of integers defined by $f : a \mapsto a^3$, then we know that f assigns to each integer its cube. The set of all functions from a nonempty set A to a nonempty set B is denoted by B^A . If $f \in B^A$ and if A' is a nonempty subset of A , then the **restriction** of f to A' is the function $f' : A' \rightarrow B$ defined by $f' : a' \mapsto f(a')$ for all $a' \in A'$.

Functions f and g in B^A are **equal** if and only if $f(a) = g(a)$ for all $a \in A$. In this case we write $f = g$. A function $f \in B^A$ is **monic** if and only if it assigns different elements of B to different elements of A , i.e. if and only if $f(a_1) \neq f(a_2)$ whenever $a_1 \neq a_2$ in A . A function $f \in B^A$ is **epic** if and only if every element of B is assigned by f to some element of A . A function which is both monic and epic is **bijective**. A bijective function from a set A to a set B determines a bijective correspondence between the elements of A and the elements of B . If $f : A \rightarrow B$ is a bijective function, then we can define the **inverse function** $f^{-1} : B \rightarrow A$ defined by the condition that $f^{-1}(b) = a$ if and only if $f(a) = b$. This inverse function is also bijective. A bijective function from a set A to itself is a **permutation** of A . Note that there is always at least one permutation of any nonempty set A , namely the identity function $a \mapsto a$.

The **cartesian product** $A_1 \times A_2$ of nonempty sets A_1 and A_2 is the set of all ordered pairs (a_1, a_2) , where $a_1 \in A_1$ and $a_2 \in A_2$. More generally, if A_1, \dots, A_n is a list of nonempty sets, then $A_1 \times \dots \times A_n$ is the set of all ordered n -tuples (a_1, \dots, a_n) satisfying the condition that $a_i \in A_i$ for each $1 \leq i \leq n$. Note that each ordered n -tuple (a_1, \dots, a_n)

uniquely defines a function $f : \{1, \dots, n\} \rightarrow \cup_{i=1}^n A_i$ given by $f : i \mapsto a_i$ for each $1 \leq i \leq n$. Conversely, each function $f : \{1, \dots, n\} \rightarrow \cup_{i=1}^n A_i$ satisfying the condition that $f(i) \in A_i$ for $1 \leq i \leq n$, defines such an ordered n -tuple, namely $(f(1), \dots, f(n))$. This suggests a method for defining the cartesian product of an arbitrary collection of nonempty sets. If $\{A_i \mid i \in \Omega\}$ is an arbitrary collection of nonempty sets, then the set $\prod_{i \in \Omega} A_i$ is defined to be the set of all those functions f from Ω to $\cup_{i \in \Omega} A_i$ satisfying the condition that $f(i) \in A_i$ for each $i \in \Omega$. The existence of such functions is guaranteed by a fundamental axiom of set theory, known as the **Axiom of Choice**. A certain amount of controversy surrounds this axiom, and there are mathematicians who prefer to make as little use of it as possible. However, we will need it constantly throughout this book, and so will always assume that it holds.

In the foregoing construction we did not assume that the sets A_i were necessarily distinct. Indeed, it may very well happen that there exists a set A such that $A_i = A$ for all $i \in \Omega$. In that case, we see that $\prod_{i \in \Omega} A_i$ is just A^Ω . If the set Ω is finite, say $\Omega = \{1, \dots, n\}$, then we write A^n instead of A^Ω . Thus, A^n is just the set of all ordered n -tuples (a_1, \dots, a_n) of elements of A .

We use the following standard notation for some common sets of numbers

\mathbb{N}	the set of all nonnegative integers
\mathbb{Z}	the set of all integers
\mathbb{Q}	the set of all rational numbers
\mathbb{R}	the set of all real numbers
\mathbb{C}	the set of all complex numbers

Other notation is introduced throughout the text, as is appropriate. See the Summary of Notation at the end of the book.

2

Fields

The way of mathematical thought is twofold: the mathematician first proceeds inductively from the particular to the general and then deductively from the general to the particular. Moreover, throughout its development, mathematics has shown two aspects – the conceptual and the computational – the symphonic interleaving of which forms one of the major aspects of the subject’s aesthetic.

Let us therefore begin with the first mathematical structure: numbers. By the Hellenistic times, mathematicians distinguished between two types of numbers: the **rational** numbers, namely those which could be written in the form $\frac{m}{n}$ for some integer m and some nonnegative integer n , and those numbers representing the geometric magnitude of segments of the line, which today we call **real** numbers and which, in decimal notation, are written in the form $m.k_1k_2k_3\dots$ where m is an integer and the k_i are digits. The fact that the set \mathbb{Q} of rational numbers is not equal to the set \mathbb{R} of real numbers was already noticed by the followers of the mathematician/mystic Pythagoras. On both sets of numbers we define operations of addition and multiplication which satisfy certain rules of manipulation. Isolating these rules as part of a formal system was a task first taken on in earnest by nineteenth-century British and German mathematicians. From their studies evolved the notion of a field, which will be basic to our considerations. However, since fields are not our primary object of study, we will

delve only minimally into this fascinating notion. A serious consideration of field theory must be deferred to an advanced course in abstract algebra.¹

A nonempty set F together with two functions $F \times F \rightarrow F$, respectively called **addition** (as usual, denoted by $+$) and **multiplication** (as usual, denoted by \cdot or by concatenation), is a **field** if the following conditions are satisfied:

(1) (**associativity of addition and multiplication**): $a + (b + c) = (a + b) + c$ and $a(bc) = (ab)c$ for all $a, b, c \in F$.

(2) (**commutativity of addition and multiplication**): $a + b = b + a$ and $ab = ba$ for all $a, b \in F$.

(3) (**distributivity of multiplication over addition**): $a(b + c) = ab + ac$ for all $a, b, c \in F$.

(4) (**existence of identity elements for addition and multiplication**): There exist distinct elements of F , which we will denote by 0 and 1 respectively, satisfying $a + 0 = a$ and $a1 = a$ for all $a \in F$.

(5) (**existence of additive inverses**): For each $a \in F$ there exists an element of F , which we will denote by $-a$, satisfying $a + (-a) = 0$.

(6) (**existence of multiplicative inverses**): For each $0 \neq a \in F$ there exists an element of F , which we will denote by a^{-1} , satisfying $a^{-1}a = 1$.

Note that we did not assume that the elements $-a$ and a^{-1} are unique, though we will soon prove that in fact they are. If a and b are elements of a field F , we will follow the usual conventions by writing $a - b$ instead of $a + (-b)$ and $\frac{a}{b}$ instead of ab^{-1} . Moreover, if $0 \neq a \in F$ and if n is a positive integer, then na denotes the sum $a + \dots + a$ (n summands) and a^n denotes the product $a \cdot \dots \cdot a$ (n factors). If n is a negative integer, then na denotes $(-n)(-a)$ and a^n denotes $(a^{-1})^{-n}$. Finally, if $n = 0$ then na denotes the field element 0 and a^n denotes the field element 1 . For $0 = a \in F$, we define $na = 0$ for all integers n and



The development of the abstract theory of fields is generally credited to the 19th-century German mathematician **Heinrich Weber**, based on earlier work by the German mathematicians **Richard Dedekind** and **Leopold Kronecker**. Another 19th-century mathematician, the British **Augustus De Morgan**, was the first to isolate the importance of such properties as associativity, distributivity, and so forth. The final axioms of a field are due to the 20th-century German mathematician **Ernst Steinitz**.

$a^n = 0$ for all positive integers n . The symbol 0^k is not defined for $k \leq 0$.

As an immediate consequence of the associativity and commutativity of addition, we see that the sum of any list a_1, \dots, a_n of elements of a field F is the same, no matter in which order we add them. We can therefore unambiguously write $a_1 + \dots + a_n$. This sum is also often denoted by $\sum_{i=1}^n a_i$. Similarly, the product of these elements is the same, no matter in which order we multiply them. We can therefore unambiguously write $a_1 \cdot \dots \cdot a_n$. This product is also often denoted by $\prod_{i=1}^n a_i$. Also, a simple inductive argument shows that multiplication distributes over arbitrary sums: if $a \in F$ and b_1, \dots, b_n is a list of elements of F then $a(\sum_{i=1}^n b_i) = \sum_{i=1}^n ab_i$.

We easily see that \mathbb{Q} and \mathbb{R} , with the usual addition and multiplication, are fields.

A subset G of a field F is a **subfield** if and only if it contains 0 and 1, is closed under addition and multiplication, and contains the additive and multiplicative inverses of all of its nonzero elements. Thus, for example, \mathbb{Q} is a subfield of \mathbb{R} . The intersection of a collection of subfields of a field F is again a subfield of F .

We now want to look at several additional important examples of fields.

Example: Let $\mathbb{C} = \mathbb{R}^2$ and define operations of addition and multiplication on \mathbb{C} by setting $(a, b) + (c, d) = (a + c, b + d)$ and $(a, b) \cdot (c, d) = (ac - bd, ad + bc)$. These operations define the structure of a field on \mathbb{C} , in which the identity element for addition is $(0, 0)$, the identity element for multiplication is $(1, 0)$, the additive inverse of (a, b) is $(-a, -b)$, and

$$(a, b)^{-1} = \left(\frac{a}{a^2 + b^2}, \frac{-b}{a^2 + b^2} \right)$$

for all $(0, 0) \neq (a, b)$. This field is called the field of **complex numbers**. The set of all elements of \mathbb{C} of the form $(a, 0)$ forms a subfield of \mathbb{C} , which we normally identify with \mathbb{R} and therefore it is standard to consider \mathbb{R} as a subfield of \mathbb{C} . In particular, we write a instead of $(a, 0)$ for any real number a . The element $(0, 1)$ of \mathbb{C} is denoted by i . This element satisfies the condition that $i^2 = (-1, 0)$ and so it is often written as $\sqrt{-1}$. We also note that any element (a, b) of \mathbb{C} can be written as $(a, 0) + b(0, 1) = a + bi$, and, indeed, that is the way complex numbers are usually written and how we will denote them from now on. If $z = a + bi$, then a is the **real part** of z , which is often denoted by $\operatorname{Re}(z)$, while bi is the **imaginary part** of z , which is often denoted by $\operatorname{Im}(z)$. The

field of complex numbers is extremely important in mathematics. From a geometric point of view, if we identify \mathbb{R} with the set of points on the Euclidean line, as one does in analytic geometry, then it is natural to identify \mathbb{C} with the set of points in the Euclidean plane.²

If $z = a + bi \in \mathbb{C}$ then we denote the complex number $a - bi$, called the **complex conjugate** of z , by \bar{z} . It is easy to see³ that for all $z, z' \in \mathbb{C}$ we have $\overline{z + z'} = \bar{z} + \bar{z}'$, $\overline{-z} = -\bar{z}$, $\overline{zz'} = \bar{z} \cdot \bar{z}'$, $\overline{z^{-1}} = (\bar{z})^{-1}$, and $\overline{\bar{z}} = z$. The number $z\bar{z}$ equals $a^2 + b^2$, which is a nonnegative real number and so has a square root in \mathbb{R} , which we will denote by $|z|$. Note that $|z|$ is nonzero whenever $z \neq 0$. From a geometric point of view, this number is just the distance from the number z , considered as a point in the euclidean plane, to the origin, just as the usual absolute value $|a|$ of a real number a is the distance between a and 0 on the real line. It is easy to see that if y and z are complex numbers then $|yz| = |y| \cdot |z|$ and $|y + z| \leq |y| + |z|$. Moreover, if $z = a + bi$ then

$$z + \bar{z} = 2a \leq 2|a| = 2\sqrt{a^2} \leq 2\sqrt{a^2 + b^2} = 2|z|.$$

We also note, as a direct consequence of the definition, that $|z| = |\bar{z}|$ for every complex number z and so $z^{-1} = |z|^{-2}\bar{z}$ for all $0 \neq z \in \mathbb{C}$.

Example: The set \mathbb{Q}^2 is a subfield of the field \mathbb{C} defined above. However, it is also possible to define field structures on \mathbb{Q}^2 in other ways. Indeed, let $F = \mathbb{Q}^2$ and let p be a fixed prime integer. Define addition and multiplication on F by setting $(a, b) + (c, d) = (a + c, b + d)$ and $(a, b) \cdot (c, d) = (ac + bdp, ad + bc)$.

Again, one can check that F is indeed a field and that, again, the set of all elements of F of the form $(a, 0)$ is a subfield, which we will identify with \mathbb{Q} . Moreover, the additive inverse of $(a, b) \in F$ is $(-a, -b)$ and



The term “imaginary” was coined by the 17th-century French philosopher and mathematician **René Descartes**. The use of i to denote $\sqrt{-1}$ was introduced by 18th-century Swiss mathematician **Leonhard Euler**. The geometric representation of the complex numbers was first proposed at the end of the 18th century by the Norwegian surveyor **Caspar Wessel**, and later by the French accountant **Jean-Robert Argand**.

³When a mathematician says that something is “easy to see” or “trivial”, it means that you are expected to take out a pencil and paper and spend some time – often considerable – checking it out by yourself.

the multiplicative inverse of $(0, 0) \neq (a, b) \in F$ is

$$\left(\frac{a}{a^2 - pb^2}, \frac{-b}{a^2 - pb^2} \right).$$

(We note that $a^2 - pb^2$ is the product of the nonzero real numbers $a + b\sqrt{p}$ and $a - b\sqrt{p}$ and so is nonzero.) The element $(0, 1)$ of F satisfies $(0, 1)^2 = (p, 0)$ and so one usually denotes it by \sqrt{p} and, as before, any element of F can be written in the form $a + b\sqrt{p}$, where $a, b \in \mathbb{Q}$. The field F is usually denoted by $\mathbb{Q}(\sqrt{p})$. Since there are infinitely-many distinct prime integers, we see that there are infinitely-many ways of defining different field structures on $\mathbb{Q} \times \mathbb{Q}$, all having the same addition.

Example: Fields do not have to be infinite. Let p be a positive integer and let $\mathbb{Z}/(p) = \{0, 1, \dots, p-1\}$. For each nonnegative integer n , let us, for the purposes of this example, denote the remainder after dividing n by p as $[n]_p$. Thus we note that $[n]_p \in \mathbb{Z}/(p)$ for each nonnegative integer n and that $[i]_p = i$ for all $i \in \mathbb{Z}/(p)$. We now define operations on $\mathbb{Z}/(p)$ by setting $[n]_p + [k]_p = [n+k]_p$ and $[n]_p \cdot [k]_p = [nk]_p$. It is easy to check that if the integer p is prime then $\mathbb{Z}/(p)$, together with these two operations, is again a field, known as the **Galois**⁴ field of order p . This field is usually denoted by $GF(p)$. While Galois fields were first considered mathematical curiosities, they have since found important applications in coding theory, cryptography, and modeling of computer processes.

These are not the only possible finite fields. Indeed, it is possible to show that for each prime integer p and each positive integer n there exists an (essentially unique) field with p^n elements, usually denoted by $GF(p^n)$.

Example: Some important structures are “very nearly” fields. For example, let $\mathbb{R}_\infty = \mathbb{R} \cup \{\infty\}$, and define operations \boxplus and \boxtimes on \mathbb{R}_∞ by setting

$$a \boxplus b = \begin{cases} \min\{a, b\} & \text{if } a, b \in \mathbb{R} \\ b & \text{if } a = \infty \\ a & \text{if } b = \infty \end{cases}$$



4

The 19th-century French mathematical genius **Evariste Galois**, who died at the age of 21, was the first to consider such structures. The study of finite and infinite fields was unified in the 1890's by **Eliakim Hastings Moore**, the first American-born mathematician to achieve an international reputation.

and

$$a \boxplus b = \begin{cases} a + b & \text{if } a, b \in \mathbb{R} \\ \infty & \text{otherwise} \end{cases}.$$

This structure, called the **optimization algebra**, satisfies all of the conditions of a field *except* for the existence of additive inverses (such structures are known as **semifields**). As the name suggests, it has important applications in optimization theory and the analysis of discrete-event dynamical systems. There are several other semifields which have important applications and which have been extensively studied.

Another possibility of generalizing the notion of a field is to consider an algebraic structure which satisfies all of the conditions of a field *except* for the existence of multiplicative inverses, and to replace that condition by the condition that if $a, b \neq 0$ then $ab \neq 0$. Such structures are known as **integral domains**. The set \mathbb{Z} of all integers is the simplest example of an integral domain which is not a field. Algebras of polynomials over a field, which we will consider later, are also integral domains. In a course in abstract algebra, one proves that any integral domain can be embedded in a field.

In the field $GF(p)$ which we defined above, one can easily see that the sum $1 + \dots + 1$ (p summands) equals 0 . On the other hand, in the field \mathbb{Q} , the sum of any number of copies of 1 is always nonzero. This is an important distinction which we will need to take into account in dealing with structures over fields. We therefore define the **characteristic** of a field F to be equal to the smallest positive integer p such that $1 + \dots + 1$ (p summands) equals 0 – if such an integer p exists – and to be equal to 0 otherwise. We will not delve deeply into this concept, which is dealt with in courses on field theory, except to note that the characteristic of a field, if nonzero, always turns out to be a prime number.

In the definition of a field, we posited the existence of distinct identity elements for addition and multiplication, but did not claim that these elements were unique. It is, however, very easy to prove that fact.

(2.1) Proposition: Let F be a field.

- (1) If e is an element of F satisfying $e + a = a$ for all $a \in F$ then $e = 0$;
- (2) If u is an element of F satisfying $ua = a$ for all $a \in F$ then $u = 1$.

Proof: By definition we note that $e = e + 0 = 0$ and $u = u1 = 1$.
 \square

Similarly, we prove that additive and multiplicative inverses, when they exist, are unique. Indeed, we can prove a stronger result.

(2.2) Proposition: If a and b are elements of a field F then:
 (1) There exists a unique element c of F satisfying $a + c = b$.
 (2) If $a \neq 0$ then there exists a unique element d of F satisfying $ad = b$.

Proof: (1) Choose $c = b - a$. Then

$$\begin{aligned} a + c &= a + (b - a) = a + [b + (-a)] \\ &= a + [(-a) + b] = [a + (-a)] + b = 0 + b = b. \end{aligned}$$

Moreover, if $a + x = b$ then

$$\begin{aligned} x &= 0 + x = [(-a) + a] + x \\ &= (-a) + (a + x) = (-a) + b = b - a, \end{aligned}$$

proving uniqueness.

(2) Choose $d = a^{-1}b$. Then $ad = a(a^{-1}b) = (aa^{-1})b = 1b = b$. Moreover, if $ay = b$ then $y = 1y = (a^{-1}a)y = a^{-1}(ay) = a^{-1}b$, proving uniqueness. \square

We now summarize some of the elementary properties of fields, which are all we will need for our discussion.

(2.3) Proposition: If a, b , and c are elements of a field F then:

- (1) $0a = 0$;
- (2) $(-1)a = -a$;
- (3) $a(-b) = -(ab) = (-a)b$;
- (4) $-(-a) = a$;
- (5) $(-a)(-b) = ab$;
- (6) $-(a + b) = (-a) + (-b)$;
- (7) $a(b - c) = ab - ac$;
- (8) If $a \neq 0$ then $(a^{-1})^{-1} = a$;
- (9) If $a, b \neq 0$ then $(ab)^{-1} = b^{-1}a^{-1}$;
- (10) If $a + c = b + c$ then $a = b$;
- (11) If $c \neq 0$ and $ac = bc$ then $a = b$.
- (12) If $ab = 0$ then $a = 0$ or $b = 0$.

Proof: (1) Since $0a + 0a = (0 + 0)a = 0a$, we can add $-(0a)$ to both sides of the equation to obtain $0a = 0$.

(2) Since $(-1)a + a = (-1)a + 1a = [(-1) + 1]a = 0a = 0$ and also $(-a) + a = 0$, we see from Proposition 2.2 that $(-1)a = -a$.

(3) By (2) we have $a(-b) = a[(-1)b] = (-1)ab = -(ab)$ and similarly $(-a)b = -(ab)$.

(4) Since $a + (-a) = 0 = -(-a) + (-a)$, this follows from Proposition 2.2.

(5) From (3) and (4) it follows that $(-a)(-b) = a[-(-b)] = ab$.

(6) Since $(a + b) + [(-a) + (-b)] = a + b + (-a) + (-b) = 0$ and $(a + b) + [-(a + b)] = 0$, the result follows from Proposition 2.2.

(7) By (3) we have $a(b - c) = ab + a(-c) = ab + [-(ac)] = ab - ac$.

(8) Since $(a^{-1})^{-1}a^{-1} = 1 = aa^{-1}$, this follows from Proposition 2.2.

(9) Since $(a^{-1}b^{-1})(ba) = a^{-1}ab^{-1}b = 1 = (ab)^{-1}(ba)$, the result follows from Proposition 2.2.

(10) This is an immediate consequence of adding $-c$ to both sides of the equation.

(11) This is an immediate consequence of multiplying both sides of the equation by c^{-1} .

(12) If $b = 0$ we are done. If $b \neq 0$ then by (1) it follows that multiplying both sides of the equation by b^{-1} will yield $a = 0$. \square

(2.4) Proposition: Let a be a nonzero element of a finite field F having q elements. Then $a^{-1} = a^{q-2}$.

Proof: If $q = 2$ then $F = GF(2)$ and $a = 1$, so the result is immediate. Hence we can assume $q > 2$. Let $B = \{a_1, \dots, a_{q-1}\}$ be the nonzero elements of F , written in some arbitrary order. Then $aa_i \neq aa_h$ for $i \neq h$ since, were they equal, we would have $a_i = a^{-1}(aa_i) = a^{-1}(aa_h) = a_h$. Therefore $B = \{aa_1, \dots, aa_{q-1}\}$ and so

$$\prod_{i=1}^{q-1} a_i = \prod_{i=1}^{q-1} (aa_i) = a^{q-1} \left[\prod_{i=1}^{q-1} a_i \right].$$

Moreover, this is a product of nonzero elements of F and so, by Proposition 2.3(12), is also nonzero. Therefore, by Proposition 2.3(11), $1 = a^{q-1}$ and so $aa^{-1} = 1 = a^{q-1} = a(a^{q-2})$, implying that $a^{-1} = a^{q-2}$. \square

Exercises

Exercise 1 Let F be a field and let $G = F \times F$. Define operations of addition and multiplication on G by setting $(a, b) + (c, d) = (a + c, b + d)$ and $(a, b) \cdot (c, d) = (ac, bd)$. Do these operations define the structure of a field on G ?

Exercise 2 Let $r \in \mathbb{R}$ and let $0 \neq s \in \mathbb{R}$. Define operations \boxplus and \boxtimes on $\mathbb{R} \times \mathbb{R}$ by setting $(a, b) \boxplus (c, d) = (a + c, b + d)$ and

$$(a, b) \boxtimes (c, d) = (ac - bd(r^2 + s^2), ad + bc + 2rbd).$$

Do these operations, considered as addition and multiplication respectively, define the structure of a field on $\mathbb{R} \times \mathbb{R}$?

Exercise 3 Define a new operation \dagger on \mathbb{R} by setting $a \dagger b = a^3b$. Show that \mathbb{R} , on which we have the usual addition and this new operation as multiplication, satisfies all of the axioms of a field with the exception of one.

Exercise 4 Let $1 < t \in \mathbb{R}$ and let $F = \{a \in \mathbb{R} \mid a < 1\}$. Define operations \oplus and \odot on F as follows:

- (1) $a \oplus b = a + b - ab$ for all $a, b \in F$;
- (2) $a \odot b = 1 - t^{\log_t(1-a)\log_t(1-b)}$ for all $a, b \in F$.

For which values of t does F , together with these operations, form a field?

Exercise 5 Show that the set of all real numbers of the form $a + b\sqrt{2} + c\sqrt{3} + d\sqrt{6}$, where $a, b, c, d \in \mathbb{Q}$, forms a subfield of \mathbb{R} .

Exercise 6 Is $\{a + b\sqrt{15} \mid a, b \in \mathbb{Q}\}$ a subfield of \mathbb{R} ?

Exercise 7 Show that the field \mathbb{R} has infinitely-many distinct subfields.

Exercise 8 Let F be a field and define a new operation $*$ on F by setting $a * b = a + b + ab$. When is $(F, +, *)$ a field?

Exercise 9 Let F be a field and let G_n be the subset of F consisting of all elements which can be written as a sum of n squares of elements of F .

- (1) Is the product of two elements of G_2 again an element of G_2 ?
- (2) Is the product of two elements of G_4 again an element of G_4 ?

Exercise 10 Let $t = \sqrt[3]{2} \in \mathbb{R}$ and let S be the set of all real numbers of the form $a + bt + ct^2$, where $a, b, c \in \mathbb{Q}$. Is S a subfield of \mathbb{R} ?

Exercise 11 Let F be a field. Show that the function $a \mapsto a^{-1}$ is a permutation of $F \setminus \{0_F\}$.

Exercise 12 Show that every $z \in \mathbb{C}$ satisfies

$$z^4 + 4 = (z - 1 - i)(z - i + i)(z + 1 + i)(z + 1 - i).$$

Exercise 13 In each of the following, find the set of all complex numbers $z = a + bi$ satisfying the given relation. Note that this set may be empty or may be all of \mathbb{C} . Justify your result in each case.

- (a) $z^2 = \frac{1}{2}(1 + i\sqrt{3})$;
 (b) $(\sqrt{2})|z| \geq |a| + |b|$;
 (c) $|z| + z = 2 + i$;
 (d) $z^4 = 2 - (\sqrt{12})i$;
 (e) $z^4 = -4$.

Exercise 14 Let y be a complex number satisfying $|y| < 1$. Find the set of all complex numbers z satisfying $|z - y| \leq |1 - \bar{y}z|$.

Exercise 15 Let z_1, z_2 , and z_3 be complex numbers satisfying the condition that $|z_i| = 1$ for $i = 1, 2, 3$. Show that $|z_1z_2 + z_1z_3 + z_2z_3| = |z_1 + z_2 + z_3|$.

Exercise 16 For any $z_1, z_2 \in \mathbb{C}$, show that $|z_1|^2 + |z_2|^2 - z_1\bar{z}_2 - \bar{z}_1z_2 = |z_1 - z_2|^2$.

Exercise 17 Show that $|z + 1| \leq |z + 1|^2 + |z|$ for all $z \in \mathbb{C}$.

Exercise 18 If $z \in \mathbb{C}$, find $w \in \mathbb{C}$ satisfying $w^2 = z$.

Exercise 19 Define new operations \circ and \diamond on \mathbb{C} by setting $y \circ z = |y|z$ and

$$y \diamond z = \begin{cases} 0 & \text{if } y = 0 \\ \frac{1}{|y|}yz & \text{otherwise} \end{cases}$$

for all $y, z \in \mathbb{C}$. Is it true that $w \diamond (y \circ z) = (w \diamond y) \circ (w \diamond z)$ and $w \circ (y \diamond z) = (w \circ y) \diamond (w \circ z)$ for all $w, y, z \in \mathbb{C}$?

Exercise 20 Let $0 \neq z \in \mathbb{C}$. Show that there are infinitely-many complex numbers y satisfying the condition $y\bar{y} = z\bar{z}$.

Exercise 21 (Abel's inequality⁵): Let z_1, \dots, z_n be a list of complex numbers and, for each $1 \leq k \leq n$, let $s_k = \sum_{i=1}^k z_i$. For real numbers a_1, \dots, a_n satisfying $a_1 \geq a_2 \geq \dots \geq a_n \geq 0$, show that

$$\left| \sum_{i=1}^n a_i z_i \right| \leq a_1 \left(\max_{1 \leq k \leq n} |s_k| \right).$$



⁵ Nineteenth-century Norwegian mathematical genius **Niels Henrik Abel** died tragically at the age of 26.

Exercise 22 Let $0 \neq z_0 \in \mathbb{C}$ satisfy the condition $|z_0| < 2$. Show that there are precisely two complex numbers, z_1 and z_2 , satisfying $|z_1| + |z_2| = 1$ and $z_1 + z_2 = z_0$.

Exercise 23 If p is a prime positive integer, find all subfields of $GF(p)$.

Exercise 24 Find elements $c, d \neq \pm 1$ in the field $\mathbb{Q}(\sqrt{5})$ satisfying $cd = 19$.

Exercise 25 Let F be the set of all real numbers of the form

$$a + b\left(\sqrt[3]{5}\right) + c\left(\sqrt[3]{5}\right)^2,$$

where $a, b, c \in \mathbb{Q}$. Is F a subfield of \mathbb{R} ?

Exercise 26 Let p be a prime positive integer and let $a \in GF(p)$. Does there necessarily exist an element b of $GF(p)$ satisfying $b^2 = a$?

Exercise 27 Let $F = GF(11)$ and let $G = F \times F$. Define operations of addition and multiplication on G by setting $(a, b) + (c, d) = (a + c, b + d)$ and $(a, b) \cdot (c, d) = (ac + 7bd, ad + bc)$. Do these operations define the structure of a field on G ?

Exercise 28 Let F be a field and let G be a finite subset of $F \setminus \{0\}$ containing 1 and satisfying the condition that if $a, b \in F$ then $ab^{-1} \in G$. Show that there exists an element $c \in G$ such that $G = \{c^i \mid i \geq 0\}$.

Exercise 29 Let F be a field satisfying the condition that the function $a \mapsto a^2$ is a permutation of F . What is the characteristic of F ?

Exercise 30 Is $\mathbb{Z}/(6)$ an integral domain?

Exercise 31 Let $F = \{a + b\sqrt{5} \in \mathbb{Q}(\sqrt{5}) \mid a, b \in \mathbb{Z}\}$. Is F an integral domain?

Exercise 32 Let F be an integral domain and let $a \in F$ satisfy $a^2 = a$. Show that $a = 0$ or $a = 1$.

Exercise 33 Let a be a nonzero element in an integral domain F . If $b \neq c$ are distinct elements of F , show that $ab \neq ac$.

Exercise 34 Let F be an integral domain and let G be a nonempty subset of F containing 0 and 1 and closed under the operations of addition and multiplication in F . Is G necessarily an integral domain?

Exercise 35 Let U be the set of all positive integers and let F be the set of all functions from U to \mathbb{C} . Define operations of addition and multiplication on F by setting $f + g : k \mapsto f(k) + g(k)$ and $fg : k \mapsto \sum_{i+j=k} f(i)g(j)$ for all $k \in U$. Is F , together with these operations, an integral domain? Is it a field?

Exercise 36 Let F be the set of all functions f from \mathbb{R} to itself of the form $f : t \mapsto \sum_{k=1}^n [a_k \cos(kt) + b_k \sin(kt)]$, where the a_k and b_k are real numbers and n is some positive integer. Define addition and multiplication on F by setting $f+g : t \mapsto f(t)+g(t)$ and $fg : t \mapsto f(t)g(t)$ for all $t \in \mathbb{R}$. Is F , together with these operations, an integral domain? Is it a field?

Exercise 37 Show that every integral domain having only finitely-many elements is a field.

Exercise 38 Let F be a field of characteristic other than 2 in which there exist elements a_1, \dots, a_n satisfying $\sum_{i=1}^n a_i^2 = -1$. (This happens, for example, in the case $F = \mathbb{C}$). Show that for any $c \in F$ there exist elements b_1, \dots, b_k of F satisfying $c = \sum_{i=1}^k b_i^2$.

Exercise 39 Let p be a prime integer. Show that for each $a \in GF(p)$ there exist elements b and c of $GF(p)$, not necessarily distinct, satisfying $a = b^2 + c^2$.

Exercise 40 Let F be a field in which we have elements a , b , and c (not necessarily distinct) satisfying $a^2 + b^2 + c^2 = -1$. Show that there exist (not necessarily distinct) elements d and e of F , satisfying $d^2 + e^2 = -1$.

Exercise 41 Is every nonzero element of the field $GF(5)$ in the form 2^i for some positive integer i ? What happens in the case of the field $GF(7)$?

Exercise 42 Find the set of all fields F in which there exists an element a satisfying the condition that $a + b = a$ for all $b \in F \setminus \{a\}$.

Exercise 43 (Binomial Formula) If a and b are elements of a field F , and if n is a positive integer, show that $(a+b)^n = \sum_{k=0}^n \binom{n}{k} a^k b^{n-k}$.

Exercise 44 Let F be a field of characteristic $p > 0$. Use the previous two exercises to show that the function $\gamma : F \rightarrow F$ defined by $\gamma : a \mapsto a^p$ is monic.

Exercise 45 Let a and b be nonzero elements of a finite field F , and let m and n be positive integers satisfying $a^m = b^n = 1$. Show that there exists a nonzero element c of F satisfying $c^k = 1$, where k is the least common multiple of m and n .

Exercise 46 If a is a nonzero element of a field F , show that $(-a)^{-1} = -(a^{-1})$.

Exercise 47 A field F is **orderable** if and only if there exists a subset P closed under addition and multiplication such that for each $a \in F$ precisely one of the following conditions holds: (i) $a = 0$; (ii) $a \in P$; (iii) $-a \in P$. Show that $GF(5)$ is not orderable.

3

Vector spaces over a field

If $n > 1$ is an integer and if F is a field, it is natural to define addition on the set F^n componentwise:

$$(a_1, \dots, a_n) + (b_1, \dots, b_n) = (a_1 + b_1, \dots, a_n + b_n).$$

More generally, if Ω is any nonempty set and if F^Ω is the set of all functions from Ω to the field F , we can define addition on F^Ω by setting $f + g : i \mapsto f(i) + g(i)$ for each $i \in \Omega$. Given these definitions, is it possible to define multiplication in such a manner that F^n or F^Ω will become a field naturally containing F as a subfield? We have seen that if $n = 2$ and if $F = \mathbb{R}$ or $F = \mathbb{Q}$, this is possible – and, indeed, in the latter case there are several different methods of doing it. If $F = GF(p)$ then it is possible to define such a field structure on F^n for every integer $n > 1$. However in general the answer is negative – as we will show in a later chapter for the specific case of \mathbb{R}^k , where $k > 2$ is an odd integer. Nonetheless, it is possible to construct another important and useful structure on these sets, and this structure will be the focus of our attention for the rest of this book. We will first give the formal definition, and then look at a large number of examples.

Let F be a field. A nonempty set V , together with a function $V \times V \rightarrow V$ called **vector addition** (denoted, as usual, by $+$) and a function $F \times V \rightarrow V$ called **scalar multiplication** (denoted, as a rule,

by concatenation) is a **vector space** over F if the following conditions are satisfied:

- (1) (**associativity of vector addition**): $v + (w + y) = (v + w) + y$ for all $v, w, y \in V$.
- (2) (**commutativity of vector addition**): $v + w = w + v$ for all $v, w \in V$.
- (3) (**existence of a identity element for vector addition**): There exists an element 0_V of V satisfying the condition that $v + 0_V = v$ for all $v \in V$.
- (4) (**existence of additive inverses**): For each $v \in V$ there exists an element of V , which we will denote by $-v$, which satisfies $v + (-v) = 0_V$.
- (5) (**distributivity of scalar multiplication over vector addition and of scalar multiplication over field addition**): $a(v + w) = av + aw$ and $(a + b)v = av + bv$ for all $a, b \in F$ and $v, w \in V$.
- (6) (**associativity of scalar multiplication**): $(ab)v = a(bv)$ for all $a, b \in F$ and $v \in V$.
- (7) (**existence of identity element for scalar multiplication**): $1v = v$ for all $v \in V$.

The elements of V are called **vectors** and the elements of F are called **scalars**¹.

Example: Note that condition (7), apparently trivial, does not follow from the other conditions. Indeed, if we take $V = F$ but define scalar multiplication by $av = 0_V$ for all $a \in F$ and $v \in V$, we would get a structure which satisfies conditions (1) - (6) but not condition (7).

If $v, w \in V$ we again write $v - w$ instead of $v + (-w)$. As we noted when we talked about fields, if v_1, \dots, v_n is a list of vectors in a vector space V over a field F , the associativity of vector addition allows us to unambiguously write $v_1 + \dots + v_n$, and this sum is often denoted by $\sum_{i=1}^n v_i$. Moreover, if $a \in F$ is a scalar then we surely have $a(\sum_{i=1}^n v_i) = (\sum_{i=1}^n av_i)$. Similarly, if a_1, \dots, a_n is a list of scalars and



The theory of vector spaces was developed in the 1880's by the American engineer and physicist, **Josiah Willard Gibbs** and the British engineer **Oliver Heaviside**, based on the work of the Scottish physicist **James Clerk Maxwell**, the German high-school teacher **Herman Grassmann**, and the French engineer **Jean Claude Saint-Venant**.

if $v \in V$, then we have $(\sum_{i=1}^n a_i)v = \sum_{i=1}^n a_iv$. We will also adopt the convention that the sum of an empty set of vectors is equal to 0_V .

Clearly any field F is a vector space over itself, where we take the vector addition to be the addition in F and scalar multiplication to be the multiplication in F .

We also note an extremely important construction. Let F be a field and let Ω be a nonempty set. Assume that, for each $i \in \Omega$, we are given a vector space V_i over F , the addition in which we will denote by $+_i$ (the vector spaces V_i need not, however, be distinct from one another). Recall that $\prod_{i \in \Omega} V_i$ is the set of all those functions f from Ω to $\bigcup_{i \in \Omega} V_i$ which satisfy the condition that $f(i) \in V_i$ for each $i \in \Omega$. We now define the structure of a vector space on $\prod_{i \in \Omega} V_i$ as follows: if $f, g \in \prod_{i \in \Omega} V_i$ then $f + g$ is the function in $\prod_{i \in \Omega} V_i$ given by $f + g : i \mapsto f(i) +_i g(i)$ for each $i \in \Omega$. Moreover, if $a \in F$ and $f \in \prod_{i \in \Omega} V_i$, then af is the function in $\prod_{i \in \Omega} V_i$ given by $af : i \mapsto a[f(i)]$ for each $i \in \Omega$. It is routine to verify that all of the axioms of a vector space are satisfied in this case. For example, the identity element for vector addition is just the function in $\prod_{i \in \Omega} V_i$ given by $i \mapsto 0_{V_i}$ for each $i \in \Omega$. This vector space is called the **direct product** of the vector spaces V_i over F . If the set Ω is finite, say $\Omega = \{1, \dots, n\}$, then we often write $V_1 \times \dots \times V_n$ instead of $\prod_{i \in \Omega} V_i$. If all of the vector spaces V_i are equal to the same vector space V , then we write V^Ω instead of $\prod_{i \in \Omega} V_i$ and if $\Omega = \{1, \dots, n\}$ we write V^n instead of V^Ω . Note that a function f from a finite set $\Omega = \{1, \dots, n\}$ to a vector space V is totally defined by the list $f(1), f(2), \dots, f(n)$ of its values. Conversely, any list v_1, \dots, v_n of elements of V uniquely defines such a function f given by $f : i \mapsto v_i$. Therefore this notation agrees with our previous use of the symbol V^n to denote sets of n -tuples of elements of V . However, to emphasize the vector space structure here, we

will write the elements of V^n as columns of the form $\begin{bmatrix} v_1 \\ \vdots \\ v_n \end{bmatrix}$, where the

v_i are (not necessarily distinct) elements of V . Usually, we will consider the case $V = F$. Vector addition and scalar multiplication in V^n are then defined by the rules

$$\begin{bmatrix} v_1 \\ \vdots \\ v_n \end{bmatrix} + \begin{bmatrix} w_1 \\ \vdots \\ w_n \end{bmatrix} = \begin{bmatrix} v_1 + w_1 \\ \vdots \\ v_n + w_n \end{bmatrix} \quad \text{and} \quad c \begin{bmatrix} v_1 \\ \vdots \\ v_n \end{bmatrix} = \begin{bmatrix} cv_1 \\ \vdots \\ cv_n \end{bmatrix}.$$

The “classical” study of vector spaces centers around the spaces \mathbb{R}^n , the vectors in which are identified with the points in n -^{dim}ensional euclidean

space². However, other vector spaces also have important applications. Vector spaces of the form \mathbb{C}^n are needed for the study of functions of several complex variables. In algebraic coding theory, one is interested in spaces of the form F^n , where F is a finite field. The vectors in this space are **words** of length n and the field F is the **alphabet** in which these words are written. Thus, for example, a popular choice for F is the Galois field $GF(2^8)$, the 256 elements of which are identified with the 256 ASCII symbols.

Let V be a vector space, let k and n be positive integers, and let $\Omega = \{(i, j) \mid 1 \leq i \leq k, 1 \leq j \leq n\}$. There exists a bijective correspondence between V^Ω and the set of all rectangular arrays of the

form $\begin{bmatrix} v_{11} & \cdots & v_{1n} \\ \vdots & & \vdots \\ v_{k1} & \cdots & v_{kn} \end{bmatrix}$ in which the entries v_{ij} are elements of V .

Such an array is called a $k \times n$ **matrix**³ over V . We will denote the set of all such matrices by $\mathcal{M}_{k \times n}(V)$. Addition and scalar multiplication in $\mathcal{M}_{k \times n}(V)$ is given by

$$\begin{aligned} & \begin{bmatrix} v_{11} & \cdots & v_{1n} \\ \vdots & & \vdots \\ v_{k1} & \cdots & v_{kn} \end{bmatrix} + \begin{bmatrix} w_{11} & \cdots & w_{1n} \\ \vdots & & \vdots \\ w_{k1} & \cdots & w_{kn} \end{bmatrix} \\ &= \begin{bmatrix} v_{11} + w_{11} & \cdots & v_{1n} + w_{1n} \\ \vdots & & \vdots \\ v_{k1} + w_{k1} & \cdots & v_{kn} + w_{kn} \end{bmatrix} \end{aligned}$$



² The first explicit statement of the geometric “parallelogram law” for adding vectors in \mathbb{R}^2 was given by the sixteenth-century Pisan scientist **Galileo Galilei**.



³ The term “matrix” was first coined by the 19th-century British mathematician **James Joseph Sylvester** in 1848. Sylvester was one of the major researchers in the theory of matrices and determinants.

and $c \begin{bmatrix} v_{11} & \dots & v_{1n} \\ \vdots & & \vdots \\ v_{k1} & \dots & v_{kn} \end{bmatrix} = \begin{bmatrix} cv_{11} & \dots & cv_{1n} \\ \vdots & & \vdots \\ cv_{k1} & \dots & cv_{kn} \end{bmatrix}$. The identity element

for vector addition in $\mathcal{M}_{k \times n}(V)$ is the **0-matrix** $O = \begin{bmatrix} 0_V & \dots & 0_V \\ \vdots & & \vdots \\ 0_V & \dots & 0_V \end{bmatrix}$.

Note that $V^n = \mathcal{M}_{n \times 1}(V)$.

If V is a vector space and if $\Omega = \mathbb{N}$, then the elements of V^Ω are infinite sequences $[v_0, v_1, \dots]$ of elements of V . We will denote this vector space, which we will need later, by V^∞ . Again, the space of particular interest will be F^∞ .

Example: If F is a subfield of a field K , then K is a vector space over F , with addition and multiplication just being the operations in K . Thus, in particular, we can think of \mathbb{C} as a vector space over \mathbb{R} and of \mathbb{R} as a vector space over \mathbb{Q} .

Example: Let A be a nonempty set and let V be the collection of all subsets of A . Let us define addition of elements of V as follows: if B and C are elements of V then $B+C = (B \cup C) \setminus (B \cap C)$. This operation is usually called the **symmetric difference** of B and C . This definition turns V into a vector space over $GF(2)$, where scalar multiplication is defined by $0B = \emptyset$ and $1B = B$ for all $B \in V$. This is actually just a special case of what we have seen before. Indeed, we note that there is a bijective function from V to $GF(2)^A$ which assigns to each subset B of A its **characteristic function**, namely the function χ_B defined by

$$\chi_B : a \mapsto \begin{cases} 1 & \text{if } a \in B \\ 0 & \text{otherwise} \end{cases}$$

and it is easy to see that $\chi_A + \chi_B = \chi_{A+B}$, while $\chi_A \chi_B = \chi_{A \cap B}$.

(3.1) Proposition: Let V be a vector space over a field F .

- (1) If $z \in V$ satisfies $z + v = v$ for all $v \in V$ then $z = 0_V$.
- (2) If $v, w \in V$ then there exists a unique element $y \in V$ satisfying $v + y = w$.

Proof: The proof is similar to the proofs of Proposition 2.1(1) and Proposition 2.2(1). \square

(3.2) Proposition: Let V be a vector space over a field F . If $v, w \in V$ and if $a \in F$, then:

- (1) $a0_V = 0_V$;
- (2) $0v = 0_V$;
- (3) $(-1)v = -v$;
- (4) $(-a)v = -(av) = a(-v)$;
- (5) $-(-v) = v$;
- (6) $av = (-a)(-v)$;
- (7) $-(v + w) = -v - w$;
- (8) $a(v - w) = av - aw$;
- (9) **If** $av = 0_V$ **then either** $v = 0_V$ **or** $a = 0$.

Proof: The proof is similar to the proof of Proposition 2.3. □

Let V be a vector space over a field F . A nonempty subset W of V is a **subspace** of V if and only if it is a vector space in its own right with respect to the addition and scalar multiplication defined on V . Thus, any vector space V is a subspace of itself, called the **improper** subspace; any other subspace is **proper**. Also, $\{0_V\}$ is surely a subspace of V , called the **trivial** subspace; any other subspace is **nontrivial**.

Note that the two conditions for a nonempty subset of a vector space to be a subspace are independent: the set of all vectors in \mathbb{R}^3 all entries of which are integers is closed under vector addition but not under scalar

multiplication; the set of all vectors $\begin{bmatrix} a \\ b \\ c \end{bmatrix} \in \mathbb{R}^3$ satisfying $abc = 0$ is closed under scalar multiplication but not under vector addition.

Example: Let V be a vector space over a field F and let Ω be a nonempty set. We have already seen that the set V^Ω of all functions from Ω to V is a vector space over F . If Λ is a subset of Ω then the set $\{f \in V^\Omega \mid f(i) = 0_V \text{ for all } i \in \Lambda\}$ is a subspace of V^Ω . In particular, if $k < n$ are positive integers, then we can think of V^k as being a subspace

of V^n , by identifying it with $\left\{ \begin{bmatrix} v_1 \\ \vdots \\ v_n \end{bmatrix} \in V^n \mid v_{k+1} = \dots = v_n = 0_V \right\}$.

Note that if $y \in V$, then $\{f \in V^\Omega \mid f(i) = y \text{ for all } i \in \Lambda\}$ is not a subspace of V^Ω unless $y = 0_V$.

Example: Let $\{V_i \mid i \in \Omega\}$ be a collection of vector spaces over a field F . The set of all functions $f \in \prod_{i \in \Omega} V_i$ satisfying the condition that $f(i) \neq 0_{V_i}$ for at most finitely-many elements i of Ω is a subspace of $\prod_{i \in \Omega} V_i$, called the **direct coproduct** of the spaces V_i and denoted by $\coprod_{i \in \Omega} V_i$. The direct coproduct is a proper subset of $\prod_{i \in \Omega} V_i$ when and only when the set Ω is infinite. If each of the spaces V_i is equal to a given vector space V , we write $V^{(\Omega)}$ instead of $\coprod_{i \in \Omega} V_i$.

Example: If V is a vector space over a field F and if $v \in V$, then the set $Fv = \{av \mid a \in F\}$ is a subspace of V which is contained in any subspace of V containing v .

Example: Let \mathbb{R} be the field of real numbers and let Ω be either equal to \mathbb{R} or to some closed interval $[a, b]$ on the real line. We have already seen that the set \mathbb{R}^Ω of all functions from Ω to \mathbb{R} is a vector space over \mathbb{R} . The set of all continuous functions from Ω to \mathbb{R} is a subspace of this vector space, as are the set of all differentiable functions from Ω to \mathbb{R} , the set of all infinitely-differentiable functions from Ω to \mathbb{R} , and the set of all analytic functions from Ω to \mathbb{R} . If $a < b$ are real numbers, we will denote the space of all continuous functions from the closed interval $[a, b]$ to \mathbb{R} by $C(a, b)$. These spaces will be very important to us later⁴.

(3.3) Proposition: If V is a vector space over a field F , then a nonempty subset W of V is a subspace of V if and only if it is closed under addition and scalar multiplication.

Proof: If W is a subspace of V then it is surely closed under addition and scalar multiplication. Conversely, suppose that it is so closed. Then for any $w \in W$ we have $0_V = 0w \in W$ and $-w = (-1)w \in W$. All of the other conditions are satisfied in W because they are satisfied in V . \square

(3.4) Proposition: If V is a vector space over a field F , and if $\{W_i \mid i \in \Omega\}$ is a collection of subspaces of V , then $\bigcap_{i \in \Omega} W_i$ is a subspace of V .

Proof: Set $W = \bigcap_{i \in \Omega} W_i$. If $w, y \in W$ then, for each $i \in \Omega$, we have $w, y \in W_i$ and so $w + y \in W_i$. Thus $w + y \in W$. Similarly, if $a \in R$ and $w \in W$ then $aw \in W_i$ for each $i \in \Omega$, and so $aw \in W$. \square

We will also set the convention that the intersection of an empty collection of subspaces of V is V itself. Subspaces W and W' are **disjoint** if and only if $W \cap W' = \{0_V\}$. More generally, a collection $\{W_i \mid i \in \Omega\}$



⁴ The first fundamental research in spaces of functions was done by the German mathematician **Erhard Schmidt**, a student of David Hilbert, whose work forms one of the bases of functional analysis.

of subspaces of V is **pairwise disjoint** if and only if $W_i \cap W_j = \{0_V\}$ for $i \neq j$ in Ω . (Note that disjointness of subspaces of a given space is not the same as disjointness of subsets!)

Now let us look at a very important method of constructing subspaces of vector spaces. Let D be a nonempty set of elements of a vector space V over a field F . A vector $v \in V$ is a **linear combination** of elements of D over F if and only if there exist elements v_1, \dots, v_n of D and scalars a_1, \dots, a_n in F such that $v = \sum_{i=1}^n a_i v_i$. We will denote the set of all linear combinations of elements of D over F by FD . Note that if $v \in V$ then $F\{v\}$ is the set Fv which we defined earlier.

It is clear that if D is a nonempty set of elements of a vector space V over a field F then $D \subseteq FD$. Also, $0_V \in FD$ for any nonempty subset D of V , and it is the only vector belonging to each of the sets FD . To simplify notation, we will therefore define $F\emptyset$ to be $\{0_V\}$. If $D' \subseteq D$ then surely $FD' \subseteq FD$. We also note that $FD = F(D \cup \{0_V\})$ for any subset D of V .

Example: If $D = \left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} \right\}$ and $D' = \left\{ \begin{bmatrix} 0 \\ 2 \\ 0 \end{bmatrix}, \begin{bmatrix} 3 \\ 3 \\ 0 \end{bmatrix} \right\}$

are subsets of \mathbb{R}^3 , then $FD = FD' = \left\{ \begin{bmatrix} a \\ b \\ 0 \end{bmatrix} \mid a, b \in \mathbb{R} \right\}$. Indeed,

if $\begin{bmatrix} a \\ b \\ 0 \end{bmatrix} \in \mathbb{R}^3$ then

$$\begin{bmatrix} a \\ b \\ 0 \end{bmatrix} = a \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} + b \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} = \left(\frac{b-a}{2} \right) \begin{bmatrix} 0 \\ 2 \\ 0 \end{bmatrix} + \frac{a}{3} \begin{bmatrix} 3 \\ 3 \\ 0 \end{bmatrix}.$$

Example: If $D = \left\{ \begin{bmatrix} 0 \\ 0 \\ 4 \end{bmatrix}, \begin{bmatrix} 2 \\ 2 \\ 0 \end{bmatrix}, \begin{bmatrix} 2 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} \right\} \subseteq \mathbb{R}^3$ then

$$\begin{aligned} \begin{bmatrix} 4 \\ 2 \\ 4 \end{bmatrix} &= 1 \begin{bmatrix} 0 \\ 0 \\ 4 \end{bmatrix} + 1 \begin{bmatrix} 2 \\ 2 \\ 0 \end{bmatrix} + 1 \begin{bmatrix} 2 \\ 0 \\ 0 \end{bmatrix} \\ &= 1 \begin{bmatrix} 2 \\ 0 \\ 0 \end{bmatrix} + (-1) \begin{bmatrix} 2 \\ 2 \\ 0 \end{bmatrix} + 4 \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}. \end{aligned}$$

Thus we see that there may be several ways of representing a vector as a linear combination of elements of a given subset of a vector space.

(3.5) Proposition: Let D be a subset of a vector space V over a field F . Then:

- (1) FD is a subspace of V ;
- (2) Every subspace of V containing D also contains FD ;
- (3) FD is the intersection of all subspaces of V containing D .

Proof: If $D = \emptyset$ then $FD = \{0_V\}$ and we are done. Thus we can assume that D is nonempty. It is an immediate consequence of the definitions that the sum of two linear combinations of elements of D over F is again a linear combination of elements of D over F , and that the product of a scalar and a linear combination of elements of D over F is again a linear combination of elements of D over F . This proves (1). Moreover, (2) is an immediate consequence of (1) and Proposition 3.3, while (3) follows directly from (2). \square

If D is a subset of a vector space V over a field F then the subspace FD of V is called the subspace **generated** or **spanned** by D , and the set D is called a **generating set** or **spanning set** for this subspace. In particular, we note that \emptyset is a generating set for $\{0_V\}$.

Example: Let F be a field. The set $\left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \right\}$ is a generating set for the vector space F^3 over F . The set

$$\left\{ \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix} \right\}$$

is also a generating set for F^3 if the characteristic of F is other than 2,

but not for $F = GF(2)$ since, in this case, $\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$ cannot be written as a linear combination of this set of vectors over F . The set

$$\left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} \right\}$$

is not a generating set for F^3 , for any field F , since $\begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$ cannot be written as a linear combination of the elements of this set.

(3.6) Proposition: Let V be a vector space over a field F and let D_1 and D_2 be subsets of V satisfying $D_1 \subseteq D_2 \subseteq FD_1$. Then $FD_1 = FD_2$.

Proof: Since FD_1 is a subspace of V containing D_2 , we know by Proposition 3.5 that $FD_2 \subseteq FD_1$. Conversely, any linear combination of elements of D_1 over F is also a linear combination of elements of D_2 over F and so $FD_1 \subseteq FD_2$, thus establishing equality. \square

In particular, we note that $FD = F(FD)$ for any subset D of V .

(3.7) Proposition (Exchange Property): Let V be a vector space over a field F and let $v, w \in V$. Let D be a subset of V satisfying $v \in F(D \cup \{w\}) \setminus FD$. Then $w \in F(D \cup \{v\})$.

Proof: Since $v \in F(D \cup \{w\})$ we know that there exist elements v_1, \dots, v_n of D and scalars a_1, \dots, a_n, b in F satisfying the condition that $v = \sum_{i=1}^n a_i v_i + bw$. Moreover, since $v \notin FD$, we know that $b \neq 0$ and so $w = b^{-1}v - \sum_{i=1}^n b^{-1}a_i v_i \in F(D \cup \{v\})$. \square

A vector space V over a field F is **finitely generated** over F if it has a finite generating set. Finitely-generated vector spaces are often much easier to deal with by purely algebraic methods and therefore, in several situations, we will have to restrict our discussion to these spaces.

Example: If F is a field and n is a positive integer, then one sees

that $\left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}, \dots, \begin{bmatrix} 0 \\ 0 \\ 0 \\ \vdots \\ 1 \end{bmatrix} \right\}$ is a finite generating set for F^n over

F , and so F^n is finitely generated. More generally, if V is a vector space finitely generated over a field F , say $V = F\{v_1, \dots, v_k\}$, and if n is a positive integer, then

$$\left\{ \begin{bmatrix} v_1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}, \begin{bmatrix} v_2 \\ 0 \\ \vdots \\ 0 \end{bmatrix}, \dots, \begin{bmatrix} v_k \\ 0 \\ \vdots \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ v_1 \\ \vdots \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ v_2 \\ \vdots \\ 0 \end{bmatrix}, \dots, \begin{bmatrix} 0 \\ 0 \\ \vdots \\ v_k \end{bmatrix}, \dots \right. \\ \left. \begin{bmatrix} 0 \\ 0 \\ \vdots \\ v_1 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ \vdots \\ v_2 \end{bmatrix}, \dots, \begin{bmatrix} 0 \\ 0 \\ \vdots \\ v_k \end{bmatrix} \right\}$$

is a generating set for V^n having kn elements.

Example: If F is a field and if k and n are positive integers, then the vector space $\mathcal{M}_{k \times n}(F)$ of all $k \times n$ matrices over F is finitely generated over F . Similarly, if V is a finitely-generated vector space over F , then the vector space $\mathcal{M}_{k \times n}(V)$ is also finitely generated over F .

Example: For any field F , the vector space F^∞ is not finitely generated over F .

Example: The field \mathbb{R} is finitely generated as a vector space over itself, but is not finitely generated as a vector space over \mathbb{Q} .

Let V be a vector space over a field F . In Proposition 3.4, we saw that if $\{W_i \mid i \in \Omega\}$ is a collection of subspaces of V then $\bigcap_{i \in \Omega} W_i$ is a subspace of V . In the same way, we can define the subspace $\sum_{i \in \Omega} W_i$ of V to be the set of all vectors in V of the form $\sum_{j \in \Lambda} w_j$, where Λ is a finite nonempty subset of Ω and $w_j \in W_j$ for each $j \in \Lambda$. In other words, $\sum_{i \in \Omega} W_i = F(\bigcup_{i \in \Omega} W_i)$. Indeed, from the definition of this sum, we see something stronger: if D_i is a generating set for W_i for each $i \in \Omega$ then $\sum_{i \in \Omega} W_i = F(\bigcup_{i \in \Omega} D_i)$.

As a special case of the above, we see that if W_1 and W_2 are subspaces of V , then $W_1 + W_2$ equals the set of all vectors of the form $w_1 + w_2$, where $w_1 \in W_1$ and $w_2 \in W_2$. If both W_1 and W_2 are finitely generated then $W_1 + W_2$ is also finitely generated. By induction, we can then show that if W_1, \dots, W_n are finitely-generated subspaces of V , then $\sum_{i=1}^n W_i$ is also finitely generated.

(3.8) Proposition: If V is a vector space over a field F and if $\{W_i \mid i \in \Omega\}$ is a collection of subspaces of V , then:

- (1) W_h is a subspace of $\sum_{i \in \Omega} W_i$ for all $h \in \Omega$;
- (2) If Y is a subspace of V satisfying the condition that W_h is a subspace of Y for all $h \in \Omega$, then $\sum_{i \in \Omega} W_i$ is a subspace of Y .

Proof: (1) is clear from the definition. As for (2), if we have a subspace Y satisfying the given condition, if Λ is a finite subset of Ω , and if $w_j \in W_j$ for each $j \in \Lambda$, then $w_j \in Y$ for each j and so $\sum_{j \in \Lambda} w_j \in Y$. Thus $\sum_{i \in \Omega} W_i \subseteq Y$. \square

(3.9) Proposition: If V is a vector space over a field F and if W_1, W_2 , and W_3 are subspaces of V , then:

- (1) $(W_1 + W_2) + W_3 = W_1 + (W_2 + W_3)$;
- (2) $W_1 + W_2 = W_2 + W_1$;
- (3) $W_3 \cap [W_2 + (W_1 \cap W_3)] = (W_1 \cap W_3) + (W_2 \cap W_3)$;

(4) (Modular law for subspaces): If $W_1 \subseteq W_3$ then

$$W_3 \cap (W_2 + W_1) = W_1 + (W_2 \cap W_3).$$

Proof: Parts (1) and (2) follow immediately from the definition, while (4) is a special case of (3). We are therefore left to prove (3). Indeed, if v belongs to $W_3 \cap [W_2 + (W_1 \cap W_3)]$, then we can write $v = w_2 + y$, where $w_2 \in W_2$ and $y \in W_1 \cap W_3$. Since $v, y \in W_3$, it follows that $w_2 = v - y \in W_3$, and so $v = y + w_2 \in (W_1 \cap W_3) + (W_2 \cap W_3)$. Thus we see that $W_3 \cap [W_2 + (W_1 \cap W_3)] \subseteq (W_1 \cap W_3) + (W_2 \cap W_3)$. Conversely, assume that $v \in (W_1 \cap W_3) + (W_2 \cap W_3)$. Then, in particular, $v \in W_3$ and we can write $v = w_1 + w_2$, where $w_1 \in W_1 \cap W_3$ and $w_2 \in W_2 \cap W_3$. Thus $v = w_1 + w_2 \in W_3 \cap W_2 + (W_1 \cap W_3)$. This shows that $(W_1 \cap W_3) + (W_2 \cap W_3) \subseteq W_3 \cap [W_2 + (W_1 \cap W_3)]$ and so we have the desired equality. \square

Exercises

Exercise 48 Let $V = \mathbb{Q}^2$, with the usual vector addition. If $a + b\sqrt{2} \in \mathbb{Q}(\sqrt{2})$ and if $\begin{bmatrix} c \\ d \end{bmatrix} \in \mathbb{Q}^2$, set $(a + b\sqrt{2}) \begin{bmatrix} c \\ d \end{bmatrix} = \begin{bmatrix} ac + 2bd \\ bc + ad \end{bmatrix}$. Do these operations turn \mathbb{Q}^2 into a vector space over $\mathbb{Q}(\sqrt{2})$?

Exercise 49 Is it possible to define on $V = \mathbb{Z}/(4)$ the structure of a vector space over $GF(2)$ in such a way that the vector addition is the usual addition in $\mathbb{Z}/(4)$?

Exercise 50 Consider the set \mathbb{Z} of integers, together with the usual addition. If $a \in \mathbb{Q}$ and $k \in \mathbb{Z}$, define $a \cdot k$ to be $[a]k$, where $[a]$ denotes the largest integer less than or equal to a . Using this as our definition of “scalar multiplication”, have we turned \mathbb{Z} into a vector space over \mathbb{Q} ?

Exercise 51 Let $V = \{0, 1\}$ and let $F = GF(2)$. Define vector addition and scalar multiplication by setting $v + v' = \max\{v, v'\}$, $0v = 0$, and $1v = v$ for all $v, v' \in V$. Does this define on V the structure of a vector space over F ?

Exercise 52 Let $V = C(0, 1)$. Define an operation \boxplus on V by setting $f \boxplus g : x \mapsto \max\{f(x), g(x)\}$. Does this operation of vector addition, together with the usual operation of scalar multiplication, define on V the structure of a vector space over \mathbb{R} ?

Exercise 53 Let $V \neq \{0_V\}$ be a vector space over \mathbb{R} . For each $v \in V$ and each complex number $a + bi$, let us define $(a + bi)v = av$. Does V , together with this new scalar multiplication, form a vector space over \mathbb{C} ?

Exercise 54 Let $V = \{i \in \mathbb{Z} \mid 0 \leq i < 2^n\}$ for some given positive integer n . Define operations of vector addition and scalar multiplication on V in such a way as to turn it into a vector space over the field $GF(2)$.

Exercise 55 Let V be a vector space over a field F . Define a function from $GF(3) \times V$ to V by setting $(0, v) \mapsto 0_V$, $(1, v) \mapsto v$, and $(2, v) \mapsto -v$ for all $v \in V$. Does this function, together with the vector addition in V , define on V the structure of a vector space over $GF(3)$?

Exercise 56 Let $V = \mathbb{R} \cup \{\infty\}$ and extend the usual addition of real numbers by defining $v + \infty = \infty + v = \infty$ for all $v \in V$. Is it possible to define an operation of scalar multiplication on V in such a manner as to turn it into a vector space over \mathbb{R} ?

Exercise 57 Let $V = \mathbb{R}^2$. If $\begin{bmatrix} a \\ b \end{bmatrix}, \begin{bmatrix} a' \\ b' \end{bmatrix} \in V$ and $r \in \mathbb{R}$, set

$$\begin{bmatrix} a \\ b \end{bmatrix} + \begin{bmatrix} a' \\ b' \end{bmatrix} = \begin{bmatrix} a + a' + 1 \\ b + b' \end{bmatrix} \quad \text{and} \quad r \begin{bmatrix} a \\ b \end{bmatrix} = \begin{bmatrix} ra + r - 1 \\ rb \end{bmatrix}. \quad \text{Do}$$

these operations define on V the structure of a vector space over \mathbb{R} ? If so, what is the identity element for vector addition in this space?

Exercise 58 Let $V = \mathbb{R}$ and let \circ be an operation on \mathbb{R} defined by $a \circ b = a^3b$. Is V , together with the usual addition and “scalar multiplication” given by \circ , a vector space over \mathbb{R} ?

Exercise 59 Show that \mathbb{Z} cannot be turned into a vector space over any field.

Exercise 60 Let V be a vector space over the field $GF(2)$. Show that $v = -v$ for all $v \in V$.

Exercise 61 Give an example of a vector space having exactly 125 elements.

Exercise 62 In the definition of a vector space, show that the commutativity of vector addition is a consequence of the other conditions.

Exercise 63 Let W be the subset of \mathbb{R}^5 consisting of all vectors an odd number of the entries in which are equal to 0. Is W a subspace of \mathbb{R}^5 ?

Exercise 64 Let $V = \mathbb{R}^{\mathbb{R}}$ and let W be the subset of V containing the constant function $x \mapsto 0$ and all of those functions $f \in V$ satisfying the condition that $f(a) = 0$ for at most finitely-many real numbers a . Is W a subspace of V ?

Exercise 65 Let $V = \left\{ \begin{bmatrix} a_1 \\ \vdots \\ a_5 \end{bmatrix} \mid 0 < a_i \in \mathbb{R} \right\}$. If $v = \begin{bmatrix} a_1 \\ \vdots \\ a_5 \end{bmatrix}$ and

$w = \begin{bmatrix} b_1 \\ \vdots \\ b_5 \end{bmatrix}$ belong to V , and if $c \in \mathbb{R}$, set $v + w = \begin{bmatrix} a_1 + b_1 \\ \vdots \\ a_5 + b_5 \end{bmatrix}$ and

$cv = \begin{bmatrix} ca_1 \\ \vdots \\ ca_5 \end{bmatrix}$. Do these operations turn V into a vector space over \mathbb{R} ?

Exercise 66 Let $F = GF(3)$. How many elements are there in the sub-

space of F^3 generated by $\left\{ \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix}, \begin{bmatrix} 2 \\ 2 \\ 1 \end{bmatrix} \right\}$?

Exercise 67 A function $f \in \mathbb{R}^{\mathbb{R}}$ is **piecewise constant** if and only if it is a constant function $x \mapsto c$ or there exist $a_1 < a_2 < \dots < a_n$ and c_0, \dots, c_n in \mathbb{R} such that

$$f : x \mapsto \begin{cases} c_0 & \text{if } x < a_1 \\ c_i & \text{if } a_i \leq x < a_{i+1} \text{ for } 1 \leq i < n \\ c_n & \text{if } a_n \leq x \end{cases}.$$

Does the set of all piecewise constant functions form a subspace of the vector space $\mathbb{R}^{\mathbb{R}}$ over \mathbb{R} ?

Exercise 68 Let V be the vector space of all continuous functions from \mathbb{R} to itself and let W be the subset of all those functions $f \in V$ satisfying the condition that $|f(x)| \leq 1$ for all $-1 \leq x \leq 1$. Is W a subspace of V ?

Exercise 69 Let $F = GF(2)$ and let W be the subspace of F^5 con-

sisting of all vectors $\begin{bmatrix} a_1 \\ \vdots \\ a_5 \end{bmatrix}$ satisfying $\sum_{i=1}^5 a_i = 0$. Is W a subspace

of F^5 ?

Exercise 70 Let $V = \mathbb{R}^{\mathbb{R}}$ and let W be the subset of V consisting of all monotonically-increasing or monotonically-decreasing functions. Is W a subspace of V ?

Exercise 71 Let $V = \mathbb{R}^{\mathbb{R}}$ and let W be the subset of V consisting of the constant function $a \mapsto 0$, and all epic functions. Is W a subspace of V ?

Exercise 72 Let $V = \mathbb{R}^{\mathbb{R}}$ and let W be the subset of V containing the constant function $a \mapsto 0$ and all of those functions $f \in V$ satisfying the condition that $f(\pi) > f(-\pi)$. Is W a subspace of V ?

Exercise 73 Let $V = \mathbb{R}^{\mathbb{R}}$ and let W be the subset of V consisting of all functions f satisfying the condition that there exists a real number c (which depends on f) such that $|f(a)| \leq c|a|$ for all $a \in \mathbb{R}$. Is W a subspace of V ?

Exercise 74 Let $V = \mathbb{R}^{\mathbb{R}}$ and let W be the subset of V consisting of all functions f satisfying the condition that there exist real numbers a and b such that $|f(x)| \leq a|\sin(x)| + b|\cos(x)|$ for all $x \geq 0$. Is W a subspace of V ?

Exercise 75 Let F be a field and let $V = F^F$, which is a vector space over F . Let W be the set of all functions $f \in V$ satisfying $f(1) = f(-1)$. Is W a subspace of V ?

Exercise 76 For any real number $0 < t \leq 1$, let V_t be the set of all functions $f \in \mathbb{R}^{\mathbb{R}}$ satisfying the condition that if $a < b$ in \mathbb{R} then there exists a real number $u(a, b)$ satisfying $|f(x) - f(y)| \leq u(a, b)|x - y|^t$ for all $a \leq x, y \leq b$. For which values of t is V_t a subspace of $\mathbb{R}^{\mathbb{R}}$?

Exercise 77 Let U be a nonempty subset of a vector space V . Show that U is a subspace of V if and only if $au + u' \in U$ for all $u, u' \in U$ and $a \in F$.

Exercise 78 Let V be a vector space over a field F and let v and w be distinct vectors in V . Set $U = \{(1-t)v + tw \mid t \in F\}$. Show that there exists a vector $y \in V$ such that $\{u + y \mid u \in U\}$ is a subspace of V .

Exercise 79 Let V be a vector space over a field F and let W and Y be subspaces of V^2 . Let U be the set of all vectors $\begin{bmatrix} v \\ v' \end{bmatrix} \in V^2$ satisfying the condition that there exists a vector $v'' \in V$ such that $\begin{bmatrix} v \\ v'' \end{bmatrix} \in W$ and $\begin{bmatrix} v'' \\ v' \end{bmatrix} \in Y$. Is U a subspace of V^2 ?

Exercise 80 Consider \mathbb{R} as a vector space over \mathbb{Q} . Given a nonempty subset W of \mathbb{R} , let \overline{W} be the set of all real numbers b for which there exists a sequence a_1, a_2, \dots of elements of W satisfying $\lim_{i \rightarrow \infty} a_i = b$. Show that \overline{W} is a subspace of \mathbb{R} whenever W is.

Exercise 81 Let V be a vector space over a field F and let P be the collection of all subsets of V , which we know is a vector space over $GF(2)$. Is the collection of all subspaces of V a subspace of P ?

Exercise 82 Let W be the set of all functions $f \in \mathbb{R}^{\mathbb{N}}$ satisfying the condition that if $f(i) \neq 0$ then $f(ji) \neq 0$ for all positive integers j . Is W a subspace of $\mathbb{R}^{\mathbb{N}}$?

Exercise 83 Let W be the set of all functions $f \in \mathbb{R}^{\mathbb{N}}$ satisfying the condition that if $f(i) \neq 0$ then $f(ji) = 0$ for all positive integers j . Is W a subspace of $\mathbb{R}^{\mathbb{N}}$?

Exercise 84 Let V be a vector space over a field F and let Y be the set of all matrices of the form
$$\begin{bmatrix} v_1 & v_2 & 0_V \\ 0_V & v_1 + v_2 & 0_V \\ 0_V & v_1 & v_2 \end{bmatrix} \quad \text{in } \mathcal{M}_{3 \times 3}(V).$$
 Is Y a subspace of $\mathcal{M}_{3 \times 3}(V)$?

Exercise 85 Let W be the set of all functions $f \in \mathbb{R}^{\mathbb{R}}$ satisfying the following conditions: there exist positive real numbers a and b such that for all $x \in \mathbb{R}$ satisfying $|x| \geq a$ we have $|f(x)| \leq b|x|$. Show that W is a subspace of $\mathbb{R}^{\mathbb{R}}$.

Exercise 86 Let W be a subspace of a vector space V over a field F . Is the set $(V \setminus W) \cup \{0_V\}$ necessarily a subspace of V ?

Exercise 87 Let V be a vector space over a field F and let f be a function from V to the unit interval $[0, 1]$ on the real line satisfying the condition that $f(au + bv) \geq \min\{f(u), f(v)\}$ for all $a, b \in F$ and all $u, v \in V$. Show that $f(0_V) \geq f(v)$ for all $v \in V$ and that if $0 \leq h \leq f(0_V)$ then $V_h = \{v \in V \mid f(v) \geq h\}$ is a subspace of V .

Exercise 88 Find subsets D and D' of \mathbb{R}^3 such that $\mathbb{R}(D \cap D') \neq \mathbb{R}D \cap \mathbb{R}D'$.

Exercise 89 Find subspaces W and Y of \mathbb{R}^3 having the property that $W \cup Y$ is not a subspace of \mathbb{R}^3 .

Exercise 90 Let V be a vector space over a field F and let $0_V \neq w \in V$. Given a vector $v \in V \setminus Fw$, find the set G of all scalars $a \in F$ satisfying $F\{v, w\} = F\{v, aw\}$.

Exercise 91 Let p be a prime integer and let V be a vector space over $F = GF(p)$. Show that V is not the union of k subspaces, for any $k \leq p$.

Exercise 92 Let V be a vector space over a field F and let c and d be fixed elements of F . Define a new operation \boxplus on V by setting $v \boxplus v' = cv + dv'$. Is V , with this new vector addition and the old scalar multiplication, still a vector space over F ?

4

Algebras over a field

In general, a vector space does not carry with it the notion of multiplying two vectors in the space to produce a third vector. However, sometimes such multiplication may be possible. A vector space K over a field F is an **F -algebra** if and only if there exists a function $(v, w) \mapsto v \bullet w$ from $K \times K$ to K such that

- (1) $u \bullet (v + w) = u \bullet v + u \bullet w$;
- (2) $(u + v) \bullet w = u \bullet w + v \bullet w$;
- (3) $a(v \bullet w) = (av) \bullet w = v \bullet (aw)$

for all $u, v, w \in K$ and $a \in F$. As in the proof of Proposition 2.3(1), these conditions suffice to show that $0_K \bullet v = v \bullet 0_K = 0_K$ for all $v \in K$.

Note that the operation \bullet need not be associative, nor need there exist an identity element for this operation. When the operation is associative, i.e. when it satisfies

- (4) $v \bullet (w \bullet y) = (v \bullet w) \bullet y$

for all $v, w, y \in K$, then the algebra is called an **associative F -algebra**. If an identity element for \bullet exists, that is to say, if there exists an element $0_K \neq e \in K$ satisfying $v \bullet e = v = e \bullet v$ for all $v \in K$, we say the F -algebra K is **unital**. In a unital F -algebra, as with the case of fields, the identity element must be unique. In this case, we can then identify F with the subset $\{ae \mid a \in F\}$ of K and we note that $a \bullet v = v \bullet a$ for all $v \in K$ and $a \in F$.

If v is an element of an associative F -algebra (K, \bullet) and if n is a positive integer, we write v^n instead of $v \bullet \dots \bullet v$ (n factors). If K

is also unital and has a multiplicative identity e , we set $v^0 = e$ for all $0_K \neq v \in K$. The element $(0_K)^0$ is not defined.

If $v \bullet w = w \bullet v$ for all v and w in some F -algebra K , then the algebra is **commutative**. An F -algebra (K, \bullet) satisfying the condition that $v \bullet w = -w \bullet v$ for all $v, w \in K$ is **anticommutative**. If the characteristic of F is other than 2, it is easy to see that this condition is equivalent to the condition that $v \bullet v = 0_K$ for all $v \in K$. Of course, in that case K cannot possibly be unital.¹

If (K, \bullet) is an associative unital F -algebra having a multiplicative identity e , and if $v \in K$ satisfies the condition that there exists an element $w \in K$ such that $v \bullet w = w \bullet v = e$, then we say that v is a **unit** of K . As with the case of fields, such an element w , if it exists, is unique and is usually denoted by v^{-1} . If v is a unit, then so is $-v$, for one immediately notes that $(-v)^{-1} = -(v^{-1})$. Also, it is easy to see that if v and w are units of K , then so is $v \bullet w$. Indeed,

$$\begin{aligned} (v \bullet w) \bullet (w^{-1} \bullet v^{-1}) &= (v \bullet (w \bullet w^{-1})) \bullet v^{-1} \\ &= (v \bullet e) \bullet v^{-1} = v \bullet v^{-1} = e \end{aligned}$$

and similarly $(w^{-1} \bullet v^{-1}) \bullet (v \bullet w) = e$, and so $(v \bullet w)^{-1} = w^{-1} \bullet v^{-1}$. If $v \in K$ is a unit and if $n > 1$ is an integer, we write v^{-n} instead of $(v^{-1})^n$.

Example: Any vector space V over a field F can be turned into an associative and commutative F -algebra which is not unital by setting $v \bullet w = 0_V$ for all $v, w \in V$.

Example: If F is a subfield of a field K , then K has the structure of an associative F -algebra, with multiplication being the multiplication in K . Thus, \mathbb{C} is an \mathbb{R} -algebra and $\mathbb{Q}(\sqrt{p})$ is a \mathbb{Q} -algebra for every prime integer p . These algebras are, of course, unital.



The first systematic study of associative algebras was initiated by the 19th century American mathematician **Benjamin Peirce** and continued by his son, the mathematician and logician **Charles Sanders Peirce**. Other major contributors at the beginning of the 20th century were the American mathematician **Leonard Dickson** and the Scottish mathematician **Joseph Henry Wedderburn**.

Example: Let F be a field, let (K, \bullet) be an F -algebra, and let Ω be a nonempty set. Then the vector space K^Ω of all functions from Ω to K has the structure of an F -algebra with respect to the operation \bullet defined by $f \bullet g : i \mapsto f(i) \bullet g(i)$ for all $i \in \Omega$. This F -algebra is associative if K is. If K is unital with multiplicative identity element e , then K^Ω is also unital, with identity element given by the constant function $i \mapsto e$. In particular, if F is a field and if Ω is a nonempty set then F^Ω is an associative unital F -algebra with respect to the operation \bullet defined by $f \bullet g : i \mapsto f(i)g(i)$ for all $i \in \Omega$.

Example: We have already seen that the collection of all subsets of a given nonempty set A is a vector space over $GF(2)$. It is in fact an associative and commutative unital $GF(2)$ -algebra with respect to the operation \cap . The identity element with respect to this operation is A itself.

Example: Let K be the vector space over \mathbb{R} consisting of all functions in $\mathbb{R}^{\mathbb{R}}$ which are infinitely differentiable, and define an operation \bullet on K by setting $f \bullet g = (fg)'$ (where $'$ denotes differentiation). Then (K, \bullet) is an algebra which is commutative but not associative.

The collection of all operations \bullet on a vector space V over a field F which turn V into an F -algebra will be studied in more detail in Chapter 20.

Let F be a field. If (K, \bullet) is an F -algebra, then a subspace L of K satisfying the condition that $w \bullet w' \in L$ for all $w, w' \in L$ is an **F -subalgebra** of K . If (K, \bullet) is a unital F -algebra, then L is a **unital subalgebra** if it contains the multiplicative identity element of K .

Let F be a field. An anticommutative F -algebra (K, \bullet) is a **Lie algebra**² over F if and only if it satisfies the additional condition



2

Sophus Lie was a 19th-century Norwegian mathematician who developed mathematical concepts that provide the basic model for quantum theory and an important tool in differential geometry. Another pioneer in the study of noncommutative algebras because of their importance in physics was 20th-century British mathematician **Dudley Littlewood**.

(Jacobi identity) $u \bullet (v \bullet w) + v \bullet (w \bullet u) + w \bullet (u \bullet v) = 0_K$;

for all $u, v, w \in K$. This algebra is not associative unless $u \bullet v = 0_K$ for all $u, v \in K$.

Example: Let F be a field and let $(K, *)$ be an associative F -algebra. Define a new operation \bullet on K by setting $v \bullet w = v * w - w * v$. Then (K, \bullet) is a Lie algebra over F , which is usually denoted by K^- . The operation in K^- is known as the **Lie product** defined on the given F -algebra K . This example is very important, because one can show that any Lie algebra over a field F can be considered as a subalgebra of a Lie algebra of the form K^- for some associative F -algebra K . (A proof of this result, known as the **Poincaré-Birkhoff-Witt Theorem**, is far beyond the scope of this book.)

If $v, w \in K$, then $v \bullet w = 0_K$ precisely when $v * w = w * v$, in other words, precisely when (v, w) forms a **commuting pair**. Of course, if the algebra K is commutative, all pairs of elements commute, but in general that will not be the case.

Lie algebras are of fundamental importance in the modeling problems in physics, and have many other applications; they are in the forefront of current mathematical research. One particular Lie algebra defined on \mathbb{R}^3 goes back to the work of Grassmann. Define the structure of an \mathbb{R} -algebra on \mathbb{R}^3 with multiplication \times given by

$$\begin{bmatrix} a_1 \\ a_2 \\ a_3 \end{bmatrix} \times \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix} = \begin{bmatrix} a_2 b_3 - a_3 b_2 \\ a_3 b_1 - a_1 b_3 \\ a_1 b_2 - a_2 b_1 \end{bmatrix}.$$

This operation, called the **cross product**, has very important applications in physics and engineering. It is easy to check that the algebra (\mathbb{R}^3, \times) is a Lie algebra over \mathbb{R} .

Note that if $v_1 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$, $v_2 = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$, and $v_3 = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$, then, surely, $v_1 \times v_2 = v_3$, $v_1 \times v_3 = -v_2$, and $v_2 \times v_3 = v_1$. Moreover, the cross product is the only possible anticommutative product which can be defined on \mathbb{R}^3 and which satisfies this condition. Indeed, if \bullet is any such product defined on \mathbb{R}^3 then

$$\begin{aligned} \begin{bmatrix} a_1 \\ a_2 \\ a_3 \end{bmatrix} \bullet \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix} &= \left(\sum_{i=1}^3 a_i v_i \right) \left(\sum_{j=1}^3 b_j v_j \right) = \sum_{i=1}^3 \sum_{j=1}^3 a_i b_j (v_i \bullet v_j) \\ &= \begin{bmatrix} a_2 b_3 - a_3 b_2 \\ a_3 b_1 - a_1 b_3 \\ a_1 b_2 - a_2 b_1 \end{bmatrix} = \begin{bmatrix} a_1 \\ a_2 \\ a_3 \end{bmatrix} \times \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix}. \end{aligned}$$

(4.1) Proposition: If v and w are nonzero elements of \mathbb{R}^3 , then $v \times w = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$ if and only if $\mathbb{R}v = \mathbb{R}w$.

Proof: Suppose $v = \begin{bmatrix} a_1 \\ a_2 \\ a_3 \end{bmatrix}$ and $w = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix}$. These vectors are nonzero and so one of the entries in w is nonzero; without loss of generality, we can assume that $b_1 \neq 0$. Then $a_2b_3 - a_3b_2 = a_3b_1 - a_1b_3 = a_1b_2 - a_2b_1 = 0$ and so, if we define $c = a_1b_1^{-1}$, we have $v = cw$. Hence $v \in \mathbb{R}w$. Moreover, $c \neq 0$ so $w = c^{-1}v \in \mathbb{R}v$, proving the desired equality. Conversely, if $\mathbb{R}v = \mathbb{R}w$ then there exists an $0 \neq d \in \mathbb{R}$ such that

$$w = dv. \text{ Then } v \times w = d(v \times v) = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}. \quad \square$$

The cross product is very particular to the vector space \mathbb{R}^3 , and does not generalize easily to spaces of the form \mathbb{R}^n for $n > 3$, with the exception of $n = 7$, which we will see in a later chapter.

An important non-associative algebra is the following: let F be a field of characteristic other than 2, and let $(K, *)$ be an associative algebra. We can define a new operation \bullet on K , called the **Jordan product**, by setting $v \bullet w = \frac{1}{2}(v * w + w * v)$. Then (K, \bullet) is a commutative F -algebra, usually denote by K^+ , called the **Jordan algebra** defined by K . It is not associative in general, but does satisfy the

$$(\text{Jordan identity}) \quad (v \bullet w) \bullet (v \bullet v) = v \bullet (w \bullet (v \bullet v))$$

for all $v, w \in K$. Jordan algebras have important applications in physics³. Note that if $v * w = w * v$, then $v \bullet w = v * w$. This observation will have important consequences later.



3

Jordan algebras were developed by 20th-century German physicist **Pascual Jordan**, one of the fathers of quantum mechanics and quantum electrodynamics. The algebraic structure of Lie algebras and Jordan algebras was studied in detail by 20th-century American mathematicians **Nathan Jacobson** and **A. Adrian Albert**.

We now come to an extremely important algebra. Let F be a field and let X be an element not in F , which we will call an **indeterminate**. A **polynomial** in X with coefficients in F is a formal sum $f(X) = \sum_{i=0}^{\infty} a_i X^i$, in which the elements a_i belong to F , and no more than a finite number of these elements differ from 0. The elements a_i are called the **coefficients** of the polynomial. If all of the a_i equal 0, then the polynomial is called the **0-polynomial**. Otherwise, there exists a nonnegative integer n satisfying the condition that $a_n \neq 0$ and $a_i = 0$ for all $i > n$. The coefficient a_n is called the **leading coefficient** of the polynomial; the integer n is called the **degree** of the polynomial, and is denoted by $\deg(f)$. If the leading coefficient of a polynomial is 1, the polynomial is **monic**. The degree of the 0-polynomial is defined to be $-\infty$, where we assume that $-\infty < i$ for each integer i and $(-\infty) + i = -\infty$ for all integers i . If $f(X)$ is a polynomial of degree $n \neq -\infty$, we often write it as $\sum_{i=0}^n a_i X^i$. The set of all polynomials in X with coefficients in F is denoted by $F[X]$. We identify the elements of F with the polynomials of degree at most 0, and so can consider F as a subdomain of $F[X]$. We can associate the 0-polynomial with the identity element 0 of F for addition and the polynomials of degree 0 with the nonzero elements of F and so, without any problems, consider F as a subset of $F[X]$.

Example: The polynomials $5X^3 + 2X^2 + 1$ and $5X^3 - X^2 + X + 4$ in $\mathbb{Q}[X]$ both have degree 3 and leading coefficient 5. Therefore they are not monic. The polynomials $X^3 + 2X^2 + 1$ and $X^3 - X^2 + X + 4$ in $\mathbb{Q}[X]$ are both monic and have the same degree 3.

We define addition and multiplication of polynomials over a field as follows: if $f(X) = \sum_{i=0}^{\infty} a_i X^i$ and $g(X) = \sum_{i=0}^{\infty} b_i X^i$ are polynomials in $F[X]$, then $f(X) + g(X)$ is the polynomial $\sum_{i=0}^{\infty} c_i X^i$, where $c_i = a_i + b_i$ for all $i \geq 0$ and $f(X)g(X)$ is the polynomial $\sum_{i=0}^{\infty} d_i X^i$, where $d_i = \sum_{j=0}^i a_j b_{i-j}$ for all $i \geq 0$. It is easy to verify that these definitions turn $F[X]$ into an associative and commutative unital F -algebra with the 0-polynomial acting as the identity element for addition and the degree-0 polynomial 1 acting as the identity element for multiplication. This algebra also has another important property, namely that the product of two nonzero elements of $F[X]$ is again nonzero. An algebra having this property is said to be **entire**.

If $f(X) = \sum_{i=0}^{\infty} a_i X^i$ and $g(X) = \sum_{i=0}^{\infty} b_i X^i$ are polynomials in $F[X]$ then we define the polynomial $f(g(X))$ to be $\sum_{i=0}^{\infty} a_i g(X)^i$. Then, for any fixed $g(X)$, the set $F[g(X)] = \{f(g(X)) \mid f(X) \in F[X]\}$ is a unital subalgebra of $F[X]$.

Commutative, associative, entire, unital F -algebras are integral domains. The converse of this is not true: \mathbb{Z} is an integral domain which is not an F -algebra for any field F .

Note that every polynomial in $F[X]$ is a linear combination of elements of the set $B = \{1, X, X^2, \dots\}$ over F , so B is a set of generators of $F[X]$ over F . On the other hand, it is clear that no finite set of polynomials can be a generating set for $F[X]$ over F , and so $F[X]$ is not finitely generated as a vector space over F .

We should remark that the formal definition of multiplication of polynomials does not translate into the fastest method of carrying out such multiplication in practice on a computer, especially for polynomials of large degree. The problem of fast polynomial multiplication has been the subject of extensive research over the years, and many interesting algorithms to perform such multiplication have been devised. A typical such algorithm is **Karatsuba's algorithm**⁴, which is easy to implement on a computer: let $f(X)$ and $g(X)$ be polynomials in $F[X]$, where F is a field. We can write these polynomials as $f(X) = \sum_{i=0}^n a_i X^i$ and $g(X) = \sum_{i=0}^n b_i X^i$, where n is a nonnegative power of 2 satisfying $n \geq \max\{\deg(f), \deg(g)\}$. (Of course, in this case a_n and b_n may equal 0.) We now calculate $f(X)g(X)$ as follows:

(1) If $n = 1$ then $f(X)g(X) = a_1 b_1 X^2 + (a_0 b_1 + a_1 b_0)X + a_0 b_0$.

(2) Otherwise, write $f(X) = f_1(X)X^{n/2} + f_0(X)$ and

$$g(X) = g_1(X)X^{n/2} + g_0(X),$$

where the polynomials $f_0(X)$, $f_1(X)$, $g_0(X)$, and $g_1(X)$ are all of degree at most $n/2$.

(3) Recursively, calculate $f_0(X)g_0(X)$, $f_1(X)g_1(X)$, and

$$(f_0 + f_1)(X)(g_0 + g_1)(X).$$

(4) Then

$$\begin{aligned} f(X)g(X) &= X^n(f_1 g_1)(X) + \\ &\quad X^{n/2}[(f_0 + f_1)(g_0 + g_1) - f_0 g_0 - f_1 g_1](X) + \\ &\quad (f_0 g_0)(X). \end{aligned}$$



4

Anatoli Karatsuba is a contemporary Russian mathematician whose research is primarily in number theory.

Indeed, if the multiplication of two polynomials of degree at most n using the definition of polynomial multiplication takes an order of $2n^2$ arithmetic operations (i.e., additions and multiplications), it is possible to prove that there exists a fixed positive integer c such that the multiplication of two polynomials of degree at most n using Karatsuba's algorithm takes at most $cn^{1.59}$ arithmetic operations. If n is sufficiently large, the difference between these two bounds can be significant.

There are other highly-sophisticated algorithms for multiplying two polynomials of degree at most n in an order of $n \log(n)$ arithmetic operations.

(4.2) Proposition (Division Algorithm): If F is a field and if $f(X)$ and $g(X) \neq 0$ are elements of $F[X]$, then there exist unique polynomials $u(X)$ and $v(X)$ in $F[X]$ satisfying $f(X) = g(X)u(X) + v(X)$ and $\deg(v) < \deg(g)$.

Proof: Assume that $f(X) = \sum_{i=0}^{\infty} a_i X^i$ and $g(X) = \sum_{i=0}^{\infty} b_i X^i$ are the given polynomials. If $f(X) = 0$ or if $\deg(f) < \deg(g)$, choose $u(X) = 0$ and $v(X) = f(X)$, and we are done. Thus we can assume that $n = \deg(f) \geq \deg(g) = k$, and will prove our result by induction on n . If $n = 0$ then $k = 0$ and therefore we can choose $u(X)$ to be $a_0 b_0^{-1}$, which is a polynomial of degree 0, and choose $v(X)$ to be the 0-polynomial. Now assume, inductively, that $n > 0$ and that the proposition has been established for all functions $f(X)$ of degree less than n . Set $h(X) = f(X) - a_n b_k^{-1} X^{n-k} g(X)$. If this is the 0-polynomial, choose $u(X) = a_n b_k^{-1} X^{n-k}$ and let $v(X)$ be the 0-polynomial. Otherwise, since $\deg(f) > \deg(h)$, we see by the induction hypothesis that there exist polynomials $v(X)$ and $w(X)$ in $F[X]$ satisfying $h(X) = g(X)w(X) + v(X)$, where $\deg(g) > \deg(v)$. Thus $f(X) = [a_n b_k^{-1} X^{n-k} + w(X)]g(X) + v(X)$, as required.

We are left to show uniqueness. Indeed, assume that

$$f(X) = g(X)u_1(X) + v_1(X) = g(X)u_2(X) + v_2(X),$$

where $\deg(v_1) < \deg(g)$ and $\deg(v_2) < \deg(g)$. Then

$$g(X)[u_1(X) - u_2(X)] + [v_1(X) - v_2(X)]$$

equals the 0-polynomial. If $u_1(X) = u_2(X)$ then $v_1(X) = v_2(X)$ and we are done. Therefore assume that $u_1(X) \neq u_2(X)$. But then, since $\deg(g[u_1 - u_2]) > \deg(v_1 - v_2)$ and since $F[X]$ is entire, this is a contradiction. Thus we have established uniqueness. \square

Let us emphasize that the set $F[X]$ is composed of formal expressions and not functions. Every polynomial $f(X) = \sum_{i=0}^{\infty} a_i X^i \in F[X]$ defines a corresponding **polynomial function** in F^F given by $c \mapsto f(c) =$

$\sum_{i=0}^{\infty} a_i c^i$, but the correspondence between polynomials and polynomial functions is not bijective. Indeed, it is possible for two distinct polynomials to define the same polynomial function. Thus, for example, if $F = GF(2)$ then the distinct polynomials X, X^2, X^3, \dots all define the same function from F to itself, namely the function given by $0 \mapsto 0$ and $1 \mapsto 1$. The **degree** of a polynomial function is the least of the degrees of the (perhaps many) polynomials which define that function.⁵

Let $p_1(X)$ and $p_2(X)$ be polynomials in $F[X]$ and let $c \in F$. If we set $f(X) = p_1(X) + p_2(X)$ and $g(X) = p_1(X)p_2(X)$ then it is clear that $f(c) = p_1(c) + p_2(c)$ and $g(c) = p_1(c)p_2(c)$.

(4.3) Proposition: Let F be a field and let $p(X)$ be a polynomial in $F[X]$. Then an element c of F satisfies the condition that $p(c) = 0$ if and only if there exists a polynomial $u(X) \in F[X]$ satisfying $p(X) = (X - c)u(X)$.

Proof: By Proposition 4.2, we know that there exist polynomials $u(X)$ and $v(X)$ in $F[X]$ satisfying $p(X) = (X - c)u(X) + v(X)$, where $\deg(v) < \deg(X - c) = 1$. Therefore $v(X) = b$ for some $b \in F$. If $b = 0$ then $p(c) = (c - c)u(c) = 0$. Conversely, if $p(c) = 0$ then $0 = p(c) = (c - c)u(c) + b = b$ and so $p(X) = (X - c)u(X)$. \square

As an immediate consequence of this result, we see that if F is a field and if $p(X) \in F[X]$, then the set of all elements c of F satisfying $p(c) = 0$ is finite and, indeed, cannot exceed the degree of $p(X)$.

Let F be a field. A polynomial $p(X) \in F[X]$ is **reducible** if and only if there exist polynomials $u(X)$ and $v(X)$ in $F[X]$, each of degree at least 1, satisfying $p(X) = u(X)v(X)$. Otherwise, the polynomial is **irreducible**. Many tests for the irreducibility of polynomials in $\mathbb{Q}[X]$ have been devised. One of the earliest and well-known is **Eisenstein's**



⁵ The first person to systematically consider the best methods of calculating $f(c)$ for a polynomial $f(X) \in F[X]$ and for $c \in F$, was the twentieth-century Russian mathematician **Alexander Ostrowski**.

criterion:⁶ if $p(X) = \sum_{i=0}^n a_i X^i \in \mathbb{Q}[X]$, where each a_i is an integer, and if there exists a prime integer q such that q does not divide a_n , q divides a_i for all $0 \leq i \leq n-1$, and q^2 does not divide a_0 , then $p(X)$ is irreducible. (A proof of this can be found in books on abstract algebra.) Thus, using this criterion, we see that $3X^3 + 7X^2 + 49X - 7$ is an irreducible polynomial in $\mathbb{Q}[X]$.

Example: If $F = GF(5)$ then the polynomial $X^3 + X + 1 \in F[X]$ is irreducible, a fact which can be established, if necessary, by testing all possibilities. However, when $F = GF(3)$ it is easy to verify the factorization $X^3 + X + 1 = (X + 2)(X^2 + X + 2)$, and thus see that the polynomial is reducible.

Example: If $p(X) = u(X)v(X)$ in $F[X]$, then surely $p(X + c) = u(X + c)v(X + c)$ for any $c \in F$, and so to prove that a polynomial $p(X)$ is irreducible it suffices to prove that $p(X + c)$ is irreducible for some $c \in F$. For example, let q be a prime integer. The q th **cyclotomic polynomial** in $\mathbb{Q}[X]$ is defined to be $\Phi_q(X) = \sum_{i=0}^{q-1} X^i$. We claim that this polynomial is irreducible. To see that this is so, we note that

$$\Phi_q(X + 1) = X^{q-1} + \sum_{i=0}^{q-2} \binom{q}{q-1-i} X^i,$$

which is irreducible by Eisenstein's criterion.

It is known that the number of monic irreducible polynomials of positive degree m in $GF(p)$ equals $N(p) = \frac{1}{m} \sum \mu(d) p^{m/d}$, where the sum ranges over all integers d which divide m and the **Möbius function** $\mu(d)$ is defined by

$$\mu(d) = \begin{cases} 1 & \text{if } d = 1 \\ (-1)^k & \text{if } d \text{ is the product of } k \text{ distinct primes} \\ 0 & \text{otherwise} \end{cases}.$$

This means that the probability of a randomly-selected monic polynomial of degree m in $GF[X]$ being irreducible is $N(p)/p^m$, which is roughly $\frac{1}{m}$.



⁶ Gauss' brilliant student, **Ferdinand Eisenstein**, died of tuberculosis at the age of 29.

Any polynomial in $F[X]$ can be written as a product of irreducible polynomials. How to find such a decomposition, especially in the case of polynomials over a finite field or over \mathbb{Q} , is a very difficult and important problem, which attracted such great mathematicians as Newton and which continues to attract many important mathematicians until this day. Indeed, the problem of factoring polynomials over finite fields into irreducible components has become even more important, since it is the basis for many current cryptographic schemes. There are algorithms, such as Berlekamp's algorithm, which factor a polynomial $f(X) \in F[X]$, where $F = GF(p^n)$, in a time polynomial in p , n , and $\deg(f)$. Moreover, under various assumptions, such as the Generalized Riemann Hypothesis, polynomials of special forms can be factored much more rapidly.

A polynomial $p(X) \in F[X]$ of positive degree is **completely reducible** if and only if it can be written as a product of polynomials in $F[X]$ of degree 1. Not every polynomial over every field is completely reducible. For example, the polynomial $X^2 + 1 \in \mathbb{Q}[X]$ is not completely reducible. The field F is **algebraically closed** if every polynomial of positive degree in $F[X]$ is completely reducible. The fields \mathbb{Q} and \mathbb{R} are not algebraically closed. The field \mathbb{C} is algebraically closed, by a theorem known as the **Fundamental Theorem of Algebra**. This theorem is in fact analytic and not algebraic, and relies on various analytic properties of functions of a complex variable. Most of the great mathematicians of the eighteenth century – d'Alembert, Euler, Laplace, Lagrange, Argand, Cauchy, and others – tried in vain to prove this theorem. The first proof was given by Gauss in his doctoral thesis in 1799. His proof was basically topological and relied on work of Euler. During his lifetime, Gauss published several proofs of this theorem⁷.

Example: The field $F = GF(2)$ is not algebraically closed since the polynomial $X^2 + X + 1 \in F[X]$ is not completely reducible.

Note that if a field F is algebraically closed then every polynomial function $F \rightarrow F$ defined by a polynomial of positive degree is epic. Indeed, let $p(X) \in F[X]$ be a polynomial of positive degree and let $d \in F$. Then



⁷ Most proofs of the Fundamental Theorem of Algebra are existence proofs and do not give a constructive method of finding the degree-one factors of a polynomial over an algebraically-closed field. The first constructive proof was given by the German mathematician **Helmut Kneser** in 1940.

$q(X) = p(X) - d$ is a polynomial of positive degree in $F[X]$ and so there exists an element c of F such that $q(c) = 0$. In other words, $p(c) = d$.

(4.4) Proposition: A monic polynomial $p(X) \in \mathbb{R}[X]$ is irreducible if and only if it is of the form $X - a$ or $(X - a)^2 + b^2$, where $a \in \mathbb{R}$ and $0 \neq b \in \mathbb{R}$.

Proof: Clearly every polynomial of the form $X - a$ is irreducible. Now assume that $f(X) = (X - a)^2 + b^2 = X^2 - 2aX + a^2 + b^2$. Were this polynomial reducible, we could find real numbers c and d satisfying

$$f(X) = (X - c)(X - d) = X^2 - (c + d)X + cd$$

and so $c + d = 2a$ and $cd = a^2 + b^2$. This implies that $c^2 - 2ac + a^2 + b^2 = 0$ and hence $c = \frac{1}{2} \left[2a \pm \sqrt{4a^2 - 4(a^2 + b^2)} \right] = a \pm \sqrt{-b^2}$, which contradicts the assumption that $c \in \mathbb{R}$ since b is assumed to be nonzero. Thus polynomials of both of the given forms are indeed irreducible.

Conversely, let $p(X) = \sum_{i=0}^n c_i X^i$ be a monic irreducible polynomial in $\mathbb{R}[X]$ that is not of the form $X - a$. By the Fundamental Theorem of Algebra we know that there exists a complex number $z = a + bi$ satisfying $p(z) = 0$. Since the coefficients of $p(X)$ are real, this means that $p(\bar{z}) = 0$ as well, since $0 = \bar{0} = \overline{p(z)} = \sum_{i=0}^n c_i \bar{z}^i = p(\bar{z})$. Thus there exists a polynomial $u(X) \in \mathbb{R}[X]$ satisfying $p(X) = (X - z)(X - \bar{z})u(X)$, where $(X - z)(X - \bar{z}) = X^2 - (z + \bar{z})X + z\bar{z} = X^2 - 2aX + a^2 + b^2$. Since $p(X)$ was assumed irreducible, we conclude that $z \notin \mathbb{R}$ (i.e. $b \neq 0$) and that $p(X)$ equals $X^2 - 2aX + a^2 + b^2$, as desired. \square

An obvious generalization of the above construction is the following: let F be a field and let (K, \bullet) be an F -algebra. If X is an element not in K , we can define a **polynomial** with coefficients in K as a formal sum $f(X) = \sum_{i=0}^{\infty} a_i X^i$, in which the elements a_i belong to K and no more than a finite number of them differ from 0_K . The set of all such polynomials will be denoted by $K[X]$. As above, we define addition and multiplication in $K[X]$ as follows: if $f(X) = \sum_{i=0}^{\infty} a_i X^i$ and $g(X) = \sum_{i=0}^{\infty} b_i X^i$ belong to $K[X]$, then $f(X) + g(X)$ is the polynomial $\sum_{i=0}^{\infty} c_i X^i$, in which $c_i = a_i + b_i$ for each $0 \leq i < \infty$, and $f(X)g(X)$ is the polynomial $\sum_{i=0}^{\infty} d_i X^i$, in which $d_i = \sum_{j=0}^i a_j \bullet b_{i-j}$ for each $0 \leq i < \infty$. Again, it is easy to check that $K[X]$ is an F -algebra. This generalization allows us to consider algebras of **polynomials in several indeterminates** with coefficients in K defined inductively by setting $K[X_1, \dots, X_n] = K[X_1, \dots, X_{n-1}][X_n]$ for each $n > 1$. Elements of this algebra are of the form

$$f(X_1, \dots, X_n) = \sum a_{i_1, \dots, i_n} X_1^{i_1} \cdots X_n^{i_n},$$

where the sum ranges over all n -tuples (i_1, \dots, i_n) of nonnegative integers and at most finitely-many of the coefficients $a_{i_1, \dots, i_n} \in K$ are nonzero.

Exercises

Exercise 93 Let F be a field and let (K, \bullet) and $(L, *)$ be F -algebras.

Define an operation \diamond on $K \times L$ by $\begin{bmatrix} a \\ b \end{bmatrix} \diamond \begin{bmatrix} a' \\ b' \end{bmatrix} = \begin{bmatrix} a \bullet a' \\ b * b' \end{bmatrix}$. Is $(K \times L, \diamond)$ an F -algebra?

Exercise 94 Let F be a field and let (K, \bullet) be a unitary, associative, commutative, and entire F -algebra which, as a vector space, is finitely generated over F . Is K a field?

Exercise 95 Let F be a field and let (K, \bullet) be an associative F -algebra which, as a vector space, is finitely generated over F . Given an element $a \in K$, do there necessarily exist elements $a_1, a_2 \in K$ satisfying $a_1 \bullet a_2 = a$?

Exercise 96 Define an operation \bullet on \mathbb{R}^2 by setting

$$\begin{bmatrix} a \\ b \end{bmatrix} \bullet \begin{bmatrix} c \\ d \end{bmatrix} = \begin{bmatrix} 2ac - bd \\ ad + bc \end{bmatrix}.$$

Show that this operation turns \mathbb{R}^2 into an \mathbb{R} -algebra. Is this algebra associative?

Exercise 97 Let F be a field and let (K, \bullet) be a unital F -algebra. Define an operation \diamond on the vector space $L = K \times F$ by setting

$$\begin{bmatrix} v \\ a \end{bmatrix} \diamond \begin{bmatrix} w \\ b \end{bmatrix} = \begin{bmatrix} v \bullet w + bv + aw \\ ab \end{bmatrix}$$

for all $v, w \in K$ and $a, b \in F$. Is L an F -algebra? Is it unital?

Exercise 98 Let F be a field. An F -algebra (K, \bullet) is a **division algebra** if and only if for every $v \in K$ and for every $0_K \neq w \in K$ there exist unique vectors $x, y \in K$, not necessarily equal, satisfying $w \bullet x = v$ and $y \bullet w = v$. Is the algebra defined in the previous exercise a division algebra?

Exercise 99 Let F be a field and let (K, \bullet) be an associative F -algebra which, as a vector space, is finitely generated over F . Suppose that there exists an element $y \in K$ satisfying the condition that for each $v \in K$ there exists an element $v' \in K$ satisfying $v' \bullet y = v$. Show that each such element v' must be unique.

Exercise 100 Let F be an infinite field and let $(K, *)$ be an associative unital F -algebra. If $v, w \in K$, show that there are infinitely-many elements w' of K satisfying $v \bullet w = v \bullet w'$ in K .

Exercise 101 Let F be a field and let (K, \bullet) be an associative unital F -algebra. If A and B are subsets of K , we let $A \bullet B$ be the set of all elements of K of the form $a \bullet b$, with $a \in A$ and $b \in B$ (in particular, $\emptyset \bullet B = A \bullet \emptyset = \emptyset$). We know that the set V of all subsets of K is a vector space over $GF(2)$. Is (V, \bullet) a $GF(2)$ -algebra? If so, is it associative? Is it unital?

Exercise 102 Let F be a field and let (K, \bullet) be an associative F -algebra. If V and W are subspaces, we let $V \bullet W$ be the set of all finite sums of the form $\sum_{i=1}^n v_i \bullet w_i$, with $v_i \in V$ and $w_i \in W$. Is $V \bullet W$ necessarily a subspace of K ?

Exercise 103 For $v, w \in \mathbb{R}^3$, simplify the expression $(v + w) \times (v - w)$.

Exercise 104 For $u, v, w \in \mathbb{R}^3$, simplify the expression $(u + v + w) \times (v + w)$.

Exercise 105 Let F be a field and let (K, \bullet) be an F -algebra satisfying the Jacobi identity. Show that K is a Lie algebra if and only if $v \bullet v = 0_K$ for all $v \in K$.

Exercise 106 Let F be a field and let $(K, *)$ be an associative F -algebra. For each $0_F \neq c \in F$ and define an operation \bullet_c on K by setting $v \bullet_c w = c(v * w + w * v)$. For which values of c is (K, \bullet) a Jordan algebra over F ?

Exercise 107 Let F be a field and let (K, \bullet) be a unitary F -algebra. For each $v \in K$, let $S(v)$ be the set of all $a \in F$ satisfying the condition that $v - a1_K$ does not have an inverse with respect to the operation \bullet . If $v \in K$ has a multiplicative inverse v^{-1} with respect to this operation, show that either $S(v) = \emptyset = S(v^{-1})$ or $S(v) \neq \emptyset$ and $S(v^{-1}) = \{a^{-1} \mid a \in S(v)\}$.

Exercise 108 Let F be a field and let L be the set of all polynomials $f(X) \in F[X]$ satisfying the condition that $f(-a) = -f(a)$ for all $a \in F$. Is L a subspace of $F[X]$?

Exercise 109 Let F be a field and let L be the set of all polynomials $f(X) \in F[X]$ satisfying the condition that $\deg(f)$ is even. Is L a subspace of $F[X]$?

Exercise 110 Let F be a field and let $f(X), g(X) \in F[X]$. Show that $\deg(fg) = \deg(f) + \deg(g)$.

Exercise 111 Let F be a field and let $f(X), g(X) \in F[X]$. Show that $\deg(f + g) \leq \max\{\deg(f), \deg(g)\}$, and give an example in which we do not have equality.

Exercise 112 Find polynomials $u(X), v(X) \in \mathbb{Q}[X]$ satisfying

$$X^4 + 3X^3 = (X^2 + X + 1)u(X) + v(X).$$

Exercise 113 Let $F = GF(2)$. Find polynomials $u(X), v(X) \in F[X]$ satisfying $X^5 + X^2 = (X^3 + X + 1)u(X) + v(X)$.

Exercise 114 Let $F = GF(7)$. Find a nonzero polynomial $p(X) \in F[X]$ such that the polynomial function defined by p is the 0-function.

Exercise 115 Is the polynomial $6X^4 + 3X^3 + 6X^2 + 2X + 5 \in GF(7)[X]$ irreducible?

Exercise 116 Is the polynomial $X^7 + X^4 + 1 \in \mathbb{Q}[X]$ irreducible?

Exercise 117 Let $t \in \mathbb{R}$ satisfy the condition that there exist $a, b \in \mathbb{R}$ satisfying $c + d = 1$ and

$$2a^3 - a^2 - 7a + t = 0 = 2b^3 - b^2 - 7b + t.$$

Find t .

Exercise 118 Compare the subsets $F[X^2]$ and $F[X^2 + 1]$ of $F[X]$, where F is a field.

Exercise 119 Let $F = GF(p)$, where p is a prime integer, and let g be an arbitrary function from F to itself. Show that there exists a polynomial $p(X) \in F[X]$ of degree less than p satisfying the condition that $g(c) = p(c)$ for all $c \in F$.

Exercise 120 Let c be a nonzero element of a field F and let $n > 1$ be an integer. Show that there exists a polynomial $p(X) \in F[X]$ satisfying $c^n + c^{-n} = p(c + c^{-1})$.

Exercise 121 Let F be a field. Find the set of all polynomials $0 \neq p(X) \in F[X]$ satisfying $p(X^2) = p(X)^2$.

Exercise 122 Let k be a positive integer and let

$$p_k(X) = \frac{1}{k!} X(X-1) \cdots (X-k+1) \in \mathbb{Q}[X].$$

Show that $p_k(n) \in \mathbb{Z}$ for every nonnegative integer n .

Exercise 123 For any positive integer n , let

$$p_n(X) = nX^{n+1} - (n+1)X^n + 1 \in \mathbb{Q}[X].$$

Show that there exists a polynomial $q_n(X) \in \mathbb{Q}[X]$ satisfying $p_n(X) = (X-1)^2 q_n(X)$.

Exercise 124 Let F be a field and let W be a nontrivial subspace of the vector space $F[X]$ over F . Let $p(X) \in F[X]$ be a given monic polynomial and let $p(X)W = \{p(X)f(X) \mid f(X) \in W\}$. Show that $p(X)W$ is a subspace of $F[X]$ and find a necessary and sufficient condition for it to equal W .

Exercise 125 Let p be a prime integer and let n be a positive integer. Does there necessarily exist an irreducible monic polynomial in $GF(p)[X]$ of degree n ?

Exercise 126 Let p be a prime integer and let n be a positive integer. Show that the product of all irreducible monic polynomials in $GF(p)[X]$ of degree dividing n is equal to $X^{p^n} - X$.

Exercise 127 Let $n > 1$ be an integer. Is the polynomial $p(X) = 1 + \sum_{h=1}^n \frac{1}{h!} X^h \in \mathbb{Q}[X]$ necessarily irreducible?

Exercise 128 Show that the polynomial $X^4 + 1$ is irreducible in $\mathbb{Q}[X]$ but reducible in $GF(p)[X]$ for every prime p .

Exercise 129 Show that the polynomial in $\mathbb{Q}[X]$ of the form

$$X^4 + 2(1 - c)X^2 + (1 + c)^2$$

is irreducible for every $c \in \mathbb{Q}$ satisfying $\sqrt{c} \notin \mathbb{Q}$.

Exercise 130 Let k be a positive integer and let $a < b$ be real numbers. A function $f \in \mathbb{R}^{[a,b]}$ is a **spline function** of degree k if and only if there exist real numbers $a = a_0 < \dots < a_n = b$ and polynomials $p_0(X), \dots, p_{n-1}(X)$ of degree k in $\mathbb{R}[X]$ satisfying the condition that $f : x \mapsto p_i(x)$ for all $a_i \leq x \leq a_{i+1}$ and all $0 \leq i \leq n - 1$. Spline functions play an important part in interpolation theory. Is the set of all spline functions of fixed degree k a subspace of the vector space $\mathbb{R}^{[a,b]}$?

5

Linear independence and dimension

In this chapter we will see how a restricted group of vectors in a vector space over a field can dictate the structure of the entire space, and we will deduce far-ranging conclusions from this. Let V be a vector space over a field F . A nonempty subset D of V is **linearly dependent** if and only if there exist distinct vectors v_1, \dots, v_n in D and scalars a_1, \dots, a_n in F , not all of which are equal to 0, satisfying $\sum_{i=1}^n a_i v_i = 0_V$. A list of elements of V is linearly dependent if it has two equal members or if its underlying subset is linearly dependent. Clearly any set of vectors containing 0_V is linearly dependent. A nonempty set of vectors which is not linearly dependent is **linearly independent**¹. That is to say, D is linearly independent if and only if $D = \emptyset$ or $D \neq \emptyset$ and we have $\sum_{i=1}^n a_i v_i = 0_V$ with the a_i in F and the v_i in V , when and only when $a_i = 0$ for all $1 \leq i \leq n$. As a consequence of this definition, we



¹ The notion of linear independence of vectors was extensively generalized to other mathematical contexts by the 20th-century American mathematician **Hassler Whitney**.

see that an infinite set of vectors is linearly dependent if and only if it has a finite linearly-dependent subset, and an infinite set of vectors is linearly independent if and only if each of its finite subsets is linearly independent. It is also clear that any set of vectors containing a linearly-dependent subset is linearly dependent and that any subset of a linearly-independent set of vectors is linearly independent.

Example: The subset $\left\{ \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix}, \begin{bmatrix} -1 \\ 3 \\ 4 \end{bmatrix}, \begin{bmatrix} -4 \\ 7 \\ 11 \end{bmatrix} \right\}$ of \mathbb{Q}^3 is linearly dependent since $\begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} = (-1) \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix} + 3 \begin{bmatrix} -1 \\ 3 \\ 4 \end{bmatrix} + (-1) \begin{bmatrix} -4 \\ 7 \\ 11 \end{bmatrix}$. Similarly, the subset $\left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} \right\}$ of \mathbb{Q}^3 is linearly independent, since if $\begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} = a \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} + b \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} + c \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$, then $\begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} = \begin{bmatrix} a+b+c \\ b+c \\ c \end{bmatrix}$ and that implies that $a = b = c = 0$.

Example: The subset $\left\{ \begin{bmatrix} 1 \\ 1 \\ 0 \\ 1 \\ 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 1 \\ 1 \\ 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 1 \\ 1 \\ 0 \\ 0 \\ 1 \end{bmatrix} \right\}$ of $GF(2)^7$ is linearly

independent and generates a subspace of V composed of eight vectors:

$$\begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ 0 \\ 1 \\ 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 1 \\ 1 \\ 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 1 \\ 1 \\ 0 \\ 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 0 \\ 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ 0 \\ 1 \\ 0 \\ 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 1 \\ 1 \\ 1 \\ 0 \\ 1 \end{bmatrix}, \text{ and } \begin{bmatrix} 0 \\ 1 \\ 1 \\ 0 \\ 1 \\ 1 \\ 0 \end{bmatrix}.$$

Note that in every element of V other than its identity element for addition, a majority of the entries are nonzero. This property makes this subspace of V important in algebraic coding theory.

Example: Let $b > 1$ be a real number, let $\{p_1, p_2, \dots\}$ be the set of prime integers and, for each i , let $u_i = \log_b(p_i)$. We claim that $D = \{u_1, u_2, \dots\}$ is a linearly-independent subset of \mathbb{R} , when it is considered as a vector space over \mathbb{Q} . Indeed, assume that this is not the case. Then there is a positive integer n and rational numbers a_1, \dots, a_n satisfying $\sum_{i=1}^n a_i u_i = 0$. If we multiply both sides by the product of the denominators of the a_i , we can assume that the a_i are integers. Then

$$1 = b^0 = b^{\sum a_i u_i} = \prod_{i=1}^n b^{a_i u_i} = \prod_{i=1}^n (b^{u_i})^{a_i} = \prod_{i=1}^n p_i^{a_i},$$

and this is a contradiction. Therefore D must be linearly independent.

Example: Let F be a field and let Ω be a nonempty set. Let V_i be a vector space over F for each $i \in \Omega$, and set $V = \prod_{i \in \Omega} V_i$. We have already seen that the identity for addition in this vector space is the function $g_0 : \Omega \rightarrow \bigcup_{i \in \Omega} V_i$ given by $g_0 : i \mapsto 0_{V_i}$. For each $i \in \Omega$, let $f_i : \Omega \rightarrow \bigcup_{i \in \Omega} V_i$ be a function satisfying the condition that $f_i(i) \neq g_0(i)$ but $f_i(h) = g_0(h)$ for all $h \in \Omega \setminus \{i\}$. We claim that the subset $\{f_i \mid i \in \Omega\}$ of V is linearly independent. To see that, assume that there exists a finite subset Λ of Ω and a family of scalars $\{c_h \mid h \in \Lambda\}$ such that $\sum_{h \in \Lambda} c_h f_h = g_0$. Then for each $k \in \Lambda$ we have $g_0(k) = (\sum_{h \in \Lambda} c_h f_h)(k) = \sum_{h \in \Lambda} c_h f_h(k) = c_k f_k(k)$ and since, by definition, $f_k(k) \neq g_0(k)$, we must have $c_k = 0$.

Example: If F is a field, the subset $\{1, X, X^2, \dots\}$ of $F[X]$ is surely linearly independent, since $\sum_{i=0}^n a_i X^i = 0$ if and only if each of the coefficients a_i equals 0.

Example: Let $V = \mathbb{R}^{\mathbb{R}}$ be the vector space, over \mathbb{R} , of all functions from \mathbb{R} to itself. Let D be the set of all functions of the form $x \mapsto e^{ax}$ for some real number a . We claim that D is linearly independent. Indeed, assume that there are distinct real numbers a_1, \dots, a_n and real numbers c_1, \dots, c_n such that the function $x \mapsto \sum_{i=1}^n c_i e^{a_i x}$ equals the 0-function $f_0 : x \mapsto 0$, which is the identity element of V for addition. We need to show that each of the c_i equals 0, and this we will do by induction on n .

If $n = 1$ then we must have $c_1 = 0$ since the function $x \mapsto e^{ax}$ is different from f_0 for each $a \in \mathbb{R}$. Assume therefore that $n > 1$ and that every subset of D having no more than $n - 1$ elements is linearly independent. For each $1 \leq i \leq n$, set $b_i = a_i - a_n$. Then

$$f_0 = e^{-a_n x} \sum_{i=1}^n c_i e^{a_i x} = \left[\sum_{i=1}^{n-1} c_i e^{b_i x} \right] + c_n$$

and if we differentiate both sides of the equation, we see that $f_0 = \sum_{i=1}^{n-1} b_i c_i e^{b_i x}$. By the induction hypothesis and the choice of the scalars a_i as being distinct, it follows that $b_i c_i = 0 \neq b_i$ for each $1 \leq i \leq n-1$ and so $c_i = 0$ for all $1 \leq i \leq n-1$. This in turn implies that $c_n = 0$ as well.

Similarly, let G be the subset of V consisting of all of the functions of the form $g_i : x \mapsto x^{i-1} 2^{x-1}$. We claim that this set too is linearly independent. Indeed, assume otherwise. Then there exists a positive integer n and there exist real numbers c_1, \dots, c_n , such that $\sum_{i=1}^n c_i g_i = f_0$. But this implies that $2^{x-1} (\sum_{i=1}^n c_i x^{i-1}) = 0$ for each real number x . Since $2^{x-1} \neq 0$ for each $x \in \mathbb{R}$, we conclude that $\sum_{i=1}^n c_i x^{i-1} = 0$ for all x . But the polynomial function $x \mapsto \sum_{i=1}^n c_i x^{i-1}$ from \mathbb{R} to itself has infinitely-many roots if and only if $c_i = 0$ for all i , proving linear independence.

Note that if $\{v, w\}$ is a linearly-dependent set of vectors in an anti-commutative algebra (K, \bullet) over a field of characteristic other than 2, then there exist scalars a and b , not both equal to 0, such that $av + bw = 0_K$. Relabeling if necessary, we can assume that $b \neq 0$. Then $0_K = a(v \bullet v) + b(v \bullet w) = b(v \bullet w)$ and so $v \bullet w = 0_K$. A simple induction argument shows that if D is a linearly-dependent subset of K then $v_1 \bullet \dots \bullet v_k = 0_K$ for any finite subset $\{v_1, \dots, v_k\}$ of D .

Note too that Proposition 3.7 can be easily iterated to get the more general result that if D is a nonempty subset of a vector space V over a field F and if B is a finite linearly-independent subset of FD having k elements, then there exists a subset D' of D also having k elements satisfying the condition that $F((D \setminus D') \cup B) = FD$. Moreover, if D is linearly independent, so is $(D \setminus D') \cup B$. This result is sometimes known as the **Steinitz Replacement Property**.

(5.1) Proposition: Let V be a vector space over a field F . A nonempty subset D of V is linearly dependent if and only if some element of D is a linear combination of the others over F .

Proof: Assume D is linearly dependent. Then there exists a finite subset $\{v_1, \dots, v_n\}$ of D and scalars a_1, \dots, a_n , not all of which equal 0, satisfying $\sum_{i=1}^n a_i v_i = 0_V$. Say $a_h \neq 0$. Then $v_h = -a_h^{-1} \sum_{i \neq h} a_i v_i$ and so we see that v_h is a linear combination of the other elements of D over F . Conversely, assume that there is some element of D is a linear combination of the others over F . That is to say, there is an element v_1 of D , elements v_2, \dots, v_n of $D \setminus \{v_1\}$ and scalars a_2, \dots, a_n in F satisfying $v_1 = \sum_{i=2}^n a_i v_i$. If we set $a_1 = -1$, we see that $\sum_{i=1}^n a_i v_i = 0_V$ and so D is linearly dependent. \square

Example: For every real number a , let f_a be the function in $\mathbb{R}^{\mathbb{R}}$ defined by $f_a : x \mapsto |x - a|$. We claim that the subset $D = \{f_a \mid a \in \mathbb{R}\}$ of $\mathbb{R}^{\mathbb{R}}$ is linearly independent. Indeed, assume that this is not the case. Then there exists a real number b such that f_b is a linear combination of other members of D . In other words, there exists a finite subset E of $\mathbb{R} \setminus \{b\}$ and scalars c_a for each $a \in E$ such that $f_b = \sum_{a \in E} c_a f_a$. But the function on the right-hand side of this equation is differentiable at b , while the function on the left-hand side is not. From this contradiction, we see that D is linearly independent.

We now introduce a new concept: if A is a nonempty set, then a relation \preccurlyeq between elements of A is called a **partial order relation** if and only if the following conditions are satisfied:

- (1) $a \preccurlyeq a$ for all $a \in A$;
- (2) If $a \preccurlyeq b$ and $b \preccurlyeq a$ then $a = b$;
- (3) If $a \preccurlyeq b$ and $b \preccurlyeq c$ then $a \preccurlyeq c$.

The term “partial” comes from the fact that, given elements a and b of A , it may happen that neither $a \preccurlyeq b$ nor $b \preccurlyeq a$. A set on which a partial order has been defined is a **partially-ordered set**. A partially-ordered set A satisfying the condition that for all $a, b \in A$ we have either $a \preccurlyeq b$ or $b \preccurlyeq a$ is called a **chain**. A nonempty subset B of a partially-ordered set A is itself partially-ordered relative to the partial order relation defined on A ; it is a **chain subset** if it is a chain relative to the partial order defined on A .

If A is a nonempty set on which we have a partial order relation \preccurlyeq defined, then an element a_0 of A is **maximal** in A if and only if $a_0 \preccurlyeq a$ when and only when $a = a_0$. An element a_1 is **minimal** if and only if $a \preccurlyeq a_1$ when and only when $a = a_1$. Maximal and minimal elements need not exist or, if they exist, need not be unique. The **Well Ordering Principle**, one of the fundamental axioms of number theory, says that any nonempty subset of \mathbb{N} , ordered with the usual partial order, has a minimal element.

Partial order relations are ubiquitous in mathematics, and often play a very important, though not usually highlighted, part in the analysis of mathematical structures.

Example: Let A be a nonempty set and let P be the collection of all subsets of A . Define a relation \preccurlyeq between elements of P by setting $B \preccurlyeq B'$ if and only if $B \subseteq B'$. It is easy to verify that this is indeed a partial order relation. Moreover, P has a unique maximal element, namely A , and a unique minimal element, namely \emptyset . The set P is not a chain whenever A has more than one element since, if a and b are distinct elements of A , then $\{a\} \not\subseteq \{b\}$ and $\{b\} \not\subseteq \{a\}$.

Example: Let $A = \{1, 2, 3\}$ and let P be the collection of all subsets of A having one or two elements. Thus P has six elements: $\{1\}, \{2\},$

$\{3\}$, $\{1, 2\}$, $\{1, 3\}$, and $\{2, 3\}$. Again, the relation \preceq between elements of P defined by setting $B \preceq B'$ if and only if $B \subseteq B'$ is a partial order relation. Moreover, P has three minimal elements: $\{1\}$, $\{2\}$, and $\{3\}$; it also has three maximal elements: $\{1, 2\}$, $\{1, 3\}$, and $\{2, 3\}$.

In general, if we have a collection of subsets of a given set, the collection is partially-ordered by setting $B \preceq B'$ if and only if $B \subseteq B'$. Therefore it makes sense for us to talk about “a minimal generating set” of a vector space V – namely a minimal element in the partially-ordered collection of all generating sets of V – and about “a maximal linearly-independent subset” of a vector space V – namely a maximal element of the partially-ordered collection of all linearly-independent subsets of V . However, we have no *a priori* guarantee that such minimal or maximal elements in fact exist.

Example: Consider the set A of all integers greater than 1, and define a relation \preceq on A by setting $k \preceq n$ if and only if there is a positive integer t satisfying $n = tk$. This is a partial order relation on A . Moreover, A has infinitely-many minimal elements, since each prime integer is a minimal element of A , while it has no maximal elements, since $n \preceq 2n$ for each $n \in A$.

(5.2) Proposition: Let V be a vector space over a field F . Then the following conditions on a subset D of V are equivalent:

- (1) D is a minimal set of generators of V ;
- (2) D is a maximal linearly-independent subset of V ;
- (3) D is a linearly-independent set of generators of V .

Proof: (1) \Rightarrow (2): Let D be a minimal set of generators of V , and assume that D is linearly dependent. By Proposition 5.1 there exists an element $v_0 \in D$ which is a linear combination of elements of the set $E = D \setminus \{v_0\}$ over F . Say $v_0 = \sum_{i=1}^n a_i u_i$, where the u_i belong to E and the a_i are scalars in F . If v is arbitrary element of V then, since D is a set of generators of V , there exists elements v_1, \dots, v_n of E and scalars b_0, b_1, \dots, b_n such that $v = \sum_{j=0}^n b_j v_j$. But this then implies that $v = b_0 v_0 + \sum_{j=1}^n b_j v_j = \sum_{i=1}^n b_0 a_i u_i + \sum_{j=1}^n b_j v_j$ and so E is also a set of generators of V , contradicting the minimality of D . This establishes the claim that D is linearly independent. If $v \in V \setminus D$, the set $D \cup \{v\}$ is linearly dependent since v is a linear combination of elements of D . Thus D is a maximal linearly-independent set.

(2) \Rightarrow (3): Assume that D is a maximal linearly-independent subset of V . Consider a vector v_0 in $V \setminus D$. By (2), we know that the set $D \cup \{v_0\}$ is linearly dependent and so $0_V \in F(D \cup \{v_0\}) \setminus FD$ by Proposition 3.7, this implies $v_0 \in F(D \cup \{0_V\}) = FD$, which proves that D is a set of generators of V .

(3) \Rightarrow (1): Assume that D is a linearly-independent set of generators of V and that E is a proper subset of D which is also a set of generators for V . Let $v_0 \in D \setminus E$. Then there exist elements v_1, \dots, v_n of E and scalars a_1, \dots, a_n such that $v_0 = \sum_{i=1}^n a_i v_i$. But, by Proposition 5.1, that implies that the set D is linearly dependent, contradicting (3). Therefore no such E exists and so D is a minimal set of generators of V . \square

(5.3) Proposition: Let V be a vector space over a field F and let D be a linearly-independent subset of V . If $v_0 \in V \setminus FD$ then the set $D \cup \{v_0\}$ is linearly independent.

Proof: Assume that this set is linearly dependent. Then there exist elements v_1, \dots, v_n of D and scalars a_0, a_1, \dots, a_n , not all equal to 0, such that $\sum_{i=0}^n a_i v_i = 0_V$. The scalar a_0 must be different from 0, for otherwise D would be linearly dependent, which is a contradiction. Therefore $v_0 = \sum_{i=1}^n -a_0^{-1} a_i v_i \in FD$, which contradicts the choice of v_0 . Thus $D \cup \{v_0\}$ must be linearly independent. \square

Proposition 5.3 has important implications. For example, let V be a vector space over a field F which is not finitely generated and let $D = \{v_1, \dots, v_n\}$ be a linearly-independent subset of V . Then $FD \neq V$, since V is not finitely generated, and so there exists a vector $v_{n+1} \in V \setminus FD$ such that $\{v_1, \dots, v_{n+1}\}$ is linearly independent. Thus we see that a vector space which is not finitely generated has linearly-independent finite subspaces of arbitrarily-large size.

A generating set for a vector space V over a field F which is also linearly independent, is called a **basis** of V over F . In Proposition 5.2 we gave some equivalent conditions for determining of a subset of a vector space is a basis. However, we have not yet proven that every (or, indeed, any) vector space must have a basis.

Example: Clearly $\left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \right\}$ and $\left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} \right\}$

are bases of F^3 for any field F . If the characteristic of F is other than 2, then $\left\{ \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix} \right\}$ is a basis of F^3 , but if the

field F has characteristic 2, then the set is linearly dependent, since

$$\begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} + \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} + \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}.$$

Example: Let F be a field and let both k and n be positive integers. For each $1 \leq s \leq k$ and each $1 \leq t \leq n$, let H_{st} be the matrix $[a_{ij}]$ in $\mathcal{M}_{k \times n}(F)$ defined by

$$a_{ij} = \begin{cases} 1 & \text{if } (i, j) = (s, t) \\ 0 & \text{otherwise} \end{cases}.$$

Then $\{H_{st} \mid 1 \leq s \leq k \text{ and } 1 \leq t \leq n\}$ is a basis of $\mathcal{M}_{k \times n}(F)$.

Example: If F is a field, then we have already seen that the subset $\{1, X, X^2, \dots\}$ of $F[X]$ is a linearly-independent generating set for $F[X]$ as a vector space over F , and so is a basis of this space. The same is true for the subset $\{1, X+1, X^2+X+1, \dots\}$ of $F[X]$. More generally, if $\{p_0(X), p_1(X), \dots\}$ is a subset of $F[X]$ satisfying the condition that $\deg(p_i(X)) = i$ for all $i \geq 0$, then it is a basis of $F[X]$ as a vector space over F .

Since every element of a vector space V over a field F has a unique representation as a linear combination of elements of a basis, if one wants to define a structure of an F -algebra on V it suffices to define the product of any pair of basis elements, and then extend the definition by distributivity and associativity. This is illustrated by the following example, and we will come back to it again in Proposition 5.5.

Example: We have already noted that if F is a field then $F[X]$ is an associative F -algebra. Let us generalize this construction. Let H be a nonempty set on which we have defined an associative operation $*$. Thus, for example, H could be the set of nonnegative integers with the operation of addition or multiplication. Let V be the vector space over F with basis $\{v_h \mid h \in H\}$ and define an operation \bullet on V as follows: if $v = \sum_{g \in H} a_g v_g$ and $w = \sum_{h \in H} b_h v_h$ are elements of V (where at most finitely-many of the a_g and the b_h are nonzero), then set $v \bullet w = \sum_{g \in H} \sum_{h \in H} a_g b_h v_{g*h}$. This turns V into an associative F -algebra. In the case $H = \{X^i \mid i \geq 0\}$, we get $F[X]$. Such constructions are very important in advanced applications of linear algebra.

Note that a vector space may have (and usually does have) many bases and so the problem arises as to whether there is a preferred basis among all of these. For vector spaces of the form F^n , there are reasons to prefer

the basis $\left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}, \dots, \begin{bmatrix} 0 \\ 0 \\ 0 \\ \vdots \\ 1 \end{bmatrix} \right\}$; for vector spaces of the form

$\mathcal{M}_{k \times n}(F)$ there are reasons to prefer the basis $\{H_{st} \mid 1 \leq s \leq k \text{ and } 1 \leq t \leq n\}$ defined above; and for vector spaces of the form $F[X]$ there are reasons to prefer the basis $\{1, X, X^2, \dots\}$. These bases are called the **canonical bases** of their respective spaces. However, in various applications – especially those involving large calculations – it is often convenient and sometimes extremely important to pick other bases which fit the problem under consideration.

It is also important to emphasize another point. When we defined the notation for this book, we stressed that when a set is defined by listing its elements, the set comes with an implicit order defined by that listing. When we deal with bases, and especially finite bases, the order in which the elements of the basis are written often plays a critical role, and one should never lose track of this.

(5.4) Proposition: Let V be a vector space over a field F and let D be a nonempty subset of V . Then D is a basis of V if and only if every vector in V can be written as a linear combination of elements of D over F in precisely one way.

Proof: First, let us assume that D is a basis of V and that there exists an element v of V which can be written as a linear combination of elements of D over F in two different ways. That is to say, that there exists a finite subset $\{v_1, \dots, v_n\}$ of D and there exist scalars $a_1, \dots, a_n, b_1, \dots, b_n$ in F such that $v = \sum_{i=1}^n a_i v_i = \sum_{i=1}^n b_i v_i$, where $a_h \neq b_h$ for at least one index h . Then $0_V = v - v = (\sum_{i=1}^n a_i v_i) - (\sum_{i=1}^n b_i v_i) = \sum_{i=1}^n [a_i - b_i] v_i$, where at least one of the scalars $a_i - b_i$ is nonzero. This contradicts the assumption that D is a basis and hence linearly independent. Therefore every vector in V can be written as a linear combination of elements of D over F in precisely one way.

Conversely, assume that every vector in V can be written as a linear combination of elements of D over F in precisely one way. That certainly implies that D is a generating set for V over F . If $\{v_1, \dots, v_n\}$ is a subset of D and if a_1, \dots, a_n are scalars satisfying $\sum_{i=1}^n a_i v_i = 0_V$, then we have $\sum_{i=1}^n a_i v_i = \sum_{i=1}^n 0 v_i$ and so, by uniqueness of representation, we have $a_i = 0$ for each $1 \leq i \leq n$. This shows that D is linearly independent and so a basis. \square

We can look at Proposition 5.4 from another point of view. Let D be a nonempty subset of a vector space V over a field F , and define a function $\theta : F^{(D)} \rightarrow V$ by setting $\theta : f \mapsto \sum_{u \in D} f(u)u$. (This sum is well-defined since only finitely-many of the summands are nonzero.) Then:

- (1) The function θ is monic if and only if D is linearly independent;
- (2) The function θ is epic if and only if D is a generating set;
- (3) The function θ is bijective if and only if D is a basis.

(5.5) Proposition: Let D be a basis for a vector space V over a field F . Then any function $f : D \times D \rightarrow V$ can be extended in a unique manner to a function $V \times V \rightarrow V$ which defines on V the structure of an F -algebra. Moreover, all F -algebra structures on V arise in this manner.

Proof: Let $D = \{y_i \mid i \in \Omega\}$. Suppose that we are given a function $f : D \times D \rightarrow V$. We define an operation \bullet on V as follows: if $v, w \in V$, then, by Proposition 5.4, we know that we can write $v = \sum_{i \in \Omega} a_i y_i$ and $w = \sum_{j \in \Omega} b_j y_j$ in a unique manner, where the a_i and b_j are scalars, only a finite number of which are nonzero; then set $v \bullet w = \sum_{i \in \Omega} \sum_{j \in \Omega} a_i b_j f(y_i, y_j)$. It is straightforward to show that this defines the structure of an F -algebra on V . Conversely, if (V, \bullet) is an F -algebra, define the function $f : D \times D \rightarrow V$ by $f : (y_i, y_j) \mapsto y_i \bullet y_j$. \square

The function f in Proposition 5.5 is the **multiplication table** of the vector multiplication operation \bullet .

Example: Let F be a field and let $a, b \in F$. Let $B = \{v_1, v_2, v_3, v_4\}$ be the canonical basis for F^4 over F . Define an operation \bullet on B according to the multiplication table:

\bullet	v_1	v_2	v_3	v_4
v_1	v_1	v_1	v_3	v_4
v_2	v_2	av_1	v_4	av_3
v_3	v_3	$-v_4$	bv_1	$-bv_2$
v_4	v_4	$-av_3$	bv_2	$-abv_1$

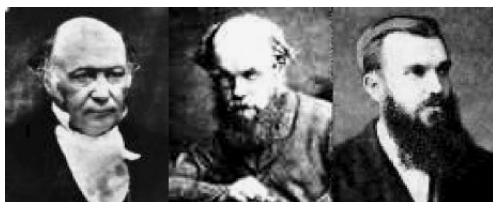
and extend this operation to F^4 by setting $\left(\sum_{i=1}^4 a_i v_i\right) \bullet \left(\sum_{j=1}^4 b_j v_j\right) = \sum_{i=1}^4 \sum_{j=1}^4 a_i b_j (v_i \bullet v_j)$. Then F^4 , together with this operation, is a unital associative algebra known as a **quaternion algebra** over F , in which v_1 is the identity element of for multiplication. In the special case of $F = \mathbb{R}$ and $a = b = -1$, we get the algebra of **real quaternions**, which is denoted by \mathbb{H} . The algebra of real quaternions was first defined by Hamilton in 1844 as a generalization of the field of complex numbers (and earlier studied by Gauss, who did not publish his results). It is a division

algebra over \mathbb{R} since every nonzero quaternion is a unit of \mathbb{H} . These were subsequently generalized by Clifford and used in his study of noneuclidean spaces.² Lately, they have also been used in computer graphics and in signal analysis. If F is a field having characteristic $p > 0$, quaternion algebras over F are not even entire. However, they arise naturally in the theory of elliptic curves, and so are of great importance in cryptography. If $p > 2$, then no quaternion algebras over F are commutative.

We now show that any vector space over a field F has a basis. Indeed, the following two propositions show somewhat stronger than that³.

(5.6) Proposition: If V is a vector space finitely generated over a field F then every finite generating set of V over F contains a basis of V .

Proof: Let V be a vector space finitely generated over a field F and let D be a finite generating set for V over F . If D is minimal among all generating sets for V , then we know by Proposition 5.2 that it is a basis of V . If not, it properly contains other generating sets for V over F , one of which, say E , has the fewest elements. Then E cannot properly contain any other generating set for V over F , and so it must be a basis of V . \square



2

Sir William Rowan Hamilton, a 19th-century Irish mathematician and physicist, helped create matrix theory in its modern formulation, together with Cayley and Sylvester. Hamilton was the first to use the terms “vector” and “scalar” in an algebraic context. His championship of quaternions as an alternative to vectors in physics was later taken up by Scottish mathematician **Peter Guthrie Tait**. Nineteenth-century British mathematician **William Kingdom Clifford** was one of the first to argue that energy and matter were just different types of curvature of space.



3

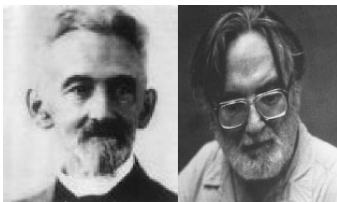
The Italian mathematician **Giuseppe Peano**, best known for his axiomatization of the natural numbers, was the first to prove that every finitely-generated vector space has a basis at the end of the 19th century. He also gave the final form for the definition of a vector space, which we used above.

(5.7) Proposition: If V is a vector space finitely generated over a field F then every linearly-independent subset B of V is contained in a basis of V over F .

Proof: By assumption, there exists a finite generating set $\{v_1, \dots, v_n\}$ for V over F . Let B be a linearly-independent subset of V . If $v_i \in FB$ for each $1 \leq i \leq n$, then $FB = V$, and B is itself a basis of V . Otherwise, let $h = \min\{i \mid v_i \notin FB\}$. By Proposition 5.3, the set $D = B \cup \{v_h\}$ is linearly independent. If it is a generating set for V , then it is a basis and we are done. If not, let $k = \min\{i \mid v_i \notin FD\}$, and replace D by $B \cup \{v_h, v_k\}$. Continuing in this manner, we see that after finitely-many steps we obtain a basis of V . \square

We now want to extend this result to vector spaces which are not finitely generated, and to do so we have to make use of an axiom of set theory known variously as the **Hausdorff Maximum Principle** or **Zorn's Lemma**⁴. To state this principle, we need another concept about partially-ordered sets. Let A be a set on which we have defined a partial order \preceq . A subset B of A is **bounded** if and only if there exists an element $a_0 \in A$ satisfying $b \preceq a_0$ for all $b \in B$. Note that we do not require that a_0 belong to B . The Hausdorff maximum principle then says that if A is a partially-ordered set in which every chain subset is bounded, then A has a maximal element. Again, this is not really a “principle” or a “lemma”; it is an axiom of set theory which has been shown to be independent of the other (Zermelo-Fraenkel) axioms one usually assumes. Indeed, it is logically equivalent to the Axiom of Choice, which we mentioned in Chapter 1 as being somewhat controversial among those mathematicians dealing with the foundations of mathematics. However, in this book, we will assume that it holds. Given that assumption, we can now extend Proposition 5.7.

(5.8) Proposition: If V is a vector space over a field F then every linearly-independent subset B of V is contained in a basis of V .



4

Felix Hausdorff, one of the leading mathematicians of the early 20th century and one of the founders of topology, died in a German concentration camp in 1942. **Max Zorn**, a German mathematician who emigrated to the United States, made skillful use of the Hausdorff Maximum Principle in his research, turning it into an important mathematical tool.

Proof: Let B be a linearly-independent subset of V and let P be the collection of all linearly-independent subsets of V which contain B , which is partially-ordered by inclusion, as usual. Then P is nonempty since $B \in P$. Let Q be a chain subset of P . We want to prove that Q is bounded in P . That is to say, we want to find a linearly independent subset E of V which contains every element of Q . Indeed, let us take E to be the union of all of the elements of Q . To show that E is linearly independent, it suffices to show that every finite subset of E is linearly independent. Indeed, let $\{v_1, \dots, v_n\}$ be a finite subset of E . Then for each $1 \leq i \leq n$, there exists an element D_i of Q containing v_i . Since Q is a chain, there exists an index h such that $D_i \subseteq D_h$ for all $1 \leq i \leq n$ and so $v_i \in D_h$ for all $1 \leq i \leq n$. Therefore this set is a subset of a linearly-independent set and so is linearly independent. Thus we have shown that every chain subset of P is bounded and so, by the Hausdorff maximum principle, the set P has a maximal element. In other words, there exists a maximal linearly-independent subset of V containing B , and this, as we know, is a basis of V over F . \square

Taking the special case of $B = \emptyset$ in Proposition 5.8, we see that every vector space has a basis. In the above proof we used the Axiom of Choice to prove this statement. In fact, one can show something considerably stronger: in the presence of the other generally-accepted axioms of set theory, the Axiom of Choice is equivalent, in the sense of formal logic, to the statement that every vector space over any field has a basis⁵.

Example: Consider the field \mathbb{R} as a vector space over its subfield \mathbb{Q} . A basis for this space is known as a **Hamel basis**. By Proposition 5.8, we know that Hamel bases exist, but nobody has been able to come up with a method of specifically constructing one. The subset C of \mathbb{R} consisting of all real numbers which can be represented in the form $\sum_{i>0} u_i 3^{-i}$, where each u_i is either 0 or 2, is called the **Cantor set**, and it can be shown to be “sparse” (in a technical sense of the word we won’t go into here) in the unit interval $[0, 1]$ in \mathbb{R} . It is possible to show that there is a Hamel basis of \mathbb{R} contained in C . Still, the mere existence of Hamel bases leads to some very interesting results, as the following shows.



5

This result is due to the contemporary American mathematician, **Andreas Blass**.

Let H be a Hamel basis of \mathbb{R} . If $r \in \mathbb{R}$ then we can write $r = \sum_{a \in H} q_a(r)a$, where $q_a(r) \in \mathbb{Q}$ and there are only finitely-many elements $a \in H$ for which $q_a(r) \neq 0$. Since such a representation is unique, we see that

$$q_a(b) = \begin{cases} 1 & \text{if } a = b \\ 0 & \text{otherwise} \end{cases}$$

for $a, b \in H$. Moreover, if $r, s \in \mathbb{R}$ and $a \in H$ then $q_a(r+s) = q_a(r) + q_a(s)$ so, if $a \neq b$ are elements of H then for any $r \in \mathbb{R}$ we have $q_a(r+b) = q_a(r) + q_a(b) = q_a(r)$. Thus we see that the function $q_a \in \mathbb{R}^{\mathbb{R}}$ is **periodic**, with period b for any $b \in H \setminus \{a\}$, and its image is contained in \mathbb{Q} . Moreover, if we pick two distinct elements c and d of H , we see that for each $r \in \mathbb{R}$, we have $r = f(r) + g(r)$, where $f, g \in \mathbb{R}^{\mathbb{R}}$ are defined by $f : r \mapsto q_c(r)c$ and $g : r \mapsto \sum_{a \in H \setminus \{c\}} q_a(r)a$. By our previous comments, f is periodic with period d and g is periodic with period c . We conclude that the identity function in $\mathbb{R}^{\mathbb{R}}$ is the sum of two periodic functions. A somewhat more sophisticated argument along the same lines shows that any polynomial function in $\mathbb{R}^{\mathbb{R}}$ of degree n is the sum of $n+1$ periodic functions. Of course, since we cannot specify H , there is no way of finding these periodic functions explicitly.⁶

We have seen that a vector space over a field can have many bases. We want to show next that if the vector space is finitely generated, then all of these bases are finite and have the same number of elements. First, however, we must prove a preliminary result.

(5.9) Proposition: Let V be a vector space over a field F which is generated by a finite set $B = \{v_1, \dots, v_n\}$ and let D be a linearly independent set of vectors in V . Then the number of elements in D is at most n .

Proof: Suppose that D has a subset $E = \{w_1, \dots, w_{n+1}\}$ having more than n elements. Since this set must also be linearly independent, we know that none of the w_i equals 0_V . For each $1 \leq k \leq n$, set $D_k = \{w_1, \dots, w_k, v_{k+1}, \dots, v_n\}$.



6

Twentieth-century German mathematician **Georg Hamel** was a student of Hilbert who worked primarily in function theory. In his later years, he became notorious for his pro-Nazi views and activities.

Since B is a generating set for V , we can find scalars a_1, \dots, a_n , not all equal to 0, such that $w_1 = \sum_{i=1}^n a_i v_i$. In order to simplify our notation, we will renumber the elements of B if necessary so that $a_1 \neq 0$. Then $v_1 = a_1^{-1} w_1 - \sum_{i=2}^n a_1^{-1} a_i v_i$ and so $D \subseteq FD_1$. But $D_1 \subseteq V = FD$ and so $V = FD_1$ by Proposition 3.6. Now assume that $1 \leq k < n$ and that we have already shown that $V = FD_k$. Then there exist scalars b_1, \dots, b_n , not all equal to 0, such that $w_{k+1} = \sum_{i=1}^k b_i w_i + \sum_{i=k+1}^n b_i v_i$. If the scalars b_{k+1}, \dots, b_n are all equal to 0, then we have shown that D is linearly dependent, which is not the case. Therefore at least one of them is nonzero and, by renumbering if necessary, we can assume that $b_{k+1} \neq 0$. Thus $v_{k+1} = b_{k+1}^{-1} w_{k+1} - \sum_{i=1}^k b_{k+1}^{-1} b_i w_i - \sum_{i=k+2}^n b_{k+1}^{-1} b_i v_i$ and so, using the above reasoning, we get $V = FD_{k+1}$. Continuing in this manner, we see that after n steps we obtain $V = FD_n = F\{w_1, \dots, w_n\}$. But then $w_{n+1} \in F\{w_1, \dots, w_n\}$ and so E is linearly dependent, contrary to our assumption. This proves that D can have at most n elements. \square

(5.10) Proposition: Let V be a vector space finitely generated over a field F . Then any two bases of V have the same number of elements.

Proof: By hypothesis, there exists a finite generating set for V over F having, say, n elements. If B is a basis of V then, by Proposition 5.9, we know that B has at most n elements and so, in particular, is finite. Suppose B and B' are two bases for V having h and k elements respectively. Since B is linearly independent and B' is a generating set, we know that $h \leq k$. But, on the other hand, B' is linearly independent and B is a generating set, so $k \leq h$. Thus $h = k$. \square

We should remark at this point that the assertion for linearly-dependent sets corresponding to Proposition 5.10 is not true. That is to say, a finite linearly-dependent set of vectors may have two minimal linearly-dependent subsets with different numbers of elements. Indeed, there is no efficient algorithm to find such subsets of a given linearly-dependent set. We should also note that Proposition 5.9 is a special case of a more general theorem: if V is a vector space (not necessarily finitely generated) over a field F then there exists a bijective function between any two bases of V . The proof of this result makes use of techniques from advanced set theory, such as transfinite induction.

If V is a vector space finitely generated over a field F then V is **finite dimensional** and the number of elements in a basis of V is called the **dimension** of V over F . If V is not finite dimensional, it is **infinite dimensional**. (In choosing this latter terminology, we are deliberately skipping over the subject of various transfinite dimensions, since the reader is not assumed to be familiar with the arithmetic of transfinite

cardinals. In certain mathematical contexts, distinction between infinite dimensions – for example the distinction between spaces of countably-infinite and uncountably-infinite dimension – can be very significant. We will not, however, need it in this book.) We denote the dimension of V over F by $\dim(V)$, or by $\dim_F(V)$ when it is important to emphasize the field of scalars⁷.

Notice that the proof of Proposition 5.9, which is in turn critical in proving Proposition 5.10, uses the fact that every nonzero element of F has a multiplicative inverse, and this cannot be avoided. If we try to weaken the notion of a vector space by allowing scalars to be, say, only integers, it may happen that such a space would have two bases of different sizes and so we could no longer define the notion of dimension in an obvious manner. We did not use, in an unavoidable manner, the commutativity of scalar multiplication and so we could weaken our notion of a vector space to allow scalars which do not commute among themselves, such as scalars coming from \mathbb{H} . However, the generality thus gained does not seem to outweigh the bother it causes, and so we will refrain from doing so. Thus, for us, the fact that scalars always come from a field is critical in the development of our theory.

Example: If F is a field then $\dim(F^n) = n$ for every positive integer n , since the canonical basis of F^n has n elements. Similarly, if k and n are positive integers then $\dim_F(\mathcal{M}_{k \times n}(F)) = kn$, since the canonical basis of $\mathcal{M}_{k \times n}(F)$ has kn elements. The dimension of the space $F[X]$ is infinite since the canonical basis of $F[X]$ has infinitely-many elements.

Example: If F is a field and n is a positive integer, then the set W of all polynomials in $F[X]$ having degree at most n is a subspace of $F[X]$ having dimension $n + 1$, since $\{1, X, \dots, X^n\}$ is a basis of W having $n + 1$ elements.

Example: The dimension of \mathbb{R} over itself is 1. Since $\{1, i\}$ is clearly a basis of \mathbb{C} as a vector space over \mathbb{R} , we see that $\dim_{\mathbb{R}}(\mathbb{C}) = 2$ and so there cannot be a proper subfield F of \mathbb{C} properly containing \mathbb{R} .



⁷ The notion of dimension was implicit in the work of Peano, but was redefined and studied in a comprehensive manner by the 20th-century German mathematician **Hermann Weyl**.

Indeed, if there were such a field, its dimension over \mathbb{R} would have to be greater than 1 and less than 2 (else it would be equal to \mathbb{C}), which is impossible. Clearly, $\dim_{\mathbb{R}}(\mathbb{H}) = 4$. It turns out that the only possible dimensions of division algebras over \mathbb{R} are 1, 2, 4, and 8. The dimension 8 case is realized by a (non-associative) Cayley algebra over \mathbb{R} , as defined in Chapter 15. There are no associative division algebras of dimension 8 over \mathbb{R} .⁸

Example: Let F be a field, let (K, \bullet) be an associative unital F -algebra, and let $v \in K$. If $p(X) = \sum_{i=0}^{\infty} a_i X^i \in F[X]$, then $p(v) = \sum_{i=0}^{\infty} a_i v^i$ is an element of K and the set of all elements of K of this form is an F -subalgebra of K , which is in fact commutative, even though K itself may not be. We will denote this algebra by $F[v]$. If the dimension of $F[v]$, considered as a vector space over F , is finite, we know that there must exist a polynomial $p(X) \in F[X]$ of positive degree satisfying $p(v) = 0_K$. In that case, we say that v is **algebraic** over F . Otherwise, if the dimension of $F[v]$ is infinite, we say v is **transcendental** over F .

Thus, for example, the real numbers π and e (the base of the natural logarithms) are transcendental over \mathbb{Q} . If F is a subfield of a field K then the set L of all elements of K which are algebraic over F is a subfield of K . Moreover, if K is algebraically closed, so is L , and in fact L is the smallest algebraically-closed subfield of K containing F . In particular, we can consider the field of all complex numbers algebraic over \mathbb{Q} . This is a proper subfield of \mathbb{C} , known as the **field of algebraic numbers**.⁹



8

Twentieth-century German mathe-

matician **Heinz Hopf** used algebraic topology to prove that the only possible dimensions were powers of 2, and the final result was obtained by twentieth-century American mathematician **Raoul Bott** and contemporary American mathematician **John Milnor**, again using nonalgebraic tools.



9

The transcendence of π was proven by German mathematician **Ferdinand von Lindemann** in 1882. The transcendence of e was proven by French mathematician **Charles Hermite** in 1873. As we shall see later, Hermite made many important contributions to linear algebra.

From the definition of dimension we see that if V is a vector space of finite dimension n over a field F then:

- (1) Every subset of V having more than n elements must be linearly dependent;
- (2) There exists a linearly-independent subset B of V having precisely n elements;
- (3) If B is as in (2) then B is also a generating set of V over F .

(5.11) Proposition: Let V be a vector space finitely generated over a field F and let W be a subspace of V . Then:

- (1) W is finitely generated over F ;
- (2) Every basis of W can be extended to a basis of V ;
- (3) $\dim(W) \leq \dim(V)$, with equality when and only when $W = V$.

Proof: Let $n = \dim(V)$.

(1) If W is not finitely generated, then, as we remarked after Proposition 5.3, W has a linearly-independent subset B having $n + 1$ elements. But B is also a subset of V , contradicting the assumption that $\dim(V) = n$.

(2) Let B be a basis of W . Then B is a linearly-independent set of elements of V and so, by Proposition 5.7, can be extended to a basis of V .

(3) By (2), we see that the number of elements of a basis of W can be no greater than the number of elements of a basis of V , and so $\dim(W) \leq \dim(V)$. Moreover, if we have equality then any basis B of W is also a basis of V , and so $W = FB = V$. \square

We now want to extend the notion of linear independence. Let U and W be subspaces of a vector space V over a field F . Any vector $v \in U + W$ can be written in the form $u + w$, where $u \in U$ and $w \in W$, but there is no reason for this representation to be unique. It will be unique, however, if U and W are disjoint. Indeed, if this condition holds and if $u, u' \in U$ and $w, w' \in W$ satisfy $u + w = u' + w'$, then $u - u' = w' - w \in U \cap W$ and so $u - u' = 0_V = w - w'$, which in turn implies that $u = u'$ and $w = w'$.

To emphasize the importance of this situation, we will introduce new notation: if U and W are disjoint subspaces of a vector space V over a field F , we will write $U \oplus W$ instead of $U + W$. The subspace $U \oplus W$ is called the **direct sum** of U and W . We note that, by this definition, $U \oplus \{0_V\} = U$ for every subspace U of V .

Example: It is easy to see that $\mathbb{R}^2 = \mathbb{R} \begin{bmatrix} 1 \\ 0 \end{bmatrix} \oplus \mathbb{R} \begin{bmatrix} 0 \\ 1 \end{bmatrix}$.

Of course, we would like to extend the notion of direct sum to cover more than two subspaces. In general, if V is a vector space over a field F ,

then a collection $\{W_h \mid h \in \Omega\}$ of subspaces of V is **independent** if and only if it satisfies the following condition: if Λ is a finite subset of Ω and if we choose elements $w_h \in W_h$ for all $h \in \Lambda$, then $\sum_{h \in \Lambda} w_h = 0_V$ when and only when $w_h = 0_V$ for each $h \in \Lambda$. Thus we see that an infinite collection of subspaces is independent if and only if every finite nonempty subcollection is independent. Clearly, a subset D of a vector space V over a field F is linearly independent if and only if the collection of subspaces $\{Fv \mid v \in D\}$ is independent.

(5.12) Proposition: Let V be a vector space over a field F and let W_1, \dots, W_n be distinct subspaces of V . Then the following conditions are equivalent:

- (1) $\{W_1, \dots, W_n\}$ is independent;
- (2) Every vector $w \in \sum_{i=1}^n W_i$ can be written as $w_1 + \dots + w_n$, with $w_i \in W_i$ for each $1 \leq i \leq n$, in exactly one way;
- (3) W_h and $\sum_{i \neq h} W_i$ are disjoint, for each $1 \leq h \leq n$.

Proof: (1) \Rightarrow (2): Let $w \in \sum_{i=1}^n W_i$ and assume that we can write $w = w_1 + \dots + w_n = y_1 + \dots + y_n$, where $w_i, y_i \in W_i$ for each $1 \leq i \leq n$. Then $\sum_{i=1}^n (w_i - y_i) = 0_V$ and so, by (1), it follows that $w_i - y_i = 0_V$ for each $1 \leq i \leq n$, proving (2).

(2) \Rightarrow (3): Assume that $0_V \neq w_h \in W_h \cap \sum_{i \neq h} W_i$. Then for each $i \neq h$ there exists an element $w_i \in W_i$ satisfying $w_h = \sum_{i \neq h} w_i$, contradicting (2).

(3) \Rightarrow (1): Suppose we can write $w_1 + \dots + w_n = 0_V$, where $w_i \in W_i$ for each $1 \leq i \leq n$, and where $w_h \neq 0_V$ for some h . Then $w_h = -\sum_{i \neq h} w_i \in W_h \cap \sum_{i \neq h} W_i$, and this contradicts (3). Thus (1) must hold. \square

If V is a vector space over a field F and if $\{W_i \mid i \in \Omega\}$ is an independent collection of subspaces of V , we write $\bigoplus_{i \in \Omega} W_i$ instead of $\sum_{i \in \Omega} W_i$. If $\Omega = \{1, \dots, n\}$, we will also write this sum as $W_1 \oplus \dots \oplus W_n$. If $V = \bigoplus_{i \in \Omega} W_i$, then we say that V has a **direct-sum decomposition** relative to the subspaces W_i .

Example: If B is a basis of a vector space V over a field F then $V = \bigoplus_{v \in B} Fv$.

The importance of direct-sum decompositions is illustrated by the following result.

(5.13) Proposition: Let V be a vector space over a field F , let $\{W_i \mid i \in \Omega\}$ be a pairwise disjoint collection of subspaces of V and, for each $i \in \Omega$, let B_i be a basis of W_i . Then $V = \bigoplus_{i \in \Omega} W_i$ if and only if $B = \bigcup_{i \in \Omega} B_i$ is a basis of V .

Proof: Assume $V = \bigoplus_{i \in \Omega} W_i$ and let $v \in V$. Then there exists a finite subset Λ of Ω such that $v \in \bigoplus_{i \in \Lambda} W_i$, and so for each $i \in \Lambda$ there is an element $w_i \in W_i$ satisfying $v = \sum_{i \in \Lambda} w_i$. Moreover, each w_i is a linear combination of elements of B_i . Thus v is a linear combination of elements of B , and so B is a generating set for V . We are left to show that B is linearly independent. If that is not the case, then there exist an element h of Ω , vectors y_1, \dots, y_t in B_h , and scalars a_1, \dots, a_t in F , not all of which equal to 0, such that $\sum_{j=1}^t a_j y_j + u = 0_V$, where u is a linear combination of elements of $\bigcup_{i \neq h} B_i$. But then $\sum_{j=1}^t a_j y_j \in W_h \cap \bigcap_{i \neq h} W_i$, contradicting our initial assumption. Thus $B = \bigcup_{i \in \Omega} B_i$.

Conversely, if $B = \bigcup_{i \in \Omega} B_i$, it then follows that every element of V can be written in a unique way as $\sum_{i \in \Lambda} w_i$, where Λ is some finite subset of Ω , which suffices to prove that $V = \bigoplus_{i \in \Omega} W_i$. \square

Let W be a subspace of a vector space V over a field F . A subspace Y of V is a **complement** of W in V if and only if $V = W \oplus Y$. We immediately note that if Y is a complement of W in V then W is a complement of Y in V . In general, a subspace of a vector space can have many complements.

Example: Each of the following subspaces of \mathbb{R}^2 is a complement of each of the others in \mathbb{R}^2 :

$$\begin{aligned} W_1 &= \left\{ \begin{bmatrix} a \\ 0 \end{bmatrix} \mid a \in \mathbb{R} \right\}; & W_2 &= \left\{ \begin{bmatrix} 0 \\ b \end{bmatrix} \mid b \in \mathbb{R} \right\}; \\ W_3 &= \left\{ \begin{bmatrix} c \\ c \end{bmatrix} \mid c \in \mathbb{R} \right\}; & \text{and } W_4 &= \left\{ \begin{bmatrix} d \\ 2d \end{bmatrix} \mid d \in \mathbb{R} \right\}. \end{aligned}$$

(5.14) Proposition: Every subspace W of a vector space V over a field F has at least one complement in V .

Proof: If W is improper, then $\{0_V\}$ is a complement of W in V . Similarly, V is a complement of $\{0_V\}$ in V . Otherwise, let B be a basis of W . By Proposition 5.8, we know that there exists a linearly-independent subset D of V such that $B \cup D$ is a basis of V . Then FD is a complement of W in V . \square

Example: Let F be a field of characteristic other than 2, let n be a positive integer, and let V be a vector space over F . Let $W = \mathcal{M}_{n \times n}(V)$, which is also a vector space over F . Let W_1 be the set of all those matrices $A = [v_{ij}]$ in W satisfying $v_{ij} = v_{ji}$ for all $1 \leq i, j \leq n$, and let W_2 be the set of all those matrices $A = [v_{ij}]$ in W satisfying $v_{ij} = -v_{ji}$ for all $1 \leq i, j \leq n$. These two subspaces are disjoint. If $A = [v_{ij}]$ is an arbitrary matrix in W , then we can write $A = B + C$, where $B = [y_{ij}]$

is the matrix defined by $y_{ij} = \frac{1}{2}(v_{ij} + v_{ji})$ for all $1 \leq i, j \leq n$, and $C = [z_{ij}]$ is the matrix defined by $z_{ij} = \frac{1}{2}(v_{ij} - v_{ji})$ for all $1 \leq i, j \leq n$. Note that $A \in W_1$ and $B \in W_2$. Thus $V = W_1 \oplus W_2$.

Example: A function $f \in \mathbb{R}^{\mathbb{R}}$ is **even** if and only if $f(a) = f(-a)$ for all $a \in \mathbb{R}$; it is **odd** if and only if $f(a) = -f(-a)$ for all $a \in \mathbb{R}$. The set W of all even functions is clearly a subspace of $\mathbb{R}^{\mathbb{R}}$, as is the set Y of all odd functions, and these two subspaces are disjoint. Moreover, if $f \in \mathbb{R}^{\mathbb{R}}$ then $f = f_1 + f_2$, where the function $f_1 : x \mapsto \frac{1}{2}[f(x) + f(-x)]$ is in W and the function $f_2 : x \mapsto \frac{1}{2}[f(x) - f(-x)]$ is in Y . Thus Y is a complement of W in $\mathbb{R}^{\mathbb{R}}$.

(5.15) Proposition: Let F be a field which is not finite and let V be a vector space over F having dimension at least 2. Then every proper nontrivial subspace W of V has infinitely-many complements in V .

Proof: By Proposition 5.14 we know that W has at least one complement U in V . Choose a basis B for U . If $0_V \neq w \in W$, then by Proposition 3.2(9) and the fact that F is infinite, we know that Fw is an infinite subset of W . Thus we know that the set W is infinite. For each $w \in W$, let $Y_w = F\{u + w \mid u \in B\}$. We claim that each of these spaces is a complement of W in V . Indeed, assume that $v \in W \cap Y_w$. Then there exist elements u_1, \dots, u_n of B and scalars c_1, \dots, c_n in F satisfying $v = \sum_{i=1}^n c_i(u_i + w)$. But then $\sum_{i=1}^n c_i u_i = v - (\sum_{i=1}^n c_i)w \in W \cap U = \{0_V\}$ and since the set $\{u_1, \dots, u_n\}$ is linearly independent, we see that $c_i = 0$ for all i . This shows that $v = 0_V$, and we have thus shown that W and Y_w are disjoint. If v is an arbitrary element of V , let us write $v = x + (\sum_{i=1}^n c_i u_i)$, where $x \in W$, the vectors u_1, \dots, u_n belong to B , and the scalars c_1, \dots, c_n belong to F . Then $v = [x - (\sum_{i=1}^n c_i)w] + \sum_{i=1}^n c_i(u_i + w) \in W + Y_w$ and thus we have shown that $V = W + Y_w$ and so Y_w is a complement of W in V .

We are left to show that all of these complements are indeed different from each other. Indeed, assume that $w \neq x$ are elements of W satisfying $Y_w = Y_x$. If $u \in B$ then there exist elements u_1, \dots, u_n of B and scalars c_1, \dots, c_n such that $u + w = \sum_{i=1}^n c_i(u_i + x)$. From this it follows that $u - \sum_{i=1}^n c_i u_i = (\sum_{i=1}^n c_i)x - w$ and this belongs to $W \cap Y_w = \{0_V\}$. But B is a linearly-independent set and so u has to equal to one of the u_h for some $1 \leq h \leq n$, and we must have $c_i = 0$ for $i \neq h$ and $c_h = 1$. Hence $x - w = 0_V$, namely $x = w$. This is a contradiction, and so the Y_w must all be distinct.

(5.16) Proposition (Grassmann's Theorem): Let V be a vector space over a field F and let W and Y be subspaces of V

satisfying the condition that $W + Y$ is finite-dimensional. Then $\dim(W + Y) = \dim(W) + \dim(Y) - \dim(W \cap Y)$.

Proof: Let $U_0 = W \cap Y$, which is a subspace both of W and of Y . In particular, U_0 has a complement U_1 in W and a complement U_2 in Y . Then $W + Y = U_0 + U_1 + U_2$. We claim that in fact $W + Y = U_0 \oplus U_1 \oplus U_2$. Indeed, assume that $u_0 + u_1 + u_2 = 0_V$, where $u_j \in U_j$ for $j = 0, 1, 2$. Then $u_1 = -u_2 - u_0 \in W \cap Y = U_0$. But U_0 and U_1 are disjoint and so $u_1 = 0_V$. Therefore $u_0 = -u_2 \in U_0 \cap U_2 = \{0_V\}$. Therefore $u_0 = 0_V$ and $u_2 = 0_V$ as well. Thus we see that the set $\{U_0, U_1, U_2\}$ is independent. Therefore, from the definition of the complement, we have

$$\dim(W + Y) = \dim(U_0) + \dim(U_1) + \dim(U_2) = \dim(W) + \dim(U_2)$$

and this equals $\dim(W) + \dim(Y) - \dim(W \cap Y)$ since $Y = U_2 \oplus (W \cap Y)$. \square

Example: Consider the subspaces $W_1 = \mathbb{R} \left\{ \begin{bmatrix} 1 \\ 0 \\ 2 \end{bmatrix}, \begin{bmatrix} 1 \\ 2 \\ 2 \end{bmatrix} \right\}$ and $W_2 = \mathbb{R} \left\{ \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix} \right\}$ of \mathbb{R}^3 . Each one of these subspaces has

dimension 2, and so we see that $2 \leq \dim(W_1 + W_2) \leq 3$. By Proposition 5.16, we see that, as a result of this, we have $1 \leq \dim(W_1 \cap W_2) \leq 2$. In order to ascertain the exact dimension of $W_1 \cap W_2$ we must find a basis for it. If $v \in W_1 \cap W_2$ then there exist scalars a, b, c, d satisfying

$$a \begin{bmatrix} 1 \\ 0 \\ 2 \end{bmatrix} + b \begin{bmatrix} 1 \\ 2 \\ 2 \end{bmatrix} = c \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} + d \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix}, \text{ and so } a + b = c, \quad 2b = c + d,$$

and $2a + 2b = d$, from which we conclude that $b = -3a$, $c = -2a$, and

$$d = -4a. \text{ Thus } v \text{ has to be of the form } (-2a) \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} + (-4a) \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix} =$$

$$a \begin{bmatrix} -2 \\ -6 \\ -4 \end{bmatrix}, \text{ which shows that } W_1 \cap W_2 = \mathbb{R} \begin{bmatrix} -2 \\ -6 \\ -4 \end{bmatrix}, \text{ and so it has dimension 1.}$$

Exercises

Exercise 131 Let v_1, v_2 , and v_3 be elements of a vector space V over a field F and let $c_1, c_2, c_3 \in F$. Is the subset

$$\{c_2v_3 - c_3v_2, c_1v_2 - c_2v_1, c_3v_1 - c_1v_3\}$$

of V necessarily linearly dependent?

Exercise 132 For which values of the real number t is the subset

$$\left\{ \begin{bmatrix} \cos(t) + i \sin(t) \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ \cos(t) - i \sin(t) \end{bmatrix} \right\}$$

of \mathbb{C}^2 linearly dependent?

Exercise 133 Consider the functions $f : x \mapsto 5^x$ and $g : x \mapsto 5^{2x}$. Is $\{f, g\}$ a linearly-dependent subset of $\mathbb{R}^{\mathbb{R}}$?

Exercise 134 Find $a, b \in \mathbb{Q}$ such that the subset $\left\{ \begin{bmatrix} 2 \\ a-b \\ 1 \end{bmatrix}, \begin{bmatrix} a \\ b \\ 3 \end{bmatrix} \right\}$

of \mathbb{Q}^3 is linearly dependent.

Exercise 135 Let V be a vector space over a field F and let $n > 1$ be

an integer. Let Y be the set of all vectors $\begin{bmatrix} v_1 \\ \vdots \\ v_n \end{bmatrix} \in V^n$ satisfying the

condition that the set $\{v_1, \dots, v_n\}$ is linearly dependent. Is Y necessarily a subspace of V^n ?

Exercise 136 Is the subset $\left\{ \begin{bmatrix} 1+i \\ 3+8i \\ 5+7i \end{bmatrix}, \begin{bmatrix} 1-i \\ 5 \\ 2+i \end{bmatrix}, \begin{bmatrix} 1+i \\ 3+2i \\ 4-i \end{bmatrix} \right\}$ of

\mathbb{C}^3 linearly independent, when we consider \mathbb{C}^3 as a vector space over \mathbb{C} ? Is it linearly independent when we consider \mathbb{C}^3 as a vector space over \mathbb{R} ?

Exercise 137 For each nonnegative integer n , let $f_n \in \mathbb{R}^{\mathbb{R}}$ be the function defined by $f_n : x \mapsto \sin^n(x)$. Is the subset $\{f_n \mid n \geq 0\}$ of $\mathbb{R}^{\mathbb{R}}$ linearly independent?

Exercise 138 Let $V = C(-1, 1)$, which is a vector space over \mathbb{R} . Let $f, g \in V$ be the functions defined by $f : x \mapsto x^2$ and $g : x \mapsto |x|x$. Is $\{f, g\}$ linearly independent?

Exercise 139 Let V be a vector space over $GF(5)$ and let $v_1, v_2, v_3 \in V$. Is the subset $\{v_1 + v_2, v_1 - v_2 + v_3, 2v_2 + v_3, v_2 + v_3\}$ of V linearly independent?

Exercise 140 Let F be a field of characteristic different from 2 and let V be a vector space over F containing a linearly-independent subset $\{v_1, v_2, v_3\}$. Show that the set $\{v_1 + v_2, v_2 + v_3, v_1 + v_3\}$ is also linearly independent.

Exercise 141 Is the subset $\left\{ \begin{bmatrix} 1 \\ 1 \\ 2 \\ 2 \end{bmatrix}, \begin{bmatrix} 1 \\ 2 \\ 1 \\ 2 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ 1 \\ 2 \end{bmatrix}, \begin{bmatrix} 0 \\ 2 \\ 2 \\ 0 \end{bmatrix} \right\}$ of $GF(3)^4$ linearly independent?

Exercise 142 Let $t \leq n$ be positive integers and, for all $1 \leq i \leq t$, let

$$v_i = \begin{bmatrix} a_{i1} \\ \vdots \\ a_{in} \end{bmatrix} \quad \text{be a vector in } \mathbb{R}^n \text{ chosen so that } 2|a_{jj}| > \sum_{i=1}^t |a_{ij}| \text{ for}$$

all $1 \leq j \leq n$. Show that is linearly independent.

Exercise 143 If $\{v_1, v_2, v_3, v_4\}$ is a linearly-independent subset of a vector space V over the field \mathbb{Q} , is the set

$$\{3v_1 + 2v_2 + v_3 + v_4, 2v_1 + 5v_2, 3v_3 + 2v_4, 3v_1 + 4v_2 + 2v_3 + 3v_4\}$$

linearly independent as well?

Exercise 144 Let A be a subset of \mathbb{R} having at least three elements and let $f_1, f_2, f_3 \in \mathbb{R}^A$ be the functions defined by $f_i : x \mapsto x^{i-1}2^{x-1}$. Is the set $\{f_1, f_2, f_3\}$ linearly independent?

Exercise 145 Let $F = GF(5)$ and let $V = F^F$, which is a vector space over F . Let $f : x \mapsto x^2$ and $g : x \mapsto x^3$ be elements of V . Find an element h of V such that $\{f, g, h\}$ is linearly independent.

Exercise 146 Consider \mathbb{R} as a vector space over \mathbb{Q} . Is the subset $\{(a - \pi)^{-1} \mid a \in \mathbb{Q}\}$ of this space linearly independent?

Exercise 147 In the vector space $V = \mathbb{R}^{\mathbb{R}}$ over \mathbb{R} , consider the functions

$$f_1 : x \mapsto \ln \left(\frac{(x^2 + 1)^3}{x^4 + 7} \right),$$

$f_2 : x \mapsto \ln(\sqrt{x^2 + 1})$, and $f_3 : x \mapsto \ln(x^4 + 7)$. Is the subset $\{f_1, f_2, f_3\}$ of V linearly independent?

Exercise 148 Show that the subset $\left\{ \begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 2 \end{bmatrix}, \begin{bmatrix} 2 \\ 0 \\ 1 \end{bmatrix} \right\}$ of $GF(p)^3$ is linearly independent if and only if $p \neq 3$.

Exercise 149 Let F be a subfield of a field K and let n be a positive integer. Show that a nonempty linearly-independent subset D of F^n remains linearly independent when considered as a subset of K^n .

Exercise 150 Let $F = GF(5)$ and let $V = F^F$. For $4 \leq k \leq 7$, let $f_k \in V$ be defined by $f_k : a \mapsto a^k$. Is the subset $\{f_k \mid 4 \leq k \leq 7\}$ of V linearly independent?

Exercise 151 Let V be a vector space over \mathbb{R} . For vectors $v \neq w$ in V , let $K(v, w)$ be the set of all vectors in V of the form $(1-a)v + aw$, where $0 \leq a \leq 1$. Given vectors $v, w, y \in V$ satisfying the condition that the set $\{w-v, y-v\}$ is linearly independent (and so, in particular, its elements are distinct), show that the set

$$K(v, \tfrac{1}{2}(w+y)) \cap K(w, \tfrac{1}{2}(v+y)) \cap K(y, \tfrac{1}{2}(v+w))$$

is nonempty, and determine how many elements it can have.

Exercise 152 Let V be a vector space finitely generated over a field F and let $B = \{v_1, \dots, v_n\}$ be a basis for V . Let $y \in V \setminus B$. Show that the set $\{v_1, \dots, v_n, y\}$ has a unique minimal linearly-dependent subset.

Exercise 153 Find all of the minimal linearly-dependent subsets of the subset $\left\{ \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 4 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 2 \\ 0 \end{bmatrix}, \begin{bmatrix} 2 \\ 2 \end{bmatrix} \right\}$ of \mathbb{Q}^2 .

Exercise 154 Let F be a field of characteristic other than 2. Let V be the subspace of $F[X]$ consisting of all polynomials of degree at most 3. Is $\{X+2, X^2+1, X^3+X^2, X^3-X^2\}$ a basis of V ?

Exercise 155 Let $\{v_1, \dots, v_n\}$ be a basis for a vector space V over a field F . Is the set $\{v_1+v_2, v_2+v_3, \dots, v_{n-1}+v_n, v_n+v_1\}$ necessarily also a basis for V over F ?

Exercise 156 Is $\{1+2\sqrt{5}, -3+\sqrt{5}\}$ a basis for $\mathbb{Q}(\sqrt{5})$ as a vector space over \mathbb{Q} ?

Exercise 157 For which values of $a \in \mathbb{R}$ is the set

$$\left\{ \begin{bmatrix} a & 2a \\ 2 & 3a \end{bmatrix}, \begin{bmatrix} 1 & 2 \\ 2a & 3 \end{bmatrix}, \begin{bmatrix} 1 & 2a \\ a+1 & a+2 \end{bmatrix}, \begin{bmatrix} 1 & a+1 \\ 2 & 2a=1 \end{bmatrix} \right\}$$

a basis for $\mathcal{M}_{2 \times 2}(\mathbb{R})$ as a vector space over \mathbb{R} ?

Exercise 158 Let A be a nonempty finite set and let P be the collection of all subsets of A . Let F be a field and let K be the vector space over F having basis $\{v_B \mid B \in P\}$. Define an operation \bullet on K by setting $(\sum_{B \in P} a_B v_B) \bullet (\sum_{C \in P} d_C v_C) = \sum_{B \in P} \sum_{C \in P} a_B d_C v_{B \cup C}$. Is K an F -algebra? If so, is it associative?

Exercise 159 Let F be an algebraically-closed field and let (K, \bullet) be an associative F -algebra having a basis $\{v_1, v_2\}$ as a vector space over F . Show that $v_2^2 = v_2$ or $v_2^2 = 0_K$.

Exercise 160 Find a basis for the subspace W of \mathbb{R}^4 generated by

$$\left\{ \begin{bmatrix} 4 \\ 2 \\ 6 \\ -2 \end{bmatrix}, \begin{bmatrix} 1 \\ -1 \\ 3 \\ -1 \end{bmatrix}, \begin{bmatrix} 1 \\ 2 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 5 \\ -3 \\ 1 \end{bmatrix} \right\}.$$

Exercise 161 For each real number a , let $f_a \in \mathbb{R}^{\mathbb{R}}$ be defined by

$$f_a : r \mapsto \begin{cases} 1 & \text{if } r = a \\ 0 & \text{otherwise} \end{cases}.$$

Is $\{f_a \mid a \in \mathbb{R}\}$ a basis for $\mathbb{R}^{\mathbb{R}}$ over \mathbb{R} ?

Exercise 162 Let A be a nonempty finite set and let V be the collection of all subsets of A , which is a vector space over $GF(2)$. For each $a \in A$, let $v_a = \{a\}$. Is $\{v_a \mid a \in A\}$ a basis for V ?

Exercise 163 Let F be a field and let $a, b, c \in F$. Determine whether

$$\left\{ \begin{bmatrix} 1 \\ a \\ b \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ c \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ a + b + c \end{bmatrix} \right\} \text{ is a basis for } F^3.$$

Exercise 164 Let V be a vector space finitely generated over a field F having a basis $\{v_1, \dots, v_n\}$. Is $\{v_1, \sum_{i=1}^2 v_i, \dots, \sum_{i=1}^n v_i\}$ necessarily a basis for V ?

Exercise 165 Let $F = GF(p)$ for some prime integer p , let n be a positive integer, and let V be a vector space of dimension n over F . In how many ways can we choose a basis for V ?

Exercise 166 Let V be a three-dimensional vector space over a field F , with basis $\{v_1, v_2, v_3\}$. Is $\{v_1 + v_2, v_2 + v_3, v_1 - v_3\}$ a basis for V ?

Exercise 167 Let V be a vector space of finite dimension n over \mathbb{C} having basis $\{v_1, \dots, v_n\}$. Show that $\{v_1, \dots, v_n, iv_1, \dots, iv_n\}$ is a basis for V , considered as a vector space over \mathbb{R} .

Exercise 168 Let V be a vector space of finite dimension $n > 0$ over \mathbb{R} and, for each positive integer i , let U_i be a proper subspace of V . Show that $V \neq \bigcup_{i=1}^{\infty} U_i$.

Exercise 169 Let V be a vector space over a field F which is not finite dimensional, and let W be a proper subspace of V . Show that there exists an infinite collection $\{Y_1, Y_2, \dots\}$ of subspaces of V satisfying $\bigcap_{i=1}^{\infty} Y_i \subseteq W$ but $\bigcap_{i=1}^n Y_i \not\subseteq W$ for all $n \geq 1$.

Exercise 170 Let V be the subspace of $\mathbb{R}[X]$ consisting of all polynomials of degree at most 5, and let $A = \{X^5 + X^4, X^5 - 7X^3, X^5 - 1, X^5 + 3X\}$. Show that this subset of V is linearly independent and extend it to a basis of V .

Exercise 171 Let V be a vector space of finite dimension n over a field F , and let W be a subspace of V of dimension $n - 1$. If U is a subspace of V not contained in W , show that $\dim(W \cap U) = \dim(U) - 1$.

Exercise 172 Let a, b, c, d be rational numbers such that

$$\{a + c\sqrt{3}, b + d\sqrt{3}\}$$

is a basis for $\mathbb{Q}(\sqrt{3})$ as a vector space over \mathbb{Q} . Is $\{c + a\sqrt{3}, d + b\sqrt{3}\}$ a basis for $\mathbb{Q}(\sqrt{3})$ as a vector space over \mathbb{Q} ? Is $\{a + c\sqrt{5}, b + d\sqrt{5}\}$ a basis for $\mathbb{Q}(\sqrt{5})$ as a vector space over \mathbb{Q} ?

Exercise 173 Find a real number a such that

$$\dim \left(\mathbb{R} \left\{ \begin{bmatrix} -9 \\ a \\ -1 \\ -5 \\ -14 \end{bmatrix}, \begin{bmatrix} 2 \\ -5 \\ 3 \\ 0 \\ 2 \end{bmatrix}, \begin{bmatrix} 1 \\ 4 \\ -1 \\ 1 \\ 2 \end{bmatrix}, \begin{bmatrix} 3 \\ -1 \\ 2 \\ 1 \\ 4 \end{bmatrix}, \begin{bmatrix} -1 \\ 9 \\ -4 \\ 1 \\ 0 \end{bmatrix} \right\} \right) = 2.$$

Exercise 174 Let $W = \mathbb{R} \left\{ \begin{bmatrix} 2 \\ 1 \\ 3 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 2 \\ 0 \\ 1 \end{bmatrix}, \begin{bmatrix} -1 \\ 1 \\ -3 \\ 0 \end{bmatrix} \right\} \subseteq \mathbb{R}^4$. Determine

the dimension of W and find a basis for it.

Exercise 175 Consider the vectors $v_1 = \begin{bmatrix} 0 \\ 1 \\ 0 \\ 1 \\ 0 \end{bmatrix}$, $v_2 = \begin{bmatrix} 7 \\ 4 \\ 1 \\ 8 \\ 3 \end{bmatrix}$, $v_3 =$

$\begin{bmatrix} 0 \\ 3 \\ 0 \\ 4 \\ 0 \end{bmatrix}$, $v_4 = \begin{bmatrix} 1 \\ 9 \\ 5 \\ 7 \\ 1 \end{bmatrix}$, and $v_5 = \begin{bmatrix} 0 \\ 1 \\ 0 \\ 5 \\ 0 \end{bmatrix}$ in the vector space \mathbb{Q}^5 . Do

there exist rational numbers a_{ij} , for $1 \leq i, j \leq 5$, such that the subset $\left\{ \sum_{j=1}^5 a_{ij} v_j \mid 1 \leq i \leq 5 \right\}$ of \mathbb{Q}^5 is linearly independent?

Exercise 176 Let F be a subfield of a field K satisfying the condition that K is finitely generated as a vector space over F . For each $c \in K$, show that there exists a nonzero polynomial $p(X) \in F[X]$ satisfying $p(c) = 0$.

Exercise 177 Let $W = \mathbb{R} \left\{ \begin{bmatrix} 1 \\ 2 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} -1 \\ 1 \\ 1 \\ 1 \end{bmatrix} \right\} \subseteq \mathbb{R}^4$ and let $V = \mathbb{R} \left\{ \begin{bmatrix} 2 \\ -1 \\ 0 \\ 1 \end{bmatrix}, \begin{bmatrix} -5 \\ 6 \\ 3 \\ 0 \end{bmatrix} \right\}$. Compute $\dim(W + V)$ and $\dim(W \cap V)$.

Exercise 178 Let F be a subfield of a field K satisfying the condition that the dimension of K as a vector space over F is finite and equal to r . Let V be a vector space of finite dimension $n > 0$ over K . Find the dimension of V as a vector space over F .

Exercise 179 Let V be a vector space over a field F having infinite dimension over F . Show that there exists a countably-infinite collection of proper subspaces of V , the union of which equals V .

Exercise 180 Let $F = GF(p)$, where p is a prime integer, and let V be a vector space over F having finite dimension n . How many subspaces of dimension 1 does V have?

Exercise 181 Let W be the subset of $\mathbb{R}^{\mathbb{R}}$ consisting of all functions of the form $x \mapsto a \cdot \cos(x - b)$, for real numbers a and b . Show that W is a subspace of $\mathbb{R}^{\mathbb{R}}$ and find its dimension.

Exercise 182 Let $W = \mathbb{R} \left\{ \begin{bmatrix} 4 \\ 3 \\ 2 \\ 1 \end{bmatrix}, \begin{bmatrix} 6 \\ 2 \\ 2 \\ 2 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ 1 \\ 2 \end{bmatrix} \right\} \subseteq \mathbb{R}^4$ and let $Y = \mathbb{R} \left\{ \begin{bmatrix} 4 \\ -2 \\ 0 \\ -2 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 3 \\ 2 \end{bmatrix} \right\} \subseteq \mathbb{R}^4$. Find $\dim(W + Y)$ and $\dim(W \cap Y)$.

Exercise 183 Let V be a vector space of finite dimension n over a field F and let W and Y be distinct subspaces of V , each of dimension $n - 1$. What is $\dim(W \cap Y)$?

Exercise 184 Let V be a finite-dimensional vector space over a field F and let B be a basis of V satisfying the condition that

$$\left\{ \left[\begin{array}{c} w \\ w \end{array} \right] \mid w \in B \right\}$$

is a basis for V^2 . What is the dimension of V ?

Exercise 185 Let F be a field and let V be the subspace of $F[X]$ consisting of all polynomials of degree at most 4. Find a complement for V in $F[X]$.

Exercise 186 Let F be a field and let V be the subspace of $F[X]$ consisting of all polynomials of the form $(X^3 + X + 1)p(X)$ for some $p(X) \in F[X]$. Find a complement for V in $F[X]$.

Exercise 187 Let B be a nonempty proper subset of a set A . Let F be a field and let $V = F^A$. Let W be the subspace of V consisting of all those functions $f \in V$ satisfying $f(b) = 0$ for all $b \in B$. Find a complement of W in V .

Exercise 188 Let F be a field of characteristic other than 2, let V be a

vector space over F , and let $U = \left\{ \left[\begin{array}{c} v \\ v' \\ v + v' \end{array} \right] \mid v, v' \in V \right\} \subseteq V^3$. Is

$Y = \left\{ \left[\begin{array}{c} v \\ v \\ v \end{array} \right] \mid v \in V \right\}$ a complement of U in V ?

Exercise 189 Let F be a field and let $p(X) \in F[X]$ have positive degree k . Let W be the subspace of $F[X]$ composed of all polynomials of the form $p(X)g(X)$ for some $g(X) \in F[X]$. Show that W has a complement in $F[X]$ of dimension k .

Exercise 190 Let V be a vector space over a field F which is not finite dimensional, and let $V \supset W_1 \supset W_2 \supset \dots$ be a chain of subspaces of V , each properly contained in the one before it. Is the subspace $\bigcap_{i=1}^{\infty} W_i$ of V necessarily finite-dimensional?

Exercise 191 Let V be a vector space finite dimensional over a field F . Let W and Y be subspaces of V and assume that there is a function $f \in F^V$ satisfying the condition that $f(w) < f(y)$ for all $0_V \neq w \in W$ and $0_V \neq y \in Y$. Show that $\dim(W) + \dim(Y) \leq \dim(V)$.

Exercise 192 Let (K, \bullet) be a division algebra of dimension 2 over \mathbb{R} containing an element v_1 which satisfies the condition that $v_1 \bullet v = v = v \bullet v_1$ for all $v \in V$. Show that $(K, +, \bullet)$ is a field.

Exercise 193 For each $a \in \mathbb{R}$, the set $\mathbb{Q}[a] = \{p(a) \mid p(X) \in \mathbb{Q}[X]\}$ is a subspace of \mathbb{R} , considered as a vector space over \mathbb{Q} . Find all pairs (a, b)

of real numbers $a \neq b$ satisfying the condition that the set $\{\mathbb{Q}[a], \mathbb{Q}[b]\}$ is independent over \mathbb{Q} .

Exercise 194 Let V be a vector space over a field F . Find a necessary and sufficient condition for there to exist subspaces W and W' of V such that $\{\{0_V\}, W, W'\}$ is independent.

Exercise 195 Let (K, \bullet) be a unital \mathbb{R} -algebra (not necessarily associative) with multiplicative identity e , and let $\{v_i \mid i \in \Omega\}$ be a basis for K over \mathbb{R} containing e (which is equal to v_t for some $t \in \Omega$). If $v = \sum_{i \in \Omega} c_i v_i \in K$, set $\bar{v} = c_t v_t - \sum_{i \neq t} c_i v_i$. If $v \in K$, is it true that $v\bar{v} = \bar{v}v$ and $\bar{\bar{v}} = v$? (Note that this construction generalizes the notion of the conjugate of a complex number.)

Exercise 196 For each nonnegative integer n , define the subsets P_n , A_n , and F_n of \mathbb{R} as follows:

- (1) $P_0 = \emptyset$, $A_0 = \{1\}$, and $F_0 = \mathbb{Q}$;
- (2) If $n > 0$, then P_n is the set of the first n prime integers, A_n consists of 1 and the set of square roots of products of distinct elements of P_n , and $F_n = \mathbb{Q}A_n$.

Show that each A_n is a linearly-independent subset of \mathbb{R} , considered as a vector space over \mathbb{Q} , and that F_n is a subfield of \mathbb{R} , having the property that every element of F_n the square of which belongs to \mathbb{Q} must belong to $\mathbb{Q}a$, for some $a \in A_n$.

6

Linear transformations

Let V and W be vector spaces over a field F . A function $\alpha : V \rightarrow W$ is a **linear transformation**¹ or **homomorphism** if and only if for all $v_1, v_2 \in V$ and $a \in F$ we have $\alpha(v_1 + v_2) = \alpha(v_1) + \alpha(v_2)$ and $\alpha(av_1) = a\alpha(v_1)$. We note that, as a consequence of the second condition, we have $\alpha(0_V) = \alpha(0 \cdot 0_V) = 0\alpha(0_V) = 0_W$. If (K, \bullet) and $(L, *)$ are F -algebras, then a linear transformation $\alpha : K \rightarrow L$ is a **homomorphism of F -algebras** if it is a linear transformation and, in addition, satisfies $\alpha(v_1 \bullet v_2) = \alpha(v_1) * \alpha(v_2)$ for all $v_1, v_2 \in K$. If both K and L are unital, then it is a **homomorphism of unital F -algebras** if it also sends the identity element of K for \bullet to the identity element of L for $*$.

Example: Let V be a vector space over a field F . Every scalar $c \in F$ defines a linear transformation $\sigma_c : V \rightarrow V$ given by $\sigma_c : v \mapsto cv$. In



¹ Linear transformations between finite-dimensional vector spaces were studied by Peano. Linear transformations between infinite-dimensional spaces were first considered in the late 19th century by Italian mathematician **Salvatore Pincherle**.

particular, σ_1 is the identity function $v \mapsto v$ and σ_0 is the 0-function $v \mapsto 0_V$.

Example: Let F be a field and let a_1, \dots, a_6 be scalars in F . The function $\alpha : F^2 \rightarrow F^3$ defined by $\alpha : \begin{bmatrix} c_1 \\ c_2 \end{bmatrix} \mapsto \begin{bmatrix} a_1c_1 + a_2c_2 \\ a_3c_1 + a_4c_2 \\ a_5c_1 + a_6c_2 \end{bmatrix}$ is a linear transformation.

The previous example can be generalized in an extremely significant manner. Let k and n be positive integers and let F be a field. Every matrix $A = [a_{ij}] \in \mathcal{M}_{k \times n}(F)$ defines a linear transformation from F^n

to F^k given by $\begin{bmatrix} c_1 \\ c_2 \\ \vdots \\ c_n \end{bmatrix} \mapsto \begin{bmatrix} a_{11}c_1 + \dots + a_{1n}c_n \\ a_{21}c_1 + \dots + a_{2n}c_n \\ \vdots \\ a_{k1}c_1 + \dots + a_{kn}c_n \end{bmatrix}$. In what follows,

we will show that every linear transformation from F^n to F^k can be defined in this manner.

Example: Let F be a field of characteristic 0. Then there are linear transformations α and β from $F[X]$ to itself defined by

$$\alpha : \sum_{i=0}^{\infty} a_i X^i \mapsto \sum_{i=0}^{\infty} i a_i X^i \quad \text{and} \quad \beta : \sum_{i=0}^{\infty} a_i X^i \mapsto \sum_{i=0}^{\infty} (1+i)^{-1} a_i X^{i+1}.$$

(By $1+i$, we mean the sum of $1+i$ copies of the identity element for multiplication of F ; since the characteristic of F is 0, we know that this element is nonzero, and so is a unit in F .)

Example: Let V and W be vector spaces over a field F and let k and n be positive integers. For all $1 \leq i \leq k$ and $1 \leq j \leq n$, let $\alpha_{ij} : V \rightarrow W$ be a linear transformation. Then there is a linear transformation from $\mathcal{M}_{k \times n}(V)$ to $\mathcal{M}_{k \times n}(W)$ defined by

$$\begin{bmatrix} v_{11} & \dots & v_{1n} \\ \vdots & & \vdots \\ v_{k1} & \dots & v_{kn} \end{bmatrix} \mapsto \begin{bmatrix} \alpha_{11}(v_{11}) & \dots & \alpha_{1n}(v_{1n}) \\ \vdots & & \vdots \\ \alpha_{k1}(v_{k1}) & \dots & \alpha_{kn}(v_{kn}) \end{bmatrix}.$$

Example: Let V be the subspace over $\mathbb{R}^{\mathbb{R}}$ consisting of all differentiable functions. For each $f \in V$, we define a function $Df : \mathbb{R} \times \mathbb{R} \rightarrow \mathbb{R}$, called the **differential** of f , by setting $Df : (a, b) \mapsto f'(a)b$, where f' is the derivative of f . Then the function $D : V \rightarrow \mathbb{R}^{\mathbb{R} \times \mathbb{R}}$ given by $f \mapsto Df$

is a linear transformation. These linear transformations play an important part in differential geometry.

Example: Linear transformations are considered nice from an algebraic point of view, but may be less so from an analytic point of view. Let B be a Hamel basis of \mathbb{R} over \mathbb{Q} . Then for each real number r there exists a unique finite subset $\{u_1(r), \dots, u_{n(r)}(r)\}$ of B and scalars $a_1(r), \dots, a_{n(r)}(r)$ in \mathbb{Q} satisfying $r = \sum_{j=1}^{n(r)} a_j(r) u_j(r)$. The function from \mathbb{R} to \mathbb{R} defined by $r \mapsto \sum_{j=1}^{n(r)} a_j(r)$ is a linear transformation, but is not continuous at any $r \in \mathbb{R}$.

Let V and W be vector spaces over a field F . To any function $f : V \rightarrow W$ can associate the subset $gr(f) = \left\{ \begin{bmatrix} v \\ f(v) \end{bmatrix} \mid v \in V \right\}$ of $V \times W$, called the **graph** of f . We can use the notion of graph to characterize linear transformations in terms of subspaces.

(6.1) Proposition: Let V and W be vector spaces over a field F and let $\alpha : V \rightarrow W$ be a function. Then α is a linear transformation if and only if $gr(\alpha)$ is a subspace of $V \times W$.

Proof: Assume that α is a linear transformation. If $v, v' \in V$ and $c \in F$ then in $V \times W$ we have

$$\begin{bmatrix} v \\ \alpha(v) \end{bmatrix} + \begin{bmatrix} v' \\ \alpha(v') \end{bmatrix} = \begin{bmatrix} v + v' \\ \alpha(v) + \alpha(v') \end{bmatrix} = \begin{bmatrix} v + v' \\ \alpha(v + v') \end{bmatrix} \in gr(\alpha)$$

and $c \begin{bmatrix} v \\ \alpha(v) \end{bmatrix} = \begin{bmatrix} cv \\ c\alpha(v) \end{bmatrix} = \begin{bmatrix} cv \\ \alpha(cv) \end{bmatrix} \in gr(\alpha)$, showing that $gr(\alpha)$

is closed under taking sums and scalar multiples, and so is a subspace of $V \times W$. Conversely, if it is such a subspace then for $v, v' \in V$ and $c \in F$

we note that $\begin{bmatrix} v \\ \alpha(v) \end{bmatrix} + \begin{bmatrix} v' \\ \alpha(v') \end{bmatrix} = \begin{bmatrix} v + v' \\ \alpha(v) + \alpha(v') \end{bmatrix} \in gr(\alpha)$ and

so we must have $\alpha(v) + \alpha(v') = \alpha(v + v')$. Similarly, $c \begin{bmatrix} v \\ \alpha(v) \end{bmatrix} =$

$\begin{bmatrix} cv \\ c\alpha(v) \end{bmatrix} \in gr(\alpha)$ and so we must have $c\alpha(v) = \alpha(cv)$. Thus α is a linear transformation. \square

Let V and W be vector spaces over a field F . If α and β be linear transformations from V to W , they are, in particular, functions in W^V

and so the function $\alpha + \beta : V \rightarrow W$ is defined by $\alpha + \beta : v \mapsto \alpha(v) + \beta(v)$ for all $v \in V$. For all $v, v' \in V$ and all $c \in F$ we have

$$\begin{aligned} (\alpha + \beta)(v + v') &= \alpha(v + v') + \beta(v + v') \\ &= \alpha(v) + \alpha(v') + \beta(v) + \beta(v') \\ &= (\alpha + \beta)(v) + (\alpha + \beta)(v'). \end{aligned}$$

and $(\alpha + \beta)(cv) = \alpha(cv) + \beta(cv) = c\alpha(v) + c\beta(v) = c[\alpha(v) + \beta(v)] = c(\alpha + \beta)(v)$.

Thus we see that $\alpha + \beta$ is a linear transformation from V to W . If $c \in F$ is a scalar then the function $c\alpha$ from V to W is defined by $c\alpha : v \mapsto c\alpha(v)$ and this, again, is a linear transformation from V to W . It is easy to check that the set of all linear transformations from V to W is a subspace of W^V , which we will denote by $\text{Hom}(V, W)$. This means that we can apply concepts we have already considered for vectors to linear transformations. For example, we can talk about a linearly-dependent or linearly-independent set of linear transformations from a vector space V over a field F to a vector space W over F . However, we must be very careful to remember that when we are doing so, we are working in the space $\text{Hom}(V, W)$, and not in either V or W . The following example illustrates the pitfalls one can encounter.

Example: Let V and W be vector spaces over the same field F . A nonempty subset $D = \{\alpha_1, \dots, \alpha_n\}$ of $\text{Hom}(V, W)$ is **locally linearly dependent** if and only if the subset $\{\alpha_1(v), \dots, \alpha_n(v)\}$ of W is linearly dependent for every $v \in V$. If D is a linearly-dependent subset of $\text{Hom}(V, W)$, then there exist scalars c_1, \dots, c_n , not all of which are equal to 0, such that $\sum_{i=1}^n c_i \alpha_i$ is the 0-function. In particular, for each $v \in V$ we see that $\sum_{i=1}^n c_i \alpha_i(v) = 0_W$ and so D is locally linearly dependent. The converse, however, is false. It may be possible for D to be linearly independent and still locally linearly dependent. To see this, take $V = W = F^2$ and let $D = \{\alpha_1, \alpha_2\} \subseteq \text{Hom}(F^2, F^2)$, where we

define $\alpha_1 : \begin{bmatrix} a \\ b \end{bmatrix} \mapsto \begin{bmatrix} a \\ 0 \end{bmatrix}$ and $\alpha_2 : \begin{bmatrix} a \\ b \end{bmatrix} \mapsto \begin{bmatrix} b \\ 0 \end{bmatrix}$. If $v \in F^2$, then $\{\alpha_1(v), \alpha_2(v)\}$ is a subset of the one-dimensional subspace $F \begin{bmatrix} 1 \\ 0 \end{bmatrix}$ of

F^2 and so cannot be linearly independent. On the other hand, D is linearly independent since if there exist scalars c and d satisfying the condition that $c\alpha_1 + d\alpha_2$ is the 0-function, then

$$\begin{bmatrix} 0 \\ 0 \end{bmatrix} = c\alpha_1 \left(\begin{bmatrix} 1 \\ 0 \end{bmatrix} \right) + d\alpha_2 \left(\begin{bmatrix} 1 \\ 0 \end{bmatrix} \right) = \begin{bmatrix} c \\ 0 \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \end{bmatrix} = \begin{bmatrix} c \\ 0 \end{bmatrix},$$

which implies that $c = 0$. Similarly,

$$\begin{bmatrix} 0 \\ 0 \end{bmatrix} = c\alpha_1 \left(\begin{bmatrix} 0 \\ 1 \end{bmatrix} \right) + d\alpha_2 \left(\begin{bmatrix} 0 \\ 1 \end{bmatrix} \right) = \begin{bmatrix} 0 \\ 0 \end{bmatrix} + \begin{bmatrix} 0 \\ d \end{bmatrix} = \begin{bmatrix} 0 \\ d \end{bmatrix},$$

which implies that $d = 0$ as well.

The following proposition shows that the operation of a linear transformation is entirely determined by its action on elements of a basis. This result is extremely important, especially if the vector spaces involved are finitely generated.

(6.2) Proposition: Let V and W be vector spaces over a field F , and let B be a basis of V . If $f \in W^B$ then there is a unique linear transformation $\alpha \in \text{Hom}(V, W)$ satisfying the condition that $\alpha(u) = f(u)$ for all $u \in B$.

Proof: Since B is a basis of V , we know that each vector $v \in V$ can be written as a linear combination $v = \sum_{i=1}^n a_i u_i$ of elements of B in a unique way. We now define the function $\alpha : V \rightarrow W$ by $\alpha : v \mapsto \sum_{i=1}^n a_i f(u_i)$. This function is well defined as a result of the uniqueness of representation of v , as was shown in Proposition 5.4. Moreover, it is clear that α is a linear transformation. If $\beta : V \rightarrow W$ is a linear transformation satisfying the condition that $\beta(u) = f(u)$ for all $u \in B$ then $\beta(v) = \beta(\sum_{i=1}^n a_i u_i) = \sum_{i=1}^n a_i \beta(u_i) = \sum_{i=1}^n a_i f(u_i) = \alpha(v)$ and so $\beta = \alpha$. Thus α is unique. \square

Example: We can use Proposition 6.2 to show how uncommon linear transformations really are. Let $F = GF(3)$ and let $V = F^4$. Then V has $3^4 = 81$ elements and so the number of functions from V to itself is 81^{81} . On the other hand, a basis B for V over F has 4 elements and so, since every linear transformation from V to itself is totally determined by its action on B and that any function from B to V defines such a linear transformation, we see that the number of linear transformations from V to itself is 81^4 . Therefore, the probability that a randomly-selected function from V to itself is a linear transformation is $81^4/81^{81} = 81^{-77}$, which is roughly 0.11134×10^{-146} .

(6.3) Proposition: Let V, W , and Y be vector spaces over a field F and let $\alpha : V \rightarrow W$ and $\beta : W \rightarrow Y$ be linear transformations. Then $\beta\alpha : V \rightarrow Y$ is a linear transformation.

Proof: If $v_1, v_2 \in V$ and if $a \in F$ then

$$\begin{aligned} (\beta\alpha)(v_1 + v_2) &= \beta(\alpha(v_1 + v_2)) = \beta(\alpha(v_1) + \alpha(v_2)) \\ &= \beta(\alpha(v_1)) + \beta(\alpha(v_2)) = (\beta\alpha)(v_1) + (\beta\alpha)(v_2) \end{aligned}$$

and $(\beta\alpha)(cv_1) = \beta(\alpha(cv_1)) = \beta(c\alpha(v_1)) = c\beta(\alpha(v_1)) = c(\beta\alpha)(v_1)$, which proves the proposition. \square

Example: It is often important and insightful to write a linear transformation as a composite of linear transformations of predetermined types. Consider the following situation: let $a < b$ be real numbers and let V be the vector space over \mathbb{R} consisting of all functions from the closed interval $[a, b]$ to \mathbb{R} . Let W be the subspace of V consisting of all differentiable functions, and let $\delta : W \rightarrow V$ be the function which assigns to each function $f \in W$ its derivative. For each real number $a < c < b$, let $\varepsilon_c : V \rightarrow \mathbb{R}$ be the linear transformation defined by $\varepsilon_c : g \mapsto g(c)$. Then the Intermediate Value Theorem from calculus says that the linear transformation $\beta : W \rightarrow \mathbb{R}$ defined by

$$\beta : f \mapsto \frac{f(b) - f(a)}{b - a}$$

is of the form $\varepsilon_c \delta$ for some c .

Let V and W be vector spaces over a field F and let $\alpha : V \rightarrow W$ be a linear transformation. For $w \in W$, we will denote the set

$$\{v \in V \mid \alpha(v) = w\}$$

by $\alpha^{-1}(w)$. Note that this set may be empty. In particular, we will be interested in $\alpha^{-1}(0_W) = \{v \in V \mid \alpha(v) = 0_W\}$. This set is called the **kernel** of α and is denoted by $\ker(\alpha)$. Then $\ker(\alpha)$ is never empty, since it always contains 0_V . If U is a nonempty subset of W , set $\alpha^{-1}(U) = \{\alpha^{-1}(u) \mid u \in U\}$. It is easy to verify that $\alpha^{-1}(U)$ is a subspace of V whenever U is a subspace of W .

Example: Let F be a field and let $\alpha \in \text{Hom}(F^3, F^4)$ be the linear transformation defined by $\alpha : \begin{bmatrix} a \\ b \\ c \end{bmatrix} \mapsto \begin{bmatrix} a - b \\ 0 \\ c \\ c \end{bmatrix}$. Then $\ker(\alpha) =$

$$\left\{ \begin{bmatrix} a \\ a \\ 0 \end{bmatrix} \mid a \in F \right\}.$$

(6.4) Proposition: Let V and W be vector spaces over a field F and let $\alpha \in \text{Hom}(V, W)$. Then $\ker(\alpha)$ is a subspace of V , which is trivial if and only if α is monic.

Proof: Let $v_1, v_2 \in \ker(\alpha)$ and let $a \in F$. Then $\alpha(v_1 + v_2) = \alpha(v_1) + \alpha(v_2) = 0_W + 0_W = 0_W$ and so $v_1 + v_2 \in \ker(\alpha)$. Similarly, $\alpha(av_1) = a\alpha(v_1) = a0_W = 0_W$ and so $av_1 \in \ker(\alpha)$. This proves that $\ker(\alpha)$ is a subspace of V .

If α is monic then $\alpha^{-1}(w)$ can have at most one element for each $w \in W$, and so, in particular, $\ker(\alpha) = \{0_V\}$. Conversely, suppose that $\ker(\alpha)$ is trivial and that there exist elements $v_1 \neq v_2$ of V satisfying $\alpha(v_1) = \alpha(v_2)$. Then $\alpha(v_1 - v_2) = \alpha(v_1) - \alpha(v_2) = 0_W$ and so $v_1 - v_2 \in \ker(\alpha)$. Thus $v_1 - v_2 = 0_V$ and so $v_1 = v_2$, which is a contradiction. Hence α must be monic. \square

Let V and W be vector spaces over a field and let $\alpha : V \rightarrow W$ be a linear transformation. The **image** of α is the subset $\text{im}(\alpha) = \{\alpha(v) \mid v \in V\}$ of W . This set is nonempty since $0_W = \alpha(0_V) \in \text{im}(\alpha)$. Note that $w \in \text{im}(\alpha)$ if and only if $\alpha^{-1}(w) \neq \emptyset$. If U is a nonempty subset of V , we denote the subset $\{\alpha(u) \mid u \in U\}$ of W by $\alpha(U)$. Thus $\alpha(V) = \text{im}(\alpha)$.

(6.5) Proposition: Let V and W be vector spaces over a field F and let $\alpha \in \text{Hom}(V, W)$. Then $\text{im}(\alpha)$ is a subspace of W , which is improper if and only if α is epic.

Proof: If $\alpha(v_1)$ and $\alpha(v_2)$ are in $\text{im}(\alpha)$ and if $a \in F$, then $\alpha(v_1) + \alpha(v_2) = \alpha(v_1 + v_2) \in \text{im}(\alpha)$ and similarly $a\alpha(v_1) = \alpha(av_1) \in \text{im}(\alpha)$, proving that $\text{im}(\alpha)$ is a subspace of W . The second part follows immediately from the definition of an epic function. \square

A monic linear transformation between vector spaces over a field F is called a **monomorphism**; an epic linear transformation between vector spaces is called an **epimorphism**. A bijective linear transformation between vector spaces is called an **isomorphism**. If both spaces are also F -algebras, then a bijective homomorphism of F -algebras is called an **isomorphism of F -algebras**. Similarly, a bijective homomorphism of unital F -algebras is an **isomorphism of unital F -algebras**.

Example: Let F be a field and let k and n be positive integers. For each matrix $A = [a_{ij}] \in \mathcal{M}_{k \times n}(F)$ we can define the **transpose** of A to be the matrix $A^T \in \mathcal{M}_{n \times k}(F)$ obtained from A by interchanging its

rows and columns. In other words, $A^T = \begin{bmatrix} a_{11} & \dots & a_{k1} \\ \vdots & & \vdots \\ a_{1n} & \dots & a_{kn} \end{bmatrix}$. It is easy

to check that the function $A \mapsto A^T$ is an isomorphism from $\mathcal{M}_{k \times n}(F)$ to $\mathcal{M}_{n \times k}(F)$.

Example: Let K and L be F -algebras. It is possible for a linear transformation $\alpha : K \rightarrow L$ to be an isomorphism of vector spaces without being an isomorphism of F -algebras. This is the case, for example, with the linear transformation $\alpha : \mathbb{Q}(\sqrt{2}) \rightarrow \mathbb{Q}(\sqrt{5})$ given by $\alpha : a + b\sqrt{2} \mapsto a + b\sqrt{5}$.

Example: Let V be a vector space over a field F . Any linear transformation $\alpha : V \rightarrow F$ other than the 0-function is an epimorphism. Indeed, if α is a nonzero linear transformation and if $v_0 \in V$ satisfies the condition that $\alpha(v_0) = c \neq 0$, then for any $a \in F$ we have $a = (ac^{-1})c = (ac^{-1})\alpha(v_0) = \alpha((ac^{-1})v_0) \in \text{im}(\alpha)$.

Example: Let F be a field and let $\alpha : F^{(\infty)} \rightarrow F[X]$ be the function defined by $\alpha : f \mapsto \sum_{i=0}^{\infty} f(i)X^i$, which is well-defined since only finitely many of the $f(i)$ are nonzero. This is easily checked to be an isomorphism of vector spaces.

We have already seen that if D is a basis of a vector space V over a field F then there exists a bijective function $\theta : F^{(D)} \rightarrow V$, and it is easy to verify that this is in fact an isomorphism of vector spaces. This leads us to the very important observation that for any nontrivial vector space V over a field F there exists a nonempty set Ω and an isomorphism $F^{(\Omega)} \rightarrow V$.

Let V and W be vector spaces over a field F and let B be a basis of V . Then we can define a function $\varphi : \text{Hom}(V, W) \rightarrow W^B$ by restriction: $\varphi(\alpha) : u \mapsto \alpha(u)$ for all $u \in B$. It is straightforward to check that φ is a linear transformation of vector spaces over F . Moreover, by Proposition 6.2 we see that any function $f \in W^B$ is of the form $\varphi(\alpha)$ for a unique element α of $\text{Hom}(V, W)$. Therefore φ is an isomorphism.

Let V and W be vector spaces over a field F . If $\alpha : V \rightarrow W$ is a linear transformation and $0_W \neq w \in W$ then $\alpha^{-1}(w)$ is not a subspace of V . However, the next result shows that, if it is nonempty, it is “close” to being a subspace.

(6.6) Proposition: Let $\alpha : V \rightarrow W$ be a linear transformation of vector spaces over a field F and let $w \in \text{im}(\alpha)$. For any $v_0 \in \alpha^{-1}(w)$ we have $\alpha^{-1}(w) = \{v + v_0 \mid v \in \ker(\alpha)\}$.

Proof: If $v \in \ker(\alpha)$ then $\alpha(v + v_0) = \alpha(v) + \alpha(v_0) = 0_W + w = w$ and so $v + v_0 \in \alpha^{-1}(w)$. Conversely, if $v_1 \in \alpha^{-1}(w)$ then $v_1 = (v_1 - v_0) + v_0$, where $v_1 - v_0 \in \ker(\alpha)$ since $\alpha(v_1 - v_0) = \alpha(v_1) - \alpha(v_0) = w - w = 0_W$. \square

Note that $\alpha^{-1}(w)$ is not a subspace of V but rather the result of “shifting” a subspace by adding a fixed nonzero vector to each of its elements. Such a subset of a vector space is called an **affine subset** of a vector space. Let V and W be vector spaces over a field F . An **affine transformation** $\zeta : V \rightarrow W$ is a function of the form $v \mapsto \alpha(v) + y$, for some fixed

$\alpha \in \text{Hom}(V, W)$ and $y \in W$. It is clear that the sum of two affine transformations is again an affine transformation, as is the product of an affine transformation by a scalar, so that the set $\text{Aff}(V, W)$ of all affine transformations from V to W is also a subspace of W^V which in turn contains $\text{Hom}(V, W)$ as a subspace. Indeed, $\text{Aff}(V, W) = F(\text{Hom}(V, W) \cup K)$, where K is the set of all constant functions from V to W .

Moreover, if $\zeta : V \rightarrow W$ is the affine transformation defined by $v \mapsto \alpha(v) + y$ and if $w \in W$, then $\zeta^{-1}(w) = \alpha^{-1}(w - y)$ and so is an affine subset of V .

Analysis of computational procedures in linear algebra often hinges on the fact that when we think we are computing the effect of some linear transformation $\alpha \in \text{Hom}(V, W)$, we are in fact computing that of an affine transformation $v \mapsto \alpha(v) + y$ where y is a vector arising from computational or random errors which, hopefully, is “very small” (in some sense) relative to $\alpha(v)$. Similarly, in linear models in statistics one must allow for such an affine transformation, where y is a random error vector, assumed to have expectation 0.

Example: Let $V = C(0, 1)$ and let W be the subspace of V composed of all differentiable functions having a continuous derivative. Let $\delta : W \rightarrow V$ be the linear transformation which assigns to each function $f \in W$ its derivative. Then $\ker(\delta)$ consists of all constant functions. If $g \in \text{im}(\delta)$ then $g = \delta(f)$, where f is the function $f : x \mapsto \int_0^x g(t) dt$. Thus $\delta^{-1}(g)$ consists of all functions of the form $f : x \mapsto \int_0^x g(t) dt + c$, where $c \in \mathbb{R}$.

(6.7) Proposition: If $\alpha : V \rightarrow W$ is an isomorphism of vector spaces over a field F then there exists an isomorphism $\beta : W \rightarrow V$ satisfying $\beta\alpha(v) = v$ and $\alpha\beta(w) = w$ for all $v \in V$ and all $w \in W$.

Proof: Define the function β by $\beta(w) = v$ if and only if $w = \alpha(v)$. This function is well-defined since every element w is of the form $\alpha(v)$ for a unique element $v \in V$. It is easy to check that the function β is an isomorphism which satisfies the stated conditions. \square

The function β defined in Proposition 6.7 is denoted by α^{-1} .

Let V and W be vector spaces over a field F . If there exists an isomorphism from V to W we, say that V and W are **isomorphic**, and write $V \cong W$. It is easy to see that if V, W , and Y are vector spaces over F then:

- (1) $V \cong V$;
- (2) If $V \cong W$ then $W \cong V$;
- (3) If $V \cong W$ and $W \cong Y$ then $V \cong Y$.

It is also clear that if $\alpha : V \rightarrow W$ is an isomorphism between vector spaces over F and if B is a basis of V then $\{\alpha(u) \mid u \in B\}$ is a basis of W . As an immediate consequence of this, we see that if $V \cong W$ then the dimensions of V and W are the same. The converse is true if V and W are finitely generated, as we shall now see.

(6.8) Proposition: If V and W are vector spaces of finite dimension n over a field F , then $V \cong W$.

Proof: Let $B = \{v_1, \dots, v_n\}$ be a basis for V and $D = \{w_1, \dots, w_n\}$ be a basis for W . Define the function $f \in W^B$ by setting $f : v_i \mapsto w_i$ for all $1 \leq i \leq n$. By Proposition 6.2, we know that there exists a linear transformation $\alpha \in \text{Hom}(V, W)$ satisfying the condition $\alpha(v_i) = f(v_i)$ for all $1 \leq i \leq n$. This linear transformation is epic since $\text{im}(\alpha)$ contains a basis of W . If $v = \sum_{i=1}^n a_i v_i \in \ker(\alpha)$ then $0_W = \alpha(v) = \alpha(\sum_{i=1}^n a_i v_i) = \sum_{i=1}^n a_i \alpha(v_i) = \sum_{i=1}^n a_i w_i$ and so $a_i = 0$ for all $1 \leq i \leq n$, since the set D is linearly independent. Therefore $\ker(\alpha)$ is trivial, and this shows that α is monic and hence an isomorphism. \square

(6.9) Proposition: If V and W are finitely-generated vector spaces over a field F then

- (1) There exists a monomorphism from V to W if and only if $\dim(V) \leq \dim(W)$;
- (2) There exists an epimorphism from V to W if and only if $\dim(V) \geq \dim(W)$.

Proof: (1) If there exists a monomorphism α from V to W then $V \cong \text{im}(\alpha)$ and so $\dim(W) \geq \dim(\text{im}(\alpha)) = \dim(V)$. Conversely, assume that $\dim(V) \leq \dim(W)$. Then there exists a basis $B = \{v_1, \dots, v_n\}$ of V and there exists a basis $D = \{w_1, \dots, w_t\}$ of W , where $n \leq t$. The function from B to W given by $v_i \mapsto w_i$ for all $1 \leq i \leq n$ can be extended to a linear transformation $\alpha : V \rightarrow W$, which is monic and so is a monomorphism.

(2) If there exists an epimorphism α from V to W and if $\{v_1, \dots, v_n\}$ is a basis of V , then $\{\alpha(v_i) \mid 1 \leq i \leq n\}$ is a generating set of W and so the dimension of W is at most $n = \dim(V)$. Conversely, if $n = \dim(V) \geq \dim(W) = t$, pick a basis $\{w_1, \dots, w_t\}$ of W and a basis $B = \{v_1, \dots, v_n\}$ of V . Define a function $f : B \rightarrow W$ by

$$f : v_i \mapsto \begin{cases} w_i & \text{for } 1 \leq i \leq t \\ w_t & \text{for } t \leq i \leq n \end{cases}.$$

From Proposition 6.2, it follows that there exists a linear transformation $\alpha : V \rightarrow W$ satisfying $\alpha(v_i) = f(v_i)$ for all $1 \leq i \leq n$, and this is the desired epimorphism. \square

(6.10) Proposition: Let V and W be vector spaces over a field F , where V is finitely generated. Then $\dim(V) = \dim(\text{im}(\alpha)) + \dim(\ker(\alpha))$ for any linear transformation $\alpha \in \text{Hom}(V, W)$.

Proof: Let $\alpha \in \text{Hom}(V, W)$. Set $V_1 = \ker(\alpha)$ and let V_2 be a complement of V_1 in V . By Proposition 5.16, we see that $\dim(V) = \dim(V_1) + \dim(V_2)$ and so it suffices for us to show that $V_2 \cong \text{im}(\alpha)$. Let α_2 be the restriction of α to V_2 . Then $\alpha_2 \in \text{Hom}(V_2, \text{im}(\alpha))$. If $v_2 \in \ker(\alpha_2)$ then $v_2 \in V_2 \cap V_1 = \{0_V\}$. Thus α_2 is a monomorphism. If $w \in \text{im}(\alpha)$ then there exists an element v of V satisfying $\alpha(v) = w$. Moreover, $v = v_1 + v_2$ for some $v_1 \in V_1$ and $v_2 \in V_2$ so $w = \alpha(v) = \alpha(v_1) + \alpha(v_2) = 0_W + \alpha(v_2) = \alpha_2(v_2) = \alpha_2(v_2)$. Therefore $\text{im}(\alpha_2) = \text{im}(\alpha)$, showing that α_2 is also an epimorphism and hence the desired isomorphism. \square

Let V and W be vector spaces over a field F . If $\alpha \in \text{Hom}(V, W)$ then we define the **rank** $\text{rk}(\alpha)$ of α to be $\dim(\text{im}(\alpha))$ and define the **nullity** $\text{null}(\alpha)$ of α to be $\dim(\ker(\alpha))$. Thus, Proposition 6.10 says that V has finite dimension n then both the rank and nullity of α are finite and their sum is n . The converse is also clearly true: if the rank and nullity of α are both finite, then the dimension of V is finite. Let us give bounds on the rank and nullity of compositions of linear transformations.

(6.11) Proposition (Sylvester's Theorem): Let V, W , and Y be vector spaces finitely-generated over a field F and let $\alpha : V \rightarrow W$ and $\beta : W \rightarrow Y$ be linear transformations. Then

- (1) $\text{null}(\beta\alpha) \leq \text{null}(\alpha) + \text{null}(\beta)$;
- (2) $\text{rk}(\alpha) + \text{rk}(\beta) - \dim(W) \leq \text{rk}(\beta\alpha) \leq \min\{\text{rk}(\alpha), \text{rk}(\beta)\}$.

Proof: (1) Let β_1 be the restriction of β to $\text{im}(\alpha)$. Then $\ker(\beta_1)$ is a subspace of $\ker(\beta)$. By Proposition 6.10, we have

$$\begin{aligned} \text{null}(\beta\alpha) &= \dim(V) - \text{rk}(\beta\alpha) = [\dim(V) - \text{rk}(\alpha)] + [\text{rk}(\alpha) - \text{rk}(\beta\alpha)] \\ &= \text{null}(\alpha) + \text{null}(\beta_1) \leq \text{null}(\alpha) + \text{null}(\beta). \end{aligned}$$

(2) Clearly $\text{im}(\beta\alpha)$ is a subspace of $\text{im}(\beta)$ and so its dimension is no greater than that of $\text{im}(\beta)$. Moreover, $\text{im}(\beta\alpha) = \text{im}(\beta_1)$ and so $\text{rk}(\beta\alpha) \leq \text{rk}(\beta)$. Thus $\text{rk}(\beta\alpha) \leq \min\{\text{rk}(\alpha), \text{rk}(\beta)\}$. Moreover, from (1) we see that

$$\begin{aligned} \dim(V) - \text{null}(\beta\alpha) &\geq \dim(V) - \text{null}(\alpha) + \dim(W) - \text{null}(\beta) - \dim(W) \\ &= \text{rk}(\alpha) + \text{rk}(\beta) - \dim(W) \end{aligned}$$

and this proves that $\text{rk}(\alpha) + \text{rk}(\beta) - \dim(W) \leq \text{rk}(\beta\alpha)$. \square

Exercises

Exercise 197 Which of the following statements are true for all vector spaces V and W over a field F and all $\alpha \in \text{Hom}(V, W)$?

- (1) $\alpha(A \cup B) = \alpha(A) \cup \alpha(B)$ for all nonempty subsets A and B of V ;
 (2) $\alpha(A \cap B) = \alpha(A) \cap \alpha(B)$ for all nonempty subsets A and B of V ;
 (3) $\alpha^{-1}(C \cup D) = \alpha^{-1}(C) \cup \alpha^{-1}(D)$ for all nonempty subsets C and D of W ;
 (4) $\alpha^{-1}(C \cap D) = \alpha^{-1}(C) \cap \alpha^{-1}(D)$ for all nonempty subsets C and D of W .

Exercise 198 Let $\alpha : \mathbb{R}^3 \rightarrow \mathbb{R}^3$ be a linear transformation satisfying

$$\alpha \left(\begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} \right) = \begin{bmatrix} -1 \\ 3 \\ 4 \end{bmatrix}, \quad \alpha \left(\begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix} \right) = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix},$$

and $\alpha \left(\begin{bmatrix} 1 \\ 2 \\ -1 \end{bmatrix} \right) = \begin{bmatrix} 3 \\ 1 \\ 4 \end{bmatrix}$. What is $\alpha \left(\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \right)$?

Exercise 199 Let $\alpha : \mathbb{R}^3 \rightarrow \mathbb{R}^3$ be a linear transformation satisfying

$$\alpha \left(\begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} \right) = \begin{bmatrix} 1 \\ 2 \\ -1 \end{bmatrix}, \quad \alpha \left(\begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix} \right) = \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix}, \quad \text{and} \quad \alpha \left(\begin{bmatrix} 0 \\ -1 \\ 1 \end{bmatrix} \right) = \begin{bmatrix} 3 \\ 3 \\ 3 \end{bmatrix}.$$

Find a vector $v \in \mathbb{R}^3$ for which $\alpha(v) = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$.

Exercise 200 Let F be a field and let V be the subspace of $F[X]$ consisting of all polynomials of degree at most 2. Let $\alpha : V \rightarrow F[X]$ be a linear transformation satisfying $\alpha(1) = X$, $\alpha(X + 1) = X^5 + X^3$, and $\alpha(X^2 + X + 1) = X^4 - X^2 + 1$. What is $\alpha(X^2 - X)$?

Exercise 201 For each $d \in \mathbb{R}$, let $\alpha_d : \mathbb{R}^2 \rightarrow \mathbb{R}^2$ be the function defined

by $\alpha_d : \begin{bmatrix} a \\ b \end{bmatrix} \mapsto \begin{bmatrix} a + b + d^2 + 1 \\ a \end{bmatrix}$. Is there a number d having the property that α_d is a linear transformation?

Exercise 202 For each $d \in \mathbb{R}$, let $\alpha_d : \mathbb{R}^2 \rightarrow \mathbb{R}^2$ be the function defined

by $\alpha_d : \begin{bmatrix} a \\ b \end{bmatrix} \mapsto \begin{bmatrix} 5da - db \\ 8d^2 - 8d - 6 \end{bmatrix}$. Is there a number d having the property that α_d is a linear transformation?

Exercise 203 Let W and W' be subspaces of a vector space V over a field F and assume that we have linear transformations $\alpha : W \rightarrow V$ and $\beta : W' \rightarrow V$ satisfying the condition that $\alpha(v) = \beta(v)$ for all $v \in W \cap W'$. Find a linear transformation $\theta : W + W' \rightarrow V$, the restriction of which to W equals α and the restriction of which to W' equals β , or show why no such linear transformation exists.

Exercise 204 Let $F = GF(3)$ and let $\theta \in F^F$ be the function defined by $\theta(0) = 0$, $\theta(1) = 2$, and $\theta(2) = 1$. Let n be a positive integer and

let $\alpha : F^n \rightarrow F^n$ be the function defined by $\alpha : \begin{bmatrix} a_1 \\ \vdots \\ a_n \end{bmatrix} \mapsto \begin{bmatrix} \theta(a_1) \\ \vdots \\ \theta(a_n) \end{bmatrix}$ Is α a linear transformation?

Exercise 205 Does there exist a linear transformation $\alpha : \mathbb{Q}^4 \rightarrow \mathbb{Q}[X]$

satisfying $\alpha \left(\begin{bmatrix} 1 \\ 2 \\ 0 \\ -1 \end{bmatrix} \right) = 2$, $\alpha \left(\begin{bmatrix} -1 \\ 1 \\ 1 \\ 1 \end{bmatrix} \right) = X$, and $\alpha \left(\begin{bmatrix} -1 \\ 4 \\ 2 \\ 1 \end{bmatrix} \right) = X + 1$?

Exercise 206 Let B be a Hamel basis for \mathbb{R} as a vector space over \mathbb{Q} and let $1 \neq a \in \mathbb{R}$. Show that there exists an element $y \in B$ satisfying $ay \notin B$.

Exercise 207 For which nonnegative integers h is the function α from $GF(3)^3$ to itself defined by $\alpha : \begin{bmatrix} a \\ b \\ c \end{bmatrix} \mapsto \begin{bmatrix} a^h \\ b \\ c^h \end{bmatrix}$ a linear transformation?

Exercise 208 For any field F , let $\theta : F \rightarrow F$ be the function defined by

$$\theta : a \mapsto \begin{cases} 0 & \text{if } a = 0 \\ a^{-1} & \text{otherwise} \end{cases}.$$

This is clearly a linear transformation when $F = GF(2)$. Does there exist a field other than $GF(2)$ for which θ is a linear transformation?

Exercise 209 Let $V = F^\infty$ and let $\alpha : V \rightarrow V$ be the function that assigns to each sequence $[a_1, a_2, \dots] \in V$ its sequence of partial sums, namely $[a_1, a_2, \dots] \mapsto [a_1, \sum_{i=1}^2 a_i, \sum_{i=1}^3 a_i, \dots]$. Is α a linear transformation?

Exercise 210 Let $Y = \mathbb{R}^{\mathbb{R}} \times \mathbb{R}$. Is the function $\alpha : Y \rightarrow \mathbb{R}$ defined by $\alpha : (f, a) \mapsto f(a)$ a linear transformation?

Exercise 211 Let $\alpha : \mathbb{R} \rightarrow \mathbb{R}$ be a continuous function satisfying

$$\alpha(a+b) = \alpha(a) + \alpha(b)$$

for all $a, b \in \mathbb{R}$. Show that α is a linear transformation.

Exercise 212 Let F be a field and let b and c be nonzero elements of F . Let $\alpha : F^\infty \rightarrow F^\infty$ be the linear transformation defined by

$$\alpha : [a_1, a_2, \dots] \mapsto [a_3 + ba_2 + ca_1, a_4 + ba_3 + ca_2, \dots].$$

Let $y \in \ker(\alpha)$ be a vector satisfying the condition that two successive entries in y equal 0. Show that $y = [0, 0, \dots]$.

Exercise 213 Consider the field $F = \mathbb{Q}(\sqrt{2})$ as a \mathbb{Q} -algebra. Show that the only homomorphisms of \mathbb{Q} -algebras from F to itself are the identity function and the function $a + b\sqrt{2} \mapsto a - b\sqrt{2}$.

Exercise 214 Let V , W , and Y be vector spaces finitely-generated over a field F and let $\alpha : V \rightarrow W$ be a linear transformation. Show that the set of all linear transformations $\beta : W \rightarrow Y$ satisfying the condition that $\beta\alpha$ is the 0-transformation is a subspace of $\text{Hom}(W, Y)$, and calculate its dimension.

Exercise 215 Let V and W be vector spaces over a field F and let V' be a proper subspace of V . Are $\{\alpha \in \text{Hom}(V, W) \mid \ker(\alpha) \subseteq V'\}$ and $\{\alpha \in \text{Hom}(V, W) \mid \ker(\alpha) \supseteq V'\}$ subspaces of $\text{Hom}(V, W)$?

Exercise 216 Let V and W be vector spaces over a field F and assume that there are subspaces V_1 and V_2 of V , both of positive dimension, satisfying $V = V_1 \oplus V_2$. For $i = 1, 2$, let $U_i = \{\alpha \in \text{Hom}(V, W) \mid \ker(\alpha) \supseteq V_i\}$. Show that $\{U_1, U_2\}$ is an independent set of subspaces of $\text{Hom}(V, W)$. Is it necessarily true that $\text{Hom}(V, W) = U_1 \oplus U_2$?

Exercise 217 Let F be a field, and let $\alpha : \mathcal{M}_{2 \times 2}(F) \rightarrow \mathcal{M}_{2 \times 2}(F)$ be a linear transformation satisfying $\alpha(AB) = \alpha(A)\alpha(B)$ for all $A, B \in$

$\mathcal{M}_{2 \times 2}(F)$. Show that $\alpha\left(\begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}\right) \neq I$.

Exercise 218 Let V and W be vector spaces over a field F and let $\alpha, \beta : V \rightarrow W$ be linear transformations satisfying the condition that for each $v \in V$ there exists a scalar $c_v \in F$ (depending on v) satisfying $\beta(v) = c_v\alpha(v)$. Show that there exists a scalar c satisfying $\beta = c\alpha$.

Exercise 219 Find the kernel of the linear transformation $\alpha : \mathbb{R}^5 \rightarrow \mathbb{R}^3$

defined by $\alpha : \begin{bmatrix} a \\ b \\ c \\ d \\ e \end{bmatrix} \mapsto \begin{bmatrix} b + c - 2d + e \\ a + 2b + 3c - 4d \\ 2a + 2c - 2e \end{bmatrix}$.

Exercise 220 Let $\alpha : \mathbb{R}^3 \rightarrow \mathbb{R}^3$ be the linear transformation defined by

$$\alpha : \begin{bmatrix} a \\ b \\ c \end{bmatrix} \mapsto \begin{bmatrix} 2a + 4b - c \\ 0 \\ 3c + 2b - a \end{bmatrix}. \text{ Are } \text{im}(\alpha) \text{ and } \ker(\alpha) \text{ disjoint?}$$

Exercise 221 Let W be the subspace $\mathbb{Q} \left\{ \begin{bmatrix} 2 \\ -1 \\ 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ 0 \\ 1 \end{bmatrix} \right\}$ of

\mathbb{Q}^4 and let $\alpha : W \rightarrow \mathbb{Q}^2$ be the linear transformation defined by setting

$$\alpha : \begin{bmatrix} a \\ b \\ c \\ d \end{bmatrix} \mapsto \begin{bmatrix} a + 2b + c \\ -a - 2b - c \end{bmatrix}. \text{ Find a basis for } \ker(\alpha).$$

Exercise 222 Let $F = GF(3)$ and let $\alpha : F^3 \rightarrow F^3$ be the linear

transformation defined by $\alpha : \begin{bmatrix} a \\ b \\ c \end{bmatrix} \mapsto \begin{bmatrix} a + b \\ 2b + c \\ 0 \end{bmatrix}$. Find the kernel of α .

Exercise 223 Let $\alpha : \mathbb{R}^4 \rightarrow \mathbb{R}^3$ be the linear transformation defined by

$$\alpha : \begin{bmatrix} a \\ b \\ c \\ d \end{bmatrix} \mapsto \begin{bmatrix} 2a + 4b + c - d \\ 3a + b - 2c \\ a + 5c + 4d \end{bmatrix}. \text{ Do there exist } a, b, d \in \mathbb{Z} \text{ such that}$$

$$\begin{bmatrix} a \\ b \\ 7 \\ d \end{bmatrix} \in \ker(\alpha)?$$

Exercise 224 Let V and W be vector spaces over a field F . Let $\alpha \in \text{Hom}(V, W)$ and $\beta \in \text{Hom}(W, V)$ satisfy the condition that $\alpha\beta\alpha = \alpha$. If $w \in \text{im}(\alpha)$, show that $\alpha^{-1}(w) = \{\beta(w) + v - \beta\alpha(v) \mid v \in V\}$.

Exercise 225 Let V , W , and Y be vector spaces over a field F and let $\alpha \in \text{Hom}(V, W)$ and $\beta \in \text{Hom}(W, Y)$ satisfy the condition that $\text{im}(\alpha)$ has a finitely-generated complement in W and $\text{im}(\beta)$ has a finitely-generated complement in Y . Does $\text{im}(\beta\alpha)$ necessarily have a finitely-generated complement in Y ?

Exercise 226 Let $\alpha : \mathcal{M}_{3 \times 3}(\mathbb{R}) \rightarrow \mathbb{R}$ be the function defined by $\alpha : [a_{ij}] \mapsto \sum_{i=1}^3 \sum_{j=1}^3 a_{ij}$. Show that α is a linear transformation and find a basis for $\ker(\alpha)$.

Exercise 227 Let $F = GF(2)$ and let $n > 2$ be an integer. Let W be

the set of all vectors $\begin{bmatrix} a_1 \\ \vdots \\ a_n \end{bmatrix}$ in F^n having an even number of nonzero

entries. Show that W is a subspace of F^n by showing that it is the kernel of some linear transformation.

Exercise 228 Let A and B be nonempty sets. Let V be the collection of all subsets of A and let W be the collection of all subsets of B , both of which, as we have seen, are vector spaces over $GF(2)$. Any function $f : A \rightarrow B$ defines a function $\alpha_f : W \rightarrow V$ by setting

$$\alpha_f : D \mapsto \{a \in A \mid f(a) \in D\}.$$

Show that each such function α_f is a linear transformation, and find its kernel.

Exercise 229 Let V be a vector space over a field F and let $\alpha : V^3 \rightarrow V$

be the function defined by $\alpha : \begin{bmatrix} v_1 \\ v_2 \\ v_3 \end{bmatrix} \mapsto v_1 + v_2 + v_3$. Show that α is a

linear transformation and find its kernel.

Exercise 230 Let n be a positive integer and let V be the subspace of $\mathbb{R}[X]$ composed of all polynomials of degree at most n . Let $\alpha : V \rightarrow V$ be the linear transformation given by $\alpha : p(X) \mapsto p(X+1) - p(X)$. Find $\ker(\alpha)$ and $\text{im}(\alpha)$.

Exercise 231 Let $\alpha : \mathbb{R}^3 \rightarrow \mathbb{R}^3$ be the linear transformation given by

$$\alpha : \begin{bmatrix} a \\ b \\ c \end{bmatrix} \mapsto \begin{bmatrix} a+b+c \\ -a-c \\ b \end{bmatrix}. \text{ Find } \ker(\alpha) \text{ and } \text{im}(\alpha).$$

Exercise 232 Find the kernel of the linear transformation $\alpha : \mathbb{Q}[X] \rightarrow \mathbb{R}$ defined by $\alpha : p(X) \mapsto p(\sqrt{3})$.

Exercise 233 Let $V = C(0,1)$. For each positive integer n , we define the n th **Bernstein function** $\beta_n : V \mapsto \mathbb{R}[X]$ by

$$\beta_n : f \mapsto \sum_{k=0}^n \frac{n!}{k!(n-k)!} f\left(\frac{k}{n}\right) X^k (1-X)^{n-k}.$$

Show that each β_n is a linear transformation and find $\bigcap_{n=1}^{\infty} \ker(\beta_n)$. (Note: the Bernstein functions are used in building polynomial approximations to continuous functions.)

Exercise 234 Let V and W be vector spaces over a field F , where $\dim(V)$ is a positive integer. Show that $W = \sum \{\text{im}(\alpha) \mid \alpha \in \text{Hom}(V, W)\}$.

Exercise 235 Let W be the subspace of $\mathbb{R}^{\mathbb{R}}$ consisting of all twice-differentiable functions and let $\alpha : W \rightarrow \mathbb{R}^{\mathbb{R}}$ be the linear transformation $\alpha : f \mapsto f''$. Find $\alpha^{-1}(f_0)$, where $f_0 \in \mathbb{R}^{\mathbb{R}}$ is defined by $f_0 : x \mapsto x + 1$.

Exercise 236 Let W be the subspace of $\mathbb{R}^{\mathbb{R}}$ consisting of all differentiable functions and let $\alpha : W \rightarrow \mathbb{R}^{\mathbb{R}}$ be the function defined by $\alpha(f) : x \mapsto f'(x) + \cos(x)f(x)$. Show that α is a linear transformation and find its kernel.

Exercise 237 Let n be a positive integer and let V be a vector space over \mathbb{C} . Does there exist a linear transformation $\alpha : V \rightarrow \mathbb{C}^n$ other than the 0-function satisfying the condition that $\text{im}(\alpha) \subseteq \mathbb{R}^n$?

Exercise 238 Let V and W be vector spaces over a field F and let $\alpha : V \rightarrow W$ be a linear transformation other than the 0-function. Find a linear transformation $\beta : V \rightarrow W$ satisfying $\text{im}(\alpha) = \text{im}(\beta) \neq \text{im}(\alpha + \beta)$.

Exercise 239 Let V be a finite-dimensional vector space over a field F and let $\alpha, \beta \in \text{Hom}(V, V)$ be linear transformations satisfying $\text{im}(\alpha) + \text{im}(\beta) = V = \ker(\alpha) + \ker(\beta)$. Show that $\text{im}(\alpha) \cap \text{im}(\beta) = \{0_V\} = \ker(\alpha) \cap \ker(\beta)$.

Exercise 240 Let V , W , and Y be vector spaces over a field F and let $\alpha \in \text{Hom}(V, W)$ and $\beta \in \text{Hom}(W, Y)$ satisfy the condition that $\ker(\alpha)$ and $\ker(\beta)$ are both finitely generated. Is $\ker(\beta\alpha)$ necessarily finitely generated?

Exercise 241 Find a linear transformation $\alpha : \mathbb{Q}^3 \rightarrow \mathbb{Q}^4$ satisfying

$$\text{im}(\alpha) = \mathbb{Q} \left\{ \begin{bmatrix} \frac{1}{2} \\ -1 \\ 3 \\ 0 \end{bmatrix}, \begin{bmatrix} 2 \\ 1 \\ 1 \\ -4 \end{bmatrix} \right\}.$$

Exercise 242 Let $F = GF(2)$ and let $\alpha \in \text{Hom}(F^7, F^3)$ be given by

$$\begin{bmatrix} a_1 \\ \vdots \\ a_7 \end{bmatrix} \mapsto \begin{bmatrix} a_4 + a_5 + a_6 + a_7 \\ a_2 + a_3 + a_6 + a_7 \\ a_1 + a_3 + a_5 + a_7 \end{bmatrix}. \quad \text{If } v \text{ is a nonzero element of } \ker(\alpha),$$

show that at least three entries in v are equal to 1.

Exercise 243 Let V and W vector spaces finite dimensional over a field F and let $\alpha \in \text{Hom}(V, W)$. If Y is a subspace of W , is it true that $\dim(\alpha^{-1}(Y)) \geq \dim(V) - \dim(W) + \dim(Y)$?

Exercise 244 Let V be a vector space over a field F and let $Y = V^\infty$. Let W be the subspace of Y consisting of all those sequences $[v_1, v_2, \dots]$ in which $v_i = 0$ for all odd i and let W' be the subspace of Y consisting of all those sequences in which $v_i = 0$ for all even i . Find a linear transformation from Y to itself, the kernel of which equals W and the image of which equals W' .

Exercise 245 Let W be the subspace of \mathbb{R}^6 composed of all vectors

$$\begin{bmatrix} a_1 \\ \vdots \\ a_6 \end{bmatrix} \text{ satisfying } \sum_{i=1}^6 a_i = 0. \text{ Does there exist a monomorphism from}$$

W to \mathbb{R}^4 ?

Exercise 246 Let n be a positive integer and let $\alpha : \mathbb{Q}^n \rightarrow \mathbb{Q}^n$ be a linear transformation which is not a monomorphism. Does there necessarily exist a nonzero element of $\ker(\alpha)$ all the entries of which are integers?

Exercise 247 Let n be a positive integer and let W be the subspace of $\mathbb{C}[X]$ consisting of all polynomials of degree less than n . Let a_1, \dots, a_n be distinct complex numbers and let $\alpha : W \rightarrow \mathbb{C}^n$ be the function defined by

$$\alpha : p(X) \mapsto \begin{bmatrix} p(a_1) \\ \vdots \\ p(a_n) \end{bmatrix}. \text{ Is } \alpha \text{ a monomorphism? Is it an isomorphism?}$$

Exercise 248 Let V be a vector space over a field F and let $\alpha : V \rightarrow V$ be a linear transformation satisfying the condition that $\alpha^2 = a\alpha + b\sigma_1$, where a and b are nonzero scalars. Show that α is a monomorphism.

Exercise 249 Let p be a prime integer and let F be a field of characteristic p . Let (K, \bullet) be an associative and commutative unital F -algebra and let $\alpha : K \rightarrow K$ be the function defined by $\alpha : v \mapsto v^p$. Show that α is an isomorphism of unital F -algebras.

Exercise 250 Let F be a field and let K and K' be fields containing F . Show that every homomorphism of F -algebras $K \rightarrow K'$ is a homomorphism of unital F -algebras.

Exercise 251 Let $F = GF(7)$. How many distinct monomorphisms can one define from F^2 to F^4 ?

Exercise 252 Let V and W be vector spaces over a field F and let $\alpha, \beta \in \text{Hom}(V, W)$ be monomorphisms. Is $\alpha + \beta$ necessarily a monomorphism?

Exercise 253 Let F be a field and let F' be a field containing F . Let (K, \bullet) be an F -algebra and let $\alpha : F' \rightarrow K$ be a nontrivial homomorphism of F -algebras. Show that α is monic.

Exercise 254 Let V and W be vector spaces over a field F and let $\alpha \in \text{Hom}(V, W)$ be an epimorphism. Show that there exists a linear transformation $\beta \in \text{Hom}(W, V)$ satisfying the condition that $\alpha\beta$ is the identity function on W .

Exercise 255 Let V be a vector space finite dimensional over a field F , the dimension n of which is even. Show that there exists an isomorphism $\alpha : V \rightarrow V$ satisfying the condition that $\alpha^2(v) = -v$ for all $v \in V$.

Exercise 256 Let $\alpha : V \rightarrow W$ be a linear transformation between vector spaces over a field F and let D be a nonempty linearly-independent subset of $\text{im}(\alpha)$. Show that there exists a basis B of V satisfying the condition that $\{\alpha(v) \mid v \in B\} = D$.

Exercise 257 Let V and W be vector spaces over a field F and let $\alpha \in \text{Hom}(V, W)$ satisfy the condition that $\alpha\beta\alpha$ is not the 0-function for any linear transformation $\beta : W \rightarrow V$ which is not the 0-function. Show that α is an isomorphism.

Exercise 258 Let F be a field and let $\alpha : F^3 \rightarrow F[X]$ be the linear transformation defined by
$$\begin{bmatrix} a \\ b \\ c \end{bmatrix} \mapsto (a+b)X + (a+c)X^5.$$
 Find the nullity and rank of α .

Exercise 259 Let F be a field and let $p(X) = X^2 + bX + c \in F[X]$ be a polynomial having distinct nonzero roots d_1 and d_2 in F . Let $\alpha : F^3 \rightarrow F$ be the linear transformation defined by

$$\begin{bmatrix} a_1 \\ a_2 \\ a_3 \end{bmatrix} \mapsto a_3 + ba_2 + ca_1$$

and let $\beta : F^\infty \rightarrow F^\infty$ be the linear transformation defined by

$$\beta : [a_1, a_2, a_3, \dots] \mapsto [\alpha(a_1, a_2, a_3), \alpha(a_2, a_3, a_4), \dots].$$

Show that the nullity of β is at least 2.

Exercise 260 Let Ω be a nonzero set and let V be the collection of all subsets of Ω , considered as a vector space over $GF(2)$. Show that this vector space is isomorphic to $GF(2)^\Omega$.

Exercise 261 Let F be a field and let V be the subspace of F^∞ consisting of all sequences $[a_1, a_2, a_3, \dots]$ in which $a_i = 0$ for all even i . Let W be the subspace of F^∞ consisting of all sequences $[a_1, a_2, a_3, \dots]$ in which $a_i = 0$ for all odd i . Show that $V \cong F^\infty \cong W$.

Exercise 262 Let V be a vector space over a field F having subspaces W and W' . Let $Y = \left\{ \begin{bmatrix} w \\ w' \end{bmatrix} \mid w \in W \text{ and } w' \in W' \right\}$, which is a subspace of V^2 . Let $\alpha : Y \rightarrow V$ be the linear transformation defined by $\alpha : \begin{bmatrix} w \\ w' \end{bmatrix} \mapsto w + w'$. Find the kernel of α , and show that it is isomorphic to $W \cap W'$.

Exercise 263 Let V be a vector space over a field F . Let W be a subspace of V and let W' be a complement of W in V . Let $\alpha : W \rightarrow W'$ be a linear transformation. Show that W is isomorphic to the subspace $Y = \{w + \alpha(w) \mid w \in W\}$ of V .

Exercise 264 Show that there is no vector space over any field F having precisely 15 elements.

Exercise 265 Let F be a field and let $V = F[X]$. Show that $V \cong V^2$.

Exercise 266 Let V , W , and Y be vector spaces over a field F . Let $\{\alpha_1, \dots, \alpha_n\}$ be a finite subset of $\text{Hom}(V, W)$ and let $\beta \in \text{Hom}(V, Y)$ be a linear transformation satisfying $\bigcap_{i=1}^n \ker(\alpha_i) \subseteq \ker(\beta)$. Show that there exist linear transformations $\gamma_1, \dots, \gamma_n$ in $\text{Hom}(W, Y)$ satisfying $\beta = \sum_{i=1}^n \gamma_i \alpha_i$.

Exercise 267 Let V be a vector space over a field F and let W be a subspace of V . For each $v \in V$, let $v + W = \{v + w \mid w \in W\}$. Let V/W be the collection of all the sets of the form $v + W$ for $v \in V$ and define operations of addition and scalar multiplication on V/W by setting $(v + W) + (v' + W) = (v + v') + W$ and $c(v + W) = (cv) + W$ for all $v, v' \in V$ and $c \in F$. Show that:

- (1) $v + W = v' + W$ if and only if $v - v' \in W$;
 - (2) V/W , with the given operations, is a vector space over F ;
 - (3) The function $v \mapsto v + W$ is an epimorphism from V to W , the kernel of which equals W ;
 - (4) Every complement of W in V is isomorphic to V/W .
- The space V/W is called the **factor space** of V by W .

Exercise 268 Let F be a field and let $m > n$ be positive integers. Let A and B be fixed matrices in $\mathcal{M}_{n \times m}(F)$ and let $\theta : \mathcal{M}_{n \times m}(F) \rightarrow \mathcal{M}_{n \times m}(F)$ be the linear transformation defined by $\theta : C \mapsto ACB$. Show that θ is not an isomorphism.

7

The endomorphism algebra of a vector space

Let V be a vector space over a field F . A linear transformation α from V to itself is called an **endomorphism** of V . We will denote the set of all endomorphisms of V by $End(V)$. This set is nonempty, since it includes the functions of the form $\sigma_c : v \mapsto cv$ for $c \in F$. In particular, it includes the 0-endomorphism $\sigma_0 : v \mapsto 0_V$ and the identity endomorphism $\sigma_1 : v \mapsto v$. If V is nontrivial, these functions are not the same. We see that we have two operations defined on $End(V)$: addition and multiplication (given by composition). Indeed, as a direct consequence of the definitions we conclude the following:

(7.1) Proposition: **If V is a nontrivial vector space over a field F , then $End(V)$ is an associative unital F -algebra with σ_0 being the identity element for addition and σ_1 being the identity element for multiplication.**

If V is a nontrivial vector space over a field F then there exists a function $\sigma : F \rightarrow End(V)$ defined by $\sigma : c \mapsto \sigma_c$ for all $c \in F$. This function is monic, for if $\sigma_c = \sigma_d$ then for any $0_V \neq v \in V$ we have $cv = \sigma_c(v) = \sigma_d(v) = dv$ and hence $(c - d)v = 0_V$. Since $0_V \neq v$, this implies that $c - d = 0$ and so $c = d$. Moreover, if $c, d \in F$ then $\sigma_c + \sigma_d = \sigma_{c+d}$ and $\sigma_c \sigma_d = \sigma_{cd}$ so σ is a monic homomorphism of unital F -algebras. We can use this function to identify F with its image under σ and consider it a subalgebra of the F -algebra $End(V)$.

If $\alpha, \beta \in \text{End}(V)$ and if $c \in F$, then we have already seen that the functions $\alpha + \beta$, $\alpha\beta$, and $c\alpha$ all belong to $\text{End}(V)$. Therefore we see that if $p(X) = \sum_{i=0}^n a_i X^i \in F[X]$ then $p(\alpha) = \sum_{i=0}^n a_i \alpha^i$ is an endomorphism of V , and, indeed, the set $F[\alpha]$ of all endomorphisms of V of this form is an F -subalgebra of $\text{End}(V)$. The function from $F[X]$ to $F[\alpha]$ given by $p(X) \mapsto p(\alpha)$ is immediately seen to be an epic homomorphism of unital F -algebras for any $\alpha \in \text{End}(V)$.

Example: Let $F = GF(2)$ and let $p(X) = X^2 + X \in F[X]$. Then $p(a) = 0$ for every $a \in F$. However, $p(\alpha) \neq \sigma_0$, where $\alpha \in \text{End}(F^2)$ is defined by $\alpha : \begin{bmatrix} a \\ b \end{bmatrix} \mapsto \begin{bmatrix} b \\ a \end{bmatrix}$.

Example: Structures of the form $F[\alpha]$ are important in many areas of mathematics. For example, let V be the collection of all infinitely-differentiable functions from \mathbb{R} to \mathbb{R} and let δ be the differentiation endomorphism on V . If $p(X) = \sum_{i=0}^n a_i X^i \in \mathbb{R}[X]$, then we have $p(\delta) : f \mapsto a_0 f + \sum_{i=1}^n a_i f^{[i]}$, where $f^{[i]}$ denotes the i th derivative of f . Such an endomorphism is called a **differential operator with constant coefficients** on V . If $c \in \mathbb{R}$ and if $f \in V$ is the function given by $f_c : x \mapsto e^{cx}$, then $\delta(f_c) = cf_c$ and so $p(\delta) : f_c \mapsto \sum_{i=0}^n a_i c^i e^{cx} = (\sum_{i=0}^n a_i c^i) f_c = p(c) f_c$. Thus, $p(\delta)$ is the 0-function whenever c is a root of $p(X)$. Hence $f_c \in \ker(p(\delta))$ for each root c of $p(X)$.

Example: Let F be a field and let (K, \bullet) be a nonassociative F -algebra. An endomorphism $\delta \in \text{End}(K)$ is a **derivation** if and only if $\delta(v \bullet w) = [\delta(v)] \bullet w + v \bullet [\delta(w)]$. Thus, for example, if K is a Lie algebra then, as a consequence of the Jacobi identity, we see that every $y \in K$ defines a derivation δ_y of K given by $\delta_y : v \mapsto y \bullet v$. Also, if K is the \mathbb{R} -algebra consisting of all infinitely-differentiable functions in $\mathbb{R}^{\mathbb{R}}$, then the endomorphism of K which assigns to each function in K its derivative is a derivation. The set of all derivations defined on K is a subspace of $\text{End}(K)$. If δ and δ' are derivations on K , then $\delta\delta'$ is not, in general, a derivation on K , but the Lie product $\delta\delta' - \delta'\delta$ is always a derivation on K , and so the set of all derivations on K is a Lie algebra over F .

Given a nontrivial vector space V over a field F , we note that the F -algebra $\text{End}(V)$ is neither necessarily commutative nor necessarily entire, as the following examples show:

Example: Let F be a field and let $V = F^3$. Let $\alpha, \beta \in \text{End}(V)$ be the endomorphisms defined by $\alpha : \begin{bmatrix} a \\ b \\ c \end{bmatrix} \mapsto \begin{bmatrix} b \\ a \\ c \end{bmatrix}$ and $\beta : \begin{bmatrix} a \\ b \\ c \end{bmatrix} \mapsto$

$\begin{bmatrix} a \\ 0 \\ 0 \end{bmatrix}$. Then $\beta\alpha : \begin{bmatrix} a \\ b \\ c \end{bmatrix} \mapsto \begin{bmatrix} b \\ 0 \\ 0 \end{bmatrix}$ and $\alpha\beta : \begin{bmatrix} a \\ b \\ c \end{bmatrix} \mapsto \begin{bmatrix} 0 \\ a \\ 0 \end{bmatrix}$, so $\beta\alpha \neq \alpha\beta$.

Example: Let F be a field and let $V = F^3$. Let $\alpha, \beta \in \text{End}(V)$ be the endomorphisms defined by $\alpha : \begin{bmatrix} a \\ b \\ c \end{bmatrix} \mapsto \begin{bmatrix} 0 \\ 0 \\ c \end{bmatrix}$ and $\beta : \begin{bmatrix} a \\ b \\ c \end{bmatrix} \mapsto \begin{bmatrix} a \\ 0 \\ 0 \end{bmatrix}$. Then $\beta\alpha = \sigma_0 = \alpha\beta$.

We do, however, have the following:

(7.2) Proposition: Let V be a nontrivial vector space over a field F . Then for all $\alpha \in \text{End}(V)$ and all $c \in F$ we have $\alpha\sigma_c = \sigma_c\alpha$.

Proof: If $v \in V$ then $\alpha\sigma_c(v) = \alpha(cv) = c\alpha(v) = \sigma_c\alpha(v)$. □

An endomorphism of a vector space V over a field F which is also an isomorphism (i.e., which is both monic and epic), is called an **automorphism** of V . Since $\alpha(0_V) = 0_V$ for any endomorphism α of V , we see that any automorphism of V induces a permutation of $V \setminus \{0_V\}$. Similarly, a homomorphism of F -algebras which is also an isomorphism is an **automorphism of F -algebras**.

By what we have already seen, we know that $\alpha \in \text{End}(V)$ is an automorphism if and only if there exists an endomorphism $\alpha^{-1} \in \text{End}(V)$ satisfying $\alpha\alpha^{-1} = \sigma_1 = \alpha^{-1}\alpha$. We will denote the set of all automorphisms of V by $\text{Aut}(V)$. This set is nonempty, since $\sigma_1 \in \text{Aut}(V)$, where $\sigma_1^{-1} = \sigma_1$. Moreover, if $\alpha, \beta \in \text{Aut}(V)$ then $(\alpha\beta)(\beta^{-1}\alpha^{-1}) = \alpha(\beta\beta^{-1})\alpha^{-1} = \alpha\alpha^{-1} = \sigma_1$ and similarly $(\beta^{-1}\alpha^{-1})(\alpha\beta) = \sigma_1$. Thus $\alpha\beta \in \text{Aut}(V)$, with $(\alpha\beta)^{-1} = \beta^{-1}\alpha^{-1}$. It is also clear that if $\alpha \in \text{Aut}(V)$ then $\alpha^{-1} \in \text{Aut}(V)$.

Example: Let V be a vector space over a field F and let $n > 1$ be an integer. Any permutation π of the set $\{1, \dots, n\}$ defines an

automorphism α_π of V^n given by $\alpha_\pi : \begin{bmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{bmatrix} \mapsto \begin{bmatrix} v_{\pi(1)} \\ v_{\pi(2)} \\ \vdots \\ v_{\pi(n)} \end{bmatrix}$ which

rearranges the entries of each vector according to the permutation π . More generally, if V is a vector space over a field F having a basis $B = \{v_i \mid i \in \Omega\}$ and if π is a permutation of Ω , then there is an automorphism of V defined by $\sum_{i \in \Lambda} a_i v_i \mapsto \sum_{i \in \Lambda} a_{\pi(i)} v_{\pi(i)}$ for each finite subset Λ of Ω .

Example: Let F be a field and let n be a positive integer. We have already seen that the function $A \mapsto A^T$ is an automorphism of $\mathcal{M}_{n \times n}(F)$, considered as a vector space over F .

Example: Let V be a vector space having finite dimension n over a field F and let v and y be nonzero elements of V . Then there exist bases $\{v_1, \dots, v_n\}$ and $\{y_1, \dots, y_n\}$ of V satisfying $v_1 = v$ and $y_1 = y$. The function $\alpha : V \rightarrow V$ defined by $\alpha : \sum_{i=1}^n a_i v_i \mapsto \sum_{i=1}^n a_i y_i$ is thus an automorphism of V satisfying $\alpha(v) = y$.

Let V be a vector space over a field F and let n be a positive integer. We will list several types automorphisms, called **elementary automorphisms**, of a vector space of the form V^n . These automorphisms will play an important part in our ensuing discussion.

$$(1) \text{ If } 1 \leq h \neq k \leq n, \text{ we define } \varepsilon_{hk} \in \text{Aut}(V^n) \text{ by } \begin{bmatrix} v_1 \\ \vdots \\ v_n \end{bmatrix} \mapsto \begin{bmatrix} w_1 \\ \vdots \\ w_n \end{bmatrix},$$

where

$$w_i = \begin{cases} v_k & \text{if } i = h \\ v_h & \text{if } i = k \\ v_i & \text{otherwise} \end{cases}.$$

This automorphism satisfies $\varepsilon_{hk}^{-1} = \varepsilon_{hk}$.

(2) If $1 \leq h \leq n$, and if $0 \neq c \in F$, we define $\varepsilon_{h;c} \in \text{Aut}(V^n)$ by

$$\begin{bmatrix} v_1 \\ \vdots \\ v_n \end{bmatrix} \mapsto \begin{bmatrix} w_1 \\ \vdots \\ w_n \end{bmatrix}, \text{ where}$$

$$w_i = \begin{cases} cv_i & \text{if } i = h \\ v_i & \text{otherwise} \end{cases}.$$

This automorphism satisfies $\varepsilon_{h;c}^{-1} = \varepsilon_{h;c^{-1}}$.

(3) If $1 \leq h \neq k \leq n$ and if $c \in F$, we define $\varepsilon_{hk;c} \in \text{Aut}(V^n)$ by

$$\begin{bmatrix} v_1 \\ \vdots \\ v_n \end{bmatrix} \mapsto \begin{bmatrix} w_1 \\ \vdots \\ w_n \end{bmatrix}, \text{ where}$$

$$w_i = \begin{cases} v_i + cv_k & \text{if } i = h \\ v_i & \text{otherwise} \end{cases}.$$

This automorphism satisfies $\varepsilon_{hk;c}^{-1} = \varepsilon_{hk;-c}$.

Identifying the automorphisms of a finite-dimensional vector space V over a field F is a problem which will be of major importance to us later, and so it is important to characterize these functions.

(7.3) Proposition: Let V be a vector space of finite dimension n over a field F . Then the following conditions on an endomorphism α of V are equivalent:

- (1) α is an automorphism of V ;
- (2) α is monic;
- (3) α is epic.

Proof: By definition, (1) implies (2). Now assume (2). By Proposition 6.10, we see that the rank of α equals n and so $\text{im}(\alpha) = V$ by Proposition 5.11, proving (3). Now assume (3). By Proposition 6.10, we see that the nullity of α equals $n - n = 0$ and so $\ker(\alpha) = \{0_V\}$, proving that α is monic as well, and so is bijective. This proves (1). \square

(7.4) Proposition: Let V be a finite-dimensional vector space over a field F and let $\alpha \in \text{End}(V)$. If there exists a $\beta \in \text{End}(V)$ satisfying $\alpha\beta = \sigma_1$ or $\beta\alpha = \sigma_1$, then $\alpha \in \text{Aut}(V)$ and $\beta = \alpha^{-1}$.

Proof: If $\beta\alpha = \sigma_1$ then $\ker(\alpha) \subseteq \ker(\sigma_1) = \{0_V\}$ and so, by Proposition 7.3, $\alpha \in \text{Aut}(V)$. Similarly, if $\alpha\beta = \sigma_1$ then $\text{im}(\alpha) \supseteq \text{im}(\sigma_1) = V$ and so, by Proposition 7.3, $\alpha \in \text{Aut}(V)$. Moreover, if $\alpha\beta = \sigma_1$ we see that $\alpha^{-1} = \alpha^{-1}\sigma_1 = \alpha^{-1}(\alpha\beta) = \beta$ and similarly $\alpha^{-1} = \beta$ when $\beta\alpha = \sigma_1$. \square

Example: Proposition 7.3 and Proposition 7.4 are no longer true if we remove the condition of finite dimensionality. For example, let F be a field and let $V = F[X]$. Define the endomorphisms α and β of V by setting $\alpha : \sum_{i=0}^n a_i X^i \mapsto \sum_{i=0}^n a_i X^{i+1}$ and $\beta : \sum_{i=0}^n a_i X^i \mapsto \sum_{i=1}^n a_i X^{i-1}$. Then $\alpha, \beta \notin \text{Aut}(V)$, despite the fact that α is monic and β is epic. Moreover, $\beta\alpha = \sigma_1$ but $\alpha\beta \neq \sigma_1$.

Let V be a vector space over a field F and let $\alpha \in \text{End}(V)$. A subspace W of V is **invariant** under α if and only if $\alpha(w) \in W$ for all $w \in W$ or, in other words, if and only if $\alpha(W) \subseteq W$. Thus, W is invariant under α if and only if the restriction of α to W is an endomorphism of W . It is clear that V and $\{0_V\}$ are both invariant under every endomorphism of V . If $\alpha \in \text{End}(V)$ then $\text{im}(\alpha)$ and $\ker(\alpha)$ are both invariant under α .

Example: Let F be a field and, for each positive integer k , let W_k be the subspace of $F[X]$ composed of all polynomials of degree at most k . Let δ be the **formal differentiation** endomorphism of $F[X]$, namely the endomorphism defined by $\delta : \sum_{i=0}^n a_i X^i \mapsto \sum_{i=0}^n i a_i X^{i-1}$. Then each of the subspaces W_k is invariant under δ . Now assume that F is of characteristic 0. If $p(X) = \sum_{i=0}^n a_i X^i \in W_k$ and if $a \in F$ then it is easy to check that $p(X) = p(a) + \sum_{h=1}^n \frac{1}{h!} [\delta^h(p)(a)] (X-a)^h$. The coefficients $\frac{1}{h!} [\delta^h(p)(a)]$ are known as the **Taylor coefficients** of $p(X)$ around a .

Example: Let $V = \mathbb{R}^2$ and let α be the endomorphism of V defined by $\alpha : \begin{bmatrix} a \\ b \end{bmatrix} \mapsto \begin{bmatrix} b \\ -a \end{bmatrix}$. Let W be a proper subspace of V which is invariant under α . Then $\dim(W) \leq 1$ and so there exists a vector $w = \begin{bmatrix} c \\ d \end{bmatrix}$ satisfying $W = \mathbb{R}w$. Since $\alpha(w) = \begin{bmatrix} d \\ -c \end{bmatrix}$, it follows that there exists a real number e such that $\alpha(w) = ew$. That is to say, $ec = d$ and $ed = -c$. From this we learn that $ce^2 = -c$ and so $c = d = 0$. This proves that $W = \left\{ \begin{bmatrix} 0 \\ 0 \end{bmatrix} \right\}$, and so we see that V has no proper nontrivial subspaces invariant under α .

Example: Let F be a field and let n be a positive integer. Let

α be the endomorphism of F^n defined by $\alpha : \begin{bmatrix} a_1 \\ a_2 \\ \vdots \\ a_n \end{bmatrix} \mapsto \begin{bmatrix} a_n \\ a_1 \\ \vdots \\ a_{n-1} \end{bmatrix}$.

A subspace W invariant under α is **cyclic**. Cyclic subspaces of F^n , where F is a finite field, are important in defining certain families of error-correcting codes.

Let V be a vector space over a field F . An endomorphism α of V is a **projection** if and only if $\alpha^2 = \alpha$. Note that if α is a projection and if $w = \alpha(v) \in \text{im}(\alpha)$ then $\alpha(w) = \alpha^2(v) = \alpha(v) = w$, so that the restriction of α to its image is just σ_1 . The converse is also true. If $\alpha \in \text{End}(V)$ satisfies the condition that the restriction of α to its image is just σ_1 , then for each $v \in V$ we have $\alpha^2(v) = \alpha(\alpha(v)) = \sigma_1(\alpha(v)) = \alpha(v)$ and so α is a projection.

Example: If F is a field then the endomorphism of F^3 defined by

$$\begin{bmatrix} a \\ b \\ c \end{bmatrix} \mapsto \begin{bmatrix} 3a - 2c \\ -a + b + c \\ 3a - 2c \end{bmatrix} \text{ is a projection.}$$

Example: The sum of two projections need not be a projection. For example, if $V = \mathbb{R}^3$ then the endomorphisms α and β of V defined

$$\text{by } \alpha : \begin{bmatrix} a \\ b \\ c \end{bmatrix} \mapsto \begin{bmatrix} a \\ b \\ 0 \end{bmatrix} \text{ and } \beta : \begin{bmatrix} a \\ b \\ c \end{bmatrix} \mapsto \begin{bmatrix} 0 \\ b \\ c \end{bmatrix} \text{ are projections, but}$$

$\alpha + \beta$ is not a projection.

Example: If W is a subspace of a vector space V over a field F having a complement Y in V , we know that every element $v \in V$ can be written in a unique way in the form $w + y$, where $w \in W$ and $y \in Y$. The endomorphism of V defined by $v \mapsto w$ is a projection the image of which is W .

In fact, all projections of a vector space are of the form in the previous example, as the following example shows.

(7.5) Proposition: Let V be a vector space over a field F and let $\alpha \in \text{End}(V)$ be a projection. Then $V = \text{im}(\alpha) \oplus \ker(\alpha)$.

Proof: If $v \in \text{im}(\alpha) \cap \ker(\alpha)$ then there exists an element y of V satisfying $v = \alpha(y)$ and so $v = \alpha(v) = 0_V$. Thus $\text{im}(\alpha)$ and $\ker(\alpha)$ are disjoint. If v is an arbitrary vector in V then $v = [v - \alpha(v)] + \alpha(v) \in \ker(\alpha) + \text{im}(\alpha)$. Therefore $V = \text{im}(\alpha) \oplus \ker(\alpha)$. \square

(7.6) Proposition: Let V be a vector space over a field F and let $\alpha \in \text{End}(V)$. A subspace W of V is invariant under α if and only if $\beta\alpha\beta = \alpha\beta$ for each projection β of V the image of which is W .

Proof: Assume that W is invariant under α and let β be a projection of V the image of which is W . By Proposition 7.5 we have $V = W \oplus \ker(\beta)$. If $v \in V$ we can therefore write $v = w + y$, where $w \in W$ and $y \in \ker(\beta)$. Hence $\alpha\beta(v) = \alpha\beta(w) + \alpha\beta(y) = \alpha(w) + 0_V = \alpha(w) = \beta\alpha(w) = \beta\alpha\beta(v)$, showing that $\beta\alpha\beta = \alpha\beta$. Conversely, if $\beta\alpha\beta = \alpha\beta$ for each projection β of V the image of which is W then, for each such β , we have $w = \beta(w)$ for all $w \in W$ and so $\alpha(w) = \alpha\beta(w) = \beta\alpha\beta(w) \in W$, showing that W is invariant under α . \square

(7.7) Proposition: Let V be a vector space over a field F and let $\{W_1, \dots, W_n\}$ be a set of subspaces of V . Then the following conditions are equivalent:

- (1) $V = W_1 \oplus \dots \oplus W_n$;
- (2) There exist projections $\alpha_1, \dots, \alpha_n$ in $\text{End}(V)$ with $W_i = \text{im}(\alpha_i)$ for all $1 \leq i \leq n$, which satisfy the conditions $\alpha_i \alpha_j = \sigma_0$ for $i \neq j$ and $\alpha_1 + \dots + \alpha_n = \sigma_1$.

Proof: (1) \Rightarrow (2): From (1) it follows that every $v \in V$ can be written in a unique manner as $\sum_{i=1}^n w_i$, where $w_i \in W_i$ for all $1 \leq i \leq n$. Define α_i to be the projection $v \mapsto w_i$ for each i . It is easy to verify that these linear transformations do indeed satisfy the required conditions.

(2) \Rightarrow (1): Since $\alpha_1 + \dots + \alpha_n = \sigma_1$ we surely have $V = \sum_{i=1}^n \text{im}(\alpha_i) = \sum_{i=1}^n W_i$. If $0_V \neq v \in W_h \cap \sum_{j \neq h} W_j$ then there exists an $i \neq h$ such that $\alpha_i(v) \neq 0_V$. But $\alpha_h(v) = v$ so $\alpha_i \alpha_h \neq \sigma_0$, which is a contradiction. Therefore $W_h \cap \sum_{j \neq h} W_j = \{0_V\}$ for each $1 \leq h \leq n$, proving (1). \square

(7.8) Proposition: Any two complements of a subspace W of a vector space V over a field F are isomorphic.

Proof: Let U and Y be complements of W in V . By Proposition 7.7 we know that there exists a projection $\beta \in \text{End}(V)$ the image of which is U and the kernel of which is W . Let α be the restriction of β to Y . The linear transformation α is a monomorphism since $\ker(\alpha) \subseteq \ker(\beta) \cap Y = W \cap Y = \{0_V\}$. Any vector $u \in U$ can be written as $w + y$, where $w \in W$ and $y \in Y$ and we have $\alpha(y) = \beta(y) = \beta(w) + \beta(y) = \beta(w + y) = \beta(u) = u$. Thus we see that α is also epic and hence is the desired isomorphism. \square

We now introduce a notion which is basic in all branches of mathematics. A relation \equiv defined between the elements of a given nonempty set U is called an **equivalence relation** if and only if the following conditions are satisfied:

- (1) $u \equiv u$ for all $u \in U$;
- (2) $u \equiv u'$ if and only if $u' \equiv u$;
- (3) If $u \equiv u'$ and $u' \equiv u''$ then $u \equiv u''$.

Example: Let B be a nonempty subset of a set A and define a relation \equiv_B on A by setting $a \equiv_B a'$ if and only if $a = a'$ or both a and a' belong to B . Then \equiv_B is an equivalence relation on A . In particular, if W is a subspace of a vector space V then the relation \equiv_W defined on V by setting $v \equiv_W v'$ if and only if $v - v' \in W$ is an equivalence relation on V .

Example: Let V and W be vector spaces over a field F and let $\alpha \in \text{Hom}_F(V, W)$. Define a relation \equiv on V by setting $v \equiv v'$ if and only if $\alpha(v) = \alpha(v')$. This is easily seen to be an equivalence relation.

Let V be a vector space over a field F . A subset G of $\text{Aut}(V)$ is a **group of automorphisms** if it is closed under taking products, contains σ_1 , and satisfies the condition that $\alpha^{-1} \in G$ whenever $\alpha \in G$. Clearly $\text{Aut}(V)$ itself is such a group. The notion of a group of automorphisms is very important in linear algebra and its applications, but here we will only touch on it.

Example: Let V be a vector space over a field F and let Ω be a nonempty set. Every permutation π of Ω defines an automorphism α_π of the vector space V^Ω over F defined by $\alpha_\pi(f) : i \mapsto f(\pi(i))$ for all $i \in \Omega$ and all $f \in V^\Omega$. The collection G of all such automorphisms is a group of automorphisms in $\text{Aut}(V^\Omega)$.

(7.9) Proposition: If V is a vector space over a field F and if G is a group of automorphisms of V then G defines an equivalence relation \sim_G on V by setting $v \sim_G v'$ if and only if there exists an element α of G satisfying $\alpha(v) = v'$.

Proof: If $v \in V$ then $\sigma_1(v) = v$ and so $v \sim_G v$. If $v, v' \in V$ satisfy $v \sim_G v'$ then there exists an element α of G satisfying $\alpha(v) = v'$ and so $v' = \alpha^{-1}(v)$. Thus $v' \sim_G v$. Finally, if $v, v', v'' \in V$ satisfy $v \sim_G v'$ and $v' \sim_G v''$ then there exist elements α and β of G satisfying $\alpha(v) = v'$ and $\beta(v') = v''$ and so $\beta\alpha(v) = v''$. Thus $v \sim_G v''$. \square

(7.10) Proposition: If V is a vector space over a field F and if G is a group of automorphisms of V then all elements of G have the same rank.

Proof: If $\alpha \in G$ then, by Proposition 6.11, $\text{rk}(\sigma_1) = \text{rk}(\alpha\alpha^{-1}) \leq \text{rk}(\alpha) = \text{rk}(\alpha\sigma_1) \leq \text{rk}(\sigma_1)$ and so $\text{rk}(\alpha) = \text{rk}(\sigma_1)$. \square

Exercises

Exercise 269 Let V be a vector space over $GF(3)$. Show that there exists an endomorphism α of V satisfying $\alpha(v) + \alpha(v) = v$ for all $v \in V$.

Exercise 270 Let V be a vector space finitely generated over a field F and let $\alpha, \beta, \gamma \in \text{End}(V)$. Find necessary and sufficient conditions for there to exist an endomorphism θ of V satisfying $\alpha\gamma\beta = \beta\theta\alpha$.

Exercise 271 Let $F = GF(2)$ and let n be a positive integer. Let

$$\alpha : F^n \rightarrow F^n \text{ be the function defined by } \alpha : \begin{bmatrix} a_1 \\ \vdots \\ a_n \end{bmatrix} \mapsto \begin{bmatrix} a'_1 \\ \vdots \\ a'_n \end{bmatrix}, \text{ where}$$

$0' = 1$ and $1' = 0$. Is α an endomorphism of F^n ?

Exercise 272 Let $\alpha, \beta : \mathbb{Q}[X] \rightarrow \mathbb{Q}[X]$ be defined by $\alpha : p(X) \mapsto Xp(X)$ and $\beta : p(X) \mapsto X^2p(X)$. Show that α , β , and $\alpha - \beta$ are all monic endomorphisms of $\mathbb{Q}[X]$.

Exercise 273 Let V be a finitely-generated vector space over a field F and let $\alpha \in \text{End}(V)$. Show that α is not monic if and only if there exists an endomorphism $\beta \neq \sigma_0$ of V satisfying $\alpha\beta = \sigma_0$.

Exercise 274 Let V be a vector space over a field F and let $\alpha \in \text{End}(V)$. Show that $\ker(\alpha) = \ker(\alpha^2)$ if and only if $\ker(\alpha)$ and $\text{im}(\alpha)$ are disjoint.

Exercise 275 Let V be a vector space over a field F and let $\alpha \in \text{End}(V)$. Show that $\text{im}(\alpha) = \text{im}(\alpha^2)$ if and only if $V = \ker(\alpha) + \text{im}(\alpha)$.

Exercise 276 Let V be a vector space over a field F and let $K = F \times V \times \text{End}(V)$, which is again a vector space over F . Define an operation \diamond on K by setting $(a, v, \alpha) \diamond (b, w, \beta) = (ab, aw + \beta(v), \beta\alpha)$. Is (K, \diamond) an F -algebra? Is it associative? Is it unital?

Exercise 277 Let V be a vector space over a field F , and let $\text{Aff}(V, V)$ be the set of all affine transformations from V to itself. Is $\text{Aff}(V, V)$, on which we have defined the operations of addition and composition of functions, an associative unital F -algebra?

Exercise 278 Let $\alpha \in \text{Aut}(\mathbb{R}^2)$ be defined by $\alpha : \begin{bmatrix} a \\ b \end{bmatrix} \mapsto \begin{bmatrix} -b \\ a \end{bmatrix}$.

Show that $\mathbb{R}\{\alpha, \sigma_1\}$ is a unital subalgebra of $\text{End}(\mathbb{R}^2)$. Show that it is proper by giving an example of an endomorphism of \mathbb{R}^2 not in this subalgebra.

Exercise 279 Let V be the space of all real-valued functions on the interval $[-1, 1]$ which are infinitely differentiable, and let δ be the endomorphism of V which assigns to each function f its derivative. Find the kernel and image of δ .

Exercise 280 Let $\alpha : \mathbb{C} \rightarrow \mathbb{C}$ be the function defined by $\alpha : a + bi \mapsto -b + ai$. Is α an endomorphism of \mathbb{C} considered as a vector space over \mathbb{R} ? Is it an endomorphism of \mathbb{C} considered as a vector space over itself?

Exercise 281 Let V be a vector space of finite dimension n over a field F and let $\alpha \in \text{End}(V)$. Show that there exists an automorphism β of V satisfying $\alpha\beta\alpha = \alpha$.

Exercise 282 Let $V = \mathcal{M}_{2 \times 2}(\mathbb{R})$, which is a vector space over \mathbb{R} . Consider the function $\alpha : V \rightarrow V$ defined by

$$\alpha : \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \mapsto \begin{bmatrix} |a_{11}| & |a_{12}| \\ |a_{21}| & |a_{22}| \end{bmatrix}.$$

Is α an endomorphism of V ?

Exercise 283 Consider \mathbb{R} as a vector space over \mathbb{Q} and let α be an endomorphism of this space satisfying the condition that there exists an $a_0 \in \mathbb{R}$ such that α is continuous at a_0 . Show that α is continuous at every $a \in \mathbb{R}$.

Exercise 284 Let A be a nonempty set and let V be the collection of all subsets of A , considered as a vector space over $GF(2)$. For which subsets C of A is the function $B \mapsto B \cup C$ an endomorphism of V ?

Exercise 285 Let V be a vector space of finite dimension n over a field F and let $\{\alpha_{ij} \mid 1 \leq i, j \leq n\}$ be a collection of endomorphisms of V , not all of which are equal to σ_0 , satisfying the condition that

$$\alpha_{ij}\alpha_{kh} = \begin{cases} \alpha_{ih} & \text{if } j = k \\ \sigma_0 & \text{otherwise} \end{cases}.$$

Show that there exists a basis $\{v_1, \dots, v_n\}$ of V such that

$$\alpha_{jk}(v_i) = \begin{cases} v_j & \text{if } i = k \\ 0_V & \text{otherwise} \end{cases}.$$

Exercise 286 Let V be a vector space of finite dimension n over a field F and choose an element $\alpha \in \text{End}(V)$. Let $\varphi : \text{End}(V) \rightarrow \text{End}(V)$ be the function defined by $\beta \mapsto \beta\alpha$. This is an endomorphism of $\text{End}(V)$, considered as a vector space over F . Show that a positive integer n satisfies $\alpha^n = \sigma_0$ if and only if φ^n is the 0-function.

Exercise 287 Let α be an endomorphism of \mathbb{R}^3 satisfying the condition that $\alpha^2 = \sigma_0$. Show that there exists a linear transformation $\beta : \mathbb{R}^3 \rightarrow \mathbb{R}^3$ and that there exists a vector $y \in \mathbb{R}^3$ satisfying $\alpha(v) = \beta(v)y$ for all $v \in \mathbb{R}^3$.

Exercise 288 For each $0 \neq a \in \mathbb{R}$, let $\beta_a : \mathbb{C} \rightarrow \mathbb{C}$ be the function defined by $\beta_a : z \mapsto z + a\bar{z}$. Show that β_a is an endomorphism of \mathbb{C} considered as a vector space over \mathbb{R} , and describe its image and kernel.

Exercise 289 Let V be a vector space finitely generated over \mathbb{Q} and let $\alpha, \beta \in \text{End}(V)$ satisfy $3\alpha^3 + 7\alpha^2 - 2\alpha\beta + 4\alpha - \sigma_1 = \sigma_0$. Show that $\alpha\beta = \beta\alpha$.

Exercise 290 Let V be a vector space over a field F and let $\alpha, \beta, \gamma \in \text{End}(V)$ satisfy $\alpha\beta = \sigma_1 = \alpha\gamma$. Show that $\beta\gamma \neq \gamma\beta$.

Exercise 291 Let F be a field of characteristic other than 2 and let V be a vector space of finite dimension n over F . Let α be an endomorphism of V satisfying the condition that $\alpha^2 = \sigma_1$. Show that $\text{rk}(\sigma_1 - \alpha) + \text{rk}(\sigma_1 + \alpha) = n$.

Exercise 292 Let V be a vector space over a field F which is not finitely generated, and let $\sigma_0 \neq \alpha \in \text{End}(V)$. Set $A = \{\beta \in \text{End}(V) \mid \alpha\beta = \sigma_1\}$. Show that if A has more than one element then it is infinite.

Exercise 293 Let V be a vector space over a field F having dimension greater than 1. Show that there exists a function $\alpha \in V^V$ which is not an endomorphism of V but which nonetheless satisfies the condition that $\alpha(av) = a\alpha(v)$ for all $a \in F$ and all $v \in V$.

Exercise 294 Let V be a vector space over a field F satisfying the condition that $\alpha\beta = \beta\alpha$ for all $\alpha, \beta \in \text{End}(V)$. Show that $\dim(V) = 1$.

Exercise 295 Let $V = \mathcal{M}_{2 \times 2}(\mathbb{R})$, considered as a vector space over \mathbb{R} . Let $\alpha : V \rightarrow V$ be the function defined by

$$\alpha : \begin{bmatrix} a & b \\ c & d \end{bmatrix} \mapsto \begin{bmatrix} a + 2b + c + 2d & 2a + 4b + 3c + 5d \\ 3a + 6b + 2c + 5d & a + 2b + c + 2d \end{bmatrix}.$$

Is α an endomorphism of V ? Is it an automorphism of V ?

Exercise 296 Let V be the vector space of all continuous functions from \mathbb{R} to itself and let $\alpha : V \rightarrow V$ be the function defined by $\alpha : f(x) \mapsto [x^2 + \sin(x) + 2] f(x)$. Show that α is an automorphism of V .

Exercise 297 Let F be a field and let $\alpha : F[X] \rightarrow F[X]$ be the function defined by $\alpha : p(X) \mapsto p(X + 1)$. Is α an endomorphism of $F[X]$? Is it an automorphism?

Exercise 298 Let F be a field and, for each $a \in F$, let θ_a be the endomorphism of $F[X]$ defined by $\theta_a : p(X) \mapsto p(X + a)$. Let $\alpha \in \text{End}(F[X])$ satisfy $\alpha(X) \in F$ and $\alpha\theta_a = \theta_a\alpha$ for all $a \in F$. Can α be a monomorphism?

Exercise 299 Let $\alpha \in \text{End}(\mathbb{R}^3)$ be defined by $\alpha : \begin{bmatrix} a \\ b \\ c \end{bmatrix} \mapsto \begin{bmatrix} a - 2b \\ c \\ a - b \end{bmatrix}$. Is

α an automorphism of \mathbb{R}^3 ?

Exercise 300 Let α be the endomorphism of $\mathbb{R}^{(\infty)}$ defined by

$$\alpha : [a_1, a_2, a_3, \dots] \mapsto [b_1, b_2, b_3, \dots],$$

where $b_h = \sum_{j \leq h} (-1)^{j-1} \binom{h-1}{j-1} a_j$ for each $h \geq 1$. Show that α is an automorphism satisfying $\alpha = \alpha^{-1}$.

Exercise 301 Let V be a vector space finitely generated over \mathbb{R} and let α be an endomorphism of V satisfying $\alpha^3 + 4\alpha^2 + 2\alpha + \sigma_1 = \sigma_0$. Show that $\alpha \in \text{Aut}(V)$.

Exercise 302 Let V be a vector space over a field F and let $\alpha, \beta \in \text{End}(V)$ satisfy $\alpha\beta = \sigma_1$. Set $\varphi = \sigma_1 - \beta\alpha$. Show that for every integer $n \geq 1$ we have $\sigma_1 = \sum_{k=0}^{n-1} \beta^k \varphi \alpha^k + \beta^n \alpha^n$.

Exercise 303 Let V be the space of all polynomial functions from the interval $[0, 1]$ on the real line to \mathbb{R} . Let α and β be the endomorphisms of V defined by $\alpha(f) : x \mapsto \int_0^x f(t) dt$ and $\beta(f) : x \mapsto \int_x^1 f(t) dt$. Find $\text{im}(\alpha + \beta)$. Is it true that $\alpha\beta = \beta\alpha$?

Exercise 304 Let F be a field and let $V = F^\infty$. Let $n > 1$ be an

integer. Each vector $y = \begin{bmatrix} d_1 \\ \vdots \\ d_n \end{bmatrix} \in F^n$ defines an endomorphism θ_y

of V by $\theta_y : [a_1, a_2, \dots] \mapsto [b_1, b_2, \dots]$, where $b_h = \sum_{i=1}^n a_{h-1+i} d_i$, for $h = 1, 2, \dots$. Show that if θ_y is a monomorphism then the polynomial $p(X) = \sum_{i=1}^n d_i X^{i-1} \in F[X]$ has no roots in F .

Exercise 305 Let $F = GF(5)$ and let $V = F^3$. How many endo-

morphisms α of V satisfy the conditions $\alpha \left(\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \right) = \begin{bmatrix} 2 \\ 1 \\ 0 \end{bmatrix}$ and

$$\alpha \left(\begin{bmatrix} 0 \\ 3 \\ 0 \end{bmatrix} \right) = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}?$$

Exercise 306 Let V be the set of all continuous functions from \mathbb{R} to itself, which is a vector space over \mathbb{R} . Let $\alpha : V \rightarrow V$ be the function defined by $\alpha(f) : x \mapsto f\left(\frac{x}{2}\right)$ for all $x \in \mathbb{R}$ and all $f \in V$. Is α an automorphism of V ?

Exercise 307 Let $V = \mathbb{R}^\infty$ and let W be the subspace of V consisting of all convergent sequences. Let $\alpha \in \text{End}(V)$ be defined by $\alpha : [a_1, a_2, \dots] \mapsto [b_1, b_2, \dots]$, where $b_h = \frac{1}{h} \left(\sum_{i=1}^h a_i \right)$ for all $h \geq 1$. If $v \in V$ satisfies $\alpha(v) \in W$, is v itself necessarily in W ?

Exercise 308 Let V be a vector space over a field F and let $\alpha \in \text{Aut}(V)$. Let W_1, \dots, W_k be subspaces of V satisfying $V = \bigoplus_{i=1}^k W_i$. For each $1 \leq i \leq k$, let $Y_i = \{\alpha(w) \mid w \in W_i\}$. Is $V = \bigoplus_{i=1}^k Y_i$?

Exercise 309 Consider \mathbb{R} as a vector space over \mathbb{Q} . An endomorphism of this space is **bounded** if and only if there exists a nonnegative real number $m(\alpha)$ satisfying the condition that $|\alpha(x)| \leq m(\alpha)|x|$ for all $x \in \mathbb{R}$. Does the set of all bounded endomorphisms of \mathbb{R} form an \mathbb{R} -subalgebra of $\text{End}(\mathbb{R})$?

Exercise 310 Let F be a field, let n be a positive integer, and let $V = \mathcal{M}_{n \times n}(F)$. Given a matrix $B \in V$, is the function $\alpha_B : V \rightarrow V$ defined by $\alpha_B : A \mapsto AB + BA$ an endomorphism of V ?

Exercise 311 Let F be a field and let $V = F[X]$. Let $\delta \in \text{End}(V)$ be the formal differentiation function and let $\alpha \in \text{End}(V)$ be defined by $\alpha : p(X) \mapsto Xp(X)$. Show that $\alpha\delta - \delta\alpha = \sigma_1$.

Exercise 312 Let V be a nontrivial vector space over a field F . Is the set of all automorphisms of V a subspace of the vector space $\text{End}(V)$ over F ?

Exercise 313 Consider $GF(3)$ as a vector space over itself. Does there exist an automorphism of this space other than σ_1 ?

Exercise 314 Let $V = F^\infty$ for some field F . Each $w = [c_1, c_2, \dots] \in V$ defines a function $\beta_w : V \rightarrow V$ by

$$\beta_w : [a_1, a_2, \dots] \mapsto [a_1, a_1c_1 + a_2, (a_1c_1 + a_2)c_2 + a_3, \dots].$$

Show that β_w is an automorphism of V .

Exercise 315 Let V be a vector space over a field F ; let $\alpha \in \text{End}(V)$ and let $\beta \in \text{Aut}(V)$. Define the function $\theta : V^2 \rightarrow V^2$ by setting

$$\theta : \begin{bmatrix} v \\ v' \end{bmatrix} \mapsto \begin{bmatrix} \beta(v) \\ \alpha(v) + v' \end{bmatrix}. \text{ Is } \theta \text{ necessarily an automorphism of } V^2?$$

Exercise 316 Let $F = GF(5)$ and let $\alpha \in \text{Aut}(F^2)$ be defined by

$$\alpha : \begin{bmatrix} a \\ b \end{bmatrix} \mapsto \begin{bmatrix} 2b \\ a + 2b \end{bmatrix}. \text{ Show that there exists a positive integer } h$$

satisfying $\alpha^{h+1} = \alpha$ and find the smallest such integer h .

Exercise 317 Let F be a field of characteristic other than 2. Let V be a vector space over F and let $\alpha, \beta, \gamma, \delta$ be endomorphisms of V satisfying the condition that $\alpha - \beta$ and $\alpha + \beta$ are automorphisms of V . Show that there exist endomorphisms φ and ψ of V satisfying $\varphi\alpha + \psi\beta = \gamma$ and $\psi\alpha + \varphi\beta = \delta$.

Exercise 318 Let V be a vector space of finite dimension n over a field F . Let $\alpha \in \text{End}(V)$ and assume that there exists a vector $y \in V$ satisfying the condition that $D = \{\alpha(y), \alpha^2(y), \dots, \alpha^n(y)\}$ is a basis for V . Show that $D' = \{y, \alpha(y), \dots, \alpha^{n-1}(y)\}$ is also a basis for V and that $\alpha \in \text{Aut}(V)$.

Exercise 319 Let F be a field and let $V = F^{\mathbb{Z}}$. Let α be the endomorphism of V defined by $\alpha(f) : i \mapsto f(i+1)$ for all $f \in V$. Show that $\alpha - c\sigma_1 \notin \text{Aut}(V)$ for all $0 \neq c \in F$.

Exercise 320 Let V be a vector space of finite dimension n over a field F , and let $0 < k \leq n$ be a positive integer. Let A_k be the set of all subspaces of V having dimension k . Let $\alpha \in \text{Aut}(V)$ and, for each $W \in A_k$, let $\theta_\alpha(W) = \{\alpha(w) \mid w \in W\}$. Show that the function θ_α is a permutation of A_k .

Exercise 321 Let r, s , and t be distinct real numbers and let α be

the endomorphism of \mathbb{R}^3 defined by $\alpha : \begin{bmatrix} a \\ b \\ c \end{bmatrix} \mapsto \begin{bmatrix} a + br + cr^2 \\ a + bs + cs^2 \\ a + bt + ct^2 \end{bmatrix}$. Is

α an automorphism of \mathbb{R}^3 ?

Exercise 322 Let V be a vector space over a field F and let $\alpha \in \text{End}(V)$. Show that $W = \bigcup_{i=1}^{\infty} \ker(\alpha^i)$ is a subspace of V which is invariant under α .

Exercise 323 Let α and β be the endomorphisms of \mathbb{Q}^4 defined by

$$\alpha : \begin{bmatrix} a \\ b \\ c \\ d \end{bmatrix} \mapsto \begin{bmatrix} 2a - 2b - 2c - 2d \\ 5b - c - d \\ -b + 5c - d \\ -b - c + 5d \end{bmatrix} \quad \text{and} \quad \beta : \begin{bmatrix} a \\ b \\ c \\ d \end{bmatrix} \mapsto \begin{bmatrix} 0 \\ -b + 2c + 3d \\ 2b - 3c + 6d \\ 3b + 6c + 2d \end{bmatrix}.$$

Find two nontrivial proper subspaces of \mathbb{Q}^4 which are invariant both under α and under β .

Exercise 324 Let F be a field and let $V = F^4$. Let α be the endomor-

phism of V defined by $\alpha : \begin{bmatrix} a \\ b \\ c \\ d \end{bmatrix} \mapsto \begin{bmatrix} a - b \\ a - b \\ a - b \\ c - b - d \end{bmatrix}$. Does there exist a

two-dimensional subspace of V invariant under α ?

Exercise 325 Let V be a vector space over a field F and let $\alpha \in \text{End}(V)$. If W and Y are subspaces of V which are invariant under α , show that both $W + Y$ and $W \cap Y$ are invariant under α .

Exercise 326 Let W be a subspace of a vector space V over a field F and let S be the set of all $\alpha \in \text{End}(V)$ such that W is invariant under α . Is S necessarily an F -subalgebra of $\text{End}(V)$?

Exercise 327 Let V be a vector space over a field F and let $\alpha \in \text{End}(V)$. If W is a subspace of V , show that the set of all subspaces of W which are invariant under α , partially ordered by inclusion, has a maximal element.

Exercise 328 Let $V = \mathbb{R}^\infty$ and let W be the subspace of V consisting of all sequences $[a_1, a_2, \dots]$ for which the series $\sum_{i=1}^\infty a_i$ converges. Let σ be a permutation of the set of all positive integers and let $\alpha \in \text{End}(V)$ be defined by $\alpha : [a_1, a_2, \dots] \rightarrow [a_{\sigma(1)}, a_{\sigma(2)}, \dots]$. Is W invariant under α ?

Exercise 329 Let V be a vector space over a field F . Let $0 \neq c \in F$ and let $\alpha \in \text{End}(V)$. Let $\{x_0, x_1, \dots, x_n\}$ be a set of vectors in V satisfying $\alpha(x_0) = cx_0$ and $\alpha(x_i) - cx_i = x_{i-1}$ for all $1 \leq i \leq n$. Show that $F\{x_0, x_1, \dots, x_n\}$ is a subspace of V which is invariant under α .

Exercise 330 Let F be a field which is not finite and let V be a vector space over F having dimension greater than 1. For each $0 \neq c \in F$, show that there exist infinitely-many distinct subspaces of V which are invariant under the endomorphism σ_c of V .

Exercise 331 Let α and β be endomorphisms of a vector space V over a field F and let $\theta \in \text{Aut}(V)$ satisfy $\theta\alpha = \beta\theta$. Show that a subspace W of V is invariant under α if and only if $W' = \{\theta(w) \mid w \in W\}$ is invariant under β .

Exercise 332 Let α and β be endomorphisms of a vector space V over a field F satisfying $\alpha\beta = \beta\alpha$. Is $\ker(\alpha)$ invariant under β ?

Exercise 333 Let V be a vector space over a field F and let $\alpha \in \text{End}(V)$ be a projection. Show that $\sigma_1 - \alpha$ is also a projection.

Exercise 334 Let V be a vector space finitely generated over a field F and let $\alpha \in \text{End}(V)$ satisfy the condition $\alpha^2(\sigma_1 - \alpha) = \sigma_0$. Is α necessarily a projection?

Exercise 335 Let V be the space of all continuous functions from \mathbb{R} to itself and let $W = \mathbb{R}\{\sin(x), \cos(x)\} \subseteq V$. Let δ be the endomorphism of W which assigns to each function its derivative. Find a polynomial $p(X) \in \mathbb{R}[X]$ of degree 2 satisfying $p(\delta) = \sigma_0$.

Exercise 336 Let V be a vector space having finite dimension over \mathbb{Q} and assume that there exists an $\alpha \in \text{Aut}(V)$ satisfying $\alpha^{-1} = \alpha^2 + \alpha$. Show that $\dim(V)$ is divisible by 3.

Exercise 337 Let n be a positive integer and let

$$G = \left\{ \left[\begin{array}{c} a_1 \\ \vdots \\ a_n \end{array} \right] \in \mathbb{R}^n \mid a_i \geq 0 \text{ for all } 1 \leq i \leq n \right\}.$$

Let α be an endomorphism of \mathbb{R}^n satisfying the condition that $\alpha(v) \in G$ implies that $v \in G$. Show that $\alpha \in \text{Aut}(\mathbb{R}^n)$.

Exercise 338 Let V be a vector space over a field F and let W and Y be subspaces of V satisfying $W + Y = V$. Let Y' be a complement of Y in V and let Y'' be a complement of $W \cap Y$ in W . Show that $Y' \cong Y''$.

Exercise 339 Let F be a field of characteristic other than 2 and let V be a vector space over F . Let $\alpha, \beta \in \text{End}(V)$ be projections satisfying the condition that $\alpha + \beta$ is also a projection. Show that $\alpha\beta = \beta\alpha = \sigma_0$.

Exercise 340 Let V be a vector space over F and let $\alpha, \beta \in \text{End}(V)$. Show that α and β are projections satisfying $\ker(\alpha) = \ker(\beta)$ if and only if $\alpha\beta = \alpha$ and $\beta\alpha = \beta$.

Exercise 341 Let V be a vector space finitely generated over a field F and let $\alpha \neq \sigma_1$ be an endomorphism of V which is a product of projections. Show that $\alpha \notin \text{Aut}(V)$.

Exercise 342 Let V be a vector space over \mathbb{Q} and let $\alpha \in \text{End}(V)$. Show that α is a projection if and only if $(2\alpha - \sigma_1)^2 = \sigma_1$.

Exercise 343 Let α and β be endomorphisms of a vector space V over a field F and let $f(X) \in F[X]$ satisfy $f(\alpha\beta) = \sigma_0$. Set $g(X) = Xf(X)$. Show that $g(\beta\alpha) = \sigma_0$.

Exercise 344 Let $V = \mathbb{R}^{\mathbb{R}}$ and let $g \in V$. Find necessary and sufficient conditions on g for the endomorphism $f \mapsto gf$ of V to be a projection.

Exercise 345 Let W be a subspace of a vector space V over a field F which is invariant under an endomorphism α of V . Let $\beta \in \text{End}(V)$ be a projection satisfying the condition that $\text{im}(\beta) = W$. Show that $\beta\alpha\beta = \alpha\beta$.

Exercise 346 Let V be a vector space of finite dimension n over a field F and let $\alpha \in \text{End}(V)$. Show that there exists an automorphism β of V and a projection θ of V satisfying $\alpha = \beta\theta$.

Exercise 347 Let F be a field of characteristic other than 2 and let V be a vector space over F . Let $\alpha \in \text{End}(V)$ be a projection satisfying the condition that $\alpha - \beta$ is a projection for all $\beta \in \text{End}(V)$. Show that $\alpha = \sigma_1$.

Exercise 348 Let V be a vector space over F and let $\alpha, \beta \in \text{End}(V)$ be projections satisfying the condition that $\text{im}(\alpha)$ and $\text{im}(\beta)$ are disjoint. Is it necessarily true that $\alpha\beta = \beta\alpha$?

Exercise 349 Let V be a vector space of finite dimension n over a field F and let $S = \text{End}(V) \setminus \text{Aut}(V)$. For $\alpha, \beta \in S$, show that $\text{im}(\alpha) = \text{im}(\beta)$ if and only if $\{\alpha\theta \mid \theta \in S\} = \{\beta\varphi \mid \varphi \in S\}$.

Exercise 350 Let F be a field. Does there exist an endomorphism α of F^3 which is not a projection satisfying the condition that α^2 is a projection equal neither to σ_0 nor to σ_1 .

Exercise 351 Let V be a vector space finite dimensional over a field F and let α be an endomorphism of V . Show that there exist a positive integer k such that $\text{im}(\alpha^k)$ and $\ker(\alpha^k)$ are disjoint.

Exercise 352 Let F be a field of characteristic other than 2, and let V be a vector space over F . Let $\alpha \in \text{End}(V)$ satisfy $\alpha^3 = \alpha$. Show that $V = W_1 \oplus W_2 \oplus W_3$, where $W_1 = \{v \in V \mid \alpha(v) = v\}$, $W_2 = \{v \in V \mid \alpha(v) = -v\}$, and $W_3 = \ker(\alpha)$.

Exercise 353 Let F be a field of characteristic other than 2 and let V be a finitely-generated vector space over F . Show that every endomorphism of V is the sum of two automorphisms of V .

Exercise 354 Let $n > 1$ be an integer and let $\theta : \mathbb{R}^n \rightarrow \mathbb{R}$ be the

function defined by $\theta : \begin{bmatrix} a_1 \\ \vdots \\ a_n \end{bmatrix} \mapsto \sum_{i=1}^n a_i^2$. Assume that we can define an

operation \bullet on \mathbb{R}^n satisfying the condition that (\mathbb{R}^n, \bullet) is an associative unital \mathbb{R} -algebra with multiplicative identity e , and also satisfying the condition that $\theta(v \bullet w) = \theta(v)\theta(w)$ for all $v, w \in \mathbb{R}^n$. Show that $(\mathbb{R}^n, +, \bullet)$ is a division algebra over \mathbb{R} .

Exercise 355 Any sequence $v = [a_1, a_2, \dots] \in \mathbb{R}^\infty$ defines an endomorphism α_v of $\mathbb{R}[X]$ which acts on elements of the canonical basis of $\mathbb{R}[X]$ according to the rule $\alpha_v : X^n \mapsto \sum_{k=0}^n \binom{n}{k} (k!) a_{k+1} X^{n-k}$ for each nonnegative integer n . Given $a \in \mathbb{R}$, find $v, w \in \mathbb{R}^\infty$ such that $\alpha_v : p(X) \mapsto p(X + a)$ and $\alpha_w : p(X) \mapsto p(X + a) - p(a)$.

8

Representation of linear transformations by matrices

In this chapter we show how we can study linear transformations between finitely-generated vector spaces by studying matrices¹. Let V and W be finitely-generated vector spaces over a field F , where $\dim(V) = n$ and $\dim(W) = k$. Fix bases $B = \{v_1, \dots, v_n\}$ of V and $D = \{w_1, \dots, w_k\}$ of W . From Proposition 5.4, we know that if we are given a linear transformation $\alpha \in \text{Hom}(V, W)$ then for each $1 \leq j \leq n$ there exist scalars a_{1j}, \dots, a_{kj} satisfying the condition $\alpha(v_j) = \sum_{i=1}^k a_{ij} w_i$, and that these scalars are in fact uniquely determined by α . Thus α defines a matrix $[a_{ij}] \in \mathcal{M}_{k \times n}(F)$. Conversely, assume we have a matrix $A = [a_{ij}] \in \mathcal{M}_{k \times n}(F)$. Then we know that every vector v in V can be written in a unique way in the form $\sum_{j=1}^n b_j v_j$ and so A defines a linear transformation $\alpha \in \text{Hom}(V, W)$ by setting $\alpha : v \mapsto \sum_{i=1}^k \left(\sum_{j=1}^n a_{ij} b_j \right) w_i$. Moreover, it is clear that different linear transformations in $\text{Hom}(V, W)$



¹ The theory of matrices and their relation to linear transformations was developed in detail by the 19th-century British mathematician **Sir Arthur Cayley**, one of the most prolific researchers in history.

define different matrices in $\mathcal{M}_{k \times n}(F)$ and different matrices in $\mathcal{M}_{k \times n}(F)$ define different linear transformations in $\text{Hom}(V, W)$. We summarize the above remarks in the following proposition.

(8.1) Proposition: Let V be a vector space of finite dimension n over a field F and let W be a vector space of finite dimension k over F . For every basis B of V and every basis D of W there exists a bijective function $\Phi_{BD} : \text{Hom}(V, W) \rightarrow \mathcal{M}_{k \times n}(F)$, which is an isomorphism of vector spaces over F .

Proof: We have already seen that if $B = \{v_1, \dots, v_n\}$ and $D = \{w_1, \dots, w_k\}$, then the function Φ_{BD} is defined by $\Phi_{BD}(\alpha) = [a_{ij}]$, where $\alpha(v_j) = \sum_{i=1}^k a_{ij}w_i$ for all $1 \leq j \leq n$, and that this function is bijective. We are therefore left to show that this is a linear transformation. Indeed, if $\Phi_{BD}(\alpha) = [a_{ij}]$ and $\Phi_{BD}(\beta) = [b_{ij}]$ then

$$(\alpha + \beta)(v_j) = \sum_{i=1}^k (a_{ij} + b_{ij})w_i = \sum_{i=1}^k a_{ij}w_i + \sum_{i=1}^k b_{ij}w_i = \alpha(v_j) + \beta(v_j)$$

for all $1 \leq j \leq n$, and so $\Phi_{BD}(\alpha + \beta) = \Phi_{BD}(\alpha) + \Phi_{BD}(\beta)$. Similarly, if $c \in F$ then $(c\alpha)(v_j) = \sum_{i=1}^k ca_{ij}w_i = c \left(\sum_{i=1}^k a_{ij}w_i \right) = c(\alpha(v_j))$ for all $1 \leq j \leq n$, and so $\Phi_{BD}(c\alpha) = c\Phi_{BD}(\alpha)$. Thus we see that Φ_{BD} is indeed a linear transformation and thus also an isomorphism. \square

We have already seen that, in the above situation, $\dim(\mathcal{M}_{k \times n}(F)) = kn$ and so, by Proposition 6.9, we also see that $\dim(\text{Hom}(V, W)) = kn$.

Example: Let $V = \mathbb{R}^3$ and let B be the canonical basis on V . Each vector $v = \begin{bmatrix} a_1 \\ a_2 \\ a_3 \end{bmatrix} \in V$ defines a linear transformation $\alpha_v : V \rightarrow V$

given by $\alpha_v : w \mapsto v \times w$. Then $\Phi_{BB}(\alpha_v) = \begin{bmatrix} 0 & -a_3 & a_2 \\ a_3 & 0 & -a_1 \\ -a_2 & a_1 & 0 \end{bmatrix}$.

Example: Let $V = \mathbb{R}^3$ and $W = \mathbb{R}^2$. Choose bases

$$B = \left\{ \begin{bmatrix} 0.5 \\ -0.5 \\ 0 \end{bmatrix}, \begin{bmatrix} 0.5 \\ 0 \\ -0.5 \end{bmatrix}, \begin{bmatrix} 0 \\ 0.5 \\ 0.5 \end{bmatrix} \right\}$$

of V and of $D = \left\{ \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \end{bmatrix} \right\}$ of W . If $\begin{bmatrix} r \\ s \\ t \end{bmatrix} \in \mathbb{R}^3$ then there exist $b_1, b_2, b_3 \in \mathbb{R}$ satisfying

$$\begin{bmatrix} r \\ s \\ t \end{bmatrix} = b_1 \begin{bmatrix} 0.5 \\ -0.5 \\ 0 \end{bmatrix} + b_2 \begin{bmatrix} 0.5 \\ 0 \\ -0.5 \end{bmatrix} + b_3 \begin{bmatrix} 0 \\ 0.5 \\ 0.5 \end{bmatrix} = \frac{1}{2} \begin{bmatrix} b_1 + b_2 \\ -b_1 + b_3 \\ -b_2 + b_3 \end{bmatrix}$$

and so we have to solve the system of linear equations

$$\begin{aligned} 2r &= b_1 + b_2 \\ 2s &= -b_1 + b_3 \\ 2t &= -b_2 + b_3 \end{aligned}$$

and if we do so, we get $b_1 = r - s + t$, $b_2 = r + s - t$, and $b_3 = r + s + t$.

The matrix $A = \begin{bmatrix} 3 & 5 & 7 \\ 4 & 8 & 2 \end{bmatrix}$ defines a linear transformation $\alpha \in \text{Hom}(V, W)$ given by

$$\begin{aligned} \alpha : b_1 \begin{bmatrix} 0.5 \\ -0.5 \\ 0 \end{bmatrix} + b_2 \begin{bmatrix} 0.5 \\ 0 \\ -0.5 \end{bmatrix} + b_3 \begin{bmatrix} 0 \\ 0.5 \\ 0.5 \end{bmatrix} \mapsto \\ (3b_1 + 5b_2 + 7b_3) \begin{bmatrix} 1 \\ 1 \end{bmatrix} + (4b_1 + 8b_2 + 2b_3) \begin{bmatrix} 1 \\ 0 \end{bmatrix}. \end{aligned}$$

so

$$\begin{aligned} \alpha \left(\begin{bmatrix} r \\ s \\ t \end{bmatrix} \right) &= (15r + 9s + 5t) \begin{bmatrix} 1 \\ 1 \end{bmatrix} + (14r + 6s - 2t) \begin{bmatrix} 1 \\ 0 \end{bmatrix} \\ &= \begin{bmatrix} 29r + 15s + 3t \\ 15r + 9s + 5t \end{bmatrix}. \end{aligned}$$

It is very important to emphasize that the matrix representation of a linear transformation depends on the bases which we fixed at the beginning, and on the order in which the elements of the bases are written! If we choose different bases or write the elements of a chosen basis in a different order, we will get a different matrix. Shortly, we will consider the relation between the matrices which represent a given linear transformation with respect to different bases.

Let V be a vector space finitely generated over a field F , let α be an endomorphism of V , and let W be a subspace of V which is invariant under α . As we have already seen, the restriction β of α to

W is an endomorphism of W . Now, let $B = \{v_1, \dots, v_k\}$ be a basis for W , which we can expand to a basis $D = \{v_1, \dots, v_n\}$ for all of V . If $\Phi_{DD}(\alpha) = [a_{ij}]$ then for all $1 \leq j \leq k$ we have $\alpha(v_j) = \sum_{i=1}^k a_{ij}v_i$ and so $a_{ij} = 0$ whenever $1 \leq j \leq k$ and $k < i \leq n$. Thus we see that the

matrix $\Phi_{DD}(\alpha)$ is of the form $\begin{bmatrix} A_{11} & A_{21} \\ O & A_{22} \end{bmatrix}$, where $A_{11} = \Phi_{BB}(\beta)$.

The subspace $Y = F\{v_{k+1}, \dots, v_n\}$ of V is a complement of W in V . If it too is invariant under α then we would also have $A_{21} = O$ and so α is represented by a matrix composed of two square matrices “strung out” along the diagonal. From a computational point of view, such a representation has distinct advantages.

Beside addition and scalar multiplication of matrices, we can also define the product of two matrices, provided that these matrices are of suitable sizes. Let (K, \bullet) be an associative unital algebra over a field F . If $A = [v_{ij}] \in \mathcal{M}_{k \times n}(K)$ and $B = [w_{jh}] \in \mathcal{M}_{n \times t}(K)$ for some positive integers k , n , and t , we define the matrix AB to be the matrix $[y_{ih}] \in \mathcal{M}_{k \times t}(K)$ where, for each $1 \leq i \leq k$ and all $1 \leq h \leq t$, we set $y_{ih} = \sum_{j=1}^n v_{ij} \bullet w_{jh}$. For the most part, we will be interested in this construction for the case $K = F$, but sometimes we will have need of the more general construction. Note that a necessary condition for the product of two matrices to be defined is that the number of columns in the first matrix be equal to the number of rows in the second matrix.

Example: If $A = \begin{bmatrix} 2 & 3 & 2 \\ -1 & 2 & 1 \end{bmatrix} \in \mathcal{M}_{2 \times 3}(\mathbb{Q})$ and

$$B = \begin{bmatrix} -1 & 0 & 2 & -1 \\ 1 & 0 & 1 & -2 \\ 2 & 1 & 0 & -3 \end{bmatrix} \in \mathcal{M}_{3 \times 4}(\mathbb{Q})$$

then $AB = \begin{bmatrix} 5 & 2 & 7 & -14 \\ 5 & 1 & 0 & -6 \end{bmatrix} \in \mathcal{M}_{2 \times 4}(\mathbb{Q})$ but BA is not defined.

Example: If we consider the matrices $A = \begin{bmatrix} 2 & 3 & 2 \\ -1 & 2 & 1 \end{bmatrix} \in \mathcal{M}_{2 \times 3}(\mathbb{Q})$ and $B = \begin{bmatrix} -1 & 0 \\ 1 & 0 \\ 2 & 1 \end{bmatrix} \in \mathcal{M}_{3 \times 2}(\mathbb{Q})$ then $AB = \begin{bmatrix} 5 & 2 \\ 5 & 1 \end{bmatrix} \in \mathcal{M}_{2 \times 2}(\mathbb{Q})$ and $BA = \begin{bmatrix} -2 & -3 & -2 \\ 2 & 3 & 2 \\ 3 & 8 & 5 \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{Q})$.

Suppose that $A = [a_{ij}] \in \mathcal{M}_{k \times n}(F)$ and $v = \begin{bmatrix} b_1 \\ \vdots \\ b_n \end{bmatrix} \in F^n$. Then

$$Av = \begin{bmatrix} c_1 \\ \vdots \\ c_k \end{bmatrix} \in F^k \text{ where, for each } 1 \leq i \leq k, \text{ we have } c_i = \sum_{j=1}^n a_{ij} b_j.$$

Denoting the columns of A by u_1, \dots, u_n , we see that $Av = \sum_{j=1}^n b_j u_j$.

So we conclude that if there exists a nonzero vector v such that $Av =$

$\begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$, then the columns of A must be linearly dependent. If every

element of F^k is of the form Av for some $v \in F^n$, then the columns of A must form a generating set for F^k .

Let (K, \bullet) be an associative unital algebra over a field F and let n be a positive integer. If $v = \begin{bmatrix} v_1 \\ \vdots \\ v_n \end{bmatrix}$ and $w = \begin{bmatrix} w_1 \\ \vdots \\ w_n \end{bmatrix}$ are elements of

K^n then $v^T w = [\sum_{i=1}^n v_i \bullet w_i] \in \mathcal{M}_{1 \times 1}(K)$. This is called the **interior product** of v and w . This 1×1 matrix is usually identified with the scalar $\sum_{i=1}^n v_i \bullet w_i \in K$, which we will denote by $v \odot w$, in a departure from usual notation².

Dually, the **exterior product** of v and w is defined to be the matrix $vv^T = [y_{ij}] \in \mathcal{M}_{n \times n}(K)$, where $y_{ij} = v_i \bullet w_j$. We will denote the exterior product of v and w by $v \wedge w$. Notice that the exterior product is not commutative, but rather $v \wedge w = (w \wedge v)^T$. Exterior products of vectors are encountered far less often than interior products, but have important applications in many areas, among them physics (in the Dirac model of quantum physics, interior products are called **bra-ket products**, whereas exterior products are called **ket-bra products**).

In particular, we note the following: let K be an algebra over a field F , let $A \in \mathcal{M}_{k \times n}(K)$, and let $B \in \mathcal{M}_{n \times t}(K)$. Let v_1^T, \dots, v_k^T be the rows of A and let w_1, \dots, w_t be the columns of B . Then $AB = [c_{ij}]$, where $c_{ij} = v_i \odot w_j$ for all $1 \leq i \leq k$ and all $1 \leq j \leq t$.

²The usual notation is $v \cdot w$, but that can cause confusion with the dot product, which we will study later, in the case that $F = \mathbb{C}$. For that reason, also, we use the term “interior product” rather than the often-seen “inner product”.

Let F be a field, let n be a positive integer, let $v = \begin{bmatrix} a_1 \\ \vdots \\ a_n \end{bmatrix}$ and $w = \begin{bmatrix} ba_1 \\ \vdots \\ b_n \end{bmatrix}$ belong to F^n , and let $C = [c_{ij}] \in \mathcal{M}_{n \times n}(F)$. Then the computation of $v \odot Cw = \sum_{i=1}^n a_i \left(\sum_{j=1}^n c_{ij} b_j \right)$ requires $n^2 + n$ multiplications and $n^2 - 1$ additions. However, if we can find vectors $u = \begin{bmatrix} u_1 \\ \vdots \\ u_n \end{bmatrix}$ and $y = \begin{bmatrix} y_1 \\ \vdots \\ y_n \end{bmatrix}$ in F^n such that $C = u \wedge y$, then, by the distributive law, $v \odot Cw = \sum_{i=1}^n a_i \left(\sum_{j=1}^n u_i y_j b_j \right) = \left(\sum_{i=1}^n a_i u_i \right) \left(\sum_{j=1}^n y_j b_j \right)$ and this requires only $2n + 1$ multiplications and $2n - 2$ additions. Similarly, if we can find vectors $u, u', y, y' \in F^n$ such that $C = u \wedge y + u' \wedge y'$, then the computation of $v \odot Cw$ requires $4n + 2$ multiplications and $4n - 4$ additions. For large values of n , this can result in considerable saving, especially if the computation is to be repeated frequently.

Example: Combinatorial optimization is the area of mathematics dealing with the computational issues arising from finding optimal solutions to such problems as the traveling salesman problem, testing Hamiltonian graphs, sphere packing, etc. The general form of combinatorial optimization problems is the following: let F be a subfield of \mathbb{R} and let n be a positive integer. Assume that we have a nonempty finite (and in general very large) subset S of $\mathbb{N}^n \subseteq F^n$. Usually, the set S arises from the characteristic functions of certain subsets of $\{1, \dots, n\}$ of interest in

the problem. Then, given a vector $v = \begin{bmatrix} a_1 \\ \vdots \\ a_n \end{bmatrix} \in F^n$, we want to find

$\min \{s \odot v \mid s \in S\}$. Note that if we consider F not as a subset of \mathbb{R} but as a subset of the optimization algebra \mathbb{R}_∞ , then the problem becomes one of computing $p(a_1, \dots, a_n)$, where

$$p(X_1, \dots, X_n) = \sum \left\{ X_1^{i_1} \cdots X_n^{i_n} \left| \begin{bmatrix} i_1 \\ \vdots \\ i_n \end{bmatrix} \in S \right. \right\}$$

is a polynomial in several indeterminates over \mathbb{R}_∞ (polynomials with coefficients in a semifield are defined in the same way as polynomials with coefficients in a field).

Observe that multiplying a $k \times n$ matrix by an $n \times t$ matrix requires $kt(n-1)$ arithmetic operations. If these numbers are all very large, as is often the case in real-life applications of matrix theory, the computational overhead – and risk of accumulated errors due to rounding and truncation – is substantial.³ We will keep this in mind throughout our discussion, and try to consider strategies of minimizing this risk. In this connection, we should note that the product of two matrices has an important property: let (K, \bullet) be an associative unital algebra over a field F assume that $A = [v_{ij}] \in \mathcal{M}_{k \times n}(K)$ and $B = [w_{ij}] \in \mathcal{M}_{n \times t}(K)$, where k , n , and t are positive integers. Furthermore, let us pick positive integers

$$\begin{aligned} 1 &= k(1) < k(2) < \dots < k(p+1) = k \\ 1 &= n(1) < n(2) < \dots < n(q+1) = n \\ 1 &= t(1) < t(2) < \dots < t(r+1) = t. \end{aligned}$$

For all $1 \leq i \leq p$ and all $1 \leq j \leq q$, let

$$A_{ij} = \begin{bmatrix} v_{k(i),n(j)} & \dots & v_{k(i),n(j+1)} \\ \vdots & & \vdots \\ v_{k(i+1),n(j)} & \dots & v_{k(i+1),n(j+1)} \end{bmatrix}.$$

This allows us to write A in **block form** $\begin{bmatrix} A_{11} & \dots & A_{1q} \\ \vdots & & \vdots \\ A_{p1} & \dots & A_{pq} \end{bmatrix}$. Note

that these blocks are not necessarily square matrices. In the same way,

we can write B as a matrix $\begin{bmatrix} B_{11} & \dots & B_{1t} \\ \vdots & & \vdots \\ B_{q1} & \dots & B_{qt} \end{bmatrix}$. Then $AB =$

$$\begin{bmatrix} C_{11} & \dots & C_{1t} \\ \vdots & & \vdots \\ C_{p1} & \dots & C_{pt} \end{bmatrix} \quad \text{where, for each } 1 \leq i \leq p \text{ and each } 1 \leq h \leq t,$$

we have $C_{ih} = \sum_{j=1}^q A_{ij}B_{jh}$. A sophisticated use of this method can

³We will often mention large matrices, without being too specific as to what that means. As a rule of thumb, a matrix is “large”, and calls for special treatment as such, when it cannot be stored in the RAM memory of whatever computer we are using for our computations. Such matrices occur in sufficiently-many applications that considerable research is devoted to dealing with them.

substantially decrease the number of operations needed to multiply two matrices, as we shall see.

Needless to say, this seemingly odd definition of the product of two matrices was not chosen at random⁴. Indeed, it satisfies certain important properties. Thus, if (K, \bullet) is an associative unital algebra over a field F , if k , n , t , and p are positive integers, and if we have matrices $A \in \mathcal{M}_{k \times n}(K)$, $B, B_1, B_2 \in \mathcal{M}_{n \times t}(K)$, and $C \in \mathcal{M}_{t \times p}(K)$, then

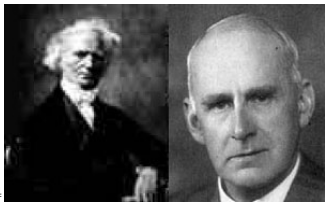
- (1) $A(BC) = (AB)C$;
- (2) $A(B_1 + B_2) = AB_1 + AB_2$;
- (3) $(B_1 + B_2)C = B_1C + B_2C$.

As a consequence, we see that if $B \in \mathcal{M}_{n \times t}(K)$ is given, then the function from $\mathcal{M}_{k \times n}(K)$ to $\mathcal{M}_{k \times t}(K)$ defined by $A \mapsto AB$ is a linear transformation of vector spaces.

We also note that if $A = [v_{ij}] \in \mathcal{M}_{k \times n}(K)$ and $B = [w_{jh}] \in \mathcal{M}_{n \times t}(K)$, then $A^T \in \mathcal{M}_{n \times k}(K)$ and $B^T \in \mathcal{M}_{t \times n}(K)$ so $B^T A^T \in \mathcal{M}_{t \times k}(K)$. Indeed, $B^T A^T = [y_{hi}]$, where $y_{hi} = \sum_{j=1}^n w_{jh} \bullet v_{ij}$. Hence, if K is also commutative (and in particular if $K = F$), we have $B^T A^T = (AB)^T$.

The definition of matrix multiplication is in fact a direct consequence of the relation between matrices and linear transformations, which we have already observed. This is best seen in the following result.

(8.2) Proposition: Let V be a vector space of finite dimension n over a field F for which we have chosen a basis $B = \{v_1, \dots, v_n\}$, let W be a vector space of finite dimension k over F for which we have chosen a basis $D = \{w_1, \dots, w_k\}$, and let Y be a vector space of finite dimension t over F , for which we have chosen a basis $E = \{y_1, \dots, y_t\}$. If $\alpha \in \text{Hom}(V, W)$ and $\beta \in \text{Hom}(W, Y)$ then $\Phi_{BE}(\beta\alpha) = \Phi_{DE}(\beta)\Phi_{BD}(\alpha)$.



⁴ Matrix multiplication was first defined by the 19th-century French mathematician **Jacques Philippe Binet**. It took some getting used to; many decades later, the father of astrophysics, **Sir Arthur Eddington**, still wrote “I cannot believe that anything so ugly as multiplication of matrices is an essential part of the scheme of nature”.

Proof: Assume that $\Phi_{BD}(\alpha) = [a_{ij}]$ and $\Phi_{DE}(\beta) = [b_{hi}]$. Then

$$\alpha : v \mapsto \sum_{i=1}^k \sum_{j=1}^n c_j a_{ij} w_i \quad \text{and} \quad \beta \alpha : v \mapsto \sum_{h=1}^t \sum_{i=1}^k \sum_{j=1}^n c_j b_{hi} a_{ij} y_h,$$

showing the desired equality. \square

We can extend the definition of matrix multiplication as follows: let h , k , and n be positive integers and let V be a vector space over a field F . If $A = [a_{ij}] \in \mathcal{M}_{h \times k}(F)$ and if $M = [v_{jt}] \in \mathcal{M}_{k \times n}(V)$, we can define $AM \in \mathcal{M}_{h \times n}(V)$ to be the matrix $[u_{it}]$, where $u_{it} = \sum_{j=1}^k a_{ij} v_{jt}$ for all $1 \leq i \leq h$ and $1 \leq t \leq n$. Notice that if $A, B \in \mathcal{M}_{k \times k}(F)$ and if $M, N \in \mathcal{M}_{k \times n}(V)$ then

- (1) $A(BM) = (AB)M$;
- (2) $A(M + N) = AM + AN$;
- (3) $(A + B)M = AM + BM$.

In general, and especially when we are talking of actual computations, it is easier to work with matrices than with linear transformations, and indeed most of the modern computer software and hardware are designed to facilitate easy and speedy matrix computation. Therefore, given finitely-generated vector spaces V and W over a field F , it is usual to fix bases for them and then identify $\text{Hom}(V, W)$ with the space of all matrices over F of the appropriate size. The choice of the correct bases then becomes critical, and we will focus on that throughout the following discussions. Such a choice usually depends on the problem at hand. In particular, the automatic choice of canonical bases, when they exist, may not be the best for a given problem, and can entail a considerable cost both in computational time and numerical accuracy.

Exercises

Exercise 356 Let V be the vector space over \mathbb{R} composed of all polynomials in $\mathbb{R}[X]$ having degree less than 3 and let W be the vector space over \mathbb{R} composed of all polynomials in $\mathbb{R}[X]$ having degree less than 4. Let $\alpha : V \rightarrow W$ be the linear transformation defined by

$$\alpha : a + bX + cX^2 \mapsto (a + b) + (b + c)X + (a + c)X^2 + (a + b + c)X^3.$$

Select bases $B = \{1, X + 1, X^2 + X + 1\}$ for V and

$$D = \{X^3 - X^2, X^2 - X, X - 1, 1\}$$

for W . Find the matrix $\Phi_{BD}(\alpha)$.

Exercise 357 Let $K = \mathcal{M}_{2 \times 2}(\mathbb{R})$ and let $A = \begin{bmatrix} a & b \\ c & d \end{bmatrix} \in K$. Let D be the canonical basis of K . If $\alpha, \beta \in \text{End}(K)$ are defined by $\alpha : X \mapsto XA$ and $\beta : X \mapsto AX$, find $\Phi_{DD}(\alpha)$ and $\Phi_{DD}(\beta)$.

Exercise 358 Given the matrix $A = \begin{bmatrix} 0 & 1 & 2 & 3 \\ 1 & 3 & 4 & 0 \\ 3 & 2 & 0 & 1 \end{bmatrix} \in \mathcal{M}_{3 \times 4}(\mathbb{R})$, find

the set of all matrices $B \in \mathcal{M}_{4 \times 3}(\mathbb{R})$ satisfying $AB = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$.

Exercise 359 Given the matrix $A = \begin{bmatrix} 1 & 8 \\ 3 & 5 \\ 2 & 2 \end{bmatrix} \in \mathcal{M}_{3 \times 2}(\mathbb{Q})$, find the set

of all matrices $B \in \mathcal{M}_{2 \times 3}(\mathbb{Q})$ satisfying $AB = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$ and find

the set of all matrices $C \in \mathcal{M}_{2 \times 3}(\mathbb{Q})$ satisfying $CA = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$.

Exercise 360 Given the matrix $A = \begin{bmatrix} 1 & -1 & 2 \\ 0 & 1 & 2 \\ -1 & 3 & 1 \\ 2 & 1 & -1 \end{bmatrix} \in \mathcal{M}_{4 \times 3}(\mathbb{R})$,

find the set of all matrices $B \in \mathcal{M}_{3 \times 4}(\mathbb{R})$ satisfying $BA = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$.

Exercise 361 Find the matrix representing the linear transformation

$\alpha : \begin{bmatrix} a \\ b \\ c \end{bmatrix} \mapsto \begin{bmatrix} a+b+c \\ b+c \end{bmatrix}$ from \mathbb{R}^3 to \mathbb{R}^2 with respect to the bases

$\left\{ \begin{bmatrix} -1 \\ 0 \\ 2 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 3 \\ -1 \\ 0 \end{bmatrix} \right\}$ of \mathbb{R}^3 and $\left\{ \begin{bmatrix} -1 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \end{bmatrix} \right\}$ of \mathbb{R}^2 .

Exercise 362 Find the set of all matrices $A \in \mathcal{M}_{4 \times 3}(\mathbb{R})$ satisfying the

condition $\begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix} A = A \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{bmatrix}$.

Exercise 363 Let α be an endomorphism of \mathbb{R}^3 represented with respect

to some basis by the matrix $\begin{bmatrix} 0 & 2 & -1 \\ -2 & 5 & -2 \\ -4 & 8 & -3 \end{bmatrix}$. Is α a projection?

Exercise 364 Find the real numbers missing from the following equation:

$$\begin{bmatrix} 1 & -3 \\ 1 & * \\ 1 & * \\ * & 1 \end{bmatrix} \begin{bmatrix} -1 & * & 7 & * \\ * & 1 & * & 0 \end{bmatrix} = \begin{bmatrix} -25 & -1 & 1 & 3 \\ -1 & * & * & * \\ * & * & 5 & * \\ * & * & * & 0 \end{bmatrix}.$$

Exercise 365 Let $\alpha : \begin{bmatrix} a \\ b \\ c \end{bmatrix} \mapsto \begin{bmatrix} 3a + 2b \\ -a - c \\ a + 3b \end{bmatrix}$ be an endomorphism of

\mathbb{R}^3 . Find the matrix representing α with respect to the basis $B = \left\{ \begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} \right\}$ of \mathbb{R}^3 .

Exercise 366 Let $D = \{v_1, v_2, v_3\}$ be a basis for \mathbb{R}^3 and let α be

the endomorphism of \mathbb{R}^3 satisfying $\Phi_{DD}(\alpha) = \begin{bmatrix} -1 & -1 & -3 \\ -5 & -2 & -6 \\ 2 & 1 & 3 \end{bmatrix}$. Find $\ker(\alpha)$.

Exercise 367 Let V be the subspace of $\mathbb{R}[X]$ consisting of all polynomials of degree less than 3 and choose the basis $B = \{1, X, X^2\}$ for V .

Let $\alpha \in \text{End}(V)$ satisfy $\Phi_{BB}(\alpha) = \begin{bmatrix} 1 & 1 & 1 \\ 0 & 2 & 2 \\ 0 & 0 & 3 \end{bmatrix}$. Let D be the basis $\{1, X + 1, 2X^2 + 4X + 3\}$ for V . What is $\Phi_{DD}(\alpha)$?

Exercise 368 Let $\alpha \in \text{End}(\mathbb{R}^3)$ be represented with respect to the canon-

ical basis by the matrix $\begin{bmatrix} 2 & 2 & 0 \\ 1 & 1 & 2 \\ 1 & 1 & 2 \end{bmatrix}$. Find a real number a such that

α is represented with respect to the basis $\left\{ \begin{bmatrix} a \\ -1 \\ 0 \end{bmatrix}, \begin{bmatrix} -2 \\ a \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ a \end{bmatrix} \right\}$

by the matrix $\begin{bmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 4 \end{bmatrix}$.

Exercise 369 Let V be the subspace of $\mathbb{R}[X]$ consisting of all polynomials of degree less than 3 and let $\alpha \in \text{End}(V)$ be defined by

$$\alpha : aX^2 + bX + c \mapsto (a + 2b + c)X^2 + (3a - b)X + (b + 2c).$$

Find $\Phi_{DD}(\alpha)$, where $D = \{X^2 + X + 1, X^2 + X, X^2\}$.

Exercise 370 Find all rational numbers a for which there exists a nonzero

matrix $B \in \mathcal{M}_{4 \times 3}(\mathbb{Q})$ satisfying $B \begin{bmatrix} a & 1 & 1 \\ 1 & 1 & a \\ 1 & a & 1 \end{bmatrix} = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}.$

Exercise 371 For which real numbers a does there exist a real number

b satisfying $\begin{bmatrix} a & 1 & 1 \\ 1 & 1 & a \end{bmatrix} \begin{bmatrix} b & -1 \\ 1 & -1 \\ 1 & b \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}?$

Exercise 372 Let $V = \mathbb{R}^{\mathbb{R}}$ and let W be the subspace of V generated by the linearly-independent set $B = \{1, x, e^x, xe^x\}$. Let δ be the endomorphism of W which assigns to each function its derivative. Find $\Phi_{BB}(\delta)$.

Exercise 373 Let $B = \{1 + i, 2 + i\}$, which is a basis for \mathbb{C} as a vector space over \mathbb{R} . Let α be the endomorphism of this space defined by $\alpha : z \mapsto \bar{z}$. Find $\Phi_{BB}(\alpha)$.

Exercise 374 Let $F = GF(3)$ and let $\alpha : F^3 \rightarrow F^2$ be the linear

transformation defined by $\alpha : \begin{bmatrix} a \\ b \\ c \end{bmatrix} \mapsto \begin{bmatrix} a - b \\ 2a - c \end{bmatrix}$. Let $\beta : F^2 \rightarrow F^4$

be the linear transformation defined by $\beta : \begin{bmatrix} a \\ b \end{bmatrix} \mapsto \begin{bmatrix} b \\ a \\ 2b \\ 2a \end{bmatrix}$. Find the

matrix representing $\beta\alpha$ with respect to the canonical bases.

Exercise 375 Let $\alpha \in \text{End}(\mathbb{R}^4)$ be represented with respect to the canon-

ical basis by the matrix $\begin{bmatrix} 3 & -1 & 0 & 0 \\ -1 & 2 & -1 & 0 \\ 0 & -1 & 2 & -1 \\ 0 & 0 & -1 & 1 \end{bmatrix}$. Given a vector $v \in \mathbb{R}^4$

satisfying the condition that all entries of $\alpha(v)$ are nonnegative, show that all entries of v are nonnegative.

Exercise 376 Let V and W be vector spaces over a field F and choose bases $\{v_i \mid i \in \Omega\}$ and $\{w_j \mid j \in \Lambda\}$ for V and W respectively. Let $p : \Omega \times \Lambda \rightarrow F$ be a function satisfying the condition that the set $\{j \in \Lambda \mid p(i, j) \neq 0\}$ is finite for each $i \in \Omega$. Let $\alpha_p : V \rightarrow W$ be the function defined as follows: if $v = \sum_{i \in \Gamma} a_i v_i$, where Γ is a finite subset of Ω and where the a_i are scalars in F , then $\alpha_p(v) = \sum_{i \in \Gamma} \sum_{j \in \Lambda} a_i p(i, j) w_j$. Show that α_p is a linear transformation and that every linear transformation from V to W is of this form.

Exercise 377 Let k and n be positive integers, let $v \in \mathbb{R}^n$, and let $A \in \mathcal{M}_{k \times n}(\mathbb{R})$. Show that $Av = 0$ if and only if $A^T Av = 0$.

Exercise 378 Let $A \in \mathcal{M}_{3 \times 2}(\mathbb{R})$ and $B \in \mathcal{M}_{2 \times 3}(\mathbb{R})$ be matrices satis-

$$\text{fying } AB = \begin{bmatrix} 8 & 2 & -2 \\ 2 & 5 & 4 \\ -2 & 4 & 5 \end{bmatrix}. \text{ Calculate } BA.$$

Exercise 379 Find matrices $A \in \mathcal{M}_{3 \times 2}(\mathbb{R})$ and $B \in \mathcal{M}_{2 \times 3}(\mathbb{R})$ satis-

$$\text{fying } AB = \begin{bmatrix} 1 & 1 & 1 \\ -2 & 0 & -6 \\ 0 & 1 & -2 \end{bmatrix}.$$

Exercise 380 Let F be a field and let n be a positive integer. Let W be a nontrivial subspace of the vector space $V = \mathcal{M}_{n \times n}(F)$ satisfying the condition that if $A \in W$ and $B \in V$ then AB and BA both belong to W . Show that $W = V$.

Exercise 381 Let $a, b, c, a', b', c' \in \mathbb{C}$ satisfy the condition that $aa' + bb' +$

$$cc' = 2, \text{ and let } A = I - \begin{bmatrix} a \\ b \\ c \end{bmatrix} \begin{bmatrix} a' & b' & c' \end{bmatrix}. \text{ Calculate } A^2.$$

Exercise 382 Find a nonzero matrix A in $\mathcal{M}_{2 \times 2}(\mathbb{R})$ satisfying $v \odot Av = 0$ for all $v \in \mathbb{R}^2$.

Exercise 383 Let α be the endomorphism of \mathbb{Q}^4 represented with respect

$$\text{to the canonical basis by the matrix } \begin{bmatrix} 1 & 0 & 1 & -1 \\ 2 & 1 & 2 & 1 \\ 0 & 1 & 6 & 1 \\ 3 & 1 & 3 & 4 \end{bmatrix}. \text{ Find a two-}$$

dimensional subspace of \mathbb{Q}^4 which is invariant under α .

Exercise 384 Find the set of all matrices A in $\mathcal{M}_{3 \times 3}(\mathbb{R})$ satisfying

$$A^2 = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}.$$

Exercise 385 Find the set of all real numbers a such that the endomorphism of \mathbb{R}^3 represented by the matrix
$$\begin{bmatrix} 1 & a & a \\ 2 & 2a & 4 \\ 3 & a & 6 \end{bmatrix}$$
 with respect to the canonical basis is an automorphism.

Exercise 386 Find the set of all real numbers a and b such that the endomorphism of \mathbb{R}^3 represented by the matrix
$$\begin{bmatrix} 1 & a & b \\ 0 & a & 1 \\ 0 & a & 1 \end{bmatrix}$$
 with respect to the canonical basis is a projection.

Exercise 387 Let n be a positive integer and let F be a field. For each $v, w \in F^n$, consider the function $\tau_{v,w} : F^n \rightarrow Fv$ defined by $\tau_{v,w} : y \mapsto (w \odot y)v$ (this function is called the **dyadic product** function). Show that $\tau_{v,w}$ is a linear transformation. Is the function $F^n \rightarrow \text{Hom}(V, Fv)$ defined $w \mapsto \tau_{v,w}$ a linear transformation?

Exercise 388 Let $k < n$ be positive integers and let F be a field. Given a matrix $A \in \mathcal{M}_{k \times n}(F)$, do there necessarily exist matrices $B, C \in \mathcal{M}_{n \times k}(F)$ satisfying the condition that $AB = O \in \mathcal{M}_{k \times k}(F)$ and $CA = O \in \mathcal{M}_{n \times n}(F)$?

Exercise 389 Let $A \in \mathcal{M}_{n \times n}(\mathbb{Q})$ be a matrix satisfying the condition that if $v \in \mathbb{Q}^n$ is a vector all of the components of which are nonnegative, then all of the components of Av are nonnegative. Are all of the entries in A necessarily nonnegative?

9

The algebra of square matrices

We are now going to concentrate on the algebraic structure of sets of the form $\mathcal{M}_{n \times n}(K)$, where n is a positive integer and (K, \bullet) is an associative unital algebra over a field F . From what we have already seen, this is again an associative unital F -algebra, which will not be commutative if $n > 1$. The additive identity of this algebra is the matrix all of the entries of which equal 0_K . The additive inverse of a matrix $A = [d_{ij}] \in \mathcal{M}_{n \times n}(K)$ is the matrix $[-a_{ij}]$. The multiplicative identity of $\mathcal{M}_{n \times n}(K)$ is the matrix $E = [d_{ij}]$ given by

$$d_{ij} = \begin{cases} e & \text{if } i = j \\ 0 & \text{otherwise} \end{cases}$$

where e is the multiplicative identity of (K, \bullet) .

The most important case is, of course, that of $K = F$. In this case, the additive identity is O and the multiplicative identity is the matrix $I = [a_{ij}]$ defined by

$$a_{ij} = \begin{cases} 1 & \text{if } i = j \\ 0 & \text{otherwise} \end{cases}.$$

If K is a vector space of dimension n over F and if B is a basis of K , then it is straightforward to verify that the function $\Phi_{BB} : \text{End}(K) \rightarrow \mathcal{M}_{n \times n}(F)$ is an isomorphism of unital F -algebras.

If F is a field and if n is a positive integer then, corresponding to the associative F -algebra $\mathcal{M}_{n \times n}(F)$, we have the Lie algebra $\mathcal{M}_{n \times n}(F)^-$. This Lie algebra is called the **general Lie algebra** defined by F^n .

Example: Let F be a field and let $A = [a_{ij}] \in \mathcal{M}_{4 \times 4}(F)$. Then A can also be written in block form as

$$\begin{bmatrix} \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} & \begin{bmatrix} a_{13} & a_{14} \\ a_{23} & a_{24} \end{bmatrix} \\ \begin{bmatrix} a_{31} & a_{32} \\ a_{41} & a_{42} \end{bmatrix} & \begin{bmatrix} a_{33} & a_{34} \\ a_{43} & a_{44} \end{bmatrix} \end{bmatrix} \in \mathcal{M}_{2 \times 2}(K),$$

where $K = \mathcal{M}_{2 \times 2}(F)$. Addition and multiplication of matrices are so defined (and not accidentally!) so that they give the same results whether performed in $\mathcal{M}_{4 \times 4}(F)$ or in $\mathcal{M}_{2 \times 2}(K)$.

We begin by identifying some particularly-important square matrices over a unital associative F -algebra K , and with them some significant subalgebras of $\mathcal{M}_{n \times n}(K)$.

Let (K, \bullet) is an associative unital F -algebra and let n be a positive integer. A matrix $A = [d_{ij}] \in \mathcal{M}_{n \times n}(K)$ is a **diagonal matrix** if and only if there exist elements c_1, \dots, c_n of K such that

$$d_{ij} = \begin{cases} c_i & \text{if } i = j \\ 0_K & \text{otherwise} \end{cases}.$$

The matrices O and E are diagonal. Moreover, the sum and product of diagonal matrices are diagonal matrices, and so the set of all diagonal matrices is an F -subalgebra of $\mathcal{M}_{n \times n}(K)$. If K is commutative (and, in particular, if $K = F$) then this algebra is also commutative. The units of the subalgebra are all diagonal matrices in which each c_i is a unit

of K (and hence surely nonzero). In this case,

$$\begin{bmatrix} c_1 & \dots & 0_K \\ \vdots & \ddots & \vdots \\ 0_K & \dots & c_n \end{bmatrix}^{-1} = \begin{bmatrix} c_1^{-1} & \dots & 0_K \\ \vdots & \ddots & \vdots \\ 0_K & \dots & c_n^{-1} \end{bmatrix}.$$

Example: Let F be a field, let (K, \bullet) is an associative unital F -algebra, and let n be a positive integer. A matrix $A = [a_{ij}] \in \mathcal{M}_{n \times n}(K)$ is a **scalar matrix** if and only if there exists a scalar $c \in K$ such that $a_{ij} = c$ when $i = j$ and $a_{ij} = 0_K$ otherwise. We denote this matrix by cE (and, in particular, cI when $K = F$). Scalar matrices are surely diagonal matrices, and both O and E are scalar matrices. Moreover, the sum and product of scalar matrices are scalar matrices. If $c, d \in K$ then $(cE)(dE) = (dE)(cE)$ and if $0_K \neq c \in K$ is a unit, then $(cE)(c^{-1}E) = E$.

Hence the set of all scalar matrices over F forms an F -subalgebra of $\mathcal{M}_{n \times n}(F)$, which is in fact a field. The function $F \rightarrow \mathcal{M}_{n \times n}(F)$ defined by $c \mapsto cI$ is a monic homomorphism of F -algebras, and so we can identify F with the subfield of all scalar matrices of $\mathcal{M}_{n \times n}(F)$. Moreover, it is also easy to see that $(cI)A = A(cI) = cA$ for any $A \in \mathcal{M}_{n \times n}(F)$.

Let (K, \bullet) is an associative unital F -algebra, let n be a positive integer, and let d be a positive integer less than n . A matrix $A = [v_{ij}] \in \mathcal{M}_{n \times n}(K)$ is a **band matrix of width** $2d - 1$ if and only if $a_{ij} = 0_K$ whenever $|i - j| > d - 1$. Thus, the band matrices of width 1

are the diagonal matrices. The matrix

$$\begin{bmatrix} 1 & 2 & 0 & 0 & 0 \\ 2 & 3 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 4 & 4 \end{bmatrix} \in \mathcal{M}_{5 \times 5}(\mathbb{R})$$

is an example of a band matrix of width 3. The set of band matrices of fixed width is closed under addition and contains O and I , but is not necessarily closed under multiplication, and so is not a subalgebra of $\mathcal{M}_{n \times n}(K)$. However, it is closed under scalar multiplication and so is a subspace of the vector space $\mathcal{M}_{n \times n}(K)$ over F .

Band matrices over F are very important for numerical computations, especially when d is small relative to n . Of particular importance are band matrices of width 3, which are also known as **tridiagonal matrices**, and have important use in the computation of quadratic splines and in the computation of extremal eigenvalues of matrices.

A special type of tridiagonal matrix in $\mathcal{M}_{2n \times 2n}(F)$, which we will see

again later, is one of the form

$$\begin{bmatrix} A_{11} & O & \dots & O \\ O & A_{22} & \dots & O \\ & & \ddots & \\ O & \dots & O & A_{nn} \end{bmatrix}, \text{ where the } A_{ii}$$

are 2×2 blocks. Note that this matrix can also be thought of as a diagonal matrix in $\mathcal{M}_{n \times n}(K)$, where $K = \mathcal{M}_{2 \times 2}(F)$. More generally, if d and n are positive integers, then any diagonal matrix in $\mathcal{M}_{n \times n}(L)$, where $L = \mathcal{M}_{d \times d}(K)$, is a band matrix of width $2d - 1$ in $\mathcal{M}_{dn \times dn}(K)$.

Let (K, \bullet) is an associative unital F -algebra and let n be a positive integer. A matrix $A = [c_{ij}] \in \mathcal{M}_{n \times n}(K)$ is an **upper-triangular matrix** if and only if $c_{ij} = 0_K$ whenever $i > j$. Thus, the matrix

$$\begin{bmatrix} 1 & 2 & 6 & 3 & 7 \\ 0 & 3 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 4 \end{bmatrix} \in \mathcal{M}_{5 \times 5}(\mathbb{R})$$

is upper triangular. The set of all upper-

triangular matrices includes the set of diagonal matrices, is closed under addition, and contains O and E . Moreover, it is closed under multiplication, and so is an F -subalgebra of $\mathcal{M}_{n \times n}(K)$. In the case that $K = F$, we see that the dimension of $\mathcal{M}_{n \times n}(F)$ as a vector space over F equals $\frac{n}{2}(n+1)$. Upper-triangular matrices arise naturally in many applications, as we will see below. In a similar manner, we say that a matrix $A = [c_{ij}] \in \mathcal{M}_{n \times n}(K)$ is a **lower-triangular matrix** if and only if $c_{ij} = 0_K$ whenever $i < j$. Again, the set of all lower-triangular matrices is a subspace of the vector space $\mathcal{M}_{n \times n}(K)$ over F and, indeed, an F -subalgebra. Note that a matrix A is upper triangular if and only if A^T is lower triangular.

A matrix $A = [c_{ij}] \in \mathcal{M}_{n \times n}(K)$ is **symmetric** if and only if $A = A^T$. That is, A is symmetric if and only if $c_{ij} = c_{ji}$ for all $1 \leq i, j \leq n$. If B is any matrix in $\mathcal{M}_{n \times n}(K)$ then $B + B^T$ is symmetric. If K is commutative and if $C \in \mathcal{M}_{k \times n}(K)$ for any positive integers k and n , then $CC^T \in \mathcal{M}_{k \times k}(K)$ and $C^TC \in \mathcal{M}_{n \times n}(K)$ are symmetric. If n is a positive integer and F is a field, then $v \wedge v$ is a symmetric matrix in $\mathcal{M}_{n \times n}(F)$ for all $v \in F^n$. Diagonal matrices are clearly symmetric and the set of symmetric matrices in $\mathcal{M}_{n \times n}(K)$ is closed under taking sums and scalar multiples, and so it is a subspace of the vector space $\mathcal{M}_{n \times n}(K)$ over F . In the case $K = F$, the dimension of $\mathcal{M}_{n \times n}(F)$ equals $\frac{n}{2}(n+1)$. However, the set of symmetric matrices is not closed under products. For

example, the matrices $A = \begin{bmatrix} 2 & 5 & 1 \\ 5 & 2 & 0 \\ 1 & 0 & 1 \end{bmatrix}$ and $B = \begin{bmatrix} 1 & 2 & 1 \\ 2 & 0 & 0 \\ 1 & 0 & 3 \end{bmatrix}$ in

$\mathcal{M}_{3 \times 3}(\mathbb{R})$ are symmetric but $AB = \begin{bmatrix} 13 & 4 & 5 \\ 9 & 10 & 5 \\ 2 & 2 & 4 \end{bmatrix}$ is not. Nonetheless,

if A and B are a commuting pair of symmetric matrices then $(AB)^T = (BA)^T = A^TB^T = AB$, so AB is again symmetric.

Similarly, a matrix $A = [c_{ij}] \in \mathcal{M}_{n \times n}(K)$ is **skew symmetric** if and only if $A = -A^T$. The set of all skew-symmetric matrices in $\mathcal{M}_{n \times n}(K)$ is again a subspace of $\mathcal{M}_{n \times n}(K)$. Note that if F is a field having characteristic other than 2, then any matrix $A \in \mathcal{M}_{n \times n}(K)$ can be written as the sum of a symmetric matrix and a skew-symmetric matrix, since $A = \frac{1}{2}(A + A^T) + \frac{1}{2}(A - A^T)$. In one of the examples after Proposition 5.14, we saw that this representation is in fact unique. The Lie product of two skew-symmetric matrices is again skew-symmetric.

Example: Let n be a positive integer. A matrix $A = [a_{ij}] \in \mathcal{M}_{n \times n}(\mathbb{R})$ is a **Markov**¹ **matrix** if and only if $a_{ij} \geq 0$ for all $1 \leq i, j \leq n$ and $\sum_{j=1}^n a_{hj} = 1$ for each $1 \leq h \leq n$; it is a **stochastic matrix** if and only if both A and A^T are Markov matrices. It is easy to show that the product of two Markov matrices is again a Markov matrix and the product of two stochastic matrices in $\mathcal{M}_{n \times n}(\mathbb{R})$ is again a stochastic matrix.

Markov matrices arise naturally in probability theory. In particular, if we have a system which, at each tick of a (discrete) clock, is in one of the distinct states s_1, \dots, s_n and if, for each $1 \leq i, j \leq n$, we denote by p_{ij} the probability that if the situation is in state i at a given time t then it will be in state j at time $t+1$, the matrix $[p_{ij}]$ is a Markov matrix.

As we have already pointed out, a matrix $O \neq A \in \mathcal{M}_{n \times n}(F)$ is not necessarily a unit. The units of the F -algebra of square matrices are known as **nonsingular** matrices; the other matrices are **singular**² matrices. By what we have already noted, the product of nonsingular matrices is again nonsingular. Let V be a vector space over F of dimension n and let B be a basis of V . Then there exists an endomorphism α of V such that $A = \Phi_{BB}(\alpha)$. If A is a unit then there also exists an endomorphism β of V satisfying $A^{-1} = \Phi_{BB}(\beta)$. This means that $I = AA^{-1} = \Phi_{BB}(\alpha)\Phi_{BB}(\beta) = \Phi_{BB}(\alpha\beta)$ and so $\alpha\beta = \sigma_1$, and similarly $\beta\alpha = \sigma_1$. Therefore $\alpha \in \text{Aut}(V)$ and $\beta = \alpha^{-1}$.

Example: Let F be a field and let n be a positive integer. If $c \in F$ and $v, w \in F^n$, then the matrix $A = I + c(v \wedge w)$ is nonsingular if and only if the scalar $1 + c(v \odot w)$ is nonzero. Indeed, direct computation shows that if $1 + c(v \odot w) \neq 0$, then $A^{-1} = I + d(v \wedge w)$, where



¹ Russian mathematician **Andrei Markov** made major contributions to probability theory at the beginning of the twentieth century.



² These terms were first used by American mathematician **Maxime Bôcher** in 1907. He was also the first to popularize the terms “linearly dependent” and “linearly independent”.

$d = -c[1 + c(v \odot w)]^{-1}$ and if $1 + c(v \odot w) = 0$ then $Av = v + c(v \odot w)v$ is the 0-vector, and so A must be singular.

Example: The multiplicative inverse of a “nice” nonsingular matrix may not be “nice”. Thus, if A is a nonsingular matrix all of the entries of which are nonnegative, it does not follow that all of the entries of A^{-1} are

nonnegative. For example, if we choose $A = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 2 & 1 \\ 1 & 1 & 2 \end{bmatrix}$ then direct

computation shows us that $A^{-1} = \begin{bmatrix} 3 & -1 & -1 \\ -1 & 1 & 0 \\ -1 & 0 & 1 \end{bmatrix}$. If $A = [a_{ij}]$ is the $n \times n$ tridiagonal matrix with $a_{ii} = 2$ for all $1 \leq i \leq n$ and $a_{ij} = -1$ whenever $|i - j| = 1$, then not only is A^{-1} not tridiagonal, but in fact no entries in A^{-1} equal 0, for any $n > 1$.

Example: Let n be a prime positive integer. The complex number $c_n = \cos\left(\frac{2\pi}{n}\right) + i \sin\left(\frac{2\pi}{n}\right)$ is called a **primitive root of unity** of degree n , since it is easy to check that $c_n^n = 1$ but $c_n^i \neq 1$ for all $0 < i < n$. Therefore, $c_n^{-1} = c_n^{n-1}$ for all n . For each $z \in \mathbb{C}$, let $F(z) \in \mathcal{M}_{n \times n}(\mathbb{C})$ be the matrix $[a_{ij}]$ defined by $a_{ij} = z^{(i-1)(j-1)}$ for all $1 \leq i, j \leq n$. It is straightforward to show that the matrix $F(c_n)$ is nonsingular and, indeed, $F(c_n)^{-1} = \frac{1}{n}F(c_n^{-1})$. The endomorphism φ_n of \mathbb{C}^n which is represented with respect to the canonical basis by the matrix $F(c_n)$ is called a **discrete Fourier transform** of \mathbb{C}^n . This endomorphism is of great importance in applied mathematics. An algorithm, known as the **fast Fourier transform (FFT)**, introduced by J. W. Cooley and John W. Tukey³ in 1965, allows one to calculate $\varphi_n(v)$ in an order of $n \log(n)$ arithmetic operations, rather than n^2 , as one would anticipate. This facilitates the use of Fourier transforms in applications. A similar construction is also possible over finite fields, and especially over fields of the form $GF(p)$. We will look at this example again in Chapter 15.



3

Joseph Fourier was a close friend of Napoleon and served for many years as permanent secretary of the Parisian Academy of Sciences. He worked primarily in applied mathematics, and developed many important tools in this area. **John W. Tukey** was a twentieth-century American statistician who developed many advanced mathematical tools in statistics.

A closely-related endomorphism, the **discrete cosine transform** is used in defining the JPEG algorithm for image compression.

Example: The subset T of $\mathcal{M}_{2 \times 2}(\mathbb{R})$ consisting of all matrices of the form $\begin{bmatrix} a & b \\ -b & a \end{bmatrix}$ is easily checked to be an \mathbb{R} -subalgebra of $\mathcal{M}_{2 \times 2}(\mathbb{R})$. Moreover, the function $\gamma : \mathbb{C} \rightarrow T$ given by $\gamma : a + bi \mapsto \begin{bmatrix} a & b \\ -b & a \end{bmatrix}$ is an \mathbb{R} -algebra isomorphism.

Let K be an associative unital algebra over a field F having multiplicative identity e , and let n be a positive integer. Let E be the multiplicative identity of $\mathcal{M}_{n \times n}(K)$. A matrix $A = [c_{ij}] \in \mathcal{M}_{n \times n}(K)$ is an **elementary matrix** if and only if it is of one of the following forms:

- (1) E_{hk} , the matrix formed from E by interchanging the h th and k th columns, where $h \neq k$;
- (2) $E_{h;c}$, the matrix formed from E by multiplying the h th column by $0_K \neq c \in K$;
- (3) $E_{hk;c}$, the matrix formed from E by adding c times the k th column to the h th column, where $h \neq k$, where $c \in K$.

It is easy to verify that matrices of the form E_{hk} and $E_{hk;c}$ are always nonsingular, with $E_{hk}^{-1} = E_{hk}$ and $E_{kh;c}^{-1} = E_{hk;-c}$. If c is a unit in K , then matrices of the form $E_{h;c}$ are nonsingular, with $E_{h;c}^{-1} = E_{h;c^{-1}}$. We note that the transpose of an elementary matrix is again an elementary matrix. Indeed, $E_{hk}^T = E_{hk}$ and $E_{h;c}^T = E_{h;c}$ for all $1 \leq h, k \leq n$ and $0_K \neq c \in K$, while $E_{hk;c}^T = E_{kh;c}$ for all $1 \leq h \neq k \leq n$ and all $c \in K$.

As the name clearly implies, there is a connection between the elementary automorphisms which we defined previously and the elementary matrices. Indeed, if $K = F$ and if B is the canonical basis of F^n , then $E_{hk} = \Phi(\varepsilon_{hk})$, $E_{h;c} = \Phi(\varepsilon_{h;c})$, and $E_{hk;c} = \Phi(\varepsilon_{hk;c})$.

Let us see what happens when one multiplies an arbitrary matrix in $\mathcal{M}_{n \times n}(K)$ on the left by an elementary matrix:

- (1) If $B \in \mathcal{M}_{n \times n}(K)$ then $E_{hk}B$ is the matrix obtained from B by interchanging the h th and k th rows of B . Thus, for example, in $\mathcal{M}_{4 \times 4}(\mathbb{Q})$ we look at the effect of E_{24} :

$$\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \end{bmatrix} \begin{bmatrix} 5 & 6 & 4 & 1 \\ 3 & 2 & 2 & 2 \\ 0 & 4 & 2 & 7 \\ 3 & 3 & 2 & 2 \end{bmatrix} = \begin{bmatrix} 5 & 6 & 4 & 1 \\ 3 & 3 & 2 & 2 \\ 0 & 4 & 2 & 7 \\ 3 & 2 & 2 & 2 \end{bmatrix}.$$

- (2) If $B \in \mathcal{M}_{n \times n}(K)$ then $E_{h;c}B$ is the matrix obtained from B by multiplying the h th row of B by c . Thus, for example, in $\mathcal{M}_{4 \times 4}(\mathbb{Q})$ we

look at the effect of $E_{2;5}$:

$$\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 5 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 5 & 6 & 4 & 1 \\ 3 & 2 & 2 & 2 \\ 0 & 4 & 2 & 7 \\ 3 & 3 & 2 & 2 \end{bmatrix} = \begin{bmatrix} 5 & 6 & 4 & 1 \\ 15 & 10 & 10 & 10 \\ 0 & 4 & 2 & 7 \\ 3 & 3 & 2 & 2 \end{bmatrix}.$$

(3) If $B \in \mathcal{M}_{n \times n}(K)$ then $E_{hk;c}B$ is the matrix obtained from B by adding c times the h th row to the k th row. Thus, for example, in $\mathcal{M}_{4 \times 4}(\mathbb{Q})$ we look at the effect of $E_{13;2}$:

$$\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 2 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 5 & 6 & 4 & 1 \\ 3 & 2 & 2 & 2 \\ 0 & 4 & 2 & 7 \\ 3 & 3 & 2 & 2 \end{bmatrix} = \begin{bmatrix} 5 & 6 & 4 & 1 \\ 3 & 2 & 2 & 2 \\ 10 & 16 & 10 & 9 \\ 3 & 3 & 2 & 2 \end{bmatrix}.$$

(9.1) Proposition: If F is a field, if n is a positive integer, and if $A, B, C, D \in M_{n \times n}(F)$ then:

(1) When A and B are nonsingular, so is AB , with $(AB)^{-1} = B^{-1}A^{-1}$;

(2) When AB is nonsingular, both A and B are nonsingular;

(3) When A and B are nonsingular,

$$A^{-1} + B^{-1} = A^{-1}(B + A)B^{-1};$$

(4) When $I + AB$ is nonsingular, so is $I + BA$, and

$$(I + BA)^{-1} = I - B(I + AB)^{-1}A;$$

(5) **(Guttman's Theorem)** If A is nonsingular and if $v, w \in F^n$ satisfy the condition that $1 + w \odot A^{-1}v \neq 0$, then the matrix $A + v \wedge w \in M_{n \times n}(F)$ is nonsingular and satisfies $(A + v \wedge w)^{-1} = A^{-1} - (1 + w \odot A^{-1}v)^{-1} (A^{-1}[v \wedge w]A^{-1})$.

(6) **(Sherman-Morrison-Woodbury Theorem⁴)** When the matrices C , D , $D^{-1} + AC^{-1}B$ and $C + BDA$ are nonsingular, then $(C + BDA)^{-1} = C^{-1} - C^{-1}B(D^{-1} + AC^{-1}B)^{-1}AC^{-1}$.



4

Louis Guttman was a 20th-century American/Israeli statistician and sociologist, who developed many advanced mathematical tools for use in statistics. **Jack Sherman**, **Winifred J. Morrison**, and **Max Woodbury** were 20th-century American statisticians.

Proof: (1) This is a special case of a general remark about units in associative F -algebras, which we have already noted.

(2) Let V a vector space of dimension n over F , and let D be a basis of V . Then there exist endomorphisms α and β of V satisfying $A = \Phi_{DD}(\alpha)$ and $B = \Phi_{DD}(\beta)$, and so $AB = \Phi_{DD}(\alpha\beta)$. Since AB is nonsingular, we know that $\alpha\beta \in \text{Aut}(V)$. Then there exists an automorphism γ of V satisfying $\gamma(\alpha\beta) = \sigma_1 = (\alpha\beta)\gamma$. Then $(\gamma\alpha)\beta = \sigma_1 = \alpha(\beta\gamma)$ and so, by Proposition 7.4, we know that both α and β are automorphisms of V and hence both A and B are nonsingular.

(3) This is an immediate consequence of the fact that $A(A^{-1} + B^{-1})B = B + A$.

(4) We note that

$$\begin{aligned} (I + BA) [I - B(I + AB)^{-1}A] &= I + BA - (B + BAB)(I + AB)^{-1}A \\ &= I + BA - B(I + AB)(I + AB)^{-1}A \\ &= I + BA - BA = I. \end{aligned}$$

(5) A simple calculation shows us that if $x, y \in F^n$ satisfy the condition that $c = 1 + y \odot x$ is nonzero, then

$$\begin{aligned} (I - c^{-1}[x \wedge y]) (I + [x \wedge y]) &= I + x \wedge y - c^{-1}[x \wedge y] - c^{-1}[x \wedge y]^2 \\ &= I + x \wedge y - c^{-1}c[x \wedge y] = I \end{aligned}$$

and so $(I + [x \wedge y])^{-1} = I - c^{-1}[x \wedge y]$. Therefore, if we set $d = 1 + w \odot A^{-1}v$, then

$$\begin{aligned} (A + v \wedge w)^{-1} &= [A(I + A^{-1}[v \wedge w])]^{-1} = (I + A^{-1}[v \wedge w])^{-1} A^{-1} \\ &= [I - d^{-1}(A^{-1}[v \wedge w])] A^{-1} \\ &= A^{-1} - d^{-1}(A^{-1}[v \wedge w]A^{-1}) \end{aligned}$$

as required.

(6) First note that $I + C^{-1}BDA = C^{-1}(C + BDA)$ and so, by (1), this matrix too is nonsingular. By (4),

$$(I + C^{-1}BDA)^{-1} = I - C^{-1}B(I + (DAC^{-1}B)DA)$$

and so

$$\begin{aligned} (C + BDA)^{-1} &= [C(I + C^{-1}BDA)]^{-1} \\ &= [I - C^{-1}B(I + DAC^{-1}B)^{-1}DA] C^{-1} \\ &= C^{-1} - C^{-1}B[D^{-1}(I + DAC^{-1}B)]^{-1}AC^{-1} \\ &= C^{-1} - C^{-1}B(D^{-1} + AC^{-1}B)^{-1}AC^{-1}, \end{aligned}$$

as required. \square

Guttman's Theorem is important in the following context: assume we have calculated A^{-1} for some square matrix A and now we have to calculate B^{-1} , where B differs from A in only one entry. With the help of this result, we can make use of our knowledge of A^{-1} to calculate B^{-1} with relative ease and speed. The Sherman-Morrison-Woodbury Theorem has similar uses.

In particular, we note from Proposition 9.1 that, if $A, B \in \mathcal{M}_{n \times n}(F)$, then AB is nonsingular if and only if BA is nonsingular. We should also note that, if $A, B \in \mathcal{M}_{n \times n}(F)$, then $B^T A^T = (AB)^T$ and so, if A is a nonsingular matrix and $B = A^{-1}$ then $AB = I$ and so $B^T A^T = I^T = I$ and so A^T is also nonsingular. Moreover, this shows that $(A^T)^{-1} = (A^{-1})^T$ for every nonsingular matrix $A \in \mathcal{M}_{n \times n}(F)$.

(9.2) Proposition: Let F be a field, let n be a positive integer, and let $A, B \in \mathcal{M}_{n \times n}(F)$, where A is nonsingular. Then there exist unique matrices C and D in $\mathcal{M}_{n \times n}(F)$ satisfying $CA = B = AD$.

Proof: Define $C = BA^{-1}$ and $D = A^{-1}B$. Then surely $CA = B = AD$. If C' and D' are matrices satisfying $C'A = B = AD'$ then $C' = (C'A)A^{-1} = BA^{-1} = C$ and $D' = A^{-1}(AD') = A^{-1}B = D$, and so we have uniqueness. \square

Example: The matrices C and D in Proposition 9.2 need not be the same. For example, if $A, B \in \mathcal{M}_{2 \times 2}(\mathbb{R})$ are defined by $A = \begin{bmatrix} 1 & 3 \\ 0 & 1 \end{bmatrix}$ and $B = \begin{bmatrix} 1 & 0 \\ 3 & 1 \end{bmatrix}$ then $A^{-1} = \begin{bmatrix} 1 & -3 \\ 0 & 1 \end{bmatrix}$, $C = \begin{bmatrix} 1 & -3 \\ 3 & -8 \end{bmatrix}$ and $D = \begin{bmatrix} -8 & -3 \\ 3 & 1 \end{bmatrix}$.

(9.3) Proposition: Let F be a field, let n be a positive integer, and let $A = [a_{ij}] \in \mathcal{M}_{n \times n}(F)$. Then the following conditions are equivalent:

- (1) A is nonsingular;
- (2) The columns of A are distinct and the set of these columns is a linearly-independent subset of F^n ;
- (3) The rows of A are distinct and the set of these rows is a linearly-independent subset of $\mathcal{M}_{1 \times n}(F)$.

Proof: (1) \Leftrightarrow (2): Denote the columns of A by y_1, \dots, y_n . Let $V = F^n$ and let $B = \{v_1, \dots, v_n\}$ be the canonical basis of V . If two columns of A are equal or if the set of columns is linearly dependent, there exist scalars

c_1, \dots, c_n , not all equal to 0, such that $A \begin{bmatrix} c_1 \\ \vdots \\ c_n \end{bmatrix} = \sum_{i=1}^n c_i y_i = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$.

But if (1) holds, then $\begin{bmatrix} c_1 \\ \vdots \\ c_n \end{bmatrix} = A^{-1} \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$, which is a

contradiction. Therefore (2) holds. Conversely, assume (2) holds. Then the endomorphism α of V given by $v \mapsto Av$ is a monic and so an automorphism of V . But $A = \Phi_{BB}(\alpha)$, and so, as we have seen, A is nonsingular.

(1) \Leftrightarrow (3): This follows directly from the equivalence of (1) and (2), given the fact that a matrix A is nonsingular if and only if A^T is nonsingular. \square

Example: Let F be a field and let $n > 1$ be an integer. If $v, w \in F^n$, then the columns of $v \wedge w \in \mathcal{M}_{n \times n}(F)$ are linearly dependent and so $v \wedge w$ is always singular.

Example: If F is a field and if $U = [u_{ij}] \in \mathcal{M}_{n \times n}(F)$ is an upper-triangular matrix satisfying the condition that $u_{ii} \neq 0$ for all $1 \leq i \leq n$ then, by Proposition 9.3, it is clear that U is nonsingular. We claim that, moreover, U^{-1} is again upper triangular. Let us prove this contention by induction on n . It is clearly true for $n = 1$. Assume therefore that $n > 1$ and that we have already shown that the inverse of any upper-triangular

matrix in $\mathcal{M}_{(n-1) \times (n-1)}(F)$ is upper-triangular. Write $U = \begin{bmatrix} A & y \\ z & u_{nn} \end{bmatrix}$,

where $A \in \mathcal{M}_{(n-1) \times (n-1)}(F)$, $y \in F^{n-1}$, and $z = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}^T$. Assume

that $U^{-1} = \begin{bmatrix} B & x \\ w^T & b \end{bmatrix}$, where $B \in \mathcal{M}_{(n-1) \times (n-1)}(F)$ and $w, x \in F^{n-1}$.

Then $AB + y \wedge w = I$, $Ax + by = z^T$, $u_{nn}w^T = z$, and $u_{nn}b = 1$, so we must have $b = u_{nn}^{-1} \neq 0$ and $w^T = z$. Therefore $y \wedge w = 0$ and so $B = A^{-1}$. By hypothesis, B is upper triangular and so U^{-1} is again upper triangular. A similar argument holds for lower-triangular matrices.

(9.4) Proposition: Let F be a field and let n be a positive integer. A matrix in $\mathcal{M}_{n \times n}(F)$ is nonsingular if and only if it is a product of elementary matrices.

Proof: Since each of the elementary matrices is nonsingular, we know that any product of elementary matrices is also nonsingular. Conversely, let $A = [a_{ij}]$ be a nonsingular matrix in $\mathcal{M}_{n \times n}(F)$ and let $B = [b_{ij}]$ be A^{-1} . Then B is also nonsingular and so, by Proposition 9.3, the columns of B are distinct and the set of columns is linearly independent in F^n . In particular, there exists a nonzero entry b_{h1} in the first column of B . Multiply B on the left by E_{h1} to get a new matrix in which the $(1, 1)$ -entry nonzero. Now multiply it on the left by $E_{1;c}$, where $c = b_{h1}^{-1}$, in

order to get a matrix of the form
$$\begin{bmatrix} 1 & * & \dots & * \\ * & * & \dots & * \\ \vdots & \vdots & & \vdots \\ * & * & \dots & * \end{bmatrix}.$$
 Now let $1 < t \leq n$,

and let $d(t)$ be the additive inverse of the $(t, 1)$ -entry of the matrix. Multiplying the matrix on the left by $E_{t1;d(t)}$, we will get a matrix with 0 in the $(t, 1)$ -entry and so, after this for each such t , we see that a matrix $B' = C_1 B$, where C_1 is a product of elementary matrices, and which is

of the form
$$\begin{bmatrix} 1 & * & \dots & * \\ 0 & * & \dots & * \\ \vdots & \vdots & & \vdots \\ 0 & * & \dots & * \end{bmatrix}.$$
 This matrix is still nonsingular, since it

is a product of two nonsingular matrices, and so its columns are distinct and form a linearly-independent subset of F^n . Therefore there exists a nonzero entry b'_{h2} in the second column, with $h > 1$. Repeating the above procedure, we can find a matrix C_2 which is a product of elementary

matrices and such that $C_2 C_1 B$ is of the form
$$\begin{bmatrix} 1 & 0 & * & \dots & * \\ 0 & 1 & * & \dots & * \\ 0 & 0 & * & \dots & * \\ \vdots & \vdots & \vdots & & \vdots \\ 0 & 0 & * & \dots & * \end{bmatrix}.$$

Continuing in this manner, we obtain matrices C_1, \dots, C_n , each of them a product of elementary matrices, such that $C_n \dots C_1 B = I$. Therefore $C_n \dots C_1 = B^{-1} = A$, as we wanted to show. \square

Example: Let F be a field and let n be a positive integer. Every permutation π of the set $\{1, \dots, n\}$ defines a matrix $A_\pi = [a_{ij}] \in \mathcal{M}_{n \times n}(F)$ by setting $a_{ij} = 1$ if $j = \pi(i)$ and $a_{ij} = 0$ otherwise, called the **permutation matrix** defined by π . This matrix is clearly a result of multiplying I by a number of elementary matrices of the form E_{hk} , and so is nonsingular.

Proposition 9.4 allows us to construct an algorithm for computing A^{-1} when A is a nonsingular matrix in $\mathcal{M}_{n \times n}(F)$. First of all, we construct the matrix $\begin{bmatrix} I & A \end{bmatrix} \in \mathcal{M}_{n \times 2n}(F)$ and on this matrix we perform a series of **elementary operations**, namely operations which are the result of multiplying it on the left by elementary matrices, which bring the right-hand block into the form I . Then the left-hand block is A^{-1} . To calculate A^{-1} by this method, we use $n^3 - 2n^2 + n$ additions and n^3 multiplications.

Example: Consider the matrix $A = \begin{bmatrix} 1 & 2 & 3 \\ 2 & 3 & 0 \\ 0 & 1 & 2 \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{Q})$. There-

fore we begin with the matrix $\begin{bmatrix} 1 & 0 & 0 & 1 & 2 & 3 \\ 0 & 1 & 0 & 2 & 3 & 0 \\ 0 & 0 & 1 & 0 & 1 & 2 \end{bmatrix} \in \mathcal{M}_{3 \times 6}(\mathbb{Q})$. Then:

(1) Get $\begin{bmatrix} 1 & 0 & 0 & 1 & 2 & 3 \\ -2 & 1 & 0 & 0 & -1 & -6 \\ 0 & 0 & 1 & 0 & 1 & 2 \end{bmatrix}$ after multiplying the first row by -2 and adding it to the second row;

(2) Get $\begin{bmatrix} 1 & 0 & 0 & 1 & 2 & 3 \\ -2 & 1 & 0 & 0 & -1 & -6 \\ -2 & 1 & 1 & 0 & 0 & -4 \end{bmatrix}$ after adding the second row to the third row;

(3) Get $\begin{bmatrix} 1 & 0 & 0 & 1 & 2 & 3 \\ 2 & -1 & 0 & 0 & 1 & 6 \\ 0.5 & -0.25 & -0.25 & 0 & 0 & 1 \end{bmatrix}$ after multiplying the second row by -1 and then multiplying the third row by -0.25 ;

(4) Get $\begin{bmatrix} 1 & 0 & 0 & 1 & 2 & 3 \\ -1 & 0.5 & 1.5 & 0 & 1 & 0 \\ 0.5 & -0.25 & -0.25 & 0 & 0 & 1 \end{bmatrix}$ after multiplying the third row by -6 and adding it to the second row;

(5) Get $\begin{bmatrix} -0.5 & 0.75 & 0.75 & 1 & 2 & 0 \\ -1 & 0.5 & 1.5 & 0 & 1 & 0 \\ 0.5 & -0.25 & -0.25 & 0 & 0 & 1 \end{bmatrix}$ after multiplying the third row by -3 and adding it to the first row;

(6) Finally, get $\begin{bmatrix} 1.5 & -0.25 & -2.25 & 1 & 0 & 0 \\ -1 & 0.5 & 1.5 & 0 & 1 & 0 \\ 0.5 & -0.25 & -0.25 & 0 & 0 & 1 \end{bmatrix}$ after multiplying the second row by -2 and adding it to the first row.

Therefore we see that $A^{-1} = \frac{1}{4} \begin{bmatrix} 6 & -1 & -9 \\ -4 & 2 & 6 \\ 2 & -1 & -1 \end{bmatrix}$.

Example: When one uses computer to compute matrix inverses, one must always be aware of hardware limitations. For example, one can show

that **Nievergelt's matrix** $A = \begin{bmatrix} 888445 & 887112 \\ 887112 & 885871 \end{bmatrix} \in \mathcal{M}_{2 \times 2}(\mathbb{Q})$ is non-

singular, while the matrix $B = A - \begin{bmatrix} c & c \\ c & c \end{bmatrix}$, where $c = \frac{1}{3548450}$ (which

is approximately $2.81813186039 \times 10^{-7}$) is not. Nonetheless, a computer or calculator capable of only 12-digit accuracy cannot differentiate between the two.

Example: For each positive integer n , let $H_n \in \mathcal{M}_{n \times n}(\mathbb{Q})$ be the matrix $[a_{ij}]$, in which $a_{ij} = \frac{1}{i+j-1}$. This matrix is called the $n \times n$ **Hilbert matrix**⁵. The Hilbert matrices are all nonsingular but, while their entries all lie between 0 and 1, the entries in their inverses are very large. For example, H_6^{-1} equals

$$\begin{bmatrix} 36 & -630 & 3360 & -7560 & 7560 & -2772 \\ -630 & 14700 & -88200 & 211680 & -220500 & 83160 \\ 3360 & -88200 & 564480 & -1411200 & 1512000 & -582120 \\ -7560 & 211680 & -1411200 & 3628800 & -3969000 & 1552320 \\ 7560 & -220500 & 1512000 & -3969000 & 4410000 & -1746360 \\ -2772 & 83160 & -582120 & 1552320 & -1746360 & 698544 \end{bmatrix}$$

Therefore these matrices are often used as benchmarks to judge the efficiency and accuracy of computer programs to calculate matrix inverses. In particular if the computer we are using has only 7-digit accuracy, it is reasonable to assume that we will have a 100% error in computing H_6^{-1} .

It is sometimes possible to use a representation of a nonsingular matrix A in block form in order to calculate A^{-1} . Indeed, suppose that $A \in$



5

German **David Hilbert** was one of the foremost mathematicians in the world at the beginning of the 20th century. He and his students were among the first to study infinite-dimensional vector spaces.

$\mathcal{M}_{n \times n}(F)$ is a matrix which can be written in block form $\begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix}$,

where $A_{11} \in \mathcal{M}_{k \times k}(F)$. In order to find A^{-1} , we must find a matrix $\begin{bmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{bmatrix}$, where $B_{11} \in \mathcal{M}_{k \times k}(F)$, such that

$$\begin{aligned} A_{11}B_{11} + A_{12}B_{21} &= I \\ A_{11}B_{12} + A_{12}B_{22} &= O \\ A_{21}B_{11} + A_{22}B_{21} &= O \\ A_{21}B_{12} + A_{22}B_{22} &= I \end{aligned}$$

and if we know that the matrix A_{11} is itself nonsingular, then from the first equation we immediately see that $B_{11} = A_{11}^{-1}(I - A_{12}B_{21})$. If we substitute this in the third equation, we obtain

$$B_{21} = -[A_{22} - A_{21}A_{11}^{-1}A_{12}]^{-1}A_{21}A_{11}^{-1},$$

under the assumption, of course, that the matrix $C = A_{22} - A_{21}A_{11}^{-1}A_{12}$ is nonsingular. In the same manner, we also obtain $B_{12} = -A_{11}^{-1}A_{12}C^{-1}$ and $B_{22} = A_{22}^{-1}(I - A_{21}B_{12})$ if the matrix A_{22} is nonsingular. Thus we see that a sufficient condition for A to be nonsingular is that we can partition it as above in such a manner that A_{11} , A_{22} , and $A_{22} - A_{21}A_{11}^{-1}A_{12}$ are all nonsingular.

Let F be a field and let k and n be positive integers. Two matrices $B, C \in \mathcal{M}_{k \times n}(F)$ are **equivalent** if and only if there exist nonsingular matrices $P \in \mathcal{M}_{k \times k}(F)$ and $Q \in \mathcal{M}_{n \times n}(F)$ such that $PBQ = C$. This is, indeed, an equivalence relation on $\mathcal{M}_{k \times n}(F)$ since:

(1) $IBI = B$ for each such matrix B , showing that B is equivalent to itself;

(2) If $PBQ = C$ then $P^{-1}CQ^{-1} = B$;

(3) If $PBQ = C$ and $P'CQ' = D'$ then $(P'P)B(QQ') = D'$, where we note that both $P'P$ and QQ' are again nonsingular.

Similarly, we say that B and C are **row equivalent** if and only if there exists a nonsingular matrix $P \in \mathcal{M}_{k \times k}(F)$ satisfying $PB = C$, and we say that B and C are **column equivalent** if and only if there exists a nonsingular matrix $Q \in \mathcal{M}_{n \times n}(F)$ satisfying $BQ = C$. Both of these relations are also equivalence relations on $\mathcal{M}_{k \times n}(F)$, and it is clear that if B and C are row equivalent then they are equivalent (take $Q = I$) and if they are column equivalent then they are equivalent (take $P = I$).

Equivalence of matrices is a very strong concept. Indeed, it is easy to show that any matrix $B \in \mathcal{M}_{k \times n}(F)$ is equivalent to one which is in block

form $\begin{bmatrix} I & O \\ O & O \end{bmatrix}$. Therefore, it is more useful to consider row equivalence of matrices as our basic tool.

Now let V be a vector space of dimension n over a field F and choose bases $B = \{v_1, \dots, v_n\}$ and $D = \{w_1, \dots, w_n\}$ of V . For each $1 \leq j \leq n$ there exist elements q_{1j}, \dots, q_{nj} of F satisfying $w_j = \sum_{i=1}^n q_{ij}v_i$. By Proposition 9.3, we know that the matrix $Q = [q_{ij}]$ is nonsingular. If $v = \sum_{i=1}^n a_i v_i = \sum_{j=1}^n b_j w_j$ is an element of V , then we see that $v = \sum_{j=1}^n b_j w_j = \sum_{j=1}^n b_j (\sum_{i=1}^n q_{ij}v_i) = \sum_{i=1}^n (\sum_{j=1}^n q_{ij}b_j)v_i$ and so we must have $a_i = \sum_{j=1}^n q_{ij}b_j$ for all $1 \leq i \leq n$. Thus we see

that $\begin{bmatrix} a_1 \\ \vdots \\ a_n \end{bmatrix} = Q \begin{bmatrix} b_1 \\ \vdots \\ b_n \end{bmatrix}$. The matrix Q is called the **change-of-basis matrix** from D to B .

Example: Let F be a field, let n be a positive integer, and let V be the subspace of the vector space $F[X]$ made up of all polynomials of degree at most $n-1$. Then $\dim(V) = n$, and it has a canonical basis $B = \{1, X, \dots, X^{n-1}\}$. Let c_1, \dots, c_n be distinct scalars, and for each $1 \leq i \leq n$, consider the polynomial

$$p_i(X) = \prod_{j \neq i} \frac{1}{c_i - c_j} (X - c_j) \in V.$$

This polynomial is called the i th **Lagrange interpolation polynomial**, and we will return to these polynomials below in another context. It is clear that

$$p_i(c_j) = \begin{cases} 1 & \text{if } i = j \\ 0 & \text{otherwise} \end{cases}.$$

Thus, for example, if $n = 4$ and if we choose $c_1 = 1$, $c_2 = 3$, $c_3 = 5$, and $c_4 = 7$, we obtain

$$\begin{aligned} p_1(X) &= -\frac{1}{48}X^3 + \frac{5}{16}X^2 - \frac{71}{48}X + \frac{35}{16} \\ p_2(X) &= \frac{1}{16}X^3 - \frac{13}{16}X^2 + \frac{47}{16}X - \frac{35}{16} \\ p_3(X) &= \frac{-1}{16}X^3 + \frac{11}{16}X^2 - \frac{31}{16}X + \frac{21}{16} \\ p_4(X) &= \frac{1}{48}X^3 - \frac{3}{16}X^2 + \frac{23}{48}X - \frac{5}{16}. \end{aligned}$$

Returning to the general case, we see that the set $D = \{p_1(X), \dots, p_n(X)\}$ of Lagrange interpolation polynomials is linearly independent since, if we

have $\sum_{i=1}^n a_i p_i(X) = 0$, then for each $1 \leq h \leq n$ we have $a_h = \sum_{i=1}^n a_i p_i(c_h) = 0$. Therefore D is also a basis of V . If $q(X)$ is an arbitrary polynomial in V then there exist scalars a_1, \dots, a_n satisfying $q(X) = \sum_{i=1}^n a_i p_i(X)$; Again, this implies that $a_i = q(c_i)$ for all i . In particular, if $q(X) = X^k$ we see that $X^k = \sum_{i=1}^n c_i^k p_i(X)$. Therefore

the change of basis matrix from D to B is
$$\begin{bmatrix} 1 & c_1 & \dots & c_1^n \\ 1 & c_2 & \dots & c_2^n \\ \vdots & \vdots & & \vdots \\ 1 & c_n & \dots & c_n^n \end{bmatrix}.$$
 A

matrix of this form is called a **Vandermonde matrix**, and such matrices are always nonsingular⁶.

Lagrange interpolation allows us to represent a polynomial $p(X)$ of degree less than n in a computer not by its list of coefficients but rather by a list of its values $p(a_1), \dots, p(a_n)$ at n preselected elements of F . Such representations can be used to obtain algorithms for rapid multiplication of polynomials, especially in the case the field F is finite.

Let us now return to the matter of change of basis, and now let us assume that we have a linear transformation $\alpha : V \rightarrow Y$, where V is a vector space of dimension n over a field F and Y is a vector space of dimension k over F . We have bases $B = \{v_1, \dots, v_n\}$ and $D = \{w_1, \dots, w_n\}$ of V . Choose a basis $E = \{y_1, \dots, y_k\}$ of Y . Then $\Phi_{BE}(\alpha)$ is a matrix $C = [c_{ij}]$. If $Q = [q_{ij}]$ is the change of basis matrix from D to B then for each $1 \leq j \leq n$ we have $\alpha(w_j) = \alpha(\sum_{h=1}^n q_{hj} v_h) = \sum_{h=1}^n q_{hj} \alpha(v_h) = \sum_{h=1}^n q_{hj} \left(\sum_{i=1}^k c_{ih} y_i \right) = \sum_{i=1}^k \left(\sum_{h=1}^n c_{ih} q_{hj} \right) y_i$ and so $\Phi_{DE}(\alpha) = CQ$, showing that $\Phi_{DE}(\alpha)$ and C are column equivalent. In the same manner, if we have another basis $G = \{z_1, \dots, z_k\}$ of Y and if $P = [p_{ij}]$ is the change of basis matrix from E to G , then $z_j = \sum_{i=1}^k p_{ij} y_i$ for all $1 \leq j \leq k$. If $\Phi_{BG}(\alpha)$ is the matrix $C' = [c'_{ij}]$, then for all $1 \leq j \leq n$ we have $\alpha(v_j) = \sum_{h=1}^k e_{hj} z_h = \sum_{h=1}^k e_{hj} \left(\sum_{i=1}^k p_{ih} y_i \right) =$



6

Joseph-Louis Lagrange was one of the applied mathematicians who surrounded Napoleon, and his book on analytical mechanics is considered a mathematical classic. **Alexandre-Théophile Vandermonde** was an 18th century French chemist and mathematician who studied determinants of matrices. Vandermonde matrices do not appear in his work, and it is not clear why they are named after him.

$\sum_{i=1}^k \left(\sum_{h=1}^k p_{ih} e_{hj} \right) y_i$ and this equals $\sum_{i=1}^k c_{ij} y_i$ and so $C = PC'$ and so $C' = P^{-1}C$. Thus $\Phi_{BG}(\alpha)$ and C are row equivalent. If we put both of these results together, we see that $\Phi_{DG}(\alpha) = P^{-1}\Phi_{BE}(\alpha)Q$, and so $\Phi_{DG}(\alpha)$ and $\Phi_{BE}(\alpha)$ are equivalent.

Example: Let $\alpha: \mathbb{R}^3 \rightarrow \mathbb{R}^2$ be the linear transformation given by $\alpha:$

$$\begin{bmatrix} a \\ b \\ c \end{bmatrix} \mapsto \begin{bmatrix} a+b \\ b+c \end{bmatrix}.$$
 Choose bases $B = \left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 0 \\ -1 \\ 0 \end{bmatrix} \right\}$ and

$$D = \left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ -1 \end{bmatrix} \right\}$$
 of \mathbb{R}^3 and bases $E = \left\{ \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ -1 \end{bmatrix} \right\}$
and $G = \left\{ \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ -1 \end{bmatrix} \right\}$ of \mathbb{R}^2 . Then $\Phi_{BE}(\alpha) = \begin{bmatrix} 1 & 1 & -1 \\ 0 & -2 & 1 \end{bmatrix}$
since

$$\begin{aligned} \alpha \left(\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \right) &= \begin{bmatrix} 1 \\ 0 \end{bmatrix} = 1 \begin{bmatrix} 1 \\ 0 \end{bmatrix} + 0 \begin{bmatrix} 0 \\ -1 \end{bmatrix} \\ \alpha \left(\begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix} \right) &= \begin{bmatrix} 1 \\ 2 \end{bmatrix} = 1 \begin{bmatrix} 1 \\ 0 \end{bmatrix} - 2 \begin{bmatrix} 0 \\ -1 \end{bmatrix} \\ \alpha \left(\begin{bmatrix} 0 \\ -1 \\ 0 \end{bmatrix} \right) &= \begin{bmatrix} -1 \\ -1 \end{bmatrix} = (-1) \begin{bmatrix} 1 \\ 0 \end{bmatrix} + 1 \begin{bmatrix} 0 \\ -1 \end{bmatrix} \end{aligned}$$

and, similarly, $\Phi_{DG}(\alpha) = \frac{1}{2} \begin{bmatrix} 1 & 2 & 1 \\ 1 & 0 & 1 \end{bmatrix},$

$$\begin{aligned} \alpha \left(\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \right) &= \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \frac{1}{2} \begin{bmatrix} 1 \\ 1 \end{bmatrix} + \frac{1}{2} \begin{bmatrix} 1 \\ -1 \end{bmatrix} \\ \alpha \left(\begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} \right) &= \begin{bmatrix} 1 \\ 1 \end{bmatrix} = 1 \begin{bmatrix} 1 \\ 1 \end{bmatrix} + 0 \begin{bmatrix} 1 \\ -1 \end{bmatrix} \\ \alpha \left(\begin{bmatrix} 0 \\ 1 \\ -1 \end{bmatrix} \right) &= \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \frac{1}{2} \begin{bmatrix} 1 \\ 1 \end{bmatrix} + \frac{1}{2} \begin{bmatrix} 1 \\ -1 \end{bmatrix} \end{aligned}$$

Further, we also see that
$$\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} = 1 \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} + 0 \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix} + 0 \begin{bmatrix} 0 \\ -1 \\ 0 \end{bmatrix},$$

$$\begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} = 1 \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} + 1 \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix} + 1 \begin{bmatrix} 0 \\ -1 \\ 0 \end{bmatrix}, \quad \text{and} \quad \begin{bmatrix} 0 \\ 1 \\ -1 \end{bmatrix} = 0 \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} -$$

$$1 \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix} - 2 \begin{bmatrix} 0 \\ -1 \\ 0 \end{bmatrix} \quad \text{so} \quad Q = \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & -1 \\ 0 & 1 & -2 \end{bmatrix}.$$

$$\text{Moreover, } \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} = 1 \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} - 1 \begin{bmatrix} 0 \\ -1 \\ -1 \end{bmatrix} \quad \text{and} \quad \begin{bmatrix} 1 \\ -1 \\ -1 \end{bmatrix} = 1 \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} +$$

$$1 \begin{bmatrix} 0 \\ -1 \\ -1 \end{bmatrix} \quad \text{so} \quad P = \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix} \quad \text{and} \quad P^{-1} = \frac{1}{2} \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix}. \quad \text{Note that} \\ P^{-1}\Phi_{BE}(\alpha)Q = \Phi_{DG}(\alpha).$$

Example: We will now see an application of linear algebra to calculus. Let V be the vector space over \mathbb{R} consisting of all infinitely-differentiable functions $f \in \mathbb{R}^{\mathbb{R}}$, and let $\delta \in \text{End}(V)$ be the differentiation endomorphism.

(1) If a and b are given real numbers, not both equal to 0, then the functions $f_0 : x \mapsto e^{ax} \sin(bx)$ and $f_1 : x \mapsto e^{ax} \cos(bx)$ belong to V and the subspace $W = \mathbb{R}\{f_0, f_1\}$ of V is invariant under δ . The restriction of δ to W can be represented with respect to the basis

$\{f_0, f_1\}$ of W by the nonsingular matrix $A = \begin{bmatrix} a & -b \\ b & a \end{bmatrix}$. It is easy to check that $A^{-1} = \frac{1}{a^2+b^2} \begin{bmatrix} a & b \\ -b & a \end{bmatrix}$. Therefore

$$\begin{aligned} \int f_0(t) dt &= \delta^{-1}(f_0) = \left(\frac{1}{a^2+b^2} \right) [af_0 - bf_1] \quad \text{and} \\ \int f_1(t) dt &= \delta^{-1}(f_1) = \left(\frac{1}{a^2+b^2} \right) [bf_0 + af_1]. \end{aligned}$$

(2) The functions $g_0 : x \mapsto x^2 e^x$, $g_1 : x \mapsto x e^x$, and $g_2 : x \mapsto e^x$ all belong to V and the subspace $Y = \mathbb{R}\{g_0, g_1, g_2\}$ of V is invariant under δ . The restriction of δ to Y can be represented with respect to the basis

$\{g_0, g_1, g_2\}$ of Y by the nonsingular matrix $B = \begin{bmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ 0 & 1 & 1 \end{bmatrix}$. Since

$$B^{-1} = \begin{bmatrix} 1 & 0 & 0 \\ -2 & 1 & 0 \\ 2 & -1 & 1 \end{bmatrix}, \text{ we see that}$$

$$\int g_0(t)dt = \delta^{-1}(g_0) = g_0 - 2g_1 + 2g_2,$$

$$\int g_1(t)dt = \delta^{-1}(g_1) = g_1 - g_2, \text{ and}$$

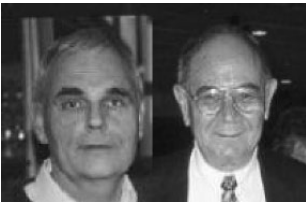
$$\int g_2(t)dt = \delta^{-1}(g_2) = g_2.$$

Let us turn to problems connected with the implementation of this theory. Let F be a field and let n be a positive integer. Let $A = [a_{ij}]$ and $B = [b_{ij}]$ belong to $\mathcal{M}_{n \times n}(F)$ and let $C = AB$. In order to calculate each one of the n^2 entries in C , we need n multiplications and $n - 1$ additions/subtractions, and so to calculate C we need n^3 multiplications and $n^3 - 2n^2 + n$ additions/subtractions. Putting this in another way, the total number of operations needed to calculate AB from the definition is on the order of n^c , where $c = 3$. If n is very large, this can entail considerable computational overhead and leaves room for the introduction of considerable error due to roundoff and truncation in the course of the calculation. It is therefore very important to find a more sophisticated method of matrix multiplication, if possible. One such method is the **Strassen-Winograd algorithm**⁷.

To illustrate the Strassen-Winograd algorithm, let us first begin with the special case $n = 2$. First, calculate

$$\begin{aligned} p_0 &= (a_{11} + a_{12})(b_{11} + b_{12}) & p_1 &= (a_{11} + a_{22})b_{11} & p_2 &= a_{11}(b_{12} - b_{22}) \\ p_3 &= (a_{21} - a_{11})(b_{11} + b_{12}) & p_4 &= (a_{11} + a_{12})b_{22} & p_5 &= a_{22}(b_{21} - b_{11}) \\ p_6 &= (a_{12} - a_{22})(b_{21} + b_{22}) \end{aligned}$$

and then note that $C = \begin{bmatrix} p_0 + p_5 - p_4 + p_6 & p_2 + p_4 \\ p_1 + p_5 & p_0 - p_1 + p_2 + p_3 \end{bmatrix}$. In this calculation, we used 7 multiplications and 18 additions/subtractions



7

Variants of this algorithm were discovered by the contemporary German mathematician **Volker Strassen** and the contemporary Israeli mathematician **Shmuel Winograd**, who later served as director of mathematical research at IBM.

(Winograd's variant of this algorithm uses only 15 additions/subtractions, but these are more interdependent, and so the algorithm is less amenable to implementation on parallel computers) instead of 8 multiplications and 4 additions/subtractions. In the early days of computers, when multiplication was several orders of magnitude slower than addition, this in itself was a great accomplishment. If $n = 4$, we write our matrices in

block form: $A = \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix}$ and $B = \begin{bmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{bmatrix}$, where each

block is a 2×2 matrix. We now calculate 2×2 matrices P_0, \dots, P_6 and then construct $C = AB$ as above. To do this, we need 49 multiplications and 198 additions/subtractions, as opposed to 64 multiplications and 46 additions/subtractions if one goes according to the definition. We continue recursively. If $n = 2^h$, then the number of multiplications needed is $M(h) = 7^h$ and the number of additions/subtractions needed is $A(h) = 6(7^h - 4^h)$ and so $M(h) + A(h) < 7^{h+1}$. (If n is not a power of 2, we can add rows and columns of 0's in order to enlarge it to the desired size.) Thus, we see that the number of arithmetic operations needed to calculate AB is on the order of n^c , where $c \leq \log_2 7 = 2.807\dots$ and so, for large n , we have a definite advantage over multiplication following from the definition. Using even more sophisticated techniques, it is possible to reduce the number of arithmetic operations to the order of n^c , where $c \leq 2.376\dots$, as was done by Winograd and Coppersmith in 1986. The size of matrices for which the Strassen-Winograd algorithm is significantly faster than the regular method depends, of course, on the particular hardware on which it is being used. The Strassen-Winograd algorithm can also be modified to multiplication of matrices which are not necessarily square.

Unfortunately, the Strassen-Winograd algorithm is no less susceptible to roundoff and truncation errors than the regular algorithm. On a computer with seven-digit accuracy, the product

$$\begin{bmatrix} 211 & 2 & 3 & 4 \\ 1 & 2 & 3 & 4 \\ 0.001 & 0.032 & 0.043 & 0.044 \\ 311 & 0.0032 & 1233 & 0.0324 \end{bmatrix} \begin{bmatrix} 50 & 0.32 & 0.0023 & 421 \\ 60 & 0.023 & 0.033 & 982 \\ 23 & 0.032 & 0.03 & 623 \\ 33 & 0.043 & 0.022 & 44 \end{bmatrix}$$

equals $\begin{bmatrix} 10871 & 67.834 & 0.7293 & 92840 \\ 371 & 0.634 & 0.2463 & 4430 \\ 4.411 & 0.0043 & 0.0033 & 60.57 \\ 43910.3 & 138.977 & 37.7061 & 899094.0 \end{bmatrix}$ using the ordinary method of matrix multiplication, whereas, using the Strassen-Winograd algo-

rithm, we obtain
$$\begin{bmatrix} 10871 & 68.54 & 0.6294 & 92840 \\ 370.9 & 1.0 & 0.2463 & 4430.18 \\ 4.411 & 0.0043 & 0 & 62.0 \\ 43910.3 & 139.047 & 37.7 & 899095.0 \end{bmatrix}.$$
 This prob-

lem can be overcome to some extent by stopping the recursion in the Strassen-Winograd algorithm early, and doing the bottom-level matrix multiplication using the ordinary method. Another disadvantage of this algorithm is that it requires a much larger amount of scratch memory space to perform its calculations.

There are other tricks that can be used to reduce the computations necessarily in matrix multiplication. For example, if n is a positive integer and if $A, B, C, D \in \mathcal{M}_{n \times n}(\mathbb{R})$, then the matrix product $(A + iB)(C + iD)$ in $\mathcal{M}_{n \times n}(\mathbb{C})$ can be calculated using only three matrix multiplications in $\mathcal{M}_{n \times n}(\mathbb{R})$, rather than the expected four, by noting that

$$(A + iB)(C + iD) = AC - BD + i[(A + B)(C + D) - AC - BD].$$

Again, we keep in mind that real and complex numbers are represented in a computer by approximations having a limited degree of accuracy. The longer calculations become, the error due to roundoff and truncation increases and limits the correctness of the calculations. It is possible to reduce the effect of roundoff and truncation errors as much as possible. Let us recall how our algorithm for inverting a matrix A worked:

- (1) We formed the matrix $\begin{bmatrix} I & A \end{bmatrix} = [b_{ij}]$;
- (2) We interchanged the first row with one of the rows below it, if necessary, such that $b_{1,n+1} \neq 0$; we then multiplied this row by $b_{1,n+1}^{-1}$ so that this element is now equal to 1, and we subtracted multiples of this row from the rows below it, in order to make $b_{i,n+1}$ equal to 0 for all $1 < i \leq n$.
- (3) We now go iterate this process for the elements $b_{h,n+h}$, where $h = 2, 3, \dots$ and so forth. If we cannot do it, i.e. if there exists an h such that $b_{i,n+h}$ for all $h \leq i \leq n$, the matrix A is nonsingular. Otherwise, at the end of the process, we have brought the matrix to the form $\begin{bmatrix} A^{-1} & I \end{bmatrix}$.

The elements $b_{h,n+h}$ are called **pivots** of the algorithm. If we are working over the real or complex numbers, we can minimize roundoff and truncation errors – to some extent – by making sure that each time we interchange rows we choose to bring into the pivot position a nonzero number having maximal absolute value. This strategy is known as **partial pivoting**. We could do better by also interchanging columns in order to bring into the pivot position $b_{h,n+h}$ the element b_{ij} ($h \leq i, j \leq n$) having maximal absolute value. This strategy is known as **full pivoting** and it

requires a certain amount of computational overhead on the side so that the columns can be returned to their proper positions at the end of the algorithm. Although there are matrices so pathological that full pivoting rather than partial pivoting is needed in order to invert them, most experts believe that the effort is not worth the effort and the computational overhead and that for such matrices it is best to use other methods altogether. Several variants of pivoting strategies for matrices having specific structures have, however, been developed and are in use.

Indeed, let us now consider another method. It is clearly easier to invert a nonsingular upper-triangular or lower-triangular matrix – namely a matrix in one of these forms all of the diagonal elements of which are nonzero. Therefore our job would be much easier if we could write A in the form LU , where L is lower triangular and U is upper triangular⁸, for then $A^{-1} = U^{-1}L^{-1}$. This is not always possible. For example, one can see

that there is no way of writing the matrix $\begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \in \mathcal{M}_{2 \times 2}(\mathbb{R})$ in this

form. However, it is always possible to write A in the form LU when A equals a product of elementary matrices of the form $E_{i;c}$ and $E_{ij;c}$ only.

How can this be done? Assume that $A = [a_{ij}]$, $U = [u_{ij}]$, and $L = [v_{ij}]$ and that $A = LU$, where U is upper-triangular and L is lower-triangular. Then for each $1 \leq i, j \leq n$ we have $a_{ij} = \sum_{h=1}^n v_{ih}u_{hj}$. In each of L and U there are only $\frac{1}{2}(n^2 + n)$ entries which can be nonzero and so our problem is one of solving n^2 nonlinear equations in $n^2 + n$ unknowns. This means that we can allow ourselves to choose the value of n of these variables arbitrarily, and we will do so by insisting that $v_{ii} = 1$ for all $1 \leq i \leq n$. Now we have a system of $n^2 + n$ nonlinear equations in $n^2 + n$ unknowns, which can be solved by a method known as **Crout's algorithm**⁹:

- (1) First set $v_{ii} = 1$ for all $1 \leq i \leq n$;
- (2) For all $2 \leq j \leq n$ and all $1 \leq i \leq j$, first calculate $u_{ij} = a_{ij} - \sum_{h=1}^{i-1} v_{ih}u_{hj}$ and then $v_{ij} = \frac{1}{u_{jj}} \left(a_{ij} - \sum_{h=1}^{j-1} v_{ih}u_{hj} \right)$ for all $j < i \leq n$.



⁸ The LU method was devised by British mathematician **Alan Turing**, who is better known as the founder of automata theory and one of the fathers of the electronic computer. It appears implicitly in the work of Jacobi on bilinear forms.

⁹This algorithm was devised by 20th-century American mathematician Prescott Crout.

We note that if A is a nonsingular matrix which can be written in the form LU , where $L = [v_{ij}]$ is a lower-triangular matrix satisfying $v_{ii} = 1$ for all $1 \leq i \leq n$ and $U = [u_{ij}]$ is upper-triangular, then this factorization must be unique. Indeed, assume that $L_1 U_1 = L_2 U_2$ where the L_h are lower triangular matrices with 1's on the diagonal, and the U_h are nonsingular upper-triangular matrices. Then $L_2^{-1} L_1 = U_2 U_1^{-1}$. Since the product of lower-triangular matrices is lower triangular and the product of upper-triangular matrices is upper triangular, this matrix must be a diagonal matrix. But then $L_2^{-1} L_1 = I$ and so $L_1 = L_2$ and that implies that $U_1 = U_2$, proving uniqueness.

Example: The above uniqueness result is no longer true if the matrix A is singular. For example,

$$\begin{bmatrix} 1 & -1 & 2 \\ -1 & 1 & -1 \\ 2 & -2 & 4 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ -1 & 1 & 0 \\ 2 & b & 1 \end{bmatrix} \begin{bmatrix} 1 & -1 & 2 \\ 0 & 0 & 1 \\ 0 & 0 & -b \end{bmatrix}$$

for any scalar $b \in \mathbb{R}$.

As was previously remarked, not all nonsingular matrices can be written in the form LU , but one can show that for any nonsingular matrix A there exists a permutation matrix P such that $A = PLU$, where L is lower triangular and U is upper triangular.

Example: It is easy to verify that $\begin{bmatrix} 0 & 1 & 1 & -3 \\ -2 & 4 & 1 & 4 \\ 0 & 0 & 0 & 1 \\ 3 & 1 & 1 & 0 \end{bmatrix} = PLU$, where

$$P = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 \end{bmatrix} \text{ is a permutation matrix, } L = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ -\frac{3}{2} & 7 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

is lower triangular, and $U = \begin{bmatrix} -2 & 4 & 1 & 4 \\ 0 & 1 & 1 & -3 \\ 0 & 0 & -\frac{9}{2} & 27 \\ 0 & 0 & 0 & 1 \end{bmatrix}$ is upper triangular.

Exercises

Exercise 390 Let $F = GF(5)$. Calculate $\begin{bmatrix} 1 & 3 & 1 \\ 2 & 1 & 1 \\ 1 & 2 & 3 \end{bmatrix} \begin{bmatrix} 1 & 2 & 2 \\ 4 & 3 & 2 \\ 1 & 4 & 2 \end{bmatrix}$ in $\mathcal{M}_{3 \times 3}(F)$.

Exercise 391 Does there exist a real number b such that the matrices

$$A = \begin{bmatrix} 1 & 0 & -1 & 0 \\ 0 & 1 & 0 & -1 \\ 1 & 0 & -1 & 0 \\ 0 & 1 & 0 & -1 \end{bmatrix} \quad \text{and} \quad B = \begin{bmatrix} b & -1 & -1 & 0 \\ -1 & b & 0 & -1 \\ 1 & 0 & \frac{b}{2} & -1 \\ 0 & 1 & -1 & \frac{b}{2} \end{bmatrix}$$

are a commuting pair in $\mathcal{M}_{4 \times 4}(\mathbb{R})$?

Exercise 392 Let $F = GF(7)$ and let K be the subalgebra of $\mathcal{M}_{2 \times 2}(F)$

consisting of all matrices of the form $\begin{bmatrix} a & b \\ -b & a \end{bmatrix}$, for $a, b \in F$. Show that K is a field. Is it a field if $F = GF(5)$?

Exercise 393 Let $A = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} \in \mathcal{M}_{2 \times 2}(\mathbb{R})$. Find the set of all matrices $B \in \mathcal{M}_{n \times n}(\mathbb{R})$ satisfying $BA = AB$.

Exercise 394 Let $A = \begin{bmatrix} 1 & i \\ -i & 1 \end{bmatrix} \in \mathcal{M}_{2 \times 2}(\mathbb{C})$. Find a complex number c satisfying $(cA)^2 = A$.

Exercise 395 Let $A = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{R})$. Find a positive integer k satisfying $A^k = A^{-1}$.

Exercise 396 Let F be a field. Find all matrices $A \in \mathcal{M}_{3 \times 3}(F)$ sat-

isfying $A^2 = \begin{bmatrix} 0 & 0 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$.

Exercise 397 Show that there are infinitely-many pairs (a, b) of real numbers satisfying the condition

$$\begin{bmatrix} a & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & b \end{bmatrix} \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 1 \end{bmatrix} \begin{bmatrix} a & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & b \end{bmatrix}.$$

Exercise 398 Does there exist a positive integer k satisfying

$$\begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \end{bmatrix} \begin{bmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{bmatrix}^k = \begin{bmatrix} 0 & 0 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}?$$

Exercise 399 Let $F = GF(3)$. Show that there exist at least 27 distinct matrices A in $\mathcal{M}_{3 \times 3}(F)$ satisfying $A^3 = I$.

Exercise 400 Let n be a positive integer and let F be a field of characteristic 0. Show that $AB - BA \neq I$ for all $A, B \in \mathcal{M}_{n \times n}(F)$ (in other words, that I is not the product of any two elements of the Lie algebra $\mathcal{M}_{n \times n}(F)^-$).

Exercise 401 If $F = GF(2)$, find the set of all pairs (A, B) of matrices in $\mathcal{M}_{2 \times 2}(F)$ satisfying $AB - BA = I$.

Exercise 402 For a field F , find $\{A \in \mathcal{M}_{2 \times 2}(F) \mid A^2 = O\}$.

Exercise 403 Find a matrix $A \in \mathcal{M}_{3 \times 3}(\mathbb{R})$ satisfying

$$A \begin{bmatrix} 1 & 1 & -1 \\ 2 & 1 & 0 \\ 1 & -1 & 1 \end{bmatrix} = \begin{bmatrix} 1 & -1 & 3 \\ 4 & 3 & 2 \\ 1 & -2 & 5 \end{bmatrix}.$$

Exercise 404 Show that if $A = \begin{bmatrix} a & 1 & 0 \\ 0 & a & 1 \\ 0 & 0 & a \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{R})$ then for each

$$n > 1 \text{ we have } A^n = \begin{bmatrix} a^n & na^{n-1} & \frac{n(n-1)}{2}a^{n-2} \\ 0 & a^n & na^{n-1} \\ 0 & 0 & a^n \end{bmatrix}.$$

Exercise 405 Let (K, \bullet) be an associative unital algebra over a field F and let S be the subset of $\mathcal{M}_{3 \times 3}(K)$ consisting of all matrices of the

form $\begin{bmatrix} v_{11} & 0_K & v_{13} \\ 0_K & v_{22} & 0_K \\ v_{31} & 0_K & v_{33} \end{bmatrix}$. Is S an F -subalgebra of $\mathcal{M}_{3 \times 3}(K)$?

Exercise 406 Let n be a positive integer and let F be a field. A matrix

in $\mathcal{M}_{n \times n}(F)$ of the form $\begin{bmatrix} a_1 & a_2 & \dots & a_n \\ a_n & a_1 & \dots & a_{n-1} \\ & & \dots & \\ a_2 & a_3 & \dots & a_1 \end{bmatrix}$ is called a **circulant**

matrix. Show that the set of all circulant matrices in $\mathcal{M}_{n \times n}(F)$ is an F -subalgebra of $\mathcal{M}_{n \times n}(F)$.

Exercise 407 Let n be a positive integer and let F be a field. If $A \in \mathcal{M}_{n \times n}(F)$ is a nonsingular circulant matrix, is A^{-1} necessarily a circulant matrix?

Exercise 408 Let K be the subset of $\mathcal{M}_{2 \times 2}(\mathbb{Q})$ consisting of all matrices of the form $\begin{bmatrix} a & b \\ 2b & a \end{bmatrix}$, where $a, b \in \mathbb{Q}$. Show that K is a \mathbb{Q} -subalgebra of $\mathcal{M}_{2 \times 2}(\mathbb{Q})$ which is, in fact, a field.

Exercise 409 Find a matrix $A \in \mathcal{M}_{2 \times 2}(\mathbb{R})$ satisfying $A^2 = \begin{bmatrix} 1 & 3 \\ 0 & 1 \end{bmatrix}$.

Exercise 410 Find all matrices $A \in \mathcal{M}_{3 \times 3}(\mathbb{R})$ satisfying

$$A \begin{bmatrix} 1 & 1 & 1 \\ 2 & 2 & 2 \\ 0 & 1 & 1 \end{bmatrix} = O.$$

Exercise 411 Let $A = \begin{bmatrix} a & b \\ c & d \end{bmatrix} \in \mathcal{M}_{2 \times 2}(\mathbb{R})$ be a matrix satisfying $A^2 = A$. Show that $a + d \in \{0, 1, 2\}$.

Exercise 412 Show that $\begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}^n = \begin{bmatrix} 2^{n-1} & 2^{n-1} \\ 2^{n-1} & 2^{n-1} \end{bmatrix}$ for all $n \geq 1$.

Exercise 413 Let F be a field and let $A = \begin{bmatrix} 0 & b \\ c & 0 \end{bmatrix} \in \mathcal{M}_{2 \times 2}(F)$. Find A^n for all $n \geq 1$.

Exercise 414 Find matrices $A, B \in \mathcal{M}_{2 \times 2}(\mathbb{Q})$ for which

$$(A - B)(A + B) \neq A^2 - B^2.$$

Exercise 415 Let F be a field and let $A = \begin{bmatrix} 1 & 0 & 0 \\ 1 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix} \in \mathcal{M}_{3 \times 3}(F)$.

Show that $A^{k+2} = A^k + A^2 - I$ for all positive integers k .

Exercise 416 Let n be a positive integer and let F be a field. Let $A, B \in \mathcal{M}_{n \times n}(F)$ satisfy $A + B = I$. Show that $AB = O$ if and only if $A = A^2$ and $B = B^2$.

Exercise 417 Let n be a positive integer and let (K, \bullet) be an associative unital algebra over a field F . Define a new operation \square on $\mathcal{M}_{n \times n}(K)$, called the **Schur product** (sometimes also called the **Hadamard product**, especially in the context of statistics), by setting $[v_{ij}] \square [w_{ij}] = [v_{ij} \bullet w_{ij}]$, for all $1 \leq i, j \leq n$. Is $(\mathcal{M}_{n \times n}(K), +, \square)$ an F -algebra? Is it associative? Is it unital? When is it commutative?

Exercise 418 Let n be a positive integer and for each $A = [a_{ij}] \in \mathcal{M}_{n \times n}(\mathbb{R})$, let $\mu(A) = \max_{1 \leq i, j \leq n} |a_{ij}|$. Show that $\mu(A^2) \leq n\mu(A)^2$ for all $A \in \mathcal{M}_{n \times n}(\mathbb{R})$.

Exercise 419 Let F be a field. Find a matrix $A \in \mathcal{M}_{3 \times 3}(F)$ satisfying $A^2 = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$ or show that no such matrix exists.

Exercise 420 Find a matrix $A \in \mathcal{M}_{2 \times 2}(\mathbb{Q})$ satisfying $A \begin{bmatrix} 0 & c \\ c & 0 \end{bmatrix} A^T = \begin{bmatrix} 2c & 0 \\ 0 & -\frac{c}{2} \end{bmatrix}$ for all $c \in \mathbb{Q}$.

Exercise 421 Let F be a field and let n be a positive integer. Show that $H_{11}AH_{11}BH_{11} = H_{11}BH_{11}AH_{11}$ for all $A, B \in \mathcal{M}_{n \times n}(F)$.

Exercise 422 Let F be a field and let n be a positive integer. Show that $\left(\sum_{i=1}^n \sum_{j=1}^n H_{ij}AH_{ji}\right)B = B\left(\sum_{i=1}^n \sum_{j=1}^n H_{ij}AH_{ji}\right)$ for all $A, B \in \mathcal{M}_{n \times n}(F)$.

Exercise 423 Is the set

$$\left\{ \begin{bmatrix} 1 & 1 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}, \begin{bmatrix} 1 & 1 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 & 0 \\ 1 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 & 0 \\ 1 & 1 & 1 \\ 0 & 0 & 0 \end{bmatrix} \right\}$$

of matrices in $\mathcal{M}_{3 \times 3}(\mathbb{Q})$ closed under taking products?

Exercise 424 Find infinitely-many triples (A, B, C) of nonzero matrices in $\mathcal{M}_{3 \times 3}(\mathbb{Q})$, the entries of which are nonnegative integers, satisfying the condition $A^3 + B^3 = C^3$.

Exercise 425 Let F be a field. Find a matrix $A \in \mathcal{M}_{4 \times 4}(F)$ satisfying $A^4 = I \neq A^3$.

Exercise 426 Let n be a positive integer and let $F = GF(p)$ for some prime integer p . Show that for any $A \in \mathcal{M}_{n \times n}(F)$ there exist positive integers $k > h$ satisfying $A^k = A^h$. Would this also be true if we chose $F = \mathbb{Q}$?

Exercise 427 Let $A = [a_{ij}] \in \mathcal{M}_{2 \times 2}(\mathbb{C})$ be a matrix satisfying

$$\frac{1}{2} [a_{11} + a_{22}] \neq \sqrt{a_{11}a_{22} - a_{12}a_{21}}.$$

Show that there exist four distinct matrices $B \in \mathcal{M}_{2 \times 2}(\mathbb{C})$ satisfying $B^2 = A$.

Exercise 428 Let c be a given complex number. Find the set of all matrices $A \in \mathcal{M}_{2 \times 2}(\mathbb{C})$ satisfying $(A - cI)^2 = O$.

Exercise 429 Show that $\begin{bmatrix} 3-4c & 2-4c & 2-4c \\ -1+2c & 2c & -1+2c \\ -3+2c & -3+2c & -2+2c \end{bmatrix}^2 = I$ for all complex numbers c .

Exercise 430 Show that there is no matrix $A \in \mathcal{M}_{2 \times 2}(\mathbb{C})$ satisfying

$$A^2 = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}.$$

Exercise 431 Let n be a positive integer and let F be a field. How many matrices $A = [a_{ij}] \in \mathcal{M}_{n \times n}(F)$ having entries in $\{0, 1\}$ satisfy the condition that each row and each column contain exactly one 1.

Exercise 432 Show that for an integer $n \geq 4$ and for a field F there exist matrices A and B in $\mathcal{M}_{n \times n}(F)$ satisfying $A^2 = B^2 = O$ but $AB = BA \neq O$.

Exercise 433 Let $F = GF(2)$ and let F' be a field of characteristic other than 2. Define a function $\varphi: \mathcal{M}_{2 \times 2}(F') \rightarrow \mathcal{M}_{2 \times 2}(F)$ as follows: if $A = [a_{ij}] \in \mathcal{M}_{2 \times 2}(F')$ then set $\varphi(A) = [b_{ij}]$, where

$$b_{ij} = \begin{cases} 1 & \text{if } a_{ij} \neq 0 \\ 0 & \text{otherwise} \end{cases}.$$

Is $\varphi(A + A') = \varphi(A) + \varphi(A')$ for all $A, A' \in \mathcal{M}_{2 \times 2}(F')$? Is $\varphi(AA') = \varphi(A)\varphi(A')$ for all $A, A' \in \mathcal{M}_{2 \times 2}(F')$?

Exercise 434 Find a matrix $I \neq A \in \mathcal{M}_{3 \times 3}(\mathbb{Q})$ satisfying

$$A \begin{bmatrix} 1 & 0 & 0 \\ 1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} A \quad \text{and} \quad A \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 1 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 1 & 1 \end{bmatrix} A.$$

Exercise 435 For each real number a , find a matrix $B(a) \in \mathcal{M}_{2 \times 2}(\mathbb{R})$

$$\text{satisfying } \begin{bmatrix} \cos(a) & -\sin(a) \\ \sin(a) & \cos(a) \end{bmatrix} = B(a) \begin{bmatrix} 1 & 0 \\ \sin(a) & 1 \end{bmatrix}.$$

Exercise 436 Let $A = \begin{bmatrix} 5 & 7 \\ -3 & -4 \end{bmatrix} \in \mathcal{M}_{2 \times 2}(\mathbb{R})$. What is A^{1024} ?

Exercise 437 Find all pairs (a, b) of rational numbers such that the

$$\text{matrix } A = \begin{bmatrix} 2a & -a \\ 2b & -b \end{bmatrix} \in \mathcal{M}_{2 \times 2}(\mathbb{Q}) \text{ satisfies } A^2 = A.$$

Exercise 438 Let F be a field and let n be a positive integer. Show that there do not exist nonsingular matrices $P, Q \in \mathcal{M}_{n \times n}(F)$ satisfying $PAQ = A^T$ for all $A \in \mathcal{M}_{n \times n}(F)$.

Exercise 439 Let F be a field and let $A, B \in \mathcal{M}_{n \times n}(F)$ be a commuting pair of matrices, where B is nonsingular. Is (A, B^{-1}) necessarily a commuting pair?

Exercise 440 Let F be a field. Is $S = \left\{ \begin{bmatrix} a & b \\ c & d \end{bmatrix} \mid a + c = b + d \right\}$ an F -subalgebra of $\mathcal{M}_{2 \times 2}(F)$?

Exercise 441 Let F be a field. Given an element a of F , show that

there exist elements $b, c, d \in F$ such that $\begin{bmatrix} a & b \\ c & d \end{bmatrix}^2 = I$ in $\mathcal{M}_{2 \times 2}(F)$.

Exercise 442 Find all rational numbers a , b and d satisfying the

condition that $\begin{bmatrix} a & b \\ 1 & d \end{bmatrix}^2 = I$ in $\mathcal{M}_{2 \times 2}(\mathbb{Q})$.

Exercise 443 Let F be a field and let K be the subset of $\mathcal{M}_{3 \times 3}(F)$ consisting of O and of all of the upper-triangular matrices $[a_{ij}]$ satisfying $0 \neq a_{11} = a_{22} = a_{33}$. Is K an F -subalgebra of $\mathcal{M}_{3 \times 3}(F)$? Is it a field?

Exercise 444 Let $F = GF(p)$, where p is a prime integer, and let K be the subset of $\mathcal{M}_{2 \times 2}(F)$ consisting of all matrices of the form

$\begin{bmatrix} a & b \\ -b & a \end{bmatrix}$, where $a, b \in F$. Show that K , together with the operations of matrix addition and multiplication, is a field when $p = 3$ and is not a field when $p = 5$. What happens when $p = 7$?

Exercise 445 Let n be a positive integer, let F be a field, and let $O \neq A, B \in \mathcal{M}_{n \times n}(F)$. Show that there exists a matrix $C \in \mathcal{M}_{n \times n}(F)$ satisfying $ACB \neq O$.

Exercise 446 Find all matrices $A, B \in \mathcal{M}_{2 \times 2}(\mathbb{R})$, the entries of which are nonnegative integers, which satisfy $AB = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix}$.

Exercise 447 Let $V = \mathcal{M}_{3 \times 3}(\mathbb{Q})$. For each rational number t , let

$\alpha_t : V \rightarrow V$ be the linear transformation $A \mapsto A \begin{bmatrix} 0 & 1 & 3 \\ t & 0 & 0 \\ 0 & -1 & 4 \end{bmatrix}$. Is

the function $t \mapsto \alpha_t$ a linear transformation from \mathbb{Q} to $\text{End}(V)$, both considered as vector spaces over \mathbb{Q} ?

Exercise 448 Let n be a positive integer, let F be a field, and for some fixed $c \in F$, let $A = [a_{ij}]$ be the matrix in $\mathcal{M}_{n \times n}(F)$ defined by

$$a_{ij} = \begin{cases} c & \text{when } i+j \text{ is even} \\ 0 & \text{otherwise} \end{cases}.$$

Show that the subset $\{A, A^2, A^3\}$ of $\mathcal{M}_{n \times n}(F)$ is linearly dependent.

Exercise 449 Let $F = GF(2)$ and let $A = \begin{bmatrix} 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 1 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \in \mathcal{M}_{4 \times 4}(F)$.

Let $L = \{O\} \cup \{A^i \mid i \geq 0\} \subseteq \mathcal{M}_{4 \times 4}(F)$. Show that L is closed under addition. Is L , under the usual definitions of addition and multiplication of matrices, a field?

Exercise 450 Let K be the set of all matrices in $\mathcal{M}_{2 \times 2}(\mathbb{Q})$ of the form $\begin{bmatrix} a & -3b \\ b & a \end{bmatrix}$. Show that K is a subalgebra of $\mathcal{M}_{2 \times 2}(\mathbb{Q})$ which is in fact a field.

Exercise 451 Find the set of all matrices $A \in \mathcal{M}_{2 \times 2}(\mathbb{Q})$ which satisfy $A^2 + A = \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}$.

Exercise 452 Let $A = \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix} \in \mathcal{M}_{2 \times 2}(\mathbb{Q})$ and let B and C be matrices in $\mathcal{M}_{2 \times 2}(\mathbb{Q})$ satisfying $AB = BA$ and $AC = CA$. Show that $BC = CB$.

Exercise 453 Find infinitely-many matrices $A \in \mathcal{M}_{3 \times 3}(\mathbb{Q})$ satisfying $A \begin{bmatrix} 1 & -1 & 2 \\ 2 & 0 & 1 \\ 3 & -1 & 3 \end{bmatrix} = \begin{bmatrix} 1 & 0 & \frac{1}{2} \\ 0 & 1 & -\frac{3}{2} \\ 0 & 0 & 0 \end{bmatrix}$.

Exercise 454 Let $A = \begin{bmatrix} 1 & -1 & -1 \\ -1 & 1 & -1 \\ -1 & -1 & 1 \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{Q})$. Find functions f and g from the set of all positive integers to \mathbb{Q} satisfying the condition

$$\text{that } A^n = \begin{bmatrix} f(n) & g(n) & g(n) \\ g(n) & f(n) & g(n) \\ g(n) & g(n) & f(n) \end{bmatrix} \quad \text{for all } n \geq 1.$$

Exercise 455 Let $F = GF(2)$. Do there exist matrices $A = [a_{ij}]$ and $B = [b_{ij}]$ in $\mathcal{M}_{2 \times 2}(F)$ satisfying $a_{11} + a_{22} = 1$, $b_{11} + b_{22} = 0$, and $AB = I$?

Exercise 456 Let F be a field and let G be the set of all matrices in

$\mathcal{M}_{3 \times 3}(F)$ of the form $\begin{bmatrix} 1 & 0 & 0 \\ a & 0 & 0 \\ 0 & 0 & b \end{bmatrix}$, where $a, b \in F$. Is G closed under

matrix multiplication? Does there exist a matrix J in G satisfying the condition that $AJ = A$ for all $A \in G$? If such a matrix J exists, is it necessarily true that $JA = A$ for all $A \in G$?

Exercise 457 Let n be a positive integer and let F be a field. Let

A and B be matrices in $\mathcal{M}_{n \times n}(F)$ of the form $\begin{bmatrix} I & A' \\ O & I \end{bmatrix}$ and

$\begin{bmatrix} I & B' \\ O & I \end{bmatrix}$ respectively, where A' and B' are (not-necessarily square)

matrices of the same size. Find necessary conditions for A and B to satisfy $AB = BA$.

Exercise 458 Let F be a field and let $A, B \in \mathcal{M}_{2 \times 2}(F)$. Show that $(AB - BA)^2$ is a diagonal matrix.

Exercise 459 Let n be a positive integer and let F be a field. Let $A \in \mathcal{M}_{n \times n}(F)$ be a diagonal matrix having distinct entries on the diagonal. Let $B \in \mathcal{M}_{n \times n}(F)$ be a matrix satisfying $AB = BA$. Show that B is also a diagonal matrix.

Exercise 460 Let n be a positive integer and let F be a field. For each integer $-n < t < n$, let $D_t(F)$ be the set of all matrices $A = [a_{ij}] \in \mathcal{M}_{n \times n}(F)$ satisfying the condition that $a_{ij} = 0$ when $j \neq i + t$. Thus, for example, $D_0(F)$ is the set of all diagonal matrices in $\mathcal{M}_{n \times n}(F)$. If $A \in D_t(F)$ and $B \in D_s(F)$, does there necessarily exist an integer $-n < u < t$ such that $AB \in D_u(F)$?

Exercise 461 Let $A = \begin{bmatrix} 1 & 2 & 3 \\ -1 & -2 & -3 \\ 2 & 4 & 6 \end{bmatrix}$ and $B = \begin{bmatrix} 1 & 2 & 3 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$ be

matrices in $\mathcal{M}_{3 \times 3}(\mathbb{R})$. Find infinitely-many lower-triangular matrices C satisfying $A = CB$.

Exercise 462 Let n be a positive integer and let F be a field. Let A_1, \dots, A_n be upper-triangular matrices in $\mathcal{M}_{n \times n}(F)$ satisfying the condition that the (i, i) -entry in A_i is equal to 0 for $1 \leq i \leq n$. Show that $A_1 \cdot \dots \cdot A_n = O$.

Exercise 463 Let F be a field in which we have elements $a \neq 0$ and b . Show that there exists an upper-triangular matrix $C \in \mathcal{M}_{2 \times 2}(F)$ satisfying

$$\begin{bmatrix} 0 & a \\ 0 & 0 \end{bmatrix} C = \begin{bmatrix} 0 & b \\ 0 & 0 \end{bmatrix}. \text{ Is } C \text{ necessarily unique?}$$

Exercise 464 Let F be a field. Find an element A of $\mathcal{M}_{2 \times 2}(F)$ satisfying $AA^T \neq A^T A$.

Exercise 465 Let F be a field and let $n > 1$. If a matrix $A \in \mathcal{M}_{n \times n}(F)$ satisfies $AA^T = O$, does it necessarily follow that $A^T A = O$?

Exercise 466 Let n be a positive integer, let F be a field, and let $A \in \mathcal{M}_{n \times n}(F)$ satisfy the condition $A = AA^T$. Show that $A^2 = A$.

Exercise 467 Let n be a positive integer, let F be a field, and let $A, B \in \mathcal{M}_{n \times n}(F)$ be symmetric matrices. Is ABA necessarily symmetric?

Exercise 468 Let n be a positive integer and let F be a field. If $A \in \mathcal{M}_{n \times n}(F)$ is symmetric, is A^h symmetric for all $h > 1$?

Exercise 469 Show that $\left\{ \begin{bmatrix} 1 & -2 \\ -2 & 1 \end{bmatrix}, \begin{bmatrix} 1 & 3 \\ 3 & 6 \end{bmatrix}, \begin{bmatrix} -1 & 1 \\ 1 & -3 \end{bmatrix} \right\}$ forms a basis for the subspace of $\mathcal{M}_{2 \times 2}(\mathbb{Q})$ consisting of all symmetric matrices.

Exercise 470 Does there exist a matrix $A \in \mathcal{M}_{2 \times 2}(\mathbb{R})$ satisfying $AA^T = \begin{bmatrix} 1 & 9 \\ 9 & 1 \end{bmatrix}$?

Exercise 471 Given real numbers a, b , and c , find all real numbers d

such that $\begin{bmatrix} 0 & 0 & 0 & -1 \\ 0 & 0 & -1 & a \\ 0 & -1 & a & b \\ -1 & a & b & c \end{bmatrix} \begin{bmatrix} a & b & c & 1 \\ 1 & 0 & 0 & 0 \\ 0 & d & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix}$ is symmetric.

Exercise 472 Find a matrix $B \in \mathcal{M}_{2 \times 2}(\mathbb{Q})$ such that the Nievergelt's matrix equals $B^T B$.

Exercise 473 Calculate $\begin{bmatrix} 1 & 2 & -3 \\ 0 & 1 & 2 \\ 0 & 0 & 1 \end{bmatrix}^{-1}$ in $\mathcal{M}_{3 \times 3}(\mathbb{R})$.

Exercise 474 Let $a \in \mathbb{R} \setminus \{1, -2\}$. Calculate $\begin{bmatrix} a & 1 & 1 \\ 1 & a & 1 \\ 1 & 1 & a \end{bmatrix}^{-1} \in \mathcal{M}_{3 \times 3}(\mathbb{R})$.

Exercise 475 Does there exist an a such that $\begin{bmatrix} -3 & 4 & 0 \\ 8 & 5 & -2 \\ a & -7 & 6 \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{R})$ is singular?

Exercise 476 Let n be a positive integer. Each complex number c defines a matrix $A(c) = [a_{ij}] \in \mathcal{M}_{n \times n}(\mathbb{C})$ given by $a_{ij} = c^{(i-1)(j-1)}$ for all $1 \leq i, j \leq n$. If $w = e^{2\pi i/n} \in \mathbb{C}$, show that $A(w)$ is nonsingular and satisfies $A(w)^{-1} = \frac{1}{n}A(w^{-1})$.

Exercise 477 Let n be a positive integer and let F be a field. Given a matrix $B \in \mathcal{M}_{n \times n}(F)$, do there exist vectors $u, v \in F^n$ such that the matrix $\begin{bmatrix} B & -Bv \\ -u^T B & u^T Bv \end{bmatrix}$ is nonsingular?

Exercise 478 Is the matrix $\begin{bmatrix} 1-a^2 & 1-a & 0 \\ 0 & 1-a^2 & 1-a \\ 1-a & 0 & 1-a^2 \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{C})$ nonsingular, where $a = -\frac{1}{2} + \frac{1}{2}\sqrt{-3} \in \mathbb{C}$.

Exercise 479 Let n be a positive integer and let F be a field. If $A \in \mathcal{M}_{n \times n}(F)$ is nonsingular, is the same necessarily true for $A + A^T$?

Exercise 480 Let n be a positive integer and let F be a field. Let $A = [a_{ij}] \in \mathcal{M}_{n \times n}(F)$ satisfy the condition that $\sum_{i=1}^n a_{ij} = 1$ for all $1 \leq j \leq n$. Show that the matrix $I - A$ is singular.

Exercise 481 Let n be a positive integer and let F be a field. If $A \in \mathcal{M}_{n \times n}(F)$ is a Markov matrix, is A^{-1} necessarily a Markov matrix?

Exercise 482 Let n be a positive integer and let F be a field. For $A \in \mathcal{M}_{n \times n}(F)$, show that A^2 is nonsingular if and only if A^3 is nonsingular.

Exercise 483 Let $F = GF(p)$, where p is a prime integer, and let n be a positive integer. What is the probability that a matrix in $\mathcal{M}_{n \times n}(F)$, chosen at random, is nonsingular?

Exercise 484 Let $P = \begin{bmatrix} 0 & -1 \\ 1 & -1 \end{bmatrix} \in \mathcal{M}_{2 \times 2}(\mathbb{R})$ and let A and Q be nonsingular matrices in $\mathcal{M}_{2 \times 2}(\mathbb{R})$. Set $B = AQ^{-1}PQ$. Show that B is nonsingular and $A^{-1} + B^{-1} = (A + B)^{-1}$.

Exercise 485 Show that there are infinitely-many matrices $A \in \mathcal{M}_{2 \times 2}(\mathbb{Q})$ satisfying $A = A^{-1}$.

Exercise 486 Let $F = GF(2)$. Is the sum of all nonsingular matrices in $\mathcal{M}_{2 \times 2}(F)$ nonsingular?

Exercise 487 Let F be a field and let U be the set of all nonsingular matrices in $\mathcal{M}_{2 \times 2}(F)$. Is the function $\theta : U \rightarrow U$ defined by $\theta : A \mapsto A^2$ a permutation of U ?

Exercise 488 Let n be a positive integer, let F be a field, and let $A \in \mathcal{M}_{2n \times 2n}(F)$ be a matrix which can be written in the form $\begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix}$, where each $A_{ij} \in \mathcal{M}_{n \times n}(F)$ is nonsingular. Is A necessarily nonsingular?

Exercise 489 Let n be a positive integer and let F be a field. Do there exist matrices $A, B \in \mathcal{M}_{n \times n}(F)$ such that the matrix $\begin{bmatrix} A^2 & AB \\ BA & B^2 \end{bmatrix} \in \mathcal{M}_{2n \times 2n}(F)$ is nonsingular?

Exercise 490 Let n be a positive integer and let F be a field. For $A, B \in \mathcal{M}_{n \times n}(F)$ with A nonsingular, show that

$$(A + B)A^{-1}(A - B) = (A - B)A^{-1}(A + B).$$

Exercise 491 Let n and p be positive integers and let F be a field. Let $A \in \mathcal{M}_{n \times n}(F)$ and let $B, C \in \mathcal{M}_{n \times p}(F)$ be matrices satisfying the condition that A and $(I + C^T A^{-1} B)$ are nonsingular. Show that $A + BC^T$ is nonsingular, and that

$$(A + BC^T)^{-1} = A^{-1} - A^{-1}B(I + C^T A^{-1} B)^{-1}C^T A^{-1}.$$

Exercise 492 Let n be a positive integer and let F be a field. If

$$\begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix} \neq v \in F^n, \text{ show that there exists a nonsingular matrix in } \mathcal{M}_{n \times n}(F)$$

the last row of which is v .

Exercise 493 Let F be a field. Show that every nonsingular matrix in

$$\mathcal{M}_{2 \times 2}(F) \text{ can be written as a product of matrices of the form } \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix},$$

$$\begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix}, \text{ or } \begin{bmatrix} a & 0 \\ 0 & 1 \end{bmatrix} \text{ for } a \in F.$$

Exercise 494 For each real number t , let $A(t) = \begin{bmatrix} 1 & 0 & t \\ -t & 1 & -\frac{1}{2}t^2 \\ 0 & 0 & 1 \end{bmatrix} \in$

$\mathcal{M}_{3 \times 3}(\mathbb{R})$. Show that each such matrix is nonsingular and that the set of all such matrices is closed under taking products.

Exercise 495 Let n be a positive integer and let F be a field. Let $A \in \mathcal{M}_{n \times n}(F)$ be a matrix for which there exists a positive integer k satisfying $A^k = O$. Show that the matrix $I - A$ is nonsingular and find $(I - A)^{-1}$.

Exercise 496 Let n be a positive integer and let F be a field. Let $A \in \mathcal{M}_{n \times n}(F)$ be a matrix for which there exists a matrix $B \in \mathcal{M}_{n \times n}(F)$ satisfying $I + A + AB = O$. Show that A is nonsingular.

Exercise 497 Let n be a positive integer and let F be a field. Let $A, B \in \mathcal{M}_{n \times n}(F)$ satisfy the condition that A and $A+B$ are nonsingular. Show that $I + A^{-1}B$ is nonsingular and that $(I + A^{-1}B)^{-1} = (A + B)^{-1}A$.

Exercise 498 Find matrices A and B in $\mathcal{M}_{2 \times 2}(\mathbb{R})$ satisfying $A^2 = B^2 = O$ and that $A + iB$ is a nonsingular matrix in $\mathcal{M}_{2 \times 2}(\mathbb{C})$.

Exercise 499 Let F be a field and let $A = \begin{bmatrix} 1 & 0 & b \\ 0 & 1 & 0 \\ a & 0 & 1 \end{bmatrix} \in \mathcal{M}_{3 \times 3}(F)$,

where $ab \neq 1$. Show that A is nonsingular and calculate A^{-1} .

Exercise 500 Let $c \neq 0$ be an element of a field F and let $A =$

$\begin{bmatrix} c & 1 & 0 & 0 \\ 0 & c & 1 & 0 \\ 0 & 0 & c & 1 \\ 0 & 0 & 0 & c \end{bmatrix} \in \mathcal{M}_{4 \times 4}(F)$. Show that A is nonsingular and find A^{-1} .

Exercise 501 Let $n > 1$ and let $B \in \mathcal{M}_{n \times n}(\mathbb{Q})$ be the matrix all of the entries of which are equal to 1. Show that there exists a matrix $A \in \mathcal{M}_{n \times n}(\mathbb{Q})$ satisfying the condition that $A + cB$ is nonsingular for all rational numbers c .

Exercise 502 Let $n > 1$ and let $B \in \mathcal{M}_{n \times n}(\mathbb{Q})$ be the matrix all of the entries of which are equal to 1. Find a rational number t such that $(I - B)^{-1} = I - tB$.

Exercise 503 Let n be a positive integer and let $A = [a_{ij}] \in \mathcal{M}_{n \times n}(\mathbb{R})$ be the matrix defined by $a_{ij} = \min\{i, j\}$ for all $1 \leq i, j \leq n$. Show that A is nonsingular.

Exercise 504 Let $A = [a_{ij}] \in \mathcal{M}_{4 \times 4}(\mathbb{R})$ be the matrix defined by

$$a_{ij} = \begin{cases} 2 & \text{if } i = j - 1 \\ 1 & \text{otherwise} \end{cases}.$$

Show that A is nonsingular and calculate A^{-1} .

Exercise 505 For each real number a , let $G(a) = \begin{bmatrix} \cos(a) & \sin(a) \\ -\sin(a) & \cos(a) \end{bmatrix} \in \mathcal{M}_{2 \times 2}(\mathbb{R})$. Given real numbers a , b , and c , show that $G(a, b, c) =$

$\begin{bmatrix} G(a) & G(b) \\ O & G(c) \end{bmatrix} \in \mathcal{M}_{4 \times 4}(\mathbb{R})$ is nonsingular, and find $G(a, b, c)^{-1}$.

Exercise 506 Find a singular matrix in $\mathcal{M}_{3 \times 3}(\mathbb{Q})$ the entries of which (in some order) are the integers $1, 2, \dots, 9$.

Exercise 507 Let n be a positive integer and let F be a field. Given elements $b, c \in F$, let $A = [a_{ij}] \in \mathcal{M}_{n \times n}(F)$ be the matrix defined by

$$a_{ij} = \begin{cases} b & \text{if } i = j \\ c & \text{otherwise} \end{cases}.$$

Find necessary and sufficient conditions for A to be nonsingular.

Exercise 508 Let F be a field and let $D = \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix} \in \mathcal{M}_{2 \times 2}(F)$.

Let $A \in \mathcal{M}_{2 \times 2}(F)$ satisfy the condition that $A^T D A = D$. Show that A is nonsingular.

Exercise 509 Let n be a positive integer and let F be a field. Is the set of all singular matrices in $\mathcal{M}_{n \times n}(F)$ closed under taking products?

Exercise 510 Let n be a positive integer, let F be a field, and let $A, B \in \mathcal{M}_{n \times n}(F)$. Show that A and B are both nonsingular if and only

if the matrix $\begin{bmatrix} A & O \\ O & B \end{bmatrix} \in \mathcal{M}_{(2n) \times (2n)}(F)$ is nonsingular.

Exercise 511 Let $A = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{Q})$. Find a rational num-

ber c satisfying $(I - A)^{-1} = I - cA$.

Exercise 512 Write the matrix $\begin{bmatrix} 1 & -2 \\ 2 & 2 \end{bmatrix} \in \mathcal{M}_{2 \times 2}(\mathbb{R})$ as a product of elementary matrices.

Exercise 513 Find the change of basis matrix from the canonical basis B

of \mathbb{R}^3 to the basis $D = \left\{ \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \right\}$ and the change of basis matrix from D to B .

Exercise 514 Let $G = \left\{ \begin{bmatrix} a & a \\ a & a \end{bmatrix} \mid 0 \neq a \in \mathbb{R} \right\}$. Show that there exists a matrix $E \in G$ satisfying the condition that $EA = A = AE$ for all $A \in G$. For each $A \in G$, show that there exists a matrix $A^\dagger \in G$ satisfying $AA^\dagger = E = A^\dagger A$.

Exercise 515 Let F be a field. Given matrices $A, B \in \mathcal{M}_{2 \times 2}(F)$, find the set of all matrices $C \in \mathcal{M}_{2 \times 2}(F)$ satisfying $(AB - BA)C = C(AB - BA)$.

Exercise 516 Let F be a field and let G be the set of all automorphisms of F^2 which are represented with respect to the canonical basis by a matrix

of the form $\begin{bmatrix} a & b \\ 0 & a^{-1} \end{bmatrix}$. Is G a group of automorphisms of F^2 ?

Exercise 517 Let G be the set of all automorphisms of \mathbb{Q}^2 which are represented with respect to the canonical basis by a matrix of the form

$\begin{bmatrix} a & b \\ 0 & d \end{bmatrix}$, where $a, d > 0$. Is G a group of automorphisms of \mathbb{Q}^2 ?

Exercise 518 Let $W_1 \subseteq W_2 \subseteq \dots \subseteq W_n$ be a fixed sequence of subspaces of a vector space V finitely generated over a field F . If $\alpha \in \text{Aut}(V)$, we say that given sequence is an α -**fan** if and only if each of the W_i is invariant under α . Show that $G = \{\alpha \in \text{Aut}(V) \mid \text{the given sequence is an } \alpha\text{-fan}\}$ is a group of automorphisms of V .

Exercise 519 For any real number t and any positive integer n , we can define the matrix $P(n, t) \in \mathcal{M}_{n \times n}(\mathbb{R})$ to equal the identity matrix I in the case $t = 0$ and otherwise to equal the matrix $[p_{ij}]$ defined by

$$p_{ij} = \begin{cases} 0 & \text{if } i < j \\ \binom{i-1}{j-1} t^{i-j} & \text{otherwise} \end{cases}.$$

Show that $P(n, s)P(n, t) = P(n, s + t)$ for all $s, t \in \mathbb{R}$. In particular, show that each matrix $P(n, t)$ is nonsingular.

Exercise 520 Let F be a field and let X be an indeterminate over F . Find matrices P and Q in $\mathcal{M}_{2 \times 2}(F[X])$ such that the matrix

$P \begin{bmatrix} 1 + X^2 & X \\ X & 1 + X \end{bmatrix} Q$ is a diagonal matrix.

Exercise 521 Let n be a positive integer and let $\alpha : \mathcal{M}_n(\mathbb{C}) \rightarrow \mathcal{M}_{2n}(\mathbb{R})$ be the function defined by

$$\alpha : \begin{bmatrix} a + bi & c + di \\ e + fi & g + hi \end{bmatrix} \mapsto \begin{bmatrix} a & b & c & d \\ -b & -a & -d & c \\ e & f & g & h \\ -f & e & -h & g \end{bmatrix}.$$

Show that α is a linear transformation of vector spaces over \mathbb{R} . Is it a homomorphism of unital \mathbb{R} -algebras?

10

Systems of linear equations

The classical problem of linear algebra is to find all solutions (if any exist) to a **system of linear equations in n unknowns** of the form

$$\begin{aligned}a_{11}X_1 + \dots + a_{1n}X_n &= b_1 \\a_{21}X_1 + \dots + a_{2n}X_n &= b_2 \\&\vdots \\a_{k1}X_1 + \dots + a_{kn}X_n &= b_k\end{aligned}$$

where the a_{ij} and the b_i are scalars belonging to some field F and the X_j are variables which take values in F .

Example: Let $a < b$ be real numbers and let $V = C(a, b)$. If W is a subspace of V of dimension n then the **interpolation problem** of V is the following: given a function $f \in V$ and given real numbers $a \leq t_1 < \dots < t_n \leq b$, find a function $g \in W$ satisfying $f(t_j) = g(t_j)$ for $1 \leq j \leq n$. If we are given a basis $\{g_1, \dots, g_n\}$ of W then we want to find real numbers c_1, \dots, c_n satisfying $\sum_{i=1}^n c_i g_i(t_j) = f(t_j)$ for all $1 \leq j \leq n$. In other words, we want to solve a system of linear equations of the above form, where $k = n$, $a_{ij} = g_j(t_i)$ and $b_i = f(t_i)$ for all $1 \leq i, j \leq n$.

Example: In Proposition 4.2 we noted that if F is a field and if $f(X)$ and $g(X) \neq 0$ are elements of $F[X]$, then there exist unique polynomials $u(X)$ and $v(X)$ in $F[X]$ satisfying $f(X) = g(X)u(X) + v(X)$ and

$\deg(v) < \deg(g)$. If we set $g(X) = \sum_{i=0}^k a_i X^i$ and $f(x) = \sum_{i=1}^n b_i X^i$, then the coefficients of $u(X) = \sum_{i=0}^{n-k} c_i X^i$ are found by solving the system of linear equations

$$\begin{aligned} a_k Y_0 + a_{k-1} Y_1 + \dots + a_0 Y_k &= b_k \\ a_k Y_1 + a_{k-1} Y_2 + \dots + a_0 Y_{k+1} &= b_{k-1} \\ &\vdots \\ a_k Y_{n-k-1} + a_{k-1} Y_{n-k} &= b_{n-1} \\ a_k Y_{n-k} &= b_n \end{aligned}$$

by any of the methods we will discuss.

Example: Sometimes we can transform systems of nonlinear equations into systems of linear equations. For example, suppose that we want to find positive real numbers r_1 , r_2 , and r_3 satisfying the following nonlinear system of equations:

$$\begin{aligned} r_1 r_2 r_3 &= 1 \\ r_1^3 r_2^2 r_3^2 &= 27 \\ r_3 / r_1 r_2 &= 81 \end{aligned}$$

Since each of the integers on the right is a power of 3, we can take the logarithm to the base 3 of both sides of each equation. Setting $X_i = \log_3(r_i)$ for $1 \leq i \leq 3$, the system now becomes linear

$$\begin{aligned} X_1 + X_2 + X_3 &= 0 \\ 3X_1 + 2X_2 + 2X_3 &= 3 \\ -X_1 - X_2 + X_3 &= 4 \end{aligned}$$

and this has a unique solution (which we can find by methods to be discussed in this chapter) $X_1 = 3$, $X_2 = -5$, and $X_3 = 2$, showing that the original system has a solution $r_1 = 27$, $r_2 = 1/243$, and $r_3 = 9$.

A system of linear equations of the above form is **homogeneous** if and only if $b_i = 0$ for all $1 \leq i \leq k$; otherwise it is **nonhomogeneous**. At this stage, we do not yet know answers to the following questions:

- (1) Does a given system of linear equations have a solution?
- (2) If it has a solution, is that solution unique?
- (3) If the solution is not unique, can we characterize the set of all solutions?
- (4) If there are solutions, how do we compute them efficiently?

In order to answer these questions, we have to move to the language of matrices. The use of matrices for this purpose was developed in Europe in the 19th century by Cayley, Sylvester, and Laguerre. However, the real pioneers were the Chinese and Japanese mathematicians. During the time

of the Han dynasty in China, around 2000 years ago, the classical book *Nine Chapters on the Mathematical Art* (Jiuzhang Suanshu,) presented a method for solving systems of linear equations using matrices. This was later expanded by the Japanese mathematician Seki¹.

How is this done? Let us write the above system in the form

$$\begin{bmatrix} a_{11} & \cdots & a_{1n} \\ \vdots & & \vdots \\ a_{k1} & \cdots & a_{kn} \end{bmatrix} \begin{bmatrix} X_1 \\ \vdots \\ X_n \end{bmatrix} = \begin{bmatrix} b_1 \\ \vdots \\ b_k \end{bmatrix}.$$

The matrix $A = [a_{ij}] \in \mathcal{M}_{k \times n}(F)$ is the **coefficient matrix** of the system. If we set $w = \begin{bmatrix} b_1 \\ \vdots \\ b_k \end{bmatrix} \in F^k$, then the matrix $\begin{bmatrix} A & w \end{bmatrix} \in \mathcal{M}_{k \times (n+1)}(F)$ is called the **extended coefficient matrix** of the system.

The set of all vectors $v = \begin{bmatrix} d_1 \\ \vdots \\ d_n \end{bmatrix} \in F^n$ satisfying $Av = w$ is the

solution set of the system. This is clearly equal to $\alpha^{-1}(w)$, where $\alpha : F^n \rightarrow F^k$ is the linear transformation satisfying $\Phi_{BD}(\alpha) = A$, where B and D are the canonical bases of F^n and F^k respectively. In particular, if the system is homogeneous then its solution set is just the kernel of α , and is called the **solution space** of the system.

We note the following simple but important point: if F is a subfield of a field K and if k and n are positive integers, then any matrix A in $\mathcal{M}_{k \times n}(F)$ also belongs to $\mathcal{M}_{k \times n}(K)$ and any vector $v \in F^n$ also belongs to K^n . Therefore, if $w \in F^k$, any element of the solution set of $Av = w$, considered as a system of linear equations over F , remains a solution when we consider this as a system of linear equations over K .



¹

Edmond Laguerre, a 19th century French mathematician, wrote an important book on systems of linear equations in 1867. **Takakazu Seki Kowa** was a 17th century Japanese mathematician, the son of a samurai warrior family, who developed matrix-based methods based on old Chinese texts.

(10.1) Proposition: The solution set of a homogeneous system of linear equations in n unknowns is a subspace of F^n .

Proof: This is a direct consequence of Proposition 6.4. □

For nonhomogeneous systems, the situation is a bit more complicated.

(10.2) Proposition: Let $AX = w$ be a nonhomogeneous system of linear equations in n unknowns over a field F and let $v_0 \in F^n$ be a solution to this systems. Then the solution set of the system is the set of all vectors in F^n of the form $v_0 + v$, where v is a

solution to the homogeneous system $AX = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$.

Proof: This is an immediate consequence of Proposition 6.6. □

We should emphasize that the solution set of a nonhomogeneous system of linear equations is not a subspace of F^n but rather an affine subset of that space.

Example: If we identify \mathbb{R}^2 with the Euclidean plane by associating each vector $\begin{bmatrix} a \\ b \end{bmatrix}$ with the point with coordinates (a, b) , then we see its subspaces of dimension 1 are precisely the straight lines going through the origin. The solutions of linear equations of the form $a_1X_1 + a_2X_2 = b$, where $b \neq 0$, and at least one of the a_i is also nonzero, are the straight lines in the plane which do not go through the origin.

We are still left with the question of how to actually find a solution to a system of linear equations. Here we can distinguish between two approaches:

(1) **Direct methods.** These methods involve the manipulation of the matrix A , either replacing it with another matrix which is easier to work with or factoring it into a product of matrices which are easier to work with, and thus reducing the difficulty of the problem.

(2) **Iterative methods.** These methods involve selecting a likely solution for the system and then repeatedly modifying it to obtain a sequence of vectors which (hopefully) will converge to an actual solution to the system. Such methods work, of course, only if our vector space is one in which the notion of convergence is meaningfully defined. As we shall see, this is possible when the field of scalars equals \mathbb{R} or \mathbb{C} .

We begin by looking at direct methods. Let P be a nonsingular matrix in $\mathcal{M}_{k \times k}(F)$. A vector $v \in F^n$ is a solution to the system $AX = w$ over F if and only if it is a solution to the system $(PA)X = Pw$. In particular, this is true for elementary matrices. Thus, given a system of linear equations, we can change the order of the equations, multiply one of the equations by a nonzero scalar, or add a scalar multiple of one equation to another, without changing the solution set of the system, so long as we do the same thing on both sides of the equal sign. In order to do this efficiently, it is best to work with the extended coefficient matrix $\begin{bmatrix} A & w \end{bmatrix}$ and perform elementary operations on it to reduce it to a convenient form.

Which form should we choose? Let F be a field, let k and n be positive integers, and let $B = [b_{ij}] \in \mathcal{M}_{k \times n}(F)$. The matrix B is in **row echelon form** if and only if the following conditions are satisfied:

- (1) If $1 < s \leq k$ then $b_{s,j} = 0$ for all $1 \leq j < s$;
- (2) If $1 \leq s \leq k$ and $1 \leq t \leq n$ satisfy the condition that $b_{sj} = 0$ for all $1 \leq j < t$, then $b_{ij} = 0$ for all $s < i \leq k$ and all $1 \leq j \leq t$.

Example: The matrices $\begin{bmatrix} 1 & 6 & 7 & 7 & 1 \\ 0 & 9 & 2 & 1 & 1 \\ 0 & 0 & 0 & 2 & 2 \\ 0 & 0 & 0 & 0 & 1 \end{bmatrix}$ and $\begin{bmatrix} 8 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 2 & 6 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}$ are in row echelon form. The matrix $\begin{bmatrix} 1 & 5 & 2 & 9 & 0 \\ 0 & 0 & 1 & 5 & 4 \\ 0 & 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 0 & 7 \end{bmatrix}$ is not in row echelon form.

Example: If n is a positive integer and if $B \in \mathcal{M}_{n \times n}(F)$ is in row echelon form, then B is surely upper triangular. However, $\begin{bmatrix} 1 & 0 & 2 & 7 \\ 0 & 0 & 3 & 8 \\ 0 & 0 & 9 & 0 \\ 0 & 0 & 0 & 5 \end{bmatrix}$ is an upper-triangular matrix which is not in row echelon form.

We claim that for any matrix $A = [a_{ij}] \in \mathcal{M}_{k \times n}(F)$ is row equivalent to a matrix in row echelon form. By Proposition 9.4, this is equivalent to saying that A can be transformed into a matrix in row echelon form by a series of elementary operations, as follows:

- (1) Find the leftmost column of A which has a nonzero entry and interchange rows if necessary, so that this entry is in the first row. Thus we now have a matrix A in which $a_{1h} \neq 0$ and $a_{ij} = 0$ for all $1 \leq i \leq k$ and all $1 \leq j < h$.

(2) For each $1 < i \leq k$, if $a_{ih} \neq 0$ then we multiply the first row by $-a_{ij}a_{1h}^{-1}$ and add it to the i th row, which creates a new row in which the (i, h) -entry is equal to 0. Thus, we now have a matrix in which $a_{ih} = 0$ for all $1 < i \leq k$.

(3) Now consider the submatrix of A from which we deleted the first row and the first h columns, and repeat the above procedure.

Example: Let us begin with the matrix $A = \begin{bmatrix} 1 & 2 & 3 & 1 \\ 2 & 1 & 4 & 2 \\ 1 & -1 & 1 & 1 \end{bmatrix} \in \mathcal{M}_{3 \times 4}(\mathbb{R})$. We already have $a_{11} \neq 0$. Multiplying the first row by -2 and adding it to the second row, we obtain $\begin{bmatrix} 1 & 2 & 3 & 1 \\ 0 & -3 & -2 & 0 \\ 1 & -1 & 1 & 1 \end{bmatrix}$ and then multiplying the first row by -1 and adding it to the third row, we obtain $\begin{bmatrix} 1 & 2 & 3 & 1 \\ 0 & -3 & -2 & 0 \\ 0 & -3 & -2 & 0 \end{bmatrix}$. We also already have $a_{22} \neq 0$. Multiplying the second row by -1 and adding it to the third row, we obtain $\begin{bmatrix} 1 & 2 & 3 & 1 \\ 0 & -3 & -2 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$, and this is in row echelon form.

If $A = [a_{ij}] \in \mathcal{M}_{k \times n}(F)$ is a matrix in row echelon form, and if the h th row of A contains nonzero entries, then the leftmost nonzero entry of the row is the **leading entry**. The matrix A is in **reduced row echelon form** if it is in row echelon form and, in addition, satisfies the following additional conditions:

- (1) The leading entry in each nonzero row is equal to 1;
- (2) If a_{hj} is a leading entry, then $a_{ij} = 0$ for all $i \neq h$.

Any matrix in row echelon is row-equivalent to one in reduced row echelon form; that is to say, such a matrix can be converted to one in reduced row echelon form by performing additional elementary operations: first we multiply each nonzero row by the multiplicative inverse of its leading entry, to obtain a matrix in which the leading entry of each nonzero row equals 1. Then, if a_{hj} is a leading entry and if $i < h$, we multiply the h th row by $-a_{ij}$ and add it to the i th row, which will give us a matrix with the (i, j) -entry equal to 0. The reduced row echelon form of any given matrix is clearly unique.

Example: Let us go back and look at the matrix $\begin{bmatrix} 1 & 2 & 3 & 1 \\ 0 & -3 & -2 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$

in row echelon form. The leading entry of the first row is already equal to

1. Multiplying the second row by $-\frac{1}{3}$ to obtain, $\begin{bmatrix} 1 & 2 & 3 & 1 \\ 0 & 1 & \frac{2}{3} & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$, a

matrix in which the leading entry of the second row is equal to 1 as well. Now multiply the second row by -2 and add it to the first row, to obtain

$$\begin{bmatrix} 1 & 0 & \frac{8}{3} & 1 \\ 0 & 1 & \frac{2}{3} & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}, \text{ which is in reduced row echelon form.}$$

Now let us return to the system of linear equations $AX = w$ in n unknowns and consider methods of solution. The most well-known is **Gaussian elimination** or the **Gauss-Jordan method**.² In this method, we first perform elementary operations on the extended coefficient matrix $[A \ w]$ to bring it to reduced row echelon form. Having done this, we now have a new system of linear equations $A'X = w'$, the solution set of which is the same as that of the original system. Let t be the greatest integer i such that the i th row has nonzero entries. There are several possibilities:

(1) $b_t \neq 0$ but $a'_{tj} = 0$ for all $1 \leq j \leq n$. Then the system has no solutions and we are done.

(2) There is precisely one index j such that $a'_{tj} \neq 0$. Then this must in fact be the leading entry of the t th row and so $a'_{tj} = 1$. This means that in any element of the solution set of the system we must have the j th entry equal to b_j . We can therefore substitute b_j for X_j in each of the other equations, and reduce the system to one of equations of $n - 1$ unknowns.

(3) There are several indices j such that $a'_{tj} \neq 0$, say those in columns $h_1 < h_2 < \cdots < h_m$. Then a'_{th_1} is the leading entry of the t th



2

Carl Friedrich Gauss, who lived in Germany at the beginning of the 19th century, is considered to be the leading mathematician of all times, as well as a physicist and astronomer of the first rank. He developed this method in connection with his work in astronomy in 1809. Gaussian elimination first appeared in print in a handbook by German geodist **Wilhelm Jordan**, who applied the method to problems in surveying.

row and so equals 1. Moreover, for any values z_1, \dots, z_m we substitute for X_{h_2}, \dots, X_{h_m} , we will get a solution to the system with these values and with $b_t - \sum_{s=2}^m z_s$ substituted for X_{h_1} . Thus we can consider the z_i as parameters of a general solution and again reduce the system to one in a smaller number of unknowns.

(4) Having reduced the system, we now recursively apply the previous steps until the system is solved.

Strassen's insight that Gaussian elimination may not be the optimal method of solving systems of linear equations, as had been previously thought, led to the development of his method of matrix multiplication.

Example: Let us consider the system of linear equations

$$\begin{aligned} 3X_1 + 2X_2 + X_3 &= 0 \\ -2X_1 + X_2 - X_3 &= 2 \\ 2X_1 - X_2 + 2X_3 &= -1 \end{aligned}$$

over the field \mathbb{R} . The extended coefficient matrix of this system is

$$\begin{bmatrix} 3 & 2 & 1 & 0 \\ -2 & 1 & -1 & 2 \\ 2 & -1 & 2 & -1 \end{bmatrix}$$

and this is row equivalent to the matrix $\begin{bmatrix} 1 & \frac{2}{3} & \frac{1}{3} & 0 \\ 0 & 7 & -1 & 6 \\ 0 & 0 & 1 & 1 \end{bmatrix}$ in row echelon

form, which is in turn row equivalent to the matrix $\begin{bmatrix} 1 & 0 & 0 & -1 \\ 0 & 1 & 0 & 1 \\ 0 & 0 & 1 & 1 \end{bmatrix}$ in

reduced row echelon form. Thus we see that the solution set of the system

is $\left\{ \begin{bmatrix} -1 \\ 1 \\ 1 \end{bmatrix} \right\}.$

Example: Let us consider the system of linear equations

$$\begin{aligned} X_1 + X_2 &= 1 \\ X_1 - X_2 &= 3 \\ -X_1 + 2X_2 &= -2 \end{aligned}$$

over the field \mathbb{R} . The extended coefficient matrix of this system equals

$$\begin{bmatrix} 1 & 1 & 1 \\ 1 & -1 & 3 \\ -1 & 2 & -2 \end{bmatrix}, \text{ and this is row equivalent to the matrix } \begin{bmatrix} 1 & 1 & 1 \\ 0 & -2 & 2 \\ 0 & 0 & 2 \end{bmatrix}$$

in row echelon form, which is row equivalent to the matrix $\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$ in reduced row echelon form. Therefore this system has no solutions at all.

Example: Let us consider the system of linear equations

$$\begin{aligned} X_1 + 2X_2 + X_3 &= -1 \\ 2X_1 + 4X_2 + 3X_3 &= 3 \\ 3X_1 + 6X_2 + 4X_3 &= 2 \end{aligned}$$

over the field \mathbb{R} . The extended coefficient matrix of this system is

$$\begin{bmatrix} 1 & 2 & 1 & -1 \\ 2 & 4 & 3 & 3 \\ 3 & 6 & 4 & 2 \end{bmatrix}$$

and this is row equivalent to the matrix $\begin{bmatrix} 1 & 2 & 1 & -1 \\ 0 & 0 & 1 & 5 \\ 0 & 0 & 0 & 0 \end{bmatrix}$ in row echelon

form, which is in turn row equivalent to the matrix $\begin{bmatrix} 1 & 2 & 0 & -6 \\ 0 & 0 & 1 & 5 \\ 0 & 0 & 0 & 0 \end{bmatrix}$ in

reduced row echelon form. From the second row we see that we must have $X_3 = 5$. From the first row we have $X_1 + 2X_2 = -6$ and so, for each value $X_2 = z$, we have a solution with $X_1 = -6 - 2z$. Therefore the

solution set to our system is $\left\{ \begin{bmatrix} -6 - 2z \\ z \\ 5 \end{bmatrix} \mid z \in \mathbb{R} \right\}$.

We note that if $A \in \mathcal{M}_{k \times n}(F)$ then the number of arithmetic operations needed to solve a system of linear equations of the form $AX = w$ using Gaussian elimination, is no more than $\frac{1}{6}k(k-1)(3n-k-2)$ if $k < n$ and no more than $\frac{1}{6}n[3kn + 3(k-n) - n^2 - 2]$ otherwise³. Of course, if



³

The first computer program to solve a system of linear equations by Gaussian elimination was written by **Lady Augusta Ada Lovelace**, a student of

the matrix A is of a special form, this procedure can be much faster. For example, if $A \in \mathcal{M}_{n \times n}(F)$ is a tridiagonal matrix, then a system of equations of the form $AX = w$ can be solved using $3n$ additions/subtractions and $5n$ multiplications.

Gaussian elimination can also be used to check if a set of vectors in F^k is linearly independent. Let $\{v_1, \dots, v_n\}$ be a set of vectors in F^k , where

$$v_j = \begin{bmatrix} a_{1j} \\ \vdots \\ a_{kj} \end{bmatrix} \text{ for all } j. \text{ We want to know if there exist scalars } b_1, \dots, b_n$$

in F , not all equal to 0, satisfying $\sum_{j=1}^n b_j v_j = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$. In other

words, we want to know if the homogeneous systems of linear equations

$$AX = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix} \text{ has a nonzero solution, where } A = [a_{ij}] \in \mathcal{M}_{k \times n}(F).$$

Example: Let us check if the subset $\left\{ \begin{bmatrix} 1 \\ -1 \\ 3 \\ 4 \end{bmatrix}, \begin{bmatrix} 3 \\ -3 \\ 6 \\ 4 \end{bmatrix}, \begin{bmatrix} -1 \\ 1 \\ 0 \\ 4 \end{bmatrix} \right\}$

of \mathbb{Q}^4 is linearly dependent, and to do so we need to consider the matrix

$$A = \begin{bmatrix} 1 & 3 & -1 & 0 \\ -1 & -3 & 1 & 0 \\ 3 & 6 & 0 & 0 \\ 4 & 4 & 4 & 0 \end{bmatrix}.$$

This matrix is row equivalent to the matrix $\begin{bmatrix} 1 & 0 & 2 & 0 \\ 0 & 1 & -1 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$ in reduced

row echelon form. Therefore the set of solutions to the homogeneous system

$$AX = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix} \text{ is } \left\{ \begin{bmatrix} -2z \\ z \\ z \end{bmatrix} \mid z \in \mathbb{Q} \right\} \text{ so that if we pick one such}$$

De Morgan and daughter of the poet Lord Byron, who developed software for Charles Babbage's (never completed) mechanical computer in the 19th century. Her program was capable of solving systems of 10 linear equations in 10 unknowns.

nonzero element, say $\begin{bmatrix} -2 \\ 1 \\ 1 \end{bmatrix}$, we see that

$$(-2) \begin{bmatrix} 1 \\ -1 \\ 3 \\ 4 \end{bmatrix} + \begin{bmatrix} 3 \\ -3 \\ 6 \\ 4 \end{bmatrix} + \begin{bmatrix} -1 \\ 1 \\ 0 \\ 4 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \end{bmatrix},$$

showing that the set is indeed linearly dependent.

If $A \in \mathcal{M}_{k \times n}(F)$ is a nonsingular matrix which can be written in the form LU , where L is lower triangular and U is upper triangular, then a system of linear equations of the form $UX = w$ is easy to solve using Gaussian elimination, since U is already in row-echelon form. Moreover, since U must also be nonsingular, this system has a unique solution $y = U^{-1}w$. Then the system $AX = w$ has a unique solution, which is also the solution to the system $LX = y$ and that system too is easy to solve. We therefore see the importance of the LU -decomposition of matrices, assuming that one exists.

Given a matrix $A \in \mathcal{M}_{k \times n}(F)$, we define the **column space** of A to be the subspace of F^k generated by the set of all columns of A . The dimension of the column space of A is called the **rank** of A . Moreover, there exists a linear transformation $\alpha : F^n \rightarrow F^k$ satisfying the condition that $\Phi_{BD}(\alpha) = A$, where B and D be the canonical bases of F^n and F^k respectively, and it is clear that the column space of A is just $\text{im}(\alpha)$. Similarly, we define the **row space** of A to be the subspace of $\mathcal{M}_{1 \times n}(F)$ generated by the rows of A . We will show that the dimension of this space is also equal to the rank of A .

(10.3) Proposition: Let F be a field, let k and n be positive

integers, and let $A \in \mathcal{M}_{k \times n}(F)$ and let $w = \begin{bmatrix} b_1 \\ \vdots \\ b_k \end{bmatrix} \in F^k$. Then

the system of linear equations $AX = w$ has a solution if and only if w belongs to the column space of A .

Proof: If $v = \begin{bmatrix} d_1 \\ \vdots \\ d_n \end{bmatrix}$ is a solution of the system $AX = w$ then

$$w = \begin{bmatrix} \sum_{j=1}^n a_{1j}d_j \\ \vdots \\ \sum_{j=1}^n a_{kj}d_j \end{bmatrix} = \sum_{j=1}^n d_j \begin{bmatrix} a_{1j} \\ \vdots \\ a_{kj} \end{bmatrix}$$

and so w is a linear combination of the columns of A . Conversely, if we assume that there exist scalars d_1, \dots, d_n in F such that $w =$

$$\sum_{j=1}^n d_j \begin{bmatrix} a_{1j} \\ \vdots \\ a_{kj} \end{bmatrix}, \text{ then } v = \begin{bmatrix} d_1 \\ \vdots \\ d_n \end{bmatrix} \text{ is a solution of the given system.}$$

□

In particular, we get the following consequence of this result.

(10.4) Proposition: Let F be a field, let k and n be positive

integers, and let $A \in M_{k \times n}(F)$ and let $w = \begin{bmatrix} b_1 \\ \vdots \\ b_k \end{bmatrix} \in F^k$. Then

the system of linear equations $AX = w$ has a solution if and only if the rank of the coefficient matrix A is equal to the rank of the extended coefficient matrix.

Now let us return to the problem of identifying the solution sets of homogeneous systems of linear equations.

(10.5) Proposition: Let F be a field, let k and n be positive integers, and let $A \in M_{k \times n}(F)$ be a matrix the columns of which are vectors y_1, \dots, y_n in F^k . Assume these columns are arranged such that $\{y_1, \dots, y_r\}$ is a basis for the column space of A , for some $r \leq n$. Moreover, for all $r < h \leq n$, let us select scalars b_{h1}, \dots, b_{hn} such that:

- (1) $y_h = b_{h1}y_1 + \dots + b_{hr}y_r$;
- (2) $b_{hh} = -1$;
- (3) $b_{hj} = 0$ otherwise.

For each $r < h \leq n$, **let** $v_h = \begin{bmatrix} b_{h1} \\ \vdots \\ b_{hn} \end{bmatrix} \in F^n$. **Then** $\{v_{r+1}, \dots, v_n\}$ is a basis for the solution space of the homogeneous system of linear equations $AX = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$.

(**Comment before the proof:** Since $\{y_1, \dots, y_n\}$ is a set of generators for the column space of A , it contains a subset that is a basis. The assumption that this subset is $\{y_1, \dots, y_r\}$ is for notational convenience only.)

Proof: If $r = n$ then the solution space of the system of linear equations is $\left\{ \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix} \right\}$ and so the result is immediate. Hence let us

assume that $r < n$. If $r < h \leq n$, then $Av_h = \sum_{j=1}^r b_{hj}y_j - y_h = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$

and so each v_h belongs to the solution space of $AX = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$. Moreover,

the set $\{v_{r+1}, \dots, v_n\}$ is linearly independent, since if $\sum_{j=r+1}^n c_j v_j = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$

then for each $r < h \leq n$ we note that the h th entry on the left-hand side is $-c_h$ whereas the corresponding entry on the right-hand side is 0, proving that $c_h = 0$ for all $r < h \leq n$.

We are therefore left to show that $\{v_{r+1}, \dots, v_n\}$ is a generating set for the solution space of the given homogeneous system. And, indeed, let

$w = \begin{bmatrix} d_1 \\ \vdots \\ d_n \end{bmatrix}$ be a vector in this solution space. Then $w + \sum_{h=r+1}^n d_h v_h = \begin{bmatrix} e_1 \\ \vdots \\ e_n \end{bmatrix}$, where $e_{r+1} = \dots = e_n = 0$. Therefore, this vector belongs to

solution space of the system, and so $\sum_{h=q}^r e_h y_h = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$. However, since the set $\{y_1, \dots, y_r\}$ is linearly independent, this implies that $e_1 = \dots = e_r = 0$ as well. Therefore $w = -\sum_{h=r+1}^n d_h v_h$, showing that $\{v_{r+1}, \dots, v_n\}$ is a generating set for the solution space, as required. \square

As an immediate consequence of Proposition 10.5, we obtain the following result.

(10.6) Proposition: Let F be a field, let k and n be positive integers, and let $A \in M_{k \times n}(F)$. Then the dimension of the solution space of the homogeneous system of linear equations

$AX = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$ is $n - r$, where r is the rank of the coefficient matrix A .

We are now ready to prove the characterization of rank which we mentioned before.

(10.7) Proposition: Let F be a field, let k and n be positive integers, and let $A \in M_{k \times n}(F)$. Then the rank of A equals the dimension of the row space of A .

Proof: Let v_1, \dots, v_k be the rows of A , which generate a subspace of $\mathcal{M}_{1 \times n}(F)$. We can reorder these rows in such a way that $\{v_1, \dots, v_t\}$ is a basis for the row space, for some $1 \leq t \leq k$. This, as we know, does not change the solution space of the homogeneous system of linear

equations $AX = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$ and hence does not change the rank r_A of A .

Let $B \in \mathcal{M}_{t \times n}(F)$ be the matrix obtained from A by deleting rows $t+1, \dots, k$. The columns of B belong to F^t and so the rank r_B of B satisfies $r_B \leq t$, which implies that $n - t \leq n - r_B$. But we have already

seen that the homogeneous systems of linear equations $AX = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$ and

$BX = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$ have the same solution space and so, by Proposition 10.6,

$n - t \leq n - r_A$. From this we conclude that $r_A \leq t$. We have thus shown that the rank of any matrix is less than or equal to the dimension of its row space. In particular, this is also true for A^T . But the rank of A^T is t , while the dimension of its row space is r_A , and so we have $t \leq r_A$ as well, proving equality. \square

Example: Let us find a basis for the solution space of the system of linear equations $\begin{bmatrix} 1 & 2 & -3 & 1 \\ 1 & 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} X_1 \\ X_2 \\ X_3 \\ X_4 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$ over \mathbb{R} . We know that

the coefficient matrix is row-equivalent to the matrix $\begin{bmatrix} 1 & 0 & 5 & 1 \\ 0 & 1 & -4 & 0 \end{bmatrix}$ in reduced row echelon form, and this matrix has rank 2. Therefore the solution space of the system has dimension $4 - 2 = 2$. Indeed, it is easy

to check that $\left\{ \begin{bmatrix} -5 \\ 4 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} -1 \\ 0 \\ 0 \\ 1 \end{bmatrix} \right\}$ is a basis for this solution space.

Gaussian elimination requires an order of magnitude of n^3 arithmetic operations to solve a system of n linear equations in n unknowns. This computational overhead is quite significant if n is large (say, over 10,000), even with the use of supercomputers. As a result, there is considerable continuing research into finding faster methods of computation, especially in those cases in which we have additional information on the structure of the matrix of coefficients, originating in knowledge of the particular problem from which the system arose. Often this structural information is immediately noticeable, but sometimes it appears only after a sophisticated consideration of the problem.

Example: It is often possible to show that the matrix we are interested in, while not itself having a special structure, is equal to the product of two matrices having a special structure, a situation which arises in many mathematical models. Let us consider one such case. An $n \times n$ **symmetric Toeplitz matrix** is a matrix $B = [b_{ij}] \in \mathcal{M}_{n \times n}(\mathbb{R})$ satisfying the condition that there exist real numbers c_0, \dots, c_{n-1} such that $b_{ij} = c_h$

whenever $|i - j| = h$. Thus, for example, the matrix

$$\begin{bmatrix} 1 & 2 & 0 & 7 \\ 2 & 1 & 2 & 0 \\ 0 & 2 & 1 & 2 \\ 7 & 0 & 2 & 1 \end{bmatrix}$$

is a symmetric Toeplitz matrix. Clearly the set of all symmetric Toeplitz matrices is a subspace of $\mathcal{M}_{n \times n}(\mathbb{R})$. However, it is not a subalgebra, since the product of two such matrices need not be a symmetric Toeplitz matrix. They are also convenient to store in a computer, since we need keep in memory only the n scalars c_0, \dots, c_{n-1} .

Many mathematical models in economics are built around solving systems of linear equations of the form $AX = w$, where A is a product of two symmetric Toeplitz matrices.⁴

The proper use of mathematical techniques, and especially computational techniques, also depends very much on a deep understanding of the particular problem one is dealing with. Also, it is crucial to emphasize once again that any method we use to solve a system of linear equations on a computer will induce errors as a result of roundoff and truncation in our computations. With some methods – such as Gaussian elimination – these errors tend to accumulate, whereas with others they often cancel each other out, within certain limits. It is therefore necessary, especially when we are dealing with large matrices, to have on hand several methods of handling such systems of equations and to be able to keep track of the way in which errors can propagate in each of the different methods at one's disposal.⁵



⁴ **Otto Toeplitz** was a 20th-century German mathematician who studied endomorphisms of infinite-dimensional vector spaces.



⁵ The problem of the numerical stability of computed solutions of systems of linear equations was the subject of considerable research in the early days of computers. Among the outstanding contributors were the Russian husband-and-wife team of **Dimitri Konstantinovich Faddeev** and **Vera Nikolaevna Faddeeva**.

We now turn to iterative methods of solution of systems of linear equations. For simplicity, we will assume that our field of scalars is always \mathbb{R} . The basic idea is, as we have already noted, to guess a possible solution and then use this initial guess to compute a sequence of further approximations to the solution which, hopefully, will converge (in some topology) with relative rapidity. Usually, the initial guess is based on knowledge of the real-life problem which gave rise to the system of equations, something that can often be done with good accuracy. In very large and computationally-difficult situations (for example, weather prediction, chip design, large-scale economic models, computational acoustics, or the modeling the chemistry of polymer chains) one can even use Monte Carlo methods, based on statistical sampling and estimation techniques, to come up with an initial guess or even an approximate solution.

To illustrate this approach, let us consider the problem of solving a system of linear equations of the form $AX = w$, where $A = [a_{ij}] \in \mathcal{M}_{n \times n}(\mathbb{R})$

is a nonsingular matrix and $w = \begin{bmatrix} b_1 \\ \vdots \\ b_n \end{bmatrix} \in \mathbb{R}^n$. We know that this system

has a unique solution, namely $A^{-1}w$, but inverting the matrix A may be computationally time-consuming and prone to error, so we are looking for another method. Suppose that we can write $A = E - D$, where E is some matrix which is easy to invert. Then if $v \in \mathbb{R}^n$ satisfies $Av = w$, we know that $Ev = Dv + w$ and so $v = E^{-1}(Dv + w)$. We now guess a value for v , call it $v^{(0)}$. Then, using this formula, we can define new vectors $v^{(1)}, v^{(2)}, \dots$ iteratively by setting $v^{(h)} = E^{-1}(Dv^{(h-1)} + w)$ for each $h > 0$. This can be done relatively quickly since, by assumption, E^{-1} was relatively easy to compute and, having computed it once for the first step of the iteration, we don't need to recompute it for subsequent steps. Our hope is that the sequence $v^{(0)}, v^{(1)}, v^{(2)}, \dots$ will in fact converge. Indeed, if this sequence does converge to some vector v then it is easy to verify that v must be the unique solution of $AX = w$.

For example, let us assume that the diagonal entries a_{ii} of A are all nonzero, and let us choose E to be the diagonal matrix having these entries on the diagonal. Then E^{-1} is also a diagonal matrix having the

entries a_{ii}^{-1} on the diagonal. If our initial guess is $v^{(0)} = \begin{bmatrix} c_1^{(0)} \\ \vdots \\ c_n^{(0)} \end{bmatrix}$,

then it is easy to see that for $h > 0$ we have $v^{(h)} = \begin{bmatrix} c_1^{(h)} \\ \vdots \\ c_n^{(h)} \end{bmatrix}$, where

$c_i^{(h+1)} = a_{ii}^{-1} \left[b_i - \sum_{j \neq i} a_{ij} c_j^{(h)} \right]$ for all $1 \leq i \leq n$. This method is known as the **Jacobi iteration method**⁶. Another possibility, again under the assumption that the diagonal entries a_{ii} of A are all nonzero, is to choose E to be the upper-triangular matrix $[e_{ij}]$ defined by setting

$$e_{ij} = \begin{cases} a_{ij} & \text{if } i \leq j \\ 0 & \text{otherwise} \end{cases}.$$

Given an initial guess $v^{(0)} = \begin{bmatrix} c_1^{(0)} \\ \vdots \\ c_n^{(0)} \end{bmatrix}$, we see that $v^{(h)} = \begin{bmatrix} c_1^{(h)} \\ \vdots \\ c_n^{(h)} \end{bmatrix}$ for

$h > 0$, where $c_i^{(h+1)} = a_{ii}^{-1} \left[b_i - \sum_{j=1}^{i-1} a_{ij} c_j^{(h+1)} - \sum_{j=i+1}^n a_{ij} c_j^{(h)} \right]$ for all $1 \leq i \leq n$. This method is known as the **Gauss-Seidel iteration method**, since it was discovered independently by Gauss and by Jacobi's student Philipp Ludwig von Seidel.

In both of the above methods, and in other iteration methods (and there are many of these), there is no guarantee that the sequence of approximations will always converge or that, even if it does converge, it will do so rapidly. Understanding the conditions for convergence and analyzing the speed of convergence requires sophisticated techniques in numerical analysis, and indeed there are many examples of matrices for which one iteration scheme converges whereas another doesn't, as well as various necessary and sufficient conditions for a given iteration method to converge.⁷ For example, a sufficient condition for the Jacobi iteration method to converge for a



6

Carl Gustav Jacob Jacobi was a 19th-century German mathematician, who worked mostly in analysis and applied mathematics. His work in astronomy led him to solve large systems of linear equations, and his papers on determinants helped make them well-known.



7

The convergence and accuracy of the Gauss-Seidel iteration method was studied in detail by the Russian mathematician and engineer **Alexander Ivanovich Nekrasov** at the beginning of the twentieth century, long before the use of electronic computers.

matrix $A = [a_{ij}]$ is that, $\sum_{j \neq i} |a_{ij}| < |a_{ii}|$ for all $1 \leq i \leq n$. As a rule of thumb, iteration methods work best for large **sparse** matrices – namely matrices in which a very large majority of the entries are 0 – such as those arising from the solution of systems of partial differential equations. As previously remarked, in iteration methods truncation and roundoff errors tend to cancel each other out, rather than accumulate.

Example: Let $A = \begin{bmatrix} 4 & 2 & 1 \\ -1 & 1 & 2 \\ 0 & 1 & 3 \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{R})$ and let $w = \begin{bmatrix} 7 \\ 2 \\ 4 \end{bmatrix}$.

The system of linear equations $Ax = w$ has a unique solution $\begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$. If

we use the Jacobi iteration method beginning with the initial guess $v^{(0)} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$, we get the sequence of vectors (written to 6 digit accuracy):

$$\begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 1.75000 \\ 2.00000 \\ 1.33333 \end{bmatrix}, \begin{bmatrix} 0.41667 \\ 1.08333 \\ 0.66667 \end{bmatrix}, \begin{bmatrix} 1.04167 \\ 1.08333 \\ 0.97222 \end{bmatrix}, \begin{bmatrix} 0.96528 \\ 1.09722 \\ 0.97222 \end{bmatrix},$$

$$\begin{bmatrix} 0.95833 \\ 1.02083 \\ 0.96759 \end{bmatrix}, \begin{bmatrix} 0.99768 \\ 1.02314 \\ 1.01157 \end{bmatrix}, \begin{bmatrix} 0.99016 \\ 1.01157 \\ 0.99228 \end{bmatrix}, \begin{bmatrix} 0.99614 \\ 1.00559 \\ 0.99614 \end{bmatrix},$$

$$\begin{bmatrix} 0.99816 \\ 1.00386 \\ 0.99814 \end{bmatrix}, \begin{bmatrix} 0.99853 \\ 1.00190 \\ 0.99871 \end{bmatrix}, \dots$$

and if we use the Gauss-Seidel iteration method with the same initial guess, we get the sequence of vectors (written to 6 digit accuracy):

$$\begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 1.75000 \\ -0.66667 \\ 1.33333 \end{bmatrix}, \begin{bmatrix} 1.04167 \\ 0.63889 \\ 1.55556 \end{bmatrix}, \begin{bmatrix} 1.06944 \\ 0.80093 \\ 1.12037 \end{bmatrix}, \begin{bmatrix} 1.01504 \\ 0.93672 \\ 1.06636 \end{bmatrix},$$

$$\begin{bmatrix} 1.00829 \\ 0.97287 \\ 1.02109 \end{bmatrix}, \begin{bmatrix} 1.00264 \\ 0.99020 \\ 1.00904 \end{bmatrix}, \begin{bmatrix} 1.00113 \\ 0.99611 \\ 1.00326 \end{bmatrix}, \begin{bmatrix} 1.00040 \\ 0.99853 \\ 1.00129 \end{bmatrix},$$

$$\begin{bmatrix} 1.00016 \\ 0.99943 \\ 1.00049 \end{bmatrix}, \begin{bmatrix} 1.00006 \\ 0.99978 \\ 1.00019 \end{bmatrix}, \dots$$

so we see that both methods converge, albeit quite differently.

Example: Let $A = \begin{bmatrix} 1 & 2 & 0 \\ 2 & 1 & 2 \\ 0 & 2 & 1 \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{R})$ and let $w = \begin{bmatrix} 0 \\ -1 \\ 3 \end{bmatrix}$.

The system of linear equations $Ax = w$ has a unique solution $\begin{bmatrix} -2 \\ 1 \\ 1 \end{bmatrix}$. If

we try to solve this system using the Gauss-Seidel method with the initial

guess $\begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$, we get the sequence of vectors

$$\begin{bmatrix} 0 \\ -1 \\ 5 \end{bmatrix}, \begin{bmatrix} 2 \\ -15 \\ 33 \end{bmatrix}, \begin{bmatrix} 30 \\ -127 \\ 257 \end{bmatrix}, \begin{bmatrix} 254 \\ -1023 \\ 2049 \end{bmatrix}, \dots$$

which clearly diverges.

A more sophisticated iteration technique is, at each stage, not to replace $v^{(i)}$ by the computed $v^{(i+1)}$ but rather by a linear combination of the form $rv^{(i+1)} + (1-r)v^{(i)}$, where $0 \leq r$ is a **relaxation parameter**. Doing this with Jacobi iteration gives us the **Jacobi overrelaxation (JOR) method**, and doing it with the Gauss-Seidel method gives us the **successive overrelaxation (SOR) method**. The relaxation parameter r is chosen on the basis of certain properties of the matrix A . By choosing this parameter wisely, one can often achieve a considerable improvement in convergence.

Example: In the beginning of this chapter we saw an example of how a nonlinear system of equations can be turned into a linear system. This can often be done in more general cases, producing large systems of linear equations of the form $AX = w$, where the matrix A is usually sparse and for which iteration methods are therefore appropriate. Consider, for example, the problem of finding real numbers a , b , and c such that the following conditions hold:

$$\begin{aligned} a^2 - b^2 + c^2 &= 6 \\ ab + ac + 4bc &= 29 \\ a^2 + 2ab - 2bc &= -7 \\ 2a^2 - 3ab + c^2 &= 5 \\ b^2 - c^2 + 5ab &= 5 \\ 2ac - 3b^2 &= -6 \end{aligned}$$

To linearize this, we begin by assigning variables to all of the terms appearing in the equations: $X_1 = a^2$, $X_2 = b^2$, $X_3 = c^2$, $X_4 = ab$, $X_5 = ac$,

and $X_6 = bc$. This then yields the system of linear equations

$$\begin{bmatrix} 1 & -1 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 1 & 4 \\ 1 & 0 & 0 & 2 & 0 & -2 \\ 2 & 0 & 1 & -3 & 0 & 0 \\ 0 & 1 & -1 & 5 & 0 & 0 \\ 0 & -3 & 0 & 0 & 2 & 0 \end{bmatrix} X = \begin{bmatrix} 6 \\ 29 \\ -7 \\ 5 \\ 5 \\ -6 \end{bmatrix}$$

which has a unique solution $X_1 = 1$, $X_2 = 4$, $X_3 = 9$, $X_4 = 2$, $X_5 = 3$, and $X_6 = 6$, from which we deduce that $a = 1$, $b = 2$, and $c = 3$.

The iterative methods we have discussed so far are all linear, in the sense that they involve only methods of linear algebra. There are, however, also families of nonlinear iterative methods, involving the calculus of functions of several variables, of which one should be aware. These include gradient (steepest-descent) methods and conjugate-direction methods. A discussion of these methods is beyond the scope of this book.

Finally, another important warning. When we attempt to solve systems of linear equations on a computer, it is important to remember that the system may be very sensitive, and small changes in the entries of the coefficient matrix may lead to large changes in the solution. Such systems are said to be **ill-conditioned**. Applied mathematicians and others who design mathematical models often take considerable pains to avoid creating ill-conditioned systems.

Example: Consider the system of linear equations

$$\begin{bmatrix} 1 & -10 & 0 & 0 & 0 & 0 \\ 0 & 1 & -10 & 0 & 0 & 0 \\ 0 & 0 & 1 & -10 & 0 & 0 \\ 0 & 0 & 0 & 1 & -10 & 0 \\ 0 & 0 & 0 & 0 & 1 & -10 \\ 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix} X = \begin{bmatrix} -9 \\ -9 \\ -9 \\ -9 \\ -9 \\ 1 \end{bmatrix}$$

over \mathbb{R} . This system has a unique solution, namely $\begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}$. However, if we

alter the coefficient matrix by changing the (6,6)-entry to $\frac{1}{1.001}$ (which is

roughly equal to 0.9990009), we will obtain a completely different solution,

$$\text{namely } \begin{bmatrix} 101 \\ 11 \\ 2 \\ 1.1 \\ 1.01 \\ 1.001 \end{bmatrix}.$$

Since real-life computations are based, as a rule, on numbers gathered through some sort of measurement process, which is as a matter of course not completely accurate and certainly beyond our control, it is extremely important to know how sensitive the system is to possible small variations in the values of the entries. The numerical analysis of matrices deals extensively with this issue, and here we can only present a simplistic measure of this sensitivity for nonsingular square matrices over \mathbb{R} . To any matrix $A = [a_{ij}] \in \mathcal{M}_{n \times n}(\mathbb{R})$, we will assign the number $\theta(A)$ defined by $\theta(A) = \max_{1 \leq j \leq n} \left\{ \sum_{i=1}^n |a_{ij}| \right\}$. The number $\theta(A)\theta(A^{-1})$ is the **condition number**⁸ of the matrix A . Note that A has the same condition number as A^{-1} and as cA , for any $0 \neq c \in \mathbb{R}$.

The condition number can be written in the form $g \times 10^t$, where $0.1 \leq g < 1$. If $t > 0$ then, as a rule of thumb, one can expect that the solution of a system of linear equations $AX = w$ will have t significant digits *fewer* than that of the entries of A . Thus, if A is the matrix in the previous example, then $\theta(A) = 11$. Moreover,

$$A^{-1} = \begin{bmatrix} 1 & 10 & 100 & 1000 & 10000 & 100000 \\ 0 & 1 & 10 & 100 & 1000 & 10000 \\ 0 & 0 & 1 & 10 & 100 & 1000 \\ 0 & 0 & 0 & 1 & 10 & 100 \\ 0 & 0 & 0 & 0 & 1 & 10 \\ 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix}$$



⁸ Condition numbers were introduced by **John von Neumann**, one of the great mathematical geniuses of the 20th century, who contributed to practically all branches of mathematics – pure and applied. Von Neumann was a major force in the the introduction of digital computers after World War II and the development of numerical methods for them.

and so $\theta(A^{-1}) = 11,1111$. Therefore $\theta(A)\theta(A^{-1})$ is roughly 12×10^7 , and so we cannot, as we have seen, expect any accuracy in our solution, if we assume our data is only good to 6-digit accuracy.

Similarly, Nievergelt's matrix $\begin{bmatrix} 888445 & 887112 \\ 887112 & 885871 \end{bmatrix}$, which we have already encountered, has condition number roughly equal to 0.39×10^5 .

Of course, computing the condition number of a given matrix may also be a problem, since it involves calculating A^{-1} . Fortunately, there are many fairly-efficient condition number estimators, algorithms that give a good estimate of the condition number of a matrix with relatively low computational overhead.

Exercises

Exercise 522 Are the matrices $\begin{bmatrix} -3 & 4 & 1 \\ -2 & -4 & -6 \\ 5 & 2 & 7 \end{bmatrix}$ and $\begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 1 \\ 0 & 0 & 0 \end{bmatrix}$ in $\mathcal{M}_{3 \times 3}(\mathbb{R})$ row equivalent?

Exercise 523 Bring the matrix $\begin{bmatrix} 1 & 2 & 3 & 4 \\ 1 & 2 & 4 & 3 \\ 2 & 3 & 1 & 4 \end{bmatrix} \in \mathcal{M}_{3 \times 4}(\mathbb{R})$ to reduced row echelon form.

Exercise 524 Let $F = GF(5)$. Bring the matrix $\begin{bmatrix} 1 & 2 & 1 & 0 \\ 2 & 3 & 1 & 1 \\ 1 & 2 & 4 & 0 \end{bmatrix} \in \mathcal{M}_{3 \times 4}(F)$ to reduced row echelon form.

Exercise 525 Solve the system of linear equations

$$\begin{aligned} (3-i)X_1 + (2-i)X_2 + (4+2i)X_3 &= 2+6i \\ (4+3i)X_1 - (5+i)X_2 + (1+i)X_3 &= 2+2i \\ (2-3i)X_1 + (1-i)X_2 + (2+4i)X_3 &= 5i \end{aligned}$$

over \mathbb{C} .

Exercise 526 Solve the system of linear equations

$$\begin{aligned} X_1 + 2X_2 + 4X_3 &= 31 \\ 5X_1 + X_2 + 2X_3 &= 29 \\ 3X_1 - X_2 + X_3 &= 10 \end{aligned}$$

over \mathbb{R} .

Exercise 527 Solve the system of linear equations

$$\begin{aligned} 3X_1 + 4X_2 + 10X_3 &= 1 \\ 2X_1 + 2X_2 + 2X_3 &= 0 \\ X_1 + X_2 + 5X_3 &= 1 \end{aligned}$$

over $GF(11)$.

Exercise 528 Find all solutions to the system

$$\begin{bmatrix} 1 & 2 & 3 & 4 \\ 2 & 1 & 2 & 3 \\ 3 & 2 & 1 & 2 \\ 4 & 3 & 2 & 1 \end{bmatrix} \begin{bmatrix} X_1 \\ X_2 \\ X_3 \\ X_4 \end{bmatrix} = \begin{bmatrix} 5 \\ 1 \\ 1 \\ -5 \end{bmatrix}$$

over \mathbb{R} .

Exercise 529 Find all solutions to the system

$$\begin{bmatrix} 1 & 2 & 3 & 4 & 5 \\ 2 & 1 & 2 & 3 & 4 \\ 2 & 2 & 1 & 2 & 3 \\ 2 & 2 & 2 & 1 & 2 \\ 2 & 2 & 2 & 2 & 1 \end{bmatrix} \begin{bmatrix} X_1 \\ X_2 \\ X_3 \\ X_4 \\ X_5 \end{bmatrix} = \begin{bmatrix} 13 \\ 10 \\ 11 \\ 6 \\ 3 \end{bmatrix}$$

over \mathbb{R} .

Exercise 530 Find all solutions to the system

$$\begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 0 \\ 0 & 0 & 1 & 1 \end{bmatrix} \begin{bmatrix} X_1 \\ X_2 \\ X_3 \\ X_4 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}$$

over $GF(2)$.

Exercise 531 Find all solutions to the system

$$\begin{bmatrix} 1 & 3 & 2 \\ 2 & -1 & 3 \\ 3 & -5 & 4 \\ 1 & 17 & 4 \end{bmatrix} \begin{bmatrix} X_1 \\ X_2 \\ X_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

over \mathbb{R} .

Exercise 532 Find a real number a so that the system of linear equations

$$\begin{aligned} 2X_1 - X_2 + X_3 + X_4 &= 1 \\ X_1 + 2X_2 - X_3 + 4X_4 &= 2 \\ X_1 + 7X_2 - 4X_3 + 11X_4 &= a \end{aligned}$$

has a solution over \mathbb{R} .

Exercise 533 Find all real numbers c such that the system of equations

$$\begin{aligned} X_1 + X_2 - X_3 &= 1 \\ X_1 + cX_2 + 3X_3 &= 2 \\ 2X_1 + 3X_2 + cX_3 &= 3 \end{aligned}$$

has a unique solution over \mathbb{R} ; find those real numbers c for which it has infinitely-many solutions over \mathbb{R} ; find those real numbers c for which it has no solution over \mathbb{R} .

Exercise 534 Solve the system of linear equations

$$\begin{aligned} X_1 + 2X_2 + X_3 &= 1 \\ X_1 + X_2 + X_3 &= 0 \end{aligned}$$

over $GF(3)$.

Exercise 535 Solve the system of linear equations

$$\begin{aligned} X_1 + (\sqrt{2})X_2 + (\sqrt{2})X_3 &= 3 \\ X_1 + (1 + \sqrt{2})X_2 + X_3 &= 3 + \sqrt{2} \\ X_1 + X_2 - (\sqrt{2})X_3 &= 4 + \sqrt{2} \end{aligned}$$

over $\mathbb{Q}(\sqrt{2})$.

Exercise 536 Solve the system of linear equations

$$\begin{aligned} 4X_1 - 3X_2 &= 3 \\ 2X_1 - X_2 + 2X_3 &= 1 \\ 3X_1 + 2X_3 &= 4 \end{aligned}$$

over $GF(5)$.

Exercise 537 Solve the system of linear equations

$$\begin{aligned} 4X_1 + 6X_2 + 2X_3 &= 8 \\ X_1 - aX_2 - 2X_3 &= -5 \\ 7X_1 + 3X_2 + (a - 5)X_3 &= 7 \end{aligned}$$

over \mathbb{R} , for various values of the real number a .

Exercise 538 For a given real number a , solve the system

$$\begin{aligned} aX_1 + X_2 + X_3 &= 1 \\ X_1 + aX_2 + X_3 &= 1 \\ X_1 + X_2 + aX_3 &= 1 \end{aligned}$$

over \mathbb{R} .

Exercise 539 For $a \in \mathbb{R}$ does the system of linear equations

$$\begin{aligned} aX_1 + X_2 + 2X_3 &= 0 \\ X_1 - X_2 + aX_3 &= 1 \\ X_1 + X_2 + X_3 &= 1 \end{aligned}$$

have a unique solution in \mathbb{R} ?

Exercise 540 Let a be an element of a field F . Find the set of all solutions to the system of linear equations

$$\begin{aligned} X_1 + X_2 + aX_3 &= a \\ X_1 + aX_2 - X_3 &= 1 \\ X_1 + X_2 - X_3 &= 1 \end{aligned}$$

over F .

Exercise 541 For which $a \in \mathbb{Q}$ does the system of linear equations

$$\begin{aligned} X_1 + 3X_2 - 2X_3 &= 2 \\ 3X_1 + 9X_2 - 2X_3 &= 2 \\ 2X_1 + 6X_2 + X_3 &= a \end{aligned}$$

have a unique solution in \mathbb{Q} ?

Exercise 542 Find real numbers a , b , c , and d such that the points $(1, 2)$, $(-1, 6)$, $(-2, 38)$, and $(2, 6)$ all lie on the curve $y = ax^4 + bx^3 + cx^2 + d$ in the euclidean plane.

Exercise 543 Find a polynomial $p(X) = a_2X^2 + a_1X + a_0 \in \mathbb{R}[X]$ satisfying $p(1) = -1$, $p(-1) = 9$, and $p(2) = -3$.

Exercise 544 Find a polynomial $p(X) = a_3X^3 + a_2X^2 + a_1X + a_0 \in \mathbb{R}[X]$ satisfying $p(0) = 2$, $p(2) = 6$, $p(4) = 3$, and $p(6) = -5$.

Exercise 545 Let $F = GF(13)$. Find a homogeneous system of linear equations over F satisfying the condition that its solution space equals

$$F \left\{ \begin{bmatrix} 2 \\ 1 \\ 9 \\ 7 \\ 4 \end{bmatrix}, \begin{bmatrix} 8 \\ 3 \\ 10 \\ 5 \\ 12 \end{bmatrix}, \begin{bmatrix} 7 \\ 6 \\ 2 \\ 11 \\ 7 \end{bmatrix} \right\}.$$

Exercise 546 Let F be a field. Let $b \in F$ and let

$$A = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \end{bmatrix} \in \mathcal{M}_{2 \times 3}(F)$$

be a matrix satisfying the condition that the sum of the entries in each row and each column of A equals b . Show that $b = 0$.

Exercise 547 Let $p(X) = X^5 - 7X^3 + 12 \in \mathbb{Q}[X]$. Find a polynomial $q(X) \in \mathbb{Q}[X]$ of degree at most 3 satisfying $p(a) = q(a)$ for all $a \in \{0, 1, 2, 3\}$.

Exercise 548 Find the rank of the matrix
$$\begin{bmatrix} 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 1 & 0 \\ 1 & 1 & 1 & 1 & 0 & 0 \\ 0 & 1 & 1 & 1 & 0 & 0 \\ 0 & 1 & 1 & 0 & 1 & 0 \\ 1 & 0 & 0 & 1 & 1 & 0 \\ 1 & 1 & 1 & 0 & 1 & 0 \end{bmatrix} \in \mathcal{M}_{7 \times 6}(GF(2)).$$

Exercise 549 Find the rank of the matrix

$$\begin{bmatrix} 1 & -1 & 2 & 3 & 4 \\ 2 & 1 & -1 & 2 & 0 \\ -1 & 2 & 1 & 1 & 3 \\ 1 & 5 & -8 & -5 & -12 \\ 3 & -7 & 8 & 9 & 13 \end{bmatrix} \in \mathcal{M}_{5 \times 5}(\mathbb{R}).$$

Exercise 550 Find the rank of the matrix

$$\begin{bmatrix} 1 & 0 & 1 & 0 & 0 \\ 1 & 1 & 0 & 0 & 0 \\ 0 & 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & -1 & 0 \\ 0 & 1 & 0 & 1 & 1 \end{bmatrix} \in \mathcal{M}_{5 \times 5}(\mathbb{R}).$$

Exercise 551 Let $F = GF(2)$. Find the rank of the matrix

$$\begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \\ 1 & 0 & 1 \end{bmatrix} \in \mathcal{M}_{3 \times 3}(F).$$

Exercise 552 Let $F = GF(5)$. Find the rank of the matrix

$$\begin{bmatrix} 1 & 2 & 3 & 4 & a \\ 4 & 3 & a & 1 & 2 \\ a & 1 & 2 & 3 & 4 \\ 2 & 3a & 2 & 4a & 1 \end{bmatrix} \in \mathcal{M}_{4 \times 5}(F)$$

for various values of $a \in F$.

Exercise 553 Do there exist a lower-triangular matrix L and an upper-triangular matrix U in $\mathcal{M}_{3 \times 3}(\mathbb{Q})$ satisfying the condition $LU =$

$$\begin{bmatrix} 1 & -1 & 2 \\ 2 & -1 & 3 \\ 0 & 1 & 8 \end{bmatrix}?$$

Exercise 554 Let $F = GF(5)$. For which values of $a \in F$ do there exist a lower-triangular matrix L and an upper-triangular matrix U in

$$\mathcal{M}_{3 \times 3}(F) \text{ satisfying the condition that } LU = \begin{bmatrix} 1 & 1 & a \\ 4 & 1 & 0 \\ a & 1 & 4 \end{bmatrix}?$$

Exercise 555 Find the LU-decomposition of the matrix

$$A = \begin{bmatrix} 4 & 2 & 3 & 4 \\ 2 & 0 & 2 & 2 \\ 3 & 4 & -4 & 5 \\ -1 & 0 & 2 & 3 \end{bmatrix} \in \mathcal{M}_{4 \times 4}(\mathbb{R}).$$

Exercise 556 Let $A \in \mathcal{M}_{n \times n}(\mathbb{R})$ be a tridiagonal matrix all diagonal entries of which are nonzero. Can we write $A = LU$, where L is a lower-triangular matrix and U is an upper-triangular matrix, both of which are also tridiagonal?

Exercise 557 Let F be a field and let $a, b, c \in F$. Find the rank of the

$$\text{matrix } \begin{bmatrix} 1 & 1 & 1 \\ b+c & c+a & a+b \\ bc & ca & ab \end{bmatrix} \in \mathcal{M}_{3 \times 3}(F).$$

Exercise 558 Let F be a field and let $a, b, c, d \in F$. Find the rank of

$$\text{the matrix } \begin{bmatrix} a & c & c \\ d & a+b & c \\ d & d & b \end{bmatrix} \in \mathcal{M}_{3 \times 3}(F).$$

$$\text{Exercise 559 Find the rank of the matrix } \begin{bmatrix} 3 & 1 & 1 & 4 \\ a & 4 & 10 & 1 \\ 1 & 7 & 17 & 3 \\ 2 & 2 & 4 & 3 \end{bmatrix} \in \mathcal{M}_{4 \times 4}(\mathbb{R})$$

for various values of the real number a .

Exercise 560 Find the rank of the matrix

$$\begin{bmatrix} a & -1 & 2 & 1 \\ -1 & a & 5 & 2 \\ 10 & -6 & 1 & 1 \end{bmatrix} \in \mathcal{M}_{3 \times 4}(\mathbb{Q})$$

for various values of the rational number a .

Exercise 561 Find the set of all real numbers a such that the rank of

$$\text{the matrix } \begin{bmatrix} a & 1 & 1 \\ -1 & -1 & -1 \\ 1 & 1 & a \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{R}) \text{ equals } 2.$$

Exercise 562 Let n be a positive integer and, for each $A \in \mathcal{M}_{n \times n}(\mathbb{R})$, let $r(A)$ be the rank of A . Define a relation \preceq on $\mathcal{M}_{n \times n}(\mathbb{R})$ by setting $B \preceq A$ if and only if $r(A - B) = r(A) - r(B)$. Is this a partial order relation?

Exercise 563 Let F be a subfield of a field K . Let k and n be positive integers and let $A \in \mathcal{M}_{k \times n}(F)$ be a matrix having rank r . If we now think of A as an element of $\mathcal{M}_{k \times n}(K)$, is its rank necessarily still equal to r ?

Exercise 564 Find an integer a such that the rank of the matrix

$$\begin{bmatrix} 1 & 7 & 17 & 3 \\ 4 & 4 & 8 & 6 \\ 3 & 1 & 1 & 4 \\ 2a & 8 & 20 & 2 \end{bmatrix} \in \mathcal{M}_{4 \times 4}(\mathbb{Q})$$

is minimal.

Exercise 565 Let F be a field and let k and n be positive integers. For a matrix $A \in \mathcal{M}_{k \times n}(F)$ having rank h , show that there exist matrices $B \in \mathcal{M}_{k \times h}(F)$ and $C \in \mathcal{M}_{h \times n}(F)$ such that $A = BC$.

Exercise 566 Let k and n be positive integers and let F be a field. For matrices $A, B \in \mathcal{M}_{k \times n}(F)$, show that the rank of $A + B$ is no more than the sum of the ranks of A and of B .

Exercise 567 Let k and n be positive integers and let F be a field. Let $A, B \in \mathcal{M}_{k \times n}(F)$ be matrices satisfying the condition that the row space of A and the row space of B are disjoint. Does it follow from this that the rank of $A + B$ equals the sum of the rank of A and the rank of B ?

Exercise 568 Find bases for the row space and column space of the matrix

$$\begin{bmatrix} 1 & 2 & -3 & -7 & -2 \\ -1 & -2 & 1 & 1 & 0 \\ 1 & 2 & 0 & 2 & 1 \end{bmatrix} \in \mathcal{M}_{3 \times 5}(\mathbb{R}).$$

Exercise 569 Find matrices $P, Q \in \mathcal{M}_{3 \times 3}(\mathbb{R})$ satisfying

$$P \begin{bmatrix} 1 & 2 & 3 \\ 2 & -2 & 1 \\ 3 & 0 & 4 \end{bmatrix} Q = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix}.$$

Exercise 570 Write the rows of the matrix $A = \begin{bmatrix} 1 & 2 & 0 \\ i-1 & 2 & i \\ 0 & 2 & -i \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{C})$ as linear combinations of the rows of A^T .

Exercise 571 Calculate $\begin{bmatrix} 1 & 2 & 3 & 4 & 5 \\ 0 & 1 & 2 & 3 & 4 \\ 0 & 0 & 1 & 2 & 3 \\ 0 & 0 & 0 & 1 & 2 \\ 0 & 0 & 0 & 0 & 1 \end{bmatrix}^{-1} \in \mathcal{M}_{5 \times 5}(\mathbb{R})$.

Exercise 572 Let k and n be positive integers and let F be a field. Let $A = \begin{bmatrix} B & C \\ D & E \end{bmatrix}$ be a matrix in $\mathcal{M}_{k \times n}(F)$, where B is a nonsingular matrix in $\mathcal{M}_{r \times r}(F)$ for some $1 \leq r < \min\{k, n\}$. Show that the rank of A equals r if and only if $DB^{-1}C = E$.

Exercise 573 Let F be a field and let a, b, c be distinct elements of F . Furthermore, let d, e, f be distinct elements of F . What is the rank of

the matrix $\begin{bmatrix} 1 & a & d & ad \\ 1 & b & e & be \\ 1 & c & f & cf \end{bmatrix} \in \mathcal{M}_{3 \times 4}(F)$?

Exercise 574 Let k and n be positive integers and let F be a field. Let $A \in \mathcal{M}_{k \times n}(F)$ and let $w \in F^k$ be such that the system of linear equations $AX = w$ has a nonempty set of solutions and that all of these solutions satisfy the condition that the h th entry in them is some fixed scalar c . What can we deduce about the columns of the matrix A ?

Exercise 575 Let n be a positive integer and let F be a field. Let $O \neq A \in \mathcal{M}_{n \times n}(F)$. Show that there exists a nonnegative integer k such that the rank of A^h equals the rank of A^k for all $h > k$.

Exercise 576 Let $A = \begin{bmatrix} 1 & -1 & 1 \\ -1 & -1 & 1 \\ 1 & 1 & 1 \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{R})$. Find the condition number of A .

Exercise 577 Let a be a positive real number. It is necessarily true that

the condition number of $A = \begin{bmatrix} a & 1 & a \\ 0 & 0 & -1 \\ a & -1 & a \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{R})$ is greater than $2a + 1$?

Exercise 578 Find a positive real number a for which the condition

number of $A = \begin{bmatrix} 1 & 1 & 0 \\ 0 & a & a \\ 1 & 1 & 1 \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{R})$ is maximal.

Exercise 579 Does there exist a system $AX = w$ of linear equations in n unknowns (for some positive integer n) over \mathbb{R} having precisely 35 distinct solutions?

11

Determinants

Let F be a field and let n be a positive integer. We would like to find a function from $\mathcal{M}_{n \times n}(F)$ to F which will serve as an oracle of singularity, namely a function that will assign a value of 0 to singular matrices and a value other than 0 to nonsingular matrices. Indeed, let F be a field and let n be a positive integer. A function $\delta_n : \mathcal{M}_{n \times n}(F) \rightarrow F$ is a **determinant function** if and only if it satisfies the following conditions:

- (1) $\delta_n(I) = 1$;
- (2) $\delta_n(A) = 0$ if A is a matrix having a row all of the entries of which are 0;
- (3) $\delta_n(E_{ij}A) = -\delta_n(A)$ for all $1 \leq i \neq j \leq n$;
- (4) $\delta_n(E_{ij;c}A) = \delta_n(A)$ for all $1 \leq i \neq j \leq n$ and all $c \in F$;
- (5) $\delta_n(E_{i;c}A) = c\delta_n(A)$ for all $1 \leq i \leq n$ and all $0 \neq c \in F$.

In particular, we note that for each $1 \leq i \neq j \leq n$ and all $c \in F$ we have $\delta_n(E_{ij}) = -1 = \delta_n(E_{ij}^T)$, $\delta_n(E_{ij;c}) = 1 = \delta_n(E_{ij;c}^T)$, and $\delta_n(E_{i;c}) = c = \delta_n(E_{i;c}^T)$.

We have yet to show that such functions exist for all values of n , but certainly they exist for a few small ones.

Example: For $n = 1$, the function $\delta_1 : [a] \mapsto a$ is a determinant function. For $n = 2$, the function $\delta_2 : \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \mapsto a_{11}a_{22} - a_{12}a_{21}$ is a determinant function.

As an immediate consequence of parts (1) and (5) of the definition, we see that if $A = [a_{ij}] \in \mathcal{M}_{n \times n}(F)$ is a diagonal matrix and if $\delta_n : \mathcal{M}_{n \times n}(F) \rightarrow F$ is a determinant function, then $\delta_n(A) = \prod_{i=1}^n a_{ii} \delta_n(I) = \prod_{i=1}^n a_{ii}$.

We now want to show that for each positive integer n there exists determinant function $\delta_n : \mathcal{M}_{n \times n}(F) \rightarrow F$, and indeed that this function is unique. We will first establish the uniqueness of these functions and check some of their properties, holding off on existence until later in this chapter.

(11.1) Proposition: Let F be a field. For each positive integer n there exists at most one determinant function $\delta_n : \mathcal{M}_{n \times n}(F) \rightarrow F$.

Proof: Let us assume that $\delta_n : \mathcal{M}_{n \times n}(F) \rightarrow F$ and $\eta_n : \mathcal{M}_{n \times n}(F) \rightarrow F$ are determinant functions and let $\beta = \eta_n - \delta_n$. Then the function β satisfies the following conditions:

- (1) $\beta(I) = 0$;
- (2) $\beta(A) = 0$ if A is a matrix having a row all of the entries of which are 0;
- (3) $\beta(E_{ij}A) = -\beta(A)$ for all $1 \leq i \neq j \leq n$;
- (4) $\beta(E_{ij;c}A) = \beta(A)$ for all $1 \leq i \neq j \leq n$ and all $c \in F$;
- (5) $\beta(E_{i;c}A) = c\beta(A)$ for all $1 \leq i \leq n$ and all $0 \neq c \in F$.

In particular, if $A \in \mathcal{M}_{n \times n}(F)$ and E is an elementary matrix, then $\beta(A)$ and $\beta(EA)$ are either both equal to 0 or both of them are different from 0. But for any matrix A we know that there exist elementary matrices E_1, \dots, E_t in $\mathcal{M}_{n \times n}(F)$ such that either $E_1 \cdots E_t A = I$ or $E_1 \cdots E_t A$ is a matrix having at least one row all of the entries of which equal 0. Therefore $\beta(A) = 0$ for every $A \in \mathcal{M}_{n \times n}(F)$. Thus β is the zero-function, and so $\delta_n = \eta_n$. \square

(11.2) Proposition: Let F be a field and let $\delta_n : \mathcal{M}_{n \times n}(F) \rightarrow F$ be a determinant function. Then $\delta_n(A) \neq 0$ if and only if A is nonsingular.

Proof: If A is nonsingular, there exist elementary matrices E_1, \dots, E_t in $\mathcal{M}_{n \times n}(F)$ such that $E_1 \cdots E_t A = I$, and so, by the definition of the determinant function, $\delta_n(A) = c\delta_n(I) = c$, where $0 \neq c \in F$, and so $\delta_n(A) \neq 0$. Now assume that $\delta_n(A) \neq 0$ and that A is singular. Then there exist elementary matrices E_1, \dots, E_t in $\mathcal{M}_{n \times n}(F)$ such that $E_1 \cdots E_t A$ is a matrix having at least one row all of the entries of which equal 0. But then, for some $0 \neq c \in F$, we have $0 \neq \delta_n(A) = c\delta_n(E_1 \cdots E_t A) = c0 = 0$, which is a contradiction, proving that A must be nonsingular. \square

Thus we see that the determinant function, to the extent it exists, is the oracle we are seeking.

Example: The subset $\left\{ \begin{bmatrix} a+bi \\ c+di \end{bmatrix}, \begin{bmatrix} -c+di \\ a-bi \end{bmatrix} \right\}$ of \mathbb{C}^2 is linearly dependent if and only if $A = \begin{bmatrix} a+bi & -c+di \\ c+di & a-bi \end{bmatrix} \in \mathcal{M}_{2 \times 2}(\mathbb{C})$ is singular.

We have already noted that $\delta_2 : \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \mapsto a_{11}a_{22} - a_{12}a_{21}$ is a determinant function, and so this happens if and only if $\delta_2(A) = a^2 + b^2 + c^2 + d^2 = 0$, i.e. if and only if $a = b = c = d = 0$.

(11.3) Proposition: Let F be a field and let $\delta_n : M_{n \times n}(F) \rightarrow F$ be a determinant function. If A is a matrix in $M_{n \times n}(F)$ having two identical rows then $\delta_n(A) = 0$.

Proof: Suppose that rows h and k of A are identical. First assume that the characteristic of F is other than 2. Then $A = E_{hk}(A)$ and so $\delta_n(A) = \delta_n(E_{hk}A) = -\delta_n(A)$, which implies that $\delta_n(A) = 0$. If the characteristic of F equals 2 then $\delta_n(A) = \delta_n(E_{hk;1}A)$, and $E_{hk;1}A$ is a matrix having a row in which the entries of one row are all 0. Therefore, by Proposition 11.2, $\delta_n(A) = 0$. \square

(11.4) Proposition: Let F be a field and let $\delta_n : M_{n \times n}(F) \rightarrow F$ be a determinant function. If $A, B \in M_{n \times n}(F)$ then

- (1) $\delta_n(AB) = \delta_n(A)\delta_n(B)$;
- (2) $\delta_n(AB) = \delta_n(BA)$.

Proof: (1) By Proposition 9.1, we know that AB is nonsingular if and only if both A and B are nonsingular. Therefore $\delta_n(A) = 0$ or $\delta_n(B) = 0$ if and only if $\delta_n(AB) = 0$. If $\delta_n(A) \neq 0 \neq \delta_n(B)$ then there exist elementary matrices $E_1, \dots, E_t, G_1, \dots, G_s$ in $M_{n \times n}(F)$ such that $B = E_1 \cdot \dots \cdot E_t I$ and $A = G_1 \cdot \dots \cdot G_s I$ and so $AB = G_1 \cdot \dots \cdot G_s E_1 \cdot \dots \cdot E_t I$, which implies that $\delta_n(AB) = \delta_n(A)\delta_n(B)$ from the definition of a determinant function.

(2) This is an immediate consequence of (1), since $\delta_n(A)\delta_n(B) = \delta_n(B)\delta_n(A)$ in F . \square

(11.5) Proposition: Let F be a field and let $\delta_n : M_{n \times n}(F) \rightarrow F$ be a determinant function. If $A \in M_{n \times n}(F)$ is nonsingular then $\delta_n(A^{-1}) = \delta_n(A)^{-1}$.

Proof: By Proposition 11.4, we see that $\delta_n(A^{-1})\delta_n(A) = \delta_n(A^{-1}A) = \delta_n(I) = 1$ and from this the result follows immediately. \square

(11.6) Proposition: Let F be a field and let $\delta_n : M_{n \times n}(F) \rightarrow F$ be a determinant function. If $A \in M_{n \times n}(F)$ then:

- (1) $\delta_n(AE_{ij}) = -\delta_n(A)$ for all $1 \leq i \neq j \leq n$;
- (2) $\delta_n(AE_{ij;c}) = \delta_n(A)$ for all $1 \leq i \neq j \leq n$ and all $c \in F$;
- (3) $\delta_n(AE_{i;c}) = c\delta_n(A)$ for all $1 \leq i \leq n$ and all $0 \neq c \in F$.

Proof: This is a direct consequence of the definition of the determinant function and Proposition 11.4(2). \square

(11.7) Proposition: Let F be a field and let $\delta_n : M_{n \times n}(F) \rightarrow F$ be a determinant function. If $A \in M_{n \times n}(F)$ then $\delta_n(A) = \delta_n(A^T)$.

Proof: If A is singular then so is A^T and so $\delta_n(A) = 0 = \delta_n(A^T)$. If A is nonsingular then there exist elementary matrices E_1, \dots, E_t in $M_{n \times n}(F)$ such that $E_1 \cdots E_t A = I = I^T = A^T E_t^T \cdots E_1^T$. By our remarks in Chapter 9 concerning the transposes of elementary matrices, and by the remarks at the beginning of this chapter, we see that $\delta_n(A) = \delta_n(E_1 \cdots E_t A) = \delta_n(A^T E_t^T \cdots E_1^T) = \delta_n(A^T)$ and so $\delta_n(A) = \delta_n(A^T)$. \square

Of course, at this stage we do not know that determinant functions $\delta_n : M_{n \times n}(F) \rightarrow F$ even exist for the case $n > 2$ and so we now have to construct them. Let us denote the set of all permutations of the set $\{1, \dots, n\}$ by S_n . We note that any $\pi \in S_n$ is a bijective function from $\{1, \dots, n\}$ to itself and so there exists a function $\pi^{-1} \in S_n$ satisfying the condition that $\pi\pi^{-1}$ and $\pi^{-1}\pi$ are equal to the identity function $i \mapsto i$. We also note that if $\pi, \pi' \in S_n$ then $\pi\pi' \in S_n$.

(11.8) Proposition: If n is a positive integer then the number of elements of S_n equals $n!$.

Proof: Suppose we wanted to construct an arbitrary element π of S_n . There are n possibilities for selecting $\pi(1)$. Once we have done that, there are $n-1$ ways of selecting $\pi(2)$, then $n-2$ ways of selecting $\pi(3)$, etc. Thus, the total number of ways in which we can define π is $n(n-1) \cdots 1 = n!$. \square

Now let $\pi \in S_n$ and let $1 \leq i < j \leq n$. The pair (i, j) is called an **inversion** with respect to π if and only if $\pi(i) > \pi(j)$. That is to say, (i, j) is an inversion with respect to π if and only if

$$\frac{i - j}{\pi(i) - \pi(j)} < 0.$$

We will denote the number of distinct inversions with respect to π by $h(\pi)$, and define the **signum** of π to be $sgn(\pi) = (-1)^{h(\pi)}$. Thus

$$sgn(\pi) = \begin{cases} 1 & \text{if there are an even number of inversions} \\ & \text{with respect to } \pi \\ -1 & \text{if there are an odd number of inversions} \\ & \text{with respect to } \pi \end{cases}.$$

It is easy to check that $sgn(\pi) = sgn(\pi^{-1})$ for all $\pi \in S_n$. If $sgn(\pi) = 1$, the permutation π is **even**; if $sgn(\pi) = -1$, the permutation π is **odd**.

Example: Let $\pi \in S_4$ be defined by $1 \mapsto 3$, $2 \mapsto 4$, $3 \mapsto 2$, and $4 \mapsto 1$. Then if we consider all possible pairs (i, j) with $1 \leq i < j \leq 4$ we get

(i, j)	$(\pi(i), \pi(j))$	inversion?
(1,2)	(3,4)	no
(1,3)	(3,2)	yes
(1,4)	(3,1)	yes
(2,3)	(4,2)	yes
(2,4)	(4,1)	yes
(3,4)	(2,1)	yes

and so we see that $sgn(\pi) = -1$.

Now let n be a positive integer and let (K, \bullet) be an associative and commutative unital F -algebra. Let $A = [a_{ij}] \in \mathcal{M}_{n \times n}(K)$. We then define the function $A \mapsto |A|$ from $\mathcal{M}_{n \times n}(K)$ to K by setting

$$|A| = \sum_{\pi \in S_n} sgn(\pi) a_{\pi(1),1} \bullet a_{\pi(2),2} \bullet \dots \bullet a_{\pi(n),n}.$$

Note that, by the commutativity of K , if $\tau = \pi^{-1}$ then

$$a_{\pi(1),1} \bullet a_{\pi(2),2} \bullet \dots \bullet a_{\pi(n),n} = a_{1,\tau(1)} \bullet a_{2,\tau(2)} \bullet \dots \bullet a_{n,\tau(n)}$$

and so $|A| = \sum_{\tau \in S_n} sgn(\tau) a_{1,\tau(1)} \bullet a_{2,\tau(2)} \bullet \dots \bullet a_{n,\tau(n)}$. Thus we see immediately that $|A| = |A^T|$ for every $A \in \mathcal{M}_{n \times n}(K)$. If $K = \mathbb{C}$ then, since $\overline{c+d} = \bar{c} + \bar{d}$ and $\overline{cd} = \bar{c}\bar{d}$, we also see that for $\bar{A} = [\bar{a}_{ij}]$ we have $|\bar{A}| = \overline{|A|}$. Defining this function for an arbitrary commutative and associative unital F -algebra is important for us, as we will need it in the case that $K = F[X]$, where F is a field.

Example: If $A = [a_{ij}] \in \mathcal{M}_{3 \times 3}(K)$, for an associative and commutative unital F -algebra (K, \bullet) , then

$$\begin{aligned} |A| &= a_{11} \bullet a_{22} \bullet a_{33} + a_{12} \bullet a_{23} \bullet a_{31} + a_{13} \bullet a_{21} \bullet a_{32} \\ &\quad - a_{11} \bullet a_{23} \bullet a_{32} - a_{13} \bullet a_{22} \bullet a_{31} - a_{12} \bullet a_{21} \bullet a_{33}. \end{aligned}$$

(11.9) Proposition: Let F be a field, let (K, \bullet) be an associative and commutative unital F -algebra, and let $A = [a_{ij}] \in M_{n \times n}(K)$. Pick $1 \leq h \leq n$ and write $a_{hj} = b_{hj} + c_{hj}$ in K for all $1 \leq j \leq n$. For all $1 \leq i \leq n$ satisfying $i \neq h$, set $b_{ij} = c_{ij} = a_{ij}$. Set $B = [b_{ij}]$ and $C = [c_{ij}]$, matrices in $M_{n \times n}(K)$. Then $|A| = |B| + |C|$.

Proof: From the definition of $|A|$, we have

$$\begin{aligned}
 |A| &= \sum_{\pi \in S_n} \operatorname{sgn}(\pi) a_{1,\pi(1)} \bullet \dots \bullet a_{h,\pi(h)} \bullet \dots \bullet a_{n,\pi(n)} \\
 &= \sum_{\pi \in S_n} \operatorname{sgn}(\pi) a_{1,\pi(1)} \bullet \dots \bullet [b_{h,\pi(h)} + c_{h,\pi(h)}] \bullet \dots \bullet a_{n,\pi(n)} \\
 &= \sum_{\pi \in S_n} \operatorname{sgn}(\pi) a_{1,\pi(1)} \bullet \dots \bullet b_{h,\pi(h)} \bullet \dots \bullet a_{n,\pi(n)} \\
 &\quad + \sum_{\pi \in S_n} \operatorname{sgn}(\pi) a_{1,\pi(1)} \bullet \dots \bullet c_{h,\pi(h)} \bullet \dots \bullet a_{n,\pi(n)} \\
 &= |B| + |C|,
 \end{aligned}$$

as required. \square

We are now ready to prove that determinant functions in fact always exist.

(11.10) Proposition: For an integer $n > 1$ and a field F , the function $M_{n \times n}(F) \rightarrow F$ defined by $A \mapsto |A|$ is a determinant function.

Proof: In order to simplify our notation, we will make the following temporary convention: if $\pi \in S_n$ and if $A = [a_{ij}] \in M_{n \times n}(F)$, we will write $u(\pi, A) = \operatorname{sgn}(\pi) a_{1,\pi(1)} \dots a_{n,\pi(n)}$. Now let us check the five conditions of a determinant function.

(1) Clearly $u(\pi, I)$ equals 1 if π is the identity permutation and 0 otherwise, and so $|I| = 1$.

(2) Let A be a matrix one of the rows of which has all of its entries equal to 0. Since a factor from each row appears in every term $u(\pi, A)$, we conclude that all of these are equal to 0 and hence $|A| = 0$.

(3) Let A be a matrix and let $B = E_{ij}A$. Let $\rho \in S_n$ be the permutation which interchanges i and j and leaves all of the other numbers between 1 and n fixed. Then $\operatorname{sgn}(\pi\rho) \neq \operatorname{sgn}(\pi)$ for all $\pi \in S_n$ and so for each $\pi \in S_n$ we have $-u(\pi, A) = u(\pi\rho, A) = u(\pi, B)$. This implies that $|B| = -|A|$.

(4) Let A be a matrix and let $B = E_{ij;c}A$. Then $B = [b_{ht}]$, where $b_{ht} = a_{ht}$ when $h \neq j$ and $1 \leq t \leq n$, and where $b_{jt} = a_{jt} + ca_{it}$ for all

$1 \leq t \leq n$. By Proposition 11.9, we have $|B| = |A| + |C|$, where C is the matrix all of the rows of which except the j th are identical with those of A , and where in the j th row we have $c_{jt} = ca_{it}$ for all $1 \leq t \leq n$. Then $|C| = c|D|$ where D is a matrix in which two rows, the i th and the j th, are equal. If the characteristic of F is other than 2, then $D = E_{ij}D$ and so, by (3), we get $|D| = -|D|$, and so we get $|C| = c|D| = 0$ and we have $|A| = |B|$, which is what we want. Therefore let us assume that the characteristic of F equals 2. Let $\rho \in S_n$ be the permutation which interchanges i and j and leaves all other numbers between 1 and n fixed. Let H be the set of all even permutations in S_n and let K be the set of all odd permutations. The function from H to K defined by $\pi \mapsto \rho\pi$ is bijective since $\rho\pi_1 = \rho\pi_2$ implies that $\pi_1 = \rho^{-1}\rho\pi_1 = \rho^{-1}\rho\pi_2 = \pi_2$. Moreover, since the characteristic of F is 2 and since $u(\pi, D) = u(\rho\pi, D)$ for all $\pi \in H$, we see that $u(\pi, D) + u(\rho\pi, D) = 0$ for all $\pi \in H$. Therefore $|D| = \sum_{\pi \in H} [u(\pi, D) + u(\rho\pi, D)] = 0$ and this implies, again, that $|C| = 0$ and so $|A| = |B|$.

(5) It is clear from the definition of $|A|$ and if $B = E_{i;c}A$ then $|B| = c|A|$. \square

Thus, in summary, we see that if F is a field and if n is a positive integer, then there exists a unique determinant function $\mathcal{M}_{n \times n}(F) \rightarrow F$, namely $A \mapsto |A|$. We call the scalar $|A|$ the **determinant**¹ of the matrix A .

Example: Let $n > 1$ be an integer. If c_1, \dots, c_n are distinct elements of a field F and if $A = [a_{ij}] \in \mathcal{M}_{n \times n}(F)$ is the Vandermonde matrix defined by $a_{ij} = c_i^{j-1}$ for all $1 \leq i, j \leq n$, then it is easy to verify that $|A| = \prod_{i < j} (c_j - c_i) \neq 0$.

Example: As a consequence of Proposition 11.7, we note that if $n > 0$ is odd and if $A \in \mathcal{M}_{n \times n}(F)$ is a skew-symmetric matrix then $|A| =$



¹ Determinants were first used in the work of the 17th-century German mathematician, philosopher, and diplomat **Gottfried von Leibniz**, who developed calculus along with Sir Isaac Newton. The common properties of determinants were first studied by 19th-century German mathematician **Heinrich Scherk**, and the first systematic analysis of the theory of determinants was done by the 19th-century French mathematician **Augustin-Louis Cauchy**, relying on the work of many mathematicians who preceded him. His work was continued by Cayley and Sylvester.

$|A^T| = |-A| = -|A|$ and so $|A| = 0$. Therefore, by Proposition 11.2, A is singular. If n is even, then one can use the definition of $|A|$ to show that $|A| = b^2$ for some b which is a sum of products of the a_{ij} . Thus, for example,

$$\begin{vmatrix} 0 & a_{12} & a_{13} & a_{14} \\ -a_{12} & 0 & a_{23} & a_{24} \\ -a_{13} & -a_{23} & 0 & a_{34} \\ -a_{14} & -a_{24} & -a_{34} & 0 \end{vmatrix} = [a_{12}a_{34} - a_{13}a_{24} + a_{14}a_{23}]^2.$$

This number b is called the **Pfaffian**² of the matrix A . Pfaffians arise naturally in combinatorics, differential geometry and other areas of mathematics.

We now give two examples of why it was worthwhile to define $|A|$ for matrices A with entries in an associative and commutative unital F -algebra, and not just a field.

Example: Let V be a vector space of finite dimension n over a field F and let $B = \{v_1, \dots, v_k\}$ be a linearly-independent subset of V . Let y_1, \dots, y_k be a list of vectors in V . We claim that there are at most finitely-many elements a of F satisfying the condition that the list $v_1 + ay_1, \dots, v_k + ay_k$ is linearly dependent. To establish this claim, we will consider determinants of matrices over $F[X]$. Indeed, extend B to a basis $D = \{v_1, \dots, v_n\}$ of V . Then, for each $1 \leq i \leq k$, we can write $y_i = \sum_{j=1}^n c_{ij}w_j$. For each $1 \leq i, j \leq k$, define the polynomial $p_{ij}(X) \in F[X]$ by setting

$$p_{ij}(X) = \begin{cases} c_{ii}X + 1 & \text{if } i = j \\ c_{ij}X & \text{otherwise} \end{cases}$$

and consider the matrix $B = [p_{ij}(X)] \in \mathcal{M}_{k \times k}(F[X])$. Then $|B|$ is a polynomial $q(X)$ in $F[X]$, which is not the 0-polynomial since $q(0) = 1$. Moreover, for any $a \in F$, we see that $q(a) = 0$ whenever the list

$$v_1 + ay_1, \dots, v_k + ay_k$$



² Pfaffians were first defined by Cayley, and named in honor of **Johann Pfaff**, an 18th-century German mathematician whose most famous doctoral student was Gauss.

is linearly dependent. Since a polynomial can have only finitely-many distinct roots, this can happen only for finitely-many values of a .

Example: Let $n > 1$ be an integer and let U be an open interval of real numbers. Let K be the set of all functions in \mathbb{R}^U which are differentiable at least $n - 1$ times. Then K is an associative and commutative unital \mathbb{R} -algebra which is not entire, let alone a field. We will denote the derivative of a function $f \in K$ by Df and, if $h > 1$, we will denote the h th derivative of f by $D^h f$. Given $f_1, \dots, f_n \in K$, the function

$$W(f_1, \dots, f_n) : t \mapsto \begin{vmatrix} f_1(t) & f_2(t) & \dots & f_n(t) \\ (Df_1)(t) & (Df_2)(t) & \dots & (Df_n)(t) \\ \vdots & \vdots & \ddots & \vdots \\ (D^{n-1}f_1)(t) & (D^{n-1}f_2)(t) & \dots & (D^{n-1}f_n)(t) \end{vmatrix}$$

is called the **Wronskian**³ of f_1, \dots, f_n . One can show that if we have $W(f_1, \dots, f_n)(t) \neq 0$ for some $t \in U$ then the subset $\{f_1, \dots, f_n\}$ of K is linearly independent over \mathbb{R} . The converse is false. To see this, let U be an open interval containing the origin, let $f_1 : t \mapsto t^3$, and let $f_2 : t \mapsto |t^3|$. Then $\{f_1, f_2\}$ is linearly independent over \mathbb{R} , but $W(f_1, f_2)(t) = 0$ for any $t \in U$.

Example: Let n be a positive integer equal to 2 or divisible by 4. A matrix $A = [a_{ij}] \in \mathcal{M}_{n \times n}(\mathbb{C})$ with $|a_{ij}| \leq 1$ for all $1 \leq i, j \leq n$ having maximal possible determinant (in absolute value) is known as a **Hadamard matrix** (though in fact such matrices were studied by Sylvester a generation before Hadamard considered them). For such a matrix, we have $|A| = n^{n/2}$, and the entries of A are all ± 1 . Indeed, a matrix A is a Hadamard matrix precisely when all of its entries are ± 1 and $AA^T = nI$.

Thus, $\begin{bmatrix} -1 & 1 & 1 & 1 \\ 1 & -1 & 1 & 1 \\ 1 & 1 & -1 & 1 \\ 1 & 1 & 1 & -1 \end{bmatrix}$ and $\begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -1 & 1 & -1 \\ 1 & 1 & -1 & -1 \\ 1 & -1 & -1 & 1 \end{bmatrix}$ are Hadamard



³ The insight of **Josef Wronski**, a 19th-century Polish mathematician living in France, was obscured by his decidedly eccentric philosophical ideas and style of writing, and was recognized only after his death. The notion of a determinant of functions was first used by Jacobi.

matrices. Indeed, for each $t \geq 1$, there exists a Hadamard matrix H_t of size $2^t \times 2^t$, defined recursively by setting $H_1 = \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}$ and $H_t = \begin{bmatrix} H_{t-1} & H_{t-1} \\ H_{t-1} & -H_{t-1} \end{bmatrix}$ for each $t > 1$.

We also note immediately that if A is a Hadamard matrix so are A^T and $-A$. Hadamard matrices have important applications in algebraic coding theory, especially in defining the error-correcting Reed-Muller codes. Needless to say, the determinants of Hadamard matrices get very big very quickly. If A is a 16×16 Hadamard matrix, then $|A| = 4,294,967,296$ and if B is a 32×32 Hadamard matrix, then $|B| = 1,208,925,819,614,629,174,706,176$.

We still are faced with the problem of actually computing the determinant of an $n \times n$ matrix A , especially when n is large. If we work using the definition, we see that we must add $n!$ summands, each of which requires $n-1$ multiplications. The total number of arithmetic operations need is therefore $(n-1)n! + (n! - 1) = n(n!) - 1$, which is a huge number even if n is relatively small. For example, if we are using a computer capable of performing a billion arithmetic operations per second, it would take us 12,200,000,000 years of nonstop computation to compute the determinant of a 25×25 matrix, based on the definition. Thus we must find better methods of computing determinants, a task which became a high priority for many nineteenth-century mathematicians.

Example: Let $A = [a_{ij}] \in \mathcal{M}_{n \times n}(F)$ be a matrix in which $a_{11} \neq 0$. Then Chiò, Dodgson⁴, and others showed that $|A| = a_{11}^{2-n}|B|$, where $B \in \mathcal{M}_{(n-1) \times (n-1)}(F)$ is the matrix obtained from A by erasing the first row and first column and replacing each other a_{ij} by $\begin{vmatrix} a_{11} & a_{1j} \\ a_{i1} & a_{ij} \end{vmatrix}$.



⁴ During the 19th century, matrix theory and the theory of determinants attracted many gifted mathematicians and mathematical amateurs. **Felice Chiò** was a 19th-century Italian mathematician and physicist. On the other hand, **Rev. Charles Lutwidge Dodgson** was an amateur who is better known by his pen name Lewis Carroll, the author of *Alice in Wonderland*. Dodgson published several works on mathematics under his own name.

Thus, for example,

$$\begin{vmatrix} 1 & 2 & 3 & 4 \\ 8 & 7 & 6 & 5 \\ 1 & 8 & 2 & 7 \\ 3 & 6 & 4 & 5 \end{vmatrix} = \begin{vmatrix} \begin{vmatrix} 1 & 2 \\ 8 & 7 \end{vmatrix} & \begin{vmatrix} 1 & 3 \\ 8 & 6 \end{vmatrix} & \begin{vmatrix} 1 & 4 \\ 8 & 5 \end{vmatrix} \\ \begin{vmatrix} 1 & 2 \\ 1 & 8 \end{vmatrix} & \begin{vmatrix} 1 & 3 \\ 1 & 2 \end{vmatrix} & \begin{vmatrix} 1 & 4 \\ 1 & 7 \end{vmatrix} \\ \begin{vmatrix} 1 & 2 \\ 3 & 6 \end{vmatrix} & \begin{vmatrix} 1 & 3 \\ 3 & 4 \end{vmatrix} & \begin{vmatrix} 1 & 4 \\ 3 & 5 \end{vmatrix} \end{vmatrix} \\
 = \begin{vmatrix} -9 & -18 & -27 \\ 6 & -1 & 3 \\ 0 & -5 & -7 \end{vmatrix} \\
 = -144.$$

This method can, of course, be iterated.

Let $A = [a_{ij}] \in \mathcal{M}_{n \times n}(K)$, where K is an associative and commutative unital F -algebra. For each $1 \leq i, j \leq n$, we define the **minor** of the entry a_{ij} of A to be $|A_{ij}|$, where $A_{ij} \in \mathcal{M}_{(n-1) \times (n-1)}(K)$ is the matrix obtained from A by erasing the i th row and the j th column.

Example: If $A = \begin{bmatrix} 4 & 3 & 1 \\ 2 & 8 & 9 \\ 7 & 3 & 4 \end{bmatrix}$, then $A_{13} = \begin{bmatrix} 2 & 8 \\ 7 & 3 \end{bmatrix}$ and $A_{22} = \begin{bmatrix} 4 & 1 \\ 7 & 4 \end{bmatrix}$.

(11.11) Proposition: Let F be a field, let (K, \bullet) be an associative and commutative unital F -algebra. If n is a positive integer, and $A = [a_{ij}] \in \mathcal{M}_{n \times n}(K)$, then $|A| = \sum_{j=1}^n (-1)^{t+j} a_{tj} \bullet |A_{tj}|$ for each $1 \leq t \leq n$.

Proof: In order to simplify our notation, let $\det(y_1, \dots, y_n)$ denote the determinant of the matrix the rows of which are y_1, \dots, y_n . We will first prove the theorem for the case $t = 1$. That is to say, we must show that $|A|$ equals $\sum_{j=1}^n (-1)^{1+j} a_{1j} \bullet |A_{1j}|$. For each $1 \leq h \leq n$, let $v_h \in \mathcal{M}_{1 \times n}(K)$ be the matrix $\begin{bmatrix} d_1 & \dots & d_n \end{bmatrix}$ defined by

$$d_i = \begin{cases} 1 & \text{if } i = h \\ 0 & \text{otherwise} \end{cases}.$$

Then the i th row of A can be written as $w_i = \sum_{j=1}^n a_{ij}v_j$. and so

$$\begin{aligned} |A| &= \det(w_1, \dots, w_n) = \det \left(\sum_{j=1}^n a_{1j}v_j, w_2, \dots, w_n \right) \\ &= \sum_{j=1}^n a_{1j} \bullet \det(v_j, w_2, \dots, w_n). \end{aligned}$$

Thus we will prove the desired result if we can show that

$$\det(v_j, w_2, \dots, w_n) = (-1)^{1+j} |A_{1j}|$$

for each $1 \leq j \leq n$. Denote the matrix the rows of which are v_j, w_2, \dots, w_n by $B = [b_{ih}]$, where

$$b_{ih} = \begin{cases} 1 & \text{if } i = 1 \text{ and } h = j \\ 0 & \text{if } i = 1 \text{ and } h \neq j \\ a_{ih} & \text{if } i > 1. \end{cases}$$

For $1 \leq j \leq n$, set $G_{1j} = \{\pi \in S_n \mid \pi(1) = j\}$.

Suppose that $j = 1$. Then, in particular, there is a bijective correspondence between G_{11} and the set of all permutations of $\{2, \dots, n\}$ which does not affect the signum of the permutation since if $\pi \in G_{11}$ then 1 does not appear in any inversion of π . Since $b_{11} = 1$ and $b_{1h} = 0$ if $h > 1$, we thus have

$$\begin{aligned} |B| &= \sum_{\pi \in S_n} \text{sgn}(\pi) b_{1, \pi(1)} \bullet \dots \bullet b_{n, \pi(n)} \\ &= \sum_{\pi \in G_{11}} \text{sgn}(\pi) b_{1, \pi(1)} \bullet \dots \bullet b_{n, \pi(n)} \\ &= \sum_{\pi \in G_{11}} \text{sgn}(\pi) b_{2, \pi(2)} \bullet \dots \bullet b_{n, \pi(n)} = |A_{11}| \end{aligned}$$

and so we have shown, as desired, that $|B| = (-1)^{1+1} |A_{11}|$. If $j > 1$ put column j of B in the position of the first column and shift columns 1 to $j-1$ of B to the right by one column position. This involves $j-1$ column interchanges, and we have

$$\det(v_j, w_2, \dots, w_n) = (-1)^{j-1} \begin{vmatrix} 1 & 0 & \dots & 0 \\ a_{2j} & a_{21} & \dots & a_{2n} \\ \vdots & \vdots & & \vdots \\ a_{nj} & a_{n1} & \dots & a_{nn} \end{vmatrix} = (-1)^{j+1} |A_{1j}|.$$

Now assume that $t > 1$. Again, we can interchange the t th row with the first row by $t-1$ exchanges with the row above, and we get $|A| =$

$(-1)^{t-1}|C|$, where C is a matrix satisfying $|C_{1j}| = |A_{tj}|$ for each $1 \leq j \leq n$. Therefore

$$|A| = (-1)^{t-1}|C| = (-1)^{t-1} \sum_{j=1}^n (-1)^{j+1} c_{1j} \bullet |C_{1j}| = \sum_{j=1}^n (-1)^{j+t} a_{tj} \bullet |A_{tj}|$$

as desired. \square

Example: For $A = \begin{bmatrix} 1 & 7 & 3 & 0 \\ 4 & 0 & 1 & 3 \\ 0 & 2 & 4 & 0 \\ 3 & 1 & 5 & 1 \end{bmatrix}$ we see that

$$\begin{aligned} |A| &= 1 \begin{vmatrix} 0 & 1 & 3 \\ 2 & 4 & 0 \\ 1 & 5 & 1 \end{vmatrix} - 7 \begin{vmatrix} 4 & 1 & 3 \\ 0 & 4 & 0 \\ 3 & 5 & 1 \end{vmatrix} + 3 \begin{vmatrix} 4 & 0 & 3 \\ 0 & 2 & 0 \\ 3 & 1 & 1 \end{vmatrix} - 0 \begin{vmatrix} 4 & 0 & 1 \\ 0 & 2 & 4 \\ 3 & 1 & 5 \end{vmatrix} \\ &= 16 + 140 - 30 + 0 = 126 \end{aligned}$$

and

$$\begin{aligned} |A| &= 0 \begin{vmatrix} 7 & 3 & 0 \\ 0 & 1 & 3 \\ 1 & 5 & 1 \end{vmatrix} - 2 \begin{vmatrix} 1 & 3 & 0 \\ 4 & 1 & 3 \\ 3 & 5 & 1 \end{vmatrix} + 4 \begin{vmatrix} 1 & 7 & 0 \\ 4 & 0 & 3 \\ 3 & 1 & 1 \end{vmatrix} - 0 \begin{vmatrix} 1 & 7 & 3 \\ 4 & 0 & 1 \\ 3 & 1 & 5 \end{vmatrix} \\ &= 0 - 2 + 128 - 0 = 126. \end{aligned}$$

Even this method of computing determinants is not easy, however, unless there is a row (or column) of the matrix a significant number of the entries in which are equal to 0. To see the computational overhead of computing the determinant of a general $n \times n$ matrix using minors, let us denote the number of arithmetic operations needed to do so by p_n . Clearly $p_1 = 1$ and $p_2 = 3$. Suppose that we have already found p_{n-1} . Then by Proposition 11.11 we see that in order to compute the determinant of an $n \times n$ matrix we have to compute the determinants of n matrices of size $(n-1) \times (n-1)$ and then perform n multiplications and $n-1$ additions/subtractions. That is to say, we obtain the recursive formula

$$p_n = np_{n-1} + n + (n-1) = np_{n-1} + 2n - 1,$$

when $n > 2$. Setting $t_n = \frac{1}{n!}p_n$, we see that

$$t_n - t_{n-1} = \frac{2}{(n-1)!} - \frac{1}{n!}$$

and so

$$\begin{aligned}
 t_n &= [t_n - t_{n-1}] + [t_{n-1} - t_{n-2}] + \dots + [t_3 - t_2] + t_2 \\
 &= \left[\frac{2}{(n-1)!} - \frac{1}{n!} \right] + \dots + \left[\frac{2}{2!} - \frac{1}{3!} \right] + 1 \\
 &= 2 \left[\frac{1}{(n-1)!} + \dots + \frac{1}{1!} \right] - \left[\frac{1}{n!} + \dots + \frac{1}{1!} \right] + 1 \\
 &= \frac{1}{(n-1)!} + \dots + \frac{1}{1!} + 1 - \frac{1}{n!} \\
 &= \left[\frac{1}{n!} + \frac{1}{(n-1)!} + \dots + \frac{1}{1!} + 1 \right] - \frac{2}{n!}
 \end{aligned}$$

and thus we see that $p_n = n! \left[\frac{1}{n!} + \frac{1}{(n-1)!} + \dots + \frac{1}{1!} + 1 \right] - 2$. But from calculus we know that e , the base of the natural logarithms, has an expansion of the form

$$e = 1 + \frac{1}{1!} + \dots + \frac{1}{n!} + \frac{e^c}{(n+1)!},$$

where $0 < c < 1$, and so

$$p_n = n! \left[e - \frac{e^c}{(n+1)!} \right] - 2.$$

If $n > 2$, we see that

$$0 < \frac{e^c}{n+1} < \frac{e}{n+1} \leq \frac{e}{3} < 1$$

and so we conclude that $en! - 3 < p_n < en! - 2$. Since p_n is a positive integer, we see that $p_n = \lfloor en! \rfloor - 2$, where $\lfloor r \rfloor$ denotes the largest whole number less than or equal to r , for any real number r . In particular, we see that p_n grows even faster than exponentially, as a function of n , which is very rapid growth indeed. For example, $p_{10} = 9,864,094$ and $p_{15} = 3,554,625,081,047$.

In special cases, it is also possible to find bounds on the value of the determinant of a matrix, without actually computing it. For example, we will see below that if $A \in \mathcal{M}_{n \times n}(\mathbb{R})$ and if g is a positive real number greater than or equal to the absolute value of each of the entries of A , then the absolute value of $|A|$ is at most $g^n \sqrt{n^n}$.

(11.12) Proposition: Let n be a positive integer, let F be a field, and let (K, \bullet) be an associative and commutative unital F -algebra. Let $A = [a_{ij}] \in \mathcal{M}_{n \times n}(K)$ be a matrix which can be

represented in block form as
$$\begin{bmatrix} A_1 & O & \dots & O \\ O & A_2 & \dots & O \\ & & \dots & \\ O & O & \dots & A_m \end{bmatrix}, \text{ where } m > 1$$

and each of the submatrices A_i is square. Then $|A| = \prod_{h=1}^m |A_h|$.

Proof: Let us first consider the case $m = 2$, and assume that $A_1 \in \mathcal{M}_{t \times t}(K)$ for some $t < n$. By definition,

$$|A| = \sum_{\pi \in S_n} \text{sgn}(\pi) a_{\pi(1),1} \bullet a_{\pi(2),2} \bullet \dots \bullet a_{\pi(n),n}.$$

However, if $\pi \in S_n$ satisfies the condition that $\pi(i) > t$ for some $1 \leq i \leq t$, then the summand $\text{sgn}(\pi) a_{\pi(1),1} \bullet a_{\pi(2),2} \bullet \dots \bullet a_{\pi(n),n}$ equals 0. Therefore, we in fact have

$$|A| = \sum_{\pi \in U} \text{sgn}(\pi) a_{\pi(1),1} \bullet a_{\pi(2),2} \bullet \dots \bullet a_{\pi(n),n},$$

where U is the subset of S_n consisting of all those permutations π satisfying $1 \leq \pi(i) \leq t$ for all $1 \leq i \leq t$ and hence, perforce, $t+1 \leq \pi(i) \leq n$ for all $t+1 \leq i \leq n$. In other words, each $\pi \in U$ can be considered as the combination of two permutations, one of $\{1, \dots, t\}$, and the other of $\{t+1, \dots, n\}$. Moreover, again as a direct consequence of the definition, $\text{sgn}(\pi)$ is the product of the signa of these two permutations. From this, our result follows immediately.

Now assume, inductively, that the result has been established for m and consider a matrix $A \in \mathcal{M}_{n \times n}(K)$ which can be written in block form

$$\text{as } \begin{bmatrix} A_1 & O & \dots & O \\ O & A_2 & \dots & O \\ & & \dots & \\ O & O & \dots & A_{m+1} \end{bmatrix}. \text{ If we set } B = \begin{bmatrix} A_1 & O & \dots & O \\ O & A_2 & \dots & O \\ & & \dots & \\ O & O & \dots & A_m \end{bmatrix}$$

then, by the case $m = 2$ and the induction hypothesis, we see that

$$|A| = |B| \bullet |A_{m+1}| = \prod_{h=1}^{m+1} |A_h|. \quad \square$$

Let $A = [a_{ij}] \in \mathcal{M}_{n \times n}(K)$ for some associative and commutative unital F -algebra (K, \bullet) . We define the **adjoint** of A to be the matrix $\text{adj}(A) = [b_{ij}] \in \mathcal{M}_{n \times n}(K)$, where $b_{ij} = (-1)^{i+j} |A_{ji}|$ for all $1 \leq i, j \leq n$.

Example: If $A = \begin{bmatrix} 1 & 0 & 3 & 5 \\ -3 & 1 & 3 & 1 \\ 4 & 2 & 1 & 2 \\ 1 & 1 & 2 & 5 \end{bmatrix} \in \mathcal{M}_{4 \times 4}(\mathbb{R})$ then

$$\text{adj}(A) = \begin{bmatrix} -20 & 9 & -17 & 25 \\ 50 & -18 & -16 & -40 \\ -40 & -18 & -16 & 50 \\ 10 & 9 & 13 & -35 \end{bmatrix}.$$

(11.13) Proposition: Let F be a field and let n be a positive integer. If $A = [a_{ij}] \in M_{n \times n}(F)$ then $A[\text{adj}(A)] = |A|I$. In particular, if the matrix A is nonsingular then $A^{-1} = |A|^{-1} \text{adj}(A)$.

Proof: Suppose that $\text{adj}(A) = [b_{ij}]$. Then $A[\text{adj}(A)] = [c_{ij}]$, where $c_{ij} = \sum_{k=1}^n a_{ik}b_{kj} = \sum_{k=1}^n (-1)^{j+k} a_{ik}|A_{jk}|$. If $i = j$, then, by Proposition 11.11, this is just $|A|$. If $i \neq j$, this is just $|A'|$, where A' is a matrix identical to A in all of its rows except the i th row, and that is equal to the j th row of A . Thus the matrix A' has two identical rows, and so by Proposition 11.3, $|A'|$ is equal to 0. Hence $A[\text{adj}(A)] = |A|I$, from which we also immediately deduce the second statement since if A is nonsingular then $|A| \neq 0$. \square

(11.14) Proposition: Let F be a field, let (K, \bullet) be an associative and commutative unital F -algebra, and let n be a positive integer. If $A = [a_{ij}] \in M_{n \times n}(K)$ is an upper-triangular matrix then $|A| = \prod_{i=1}^n a_{ii}$.

Proof: We can prove this by induction on n . For the case $n = 1$, it is immediate. Assume therefore that we have already established it for all matrices in $\mathcal{M}_{n \times n}(K)$. Then, by Proposition 11.11, $|A| = |A^T| = \sum_{j=1}^n (-1)^{1+j} a_{j1} \bullet |A_{j1}| = a_{11} \bullet |A_{11}|$. But, by the induction hypothesis, $|A_{11}| = \prod_{i=2}^n a_{ii}$, and we are done. \square

By Proposition 11.14 we see that in general, from a computational point of view, it is much faster to first perform elementary operations on a matrix to reduce it to upper-triangular form, and then calculate the determinant (making use of the fact, from the definition of a determinant function and from Proposition 11.4, we easily know the determinants of the elementary matrices), than to calculate the determinant directly. When working in associative and commutative unital algebras over a field, or when working with matrices of integers, this presents somewhat of a problem since it is not always possible to divide by nonzero scalars in such contexts. However, various variants on Gaussian elimination which do not involve division have been developed to overcome this.

Combining Propositions 11.4 and 11.14, we see that if $A \in M_{n \times n}(F)$ can be written in the form LU , where L is a lower-triangular matrix and U is an upper-triangular matrix, then $|A|$ is the product of the diagonal elements of L and the diagonal elements of U .

(11.15) Proposition (Cramer's Theorem⁵): Let F be a field and let n be a positive integer. If $A = [a_{ij}] \in M_{n \times n}(F)$ is a

nonsingular matrix and if $w = \begin{bmatrix} b_1 \\ \vdots \\ b_n \end{bmatrix} \in F^n$, then the system of

linear equations $AX = w$ has the unique solution $v = \begin{bmatrix} d_1 \\ \vdots \\ d_n \end{bmatrix}$ in

which, for each $1 \leq i \leq n$, we have $d_i = |A|^{-1}|A_{(i)}|$, where $A_{(i)}$ is the matrix formed from A by replacing the i th column of A by w .

Proof: If $Av = w$ then $|A|v = (|A|A^{-1})Av = \text{adj}(A)Av = \text{adj}(A)w$ and so for each $1 \leq i \leq n$, we have $|A|d_i = \sum_{j=1}^n (-1)^{i+j} b_j |A_{ji}|$. But the expression on the right-hand side of this equation is just, by Proposition 11.11, $|A_{(i)}|$, developed by minors on the i th column. \square

Cramer's theorem, published in 1750, was the first systematic method for solving a system of linear equations, though special cases of it were known to Leibnitz 75 years earlier. While it is elegant mathematically, it is clearly not computationally feasible, even when n is only moderately large, as was immediately realized by mathematicians of the time. Indeed, solving a system of linear equations $AX = w$ by Cramer's method, where A is a nonsingular $n \times n$ matrix over a field F , requires $\frac{1}{3}n^4 - \frac{1}{6}n^3 - \frac{1}{3}n^2 + \frac{1}{6}n$ additions and $\frac{1}{3}n^4 + \frac{1}{3}n^3 + \frac{2}{3}n^2 + \frac{2}{3}n - 1$ multiplications, which is considerably worse than the methods we have previously studied, for which the number of arithmetic operations necessary grows as n^3 , rather than as n^4 .



5

Gabriel Cramer was an 18th-century Swiss mathematician and friend of Johann Bernoulli (one of the formulators of calculus) who was among the first to study determinants and their use to solve systems of linear equations.

Example: Consider the system of linear equations $AX = \begin{bmatrix} 2 \\ 1 \\ 4 \end{bmatrix}$,

where $A = \begin{bmatrix} 1 & -1 & 1 \\ 1 & 2 & 0 \\ 1 & 0 & -1 \end{bmatrix}$. Then $|A| = -5$ and

$$|A_{(1)}| = \begin{vmatrix} 2 & -1 & 1 \\ 1 & 2 & 0 \\ 4 & 0 & -1 \end{vmatrix} = -13, \quad |A_{(2)}| = \begin{vmatrix} 1 & 2 & 1 \\ 1 & 1 & 0 \\ 1 & 4 & -1 \end{vmatrix} = 4, \quad \text{and}$$

$$|A_{(3)}| = \begin{vmatrix} 1 & -1 & 2 \\ 1 & 2 & 1 \\ 1 & 0 & 4 \end{vmatrix} = 7.$$

As a consequence, we see that the unique solution to the equation is

$$\frac{1}{5} \begin{bmatrix} 13 \\ -4 \\ -7 \end{bmatrix}.$$

We end this chapter with an interesting application of determinants of matrices over the $\mathbb{R}[X]$ -algebra $\mathbb{R}[X]$. Let c_0, c_1, \dots be real numbers and let us consider the analytic function $f : x \mapsto \sum_{i=0}^{\infty} c_i x^i$, which converges for all x in some subset U of \mathbb{R} . We know that $U \neq \emptyset$ since surely $0 \in U$. Given positive integers k and n , we want to find polynomials $p(X), q(X) \in \mathbb{R}[X]$ of degrees at most k and n respectively, such that the function $x \mapsto p(x)q(x)^{-1} - f(x)$ also converges for all $x \in U$ and is representable there by a power series of the form $x \mapsto \sum_{i=1}^{\infty} d_i x^{k+n+i}$. If we find such p and q , then the function $x \mapsto p(x)q(x)^{-1}$ is called the **Padé approximant**⁶ to f of type k/n . Padé approximants are very important tools in differential equations and in approximation theory.

Example: If $f : x \mapsto e^x = \sum_{i=0}^{\infty} \frac{1}{i!} x^i$ then the function

$$g_1 : x \mapsto \frac{x^2 + 4x + 6}{6 - 2x}$$



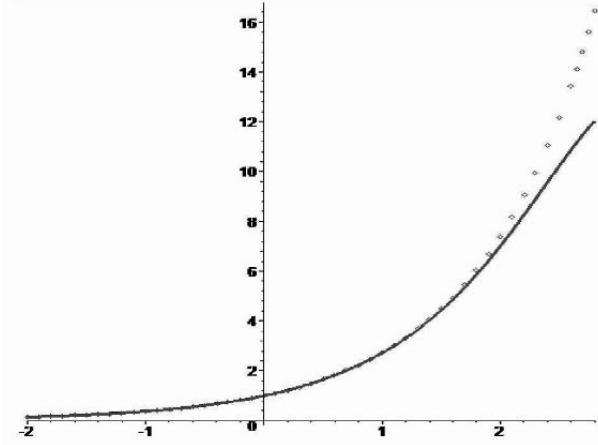
6

Henri Padé was a 19th-century French engineer who developed these approximants in the course of his work. Interest in them intensified in the early 20th century when the French mathematician **Émile Borel** made extensive use of them in his work on analysis.

is a Padé approximant to f of type $2/1$ and the function

$$g_2 : x \mapsto \frac{x^2 + 6x + 12}{x^2 - 6x + 12}$$

is a Padé approximant to f of type $2/2$. The following diagram shows a part of the graph of f (broken line) and of g_2 (solid line):



If we are given f as above, how do we calculate Padé approximants to it? First of all, define $c_{-i} = 0$ for all positive integers i . Then, given positive integers k and n , define the matrices $P_{k/n}(X), Q_{k/n}(X) \in \mathcal{M}_{(n+1) \times (n+1)}(\mathbb{R}[X])$ as follows:

$$P_{k/n}(X) = \begin{bmatrix} c_{k-n+1} & c_{k-n+2} & \cdots & c_{k+1} \\ c_{k-n+2} & c_{k-n+3} & \cdots & c_{k+2} \\ & & \cdots & \\ & c_k & c_{k+1} & \cdots & c_{k+n} \\ \sum_{i=0}^{k-n} c_i X^{n+i} & \sum_{i=0}^{k-n+1} c_i X^{n+i-1} & \cdots & \sum_{i=0}^k c_i X^i \end{bmatrix}$$

and

$$Q_{k/n}(X) = \begin{bmatrix} c_{k-n+1} & c_{k-n+2} & \cdots & c_{k+1} \\ c_{k-n+2} & c_{k-n+3} & \cdots & c_{k+2} \\ & & \cdots & \\ & c_k & c_{k+1} & \cdots & c_{k+n} \\ X^n & X^{n-1} & \cdots & 1 \end{bmatrix}.$$

Then the polynomials $p(X) = |P_{k/n}(X)|$ and $q(X) = |Q_{k/n}(X)|$ are of the desired size, and our approximant is given by $x \mapsto p(x)q(x)^{-1}$.

Exercises

Exercise 580 For real numbers a and b , calculate

$$\begin{vmatrix} \sin(a) & \cos(a) \\ \sin(b) & \cos(b) \end{vmatrix} \quad \text{and} \quad \begin{vmatrix} \cos(a) & \sin(a) \\ \sin(b) & \cos(b) \end{vmatrix}.$$

Exercise 581 Calculate $\begin{vmatrix} 1 & i & 1+i \\ -i & 1 & 0 \\ 1-i & 0 & 1 \end{vmatrix} \in \mathbb{C}^3$.

Exercise 582 For any real number a , calculate

$$\begin{vmatrix} a-6 & 0 & 0 & -8 \\ 5 & a-4 & 0 & 12 \\ -1 & 3 & a-2 & -6 \\ 0 & -\frac{1}{2} & 1 & 1 \end{vmatrix}.$$

Exercise 583 Find the image of the function f from \mathbb{R} to itself defined

$$\text{by } f : t \mapsto \begin{vmatrix} 1 & 0 & -t \\ 1 & 1 & -1 \\ t & 0 & -1 \end{vmatrix}.$$

Exercise 584 For real numbers a , b , c , and d , show that

$$\begin{vmatrix} a^2 & (a+1)^2 & (a+2)^2 & (a+3)^2 \\ b^2 & (b+1)^2 & (b+2)^2 & (b+3)^2 \\ c^2 & (c+1)^2 & (c+2)^2 & (c+3)^2 \\ d^2 & (d+1)^2 & (d+2)^2 & (d+3)^2 \end{vmatrix} = 0.$$

Exercise 585 Let n be a positive integer and let c be a fixed real number. Let $A = [a_{ij}] \in \mathcal{M}_{n \times n}(\mathbb{R})$ be the matrix defined by

$$a_{ij} = \begin{cases} c & \text{if } i < j \\ i & \text{if } i = j \\ 0 & \text{if } i > j \end{cases}.$$

Calculate $|A|$.

Exercise 586 If n is a positive integer, we define the n th **Hankel matrix** $H_n \in \mathcal{M}_{n \times n}(\mathbb{R})$ to be the matrix $[a_{ij}]$ satisfying

$$a_{ij} = \begin{cases} 0 & \text{if } i + j - 1 > n \\ i + j - 1 & \text{otherwise} \end{cases}.$$

Calculate $|H_n|$.

Exercise 587 For $a, b \in \mathbb{R}$, calculate $\begin{vmatrix} a & b & a+b \\ b & a+b & a \\ a+b & a & b \end{vmatrix}$.

Exercise 588 Let $p(X) = a_0 + a_1X + a_2X^2$ and $q(X) = b_0 + b_1X + b_2X^2$ be polynomials in $\mathbb{C}[X]$. Show that there exists a complex number c

satisfying $p(c) = q(c) = 0$ if and only if $\begin{vmatrix} a_0 & a_1 & a_2 & 0 \\ 0 & a_0 & a_1 & a_2 \\ b_0 & b_1 & b_2 & 0 \\ 0 & b_0 & b_1 & b_2 \end{vmatrix} = 0$.

Exercise 589 Find the set of all pairs (a, b) of real numbers such that

$$\begin{vmatrix} a+1 & 3a & b+3a & b+1 \\ 2b & b+1 & 2-b & 1 \\ a+2 & 0 & 1 & a+3 \\ b-1 & 1 & a+2 & a+b \end{vmatrix} = 0.$$

Exercise 590 For $a, b, c \in \mathbb{R}$, show that

$$\begin{vmatrix} 0 & (a-b)^2 & (a-c)^2 \\ (b-a)^2 & 0 & (b-c)^2 \\ (c-a)^2 & (c-b)^2 & 0 \end{vmatrix} \leq 0.$$

Exercise 591 Let n be a positive even integer and let $c, d \in \mathbb{Q}$. Let $A = [a_{ij}] \in \mathcal{M}_{n \times n}(\mathbb{Q})$ be the matrix defined by

$$a_{ij} = \begin{cases} c & \text{if } i = j \\ d & \text{if } i + j = n + 1 \\ 0 & \text{otherwise} \end{cases}.$$

Calculate $|A|$.



7 Nineteenth-century German mathematician **Hermann Hankel** was among the first to recognize and popularize the work of Grassmann.

Exercise 592 Let n be a positive integer and let $A = [a_{ij}] \in \mathcal{M}_{n \times n}(\mathbb{Q})$ be the tridiagonal matrix defined by

$$a_{ij} = \begin{cases} 1 & \text{if } |i - j| \leq 1 \\ 0 & \text{otherwise} \end{cases}.$$

Show that

$$|A| = \begin{cases} -1 & \text{if } n = 3k \\ 1 & \text{if } n = 3k + 1 \\ 0 & \text{if } n = 3k + 2 \end{cases}$$

for some nonnegative integer k .

Exercise 593 Find integers a , b , and c for which

$$\begin{vmatrix} a+b & c & c \\ a & b+c & a \\ b & b & a+c \end{vmatrix}$$

is divisible by 8.

Exercise 594 Let n be a positive integer and let $A \in \mathcal{M}_{n \times n}(\mathbb{Q})$ be a nonsingular matrix satisfying the condition that all of the entries of A and of A^{-1} are integers. Show that $|A| = \pm 1$.

Exercise 595 For elements a , b , and c of a field F , calculate

$$\begin{vmatrix} -2a & a+b & a+c \\ a+b & -2b & b+c \\ a+c & b+c & -2c \end{vmatrix}.$$

Exercise 596 We know that the integers 23028, 31882, 86469, 6327,

and 61902 are all divisible by 19. Show that

$$\begin{vmatrix} 2 & 3 & 0 & 2 & 8 \\ 3 & 1 & 8 & 8 & 2 \\ 8 & 6 & 4 & 6 & 9 \\ 0 & 6 & 3 & 2 & 7 \\ 6 & 1 & 9 & 0 & 2 \end{vmatrix} \text{ is also}$$

divisible by 19.

Exercise 597 Let $q \in \mathbb{Q}$. Show that there are infinitely-many matrices in $\mathcal{M}_{3 \times 3}(\mathbb{Q})$ of the form $\begin{bmatrix} 2 & 2 & 3 \\ 3q+2 & 4q+2 & 5q+3 \\ a & b & c \end{bmatrix}$, where $a < b < c$, the determinant of which equals q .

Exercise 598 Let n be a positive integer and let F be a field. Let $A \in \mathcal{M}_{n \times n}(F)$ be a nonsingular matrix which can be written in block form

$$A = \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix}, \text{ where } A_{11} \in \mathcal{M}_{k \times k}(F) \text{ for some integer } k < n.$$

Write A^{-1} as $\begin{bmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{bmatrix}$, where $B_{11} \in \mathcal{M}_{k \times k}(F)$. Show that $|A_{11}| = |A| \cdot |B_{22}|$.

Exercise 599 Find all real numbers a for which $\begin{vmatrix} a & 1 & 1 & 1 \\ 1 & a & 2 & 3 \\ 0 & -1 & 0 & 1 \\ -1 & 1 & 1 & 2 \end{vmatrix} = 0$.

Exercise 600 Let a_1, a_2, \dots be a sequence of real numbers. For each positive integer n , define the **n th continuant** c_n of the sequence to be the determinant of the tridiagonal matrix $A_n = [a_{ij}] \in \mathcal{M}_{n \times n}(\mathbb{R})$ given by

$$a_{ij} = \begin{cases} a_i & \text{if } i = j \\ -1 & \text{if } i = j - 1 \\ 1 & \text{if } i = j + 1 \\ 0 & \text{otherwise} \end{cases}.$$

Show that $c_n = a_n c_{n-1} + c_{n-2}$ for all $n > 2$.

Exercise 601 Let $n > 1$ be an integer, let d be a real number, and let $A = [a_{ij}] \in \mathcal{M}_{n \times n}(\mathbb{R})$ be the matrix defined as follows:

$$a_{ij} = \begin{cases} 0 & \text{if } i = j \\ 1 & \text{if } i > 1 \text{ and } j = 1 \text{ or } i = 1 \text{ and } j > 1 \\ d & \text{otherwise} \end{cases}.$$

Show that $|A| = (-1)^{n-1}(n-1)d^{n-2}$.

Exercise 602 Let b_1, \dots, b_n be nonzero real numbers and let $A = [a_{ij}] \in \mathcal{M}_{n \times n}(\mathbb{R})$ be the matrix defined as follows:

$$a_{ij} = \begin{cases} 1 + b_j & \text{if } i = j \\ 1 & \text{otherwise} \end{cases}.$$

Calculate $|A|$.

Exercise 603 Let $A = [a_{ij}] \in \mathcal{M}_{4 \times 4}(\mathbb{Q})$ be a matrix each entry of which is either -2 or 3 . Show that $|A|$ is an integer multiple of 125 .

Exercise 604 Let a, b, c , and d be real numbers not all of which are

equal to 0 . Show that the matrix $\begin{bmatrix} a & b & c & d \\ b & -a & d & -c \\ c & -d & -a & b \\ d & c & -b & -a \end{bmatrix} \in \mathcal{M}_{4 \times 4}(\mathbb{R})$ is nonsingular.

Exercise 605 Does there exist a rational number a satisfying the condition that the matrix

$$\begin{bmatrix} 1 & a & 0 \\ a & 1 & 1 \\ -1 & a & -1 \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{Q}) \text{ is nonsingular?}$$

Exercise 606 Find all matrices $I \neq A \in \mathcal{M}_{2 \times 2}(\mathbb{R})$ satisfying $A^3 = I$.

Exercise 607 Find all triples (a, b, c) of real numbers satisfying the condition

$$\begin{vmatrix} 1 & a & a^3 \\ 1 & b & b^3 \\ 1 & c & c^3 \end{vmatrix} = (b-c)(c-a)(a-b)(a+b+c).$$

Exercise 608 Let n be a positive integer, let $A = [a_{ij}] \in \mathcal{M}_{n \times n}(\mathbb{C})$ and let $B = [b_{ij}] \in \mathcal{M}_{n \times n}(\mathbb{C})$ be defined by $b_{ij} = \bar{a}_{ji}$ for each $1 \leq i, j \leq n$. Show that $|AB|$ is a nonnegative real number.

Exercise 609 Calculate $\begin{vmatrix} 1 & \log_b a \\ \log_a b & 1 \end{vmatrix}$ for given positive real numbers a and b .

Exercise 610 Let F be a field. Calculate $\begin{vmatrix} 1 & a & a^2 & a^3 \\ a^3 & a^2 & a & 1 \\ 1 & 2a & 3a^2 & 4a^3 \\ 4a^3 & 3a^2 & 2a & 1 \end{vmatrix}$ for any $a \in F$.

Exercise 611 Calculate $\begin{vmatrix} \cos(a) & \sin(a) & \cos(a) & \sin(a) \\ \cos(2a) & \sin(2a) & 2\cos(2a) & 2\sin(2a) \\ \cos(3a) & \sin(3a) & 3\cos(3a) & 3\sin(3a) \\ \cos(4a) & \sin(4a) & 4\cos(4a) & 4\sin(4a) \end{vmatrix}$ for $a \in \mathbb{R}$.

Exercise 612 Let n be a positive integer and let $A = [a_{ij}] \in \mathcal{M}_{n \times n}(\mathbb{R})$ be the matrix defined by

$$a_{ij} = \begin{cases} 0 & \text{if } i = j \\ 1 & \text{otherwise} \end{cases}.$$

Calculate $|A|$.

Exercise 613 Let $A = \begin{bmatrix} 3 & -1 & 1 \\ 0 & 2 & 4 \\ 1 & -1 & 1 \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{R})$. Calculate $\text{adj}(A)$.

Exercise 614 Let $F = GF(2)$ and let $A = \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix} \in \mathcal{M}_{3 \times 3}(F)$. Calculate $\text{adj}(A)$.

Exercise 615 Let F be a field, let n be a positive integer, and let $A, B \in \mathcal{M}_{n \times n}(F)$. Is it necessarily true that $\text{adj}(AB) = \text{adj}(A)\text{adj}(B)$?

Exercise 616 Let F be a field, let n be a positive integer, and let $A \in \mathcal{M}_{n \times n}(F)$. Is it necessarily true that $\text{adj}(A^T) = \text{adj}(A)^T$?

Exercise 617 Let F be a field, let n be a positive integer, and let the matrices $A, B \in \mathcal{M}_{n \times n}(F)$ be nonsingular. Show that $\text{adj}(B^{-1}AB) = B^{-1}\text{adj}(A)B$.

Exercise 618 Let F be a field, let n be a positive integer, and let $A, B \in \mathcal{M}_{n \times n}(F)$ be matrices satisfying $B \neq O$ and $AB = O$. Show that $|A| = 0$.

Exercise 619 Let $A = \begin{bmatrix} 1 & 2 & 3 \\ 1 & 3 & 4 \\ 1 & 4 & 3 \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{R})$. Use the adjoint of A to calculate A^{-1} .

Exercise 620 Let F be a field, let n be a positive integer, and let $A = [a_{ij}] \in \mathcal{M}_{n \times n}(F)$. Let $B = [b_{ij}] \in \mathcal{M}_{n \times n}(F)$ defined by $b_{ij} = (-1)^{i+j}a_{ij}$ for all $1 \leq i, j \leq n$. Show that $|A| = |B|$.

Exercise 621 Let F be a field, let n be a positive integer, and let $A = [a_{ij}] \in \mathcal{M}_{n \times n}(F)$. Let $B = [b_{ij}] \in \mathcal{M}_{n \times n}(F)$ defined by $b_{ij} = (-1)^{i+j+1}a_{ij}$ for all $1 \leq i, j \leq n$. Show that $(-1)^n|A| = |B|$.

Exercise 622 Let n be a positive integer and let $\pi \in S_n$. Let $A \in \mathcal{M}_{n \times n}(\mathbb{Q})$ be the permutation matrix defined by π . Calculate $|A|$.

Exercise 623 Is the set of all permutation matrices in $\mathcal{M}_{n \times n}(\mathbb{Q})$ closed under multiplication? Is the inverse of a permutation matrix a permutation matrix?

Exercise 624 Let $A = [a_{ij}] \in \mathcal{M}_{3 \times 3}(\mathbb{R})$ be a matrix in which $a_{i2} \neq 0$ for all $1 \leq i \leq 3$. Denote the minor of a_{ij} for all $1 \leq i, j \leq n$ by A_{ij} . Show that

$$|A| = \frac{1}{a_{12}} \begin{vmatrix} A_{21} & A_{23} \\ A_{31} & A_{32} \end{vmatrix} + \frac{1}{a_{22}} \begin{vmatrix} A_{11} & A_{13} \\ A_{31} & A_{33} \end{vmatrix} + \frac{1}{a_{32}} \begin{vmatrix} A_{11} & A_{13} \\ A_{21} & A_{22} \end{vmatrix}.$$

Exercise 625 Let F be a field, let n be a positive integer, and let $A = [a_{ij}] \in \mathcal{M}_{n \times n}(F)$ be nonsingular. Show that $\text{adj}(\text{adj}(A)) = |A|^{n-2}A$.

Exercise 626 Let a and b be real numbers and let n be an integer greater than 2. Let $D = [d_{ij}] \in \mathcal{M}_{n \times n}(\mathbb{R})$ be the matrix defined by $d_{ij} = \sin(ia + jb)$ for all $1 \leq i, j \leq n$. Show that $|D| = 0$.

Exercise 627 Let F be a field and let $a, b, c, d, e, f, g \in F$. Show that

$$\begin{vmatrix} a & b & b \\ c & d & e \\ f & g & g \end{vmatrix} + \begin{vmatrix} a & b & b \\ e & c & d \\ f & g & g \end{vmatrix} + \begin{vmatrix} a & b & b \\ d & e & c \\ f & g & g \end{vmatrix} = 0.$$

Exercise 628 Let F be a field, let n be a positive integer, and let $A = [a_{ij}] \in \mathcal{M}_{n \times n}(F)$. Let $B = [b_{ij}] \in \mathcal{M}_{n \times n}(F)$ be the matrix defined by

$$b_{ij} = \begin{cases} a_{ij} + a_{i,j+1} & \text{if } j < n \\ a_{in} & \text{otherwise} \end{cases}.$$

Show that $|B| = |A|$.

Exercise 629 Let k and n be integers greater than 1. Let F be a field and let $A = [a_{ij}]$ be a matrix in $\mathcal{M}_{k \times n}(F)$, the upper row of which contains at least one nonzero entry. For each $2 \leq i \leq k$ and each

$2 \leq j \leq n$, let $d_{ij} = \begin{vmatrix} a_{11} & a_{1j} \\ a_{i1} & a_{ij} \end{vmatrix}$. Show that the rank of the matrix

$$D = \begin{bmatrix} d_{22} & \dots & d_{2n} \\ \vdots & & \vdots \\ d_{k2} & \dots & d_{kn} \end{bmatrix} \in \mathcal{M}_{(k-1) \times (n-1)}(F)$$

is $r - 1$, where r is the rank of A .

Exercise 630 Let F be a field, let $a \neq b$ be elements of F , and let $A, B \in \mathcal{M}_{2 \times 2}(F)$ be matrices satisfying the condition that $|A + hB| \in \{a, b\}$ for $h = 1, 2, 3, 4, 5$. Show that $|A + 9B| \in \{a, b\}$.

Exercise 631 Let F be a field and let $a, b, c \in F$. Make use of the

matrix $\begin{bmatrix} b & c & 0 \\ a & 0 & c \\ 0 & a & b \end{bmatrix}$ in order to calculate the determinant of the matrix

$$\begin{bmatrix} b^2 + c^2 & ab & ac \\ ab & a^2 + c^2 & bc \\ ac & bc & a^2 + b^2 \end{bmatrix}.$$

Exercise 632 Let $A \in \mathcal{M}_{n \times n}(\mathbb{C})$ be a nonsingular matrix, which we will write in the form $B + iC$, where $B, C \in \mathcal{M}_{n \times n}(\mathbb{R})$. Show that there is a real number d such that the matrix $B + dC \in \mathcal{M}_{n \times n}(\mathbb{R})$ is nonsingular.

Exercise 633 Let F be a field and let n be a positive integer. Let $A \in \mathcal{M}_{n \times n}(F)$ be a matrix having the property that the sum of all even-numbered columns (considered as vectors in F^n) of A equals the sum of all odd-numbered columns of A . What is $|A|$?

Exercise 634 Let $V = \mathbb{R}^2$ and let $f : V^3 \rightarrow \mathbb{R}$ be the function defined as

$$\text{follows: if } v_i = \begin{bmatrix} a_i \\ b_i \end{bmatrix} \text{ for } i = 1, 2, 3, \text{ then } f : \begin{bmatrix} v_1 \\ v_2 \\ v_3 \end{bmatrix} \mapsto \begin{vmatrix} a_1 & b_1 & 1 \\ a_2 & b_2 & 1 \\ a_3 & b_3 & 1 \end{vmatrix}.$$

Show that $f(v_1, v_2, v_3) = f(v_4, v_2, v_3) + f(v_1, v_4, v_3) + f(v_1, v_2, v_4)$ for all $v_1, v_2, v_3, v_4 \in V$.

Exercise 635 Let F be a field and let n be a positive integer. Let $D = [d_{ij}] \in \mathcal{M}_{n \times n}(F)$ be the matrix defined by $d_{ij} = 1$ for all $1 \leq i, j \leq n$. Show that for any matrix $A \in \mathcal{M}_{n \times n}(F)$ precisely one of the following conditions holds: (1) There is a unique scalar $a \in F$ such that $A + aD$ is singular; (2) $A + aD$ is singular for all scalars $a \in F$; (3) $A + aD$ is nonsingular for all scalars $a \in F$.

Exercise 636 Let $A, B, C, D \in \mathcal{M}_{2 \times 2}(\mathbb{R})$ and let M be the matrix

$\begin{bmatrix} A & B \\ C & D \end{bmatrix} \in \mathcal{M}_{4 \times 4}(\mathbb{R})$. If all of the “formal determinants” $AD - BC$, $AD - CB$, $DA - BC$, and $DA - CB$ are nonsingular, is M necessarily nonsingular?

Exercise 637 Let $A, B, C, D \in \mathcal{M}_{2 \times 2}(\mathbb{R})$ and let M be the matrix

$\begin{bmatrix} A & B \\ C & D \end{bmatrix} \in \mathcal{M}_{4 \times 4}(\mathbb{R})$. If M is a nonsingular matrix, is at least one of the “formal determinants” $AD - BC$, $AD - CB$, $DA - BC$, and $DA - CB$ also nonsingular?

Exercise 638 Let $n > 1$ be an integer and let $A = [a_{ij}] \in \mathcal{M}_{n \times n}(\mathbb{Q})$ be a matrix satisfying the condition that each a_{ij} is either equal to 1 or to -1 . Show that $|A|$ is an integer multiple of 2^{n-1} .

Exercise 639 If a, b, c, d, e, f are nonzero elements of a field F , show

$$\text{that } \begin{vmatrix} 0 & a^2 & b^2 & c^2 \\ a^2 & 0 & f^2 & e^2 \\ b^2 & f^2 & 0 & d^2 \\ c^2 & e^2 & d^2 & 0 \end{vmatrix} = \begin{vmatrix} 0 & ad & be & cf \\ ad & 0 & cf & be \\ be & cf & 0 & ad \\ cf & be & ad & 0 \end{vmatrix}.$$

Exercise 640 Let n be a positive integer and let c_1, \dots, c_n be distinct real numbers transcendental over \mathbb{Q} . For $1 \leq h \leq n$, let $p_h(X) = \sum_{i=0}^{h-1} a_i X^i \in \mathbb{Q}[X]$ be a polynomial of degree $h-1$. Let $A = [p_i(c_j)] \in \mathcal{M}_{n \times n}(\mathbb{R})$. Show that $|A| = (a_0 \cdots a_{n-1}) \prod_{i < j} (c_j - c_i)$.

Exercise 641 Let n be a positive integer and let c_1, \dots, c_n be distinct real numbers transcendental over \mathbb{Q} . For each $1 \leq i, j \leq n$, set $d_{ij} =$

$c_i^j - c_i^{-j}$ and let $A = [d_{ij}] \in \mathcal{M}_{n \times n}(\mathbb{R})$. Show that

$$|A| = (c_1 \cdot \dots \cdot c_n)^{-n} \prod_{i < j} [(c_i - c_j)(1 - c_i c_j)] \prod_{i=1}^n (c_i^2 - 1).$$

Exercise 642 Let a, b, c, d be elements of a field F . Solve the equation

$$\begin{vmatrix} a & b & c & d \\ b & c & d & a \\ c & d & a & b \\ d & a & b & c \end{vmatrix} = X \begin{vmatrix} 0 & 1 & -1 & 1 \\ 1 & c & d & a \\ 1 & d & a & b \\ 1 & a & b & c \end{vmatrix}.$$

Exercise 643 Let a, b , and c be nonzero real numbers. Under which

conditions does the equation $\begin{vmatrix} 0 & a - X & b - x \\ -a - X & 0 & c - X \\ -b - X & -c - X & 0 \end{vmatrix} = 0$ have more than one solution?

Exercise 644 Use determinants to show that there is no matrix $A \in$

$$\mathcal{M}_{4 \times 4}(\mathbb{Q}) \text{ satisfying the condition that } A^4 = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 2 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}.$$

Exercise 645 Let $A = [a_{ij}] \in \mathcal{M}_{n \times n}(\mathbb{R})$ be a matrix satisfying the condition $|a_{ii}| > \sum_{j \neq i} |a_{ij}|$ for all $1 \leq i \leq n$. Show that $|A| \neq 0$.

Exercise 646 Let F be a field and let $a, b, c \in F$. Is it true that

$$\begin{vmatrix} a & b & c & 0 \\ b & a & 0 & c \\ c & 0 & a & b \\ 0 & c & b & a \end{vmatrix} = \begin{vmatrix} -a & b & c & 0 \\ b & -a & 0 & c \\ c & 0 & -a & b \\ 0 & c & b & -a \end{vmatrix}?$$

Exercise 647 Let F be a field and let $n > 2$ be an integer. Give an example of a matrix $A \in \mathcal{M}_{n \times n}(F)$ all of the entries in which are nonzero, satisfying $\text{adj}(A) = O$.

Exercise 648 Let F be a field and let $n > 2$ be an integer. Show that $|\text{adj}(A)| = |A|^{n-1}$ for all $A \in \mathcal{M}_{n \times n}(F)$.

Exercise 649 Is the function $\text{adj} : \mathcal{M}_{2 \times 2}(\mathbb{R}) \rightarrow \mathcal{M}_{2 \times 2}(\mathbb{R})$ epic?

Exercise 650 For real numbers s and t , let $A(s, t) = \begin{bmatrix} s & 0 & t \\ 1 & 1 & 1 \\ t & 0 & 1 \end{bmatrix}$

and let $B(s, t) = \text{adj}(A(s, t))$. Find the set of all real numbers s satisfying the condition that $|A(s, t)| \neq |B(s, t)|$ for all $t \in \mathbb{R}$.

Exercise 651 Let F be a field. Does there exist a matrix A in $\mathcal{M}_{3 \times 3}(F)$ satisfying the condition that the rank of $\text{adj}(A)$ equals 2?

Exercise 652 Let n be a positive integer and for all $1 \leq j \leq n$, let m_j be a positive integer. Define the matrix $A = [a_{ij}] \in \mathcal{M}_{n \times n}(\mathbb{Q})$ by setting $a_{ij} = \binom{m_j + i - 1}{j - 1}$ for all $1 \leq i, j \leq n$. Calculate $|A|$.

Exercise 653 Let a and b be distinct elements of a field F and let n be a positive integer. Let $A(n) = [a_{ij}] \in \mathcal{M}_{n \times n}(F)$ be the matrix defined by

$$a_{ij} = \begin{cases} a & \text{if } i = j \\ b & \text{otherwise} \end{cases}.$$

Use induction on n to prove that $|A(n)| = [a + (n - 1)b](a - b)^{n-1}$.

Exercise 654 Let n be a positive integer and pick integers $1 \leq h, k \leq n$. Let $f, g \in \mathbb{R}^{\mathbb{R}}$ be the functions defined by

$$f : c \mapsto \begin{cases} |E_{h,c}| & \text{if } c \neq 0 \\ 0 & \text{if } c = 0 \end{cases}$$

and $g : c \mapsto |E_{hk;c}|$. Are these functions continuous?

Exercise 655 Let n be a positive integer and let $A \in \mathcal{M}_{n \times n}(\mathbb{Q})$ be a nonsingular matrix the entries of which are integers and the determinant of which is ± 1 . Show that all of the entries of A^{-1} are integers.

Exercise 656 Let F be a field and let $A \in \mathcal{M}_{2 \times 2}(F)$. Show that the matrix $A^2 + |A|I$ belongs to the subspace of $\mathcal{M}_{2 \times 2}(F)$ generated by $\{A\}$.

Exercise 657 Let $A = [a_{ij}] \in \mathcal{M}_{3 \times 3}(\mathbb{Q})$ be a matrix all of the entries of which are nonnegative one-digit integers. Let d be a positive integer dividing the three-digit integers $a_{11}a_{12}a_{13}$, $a_{21}a_{22}a_{23}$, and $a_{31}a_{32}a_{33}$. Show that d divides $|A|$.

Exercise 658 Let n be a positive integer and let F be a field. Let $A \in \mathcal{M}_{n \times n}(F)$ be a matrix satisfying the condition that $|A + B| = |A|$ for all $B \in \mathcal{M}_{n \times n}(F)$. Show that $A = O$.

Exercise 659 Let n be an odd positive integer let $A \in \mathcal{M}_{n \times n}(\mathbb{R})$. Show that there exists a diagonal matrix B the diagonal entries of which are ± 1 such that $A + B$ is nonsingular.

Exercise 660 Let $n > 1$ be an integer and let F be a field. Show that there exist subspaces W and Y of $\mathcal{M}_{n \times n}(F)$ satisfying $\mathcal{M}_{n \times n}(F) = W \oplus Y$ such that the restrictions of the determinant function δ_n to W and to Y are linear transformations.

Exercise 661 Let $n > 1$ be an integer and let B be the set of all of the nonsingular matrices in $\mathcal{M}_{n \times n}(\mathbb{R})$ all of the entries of which are either 1 or 0. Show that in every matrix in B there are at least $n-1$ entries which are equal to 0 and that there exists a matrix in B in which there are precisely $n-1$ entries equal to 0.

Exercise 662 Let A be a matrix formed by permuting the rows or columns of a Hadamard matrix. Is A necessarily a Hadamard matrix?

Exercise 663 Let V be a vector space of finite dimension n over a field F and let $\{v_1, \dots, v_n\}$ be a given basis for V . Let U be the subset of V consisting of all vectors of the form $y_a = \sum_{i=1}^n a^{i-1} v_i$, for $0 \neq a \in F$. Show that any subset of U having n elements is a basis for V .

Exercise 664 Let F be a field and let n be an even positive integer. Let $A \in \mathcal{M}_{n \times n}(F)$ be a matrix which can be written in block form as $[A_{ij}]$,

where $A_{ij} = \begin{bmatrix} 0 & c_i \\ -c_i & 0 \end{bmatrix}$ if $i = j$, and $A_{ij} = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$ otherwise.

Calculate the Pfaffian of A .

Exercise 665 Let $c = \frac{1}{2}(1 + i\sqrt{3})$. Find the set of all real numbers a such that $\begin{vmatrix} a & 1 & 1 \\ 1 & c & c^2 \\ 1 & c^2 & c \end{vmatrix} \in \mathbb{R}$.

Exercise 666 Let a, b, c, d be elements of a field F . Calculate

$$\begin{vmatrix} a & b & c & d \\ b & a & d & c \\ c & d & a & b \\ d & c & b & a \end{vmatrix}.$$

12

Eigenvalues and eigenvectors

One of the central problems in linear algebra is this: given a vector space V finitely generated over a field F , and given an endomorphism α of V , is there a way to select a basis B of V so that the matrix $\Phi_{BB}(\alpha)$ is as nice as possible? In this chapter we will begin by defining some basic notions which will help us address this problem.

Let V be a vector space over a field F and let $\alpha \in \text{End}(V)$. A scalar $c \in F$ is an **eigenvalue** of α if and only if there exists a vector $v \neq 0_V$ satisfying $\alpha(v) = cv$. Such a vector is called an **eigenvector**¹ of α associated with the eigenvalue c . Thus we see that a nonzero vector $v \in V$ is an eigenvector of α if and only if the subspace Fv of V is invariant under α . Every eigenvector of α is associated with a unique eigenvalue of α but any eigenvalue has, as a rule, many eigenvectors associated with it. The set of all eigenvalues of α is called the **spectrum** of α and is denoted by $\text{spec}(\alpha)$. Thus, $c \in \text{spec}(\alpha)$ if and only if the endomorphism $c\sigma_1 - \alpha$ of V is not monic.

¹The terms “eigenvalue” and “eigenvector” are due to Hilbert. Eigenvalues and eigenvectors are sometimes called **characteristic values** and **characteristic vectors** respectively, based on terminology used by Cauchy. Sylvester coined the term “latent values” since, as he put it, such scalars are “latent in a somewhat similar sense as vapour may be said to be latent in water or smoke in a tobacco-leaf”.

Example: The spectrum of $\alpha \in \text{End}(V)$ may be empty. For example, if $\alpha \in \text{End}(\mathbb{R}^2)$ is defined by $\alpha : \begin{bmatrix} a \\ b \end{bmatrix} \mapsto \begin{bmatrix} -b \\ a \end{bmatrix}$ then $\text{spec}(\alpha) = \emptyset$.

More generally, if V is any vector space over \mathbb{R} and if $\alpha \in \text{End}(V)$ satisfies $\alpha^2 = -\sigma_1$, then $\text{spec}(\alpha) = \emptyset$. To see this, note that if v is an eigenvector corresponding to an eigenvalue c then $-v = \alpha^2(v) = c^2v$ and so $(c^2 + 1)v = 0_V$, implying that $c^2 = -1$, which is impossible for a real number c .

Example: Let $\alpha \in \text{End}(\mathbb{R}^2)$ be defined by $\alpha : \begin{bmatrix} a \\ b \end{bmatrix} \mapsto \begin{bmatrix} b \\ a \end{bmatrix}$.

Then $c \in \text{spec}(\alpha)$ if and only if there exists a vector $\begin{bmatrix} a \\ b \end{bmatrix}$ satisfying $\begin{bmatrix} b \\ a \end{bmatrix} = \begin{bmatrix} ca \\ cb \end{bmatrix}$. Therefore we see that $\text{spec}(\alpha) = \{-1, 1\}$, where $\begin{bmatrix} a \\ a \end{bmatrix}$

is an example of an eigenvector of α associated with -1 and $\begin{bmatrix} a \\ a \end{bmatrix}$ is an example of an eigenvector of α associated with 1 , for any $0 \neq a \in \mathbb{R}$.

Example: Let V be the vector space of all infinitely-differentiable functions from \mathbb{R} to itself and let δ be the endomorphism of V which assigns to each such function its derivative. Then a function f , which is not the 0-function, is an eigenvector of δ if and only if there exists a scalar $c \in \mathbb{R}$ such that $\delta(f) = cf$. For any real number c , there is indeed such a function in V , namely the function $x \mapsto e^{cx}$. Thus $\text{spec}(\delta) = \mathbb{R}$. The set of all eigenvectors of δ associated with c is $\{ae^{cx} \mid a \neq 0\}$. This fact has important applications in the theory of differential equations².

Let α be an endomorphism of a vector space V of a field F having an eigenvalue c . If $\beta \in \text{Aut}(V)$ then c is also an eigenvalue of $\beta\alpha\beta^{-1}$. Indeed, if v is an eigenvector of α associated with c then



2

The first use of eigenvalues to study differential equations is due to the French mathematician **Jean d'Alembert**, one of the foremost researchers of the 18th century. Important solutions of eigenvalue problems for second-order differential equations were obtained in the 19th century by Swiss mathematician **Charles-François Sturm** and French mathematician **Joseph Liouville**.

$\beta\alpha\beta^{-1}(\beta(v)) = \beta\alpha(v) = \beta(cv) = c\beta(v)$ and $\beta(v) \neq 0_V$ since β is an automorphism. Therefore $\beta(v)$ is an eigenvector of $\beta\alpha\beta^{-1}$ associated with c .

Similarly, let $p(X) = \sum_{i=0}^n b_i X^i \in F[X]$. If $v \in V$ is an eigenvector of α associated with an eigenvalue c , then v is also an eigenvector of $p(\alpha) \in \text{End}(V)$ associated with the eigenvalue $p(c)$, since $p(\alpha)v = \sum_{i=0}^n b_i \alpha^i(v) = \sum_{i=0}^n b_i c^i v = p(c)v$. In particular, we see that, for any positive integer n , the vector v is an eigenvector of α^n associated with the eigenvalue c^n .

Let V be a vector space over a field F and let α be an endomorphism of V . A vector $v \in V$ is a **fixed point** of α if and only if $\alpha(v) = v$. It is clear that 0_V is a fixed point of every endomorphism of V and a nonzero vector v is a fixed point of α if and only if $1 \in \text{spec}(\alpha)$ and v is an eigenvector of α associated with 1 .

(12.1) Proposition: Let V be a vector space over a field F and let α be an endomorphism of V having an eigenvalue c . The subset W composed of 0_V and all eigenvectors of α associated to c is a subspace of V .

Proof: If $w, w' \in W$ and $a \in F$ then $\alpha(w + w') = \alpha(w) + \alpha(w') = cw + cw' = c(w + w')$ and $\alpha(aw) = a\alpha(w) = a(cw) = c(aw)$ and so both $w + w'$ and aw belong to W , proving that W is a subspace of V . \square

Let V be a vector space over a field F and let α be an endomorphism of V having an eigenvalue c . The subset W composed of 0_V and all eigenvectors of α associated with c , which we know by Proposition 12.1 is a subspace of V , is called the **eigenspace** of α associated with c . In particular, if 1 is an eigenvalue of α then the **fixed space** of α is the eigenspace associated with 1 . If $1 \notin \text{spec}(\alpha)$ then the fixed space of α is taken to be $\{0_V\}$.

Example: Let α be the endomorphism of \mathbb{R}^3 defined by

$$\alpha : \begin{bmatrix} a \\ b \\ c \end{bmatrix} \mapsto \begin{bmatrix} a \\ 0 \\ c \end{bmatrix}.$$

Then $1 \in \text{spec}(\alpha)$ and the eigenspace of α associated with 1 (namely

$$\text{the fixed space of } \alpha) \text{ is } \mathbb{R} \left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \right\}.$$

Example: Let $V = C(0, 1)$ and let α be the endomorphism of V defined by $\alpha(f) : x \mapsto \int_0^x \cos(\pi[x-t])f(t)dt$ for all $f \in V$. To find the eigenvalues of α , recall the trigonometric identity

$$\cos(\pi[x-t]) = \cos(\pi x)\cos(\pi t) + \sin(\pi x)\sin(\pi t).$$

Using this identity, we see that if $f \in V$ then

$$\alpha(f) : x \mapsto \left[\int_0^x \cos(\pi t)f(t)dt \right] \cos(\pi x) + \left[\int_0^x \sin(\pi t)f(t)dt \right] \sin(\pi x)$$

and so the image of α is contained in the subspace $W = \mathbb{R}\{g_1, g_2\}$ of V , where $g_1 : x \mapsto \cos(\pi x)$ and $g_2 : x \mapsto \sin(\pi x)$. It is easy to see that $\alpha(g_1) = \frac{1}{2}g_1$ and $\alpha(g_2) = \frac{1}{2}g_2$, so both of these functions are eigenvectors of α associated with the eigenvalue $\frac{1}{2}$. Moreover, $\{g_1, g_2\}$ is linearly independent. Thus we see that $\text{spec}(\alpha) = \{\frac{1}{2}\}$ and the eigenspace associated with this sole eigenvalue is W .

(12.2) Proposition: Let V be a vector space finitely generated over a field F and let α be an endomorphism of V . Then the following conditions on a scalar c are equivalent:

- (1) c is an eigenvalue of α ;
- (2) $c\sigma_1 - \alpha \notin \text{Aut}(V)$;
- (3) If $A = \Phi_{BB}(\alpha)$ for some basis B of V , then $|cI - A| = 0$.

Proof: (1) \Leftrightarrow (2): Condition (1) is satisfied if and only if there exists a nonzero vector $v \in V$ satisfying $\alpha(v) = cv$, i.e. if and only if $(c\sigma_1 - \alpha)(v) = 0_V$. This is true if and only if $\ker(c\sigma_1 - \alpha) \neq \{0_V\}$. Since V is finitely generated, by Proposition 7.3 we know that this is true if and only if condition (2) holds.

(2) \Leftrightarrow (3): This is a direct consequence of the fact that a matrix is nonsingular if and only if its determinant is nonzero. \square

From Proposition 12.2, we see how to define eigenvalues of square matrices over a field: if F is a field and n is a positive integer, then $c \in F$ is an **eigenvalue** of a matrix $A \in \mathcal{M}_{n \times n}(F)$ if and only if $|cI - A| = 0$, namely if and only if the matrix $cI - A$ is singular. The set of all eigenvalues of A will be denoted by $\text{spec}(A)$. In particular, we observe that a matrix

A is nonsingular if and only if $0 \notin \text{spec}(A)$. A vector $\begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix} \neq v \in F^n$

is an **eigenvector** of A associated with the eigenvalue c if and only if

$Av = cv$. The subset of F^n consisting of $\begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$ and all eigenvectors

of A associated with c is a subspace of F^n called the **eigenspace** associated with c . In the case that F equals \mathbb{R} or \mathbb{C} , the number $\rho(A) = \max\{|c| \mid c \in \text{spec}(A)\}$ is called the **spectral radius** of the matrix A , and plays a very important part in the numerical analysis of matrices. Note that if $F = \mathbb{C}$, then $\rho(A)$ is just the radius of the smallest circle in the complex plane, centered at the origin, containing $\text{spec}(A)$. Moreover, $\text{spec}(A)$ consists precisely of the poles of the function $z \mapsto |zI - A|^{-1}$. This observation allows the use of powerful techniques of complex analysis in the study of the spectra of complex matrices.

Example: It is not necessarily true that $\rho(AB) = \rho(A)\rho(B)$ for square matrices A and B . For example, if $A = \begin{bmatrix} 0 & 2 \\ 0 & 0 \end{bmatrix}$ and $B = \begin{bmatrix} 0 & 0 \\ 2 & 0 \end{bmatrix}$ in $\mathcal{M}_{2 \times 2}(\mathbb{R})$, then $\rho(A) = 0 = \rho(B)$, whereas $\rho(AB) = 4$.

Given a matrix $A \in \mathcal{M}_{n \times n}(F)$, we note that $|cI - A| = |(cI - A)^T| = |cI - A^T|$ and so $\text{spec}(A) = \text{spec}(A^T)$. However, for each such common eigenvalue, the associated eigenvectors may be different.

Example: Let F be a field and let n be a positive integer. If $v = \begin{bmatrix} a_1 \\ \vdots \\ a_n \end{bmatrix}$ and $w = \begin{bmatrix} b_1 \\ \vdots \\ b_n \end{bmatrix}$ are vectors in F^n , then we have already

noted that their exterior product $v \wedge w = vw^T \in \mathcal{M}_{n \times n}(F)$ and their interior product $v \odot w$, satisfies $v^T w = [v \odot w] \in \mathcal{M}_{1 \times 1}(F)$. Direct calculation shows that $(v \wedge w)v = (v \odot w)v$ and so $v \odot w$ is an eigenvalue of $v \wedge w$ associated with the eigenvector v .

Example: Let $A = \begin{bmatrix} 1 & 1 & -2 \\ -1 & 2 & 1 \\ 0 & 1 & -1 \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{R})$. Then $\text{spec}(A) = \{-1, 1, 2\}$ and so this is also $\text{spec}(A^T)$.

(1) The eigenspace of A associated with -1 is $\mathbb{R} \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}$ and the eigenspace of A^T associated with -1 is $\mathbb{R} \begin{bmatrix} 1 \\ 2 \\ -7 \end{bmatrix}$;

- (2) The eigenspace of A associated with 1 is $\mathbb{R} \begin{bmatrix} 3 \\ 2 \\ 1 \end{bmatrix}$ and the eigenspace of A^T associated with 1 is $\mathbb{R} \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}$;
- (3) The eigenspace of A associated with 2 is $\mathbb{R} \begin{bmatrix} 1 \\ 3 \\ 1 \end{bmatrix}$ and the eigenspace of A^T associated with 2 is $\mathbb{R} \begin{bmatrix} -1 \\ 1 \\ 1 \end{bmatrix}$.

Example: Let n be a positive integer and let $A = [a_{uj}]$ be an $n \times n$ Markov matrix, which we will consider as an element of $\mathcal{M}_{n \times n}(\mathbb{C})$. We

claim that $\rho(A) \leq 1$. Indeed, let $c \in \text{spec}(A)$ and let $v = \begin{bmatrix} b_1 \\ \vdots \\ b_n \end{bmatrix} \in \mathbb{C}^n$

be an eigenvector associated with c . Let $1 \leq h \leq n$ satisfy the condition that $|b_i| \leq |b_h|$ for all $1 \leq i \leq n$. Then $Av = cv$ implies, in particular, that $\sum_{j=1}^n a_{hj}b_j = cb_h$ and so

$$|c| \cdot |b_h| = |cb_h| = \left| \sum_{j=1}^n a_{hj}b_j \right| \leq \sum_{j=1}^n a_{hj}|b_j| \leq \left(\sum_{j=1}^n a_{hj} \right) |b_h| = |b_h|.$$

Hence $|c| \leq 1$, as claimed.

Example: Let n be a positive integer and let $A \in \mathcal{M}_{n \times n}(\mathbb{R})$ be a skew-symmetric matrix. We claim that $\text{spec}(A) \subseteq \{0\}$, with equality when n is odd. Indeed, let $c \in \text{spec}(A)$ and let $v \in \mathbb{R}^n$ be an eigenvector of A associated with c . Then $-A^T v = Av = cv$ and so $-A^T(Av) = -A^T(cv) = c(-A^T v) = c^2 v$. Therefore $-(Av \odot Av) =$

$-v^T A^T Av = c^2 v^T v = c^2(v \odot v)$. But if $y = \begin{bmatrix} b_1 \\ \vdots \\ b_n \end{bmatrix}$ is any vector in \mathbb{R}^n ,

then $y \odot y = \sum_{i=1}^n b_i^2 \geq 0$, with equality if and only if $y = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$. Since v

is nonzero, we conclude that we must have $c^2 = 0$ and so $c = 0$. Therefore $\text{spec}(A) \subseteq \{0\}$. If n is odd then, by the remark after Proposition 11.7, we know that A is singular and so $0 \in \text{spec}(A)$, establishing equality.

Example: Let n be a positive integer and let $A \in \mathcal{M}_{n \times n}(\mathbb{C})$. If c is a nonzero eigenvalue of A and if $v \in \mathbb{C}^n$ is an eigenvector associated with c then, by Proposition 11.13, we know that $|A|v = \text{adj}(A)Av = c[\text{adj}(A)]v$ and so $[\text{adj}(A)]v = c^{-1}|A|v$. Thus v is also an eigenvector of $\text{adj}(A)$ associated with the eigenvalue $c^{-1}|A|$.

If F is a field, if n is a positive integer, and if $A \in \mathcal{M}_{n \times n}(F)$ is a matrix having eigenvalue c , then $|cI - A| = 0$ and so by Proposition 11.13 we see that $(cI - A)\text{adj}(cI - A) = O$, and so $A[\text{adj}(cI - A)] = c[\text{adj}(cI - A)]$. From this we conclude that each of the columns of $\text{adj}(cI - A)$ must belong to the eigenspace of A associated with c .

Example: Let $A = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 4 & -17 & 8 \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{R})$. Then one can

calculate that $\text{spec}(A) = \{2 - \sqrt{3}, 2 + \sqrt{3}, 4\}$. Moreover, $\text{adj}(4I - A) =$

$\begin{bmatrix} 1 & -4 & 1 \\ 4 & -16 & 4 \\ 16 & -64 & 16 \end{bmatrix}$ and it is easy to check that the columns of this matrix

are indeed eigenvectors of A associated with 4.

(12.3) Proposition: If V is a vector space finitely generated over a field F and if $\alpha, \beta \in \text{End}(V)$ then $\text{spec}(\alpha\beta) = \text{spec}(\beta\alpha)$.

Proof: Let $c \in \text{spec}(\alpha\beta)$. If $c = 0$, this means that $\alpha\beta \notin \text{Aut}(V)$. Therefore either α or β is not an automorphism of V , and so $\beta\alpha \notin \text{Aut}(V)$ as well. Therefore we can assume $c \neq 0$. Let v be an eigenvector of $\alpha\beta$ associated with c and let $w = \beta(v)$. Then $\alpha(w) = \alpha\beta(v) = cv \neq 0_V$ and so $w \neq 0_V$. Moreover, $\beta\alpha(w) = \beta\alpha\beta(v) = \beta(cv) = c\beta(v) = cw$ and so w is an eigenvector of $\beta\alpha$ associated with c . Thus $\text{spec}(\alpha\beta) \subseteq \text{spec}(\beta\alpha)$. A similar argument shows the reverse inclusion, and so we have equality. \square

In particular, as a consequence of Proposition 12.3, we see that if F is a field, if n is a positive integer, and if $A, B \in \mathcal{M}_{n \times n}(F)$ then $\text{spec}(AB) = \text{spec}(BA)$.

As we noted at the beginning of the chapter, if we are given a vector space V finitely generated over a field F and an endomorphism α of V , we would like to find, to the extent possible, a basis B of V such that the matrix $\Phi_{BB}(\alpha)$ is nice, in the sense that it is amenable to quick and accurate calculations. Let V be a vector space over a field F (not necessarily finitely generated) and let $\alpha \in \text{End}(V)$. Then α is

diagonalizable if and only if there exists a basis B of V composed of eigenvectors of α .

Example: We have already seen that the set B of all functions in $\mathbb{R}^{\mathbb{R}}$ of the form $x \mapsto e^{ax}$, for some $a \in \mathbb{R}$, is linearly independent. Therefore $W = \mathbb{R}B$ is a subspace of $\mathbb{R}^{\mathbb{R}}$ which is not finitely generated, and B is a basis for W . Let α be the endomorphism of W which assigns to each $f \in W$ its derivative. Since each element of B is an eigenvector of α , we see that α is diagonalizable.

The following result characterizes the diagonalizable endomorphisms of finitely-generated vector spaces.

(12.4) Proposition: Let V be a vector space finitely generated over a field F and let $\alpha \in \text{End}(V)$. Then the following conditions on a basis $B = \{v_1, \dots, v_n\}$ are equivalent:

- (1) v_i is an eigenvector of α for each $1 \leq i \leq n$;
- (2) $\Phi_{BB}(\alpha)$ is a diagonal matrix.

Proof: (1) \Rightarrow (2): By (1) we know that for each $1 \leq i \leq n$ there exists a scalar c_i satisfying $\alpha(v_i) = c_i v_i$ and so, by definition, $\Phi_{BB}(\alpha)$ is the diagonal matrix $[a_{ij}]$ given by

$$a_{ij} = \begin{cases} c_i & \text{if } i = j \\ 0 & \text{otherwise} \end{cases}.$$

(2) \Rightarrow (1): If $\Phi_{BB}(\alpha) = [a_{ij}]$ is a diagonal matrix then for each $1 \leq i \leq n$ we have $\alpha(v_i) = a_{ii}v_i$ and so v_i is an eigenvector of α for each $1 \leq i \leq n$. \square

Let V be a vector space over a field F and let $\alpha \in \text{End}(V)$. If B is a basis of V made up of eigenvectors of α then, as we have seen above, the elements of B are also eigenvectors of $p(\alpha)$ for any polynomial $p(X) \in F[X]$. We need not stick to polynomials: suppose that each $v \in B$ is an eigenvector of α associated with an eigenvalue c_v of α . Given any function whatsoever $f : \text{spec}(\alpha) \rightarrow F$, we can define the endomorphism $f(\alpha)$ of V by setting $f(\alpha) : \sum_{v \in B} a_v v \mapsto \sum_{v \in B} a_v f(c_v) v$ and the elements of B are also eigenvectors of $f(\alpha)$. We note that if f and g are functions from $\text{spec}(\alpha)$ to F then $f(\alpha)g(\alpha) = g(\alpha)f(\alpha)$.

Now assume that V is finitely generated over F and that $B = \{v_1, \dots, v_n\}$ is a basis of V made up of eigenvectors of $\alpha \in \text{End}(V)$. For each $1 \leq i \leq n$, let c_i be the eigenvalue of α associated with v_i . We have already seen that for each such i there exists a polynomial $p_i(X)$, namely the Lagrange interpolation polynomial, satisfying the condition that

$$p_i(c_j) = \begin{cases} 1 & \text{if } i = j \\ 0 & \text{otherwise} \end{cases}.$$

Thus, given a function $f : \text{spec}(\alpha) \rightarrow F$, the polynomial $p(X) = \sum_{i=1}^n f(c_i)p_i(X)$ satisfies $p(c_i) = f(c_i)$ for all $1 \leq i \leq n$, and so $p(\alpha) = f(\alpha)$. Thus, for finitely-generated vector spaces, the above generalization does not in fact contribute anything new; it is important however, in the case of vector spaces which are not finitely generated.

We now show that the size of the spectrum of an endomorphism of a finitely-generated vector space is limited.

(12.5) Proposition: Let V be a vector space over a field F and let $\alpha \in \text{End}(V)$. If c_1, \dots, c_k are distinct eigenvalues of α and if v_i is an eigenvector of α associated with c_i for each $1 \leq i \leq k$, then the set $\{v_1, \dots, v_k\}$ is linearly independent.

Proof: Assume that the set $\{v_1, \dots, v_k\}$ is linearly dependent. Since $v_1 \neq 0_V$, we know that the set $\{v_1\}$ is linearly independent. Thus there exists an integer $1 \leq t < k$ such that the set $\{v_1, \dots, v_t\}$ is linearly independent but $\{v_1, \dots, v_{t+1}\}$ is linearly dependent. In other words, there exist scalars a_1, \dots, a_{t+1} , not all of which are equal to 0, such that $\sum_{i=1}^{t+1} a_i v_i = 0_V$ and so $0_V = c_{t+1} \left(\sum_{i=1}^{t+1} a_i v_i \right) = \sum_{i=1}^{t+1} a_i c_{t+1} v_i$. On the other hand, $0_V = \alpha \left(\sum_{i=1}^{t+1} a_i v_i \right) = \sum_{i=1}^{t+1} a_i \alpha(v_i) = \sum_{i=1}^{t+1} a_i c_i v_i$. Therefore $0_V = \sum_{i=1}^{t+1} a_i c_i v_i - \sum_{i=1}^{t+1} a_i c_{t+1} v_i = \sum_{i=1}^t a_i (c_i - c_{t+1}) v_i$. But the set $\{v_1, \dots, v_t\}$ is linearly independent and so $a_i (c_i - c_{t+1}) = 0$ for all $1 \leq i \leq t$. Since, by assumption, $c_i \neq c_{t+1}$ for all $1 \leq i \leq t$, this implies that $a_i = 0$ for all $1 \leq i \leq t$ and hence $a_{t+1} = 0$ as well, which is a contradiction. Thus the set $\{v_1, \dots, v_k\}$ must be linearly independent. \square

Thus we see that if F is a field and if $A \in \mathcal{M}_{n \times n}(F)$, then $\text{spec}(A)$ can have at most n elements. In particular, if F has more than n elements,

then there exists an element $c \in F \setminus \text{spec}(A)$, and so $(cI - A)v \neq \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$

for all $v \neq \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$. This implies that $cI - A$ is nonsingular.

From Proposition 12.5 we see that if α is an endomorphism of a vector space V over a field F having distinct eigenvalues c_1, \dots, c_k , and if W_i is the eigenspace associated with c_i for all $1 \leq i \leq k$, then the collection $\{W_1, \dots, W_k\}$ of subspaces of V is independent. Moreover, if

V is finitely generated over F then the number of elements in $\text{spec}(\alpha)$ is no greater than $\dim(V)$.

(12.6) Proposition: Let V be a vector space of finite dimension n over a field F . Then any endomorphism α of V having n distinct eigenvalues is diagonalizable.

Proof: This is a direct consequence of Proposition 12.4 and Proposition 12.5. \square

Example: Let $\alpha \in \text{End}(\mathbb{R}^2)$ be defined by $\alpha : \begin{bmatrix} a \\ b \end{bmatrix} \mapsto \begin{bmatrix} 3a - b \\ 3b - a \end{bmatrix}$. Then $\alpha \left(\begin{bmatrix} 1 \\ 1 \end{bmatrix} \right) = \begin{bmatrix} 2 \\ 2 \end{bmatrix}$ and so $\begin{bmatrix} 1 \\ 1 \end{bmatrix}$ is an eigenvector of α associated with the eigenvalue 2. Also, $\alpha \left(\begin{bmatrix} 1 \\ -1 \end{bmatrix} \right) = \begin{bmatrix} 4 \\ -4 \end{bmatrix}$ and so $\begin{bmatrix} 1 \\ -1 \end{bmatrix}$ is an eigenvector of α associated with the eigenvalue 4. Thus $B = \left\{ \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ -1 \end{bmatrix} \right\}$ is a basis for \mathbb{R}^2 and $\Phi_{BB}(\alpha) = \begin{bmatrix} 2 & 0 \\ 0 & 4 \end{bmatrix}$.

Example: Let $\alpha \in \text{End}(\mathbb{R}^2)$ be defined by $\alpha : \begin{bmatrix} a \\ b \end{bmatrix} \mapsto \begin{bmatrix} a + b \\ b \end{bmatrix}$. If $\begin{bmatrix} a \\ b \end{bmatrix} \neq \begin{bmatrix} 0 \\ 0 \end{bmatrix}$ and $\alpha \left(\begin{bmatrix} a \\ b \end{bmatrix} \right) = c \begin{bmatrix} a \\ b \end{bmatrix}$ then $cb = b$ and $a + b = ca$, and this can happen only when $b = 0$ and $c = 1$. Thus $\text{spec}(\alpha) = \{1\}$ and the eigenspace associated with this sole eigenvalue is $\mathbb{R} \begin{bmatrix} 0 \\ 1 \end{bmatrix}$. Since this is not all of \mathbb{R}^2 , we know that there is no basis of \mathbb{R}^2 made up of eigenvectors of α and hence α is not diagonalizable.

Note that the converse of Proposition 12.6 is false, as we easily see by taking $\alpha = \sigma_1$.

From the above, we know that if $A \in \mathcal{M}_{n \times n}(\mathbb{R})$ then the matrix has at most n distinct eigenvalues. However, it may have many less than that. If we assume that the entries of this matrix were chosen independently and randomly from a standard normal distribution, how many distinct eigenvalues should we expect? American mathematicians Alan Edelman, Eric Kostlan and Michael Shub have shown that if ε_n denotes the mathematical expectancy for the number of eigenvalues of such a matrix in \mathbb{R} , then $\lim_{n \rightarrow \infty} \frac{1}{\sqrt{n}} \varepsilon_n = \sqrt{\frac{2}{\pi}}$. The situation over the complex numbers is quite different. Given a matrix $A \in \mathcal{M}_{n \times n}(\mathbb{C})$ one can, with probability 1, pick

a matrix $B \in \mathcal{M}_{n \times n}(\mathbb{C})$ as near to A as we wish, which has n distinct eigenvalues in \mathbb{C} .

If F is a field, if n is a positive integer, and if $A \in \mathcal{M}_{n \times n}(F)$, then we can consider the matrix of polynomials $XI - A \in \mathcal{M}_{n \times n}(F[X])$. The determinant of this matrix, $|XI - A|$, is a polynomial in $F[X]$ called the **characteristic polynomial** of A . Note that this polynomial is always monic and of degree n .

Example: The characteristic polynomial of $\begin{bmatrix} 1 & -1 & 0 \\ 2 & 1 & 5 \\ 4 & 2 & 1 \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{R})$ is $X^3 - 3X^2 - 5X + 27$.

Example: The characteristic polynomial of $A = \begin{bmatrix} 1 & 2 & 1 & 2 \\ 0 & 1 & 2 & 3 \\ 3 & 2 & 1 & 1 \\ 1 & 1 & 2 & 0 \end{bmatrix} \in \mathcal{M}_{4 \times 4}(\mathbb{R})$ is $X^4 - 3X^3 - 11X^2 - 25X - 15$. If we sketch the graph of the polynomial function $t \mapsto t^4 - 3t^3 - 11t^2 - 25t - 15$, we see that it has real roots in the neighborhoods of -0.8 and 5.8 . (More precisely, they are approximately equal to -0.8062070604 and 5.7448832706 .) These are the only real eigenvalues of the matrix A .

Example: Let $F = GF(3)$. The characteristic polynomial of

$$A = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 2 & 0 & 1 & 0 \\ 0 & 1 & 1 & 0 \\ 1 & 1 & 1 & 0 \end{bmatrix} \in \mathcal{M}_{4 \times 4}(F)$$

equals $X^4 + X^3 + 1 = (X + 2)(X^3 + 2X^2 + 2X + 2)$, and so A has only one eigenvalue, namely 1 .

Example: The characteristic polynomial of $A = \begin{bmatrix} 5 & 4 & 2 \\ 4 & 5 & 2 \\ 2 & 2 & 2 \end{bmatrix}$ in $\mathcal{M}_{3 \times 3}(\mathbb{Q})$ is $(X - 10)(X - 1)^2$ and so $\text{spec}(A) = \{1, 10\}$. The eigenspace

of A associated with 10 is $\mathbb{Q} \begin{bmatrix} 2 \\ 2 \\ 1 \end{bmatrix}$, while the eigenspace of A

associated with 1 is $\mathbb{Q} \left\{ \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} -1 \\ 0 \\ 2 \end{bmatrix} \right\}$.

Example: Let $F = GF(2)$ and let $A = \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix} \in \mathcal{M}_{2 \times 2}(F)$.

The characteristic polynomial of A is $p(X) = X^2 + X + 1$ and, since $p(0) = p(1) = 1$ we see that $\text{spec}(A) = \emptyset$. In fact, it is possible to show that for every prime integer p there is a symmetric 2×2 matrix A over $GF(p)$ satisfying $\text{spec}(A) = \emptyset$. Later, we will show that any symmetric matrix over \mathbb{R} must have an eigenvalue.

Example: Let α be the endomorphism of \mathbb{C}^2 represented with respect to the canonical basis by the matrix $A = \begin{bmatrix} 1+i & 1 \\ 1 & 1-i \end{bmatrix} \in \mathcal{M}_{2 \times 2}(\mathbb{C})$. The characteristic polynomial of A is $(X-1)^2$ and so $\text{spec}(A) = 1$. The eigenspace associated with it is $\mathbb{C} \begin{bmatrix} i \\ 1 \end{bmatrix}$, which has dimension 1. Therefore α is not diagonalizable.

(12.7) Proposition: Let F be a field and let n be a positive integer. If $A \in M_{n \times n}(F)$ has characteristic polynomial $p(X) = \sum_{i=0}^n a_i X^i$, then $|A| = (-1)^n a_0$.

Proof: We note that $a_0 = p(0) = |0I - A| = |-A| = (-1)^n |A|$ and so $|A| = (-1)^n a_0$. \square

Any monic polynomial in $F[X]$ of positive degree is the characteristic polynomial of some square matrix over F . To see this, consider a polynomial $p(X) = \sum_{i=0}^n a_i X^i$, for $n > 0$. If $p(X)$ is monic, define the **companion matrix** of $p(X)$ to be the matrix $\text{comp}(p) \in \mathcal{M}_{n \times n}(F)$ to be the matrix $[a_{ij}]$ given by

$$a_{ij} = \begin{cases} 1 & \text{if } i = j + 1 \text{ and } j < n \\ -a_{i-1} & \text{if } j = n \\ 0 & \text{otherwise} \end{cases}.$$

Otherwise, define $\text{comp}(p)$ to be $\text{comp}(a_n^{-1}p)$.

(12.8) Proposition: Let F be a field and let n be a positive integer. If $p(X) = \sum_{i=0}^n a_i X^i \in F[X]$ is monic, then $p(X)$ is the characteristic polynomial of $\text{comp}(p) \in M_{n \times n}(F)$.

Proof: We will proceed by induction on n . For $n = 1$, the result is immediate. If $n = 2$ and if $p(X) = X^2 + a_1 X + a_0$, then

$\text{comp}(p) = \begin{bmatrix} 0 & -a_0 \\ 1 & -a_1 \end{bmatrix}$ and so the characteristic polynomial of $\text{comp}(p)$

is $\begin{vmatrix} X & a_0 \\ -1 & X + a_1 \end{vmatrix} = p(X)$ and we are done. Assume now that $n > 2$ and

the result has been established for $n - 1$. Then the characteristic poly-

mial of $\text{comp}(p)$ is $\begin{vmatrix} X & 0 & \dots & a_0 \\ -1 & X & \dots & a_1 \\ & \ddots & \ddots & \vdots \\ 0 & \dots & -1 & X + a_{n-1} \end{vmatrix}$. By Proposition 11.11,

this equals $X|\text{comp}(q)| + a_0(-1)^{n-1}|B|$, where $q(X) = \sum_{i=0}^{n-1} a_{i+1}X^i$ and where $B \in \mathcal{M}_{(n-1) \times (n-1)}(F)$ is an upper-triangular matrix with diagonal entries all equal to -1 . Thus $|B| = (-1)^{n-1}$ and, by the induction hypothesis, $|\text{comp}(q)| = q(X)$. Thus the characteristic polynomial of $\text{comp}(p)$ is $Xq(X) + a_0 = p(X)$, as desired. \square

Let F be a field and let n be a positive integer. Every nonsingular matrix $P \in \mathcal{M}_{n \times n}(F)$ defines a function ω_P from $\mathcal{M}_{n \times n}(F)$ to itself given by $\omega_P : A \mapsto P^{-1}AP$. In fact, $\omega_P \in \text{Aut}(\mathcal{M}_{n \times n}(F))$, where $\omega_P^{-1} = \omega_{P^{-1}}$. This is an automorphism of F -algebras and, indeed, it can be shown that every automorphism of unital F -algebras in $\text{Aut}(\mathcal{M}_{n \times n}(F))$ is of this form. Therefore the set of all automorphisms of the form ω_P is a group of automorphisms of $\mathcal{M}_{n \times n}(F)$ and so defines an equivalence relation \sim by setting $A \sim B$ if and only if $B = P^{-1}AP$. In this case, we say that the matrices A and B are **similar**. From what we have already seen, two matrices in $\mathcal{M}_{n \times n}(F)$ are similar if and only if they represent the same endomorphism of an n -dimensional vector space over F with respect to different bases. One of the problems before us is to decide, given two square matrices of the same size, if they are similar or not.

Note that if a matrix $A \in \mathcal{M}_{n \times n}(F)$ is similar to O , then it must equal O . Indeed, if $P^{-1}AP = O$ then $A = (PP^{-1})A(PP^{-1}) = P(P^{-1}AP)P^{-1} = POP^{-1} = O$.

Example: In $\mathcal{M}_{3 \times 3}(\mathbb{Q})$, the matrices $A = \begin{bmatrix} 20 & 10 & 10 \\ 10 & 0 & 10 \\ 10 & 10 & 10 \end{bmatrix}$ and $B = \begin{bmatrix} 80 & 130 & 100 \\ 10 & 10 & 10 \\ -50 & -80 & -60 \end{bmatrix}$ are similar, since $B = P^{-1}AP$, where $P = \begin{bmatrix} 1 & 2 & 1 \\ 1 & 0 & 1 \\ 2 & 3 & 3 \end{bmatrix}$. Thus we note that a symmetric matrix may be similar to

a matrix which is not symmetric.

Example: The matrices

$$A = \begin{bmatrix} 1 & 0 & 0 \\ -1 & 1 & 1 \\ -1 & 0 & 2 \end{bmatrix} \quad \text{and} \quad B = \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 2 \end{bmatrix}$$

in $\mathcal{M}_{3 \times 3}(\mathbb{Q})$ are not similar since, were they similar, the matrices $A - I$ and $B - I$ would also be similar, and thus have the same rank. But it is easy to see that the rank of $A - I$ equals 1, while the rank of $B - I$ equals 2.

Example: If matrices $A, B \in \mathcal{M}_{n \times n}(F)$ are similar, it does not follow that they commute. For example, let $A = \begin{bmatrix} 1 & 0 & -1 \\ 2 & 3 & 0 \\ -1 & 0 & -2 \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{R})$. Then $P = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 1 & 1 \end{bmatrix}$ is nonsingular and so $B = PAP^{-1} = \begin{bmatrix} 1 & 1 & -1 \\ 2 & 3 & 0 \\ 1 & 5 & -2 \end{bmatrix}$ is similar to A . However, $AB \neq BA$.

(12.9) Proposition: Let F be a field and let $k < n$ be positive integers. Let $A \in \mathcal{M}_{n \times n}(F)$ be a matrix which can be written in block form as $A = \begin{bmatrix} A_{11} & A_{12} \\ O & A_{22} \end{bmatrix}$, where $A_{11} \in \mathcal{M}_{k \times k}(F)$ and $A_{22} \in \mathcal{M}_{(n-k) \times (n-k)}(F)$. Then $\text{spec}(A) = \text{spec}(A_{11}) \cup \text{spec}(A_{22})$.

Proof: Let $c \in \text{spec}(A)$ and let $v \in F^n$ be an eigenvector associated with c . Write $v = \begin{bmatrix} v_1 \\ v_2 \end{bmatrix}$, where $v_1 \in F^k$ and $v_2 \in F^{n-k}$. Then

$$\begin{bmatrix} A_{11}v_1 + A_{12}v_2 \\ A_{22}v_2 \end{bmatrix} = \begin{bmatrix} A_{11} & A_{12} \\ O & A_{22} \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = Av = cv = \begin{bmatrix} cv_1 \\ cv_2 \end{bmatrix}.$$

From this we see immediately that if $v_2 \neq \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$ then $c \in \text{spec}(A_{22})$,

while if $v_2 = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$ then $c \in \text{spec}(A_{11})$. Therefore $\text{spec}(A)$ is contained in $\text{spec}(A_{11}) \cup \text{spec}(A_{22})$.

Conversely, let $c \in \text{spec}(A_{11})$ and let $v_1 \in F^k$ be an eigenvector associated with c . Then $A \begin{bmatrix} v_1 \\ 0 \end{bmatrix} = \begin{bmatrix} A_{11}v_1 \\ 0 \end{bmatrix} = c \begin{bmatrix} v_1 \\ 0 \end{bmatrix}$, proving that $c \in \text{spec}(A)$. Now assume that $d \in \text{spec}(A_{22}) \setminus \text{spec}(A_{11})$ and let $v_2 \in F^{n-k}$ be an eigenvector associated with d . Since $d \notin \text{spec}(A_{11})$, we know that the matrix $B = A_{11} - dI \in \mathcal{M}_{k \times k}(F)$ is nonsingular. Set

$$v_1 = B^{-1}A_{12}(-v_2). \text{ Then } (A - dI) \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = \begin{bmatrix} Bv_1 + A_{12}v_2 \\ (A_{22} - dI)v_2 \end{bmatrix} = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix},$$

showing that $d \in \text{spec}(A)$. Therefore $\text{spec}(A_{11}) \cup \text{spec}(A_{22}) \subseteq \text{spec}(A)$, proving equality. \square

Example: Let $A = \begin{bmatrix} 1 & 1 & 5 & 6 \\ -1 & 1 & 7 & 3 \\ 0 & 0 & 2 & 1 \\ 0 & 0 & -4 & 3 \end{bmatrix} \in \mathcal{M}_{4 \times 4}(\mathbb{C})$. Then

$$\begin{aligned} \text{spec}(A) &= \text{spec}\left(\begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix}\right) \cup \text{spec}\left(\begin{bmatrix} 2 & 1 \\ -4 & 3 \end{bmatrix}\right) \\ &= \{1 \pm i\} \cup \left\{\frac{1}{2}[5 \pm i\sqrt{15}]\right\}. \end{aligned}$$

(12.10) Proposition: Similar matrices in $M_{n \times n}(F)$, where F is a field and where n is a positive integer, have identical characteristic polynomials.

Proof: If $A, B \in \mathcal{M}_{n \times n}(F)$ satisfy $B = P^{-1}AP$ then

$$|XI - B| = |XI - P^{-1}AP| = |P^{-1}(XI - A)P| = |P|^{-1}|XI - A||P| = |XI - A|$$

as required. \square

Example: The converse of Proposition 12.10 is false. Indeed, the matrices $\begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix}$ and $\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$ are not similar, despite the fact that both of them have the same characteristic polynomial, $(X - 1)^2$.

Example: If A and B are square matrices over a field F , then we know that the matrices AB and BA are not necessarily equal.

They are also not necessarily similar. For example, if $A = \begin{bmatrix} 1 & 0 \\ 1 & 0 \end{bmatrix}$ and

$B = \begin{bmatrix} 0 & 0 \\ 1 & 1 \end{bmatrix}$ then $AB = O \neq BA$ and so AB and BA are not similar.

Nonetheless, by Proposition 12.3, we see that $\text{spec}(AB) = \text{spec}(BA)$.

Proposition 12.10 can be used to facilitate computation, as the following example shows.

Example: Let n be a positive integer, let F be a field, and let $A = [a_{ij}] \in \mathcal{M}_{n \times n}(F)$ be a symmetric tridiagonal matrix. That is to say, the entries of A satisfy the condition that $a_{ij} = a_{ji}$ when $|i - j| = 1$ and $a_{ij} = 0$ when $|i - j| > 1$. Set $p_0(X) = 0$ and, for each $1 \leq k \leq n$, let $p_k(X)$ be the characteristic polynomial of the $k \times k$ submatrix of A consisting of the first k rows and first k columns of the matrix $XI - A \in F[X]$. Then $p_n(X)$ is the characteristic polynomial of A and we have $p_1(X) = X - a_{11}$ and $p_k(X) = (X - a_{kk})p_{k-1}(X) - a_{ij}^2 p_{k-2}(X)$ for each $2 \leq k \leq n$. This recursion relation allows us to compute the characteristic polynomial of A quickly. Therefore, if A is any symmetric matrix, a good strategy is to try and find a symmetric tridiagonal matrix similar to it and then compute its characteristic polynomial.

Let α be an endomorphism of a vector space V finitely generated over a field F and let $c \in \text{spec}(\alpha)$. The **algebraic multiplicity** of c is the largest integer k such that $(X - c)^k$ divides the characteristic polynomial of α . The **geometric multiplicity** of c is the dimension of the eigenspace of α associated with c . The geometric multiplicity of c is not greater than its algebraic multiplicity, but these two numbers need not be equal, as the following examples will show. If these two multiplicities are equal, we say that c is a **semisimple eigenvalue** of α ; an eigenvalue which is not semisimple is **defective**. In particular, if the algebraic multiplicity of c is 1 then the same must be true for its geometric multiplicity. In that case, we say that c is a **simple eigenvalue** of α . If at least one eigenvalue of α has geometric multiplicity greater than 1, then α is **derogatory**.

Example: If $\alpha \in \text{End}(\mathbb{R}^2)$ is defined by $\alpha : \begin{bmatrix} a \\ b \end{bmatrix} \mapsto \begin{bmatrix} a + b \\ b \end{bmatrix}$

then $c = 1$ is an eigenvalue of α with associated eigenspace $\mathbb{R} \begin{bmatrix} 0 \\ 1 \end{bmatrix}$

and so the geometric multiplicity of c is 1. On the other hand, α is

represented with respect to the canonical basis by the matrix $\begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix}$,

so its characteristic polynomial is $(X - 1)^2$, implying that the algebraic multiplicity of c is 2.

Example: Let $\alpha \in \text{End}(\mathbb{R}^3)$ be the endomorphism represented with respect to the canonical basis by the matrix $\begin{bmatrix} 2 & 3 & 1 \\ 3 & 2 & 4 \\ 0 & 0 & -1 \end{bmatrix}$. The characteristic polynomial of α is $(X-5)(X+1)^2$ and so $\text{spec}(\alpha) = \{-1, 5\}$, where the algebraic multiplicity of -1 equals 2 and the algebraic multiplicity of 5 equals 1. The eigenspace associated with -1 is $\mathbb{R} \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix}$ and the eigenspace associated with 5 is $\mathbb{R} \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}$. Thus both eigenvalues have geometric multiplicity 1. Hence, 5 is a simple eigenvalue of α whereas -1 is defective.

Let n be a positive integer. If $\alpha \in \text{End}(\mathbb{R}^n)$ is represented with respect to a given basis of \mathbb{R}^n by a matrix all entries in which are positive, then Perron, using analytic methods, showed that the eigenvalue of largest absolute value of α is simple and positive, and has an associated eigenvector all entries of which are positive. This result has many important applications in statistics and economics, especially in input-output analysis. It was also used by Thurston in his classification of surface diffeomorphisms in topology. Perron's results were later extended by Frobenius to certain matrices all entries in which are nonnegative, and later by Karlin to certain endomorphisms of spaces which are not finite-dimensional.³

(12.11) Proposition: Let V be a vector space finitely generated over a field F and let α be an endomorphism of V satisfying the condition that the characteristic polynomial of α



3

Twentieth-century

German mathematician **Oskar Perron** worked in many areas of algebra and geometry. Fellow German mathematician **Georg Frobenius** is known for his important work in group theory and his work on bilinear forms. He was also the first to consider the rank of a matrix. **William Thurston** is a contemporary American geometer; contemporary American applied mathematician **Samuel Karlin** has published extensively in probability and statistics, as well as mathematical biology.

is completely reducible. Then α is diagonalizable if and only if every eigenvalue of α is semisimple.

Proof: Let $\text{spec}(\alpha) = \{c_1, \dots, c_k\}$. First of all, we will assume that there exists a basis D of V such that $\Phi_{DD}(\alpha)$ is a diagonal matrix. For each $1 \leq j \leq k$, denote by $m(j)$ the number of times that c_j appears on the diagonal of $\Phi_{DD}(\alpha)$. Then $\sum_{j=1}^k m(j) = n$ and by Proposition 12.4, we know that for each $1 \leq j \leq k$ there exists a subset of D , having $m(j)$ elements, which is a basis for the eigenspace of α associated with c_j . Moreover, the characteristic polynomial of α is $\prod_{j=1}^k (X - c_j)^{m(j)}$ and so $m(j)$ equals both the algebraic multiplicity and the geometric multiplicity of c_j for each $1 \leq j \leq k$, proving that each such c_j is semisimple. Conversely, assume that each c_j is semisimple, and for each $1 \leq j \leq k$ let $m(j)$ be the algebraic (and geometric) multiplicity of c_j . Let D_j be a basis for the eigenspace of α associated with c_j , and let $D = \bigcup_{j=1}^k D_j$. Then D is a linearly-independent subset of V having n elements, and so is a basis of V over F . The result then follows from Proposition 7.5. \square

Example: The condition in Proposition 12.11 that the characteristic polynomial of α be completely reducible is essential. To see this, consider the endomorphism α of \mathbb{R}^3 represented with respect to the canonical

basis by the matrix $A = \begin{bmatrix} 0 & 1 & 0 \\ -1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}$. The characteristic polynomial

of α is $(X - 1)(X^2 + 1) \in \mathbb{R}[X]$ and so $\text{spec}(\alpha) = \{1\}$, where 1 is a simple eigenvalue of α and so it is surely semisimple. The eigenspace of

α associated with this eigenvalue is $\mathbb{R} \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$ and so its dimension is 1.

Hence α is not diagonalizable.

Example: Consider the endomorphism α of \mathbb{R}^3 represented with

respect to the canonical basis by the matrix $\begin{bmatrix} -1 & -1 & -2 \\ 8 & -11 & -8 \\ -10 & 11 & 7 \end{bmatrix}$ and let

β be the endomorphism of \mathbb{R}^3 represented with respect to the canonical

basis by the matrix $\begin{bmatrix} 1 & -4 & -4 \\ 8 & -11 & -8 \\ -8 & 8 & 5 \end{bmatrix}$. These two endomorphisms have

the same characteristic polynomial $X^3 + 5X^2 + 3X - 9 = (X - 1)(X + 3)^2$. Thus the algebraic multiplicity of the eigenvalue 1 equals 1 and the

algebraic multiplicity of the eigenvalue -3 equals 2 . But for α , the geometric multiplicity of -2 equals 1 , so α is not diagonalizable. On the other hand, for β the geometric multiplicity of -2 equals 2 , and so β is diagonalizable.

Let F be a field and let (K, \bullet) be an associative unital F -algebra. If $v \in K$ and if $p(X) = \sum_{i=0}^k c_i X^i \in F[X]$, then $p(v) = \sum_{i=0}^k c_i v^i \in K$. For any polynomial $q(X) \in F[X]$ we have $p(v) \bullet q(v) = q(v) \bullet p(v)$. In particular, $v \bullet p(v) = p(v) \bullet v$. It is clear that $\text{Ann}(v) = \{p(X) \in F[X] \mid p(v) = 0_K\}$ is a subspace of $F[X]$. If $p(v) = 0_K$, we say that v **annihilates** the polynomial $p(X)$.

In particular, we note that all of the above is true for the associative unital F -algebra $\mathcal{M}_{n \times n}(F)$, where n is a positive integer. We note that if $A \sim B$ in $\mathcal{M}_{n \times n}(F)$ then there is a nonsingular matrix P such that $B = P^{-1}AP$ and so $p(B) = P^{-1}p(A)P$ so that if $p(A) = O$ then $p(B) = O$. Thus we see that $\text{Ann}(A) = \text{Ann}(B)$ whenever the matrices A and B are similar.

Example: Let $A = \begin{bmatrix} 2 & 1 \\ 1 & 0 \end{bmatrix} \in \mathcal{M}_{2 \times 2}(\mathbb{R})$ and let $p(X) = X^2 - X + 2 \in \mathbb{R}[X]$. Then $p(A) = A^2 - A + 2I = \begin{bmatrix} 5 & 1 \\ 1 & 3 \end{bmatrix}$. If $q(X) = X^2 - 2X - 1$ then $q(A) = O$ so $q(X) \in \text{Ann}(A)$.

(12.12) Proposition: Let F be a field and let (K, \bullet) be an associative F -algebra of finite dimension over F . Then $\text{Ann}(v)$ is nontrivial for each $v \in K$.

Proof: Let $\dim(V) = n$. If $v \in K$ then v^0, v^1, \dots, v^n must be linearly dependent and so there exist scalars a_0, \dots, a_n , not all equal to 0 , such that $\sum_{i=0}^n a_i v^i = 0_K$. In other words, there exists a nonzero polynomial $p(X) = \sum_{i=0}^n a_i X^i$ in $\text{Ann}(v)$. \square

We now show why one cannot define “three-dimensional complex numbers”.

(12.13) Proposition: If n is an odd integer greater than 1 then there is no way of defining on \mathbb{R}^n the structure of an \mathbb{R} -algebra which is also a field.

Proof: Assume that we can define an operation on \mathbb{R}^n (which we will denote by concatenation) which turns it into an \mathbb{R} -algebra which is also a field, and let v_1 be the identity element for this operation. Then $V \neq \mathbb{R}v_1$ since $\dim(V) > 1$. Pick an element $y \in V \setminus \mathbb{R}v_1$ and let $\alpha \in \text{End}(V)$ be

given by $\alpha : v \mapsto yv$, which is represented with respect to the canonical basis of \mathbb{R}^n by a matrix A . The characteristic polynomial $p(X)$ of A belongs to $\mathbb{R}[X]$ and has odd degree; therefore it has a root c in \mathbb{R} . Thus $p(X) = (X - c)^k q(X)$ for some $k \geq 1$ and some $q(X) \in \mathbb{R}[X]$ satisfying $q(c) \neq 0$. Let $\beta \in \text{End}(V)$ be given by $\beta : v \mapsto (y - cv_1)^k v$. Then $\beta \neq \sigma_0$ since $y \notin \mathbb{R}v_1$ and $0_V \neq q(c) = q(cv_1)$. But then $(y - cv_1)^k q(cv_1) = 0_V$, contradicting Proposition 2.3(12). \square

Let F be a field and let (K, \bullet) be an associative unital F -algebra. If $v \in K$ satisfies the condition that $\text{Ann}(v)$ is nontrivial then $\text{Ann}(v)$ must contain a polynomial $p(X) = \sum_{i=0}^n a_i X^i$ of minimal degree. This means, in particular, that $a_n \neq 0$ and so the monic polynomial $a_n^{-1}p(X)$ also belongs to $\text{Ann}(v)$. We claim that it is the unique monic polynomial of minimal degree in $\text{Ann}(v)$. Indeed, if $q(X)$ is a monic polynomial of degree n belonging to $\text{Ann}(v)$ not equal to $a_n^{-1}p(X)$, then $r(X) = q(X) - a_n^{-1}p(X) \in \text{Ann}(v)$. But $\deg(r) < n$, contradicting the minimality of the degree n of $p(X)$. Thus we see that $\text{Ann}(v)$, if nonempty, contains a unique monic polynomial of minimal positive degree, which we call the **minimal polynomial** of v over F and denote by $m_v(X)$.

In particular, if F is a field and if n is a positive integer, then any matrix $A \in \mathcal{M}_{n \times n}(F)$ has a minimal polynomial, which we denote by $m_A(X)$. If A and B are similar matrices, then $m_A(X) = m_B(X)$. Similarly, if V is a vector space finitely generated over a field F , and if $\alpha \in \text{End}(V)$ then α has a minimal polynomial $m_\alpha(X)$, and this equals the minimal polynomial of $\Phi_{DD}(\alpha)$ for any basis D of V .

Example: Let (K, \bullet) be an associative unital entire \mathbb{R} -algebra. Assume that $v \in K$ has a minimal polynomial $m_v(X) \in \mathbb{R}[X]$. By the Proposition 4.4, we know that $m_v(X) = \prod_{i=1}^t p_i(X)$, where the $p_i(X)$ are irreducible polynomials of degree at most 2. But then $\prod_{i=1}^t p_i(v) = 0_K$ and, since K is entire, there is some index h such that $p_h(v) = 0_K$. By minimality, this means that $m_v(X) = p_h(X)$. We thus conclude that any element of v having a minimal polynomial has one of degree at most 2.

(12.14) Proposition: Let F be a field and let (K, \bullet) be an associative F -algebra of finite dimension over F . If $v \in K$ satisfies the condition that $\text{Ann}(v)$ is nontrivial and if $p(X) \in \text{Ann}(v)$, then there is a polynomial $q(X) \in F[X]$ satisfying $p(X) = m_v(X)q(X)$.

Proof: If $p(X)$ is the 0-polynomial, pick $u(X)$ to be the 0-polynomial and we are done. Therefore assume that $\deg(p) > 0$. From Proposition 4.2, we know that we can write $p(X) = m_v(X)q(X) + r(X)$, where $q(X), r(X) \in F[X]$, with $\deg(r) < \deg(m_v)$. Since $p(v) = 0_K$ we

see that $0_K = m_v(v) \bullet q(v) + r(v) = r(v)$. Since $\deg(v) < \deg(m_A)$, we must have $\deg(v) = -\infty$ and so $p(X) = m_v(X)q(X)$. \square

(12.15) Proposition: Let F be a field and let (K, \bullet) be an associative unital F -algebra with multiplicative identity e . If $v \in K$ has a minimal polynomial $m_v(X) = \sum_{i=0}^n a_i X^i$ then:

- (1) v is a unit of K if and only if $a_0 \neq 0$; and
- (2) If v is a unit of K then $v^{-1} = g(v)$, where

$$g(X) = \sum_{i=1}^n (-a_0^{-1} a_i) X^{i-1} \in F[X].$$

Proof: If $a_0 \neq 0$ then $m_v(v) = 0_K$ implies that

$$e = a_0^{-1} \left[-\sum_{i=1}^n a_i v^i \right] = a_0^{-1} \left[-\sum_{i=1}^n a_i v^{i-1} \right] \bullet v = g(v) \bullet v = v \bullet g(v)$$

and so v is a unit and $v^{-1} = g(v)$. Conversely, assume that v is a unit. Had we $a_0 = 0$, we would have $0_V = m_v(v) = v \bullet \left[\sum_{i=1}^n a_i v^{i-1} \right]$ and so $0_K = v^{-1} m_v(v) = \sum_{i=1}^n a_i v^{i-1}$. Thus $\sum_{i=1}^n a_i X^{i-1} \in \text{Ann}(v)$, contradicting the minimality of the degree $m_v(X)$. Hence $a_0 \neq 0$. \square

It is important to note that the minimal polynomial of a matrix over a field need not equal its characteristic polynomial. For example, if we consider $I \in \mathcal{M}_{n \times n}(F)$ for any field F and any integer $n > 1$, then the characteristic polynomial of I is $(X-1)^n$ whereas its minimal polynomial is $X-1$.

Example: Let F be a field. The matrix $A = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} \in \mathcal{M}_{2 \times 2}(F)$ annihilates the polynomial $X(X-1)$, and this is in fact its minimal polynomial. It is also the characteristic polynomial of A . Thus we see that the minimal polynomial of a matrix does not have to be irreducible. Notice too that the rank of A equals 1, but the degree of its minimal polynomial is 2. Thus the degree of the minimal polynomial of a matrix may be larger than its rank.

Example: Let F be a field. The matrix $A = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \in \mathcal{M}_{3 \times 3}(F)$ annihilates the polynomial $X(X-1)$, and this is in fact its minimal polynomial. The characteristic polynomial of A is $X^2(X-1)$.

Example: One can check that $\begin{bmatrix} 1 & 0 & 0 \\ 0 & 3 & 0 \\ 0 & 0 & 3 \end{bmatrix}, \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 3 \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{Q})$

are not similar, but they both have the same minimal polynomial, namely $(X - 1)(X - 3)$.

Example: Proposition 12.15 can be used to calculate the inverse of a nonsingular matrix, though it is rarely the most efficient method of doing so. For example, the matrix $A = \begin{bmatrix} 2 & -2 & 4 \\ 2 & 3 & 2 \\ -1 & 1 & -1 \end{bmatrix}$ has minimal

polynomial $X^3 - 4X^2 + 7X - 10 = 0$ so $A^{-1} = \frac{1}{10}(A^2 - 4A + 7I) = \frac{1}{10} \begin{bmatrix} -5 & 2 & -16 \\ 0 & 2 & 4 \\ 5 & 0 & 10 \end{bmatrix}$.

(12.16) Proposition (Cayley-Hamilton Theorem): Let F be a field and let n be a positive integer. Then every matrix in $\mathcal{M}_{n \times n}(F)$ annihilates its characteristic polynomial.

Proof: Let A be a matrix in $\mathcal{M}_{n \times n}(F)$ having minimal polynomial $p(X) = X^n + \sum_{i=0}^{n-1} a_i X^i$. Let us look at the matrix

$$[g_{ij}(X)] = \text{adj}(XI - A) \in \mathcal{M}_{n \times n}(F[X]),$$

where each $g_{ij}(X)$ is a polynomial of degree at most $n - 1$. Then we can write this matrix in the form $\sum_{i=1}^n B_i X^{n-i}$, where the B_i are matrices in $\mathcal{M}_{n \times n}(F)$. Moreover, we know that

$$\begin{aligned} p(X)I &= |XI - A|I = (XI - A)\text{adj}(XI - A) \\ &= (XI - A) \left(\sum_{i=1}^n B_i X^{n-i} \right). \end{aligned}$$

Equating coefficients of the various powers of X , we thus see

$$\begin{aligned} B_1 &= I \\ B_2 - AB_1 &= a_{n-1}I \\ B_3 - AB_2 &= a_{n-2}I \\ &\vdots \\ B_n - AB_{n-1} &= a_1I \\ -AB_n &= a_0I \end{aligned}$$

For $1 \leq h \leq n$, multiply both sides of the h th equation in the above list on the left by A^{n+1-h} and then sum both sides, to obtain $O = p(A)$, as desired. \square

In particular, we conclude from Proposition 12.14 and Proposition 12.16 that the minimal polynomial of any $n \times n$ matrix over a field divides its characteristic polynomial and so the degree of the minimal polynomial is at most n .

Let V be a vector space finitely generated over a field F and let $\sigma_0 \neq \alpha \in \text{End}(V)$. In Proposition 12.4 we saw that α is diagonalizable if and only if that basis is composed of eigenvectors of V . Moreover, if $\text{spec}(\alpha) = \{c_1, \dots, c_k\}$ and if, for each $1 \leq i \leq k$, we denote the eigenspace of α associated with c_i by W_i , then for each $1 \leq i \leq k$ we have a projection $\pi_i \in \text{End}(V)$ satisfying the following conditions:

- (1) $\text{im}(\pi_i) = W_i$;
- (2) $\pi_1 + \dots + \pi_k = \sigma_0$;
- (3) $\pi_i \pi_j = \sigma_0$ whenever $i \neq j$;
- (4) $\alpha = c_1 \pi_1 + \dots + c_k \pi_k$.

For each $1 \leq h \leq k$, let $p_h(X)$ be the h th Lagrange interpolation polynomial determined by c_1, \dots, c_k . Then we can check that $\pi_h = p_h(\alpha)$ for each h , since $p_h(X)(X - c_h)$ is just a scalar multiple of the minimal polynomial of α .

Is it possible to simultaneously diagonalize two distinct endomorphisms of V ? Indeed, let V be a vector space finitely generated over a field F and let α and β be distinct elements of $\text{End}(V) \setminus \{\sigma_0\}$. There exists a basis D of V such that both $\Phi_{DD}(\alpha)$ and $\Phi_{DD}(\beta)$ are diagonal matrices if and only if the elements of D are eigenvectors of α as well as of β . Suppose that we have in hand such a basis $D = \{u_1, \dots, u_k\}$. Since diagonal matrices commute with each other, we see that $\Phi_{DD}(\alpha\beta) = \Phi_{DD}(\alpha)\Phi_{DD}(\beta) = \Phi_{DD}(\beta)\Phi_{DD}(\alpha) = \Phi_{DD}(\beta\alpha)$ and so $\alpha\beta = \beta\alpha$. Therefore a necessary condition for both endomorphisms of V to be represented by diagonal matrices with respect to the same basis is that they form a commuting pair.

We also note that if D is a basis for a vector space V over a field F . Then the set of all endomorphisms α of V satisfying the condition that $\Phi_{DD}(\alpha)$ is a diagonal matrix is a subspace of $\text{End}(V)$. Indeed, this is an immediate consequence of the fact that the set of all diagonal $n \times n$ matrices is a subspace of $\mathcal{M}_{n \times n}(F)$.

(12.17) Proposition: Let V be a vector space over a field F and let (α, β) be a commuting pair of endomorphisms of V . Then $p(\alpha)q(\beta) = q(\beta)p(\alpha)$ for any $p(X), q(X) \in F[X]$.

Proof: Initially, we will consider the special case of $q(X) = X$. If $p(X) = \sum_{i=0}^n a_i X^i$ then $\beta\alpha^2 = (\beta\alpha)\alpha = (\alpha\beta)\alpha = \alpha(\beta\alpha) = \alpha(\alpha\beta) = \alpha^2\beta$, and by induction we similarly have $\beta\alpha^k = \alpha^k\beta$ for every positive integer k . Therefore

$$\beta p(\alpha) = \beta \left(\sum_{i=0}^n a_i \alpha^i \right) = \sum_{i=0}^n a_i \beta \alpha^i = \sum_{i=0}^n a_i \alpha^i \beta = \left(\sum_{i=0}^n a_i \alpha^i \right) \beta = p(\alpha) \beta.$$

Now a proof similar to the first part shows that $p(\alpha)\beta^k = \beta^k p(\alpha)$ for every positive integer k and hence, by a proof similar to the second part, we get $p(\alpha)q(\beta) = q(\beta)p(\alpha)$ for any $p(X), q(X) \in F[X]$. \square

As a consequence of this we note that if $\alpha, \beta \in \text{End}(V)$ are commuting projections then $(\alpha\beta)^2 = (\alpha\beta)(\alpha\beta) = \alpha(\beta\alpha)\beta = \alpha(\alpha\beta)\beta = \alpha^2\beta^2 = \alpha\beta$ and so $\alpha\beta$ is a projection as well.

(12.18) Proposition: Let V be a vector space finitely generated over a field F and let $\alpha, \beta \in \text{End}(V)$ be diagonalizable endomorphisms of V . Then there exists a basis of V relative to which both α and β can be represented by diagonal matrices if and only if $\alpha\beta = \beta\alpha$.

Proof: We have already noted that if α and β can both be represented by diagonal matrices with respect to a given basis of V then we must have $\alpha\beta = \beta\alpha$. Conversely, assume that α and β are diagonalizable endomorphisms of V satisfying $\alpha\beta = \beta\alpha$. Then, as we have already seen, there exist distinct scalars c_1, \dots, c_k and projections $\pi_1, \dots, \pi_k \in \text{End}(V)$ such that $\pi_1 + \dots + \pi_k = \sigma_1$, $\pi_i \pi_j = \sigma_0$ for $i \neq j$, and $c_1 \pi_1 + \dots + c_k \pi_k = \alpha$. Similarly, there exist scalars d_1, \dots, d_t and projections $\eta_1, \dots, \eta_t \in \text{End}(V)$ such that $\eta_1 + \dots + \eta_t = \sigma_1$, $\eta_i \eta_j = \sigma_0$ for $i \neq j$, and $d_1 \eta_1 + \dots + d_t \eta_t = \beta$. Therefore

$$\alpha = \alpha\sigma_1 = \left(\sum_{i=1}^k c_i \pi_i \right) \left(\sum_{j=1}^t \eta_j \right) = \sum_{i=1}^k \sum_{j=1}^t c_i \pi_i \eta_j$$

and

$$\beta = \beta\sigma_1 = \left(\sum_{j=1}^t d_j \eta_j \right) \left(\sum_{i=1}^k \pi_i \right) = \sum_{j=1}^t \sum_{i=1}^k d_j \eta_j \pi_i.$$

Since we saw that, for each $1 \leq i \leq k$ we have $\pi_i = p_i(\alpha)$ for some $p_i(X) \in F[X]$ and similarly for each $1 \leq j \leq t$ we have $\eta_j = q_j(\beta)$ for some $q_j(X) \in F[X]$, we conclude that $\pi_i \eta_j = \eta_j \pi_i$ for each such i and j . Call this common value θ_{ij} . By the comments after Proposition 12.17, we see that θ_{ij} is also a projection in $\text{End}(V)$.

We note that $\theta_{ij}\theta_{hm} = \pi_i\eta_j\pi_h\eta_m = \pi_i\pi_h\eta_j\eta_m$ and this equals σ_0 when $i \neq j$ or $h \neq m$. We also note that

$$\sum_{i=1}^k \sum_{j=1}^t \theta_{ij} = \left(\sum_{i=1}^k \pi_i \right) \left(\sum_{j=1}^t \eta_j \right) = \sigma_1.$$

Thus we have shown that α and β are simultaneously diagonalizable, using those projections θ_{ij} which are nonzero (as some of them may be). \square

Algorithms for the computation of the eigenvalues and eigenvectors of a given matrix are usually very complicated, especially if speed of computation is a major consideration. Therefore, we shall not go into the description of such algorithms in detail. As a rule of thumb, it is best to try to compute eigenvectors directly, and not through finding roots of the characteristic polynomial, since small errors in the computation of eigenvalues may often lead to large errors in the computation of the corresponding eigenvectors. For matrices over \mathbb{R} , there are often reasonably efficient iterative methods to find at least some of the eigenvectors. We will bring here one example to find an eigenvector associated with the real eigenvalue of a matrix over \mathbb{R} having greatest absolute value, under assumption that such an eigenvalue indeed exists. The algorithm is based on the observation that if c is an eigenvalue of a matrix $A \in \mathcal{M}_{n \times n}(\mathbb{R})$ then c^k is an eigenvalue of A^k . Therefore, if k is sufficiently large, the matrix $A(A^k)$ is approximately equal to cA^k . Therefore, if we select an arbitrary vector $v^{(0)} \in \mathbb{R}^n$ and successively define vectors $v^{(1)}, v^{(2)}, \dots$ by setting $v^{(i+1)} = Av^{(i)}$ for each $i \geq 0$, then $Av^{(k)} = A^{k+1}v^{(0)}$ and this is roughly equal to $cv^{(k)}$. Therefore, if the conditions are amenable (and we will not go into the precise conditions necessary for this to happen), the vector $v^{(k)}$ is a reasonable approximation to an eigenvector of A associated with c . Of course, we must always remember that repeated computations lead to accumulating roundoff and truncation errors; one way of combating these is to divide each entry in $v^{(i)}$ by the absolute value of the largest entry, and use this “normalized” vector in the next iteration⁴.



4

Of the many numerical analysts who studied computational methods for finding eigenvalues, one of the most important is the British mathematician **James H. Wilkinson**, a former assistant of Alan Turing and one of the major early innovators in numerical linear algebra. The iteration algorithm given here

Example: Consider $A = \begin{bmatrix} 5 & 1 \\ -3 & 1 \end{bmatrix} \in \mathcal{M}_{2 \times 2}(\mathbb{R})$ and let us pick $v^{(0)} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$ then

$$Av^{(0)} = \begin{bmatrix} 6 \\ -2 \end{bmatrix} \text{ and so we will take } v^{(1)} = \begin{bmatrix} 1 \\ -\frac{1}{3} \end{bmatrix};$$

$$Av^{(1)} = \frac{1}{3} \begin{bmatrix} 14 \\ -10 \end{bmatrix} \text{ and so we will take } v^{(2)} = \begin{bmatrix} 1 \\ -\frac{5}{7} \end{bmatrix};$$

$$Av^{(2)} = \frac{1}{7} \begin{bmatrix} 30 \\ -26 \end{bmatrix} \text{ and so we will take } v^{(3)} = \begin{bmatrix} 1 \\ -\frac{13}{13} \end{bmatrix};$$

$$Av^{(3)} = \frac{1}{13} \begin{bmatrix} 62 \\ -58 \end{bmatrix} \text{ and so we will take } v^{(4)} = \begin{bmatrix} 1 \\ -\frac{29}{31} \end{bmatrix};$$

$$Av^{(4)} = \frac{1}{31} \begin{bmatrix} 126 \\ -122 \end{bmatrix} \text{ and so we will take } v^{(5)} = \begin{bmatrix} 1 \\ -\frac{61}{63} \end{bmatrix}.$$

It seems that this sequence of vectors is converging to $\begin{bmatrix} 1 \\ -1 \end{bmatrix}$ and, indeed, one can check that this is an eigenvector of A associated with the eigenvalue 4.

Example: Let n be a positive integer and let $A \in \mathcal{M}_{n \times n}(\mathbb{R})$ be a matrix of the form $[cB + (1-c)D]^T$, where $B \in \mathcal{M}_{n \times n}(\mathbb{R})$ is a Markov matrix, $c \in \mathbb{R}$ satisfies $0 \leq c \leq 1$, and $D = \begin{bmatrix} 1 \\ \vdots \\ 1 \end{bmatrix} \wedge \begin{bmatrix} d_1 \\ \vdots \\ d_n \end{bmatrix}$ for

nonnegative real numbers d_i satisfying $\sum_{i=1}^n d_i = 1$. Such matrices have been called **Google matrices** since they are needed for the *Page-Rank* algorithm used by the internet search engine *Google*TM to compute an estimate of web-page importance for ranking search results (for these purposes, a typical value for c is 0.85). The value of n can be very large, often far larger than 10^9 .

One can show that the eigenvalues e_1, \dots, e_n of such a matrix satisfy $1 = |e_1| \geq |e_2| \geq \dots \geq |e_n| \geq 0$ and so the power method mentioned above can (and is) used by *Google* to rapidly compute an eigenvector associated to e_1 . Recently, Stanford University researchers Taher Haveliwala and Sepandar Kamvar have shown that for any Google matrix, $|e_2| \leq c$, with

was first studied in the 1920's by the Austrian applied mathematician **Richard von Mises**, who later emigrated to the United States.

equality happening under conditions that hold in the case of those matrices arising in this particular application. Eigenvectors corresponding to this second eigenvalue can be used to detect and combat link spamming on the internet.

One can also consider a generalization of the eigenvalue problem: given endomorphisms α and β of a vector space V , find all scalars c such that $c\beta - \alpha$ is not monic. Problems of this sort arise naturally, for example, in plasma physics and in the design of control systems. When β is an automorphism, as is usually the case, this can be reduced to the usual eigenvalue problem for the endomorphism $\beta^{-1}\alpha$, but there are sometimes reasons for not wanting to do so. For example, even if both α and β are represented with respect to a given basis by symmetric matrices, the matrix representing $\beta^{-1}\alpha$ may not be symmetric. Therefore, some specialized algorithms have been developed to find solutions of the generalized eigenvalue problem directly.

Exercises

Exercise 667 Find the characteristic polynomial of the matrix

$$\begin{bmatrix} 3 & 2 & 2 \\ 1 & 4 & 1 \\ 2 & -4 & -1 \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{R}).$$

Exercise 668 Find the characteristic polynomial of the matrix

$$\begin{bmatrix} 1 & 3 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 2 & 2 & 0 \\ 0 & 0 & 0 & 0 & 2 \end{bmatrix} \in \mathcal{M}_{5 \times 5}(\mathbb{R}).$$

Exercise 669 Let $a, b, c \in \mathbb{R}$. Find the characteristic polynomial of the

$$\text{matrix } \begin{bmatrix} 0 & 0 & 0 & a \\ a & 0 & 0 & b \\ 0 & b & 0 & c \\ 0 & 0 & c & 0 \end{bmatrix} \in \mathcal{M}_{4 \times 4}(\mathbb{R}).$$

Exercise 670 Let n be a positive integer and let $A \in \mathcal{M}_{n \times n}(\mathbb{Q})$. Let c_1, \dots, c_n be the list of (not necessarily distinct) eigenvalues of A , considered as a matrix in $\mathcal{M}_{n \times n}(\mathbb{C})$. Show that $\sum_{i=1}^n c_i$ and $\prod_{i=1}^n c_i$ are rational numbers.

Exercise 671 Let F be a field, let n be a positive integer, and let $A, B \in \mathcal{M}_{n \times n}(F)$. Assume that A and B have the same characteristic polynomial $p(X) \in F[X]$. Is it necessarily true that $p(X)$ is the characteristic polynomial of AB ?

Exercise 672 Find infinitely-many matrices in $\mathcal{M}_{3 \times 3}(\mathbb{R})$, all of which have characteristic polynomial $X(X-1)(X-2)$.

Exercise 673 Let n be a positive integer. Show that every matrix $A \in \mathcal{M}_{n \times n}(\mathbb{R})$ can be written as the sum of two nonsingular matrices.

Exercise 674 Let $F = GF(3)$ and let n be a positive integer. Let $D = [d_{ij}] \in \mathcal{M}_{n \times n}(F)$ be a nonsingular diagonal matrix and let $A \in \mathcal{M}_{n \times n}(F)$. Show that $1 \notin \text{spec}(DA)$ if and only if $D - A$ is nonsingular.

Exercise 675 Let F be a field of characteristic other than 2. For each positive integer n , let T_n be the set of all diagonal matrices in $\mathcal{M}_{n \times n}(\mathbb{R})$ the diagonal entries of which belong to $\{-1, 1\}$. For any $A \in \mathcal{M}_{n \times n}(\mathbb{R})$, show that there exists a matrix $D \in T_n$ satisfying $1 \notin \text{spec}(DA)$.

Exercise 676 Let n be a positive integer and let $\alpha : \mathcal{M}_{n \times n}(\mathbb{C}) \rightarrow \mathbb{C}^n$

be the function defined by $\alpha : A \mapsto \begin{bmatrix} a_0 \\ \vdots \\ a_{n-1} \end{bmatrix}$, where $X^n + \sum_{i=0}^{n-1} a_i X^i$

is the characteristic polynomial of A . Is α a linear transformation?

Exercise 677 Let α be the endomorphism of \mathbb{R}^3 defined by

$$\alpha : \begin{bmatrix} a \\ b \\ c \end{bmatrix} \mapsto \begin{bmatrix} a - b \\ a + 2b + c \\ -2a + b - c \end{bmatrix}.$$

Find the eigenvalues of α and, for each eigenvalue, find the associated eigenspace.

Exercise 678 Let A is a nonempty set and let V be the collection of all subsets of A , which is a vector space over $GF(2)$. Let B be a fixed subset of A and let $\alpha : V \rightarrow V$ be the endomorphism defined by $\alpha : Y \mapsto Y \cap B$. Find the eigenvalues of α and, for each eigenvalue, find the associated eigenspace.

Exercise 679 Let α be the endomorphism of \mathbb{R}^4 defined by

$$\alpha : \begin{bmatrix} a \\ b \\ c \\ d \end{bmatrix} \mapsto \begin{bmatrix} b + c \\ c \\ 0 \\ 0 \end{bmatrix}.$$

Find the eigenvalues of α . Do there exist two-dimensional subspaces W and Y of \mathbb{R}^4 , both invariant under α , such that $\mathbb{R}^4 = W \oplus Y$?

Exercise 680 Let V be a vector space finitely generated over a field F and let α be an endomorphism of V having an eigenvalue c . For any $p(X) \in F[X]$, show that $p(c)$ is an eigenvalue of $p(\alpha)$.

Exercise 681 Let V be the vector space of all functions in $\mathbb{R}^{\mathbb{R}}$ which are infinitely differentiable and let $\alpha : V \rightarrow V$ be the endomorphism of V defined by $\alpha : f \mapsto f''$. If n is an integer, show that the function $f : x \mapsto \sin(nx)$ is an eigenvector of α^2 and find the associated eigenvalue.

Exercise 682 Let F be a field and let $V = F^{\mathbb{Z}}$. Let α be the endomorphism of V defined by $\alpha(f) : i \mapsto f(i+1)$ for all $i \in \mathbb{Z}$. Show that $\text{spec}(\alpha) = \emptyset$.

Exercise 683 Let V be the vector space composed of all polynomial functions from \mathbb{R} to itself, let $a \in \mathbb{R}$, and let α be the endomorphism of V defined by $\alpha(p) : x \mapsto (x-a)[p'(x) + p'(a)] - 2[p(x) - p(a)]$, where p' denotes the derivative of p . Find the eigenvalues of α and for each such eigenvalue, find the associated eigenspace.

Exercise 684 Let α be the endomorphism of $\mathcal{M}_{2 \times 2}(\mathbb{R})$ defined by $\alpha : \begin{bmatrix} a & b \\ c & d \end{bmatrix} \mapsto \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$. Find the eigenvalues of α and for each such eigenvalue, find the associated eigenspace.

Exercise 685 Let V be a vector space over \mathbb{Q} and let $\alpha \in \text{End}(V)$ be a projection. Show that $\text{spec}(\alpha) \subseteq \{0, 1\}$.

Exercise 686 Let V be a vector space of dimension $n > 0$ over a field F . Let α be an endomorphism of V for which there exists a set A of $n+1$ distinct eigenvectors satisfying the condition that every subset of A of size n is a basis for V . Show that all of the eigenvectors in V are associated with the same eigenvalue c of α and that $\alpha = c\sigma_1$.

Exercise 687 For $a, b \in \mathbb{R}$, let $A = \begin{bmatrix} a & b & 0 \\ b & a & b \\ 0 & b & a \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{R})$. Find the eigenvalues of A .

Exercise 688 Let $A \in \mathcal{M}_{2 \times 2}(\mathbb{R})$ be a matrix of the form $\begin{bmatrix} a & b \\ c & a \end{bmatrix}$, where $a > 0$ and $bc > 0$. Show that A has two distinct eigenvalues in \mathbb{R} .

Exercise 689 Find the eigenvalues of the matrix $\begin{bmatrix} 5 & 6 & -3 \\ -1 & 0 & 1 \\ 2 & 2 & -1 \end{bmatrix} \in$

$\mathcal{M}_{3 \times 3}(\mathbb{R})$ and, for each such eigenvalue, find the associated eigenspace.

Exercise 690 Find the eigenvalues of the matrix $\begin{bmatrix} 0 & 2 & 1 \\ -2 & 0 & 3 \\ -1 & -3 & 0 \end{bmatrix} \in$

$\mathcal{M}_{3 \times 3}(\mathbb{C})$ and, for each such eigenvalue, find the associated eigenspace.

Exercise 691 Let W be the subspace of \mathbb{R}^∞ consisting of all convergent sequences and let α be the endomorphism of W defined by

$$\alpha : [a_1, a_2, \dots] \mapsto \left[\left(\lim_{i \rightarrow \infty} a_i \right) - a_1, \left(\lim_{i \rightarrow \infty} a_i \right) - a_2, \dots \right].$$

Find all eigenvalues of α and, for each eigenvalue, find the corresponding eigenspace.

Exercise 692 Find the eigenvalues of the matrix $\begin{bmatrix} 1 & -1 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}$ in

$\mathcal{M}_{3 \times 3}(\mathbb{C})$ and, for each such eigenvalue, find the associated eigenspace.

Exercise 693 Does there exist a real number a such that

$$\text{spec} \left(\begin{bmatrix} 1 & -1 & 0 \\ 0 & a & -1 \\ -6 & 11 & -5 \end{bmatrix} \right) = \{-2, -1, 0\}?$$

Exercise 694 Let α be an endomorphism of a vector space V over a field F and let v and w be eigenvectors of α . If $v + w \neq 0_V$, show that $v + w$ is an eigenvector of α if and only if both v and w correspond to the same eigenvalue.

Exercise 695 Show that the matrix $A = \begin{bmatrix} 1 & 0 & a \\ a & a & a \\ a & 0 & -1 \end{bmatrix}$ has three dis-

tinct eigenvalues for any real number a .

Exercise 696 Let n be a positive integer and let t be a nonzero real number. Let $A \in \mathcal{M}_{n \times n}(\mathbb{R})$ be the matrix all of the entries of which equal t . Find the eigenvalues of A and, for each such eigenvalue, find the associated eigenspace.

Exercise 697 Let n be a positive integer and let F be a field. Let A be a nonsingular matrix in $\mathcal{M}_{n \times n}(F)$. Given the eigenvalues of A , find the eigenvalues of A^{-1} .

Exercise 698 Let n be a positive integer and let F be a field. Let $A = [a_{ij}] \in \mathcal{M}_{n \times n}(F)$, and let $c \in \text{spec}(A)$. If $b, d \in F$, show that $bc + d \in \text{spec}(bA + dI)$.

Exercise 699 Let $A = \begin{bmatrix} a & b \\ c & d \end{bmatrix} \in \mathcal{M}_{2 \times 2}(\mathbb{R})$. If $t \in \mathbb{R}$ is a root of the polynomial $bX^2 + (a-d)X - c \in \mathbb{R}[X]$, show that $\begin{bmatrix} 1 \\ t \end{bmatrix}$ is an eigenvector of A associated with the eigenvalue $a + bt$.

Exercise 700 Let $A \in \mathcal{M}_{2 \times 2}(\mathbb{C})$ be a matrix having two distinct eigenvalues. Show that there are precisely four distinct matrices $B \in \mathcal{M}_{2 \times 2}(\mathbb{C})$ satisfying $B^2 = A$.

Exercise 701 Find all $a \in \mathbb{R}$ such that $\begin{bmatrix} a & 0 & 0 \\ 2a & 2a & 2a \\ 0 & 0 & a \end{bmatrix}$ has a unique eigenvalue.

Exercise 702 Find a real number a such that the only eigenvalue of the

matrix $\begin{bmatrix} a & 1 & 0 \\ -1 & 0 & -1 \\ 0 & 1 & -a \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{R})$ is 0.

Exercise 703 For $1 \leq i \leq 3$ and $2 \leq j \leq 3$, find a real number a_{ij}

such that $\begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix}$, $\begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix}$, and $\begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$ are all eigenvectors of the

matrix $\begin{bmatrix} 1 & a_{12} & a_{13} \\ 1 & a_{22} & a_{23} \\ 1 & a_{32} & a_{33} \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{R})$.

Exercise 704 Let $0 \neq r \in \mathbb{C}$ and let n and m be positive integers. Let $A = [a_{ij}] \in \mathcal{M}_{n \times n}(\mathbb{C})$ be given and let $B = [b_{ij}] \in \mathcal{M}_{n \times n}(\mathbb{C})$ be the matrix defined by $b_{ij} = r^{m+i-j}a_{ij}$ for all $1 \leq i, j \leq n$. Show that if $d \in \mathbb{C}$ is an eigenvalue of A then $r^m d$ is an eigenvalue of B .

Exercise 705 Let n be a positive integer and let F be a field. A matrix $A \in \mathcal{M}_{n \times n}(F)$ is a **magic matrix** if and only if there exists a scalar $c \in F$ such that the sum of the entries in each row and each column is c . Characterize magic matrices in terms of their eigenvalues.

Exercise 706 Show that all matrices of the form $\begin{bmatrix} a-1 & -1 \\ a^2-a+1 & -a \end{bmatrix} \in \mathcal{M}_{2 \times 2}(\mathbb{C})$ have the same eigenvalues.

Exercise 707 Let $A \in \mathcal{M}_{2 \times 2}(\mathbb{C})$ be a matrix having distinct eigenvalues $a \neq b$. Show that, for all $n > 0$,

$$A^n = \frac{a^n}{a-b}(A-bI) + \frac{b^n}{b-a}(A-aI).$$

Exercise 708 Let $A \in \mathcal{M}_{2 \times 2}(\mathbb{C})$ be a matrix having a unique eigenvalue c . Show that $A^n = c^{n-1} [nA - (n-1)cI]$ for all $n > 0$.

Exercise 709 Let n be a positive integer and let $A \in \mathcal{M}_{n \times n}(\mathbb{C})$. Show that every eigenvector of A is also an eigenvector of $\text{adj}(A)$.

Exercise 710 Let n be a positive integer. Let G be the set of all matrices $A \in \mathcal{M}_{n \times n}(\mathbb{C})$ satisfying the condition that \mathbb{C}^n has a basis composed of eigenvectors of A . Is G closed under taking sums? Is it closed under taking products?

Exercise 711 Let $p(X) \in \mathbb{C}[X]$ and let $A \in \mathcal{M}_{n \times n}(\mathbb{C})$ for some positive integer n . Calculate the determinant of the matrix $p(A)$ using the eigenvalues of A .

Exercise 712 Let $-1 \neq a \in \mathbb{R}$ and let $A = \begin{bmatrix} 1-a+a^2 & 1-a \\ a-a^2 & a \end{bmatrix} \in \mathcal{M}_{2 \times 2}(\mathbb{R})$. Calculate A^n for all $n \geq 1$.

Exercise 713 Let n be a positive integer. Given a matrix $A \in \mathcal{M}_{n \times n}(\mathbb{Q})$, find infinitely-many distinct matrices having the same eigenvalues as A .

Exercise 714 Let $c \in \mathbb{R}$. Find the spectral radius of

$$A = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & c & 0 \\ 0 & -c & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \in \mathcal{M}_{4 \times 4}(\mathbb{C}).$$

Exercise 715 Let $A \in \mathcal{M}_{n \times n}(\mathbb{R})$ be a matrix all entries in which are positive and let c be a positive real number greater than the spectral radius of A . Show that $|cI - A| > 0$.

Exercise 716 Let n be a positive integer. Show that 1 is an eigenvalue of any Markov matrix in $\mathcal{M}_{n \times n}(\mathbb{R})$.

Exercise 717 Let n be a positive integer. Let $A = [a_{ij}] \in \mathcal{M}_{n \times n}(\mathbb{R})$ satisfy the condition that $\sum_{j=1}^n |a_{ij}| \leq 1$ for all $1 \leq i \leq n$. Show that $|c| \leq 1$ for all $c \in \text{spec}(A)$.

Exercise 718 Let n be a positive integer and let F be a field. Let $A = [a_{ij}] \in \mathcal{M}_{n \times n}(F)$ be a matrix satisfying the condition that the sum of

the entries in each row equals 1. Let $1 \neq c \in \text{spec}(A)$ and let $\begin{bmatrix} b_1 \\ \vdots \\ b_n \end{bmatrix}$ be an eigenvalue of A associated with c . Show that $\sum_{j=1}^n b_j = 0$.

Exercise 719 Give an example of a matrix $A \in \mathcal{M}_{2 \times 2}(\mathbb{R})$ satisfying the condition that $\text{spec}(A) = \emptyset$ but $\text{spec}(A^4) \neq \emptyset$.

Exercise 720 Find an example of matrices $A, B \in \mathcal{M}_{2 \times 2}(\mathbb{R})$ satisfying the condition that every element of $\text{spec}(A) \cup \text{spec}(B)$ is positive but every element of $\text{spec}(AB)$ is negative.

Exercise 721 Find a polynomial $p(X) \in \mathbb{C}[X]$ of degree 2 satisfying the condition that all matrices in $\mathcal{M}_{2 \times 2}(\mathbb{C})$ of the form $\begin{bmatrix} 1-a & 1 \\ p(a) & a \end{bmatrix}$, for $a \in \mathbb{C}$, have the same characteristic polynomial.

Exercise 722 Let F be a field and let n be an even positive integer. Let $A, B \in \mathcal{M}_{n \times n}(F)$ be matrices satisfying $A = B^2$. Let $p(X)$ be the characteristic polynomial of A and let $q(X)$ be the characteristic polynomial of B . Show that $p(X^2) = q(X)q(-X)$.

Exercise 723 Let F be a field. Characterize the matrices in $\mathcal{M}_{2 \times 2}(F)$ having the property that their characteristic polynomial is not equal to their minimal polynomial.

Exercise 724 Let (K, \bullet) be an associative unital F -algebra, let $v \in K$, and let $\alpha : K \rightarrow K$ be a homomorphism of F -algebras. Show that $\text{Ann}(v) \subseteq \text{Ann}(\alpha(v))$.

Exercise 725 Are the matrices $\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$ and $\begin{bmatrix} 4 & 2 \\ 3 & 1 \end{bmatrix}$ in $\mathcal{M}_{2 \times 2}(\mathbb{R})$ similar?

Exercise 726 Are the matrices $\begin{bmatrix} 1 & i & 0 \\ i & 2 & -1 \\ 0 & i & 1 \end{bmatrix}$ and $\begin{bmatrix} 1+i & 7 & 2 \\ 0 & 1 & 9 \\ 0 & 0 & 2-i \end{bmatrix}$ in $\mathcal{M}_{3 \times 3}(\mathbb{C})$ similar?

Exercise 727 Let n be a positive integer and let $A, B \in \mathcal{M}_{n \times n}(\mathbb{R})$. Show that if A and B are similar when considered as elements of $\mathcal{M}_{n \times n}(\mathbb{C})$, they are also similar in $\mathcal{M}_{n \times n}(\mathbb{R})$.

Exercise 728 Find a diagonal matrix in $\mathcal{M}_{3 \times 3}(\mathbb{R})$ similar to the matrix $\begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 1 \end{bmatrix}$.

Exercise 729 Find a diagonal matrix in $\mathcal{M}_{3 \times 3}(\mathbb{R})$ similar to the matrix

$$\begin{bmatrix} 0 & 0 & 1 \\ 0 & 0 & 0 \\ 1 & 0 & 0 \end{bmatrix}.$$

Exercise 730 Is there a diagonal matrix in $\mathcal{M}_{3 \times 3}(\mathbb{R})$ similar to the

$$\text{matrix } \begin{bmatrix} 8 & 3 & -3 \\ -6 & -1 & 3 \\ 12 & 6 & -4 \end{bmatrix}?$$

Exercise 731 Show that every matrix in the subspace of $\mathcal{M}_{2 \times 2}(\mathbb{R})$ generated by $\left\{ \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}, \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix} \right\}$ is similar to a diagonal matrix.

Exercise 732 Determine if the matrices $\begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{bmatrix}$ and $\begin{bmatrix} 0 & 0 & 0 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}$ in $\mathcal{M}_{3 \times 3}(GF(5))$ are similar.

Exercise 733 Is there a diagonal matrix in $\mathcal{M}_{3 \times 3}(\mathbb{R})$ similar to the

$$\text{matrix } \begin{bmatrix} 1 & -1 & 1 \\ -2 & 1 & 2 \\ -2 & -1 & 4 \end{bmatrix}?$$

Exercise 734 Let $A = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 1 \\ 0 & 1 & 1 \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{R})$. Find a nonsingular matrix $P \in \mathcal{M}_{3 \times 3}(\mathbb{R})$ such that $P^{-1}AP$ is a diagonal matrix.

Exercise 735 Show that the matrix $A = \begin{bmatrix} 1 & i \\ i & -1 \end{bmatrix} \in \mathcal{M}_{2 \times 2}(\mathbb{C})$ is not similar to a diagonal matrix.

Exercise 736 Are the matrices $\begin{bmatrix} 1 & -1 & 0 \\ 0 & 2 & 5 \\ 0 & 0 & 3 \end{bmatrix}$ and $\begin{bmatrix} 2 & 0 & 0 \\ -1 & 4 & 0 \\ 0 & 3 & 7 \end{bmatrix}$ in $\mathcal{M}_{3 \times 3}(\mathbb{Q})$ similar?

Exercise 737 Let k and n be positive integers and let F be a field.

Let $A \in \mathcal{M}_{k \times n}(F)$ and $B \in \mathcal{M}_{n \times k}(F)$. Is the matrix $\begin{bmatrix} AB & O \\ B & O \end{bmatrix}$

similar to the matrix $\begin{bmatrix} O & O \\ B & BA \end{bmatrix}$?

Exercise 738 Let F be a field and let $A = \begin{bmatrix} a & 1 & 0 \\ 0 & a & 1 \\ 0 & 0 & a \end{bmatrix} \in \mathcal{M}_{3 \times 3}(F)$. For

any $p(X) \in F[X]$, show that $p(A) = \begin{bmatrix} p(a) & p'(a) & \frac{1}{2}p''(a) \\ 0 & p(a) & p'(a) \\ 0 & 0 & p(a) \end{bmatrix}$, where $p'(X)$ denotes the formal derivative of the polynomial $p(X)$ and $p''(X)$ is the formal derivative of $p'(X)$.

Exercise 739 Find the characteristic and minimal polynomials of the ma-

trix $\begin{bmatrix} 7 & 4 & -4 \\ 4 & -8 & -1 \\ -4 & -1 & -8 \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{R})$.

Exercise 740 Let n be a positive integer. Let V be the vector space over \mathbb{R} consisting of all polynomial functions from \mathbb{R} to itself having degree at most n . Let α be the endomorphism of V which assigns to each $f \in V$ its derivative, and let A be a matrix representing α with respect to some basis of V . Find the minimal polynomial of A .

Exercise 741 Find six distinct matrices in $\mathcal{M}_{2 \times 2}(\mathbb{R})$ which annihilate the polynomial $X^2 - 1$.

Exercise 742 Let n be a positive integer and let c be an element of a field F . Find a matrix $A \in \mathcal{M}_{n \times n}(F)$ having minimal polynomial $(X - c)^n$.

Exercise 743 Use the Cayley-Hamilton Theorem to find the inverse of

$\begin{bmatrix} 5 & 1 & -1 \\ -6 & 0 & 2 \\ 0 & 0 & 2 \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{R})$.

Exercise 744 Let n be a positive integer and let $A \in \mathcal{M}_{n \times n}(F)$ be a matrix of rank h . Show that the degree of the minimal polynomial of A is at most $h + 1$.

Exercise 745 Let n be a positive integer and let F be a field. Show that a matrix $A \in \mathcal{M}_{n \times n}(F)$ is nonsingular if and only if $m_A(0) \neq 0$.

Exercise 746 Find the eigenvalues of the matrix $\begin{bmatrix} 0 & 0 & 1 & 1 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ 1 & 1 & 0 & 0 \end{bmatrix} \in$

$\mathcal{M}_{4 \times 4}(\mathbb{Q})$ and determine the algebraic multiplicity of each.

Exercise 747 Find the minimal polynomial of the matrix $\begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 3 & -3 \end{bmatrix}$ in $\mathcal{M}_{3 \times 3}(\mathbb{R})$.

Exercise 748 Let α and β be the endomorphisms of \mathbb{Q}^4 represented

with respect to the canonical basis by the matrices $\begin{bmatrix} 1 & 0 & -1 & 0 \\ 0 & 1 & 0 & -1 \\ 1 & 0 & -1 & 0 \\ 0 & 1 & 0 & -1 \end{bmatrix}$

and $\begin{bmatrix} 4 & -1 & -1 & 0 \\ -1 & 4 & 0 & -1 \\ 1 & 0 & 2 & -1 \\ 0 & 1 & -1 & 2 \end{bmatrix}$ respectively. Does there exist a basis of \mathbb{Q}^4

with respect to which both of them can be represented by diagonal matrices?

Exercise 749 Let α be the endomorphisms of \mathbb{R}^3 represented with

respect to the canonical basis by the matrix $\begin{bmatrix} -6 & 2 & -5 \\ 4 & 4 & -2 \\ 10 & -3 & 8 \end{bmatrix}$. Calculate the algebraic and geometric multiplicities of each of the eigenvalues of α .

Exercise 750 Let $A = [a_{ij}] \in \mathcal{M}_{n \times n}(\mathbb{C})$ be a symmetric tridiagonal matrix having an eigenvalue c with algebraic multiplicity k . Show that $a_{i-1,i} = 0$ for at least $k-1$ values of i .

Exercise 751 Let α be the endomorphisms of \mathbb{R}^3 represented with

respect to the canonical basis by the matrix $\begin{bmatrix} -8 & -13 & -14 \\ -6 & -5 & -8 \\ 14 & 17 & 21 \end{bmatrix}$. Does

there exist a basis of \mathbb{R}^3 with respect to which α can be represented by a diagonal matrix?

Exercise 752 Let $A = \begin{bmatrix} 17 & -8 & -12 & 14 \\ 46 & -22 & -35 & 41 \\ -2 & 1 & 4 & -4 \\ 4 & -2 & -2 & 3 \end{bmatrix} \in \mathcal{M}_{4 \times 4}(\mathbb{Q})$. Find the minimal polynomial A .

Exercise 753 For each $t \in \mathbb{R}$, set $A(t) = \begin{bmatrix} \cos^2(t) & \cos(t)\sin(t) \\ \cos(t)\sin(t) & \sin^2(t) \end{bmatrix} \in \mathcal{M}_{2 \times 2}(\mathbb{R})$. Show that all of these matrices have the same characteristic and minimal polynomials.

Exercise 754 Let $a, b, c \in \mathbb{C}$. Find a necessary and sufficient condition for the minimal polynomial of $\begin{bmatrix} 2 & 0 & 0 \\ a & 2 & 0 \\ b & c & 1 \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{C})$ to be equal to $(X - 1)(X - 2)$.

Exercise 755 Let $A = \begin{bmatrix} 1 & a & 0 \\ a & a & 1 \\ a & a & -1 \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{R})$. Find the set of all real numbers a for which the minimal and characteristic polynomials of A are equal.

Exercise 756 Let $F = GF(5)$. For which values of $a, b \in F$ are the characteristic polynomial and minimal polynomial of the matrix

$$\begin{bmatrix} a & b & 4 & 2 & 0 \\ b & b & b & 3 & 3 \\ 3 & 4 & 2b & 1 & 3 \\ 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 3b & 0 \end{bmatrix}$$

equal? What if $F = GF(7)$?

Exercise 757 Let F be a field and let $O \neq A \in \mathcal{M}_{3 \times 3}(F)$ be a matrix satisfying $A^k = O$ for some positive integer k . Show that $A^3 = O$.

Exercise 758 Let F be a field and let $A \in \mathcal{M}_{3 \times 3}(F)$ be a matrix which can be written in the form BC , where B and C are matrices in $\mathcal{M}_{3 \times 3}(F)$ satisfying $B^2 = I = C^2$. Show that A is nonsingular and similar to A^{-1} .

Exercise 759 Let $A \in \mathcal{M}_{3 \times 3}(\mathbb{Q})$ be a matrix satisfying the condition that $A^5 = I$. Show that $A = I$.

Exercise 760 Let n be a positive integer and let α be an endomorphism of $\mathcal{M}_{n \times n}(\mathbb{C})$, considered as a vector space over \mathbb{C} , which satisfies the condition that $\alpha(A)$ is nonsingular if and only if A is nonsingular. Show that α is an automorphism.

Exercise 761 Let F be a field and let n be a positive integer. Let $A \in \mathcal{M}_{n \times n}(F)$ be a matrix having characteristic polynomial $p(X) = X^n + \sum_{i=0}^{n-1} c_i X^i$. Show that, for each $k \geq n$, we have $A^k = \sum_{j=0}^{n-1} b_j(k) A^j$, where

- (1) $b_j(n) = -c_j$ for all $0 \leq j \leq n-1$;
- (2) $b_{-1}(k) = 0$ for all $k \geq n$;
- (3) $b_j(k+1) = b_{j-1}(k) - a_j b_{n-1}(k)$ for all $k \geq n$ and all $0 \leq j \leq n-1$.

Exercise 762 Show that there is no matrix $A \in \mathcal{M}_{2 \times 2}(\mathbb{R})$ satisfying the condition that $A^2 = \begin{bmatrix} -1 & 0 \\ 0 & -c \end{bmatrix}$, where $c > 1$.

Exercise 763 Let V be a vector space finitely generated over \mathbb{C} and let $\alpha \in \text{End}(V)$ be diagonalizable. If W is a nontrivial subspace of V invariant under α , is the restriction of α to W necessarily diagonalizable?

Exercise 764 Find all rational numbers a satisfying the condition the endomorphism of \mathbb{Q}^3 represented with respect to some basis by the matrix $\begin{bmatrix} 1 & 0 & 0 \\ 1 & a & 0 \\ 0 & 0 & 1 \end{bmatrix}$ is diagonalizable.

Exercise 765 Let n be a positive integer and let $B \in \mathcal{M}_{n \times n}(\mathbb{R})$ be a matrix all entries of which are positive. Let $r > \rho(B)$. Show that

- (1) All the matrix $A = rI - B$ is nonsingular;
- (2) All nondiagonal entries of A are nonpositive;
- (3) All entries of A^{-1} are nonnegative; and
- (4) If $a + bi \in \mathbb{C}$ is an eigenvalue of A , then $a > 0$.

Exercise 766 Let V be a vector space of finite odd dimension over \mathbb{R} and let $\alpha_1, \dots, \alpha_k$ be distinct mutually-commuting endomorphisms of V , for some $k > 1$. Show that these endomorphisms have a common eigenvector.

Exercise 767 Show that the endomorphisms $\alpha : \begin{bmatrix} a \\ b \\ c \\ d \end{bmatrix} \mapsto \begin{bmatrix} 2b \\ 2a \\ 2d \\ 2c \end{bmatrix}$ and

$\beta : \begin{bmatrix} a \\ b \\ c \\ d \end{bmatrix} \mapsto \begin{bmatrix} c \\ d \\ a \\ b \end{bmatrix}$ of \mathbb{Q}^4 are diagonalizable and commute. Find a

basis of \mathbb{Q}^4 relative to which both α and β are represented by diagonal matrices.

Exercise 768 Let $A \in \mathcal{M}_{3 \times 3}(\mathbb{Q})$ have characteristic polynomial $X^3 - bX^2 + cX - d$. For all $n \geq 3$, show that

$$A^n = t_{n-1}A + t_{n-2}\text{adj}(A) + (t_n - bt_{n-1})I,$$

where $t_n = \sum_{2i+3j \leq n} (-1)^i \binom{i+j}{j} \binom{n-i-2j}{i+j} b^{n-2i-3j} c^i d^j$.

13

Krylov subspaces

Let V be a vector space over a field F and let $\alpha \in \text{End}(V)$. If $0_V \neq v_0 \in V$ then the subspace $F\{v_0, \alpha(v_0), \alpha^2(v_0), \dots\}$ of V is called the **Krylov¹ subspace** of V defined by α and v_0 . The elements of this subspace are precisely those vectors in V of the form $p(\alpha)(v_0)$, where $p(X) \in F[X]$, and so it is natural to denote it by $F[\alpha]v_0$. It is clear that $F[\alpha]v_0$ is invariant under α . We claim that $\dim(F[\alpha]v_0) = 1$ if and only if v_0 is an eigenvector of α . Indeed, if v_0 is an eigenvector of α associated with an eigenvalue c then for each $p(X) = \sum_{j=0}^k a_j X^j \in F[X]$ we have $p(\alpha)(v_0) = \sum_{j=0}^k a_j \alpha^j(v_0) = \sum_{j=0}^k a_j c^j v_0 \in Fv_0$, proving that $F[\alpha]v_0 = Fv_0$ and so $\dim(F[\alpha]v_0) = 1$. Conversely, assume that $\dim(F[\alpha]v_0) = 1$. Then $F[\alpha]v_0 = Fv_0$ since Fv_0 is a one-dimensional subspace of $F[\alpha]v_0$. In particular, $\alpha(v_0) \in Fv_0$ and so there exists a scalar c such that $\alpha(v_0) = cv_0$, which proves that v_0 is an eigenvector



¹ **Alexei Nikolaevich Krylov** was a Russian applied mathematician who, at the end of the 19th century, developed many of the methods mentioned here in connection with the solution of differential equations.

of α . Note that in any case the set $\{v_0, \alpha(v_0), \alpha^2(v_0), \dots\}$ is a generating set for $F[\alpha]v_0$ over F and so $\dim(F[\alpha]v_0)$ can be used to measure how “far” v_0 is from being an eigenvector of α .

(13.1) Proposition: Let V be a vector space over a field F and let $\alpha \in \text{End}(V)$. If $0_V \neq v_0 \in V$ then $F[\alpha]v_0$ is the intersection of all subspaces of V containing v_0 and invariant under α .

Proof: Since $F[\alpha]v_0$ contains v_0 and is invariant under α , it certainly contains the intersection of all such subspaces of V . Conversely, if W is a subspace of V which contains v_0 and is invariant under α , then $p(\alpha)(v_0) \in W$ for all $p(X) \in F[X]$ and so $F[\alpha]v_0 \subseteq W$. Thus we have the desired equality. \square

As a first example of the use to which we can put Krylov subspaces, we will see how to use the minimal polynomial to solve systems of linear equations. Let V be a vector space over a field F and let V^∞ be the space of all infinite sequences of elements of V . Every polynomial $p(X) = \sum_{j=0}^k a_j X^j \in F[X]$ defines an endomorphism θ_p of V^∞ by

$$\theta_p : [v_0, v_1, \dots] \mapsto \left[\sum_{j=0}^k a_j v_j, \sum_{j=0}^k a_j v_{j+1}, \sum_{j=0}^k a_j v_{j+2}, \dots \right].$$

Note that if $p(X) = c$ is a polynomial of degree no greater than 0, then $\theta_p = \sigma_c$. It is also easy to verify that $\theta_{pq} = \theta_p \theta_q = \theta_q \theta_p$ for all $p(X), q(X) \in F[X]$.

A sequence $y \in V^\infty$ is **linearly recurrent** if and only if there exists a polynomial $p(X) \in F[X]$ with $y \in \ker(\theta_p)$. In this case, we say that $p(X)$ is a **characteristic polynomial** of y . If $p(X) \in F[X]$ is a characteristic polynomial of $y \in V^\infty$ and if $q(X) \in F[X]$ is a characteristic polynomial of $z \in V^\infty$ then $\theta_{pq}(y + z) = \theta_q \theta_p(y) + \theta_p \theta_q(z) = [0, 0, \dots]$ and so $p(X)q(X)$ is a characteristic polynomial of $y + z$. It is also clear that $p(X)$ is a characteristic polynomial of cy for all $c \in F$. Thus we see that the set of all linearly recurrent sequences in V^∞ is a subspace of V^∞ , which we will denote by $LR(V)$. If $y \in LR(V)$, there is precisely one characteristic polynomial which is monic and of minimal degree. This polynomial will be called the **minimal polynomial** of y . The degree of the minimal polynomial of y will be called the **order of recurrence** of y .

Example: Let F be a field, let n be a positive integer, and let $V = \mathcal{M}_{n \times n}(F)$. If $A \in V$ then a polynomial $p(X) \in F[X]$ is a characteristic [resp. minimal] polynomial of the sequence $[I, A, A^2, \dots]$ if and only if it was the characteristic [minimal] polynomial of A in the sense of the previous chapter.

Example: Let $V = F = \mathbb{Q}$ and let $y = [a_0, a_1, \dots] \in V^\infty$ be the sequence defined by $a_0 = 0$, $a_1 = 1$, and $a_{i+2} = a_{i+1} + a_i$ for all $i \geq 0$. This sequence is called the **Fibonacci sequence**. Its minimal polynomial is $X^2 - X - 1$. The roots of this polynomial are $\frac{1}{2}(1 \pm \sqrt{5})$. The number $\frac{1}{2}(1 + \sqrt{5})$ is called the **golden ratio** and artists consider rectangles the sides of which are related by the golden ratio to be of high aesthetic value. This ratio – which appears in ancient Egyptian and Babylonian texts – appears in nature and is basic in the analysis of certain patterns of growth in nature (such as the spirals of a snail shell or a sunflower), of Greek architecture, of Renaissance painting², and even such modern designs as the ratio of the dimensions of a credit card or of A4 paper. Notice that

$X^2 - X - 1$ is also the characteristic polynomial of the matrix $\begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix} \in \mathcal{M}_{2 \times 2}(\mathbb{R})$, and so the eigenvalues of this matrix are also precisely $\frac{1}{2}(1 \pm \sqrt{5})$.

The eigenspace associated with $\frac{1}{2}(1 + \sqrt{5})$ is $\mathbb{R} \begin{bmatrix} \frac{1}{2}(1 + \sqrt{5}) \\ 1 \end{bmatrix}$ and the eigenspace associated with $\frac{1}{2}(1 - \sqrt{5})$ is $\mathbb{R} \begin{bmatrix} \frac{1}{2}(1 - \sqrt{5}) \\ 1 \end{bmatrix}$.

We note that if $V = F$ and if $y \in LR(F)$ is a sequence having order of recurrence at most n , then there exist algorithms, which are essentially extensions of the euclidean algorithm, to calculate the coefficients of the minimal polynomial of y in an order of n^2 arithmetic operations in F .

Now let V be a vector space of finite dimension n over a field F and let α be an automorphism of V having minimal polynomial $p(X) \in F[X]$. If $w \in V$ then the sequence $y = [w, \alpha(w), \alpha^2(w), \dots]$ belongs to $\ker(\theta_p)$ and hence to $LR(V)$. Therefore this sequence has a minimal polynomial $q(X) = \sum_{j=0}^d c_j X^j$, which divides the polynomial $p(X)$ in $F[X]$. Since α is an automorphism, we can assume that $c_0 \neq 0$ and so we see that if $u = -c_0^{-1} \sum_{j=1}^d c_j \alpha^{j-1}(w)$ then $\alpha(u) = w$ and so



² **Leonardo Fibonacci** was born in Italy in the 12th century and educated in Tunis, bringing back the fruits of Arab mathematics to Europe. His book *Liber Abaci*, written in 1202, contained the first new mathematical research in Christian Europe in over 1000 years. In 1509, **Fra Luca Pacioli**, one of the most important Renaissance mathematicians, wrote a book, *The Divine Proportion*, illustrated by his friend Leonardo da Vinci, about the golden ratio.

$u = \alpha^{-1}(w)$. In particular, if $V = F^n$ for some positive integer n and if α is represented by a matrix A with respect to the canonical basis, then $u = -c_0^{-1} \sum_{j=1}^d c_j A^{j-1} w$ is the unique solution of the system of linear equations $AX = w$. If we set $q^*(X) = -c_0^{-1} \sum_{j=1}^d c_j X^{j-1}$ then $u = q^*(A)w$, and this could be computed quickly were we to already know $q(X)$.

How does one calculate $q^*(X)$ in practice? One method in use is basically probabilistic: we randomly choose a vector $u \in F^n$ and compute the minimal polynomial $q_u(X)$ of the sequence

$$y_u = [u \odot w, u \odot (Aw), u \odot A^2 w, \dots]$$

in F^∞ , something which can be done, as we have already observed, in an order of n^2 arithmetic operations in F . After that, we check whether the minimal polynomial of y_u is also the minimal polynomial of y . In general, it will not be so, but it will divide the minimal polynomial of y and so after a reasonable number of such attempts we will, usually, have enough information on hand to reconstruct the minimal polynomial of y .

Example: Let $F = GF(5)$, let $A = \begin{bmatrix} 1 & 4 & 4 \\ 4 & 0 & 3 \\ 1 & 2 & 4 \end{bmatrix} \in \mathcal{M}_{3 \times 3}(F)$, and let $w = \begin{bmatrix} 3 \\ 1 \\ 2 \end{bmatrix} \in F^3$. The sequence $w, Aw, A^2 w, \dots$ looks like

$$\begin{bmatrix} 3 \\ 1 \\ 2 \end{bmatrix}, \begin{bmatrix} 0 \\ 3 \\ 3 \end{bmatrix}, \begin{bmatrix} 4 \\ 4 \\ 3 \end{bmatrix}, \begin{bmatrix} 2 \\ 0 \\ 4 \end{bmatrix}, \begin{bmatrix} 3 \\ 0 \\ 3 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \dots$$

If we choose $u = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$ we obtain the sequence $y_u = [3, 0, 4, 2, 3, 0, \dots]$

in F^∞ and the minimal polynomial $q_u(X)$ of this sequence equals

$X^2 + 2X + 2$. Since $q_u(A)w = \begin{bmatrix} 0 \\ 2 \\ 3 \end{bmatrix}$, we see that this polynomial is

not the minimal polynomial of y . We will try again with $u = \begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix}$.

For this choice, we get $y_u = [0, 1, 2, 2, 3, 2, \dots]$ and this has minimal polynomial $X^3 + 3X + 1$. Since the minimal polynomial of y has to be a multiple of this polynomial, and has to be of degree 3, it must equal

$X^3 + 3X + 1$ and, indeed, $q_u(A)w = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$. Therefore $q^*(X) = X^2 + 3$

and $q^*(A)w = q_u(A)w = \begin{bmatrix} 2 \\ 3 \\ 1 \end{bmatrix}$.

Now let us return to an important problem which was considered in the previous chapter. Let V be a vector space finitely generated over a field F . Given an endomorphism $\alpha \in \text{End}(V)$, how can we find a basis of V relative to which α is represented by a matrix which is as nice as possible? We have already found out when α is diagonalizable. But what if α is not diagonalizable? Given a vector $0_V \neq w \in V$, there exists a positive integer k such that the set $\{w, \alpha(w), \dots, \alpha^{k-1}(w)\}$ is linearly independent but the set $\{w, \alpha(w), \dots, w^k(w)\}$ is linearly dependent. Then $\{w, \alpha(w), \dots, \alpha^{k-1}(w)\}$ is a basis for the Krylov subspace $F[\alpha]w$ of V , which is called the **canonical basis** of this subspace. The restriction of α to $F[\alpha]w$ is represented by a matrix of the form

$$\begin{bmatrix} 0 & 0 & \dots & 0 & c_1 \\ 1 & 0 & \dots & 0 & c_2 \\ 0 & 1 & \dots & 0 & c_3 \\ & & \dots & & \\ 0 & \dots & 1 & 0 & c_{k-1} \\ 0 & \dots & 0 & 1 & c_k \end{bmatrix} \quad \text{with respect to the canonical basis, where the}$$

scalars c_1, \dots, c_k satisfy $\alpha^k(w) = \sum_{i=1}^k c_i \alpha^{i-1}(w)$. This, of course, is just the companion matrix of the polynomial $X^k - \sum_{i=1}^k c_i X^{i-1}$.

Krylov subspaces are also the basis for a family of iterative algorithms, known as **Krylov algorithms**, used for approximating solutions to systems of equations of the form $AX = w$, where $A \in \mathcal{M}_{n \times n}(\mathbb{R})$ or $A \in \mathcal{M}_{n \times n}(\mathbb{C})$. Similarly, Krylov subspaces are a basis for a family of iterative algorithms, known as **Lanczos³ algorithms**, used for approximating eigenvalues of sufficiently-nice (e.g. symmetric) matrices. Such algorithms



3

Hungarian-born applied mathematician **Cornelius Lanczos** developed many important numerical methods for computers in the period after World War II, while working at the US National Bureau of Standards and, later, at the University of Dublin in Ireland.

work even under the assumption that we don't even have direct access to the entries of A but do have a "black box" ability to compute Av or $A^T v$ for any given vector $v \in \mathbb{R}^n$. Of course, they do not work for all matrices, but when they work they tend to be fairly efficient and rapid, and are especially good for large sparse matrices. Moreover, they are also amenable to implementation on parallel computers.

Let V be a vector space over a field F . An endomorphism $\alpha \in \text{End}(V)$ is **nilpotent** if and only if there exists a positive integer k satisfying $\alpha^k = \sigma_0$. The smallest such integer k , if one exists, is called the **index of nilpotence** of the endomorphism.

Example: Let F be a field and let α be the endomorphism of F^3 defined by $\alpha : \begin{bmatrix} a \\ b \\ c \end{bmatrix} \mapsto \begin{bmatrix} 0 \\ a \\ b \end{bmatrix}$. Then α is a nilpotent endomorphism, having index of nilpotence 3. The endomorphism β of F^3 defined by $\beta : \begin{bmatrix} a \\ b \\ c \end{bmatrix} \mapsto \begin{bmatrix} -a + 2b + c \\ 0 \\ -a + 2b + c \end{bmatrix}$ is a nilpotent endomorphism, having index of nilpotence 2.

Example: Let F be a field and let α and β be the endomorphisms of F^2 defined by $\alpha : \begin{bmatrix} a \\ b \end{bmatrix} \mapsto \begin{bmatrix} b \\ 0 \end{bmatrix}$ and $\beta : \begin{bmatrix} a \\ b \end{bmatrix} \mapsto \begin{bmatrix} 0 \\ a \end{bmatrix}$. Both endomorphisms are nilpotent, but $\alpha + \beta$ is clearly not nilpotent.

If α is a nilpotent endomorphism of a vector space V and $w \in V \setminus \ker(\alpha)$ then the restriction of α to $F[\alpha]w$ is represented with respect to the canonical basis of $F[\alpha]w$ by a matrix of the form

$$\begin{bmatrix} 0 & 0 & 0 & \cdots & 0 & 0 \\ 1 & 0 & 0 & \cdots & 0 & 0 \\ 0 & 1 & 0 & \cdots & 0 & 0 \\ & & & \cdots & & \\ 0 & 0 & 0 & \cdots & 0 & 0 \\ 0 & 0 & 0 & \cdots & 1 & 0 \end{bmatrix}.$$

(13.2) Proposition: Let V be a vector space over a field F and let α be a nilpotent endomorphism of V having index of nilpotence k . Then there exists a vector $w \in V$ satisfying the condition that $\dim(F[\alpha]w) = k$.

Proof: We know that $\alpha^k = \sigma_0$ but that there exists a vector $0_V \neq w \in V$ such that $\alpha^{k-1}(w) \neq 0_V$. We will have proven the theorem should we be able to show that the set $\{w, \alpha(w), \dots, \alpha^{k-1}(w)\}$ is linearly independent. And, indeed, assume that we have scalars $a_0, \dots, a_{k-1} \in F$ satisfying $\sum_{i=0}^{k-1} a_i \alpha^i(w) = 0_V$. Let t be the smallest index such that $a_t \neq 0$. Then if we apply the endomorphism α^{k-t-1} to $\sum_{i=0}^{k-1} a_i \alpha^i(w)$ we get $0_V = a_t \alpha^{k-1}(w) + a_{t+1} \alpha^k(w) + \dots + a_{k-2} \alpha^{2k-t-2}(w)$ and so $a_t = 0$, which is a contradiction. Therefore we conclude that $a_i = 0$ for all i , and so the set is linearly independent, as required. \square

In particular, we see that if V is a vector space of finite dimension over a field F and if α is a nilpotent endomorphism of V , then the index of nilpotence of α is no greater than $\dim(V)$.

(13.3) Proposition: Let V be a vector space finitely generated over a field F and let α be a nilpotent endomorphism of V having index of nilpotence k . If $w \in V$ satisfies the condition that $\dim(F[\alpha]w) = k$ then the subspace $F[\alpha]w$ of V has a complement in V which is invariant under α .

Proof: We will proceed by induction on k . If $k = 1$ then $\alpha = \sigma_0$ and so $F[\alpha]w = Fw$. Then there is a subset B of $V \setminus \{Fw\}$ such that $B \cup \{w\}$ is a basis for V , and B is a basis for a complement of Fw in V . Thus we can assume that $k > 1$ and that the result has been established for any vector space finitely generated over F and any nilpotent endomorphism of that space having index of nilpotence less than k .

We know that $\text{im}(\alpha)$ is invariant under α and that the restriction of α to $\text{im}(\alpha)$ is nilpotent, having index of nilpotence $k-1$. We know that the set $\{w, \alpha(w), \dots, \alpha^{k-1}(w)\}$ forms a basis for $F[\alpha]w$ and so the set $\{\alpha(w), \dots, \alpha^{k-1}(w)\}$ forms a basis for the image U of $F[\alpha]w$ under α . Therefore, $U = F[\alpha]\alpha(w)$ is a subspace of $\text{im}(\alpha)$ and, by the induction hypothesis, it has a complement W_2 in $\text{im}(\alpha)$ invariant under α .

Let $W_0 = \{v \in V \mid \alpha(v) \in W_2\}$. This is a subspace of V containing W_2 , since W_2 is invariant under α . But $\alpha(v) \in W_2 \subseteq W_0$ for all $v \in W_0$ and so W_0 is also invariant under α .

Our first assertion is that $V = F[\alpha]w + W_0$. And, indeed, if $x \in V$ then $\alpha(x) \in \text{im}(\alpha) = U \oplus W_2$ and so $\alpha(x) = u + w_2$, where $u \in U$ and $w_2 \in W_2$. But $u = \alpha(y)$ for some $y \in F[\alpha]w$ and $x = y + (x - y)$. The first summand belongs to $F[\alpha]w$, whereas, as to the second, we have $\alpha(x - y) = \alpha(x) - \alpha(y) = \alpha(x) - u = w_2 \in W_2$ and so $x - y \in W_0$, proving the assertion.

Our second assertion is that $F[\alpha]w \cap W_0 \subseteq U$. Indeed, if $x \in F[\alpha]w \cap W_0$ then $\alpha(x) \in U \cap W_2 = \{0_V\}$ and so $x \in \ker(\alpha)$. Since $x \in F[\alpha]w$, we

know that there exist scalars a_0, \dots, a_{k-1} such that $x = \sum_{i=0}^{k-1} a_i \alpha^i(w)$ and hence $0_V = \alpha(x) = \sum_{i=0}^{k-2} a_i \alpha^{i+1}(w)$, which implies that $a_0 = \dots = a_{k-2} = 0$. Therefore $x = a_{k-1} \alpha^{k-1}(w) \in U$, proving the second assertion.

In particular, from what we have seen, we deduce that the subspaces W_2 and $F[\alpha]w \cap W_0$ are disjoint. Therefore $W_2 \oplus (F[\alpha]w \cap W_0)$ is a subspace of W_0 . This subspace has a complement W_1 in W_0 . Thus we have $W_0 = W_1 \oplus W_2 \oplus (F[\alpha]w \cap W_0)$.

Our third assertion is that $W = W_1 \oplus W_2$ is a complement of $F[\alpha]w$ in V which is invariant under α , and should we prove this, we will have proven the proposition. Indeed, we immediately note that $\alpha(W) \subseteq \alpha(W_0) \subseteq W_2 \subseteq W$ and so W is surely invariant under α . Moreover, $F[\alpha]w \cap W = \{0_V\}$ since this subspace is contained in the intersection of W and $F[\alpha]w \cap W_0$, which, by the choice of W , equals $\{0_V\}$. Finally,

$$\begin{aligned} V &= F[\alpha]w + W_0 = F[\alpha]w + [W_1 + W_2 + (F[\alpha]w \cap W_0)] \\ &= F[\alpha]w + W_1 + W_2 = F[\alpha]w + W \end{aligned}$$

and so $V = F[\alpha]w \oplus W$. □

(13.4) Proposition (Rational Decomposition Theorem): Let V be a vector space of finite dimension n over a field F let α be a nilpotent endomorphism of V having index of nilpotence k . Then there exist natural numbers $k = k_1 \geq \dots \geq k_t$ satisfying $k_1 + \dots + k_t = n$, and there exist vectors v_1, \dots, v_t in V such that $\{v_1, \alpha(v_1), \dots, \alpha^{k_1-1}(v_1), v_2, \alpha(v_2), \dots, \alpha^{k_2-1}(v_2), \dots, v_t, \alpha(v_t), \dots, \alpha^{k_t-1}(v_t)\}$ forms a basis for V . The matrix which represents

α with respect to this basis is of the form
$$\begin{bmatrix} A_1 & O & \dots & O \\ O & A_2 & \dots & O \\ & & \dots & \\ O & O & \dots & A_t \end{bmatrix},$$

where each A_i is of the form
$$\begin{bmatrix} 0 & 0 & \dots & 0 & 0 \\ 1 & 0 & \dots & 0 & 0 \\ 0 & 1 & \dots & 0 & 0 \\ & & \dots & & \\ 0 & 0 & \dots & 1 & 0 \end{bmatrix} \quad \text{in } \mathcal{M}_{k_i \times k_i}(F).$$

Proof: Choose $k_1 = k$ and choose v_1 to be any vector not in $\ker(\alpha^{k_1-1})$. Then $U_1 = F[\alpha]v_1$ has a basis $\{v_1, \alpha(v_1), \dots, \alpha^{k_1-1}(v_1)\}$. This subspace is invariant under α and of dimension k_1 . By Proposition 13.3, we know that we can write $V = U_1 \oplus W_1$, where W_1 is a subspace of V which is also invariant under α . The restriction of α to W_1 is a nilpotent endomorphism of W_1 having index of nilpotence $k_2 \leq k_1$. We now return on the above procedure for W_1 . We pick an element $v_2 \in W_1 \setminus \ker(\alpha^{k_2-1})$. Then the subspace $U_2 = F[\alpha]v_2$ of W_1 has a

basis $\{v_2, \alpha(v_2), \dots, \alpha^{k_2-1}(v_2)\}$. This subspace is invariant under α and of dimension k_2 . Moreover, we can write $W_1 = U_2 \oplus W_2$, where W_2 is invariant under α . Continuing in this manner, we end up with a decomposition $V = U_1 \oplus \dots \oplus U_t$, where each U_i is a subspace of V invariant under α having a basis of the form $\{v_i, \alpha(v_i), \dots, \alpha^{k_i-1}(v_i)\}$ as above. This proves the first contention of the proposition. The second one follows since $U_i = F[\alpha]v_i$ for all i , which leads to a matrix of the desired form. \square

A matrix of the form given in Proposition 13.4 is called a representation of the nilpotent endomorphism α in **Jordan⁴ canonical form**. Let V be a vector space over a field F and let α be an endomorphism of V having an eigenvalue c . A vector $0_V \neq v \in V$ is a **generalized eigenvector** of α associated with c of **degree** $k > 0$ if and only if v is in $\ker((\alpha - c\sigma_1)^k) \setminus \ker((\alpha - c\sigma_1)^{k-1})$. Thus, in particular, the eigenvectors of α associated with c , in the previous sense, are just the generalized eigenvectors of α of degree 1 associated with c .

Example: Let α be the endomorphism of \mathbb{R}^4 represented with respect to the canonical basis by the matrix
$$\begin{bmatrix} 2 & -2 & 1 & 1 \\ 0 & 1 & 1 & 1 \\ 0 & 0 & 2 & 1 \\ 0 & 0 & 0 & 2 \end{bmatrix}.$$
 This endomorphism has an eigenvector $\begin{bmatrix} 2 \\ 1 \\ 0 \\ 0 \end{bmatrix}$ associated with the eigenvalue 1 and an eigenvector $\begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}$ associated with the eigenvalue 2. It also has a generalized eigenvector $\begin{bmatrix} 0 \\ -1 \\ -1 \\ 0 \end{bmatrix}$ of degree 2 associated with the eigenvalue 2



⁴ The 19th-century French mathematician **Camille Jordan** made major contributions to linear algebra, group theory, the theory of finite fields, and the beginnings of topology.

and a generalized eigenvector $\begin{bmatrix} 0 \\ -1 \\ -1 \\ -1 \end{bmatrix}$ of degree 3 associated with the eigenvalue 2.

We now prove a generalization of Proposition 12.1.

(13.5) Proposition: Let V be a vector space over a field F and let $\alpha \in \text{End}(V)$ have an eigenvalue c . Then the set of all generalized eigenvectors of α (of all degrees) associated with c , together with 0_V , forms a subspace of V .

Proof: Let $a \in F$ and let $v, w \in V$ be generalized eigenvectors of α associated with c , of degrees k and h respectively. Then both v and w belong to $\ker(\alpha - c\sigma_1)^{h+k}$ and hence the same is true for $v + w$ and av . This means that there exist positive integers $s, t \leq h + k$ such that $v + w \in \ker((\alpha - c\sigma_1)^s) \setminus \ker((\alpha - c\sigma_1)^{s-1})$ and $av \in \ker((\alpha - c\sigma_1)^t) \setminus \ker((\alpha - c\sigma_1)^{t-1})$, which is what we needed. \square

Let V be a vector space over a field F and let $\alpha \in \text{End}(V)$ have an eigenvalue c . The subspace of V defined in Proposition 13.5 is called the **generalized eigenspace** of α associated with c .

(13.6) Proposition: Let V be a vector space over a field F and let $\alpha \in \text{End}(V)$ have an eigenvalue c . Let v be a generalized eigenvector of degree k associated with c . Then the set of vectors $\{v, (\alpha - c\sigma_1)(v), \dots, (\alpha - c\sigma_1)^{k-1}(v)\}$ is linearly independent.

Proof: Set $\beta = \alpha - c\sigma_1$ and, for each $1 \leq j \leq k$, let $v_j = \beta^{k-j}(v)$. Assume that there exist scalars $c_1, \dots, c_k \in F$ satisfying $\sum_{j=1}^k c_j v_j = 0_V$. Then $0_V = \beta^{k-1} \left(\sum_{j=1}^k c_j v_j \right) = \beta^{k-1}(c_k v_k) = c_k \beta^{k-1}(v_k)$ and so, since $\beta^{k-1}(v_k) \neq 0_V$, we conclude that $c_k = 0$. We work backwards in this manner to see that $c_j = 0$ for all $1 \leq j \leq k$, and so the given set is linearly independent. \square

In particular, let V be a vector space of finite dimension n over a field F and let $\alpha \in \text{End}(V)$. If v is a generalized eigenvector of α of degree k associated to an eigenvalue c of α , then we must have $k \leq n$. Thus we see that $\dim(V)$ is an upper bound to the degree of generalized eigenvalue of α and we see that the generalized eigenspace of α associated to an eigenvalue c is just $\ker((\alpha - c\sigma_1)^n)$.

(13.7) Proposition: Let V be a vector space of finite dimension n over a field F and let $\alpha \in \text{End}(V)$ satisfy the condition that the characteristic polynomial $p(X)$ of α is completely reducible, say $p(X) = \prod_{j=1}^m (X - c_j)^{n_j}$, where $\text{spec}(\alpha) = \{c_1, \dots, c_m\}$. Then there exist subspaces U_1, \dots, U_m of V , each of which invariant under α , such that:

- (1) $V = U_1 \oplus \dots \oplus U_m$;
- (2) $\dim(U_h) = n_h$ for each $1 \leq h \leq m$;
- (3) For each $1 \leq h \leq m$, the restriction of α to U_h is of the form $c_h \tau_h + \beta_h$, where $\beta_h \in \text{End}(U_h)$ is nilpotent and τ_h is the restriction of σ_1 to U_h .

Proof: For each $1 \leq h \leq m$, consider the endomorphism $\beta_h = \alpha - c_h \sigma_1$ of V , and let U_h be the generalized eigenspace of α associated with c_h . Then U_h is a subspace of V invariant under β_h and also invariant under α since for all $v \in U_h$ we have $\beta_h^n \alpha(v) = \alpha \beta_h^n(v) = \alpha(0_V) = 0_V$. We claim that there exists a positive integer k , independent of h , such that all elements of U_h are generalized eigenvectors of α of degree at most k . Indeed, we see that $\ker(\beta_h) \subseteq \ker(\beta_h^2) \subseteq \ker(\beta_h^3) \subseteq \dots$ and since V is finitely-generated, there are at most a finite number of proper containments. Thus there exists a k such that $\ker(\beta_h^k) = \ker(\beta_h^{k+1}) = \dots$. From here it is clear that $\ker(\beta_h^k) = U_h$, proving the claim.

In particular, this claim shows that the restriction of β_h to U_h is a nilpotent endomorphism having index of nilpotence k . More than that, the restriction of α to U_h equals $c_h \tau_h + \beta_h$, proving (3). We now notice that if $t \neq h$ then U_t is invariant under β_h . We claim that the restriction of β_h to U_t is an automorphism. Since U_t is finite-dimensional, it is sufficient to prove that it is a monomorphism. Indeed, suppose that $v \in U_t \cap \ker(\beta_h)$. Then there exists a positive integer k such that $\beta_t^k(v) = 0_V$ and so $0_V = \beta_t^k(v) = [\beta_h + (c_h - c_t)]^k(v) = (c_h - c_t)^k(v)$ and, since $c_h - c_t \neq 0$, we must have $v = 0_V$, proving the claim.

The next step is to show that the collection $\{U_1, \dots, U_m\}$ of subspaces of V is independent. Indeed, let $1 \leq h \leq m$ and let $Y = U_h \cap \sum_{j \neq h} U_j$. Then Y is a subspace of V invariant under β_h on which β_h is monic (since $Y \subseteq \sum_{j \neq h} U_j$) and nilpotent (since $Y \subseteq U_h$), which is possible only if $Y = \{0_V\}$. This proves independence, and we will set $U = U_1 \oplus \dots \oplus U_m$. We want to show that $U = V$. Let $v \in U$. By the Cayley-Hamilton Theorem (Proposition 12.16) we see that α annihilates its characteristic polynomial $p(X)$ and so $[\prod_{i=1}^m \beta_i^{n_i}](v) = 0_V \in U$. Suppose that $\beta_t^{n_1}(v) \in U$, say that it is equal to $\sum_{i=1}^m u_i$, where $u_h \in U_h$ for all $1 \leq h \leq m$. Since $\beta_t^{n_1}$ is epic when restricted to U_h , for each $h \neq t$, we can find an element w_h of U_h for each $1 \leq h \leq m$, such that $u_h = \beta_t^{n_1}(w_h)$ for all such h . Therefore $\beta_t^{n_1}(u - \sum_{h=2}^m w_h) = u_1 \in U_1$. By definition of U_1 , it follows that $w_1 = u - \sum_{h=2}^m w_h \in U_1$. Therefore $v = \sum_{h=1}^m w_h \in U$. If, on the other hand, $\beta_t^{n_1}(v) \notin U$, then let

t be the smallest element of $\{2, \dots, m\}$ satisfying the condition that $\left[\prod_{i=1}^t \beta_i^{n_i}\right](v) \in U$ and $\left[\prod_{i=1}^{t-1} \beta_i^{n_i}\right](v) \notin U$. A similar argument to the preceding then shows that we must have $v \in U$.

We are left to show that $\dim(U_h) = n_h$ for all $1 \leq h \leq m$. Pick a basis for V which is a union of bases of the U_h . With respect to this basis, the endomorphism α is represented by a matrix of the form

$$\begin{bmatrix} A_1 & O & \dots & O \\ O & A_2 & \dots & O \\ & & \ddots & \\ O & O & \dots & A_m \end{bmatrix}, \text{ where each } A_h \text{ is a matrix representing the re-}$$

striction of α to U_h . By Proposition 11.12, the characteristic polynomial of α is therefore of the form $|XI - A| = \prod_{h=1}^m |XI - A_h|$. From this decomposition and from the fact that each β_h restricts to an automorphism of U_t for all $t \neq h$, it follows that the only eigenvalue of the restriction of α to U_h is c_h , and the algebraic multiplicity of this eigenvalue is at most n_h . Since $\sum_{h=1}^m \dim(U_h) = \sum_{h=1}^m n_h$, it then follows that $\dim(U_h) = n_h$ for each h . \square

Proposition 13.7 shows that when conditions are right – for example when the field F is algebraically closed – and when we are given an endomorphism α of a finite-dimensional vector space, it is possible to choose a basis for V relative to which α is represented in a particularly simple form. We do this in two steps.

(I) Write V as a direct sum $U_1 \oplus \dots \oplus U_m$ as above. By choosing a basis for V which is a union of bases of the U_h , we get a matrix representing α composed of blocks strung out along the main diagonal, each representing the restriction of α to one of the subspaces U_h .

(II) For each h , we have $\alpha = c_h \tau_h + \beta_h$, where β_h is a nilpotent endomorphism of U_h . We now choose a basis of U_h relative to which β_h is represented in Jordan canonical form.

Thus, in the end, we have a representation of α by a matrix of the form

$$\begin{bmatrix} A_1 & O & \dots & O \\ O & A_2 & \dots & O \\ & & \ddots & \\ O & O & \dots & A_m \end{bmatrix}, \text{ where each block } A_h \text{ is a matrix with blocks of}$$

$$\text{the form } \begin{bmatrix} c_h & 0 & 0 & \dots & 0 \\ 1 & c_h & 0 & \dots & 0 \\ 0 & 1 & c_h & \dots & 0 \\ & & \ddots & \ddots & \\ 0 & \dots & 0 & 1 & c_h \end{bmatrix} \text{ on its diagonal (these may be } 1 \times 1!).$$

and all other entries equal to 0. A matrix of this form is called the **Jordan canonical form** of α . By Proposition 13.7, we see that if V is a vector

space finitely generated over a field F and if α is an endomorphism of V having a completely reducible characteristic polynomial in $F[X]$, then there is a basis of V relative to which α can be represented by a matrix in Jordan canonical form. Thus, this can always be done if the field F is algebraically closed. If F is not algebraically closed then it is always possible to extend the field F to a larger field K such that the characteristic polynomial of α is completely reducible in $K[X]$.

Example: Consider the endomorphism α of \mathbb{R}^4 which is represented

with respect to the canonical basis by the matrix $A = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ -1 & 4 & -6 & 4 \end{bmatrix}$.

The characteristic polynomial of this matrix is $X^4 - 4X^3 + 6X^2 - 4X + 1 = (X - 1)^4$ and so its only eigenvalue is 1. The matrix A is similar to

the matrix $B = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 \\ 0 & 1 & 1 & 0 \\ 0 & 0 & 1 & 1 \end{bmatrix}$ in Jordan canonical form. Indeed,

$$B = PAP^{-1}, \text{ where } P = \begin{bmatrix} -1 & 3 & -3 & 1 \\ 1 & -2 & 1 & 0 \\ -1 & 1 & 0 & 0 \\ 1 & 0 & 0 & 0 \end{bmatrix}.$$

Example: Consider the endomorphism α of \mathbb{R}^5 which is represented

with respect to the canonical basis by the matrix $A = \begin{bmatrix} 3 & 0 & 0 & 0 & 0 \\ 0 & 4 & 2 & -1 & 4 \\ 0 & 0 & 2 & 0 & 0 \\ 0 & 1 & 3 & 2 & 1 \\ 0 & 0 & 0 & 0 & 2 \end{bmatrix}$.

The characteristic polynomial of this matrix is $(X - 3)^3(X - 2)^2$. The matrix

A is similar to the matrix $B = \begin{bmatrix} 2 & 0 & 0 & 0 & 0 \\ 0 & 3 & 0 & 0 & 0 \\ 0 & 1 & 3 & 0 & 0 \\ 0 & 0 & 0 & 2 & 0 \\ 0 & 0 & 0 & 0 & 3 \end{bmatrix}$ in Jordan canonical form. Indeed, $B = PAP^{-1}$, where $P =$

$$P = \begin{bmatrix} 0 & 0 & -3 & 0 & -4 \\ 0 & 1 & -1 & -1 & 3 \\ 2 & 1 & 3 & 0 & 1 \\ 0 & 0 & 0 & 0 & -1 \\ -1 & 0 & 0 & 0 & 0 \end{bmatrix}.$$

Example: Consider the endomorphism α of \mathbb{C}^4 which is represented with respect to the canonical basis by $A = \begin{bmatrix} 0 & 0 & 2i & 0 \\ 1 & 0 & 0 & 2i \\ -2i & 0 & 0 & 0 \\ 0 & -2i & 1 & 0 \end{bmatrix}$. The characteristic polynomial of this matrix is $(X-2)^2(X+2)^2$. The matrix A

is similar to the matrix $B = \begin{bmatrix} -2 & 0 & 0 & 0 \\ 1 & -2 & 0 & 0 \\ 0 & 0 & 2 & 0 \\ 0 & 0 & 1 & 2 \end{bmatrix}$. Indeed, $B = PAP^{-1}$, where $P = \frac{1}{2} \begin{bmatrix} 1 & 0 & -i & 0 \\ 0 & 1 & 0 & -i \\ 1 & 0 & i & 0 \\ 0 & 1 & 0 & i \end{bmatrix}$.

We now use Jordan canonical forms to prove a result interesting in its own right.

(13.8) Proposition: Let n be a positive integer and let $A \in M_{n \times n}(\mathbb{C})$. Then A can be written as a product of two symmetric matrices.

Proof: By Proposition 13.7 we know that A is similar to a matrix B in Jordan canonical form. In other words, there exists a nonsingular matrix $Q \in M_{n \times n}(\mathbb{C})$ satisfying $A = QBQ^{-1}$. If we can write $B = CD$, where both C and D are symmetric, then $A = QBDQ^{-1} = (QCQ^T)((Q^T)^{-1}DQ^{-1}) = (QCQ^T)((Q^{-1})^T DQ^{-1})$, where both QCQ^T and $(Q^{-1})^T DQ^{-1}$ are symmetric. Therefore, without loss of generality, we can assume that A is in Jordan canonical form, say

$$A = \begin{bmatrix} A_1 & O & \dots & O \\ O & A_2 & \dots & O \\ & & \dots & \\ O & O & \dots & A_m \end{bmatrix}, \quad \text{where each block } A_h \text{ is of the form}$$

$$\begin{bmatrix} a_h & 0 & 0 & \dots & 0 \\ 1 & a_h & 0 & \dots & 0 \\ 0 & 1 & a_h & \dots & 0 \\ & & \ddots & \ddots & \\ 0 & \dots & 0 & 1 & a_h \end{bmatrix} \in M_{n_h \times n_h}(\mathbb{C}).$$

Define the matrix $D_h \in M_{n_h \times n_h}(\mathbb{C})$ to be $[d_{ij}]$, where

$$d_{ij} = \begin{cases} 1 & \text{if } i+j = n_h + 1 \\ 0 & \text{otherwise} \end{cases}.$$

Then D_h is a symmetric matrix satisfying $D_h^{-1} = D_h$. Moreover, the

matrix $D = \begin{bmatrix} D_1 & O & \dots & O \\ O & D_2 & \dots & O \\ & & \ddots & \\ O & O & \dots & D_m \end{bmatrix} \in \mathcal{M}_{n \times n}(\mathbb{C})$ is also symmetric and satisfies $D^{-1} = D$. Furthermore, the matrix

$$C = \begin{bmatrix} A_1 D_1 & O & \dots & O \\ O & A_2 D_2 & \dots & O \\ & & \ddots & \\ O & O & \dots & A_m D_m \end{bmatrix} \in \mathcal{M}_{n \times n}(\mathbb{C})$$

is also symmetric and $A = CD$, as required. \square

Exercises

Exercise 769 Let $V = \mathbb{R}^3$. Find endomorphisms α and β of V satisfying the condition that $\alpha\beta$ is not nilpotent but $c\alpha + d\beta$ is nilpotent for all $c, d \in \mathbb{R}$.

Exercise 770 Let V be a vector space over a field F and let $\alpha \in \text{End}(V)$ be nilpotent, having index of nilpotence $k > 0$. Show that $\sigma_1 + \alpha \in \text{Aut}(V)$.

Exercise 771 Let V be a vector space finitely-generated over \mathbb{C} . Do there exist endomorphisms α and β of V satisfying the condition that $\sigma_1 + \alpha\beta - \beta\alpha$ is nilpotent?

Exercise 772 Let F be a field. Give an example of a nilpotent endomorphism of F^5 having index of nilpotence 3.

Exercise 773 Let α be the endomorphism of $V = \mathbb{R}^4$ represented with

respect to a basis B of V by the matrix $\begin{bmatrix} 2 & -8 & 12 & -60 \\ 2 & -5 & 9 & -48 \\ 6 & -17 & 29 & -152 \\ 1 & -3 & 5 & -26 \end{bmatrix}$.

Show that α is nilpotent and find its index of nilpotence.

Exercise 774 Let V be a vector space over a field F and let $\alpha \in \text{End}(V)$ be nilpotent. Does $\beta\alpha$ have to be nilpotent for all $\beta \in \text{End}(V)$?

Exercise 775 Let V be a vector space over a field F and let $\alpha \in \text{End}(V)$ be nilpotent. Find $\text{spec}(\alpha)$.

Exercise 776 Let V be a vector space finitely generated over \mathbb{C} and let $\alpha \in \text{End}(V)$ satisfy $\text{spec}(\alpha) = \{0\}$. Show that α is nilpotent.

Exercise 777 Let V be a vector space over a field F and let $\alpha \in \text{End}(V)$ be nilpotent, having index of nilpotence k . Find the minimal polynomial of α .

Exercise 778 Let V be a vector space finitely generated over a field F and let $\alpha \in \text{End}(V)$ satisfy the condition that for each $v \in V$ there exists a positive integer $n(v)$ satisfying $\alpha^{n(v)}(v) = 0_V$. Show that α is nilpotent.

Exercise 779 Let F be a field and let α be an endomorphism of F^3

represented with respect to a basis B of F^3 by the matrix $\begin{bmatrix} 0 & a & 0 \\ 0 & 0 & b \\ 0 & 0 & 0 \end{bmatrix}$,

where a and b are nonzero scalars. Does there exist a endomorphism β of F^3 satisfying $\beta^2 = \alpha$?

Exercise 780 Let α be a nilpotent endomorphism of a vector space V over a field F having characteristic 0. Show that there exists an endomorphism β of V belonging to $F[\alpha]$ and satisfying $\beta^2 = \sigma_1 + \alpha$.

Exercise 781 Let $\alpha \in \text{End}(\mathbb{R}^3)$ be represented with respect to the canon-

ical basis by the matrix $\begin{bmatrix} 1 & 2 & -2 \\ 3 & 0 & 3 \\ 1 & 1 & -2 \end{bmatrix}$. Calculate $\mathbb{R}[\alpha] \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$.

Exercise 782 Let V be the space of all infinitely-differentiable functions from \mathbb{R} to itself. Let δ be the endomorphism of V which assigns to each function its derivative. What is $\mathbb{R}[\delta] \sin(x)$?

Exercise 783 Define $\alpha \in \text{End}(\mathbb{R}^3)$ by $\alpha : \begin{bmatrix} a \\ b \\ c \end{bmatrix} \mapsto \begin{bmatrix} a+c \\ b-a \\ b \end{bmatrix}$. Find

$$\mathbb{R}[\alpha] \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}.$$

Exercise 784 Let V be a vector space over a field F and let $\alpha \in \text{End}(V)$. Let $v \in V$ be a vector satisfying $F[\alpha^2]v = V$. Show that $F[\alpha]v = V$.

Exercise 785 Given $a \in \mathbb{R}$, let $\alpha_a \in \text{End}(\mathbb{R}^4)$ be represented with

respect to the canonical basis by the matrix
$$\begin{bmatrix} 0 & a & 1 & 0 \\ 1 & -2 & 1 & 1 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & -2 \end{bmatrix}.$$
 For

which values of a is the dimension of $\mathbb{R}[\alpha_a]$ $\begin{bmatrix} 0 \\ 0 \\ 0 \\ -1 \end{bmatrix}$ equal to 3?

Exercise 786 Let $\alpha \in \text{End}(\mathbb{R}^3)$ be represented with respect to the canon-

ical basis by the matrix
$$\begin{bmatrix} 1 & 1 & 1 \\ 0 & 1 & 0 \\ 0 & 0 & 2 \end{bmatrix}.$$
 Find the eigenvalues of α and

the generalized eigenspace associated with each.

Exercise 787 Let V be a vector space over a field F and let $\alpha, \beta \in \text{End}(V)$ satisfy $\alpha\beta = \beta\alpha$. Let W be the generalized eigenspace of α associated with an eigenvalue a . Show that W is invariant under β .

Exercise 788 Let V be a vector space finitely generated over \mathbb{C} and let $\alpha \in \text{End}(V)$. Show that V is diagonalizable if and only if every generalized eigenvector of α is an eigenvector of α .

Exercise 789 Let $B = \{v_1, v_2, v_3\}$ be the canonical basis of \mathbb{R}^3 and

let α be the endomorphism of \mathbb{R}^3 satisfying $\Phi_{BB}(\alpha) = \begin{bmatrix} 1 & 0 & 2 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}.$

Show that $W = \mathbb{R}\{v_1, v_3\}$ and $Y = \mathbb{R}v_2$ are complements of each other in \mathbb{R}^3 and that each of these spaces is invariant under α .

Exercise 790 Let $A = \begin{bmatrix} 1 & 2 & 0 \\ 0 & 2 & 0 \\ 2 & -2 & -1 \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{R})$. Find the Jordan canonical form of A .

Exercise 791 Let $A = \begin{bmatrix} -2 & 8 & 6 \\ -4 & 10 & 6 \\ 4 & -8 & -4 \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{R})$. Find the Jordan canonical form of A .

Exercise 792 Let $O \neq A \in \mathcal{M}_{3 \times 3}(\mathbb{C})$ be of the form
$$\begin{bmatrix} 0 & a & -b \\ -a & 0 & c \\ b & -c & 0 \end{bmatrix},$$

where a , b , and c are real numbers. What is the Jordan canonical form of A ?

Exercise 793 Let n be a positive integer. If $A \in \mathcal{M}_{n \times n}(\mathbb{C})$, show that A and A^T are similar.

Exercise 794 Let $A \in \mathcal{M}_{5 \times 5}(\mathbb{Q})$ be a matrix in Jordan canonical form having minimal polynomial $(X - 3)^2$. What does A look like?

Exercise 795 Give an example of a matrix in $\mathcal{M}_{4 \times 4}(\mathbb{R})$ which is not similar to a matrix in Jordan canonical form.

Exercise 796 Let V be a vector space finitely generated over a field F and let α and β be nilpotent endomorphisms of V represented with respect to some given basis by matrices A and B respectively. If the matrices A and B are similar, does the index of nilpotence of α have to equal that of β ?

Exercise 797 Let F be a field and let $A = \begin{bmatrix} a & 0 & 0 \\ 1 & a & 0 \\ 0 & 1 & a \end{bmatrix} \in \mathcal{M}_{3 \times 3}(F)$.

Show that $A^k = \begin{bmatrix} a^k & 0 & 0 \\ ka^{k-1} & a^k & 0 \\ \frac{1}{2}k(k-1)a^{k-2} & ka^{k-1} & a^k \end{bmatrix}$ for all $k > 0$.

Exercise 798 Let n be a positive integer and, for all $1 \leq i, j \leq n$, let $p_{ij}(X) \in \mathbb{C}[X]$. Let $\varphi : \mathbb{C} \rightarrow \mathcal{M}_{n \times n}(\mathbb{C})$ be the function defined by $\varphi : z \mapsto [p_{ij}(z)]$. Furthermore, let us assume that $\varphi(z)$ is nonsingular for each $z \in \mathbb{C}$. Show that there exists a nonzero complex number d such that $|\varphi(z)| = d$ for all $z \in \mathbb{C}$.

Exercise 799 For each $t \in \mathbb{C}$, let α_t be the endomorphism of \mathbb{C}^3

represented with respect to the canonical basis by the matrix $\begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 0 \\ t & 0 & 0 \end{bmatrix}$.

Is the representation of α_t in Jordan canonical form dependent on t ?

Exercise 800 Let $A = \begin{bmatrix} 0 & 0 & 4 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{C})$. Find the set of all

$c \in \mathbb{C}$ satisfying the condition that cA is similar to A .

14

The dual space

Let V be a vector space over a field F . A linear transformation from V to F (considered as a vector space over itself) is a **linear functional** on V . The space $\text{Hom}(V, F)$ of all such linear functionals is called the **dual space** of V and will be denoted by $D(V)$. Note that $D(V)$ is a vector space over F , the identity element of which for addition is the **0-functional**, $v \mapsto 0$. Since $\dim(F) = 1$, we immediately see that every linear functional other than the 0-functional must be an epimorphism¹.

Example: Let F be a field and let n be a positive integer. Any $v \in F^n$ defines a linear functional in $D(F^n)$ by $w \mapsto v \odot w$.



¹ Linear functionals were first studied systematically by the French mathematician **Jacques Hadamard**, whose long life ranged from the mid 19th century to the mid 20th century. His work on functionals turned them into an important tool in analysis.

Example: Let V be a vector space over a field F and let B be a basis for V . Each $u \in B$ defines a function $f_u \in F^B$ defined by

$$f_u : u' \mapsto \begin{cases} 1 & \text{if } u' = u \\ 0 & \text{otherwise} \end{cases}$$

and by Proposition 6.2 we know that this function in turn defines a linear functional $\delta_u \in D(V)$. In particular, if $V = F^n$ and if $B = \{u_1, \dots, u_n\}$

is the canonical basis for V , then $\delta_{u_h} : \begin{bmatrix} a_1 \\ \vdots \\ a_n \end{bmatrix} \mapsto a_h$ for each $1 \leq h \leq n$.

Example: Suppose that $V = C(a, b)$ and that $g_0 \in V$. Then the function $\eta : V \rightarrow \mathbb{R}$ defined by $\eta : f \mapsto \int_a^b f(x)g_0(x)dx$ belongs to $D(V)$. Hadamard's initial work on linear functionals concerned those of the form $f \mapsto \lim_{n \rightarrow \infty} \int_a^b f(x)g_n(x)dx$ for suitable sequences g_1, g_2, \dots in V .

Example: Let V be the subspace of $\mathbb{R}^{\mathbb{R}}$ consisting of all infinitely-differentiable functions f satisfying the condition that there exist real numbers $a \leq b$ such that $f(x) = 0$ if $x \notin [a, b]$. Then the function $f \mapsto \int_{-\infty}^{\infty} f(x)dx$ belongs to $D(V)$. Elements of $D(V)$ are known as **distributions** and play an important role in analysis and theoretical physics.

Let F be a field and let n be a positive integer. Then there exists a linear functional $tr : \mathcal{M}_{n \times n}(F) \rightarrow F$ which assigns to each matrix the sum of the elements of its diagonal, i.e. $tr : [a_{ij}] \mapsto \sum_{i=1}^n a_{ii}$. This linear functional is called the **trace**. This functional will play an important part in our later discussion. Note that $v \odot w = tr(v \wedge w)$ for all $v, w \in F^n$.

The set of all matrices $A \in \mathcal{M}_{n \times n}(F)$ satisfying $tr(A) = 0$ forms a subalgebra of the general Lie algebra $\mathcal{M}_{n \times n}(F)^-$, called the **special Lie algebra** defined by F^n .

If $A = [a_{ij}]$ and $B = [b_{ij}]$ are matrices in $\mathcal{M}_{n \times n}(F)$, then it is easy to see that $tr(AB) = \sum_{i=1}^n \sum_{h=1}^n a_{ih}b_{hi} = tr(BA)$. We also notice that $tr(I) = n$, where I is the identity matrix of $\mathcal{M}_{n \times n}(F)$. If the characteristic of the field F does not divide n , we claim that these conditions uniquely characterize the trace.

(14.1) Proposition: Let n be a positive integer and let F be a field the characteristic of which does not divide n . Let δ be a linear functional on $\mathcal{M}_{n \times n}(F)$ satisfying the conditions that

$\delta(AB) = \delta(BA)$ for all $A, B \in \mathcal{M}_{n \times n}(F)$ and that $\delta(I) = n$. Then $\delta = \text{tr}$.

Proof: If $\{H_{ij} \mid 1 \leq i, j \leq n\}$ is the canonical basis of $\mathcal{M}_{n \times n}(F)$, then it suffices to show that $\delta(H_{ij}) = \text{tr}(H_{ij})$ for all $1 \leq i, j \leq n$. In particular, if $1 \leq i, j \leq n$ then $\delta(H_{ii}) = \delta(H_{ij}H_{ji}) = \delta(H_{ji}H_{ij}) = \delta(H_{jj})$. Since $I = \sum_{i=1}^n H_{ii}$, this implies that $n = \delta(I) = \sum_{i=1}^n \delta(H_{ii})$ and so $\delta(H_{ii}) = 1 = \text{tr}(H_{ii})$ for all $1 \leq i \leq n$. If $i \neq j$ then $H_{1j}H_{i1} = O$ and so $\delta(H_{ij}) = \delta(H_{i1}H_{1j}) = \delta(H_{1j}H_{i1}) = \delta(O) = 0 = \text{tr}(H_{ij})$ and we are done. \square

By the above, we see that if F is a field, if n is a positive integer, and if $A, B \in \mathcal{M}_{n \times n}(F)$, then $\text{tr}(A \bullet B) = \text{tr}(AB) - \text{tr}(BA) = 0$, where \bullet denotes the Lie product on $\mathcal{M}_{n \times n}(F)$. In fact, the converse of this result is also true – namely that if a matrix C satisfies $\text{tr}(C) = 0$, then C is the Lie product of two other matrices – as has been proven in several more general contexts².

If n is a positive integer, if F is a field, and if P is a nonsingular matrix in $\mathcal{M}_{n \times n}(F)$, then $\text{tr}(PAP^{-1}) = \text{tr}(PP^{-1}A) = \text{tr}(A)$ and so similar matrices have identical traces. In general, if B and C are fixed matrices in $\mathcal{M}_{n \times n}(F)$ then the functions $A \mapsto \text{tr}(BA)$ and $A \mapsto \text{tr}(AC)$ belong to $D(\mathcal{M}_{n \times n}(F))$.

The following result shows that traces essentially define all linear functionals on spaces of square matrices.

(14.2) Proposition: Let F be a field, let n be a positive integer, and let $\delta \in D(\mathcal{M}_{n \times n}(F))$. Then there exists a matrix $C \in \mathcal{M}_{n \times n}(F)$ satisfying $\delta : A \mapsto \text{tr}(AC)$ for all $A \in \mathcal{M}_{n \times n}(F)$.

Proof: For each $1 \leq i, j \leq n$, let H_{ij} be the matrix having (i, j) -entry equal to 1 and all of the other entries equal to 0. Then we know that the set of all such matrices is a basis for $\mathcal{M}_{n \times n}(F)$. Let $C = [c_{ij}] \in \mathcal{M}_{n \times n}(F)$ be the matrix defined by $c_{ij} = \delta(H_{ji})$ for all $1 \leq i, j \leq n$. Then for each matrix $A = [a_{ij}] \in \mathcal{M}_{n \times n}(F)$ we have $\delta(A) = \delta\left(\sum_{i=1}^n \sum_{j=1}^n a_{ij} H_{ij}\right) =$



2

The first of many proofs of this result was given by **Kenjiro Shoda**, one of the major figures in 20th-century Japanese algebra.

$\sum_{i=1}^n \sum_{j=1}^n a_{ij} \delta(H_{ij}) = \sum_{i=1}^n \sum_{j=1}^n a_{ij} c_{ji} = \text{tr}(AC)$, as we needed to prove. \square

(14.3) Proposition: Let F be a field, let n be a positive integer, and let $\delta \in D(\mathcal{M}_{n \times n}(F))$ be a linear functional satisfying $\delta(AB) = \delta(BA)$ for all $A, B \in \mathcal{M}_{n \times n}(F)$. Then there exists a scalar $c \in F$ such that $\delta(A) = c \cdot \text{tr}(A)$ for all $A \in \mathcal{M}_{n \times n}(F)$.

Proof: Again, for each $1 \leq i, j \leq n$, let H_{ij} be the matrix having (i, j) -entry equal to 1 and all of the other entries equal to 0. If $1 \leq i \neq j \leq n$ then $\delta(H_{ij}) = \delta(H_{ii}H_{ij}) = \delta(H_{ij}H_{ii}) = \delta(O) = 0$. Moreover, for all $1 \leq j, k \leq n$ we have $\delta(H_{jj}) = \delta(H_{jk}H_{kj}) = \delta(H_{kj}H_{jk}) = \delta(H_{kk})$. Thus we see that there exists a $c \in F$ such that $\delta(H_{jj}) = c$ for all $1 \leq j \leq n$ and from Proposition 14.2 we conclude that $\delta(A) = \text{tr}(A \cdot cI) = c \cdot \text{tr}(A)$ for all $A \in \mathcal{M}_{n \times n}(F)$. \square

(14.4) Proposition: Let F be a field, let n be a positive integer, and let $A \in \mathcal{M}_{n \times n}(F)$ be a matrix the characteristic polynomial of which is completely reducible. Then $\text{tr}(A)$ is the sum of the eigenvalues of A (with the appropriate multiplicities).

Proof: Let $p(X) = \sum_{i=0}^n c_i X^i$ be the characteristic polynomial of A . We know that this polynomial is completely reducible, say $p(X) = \prod_{i=1}^n (X - b_i)$, and after multiplying this out, we see that $c_{n-1} = -\sum_{i=1}^n b_i$. But from the definition of the characteristic polynomial, we also see that $c_{n-1} = -\text{tr}(A)$. Thus we see that, for any such matrix, $\text{tr}(A)$ is the sum of the eigenvalues of A (with the appropriate multiplicities). \square

Example: Let n be a positive integer and let c and d be complex numbers. Can we find all matrices $A \in \mathcal{M}_{n \times n}(\mathbb{C})$ having the property that c is an eigenvalue of A having geometric multiplicity $n-1$ and d is an eigenvalue of A having geometric multiplicity 1? (Certainly one such matrix always exists, namely a diagonal matrix with c appearing $n-1$ times on the diagonal and d once.) In general, in order for c to be an eigenvalue of A of geometric multiplicity $n-1$, the eigenspace connected with it has to be of dimension $n-1$. In other words, the nullity of the matrix $A - cI$ must equal $n-1$. From this we see that the dimension of the column space of $A - cI$ must equal 1, and so there must exist nonzero

vectors $u = \begin{bmatrix} b_1 \\ \vdots \\ b_n \end{bmatrix}$ and $v = \begin{bmatrix} e_1 \\ \vdots \\ e_n \end{bmatrix}$ in \mathbb{C}^n such that $A - cI = u \wedge v$,

whence $A = u \wedge v + cI$. Conversely, if A is a matrix of the form $u \wedge v + cI$, then c is an eigenvalue of A having multiplicity at least $n-1$. Note

that $\text{tr}(A) = \sum_{i=1}^n b_i e_i + nc$. But, as we just noted, $\text{tr}(A)$ is also the sum of the eigenvalues of A , counted by multiplicity, and so we want it to equal $d + (n-1)c$. Thus we are reduced to finding vectors u and v as above satisfying the condition that $\sum_{i=1}^n b_i e_i = d - c$. This is easy to do in concrete cases.

Example: Let F be a field, let Ω be a nonempty set, and let $V = F^\Omega$. For each $a \in \Omega$ there exists a linear functional $\delta_a \in D(V)$ defined by evaluation: $\delta_a : f \mapsto f(a)$. In the case that F is \mathbb{R} and Ω is the unit interval of the real line, this functional is known to physicists as the **Dirac³ functional**. In analysis, evaluation functionals are often used to establish boundary conditions on classes of functions being studied.

The following proposition shows that there are always enough linear functionals to enable us to distinguish between vectors.

(14.5) Proposition: Let V be a vector space of a field F . If $v \neq w$ are elements of V then there exists a linear functional $\delta \in D(V)$ satisfying $\delta(v) \neq \delta(w)$.

Proof: Since the set $\{v-w\}$ is linearly independent, it can be completed to a basis B of V . By Proposition 6.2, there exists a linear functional $\delta \in D(V)$ satisfying $\delta(v-w) = 1$ and $\delta(u) = 0$ for all $u \in B \setminus \{v-w\}$. This is the linear functional we want. \square

In particular, if $0_V \neq v \in V$ then there is a linear functional $\delta \in D(V)$ satisfying $\delta(v) \neq 0$.

We can also extend the idea we used in the proof of Proposition 14.5. Let V be a vector space over a field F and let B be a given basis for V . For each $v \in B$, let $\delta_v \in D(V)$ satisfy $\delta_v(v) = 1$ and $\delta_v(u) = 0$ for all $v \neq u \in B$. We claim that $E = \{\delta_v \mid v \in B\}$ is a linearly-independent subset of $D(V)$. Indeed, if c_1, \dots, c_n are scalars in F and u_1, \dots, u_n are elements of B satisfying the condition that $\sum_{i=1}^n c_i \delta_{u_i}$ is



3

Paul Dirac, the Nobel-prize-winning 20th-century British physicist, built the first accepted model of quantum mechanics, in which linear functions played a fundamental part.

the 0-functional. Then for all $1 \leq h \leq n$ we have

$$0 = \left(\sum_{i=1}^n c_i \delta_{u_i} \right) (u_h) = \sum_{i=1}^n c_i \delta_{u_i}(u_h) = c_h.$$

This establishes the claim. If V is finitely-generated then B is finite and so E is a basis for $D(V)$, since it is easy to check that $\delta = \sum_{u \in B} \delta(u) \delta_u$ for all $\delta \in D(V)$. Such a basis E for $D(V)$ is called the **dual basis** of the basis B for V . If V is not finitely generated, then FE is a subspace of $D(V)$ composed of all those linear functionals $\delta \in D(V)$ satisfying the condition that $\delta(u) \neq 0$ for at most finitely-many elements u of B . This subspace is called the **weak dual space** of V .

As a consequence of these remarks, we immediately see:

(14.6) Proposition: If V is a vector space finitely generated over a field F then $\dim(V) = \dim(D(V))$.

Example: Let $a < b$ be real numbers and let t_1, \dots, t_n be distinct real numbers in the closed interval $[a, b]$ of the real line and let W be the subspace of $\mathbb{R}^{\mathbb{R}}$ consisting of all polynomial functions of degree less than n . Then $\dim(W) = n$. For each $1 \leq i \leq n$, let $\delta_i \in D(W)$ be the linear functional defined by $\delta_i : p \mapsto p(t_i)$. We claim that the subset $B = \{\delta_1, \dots, \delta_n\}$ of $D(W)$ is linearly independent. Indeed, if $\sum_{i=1}^n c_i \delta_i$ is the 0-functional and if $1 \leq h \leq n$, then $0 = (\sum_{i=1}^n c_i \delta_i) \prod_{j \neq h} (X - t_j) = c_h \prod_{j \neq h} (t_h - t_j)$, which implies that $c_h = 0$ since the t_i are distinct. Therefore, by Proposition 14.6, B is a basis for $D(W)$. Since the function $p \mapsto \int_a^b p(x) dx$ also belongs to $D(W)$, we conclude that there exist real numbers c_1, \dots, c_n satisfying $\int_a^b p(x) dx = \sum_{i=1}^n c_i p(t_i)$ for any $p \in W$.

Example: Let V be the vector space of all polynomial functions in $\mathbb{R}^{\mathbb{R}}$ having degree at most 4. Suppose that $B = \{a_1, \dots, a_5\}$ is a set of distinct positive real numbers and, for each $1 \leq i \leq 5$, let $\delta_i \in D(V)$ be the linear transformation defined by $\delta_i : p(t) \mapsto \int_0^{a_i} p(t) dt$. We claim that $B = \{\delta_1, \dots, \delta_5\}$ is a basis for $D(V)$. Indeed, since we know by Proposition 14.6 that $\dim(D(V)) = 5$, all we have to show is that the set B is linearly independent. That is to say, we must show that if there exist real numbers b_1, \dots, b_5 satisfying the condition that $\sum_{i=1}^5 b_i \delta_i$ is the 0-functional, then $b_i = 0$ for all $1 \leq i \leq 5$. Since $\sum_{i=1}^5 b_i \delta_i(t^h) = \sum_{i=1}^5 \left[\frac{1}{h+1} a_i^{h+1} \right] b_i$ for all $0 \leq h \leq 4$, we must show that the matrix

$$A = \begin{bmatrix} a_1 & \frac{1}{2}a_1^2 & \frac{1}{3}a_1^3 & \frac{1}{4}a_1^4 & \frac{1}{5}a_1^5 \\ a_2 & \frac{1}{2}a_2^2 & \frac{1}{3}a_2^3 & \frac{1}{4}a_2^4 & \frac{1}{5}a_2^5 \\ & & \vdots & & \\ a_5 & \frac{1}{2}a_5^2 & \frac{1}{3}a_5^3 & \frac{1}{4}a_5^4 & \frac{1}{5}a_5^5 \end{bmatrix}$$

is nonsingular, which we know in fact to be the case since it is just a Vandermonde matrix, which is always nonsingular.

Note that Proposition 14.6 is never true for vector spaces which are not finitely generated. The proof of this fact requires a knowledge of the arithmetic of infinite cardinals.

Now let V and W be vector spaces over a field F and let $\alpha \in \text{Hom}(V, W)$. If $\delta \in D(W)$ then $\delta\alpha \in D(V)$. Moreover, if $\delta_1, \delta_2 \in D(W)$ and if $v \in V$ then $[(\delta_1 + \delta_2)\alpha](v) = (\delta_1 + \delta_2)\alpha(v) = \delta_1\alpha(v) + \delta_2\alpha(v) = [\delta_1\alpha + \delta_2\alpha](v)$ and so $(\delta_1 + \delta_2)\alpha = \delta_1\alpha + \delta_2\alpha$. Similarly, if $c \in F$ and if $\delta \in D(W)$ then $c(\delta\alpha) = (c\delta)\alpha$. Therefore we see that α defines a linear transformation $D(\alpha) \in \text{Hom}(D(W), D(V))$ by setting $D(\alpha) : \delta \mapsto \delta\alpha$. If V, W , and Y are vector spaces over F and if $\alpha \in \text{Hom}(V, W)$ and $\beta \in \text{Hom}(W, Y)$ then it is straightforward to show that $D(\beta\alpha) = D(\alpha)D(\beta)$. If α is an isomorphism, then $D(\alpha)$ is also an isomorphism, where $D(\alpha)^{-1} = D(\alpha^{-1})$.

(14.7) Proposition: Let F be a field and let V and W be vector spaces finitely generated over F . Let $B = \{v_1, \dots, v_k\}$ be a basis for V , the dual basis of which is $C = \{\delta_1, \dots, \delta_k\}$, and let $D = \{w_1, \dots, w_n\}$ be a basis for W , the dual basis of which is $E = \{\eta_1, \dots, \eta_n\}$. If $\alpha : V \rightarrow W$ is a linear transformation then $\Phi_{EC}(D(\alpha)) = \Phi_{BD}(\alpha)$.

Proof: Let $\Phi_{BD}(\alpha) = [a_{ij}]$. For each $1 \leq i \leq k$ we have $\alpha(v_i) = \sum_{h=1}^n a_{hi}w_h$ and so for all $1 \leq i \leq k$ and all $1 \leq j \leq n$ we have $[D(\alpha)(\eta_j)](v_i) = \eta_j\alpha(v_i) = \sum_{h=1}^n a_{hi}\eta_j(w_h) = a_{ji}$. But each $\delta \in D(V)$ satisfies $\delta = \sum_{i=1}^k \delta(v_i)\delta_i$ and so, in particular, $D(\alpha)(\eta_j) = \sum_{i=1}^k [D(\alpha)(\eta_j)](v_i)\delta_i = \sum_{i=1}^k a_{ji}\delta_i$, which gives the desired result. \square

We have already seen that, given a vector space V over a field F , we can build the dual space $D(V)$. Since this too is a vector space over F , we can go on to build its dual space, $D^2(V) = D(D(V))$. What do some elements of this space look like? Each $v \in V$ defines a function $\theta_v : D(V) \rightarrow F$ by setting $\theta_v : \delta \mapsto \delta(v)$. This is indeed a linear function and so is an element of $D^2(V)$, which we call the **evaluation functional** at v .

(14.8) Proposition: Let V be a vector space over a field F . The function $v \mapsto \theta_v$ is a monomorphism from V to $D^2(V)$, which is an isomorphism in the case V is finitely generated.

Proof: We first have to show that this function is a linear transformation. And, indeed, if $v, w \in V$, if $a \in F$, and if $\delta \in D(V)$, then as a direct consequence of the definitions we obtain $\theta_{v+w}(\delta) = \delta(v+w) = \delta(v) + \delta(w) = \theta_v(\delta) + \theta_w(\delta) = [\theta_v + \theta_w](\delta)$ and so $\theta_{v+w} = \theta_v + \theta_w$. Similarly, $\theta_{av}(\delta) = \delta(av) = a\delta(v) = a\theta_v(\delta)$ and so $\theta_{av} = a\theta_v$. Thus we have shown that we do indeed have a homomorphism. If v belongs to the kernel of this function then $\theta_v(\delta) = \delta(v)$ for all $\delta \in D(V)$ and so, by Proposition 14.5, we know that $v = 0_V$. Thus it is a monomorphism. Finally, if V is finitely generated then by Proposition 14.6, we see that $\dim(D^2(V)) = \dim(D(V)) = \dim(V)$ and so any monomorphism from V to $D^2(V)$ has to be an isomorphism. \square

We should note that the importance of Proposition 14.8 lies not in the existence of an isomorphism between V and $D^2(V)$, which could be inferred from dimension arguments alone, but in finding a specific, natural, such isomorphism.

A proper subspace W of a vector space V over a field F is a **maximal subspace** if and only if there is no subspace of V properly contained in V and properly containing W . By the Hausdorff Maximum Principle we know that any nontrivial vector space contains a maximal subspace. The maximal subspaces of finitely-generated vector spaces are usually called **hyperplanes** of the space. We will now use linear functionals in order to characterize these subspaces of V .

(14.9) Proposition: A subspace W of a vector space V over a field F is maximal if and only if there exists a linear functional $\delta \in D(V)$ which is not the 0-functional, the kernel of which equals W .

Proof: Let us assume that $W = \ker(\delta)$, where δ is a linear functional which is not the 0-functional, and assume that there exists a proper subspace Y of V which properly contains W . Pick $y \in Y \setminus W$ and $x \in V \setminus Y$. These two vectors have to be nonzero and the set $\{x, y\}$ is linearly independent by Proposition 5.3, since $Fy \subseteq Y$ and $x \notin Y$. Set $U = F\{x, y\}$. Then $\ker(\delta)$ and U are disjoint, so the restriction of δ to U is a monomorphism, which is impossible since $\dim(U) = 2$ and $\dim(F) = 1$. Therefore W must be a maximal subspace of V .

Conversely, let W be a maximal subspace of V and let $y \in V \setminus W$. Then $Fy \cap W = \{0_V\}$ and $Fy + W = V$ by the maximality of W . Therefore $V = Fy \oplus W$ and so every vector in V can be written in the

form $ay + w$, where $a \in F$ and $w \in W$. The function $\delta : ay + w \mapsto a$ is a linear functional in $D(V)$ the kernel of which equals W , as we need. \square

(14.10) Proposition: Let V be a vector space over a field F and let $\delta, \delta_1, \dots, \delta_n$ be elements of $D(V)$. Then $\delta \in F\{\delta_1, \dots, \delta_n\}$ if and only if $\bigcap_{i=1}^n \ker(\delta_i) \subseteq \ker(\delta)$.

Proof: Assume that $\delta \in F\{\delta_1, \dots, \delta_n\}$. Then there exist scalars a_1, \dots, a_n such that $\delta = \sum_{i=1}^n a_i \delta_i$. If $v \in \bigcap_{i=1}^n \ker(\delta_i)$ then $\delta_i(v) = 0$ for all $1 \leq i \leq n$ and so $\delta(v) = \sum_{i=1}^n a_i \delta_i(v) = 0$. Thus $v \in \ker(\delta)$. Conversely, suppose that $\bigcap_{i=1}^n \ker(\delta_i) \subseteq \ker(\delta)$. We will proceed by induction on n . First, assume that $n = 1$. If δ is the 0-functional; then surely we are done. Thus let us assume that this is not the case and let $v \in V \setminus \ker(\delta)$. Since $\ker(\delta_1) \subseteq \ker(\delta)$, this means that $\delta_1(v) \neq 0$. Set $a = \delta_1(v)^{-1} \delta(v)$. Then $\delta(v) = a \delta_1(v) = (a \delta_1)(v)$ and so $v \in \ker(\delta - a \delta_1)$. But $\ker(\delta_1) \subseteq \ker(\delta - a \delta_1)$ and so this containment is again proper. By Proposition 14.9, $\ker(\delta_1)$ is a maximal subspace of V and so $\ker(\delta - a \delta_1) = V$, which shows that $\delta = a \delta_1$.

Now let us assume that we have prove the result for a given n and assume we have linear functionals $\delta, \delta_1, \dots, \delta_{n+1}$ in $D(V)$ satisfying $\bigcap_{i=1}^{n+1} \ker(\delta_i) \subseteq \ker(\delta)$. Set $W = \ker(\delta_{n+1})$ and for each $1 \leq i \leq n$ let β_i be the restriction of δ_i to W . Also, let β be the restriction of δ to W . Then $\bigcap_{i=1}^n \ker(\beta_i) \subseteq \ker(\beta)$ and so, by the induction hypothesis, we know that there exist scalars a_1, \dots, a_n such that $\beta = \sum_{i=1}^n a_i \beta_i$. Therefore $\ker(\delta_{n+1}) \subseteq \ker(\delta - \sum_{i=1}^n a_i \delta_i)$ and, as in the case $n = 1$, it follows that there exists a scalar a_{n+1} such that $\delta - \sum_{i=1}^n a_i \delta_i = a_{n+1} \delta_{n+1}$, proving that $\delta = \sum_{i=1}^{n+1} a_i \delta_i$. \square

In the context of functional analysis, the following consequence of Proposition 14.9, taken together with the Riesz Representation Theorem (Proposition 16.13), is known as the **Fredholm⁴ alternative**, and has many important applications.

(14.11) Proposition: Let V and W be vector spaces over a field F , let $\alpha \in \text{Hom}(V, W)$, and let $w \in W$. Then $w \in \text{im}(\alpha)$ if and only if $w \in \ker(\delta)$ for any $\delta \in D(W)$ satisfying $\text{im}(\alpha) \subseteq \ker(\delta)$.



4

Swedish mathematician **Ivar Fredholm** was active in the late 19th century and studied the solvability of integral equations.

Proof: If $w \in \text{im}(\alpha)$ then the given condition clearly holds. Conversely, assume that $w \notin \text{im}(\alpha)$ and let B be a basis for $\text{im}(\alpha)$. By Proposition 5.3, the set $\{w\} \cup B$ is linearly independent and so there exists a subset B' of W containing B such that $\{w\} \cup B'$ is a basis for W . Then FB' is a maximal subspace of W and so, by Proposition 14.9, there exists a $\delta \in D(W)$ satisfying $\delta(w) \neq 0$ and $\text{im}(\alpha) \subseteq FB' = \ker(\delta)$. \square

Exercises

Exercise 801 Let $V = C(0,1)$. From calculus we know that each for $f \in V$ there exists a maximal element a_f of $\{f(t) \mid 0 \leq t \leq 1\}$. Is the function $f \mapsto a_f$ a linear functional on V ?

Exercise 802 Let $F = GF(2)$ and let $\delta : F^3 \rightarrow F$ be the function which

assigns to each vector $v = \begin{bmatrix} a \\ b \\ c \end{bmatrix}$ the value (0 or 1) appearing in the majority of entries of v . Is δ a linear functional?

Exercise 803 Let W be a subspace of $\mathbb{Q}[X]$ generated by a countably-infinite linearly-independent set $\{p_1(X), p_2(X), \dots\}$ of polynomials. Let $\delta : W \rightarrow \mathbb{Q}$ be the function defined by $\delta : \sum_{i=1}^{\infty} a_i p_i(X) \mapsto \sum_{i=1}^{\infty} a_i \deg(p_i)$ (where only finitely-many of the a_i are nonzero). Does δ belong to $D(W)$?

Exercise 804 Find a linear functional $\delta \in D(\mathbb{R}^3)$ which is not the 0-

functional but which satisfies $\delta \left(\begin{bmatrix} 3 \\ 2 \\ -1 \end{bmatrix} \right) = \delta \left(\begin{bmatrix} 3 \\ 2 \\ 1 \end{bmatrix} \right) = 0$.

Exercise 805 Let $V = \mathbb{Q}[X]$ and to each vector $v = [b_1, b_2, \dots] \in \mathbb{Q}^{\infty}$, let us assign a linear functional $\delta_v \in D(V)$ defined by

$$\delta_v : \sum_{n=1}^{\infty} a_n X^n \mapsto \sum_{n=1}^{\infty} n! a_n b_{n+1}.$$

Is the function $\alpha : \mathbb{Q}^{\infty} \rightarrow D(V)$ defined by $v \mapsto \delta_v$ an isomorphism?

Exercise 806 Let V be a vector space over a field F and let $\alpha, \beta \in D(V)$ satisfy the condition that $\ker(\beta) \subseteq \ker(\alpha)$. Show that $\alpha \in F\beta$.

Exercise 807 Let F be a field and let $0 \neq a \in F$. Let $\alpha : F[X] \rightarrow F$ be the function defined by $\alpha : p(X) \mapsto p(a) - p(0)$. Is α a linear functional?

Exercise 808 Let $V = \mathbb{R}^3$ and consider the linear functionals

$$\begin{aligned}\delta_1 &: \begin{bmatrix} a \\ b \\ c \end{bmatrix} \mapsto 2a - b + 3c, & \delta_2 &: \begin{bmatrix} a \\ b \\ c \end{bmatrix} \mapsto 3a - 5b + c \text{ and} \\ \delta_3 &: \begin{bmatrix} a \\ b \\ c \end{bmatrix} \mapsto 4a - 7b + c\end{aligned}$$

on V . Is $\{\delta_1, \delta_2, \delta_3\}$ a basis for $D(V)$?

Exercise 809 Let V be a vector space finitely generated over a field F and let W be a subspace of V having a complement Y in V . Show that $D(V) = W' \oplus Y'$, where W' is a subspace of $D(V)$ isomorphic to W and Y' is a subspace of $D(V)$ isomorphic to Y .

Exercise 810 Let n be a positive integer and let V be the vector space of all polynomial functions from \mathbb{R} to itself of degree no more than n . For all $0 \leq k \leq n$, let $\delta_k : V \rightarrow \mathbb{R}$ be the function defined by $\delta_k : p \mapsto \int_{-1}^1 t^k p(t) dt$. Show that $\{\delta_1, \dots, \delta_n\}$ is a basis of $D(V)$.

Exercise 811 Let $B = \left\{ \begin{bmatrix} 0 \\ 3 \\ -2 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ -1 \end{bmatrix}, \begin{bmatrix} 1 \\ -1 \\ 3 \end{bmatrix} \right\} \subseteq \mathbb{R}^3$. Find the dual basis of B .

Exercise 812 Let n be a positive integer and let V be a vector space of dimension n over a field F . Let $B = \{\delta_1, \dots, \delta_n\}$ be a subset of $D(V)$ and assume that there exists a vector $0_V \neq v \in V$ satisfying $\delta_i(v) = 0$ for all $0 \leq i \leq n$. Show that B is linearly dependent.

Exercise 813 Let V be a vector space over a field F . For every subspace W of V , let $E(W) = \{\delta \in D(V) \mid \ker(\delta) \supseteq W\}$. Show that $E(W)$ is a subspace of $D(V)$. Moreover, if W' and W'' are subspaces of V , show that $E(W' + W'') = E(W') \cap E(W'')$.

Exercise 814 Let V be a vector space finitely generated over a field F and let W be a subspace of V . For $E(W) = \{\delta \in D(V) \mid \ker(\delta) \supseteq W\}$, show that $\dim(W) + \dim(E(W)) = \dim(V)$.

Exercise 815 Let $A, B \in \mathcal{M}_{2 \times 2}(\mathbb{R})$. Show that $\text{tr}(AB) = \text{tr}(A) \cdot \text{tr}(B)$ if and only if $|A + B| = |A| + |B|$.

Exercise 816 Let n be a positive integer and let U be a finite subset of $\mathcal{M}_{n \times n}(\mathbb{C})$ which is closed under multiplication of matrices. Show that there exists a matrix A in U satisfying $\text{tr}(A) \in \{1, \dots, n\}$.

Exercise 817 Let n be a positive integer and let F be a field. For any matrix $A = [a_{ij}] \in \mathcal{M}_{n \times n}(F)$, define the **antitrace** of A to be $\text{antitr}(A) = \sum_{i=1}^n a_{i, n+1-i}$. Is the function $A \mapsto \text{antitr}(A)$ a linear functional on $\mathcal{M}_{n \times n}(F)$?

Exercise 818 Let F be a field and let $A \in \mathcal{M}_{2 \times 2}(F)$ be a matrix satisfying $\text{tr}(A) = \text{tr}(A^2) = 0$. Is it necessarily true that $A = O$?

Exercise 819 Let k and n be positive integers. If $O \neq A \in \mathcal{M}_{k \times n}(\mathbb{R})$, does there necessarily exist a matrix $B \in \mathcal{M}_{n \times k}(\mathbb{R})$ satisfying $\text{tr}(AB) \neq 0$?

Exercise 820 Let F be a field and let $k \neq n$ be positive integers. Let $A \in \mathcal{M}_{k \times n}(F)$ and $B \in \mathcal{M}_{n \times k}(F)$. Are $\text{tr}(AB)$ and $\text{tr}(BA)$ necessarily equal?

Exercise 821 Show that the matrices
$$\begin{bmatrix} 1 & 2-i & 1+i \\ 4+i & 1+i & 0 \\ 1+i & 1 & 1 \end{bmatrix} \quad \text{and} \quad \begin{bmatrix} 1 & 1+i & 2-i \\ 3-i & 1+i & 0 \\ 1 & 27 & 1-i \end{bmatrix}$$
 in $\mathcal{M}_{3 \times 3}(\mathbb{C})$ are not similar.

Exercise 822 Let n be a positive integer and let V be the subspace of $\mathbb{R}[X]$ composed of all polynomials of degree at most n . What is the dual basis of $\{1, X, \dots, X^n\}$?

Exercise 823 Let V be a vector space over a field F having a maximal subspace W . Show that there exists a linear functional $\delta \in D(V)$ satisfying $\ker(\delta) = W$.

Exercise 824 Let $V = \mathbb{R}[X]$. For all $c \in \mathbb{R}$, let ε_c be the linear functional on V defined by $\varepsilon_c : p(X) \mapsto p(c)$. Is $\{\varepsilon_c \mid c \in \mathbb{R} \setminus \{0, 1\}\}$ a linearly-independent subset of $D(V)$?

Exercise 825 Let n be a positive integer. If B and C are elements of $\mathcal{M}_{n \times n}(\mathbb{R})$ satisfying $\text{tr}(B) \leq \text{tr}(C)$, and if $A \in \mathcal{M}_{n \times n}(\mathbb{R})$, is it necessarily true that $\text{tr}(AB) \leq \text{tr}(AC)$?

Exercise 826 For a matrix $A \in \mathcal{M}_{3 \times 3}(\mathbb{R})$, find a positive integer c satisfying $|A| = \frac{1}{c} \begin{vmatrix} \text{tr}(A) & 1 & 0 \\ \text{tr}(A^2) & \text{tr}(A) & 2 \\ \text{tr}(A^3) & \text{tr}(A^2) & \text{tr}(A) \end{vmatrix}$.

Exercise 827 Let k and n be positive integers and let F be a field. Define a function $\alpha : \mathcal{M}_{kn \times kn}(F) \rightarrow \mathcal{M}_{n \times n}(F)$ as follows: if $A \in \mathcal{M}_{kn \times kn}(F)$, write $A = [A_{ij}]$, where each A_{ij} is a $(k \times k)$ -block. Then set $\alpha(A) = [b_{ij}] \in \mathcal{M}_{n \times n}(F)$, where $b_{ij} = \text{tr}(A_{ij})$ for each $1 \leq i, j \leq n$. Is α a linear transformation? Is it a homomorphism of unital F -algebras?

Exercise 828 Let A be a nonempty set and let V be the collection of all subsets of A , which is a vector space over $GF(2)$. Is the characteristic function of $\emptyset \neq D \subseteq A$ a linear functional on V ?

Exercise 829 For each integer $n > 1$, find a nonsingular matrix $A \in \mathcal{M}_{n \times n}(\mathbb{Q})$ satisfying $\text{tr}(A) = 0$.

Exercise 830 Let $n > 1$ be an integer and let $A \in \mathcal{M}_{n \times n}(\mathbb{R})$. Find a symmetric matrix $B \in \mathcal{M}_{n \times n}(\mathbb{R})$ satisfying $\text{tr}(A) = \text{tr}(B)$.

Exercise 831 Let V be a vector space finitely generated over \mathbb{Q} and let $\alpha \in \text{End}(V)$ be a projection. Show that there is a basis D of V satisfying the condition that the rank of α equals $\text{tr}(\Phi_{DD}(\alpha))$.

Exercise 832 Let W be a proper subspace of a vector space V over a field F and let $v \in V \setminus W$. Show that there is a linear functional $\delta \in D(V)$ satisfying $\delta(v) \neq 0$ but $\delta(w) = 0$ for all $w \in W$.

Exercise 833 Let V be a vector space finitely generated over a field F and let W_1 and W_2 be proper subspaces of V satisfying $V = W_1 \oplus W_2$. Show that $D(V) = E_1 \oplus E_2$, where $E_j = \{\delta \in D(V) \mid W_j \subseteq \ker(\delta)\}$ for $j = 1, 2$.

Exercise 834 Let F be a field and let $A \in \mathcal{M}_{2 \times 2}(F)$. Show that we always have $A^2 - \text{tr}(A)A + |A|I = O$.

Exercise 835 Let V be the subspace of \mathbb{R}^∞ consisting of all sequences $[a_1, a_2, \dots] \in \mathbb{R}^\infty$ satisfying the condition that $\lim_{i \rightarrow \infty} a_i$ exists in \mathbb{R} . Define linear functionals $\delta_1, \delta_2, \dots, \delta_\infty \in D(V)$ by setting $\delta_h : [a_1, a_2, \dots] \mapsto a_h$ for each $h = 1, 2, \dots$ and $\delta_\infty : [a_1, a_2, \dots] \mapsto \lim_{i \rightarrow \infty} a_i$. Is the subset $\{\delta_1, \delta_2, \dots, \delta_\infty\}$ of $D(V)$ necessarily linearly independent?

Exercise 836 Let F be a field and, for each $a \in F$, let $\varepsilon_a : F[X] \rightarrow F$ be the linear functional defined by $\varepsilon_a : p(X) \mapsto p(a)$. Show that the subset $\{\varepsilon_a \mid a \in F\}$ of $D(F[X])$ is linearly independent.

Exercise 837 Let V be a vector space over a field F and let $\delta_1, \delta_2 \in D(V)$ be linear functionals satisfying the condition that $\delta_1(v)\delta_2(v) = 0$ for all $v \in V$. Show that one of the δ_i must be the 0-functional.

Exercise 838 Let $n > 1$ be an integer and let $f : \mathbb{R}^n \rightarrow \mathbb{R}$ be a continuous function which maps the 0-vector to 0 and which satisfies the condition that $f(v+w) + f(v-w) = 2f(v)$ for all $v, w \in \mathbb{R}^n$. Show that $f \in D(\mathbb{R}^n)$.

Exercise 839 Let $a \in \mathbb{R}$, let n be a positive integer, and let $A, B \in \mathcal{M}_{n \times n}(\mathbb{R})$. Does there necessarily exist a matrix $C \in \mathcal{M}_{n \times n}(\mathbb{R})$ satisfying $AC + \text{tr}(C)A = B$?

Exercise 840 Let F be a field and let n be a positive integer. Let $\delta : \mathcal{M}_{n \times n}(F) \rightarrow F$ be the linear functional given by $\delta : [a_{ij}] \mapsto \sum_{i=1}^n \sum_{j=1}^n a_{ij}$. Find an endomorphism α of $\mathcal{M}_{n \times n}(F)$ satisfying the condition that $\delta(A) = a \cdot \text{tr}(\alpha(A))$ for all $A \in \mathcal{M}_{n \times n}(F)$.

Exercise 841 Let F be a field and let n be a positive integer. Let $A, B \in \mathcal{M}_{n \times n}(F)$ be matrices satisfying $A^2 + B^2 = I$ and $AB + BA = O$. Show that $\text{tr}(A) = \text{tr}(B) = 0$.

Exercise 842 Let F be a field and let n be a positive integer. Given a positive integer k , is it necessarily true that $\text{tr}(AB)^k = \text{tr}(A^k)\text{tr}(B^k)$ for all $A, B \in \mathcal{M}_{n \times n}(F)$?

Exercise 843 Let V be a vector space finitely generated over a field F and let $\alpha \in \text{End}(V)$. Show that α and $D(\alpha)$ have identical minimal polynomials.

Exercise 844 Let V be a vector space over a field F and let n be a positive integer. Let v_1, \dots, v_n be distinct vectors in V and assume that there exist $\alpha \in \text{End}(V)$ and $\delta \in D(V)$ such that the matrix $[\delta\alpha^{i-1}(v_j)] \in \mathcal{M}_{n \times n}(F)$ is nonsingular. Show that the set $\{v_1, \dots, v_n\}$ is linearly independent.

Exercise 845 Let W be a subspace of a vector space V over a field F . Show that W is a maximal subspace of V if and only if every complement of W in V has dimension 1.

Exercise 846 Let F be a field and let n be a positive integer. For matrices $A, B \in \mathcal{M}_{n \times n}(F)$, calculate $\text{tr}([AB - BA][AB + BA])$.

Exercise 847 Let F be a field and let n be a positive integer. For matrices $A, B \in \mathcal{M}_{n \times n}(F)$, do we necessarily have $\text{tr}((AB)^2) = \text{tr}(A^2B^2)$?

Exercise 848 Let n be a positive integer. Can we find matrices $A, B \in \mathcal{M}_{n \times n}(\mathbb{C})$ satisfying the condition that all eigenvalues of A and of B are positive real numbers, but not all eigenvalues of $A + B$ are positive real numbers?

Exercise 849 Let k and n be positive integers, let F be a field, and let $O \neq A \in \mathcal{M}_{k \times n}(F)$. Does there necessarily exist a matrix $B \in \mathcal{M}_{n \times k}(F)$ satisfying $\text{tr}(AB) \neq 0$.

Exercise 850 Let V be a vector space of finite dimension n over a field F . A nonempty finite collection $\{W_1, \dots, W_k\}$ of hyperplanes of V is **co-independent** if and only if $\dim\left(\bigcap_{i=1}^k W_i\right) = n - k$. Is a nonempty subcollection of an co-independent collection of hyperplanes necessarily co-independent?

15

Inner product spaces

In this chapter, we will have to restrict the set of fields over which we work. A subfield F of \mathbb{R} is **real euclidean** if and only if for each $0 \leq c \in F$ there exists an element $d \in F$ satisfying $d^2 = c$ and a subfield K of \mathbb{C} is **euclidean** if and only if there exists a real euclidean field F such that $K = \{a + bi \mid a, b \in F\}$. It is immediately clear that if K is a euclidean field and $c \in K$, then $\bar{c} \in K$. Being a euclidean field is intimately tied in with the constructibility of elements of the complex plane by straightedge and compass constructions, and in fact every real euclidean field must contain all those real numbers which are then lengths of line segments obtainable from the unit line segment by straightedge and compass construction methods. Clearly \mathbb{R} itself is real euclidean, while \mathbb{Q} , as we have already noted, is not; the set real numbers algebraic over \mathbb{Q} is real euclidean and properly contained in \mathbb{R} . The field \mathbb{C} is euclidean, and set of all algebraic numbers is euclidean and properly contained in \mathbb{C} .

Let V be a vector space over an euclidean field F . A function μ from $V \times V$ to F is an **inner product** on V if and only if:

- (1) For each $w \in V$, the function $v \mapsto \mu(v, w)$ from V to F is a linear functional;
- (2) If $v, w \in V$ then $\mu(v, w) = \overline{\mu(w, v)}$;
- (3) If $v \in V$ then $\mu(v, v)$ is a nonnegative real number, which equals 0 if and only if $v = 0_V$.

Note that, in the above situation, if $v, w \in V$ then, as a consequence of (2), $\mu(v, w) + \mu(w, v) = 2 \operatorname{Re}(\mu(v, w))$ is also always a real number, though it may of course be negative.

In general, once we have fixed an inner product on a space, we will write $\langle v, w \rangle$ instead of $\mu(v, w)$. A vector space over an euclidean subfield F of \mathbb{C} on which we have an inner product defined is called an **inner product space**. Another term for such spaces, coming from functional analysis, is a **pre-Hilbert spaces**. Abstract inner product spaces were first studied in an axiomatic manner by Von Neumann. While inner product spaces over general euclidean fields may prove to be interesting in the future, at the moment the study of such spaces is almost universally restricted to spaces over \mathbb{R} or \mathbb{C} , and so from now on we will do the same and consider only these as possible fields of scalars. When we talk about an inner product space without specifying the field of scalars, we will always assume that it is one of these two fields.

Example: Let n be a positive integer and let F be either \mathbb{R} or \mathbb{C} . We define an inner product on F^n , called the **dot product**, as

follows: if $v = \begin{bmatrix} a_1 \\ \vdots \\ a_n \end{bmatrix}$ and $w = \begin{bmatrix} b_1 \\ \vdots \\ b_n \end{bmatrix}$, then we set $v \cdot w = \sum_{i=1}^n a_i \bar{b}_i$.

Note that if $F = \mathbb{R}$, then this product just coincides with the interior product $v \odot w$ which defined earlier. However, that is not true for the case $F = \mathbb{C}$, so we must be very careful to distinguish between the two products. This modification of the definition is necessary since, over \mathbb{C} ,

we have $\begin{bmatrix} 1 \\ i \end{bmatrix} \odot \begin{bmatrix} 1 \\ i \end{bmatrix} = 0$, even though $\begin{bmatrix} 1 \\ i \end{bmatrix} \neq \begin{bmatrix} 0 \\ 0 \end{bmatrix}$. Hence the

interior product \odot is not an inner product as we have defined it in this chapter¹.

We can generalize the previous example. If F is either \mathbb{R} or \mathbb{C} , and if $D = [d_{ij}]$ is a nonsingular matrix in $\mathcal{M}_{n \times n}(F)$, we can define an inner



¹ The problem arises because, in \mathbb{C} , 0 can be written as the sum of squares of nonzero elements. A field F in which 0 cannot be the sum of squares of nonzero elements of F is **formally real**; so \mathbb{R} is formally real while \mathbb{C} is not. The theory of formally real fields was developed in the 1920's by the Austrian mathematicians **Emil Artin** and **Otto Schreier**.

product on F^n by setting

$$\left\langle \begin{bmatrix} a_1 \\ \vdots \\ a_n \end{bmatrix}, \begin{bmatrix} b_1 \\ \vdots \\ b_n \end{bmatrix} \right\rangle = [a_1 \ \dots \ a_n] D D^H \begin{bmatrix} \bar{b}_1 \\ \vdots \\ \bar{b}_n \end{bmatrix},$$

where $D^H = [\bar{d}_{ij}]^T$. The matrix D^H is called the **conjugate transpose** or **Hermitian transpose** of D , and it again belongs to $\mathcal{M}_{n \times n}(F)$. Conjugate transposes of matrices over \mathbb{C} will play an important part in the following discussion; of course, $D^H = D^T$ for any matrix $D \in \mathcal{M}_{n \times n}(\mathbb{R})$.

The properties of the conjugate transpose are very much like those of the transpose. Indeed, we note that if $A, B \in \mathcal{M}_{n \times n}(\mathbb{C})$ and $c \in \mathbb{C}$, then $(A + B)^H = A^H + B^H$, $(cA)^H = \bar{c}A^H$, $A^{HH} = A$, and $(AB)^H = B^H A^H$. In particular, if A is nonsingular then $I = I^H = (AA^{-1})^H = (A^{-1})^H A^H$, proving that $(A^{-1})^H = (A^H)^{-1}$.

Example: If we are given positive real numbers c_1, \dots, c_n and consider the diagonal matrix $D = [d_{ij}] \in \mathcal{M}_{n \times n}(\mathbb{R})$ the diagonal entries of which are given by $d_{ii} = \sqrt{c_i}$ for $1 \leq i \leq n$, then, by the above, we have an inner product on \mathbb{C}^n given by

$$\left\langle \begin{bmatrix} a_1 \\ \vdots \\ a_n \end{bmatrix}, \begin{bmatrix} b_1 \\ \vdots \\ b_n \end{bmatrix} \right\rangle = \sum_{i=1}^n c_i a_i \bar{b}_i.$$

Such a product is called a **weighted dot product**. Weighted dot products are extremely important in statistics and data analysis, where we often want to emphasize the values of certain parameters and de-emphasize others.

Example: Let $a < b$ be real numbers and let $V = C(a, b)$. This is, as we have seen, a vector space over \mathbb{R} , on which we can define an inner product $\langle f, g \rangle = \int_a^b f(x)g(x)dx$. Continuity is important here. The set Y of all functions from $[a, b]$ to \mathbb{R} which are continuous at all but finitely-many points is a subspace of $\mathbb{R}^{[a, b]}$ properly containing $C(a, b)$ but $\langle f, g \rangle = \int_a^b f(x)g(x)dx$ is not an inner product on Y . Indeed, if we select a real number c satisfying $a < c < b$ and define the function $f \in Y$ by

$$f : x \mapsto \begin{cases} 1 & \text{if } x = c \\ 0 & \text{otherwise} \end{cases}$$

then f is a nonzero element of Y but $\langle f, f \rangle = 0$.

Similarly, if V be the set of all continuous complex-valued functions defined on the closed interval $[a, b]$ in \mathbb{R} , then V is a vector space over \mathbb{C} , on which we can define an inner product $\langle f, g \rangle = \int_a^b f(x)\overline{g(x)}dx$.

Example: Let F be \mathbb{R} or \mathbb{C} , and let $V = \mathcal{M}_{n \times n}(F)$, which is a vector space over F . Define an inner product on V by setting $\langle A, B \rangle = \text{tr}(AB^H) = \text{tr}(B^H A)$. If $A = [a_{ij}]$ and $B = [b_{ij}]$, then $\langle A, B \rangle = \sum_{i=1}^n \sum_{j=1}^n a_{ij} \bar{b}_{ij}$.

Example: Let V be the subspace of \mathbb{C}^∞ composed of all those sequences $[c_0, c_1, \dots]$ of complex numbers satisfying $\sum_{i=0}^\infty |c_i|^2 < \infty$. This vector space is very important in analysis, and we can define an inner product on it by setting $\langle [c_0, c_1, \dots], [d_0, d_1, \dots] \rangle = \sum_{i=0}^\infty c_i \bar{d}_i$.

Let F be \mathbb{R} or \mathbb{C} , and let W be a subspace of an inner product space V over F . The restriction of this inner product to a function from $W \times W$ to F is an inner product on W . Thus we can always assume that any subspace of an inner product space V inherits the inner-product-space structure of V .

Example: Let V be an inner product space over \mathbb{R} and let K be the set of all matrices of the form $\begin{bmatrix} a & v \\ w & b \end{bmatrix}$, where $a, b \in \mathbb{R}$ and $v, w \in V$. Then K is a vector space over \mathbb{R} , where addition and scalar multiplication are defined by $\begin{bmatrix} a & v \\ w & b \end{bmatrix} + \begin{bmatrix} a' & v' \\ w' & b' \end{bmatrix} = \begin{bmatrix} a + a' & v + v' \\ w + w' & b + b' \end{bmatrix}$ and $c \begin{bmatrix} a & v \\ w & b \end{bmatrix} = \begin{bmatrix} ca & cv \\ cw & cb \end{bmatrix}$. We create the structure of an \mathbb{R} -algebra on K by defining an operation \bullet as follows: $\begin{bmatrix} a & v \\ w & b \end{bmatrix} \bullet \begin{bmatrix} a' & v' \\ w' & b' \end{bmatrix} = \begin{bmatrix} aa' + \langle v, w' \rangle & av' + b'v \\ a'w + bw' & bb' + \langle w, v' \rangle \end{bmatrix}$. This algebra is a division algebra, called the **Cayley algebra**, and it is not associative. Indeed, it is the only nonassociative division algebra of finite dimension over \mathbb{R} .

We now look at some properties of general inner product spaces.

(15.1) Proposition: Let V be an inner product space. For $v, w_1, w_2 \in V$ and for a scalar a , we have:

- (1) $\langle v, w_1 + w_2 \rangle = \langle v, w_1 \rangle + \langle v, w_2 \rangle$;
- (2) $\langle v, aw_1 \rangle = a \langle v, w_1 \rangle$;
- (3) $\langle 0_V, w_1 \rangle = \langle v, 0_V \rangle = 0$.

Proof: From the definition of the inner product, we have

$$\begin{aligned} \langle v, w_1 + w_2 \rangle &= \overline{\langle w_1 + w_2, v \rangle} = \overline{\langle w_1, v \rangle + \langle w_2, v \rangle} \\ &= \overline{\langle v, w_1 \rangle} + \overline{\langle v, w_2 \rangle} = \langle v, w_1 \rangle + \langle v, w_2 \rangle, \end{aligned}$$

which proves (1). We also have

$$\langle v, aw_1 \rangle = \overline{\langle aw_1, v \rangle} = \overline{a \langle w_1, v \rangle} = \bar{a} \overline{\langle w_1, v \rangle} = \bar{a} \langle v, w_1 \rangle,$$

which proves (2). Finally, $\langle 0_V, w_1 \rangle = \langle 00_V, w_1 \rangle = 0$, $\langle 0_V, w_1 \rangle = 0$, and similarly $\langle v, 0_V \rangle = 0$, proving (3). \square

By Proposition 15.1, we see that if V is an inner product space over \mathbb{R} then for each $v \in V$ the function $w \mapsto \langle v, w \rangle$ from V to F is a linear transformation, but that is not the case for inner product spaces over \mathbb{C} .

Let V be a finitely-generated inner product space, having a basis $\{v_1, \dots, v_n\}$. Given vectors $v = \sum_{i=1}^n a_i v_i$ and $w = \sum_{j=1}^n b_j v_j$ in V , we have

$$\langle v, w \rangle = \sum_{i=1}^n \sum_{j=1}^n a_i \bar{b}_j \langle v_i, v_j \rangle = \begin{bmatrix} a_1 & \dots & a_n \end{bmatrix} G \begin{bmatrix} \bar{b}_1 \\ \vdots \\ \bar{b}_n \end{bmatrix},$$

where $G = [g_{ij}]$ is the matrix defined by $g_{ij} = \langle v_i, v_j \rangle$ for all $1 \leq i, j \leq n$. This matrix is known as the **Gram² matrix** defined by the given basis.

Example: Let V be the subspace of $\mathbb{C}[X]$ consisting of all polynomials of degree at most 5, and let B be the canonical basis for V . Define an inner product on V by setting $\langle f, g \rangle = \int_0^1 f(x) \overline{g(x)} dx$. (Note that we are using the same notation for a polynomial and its corresponding polynomial function in $\mathbb{C}^{\mathbb{R}}$.) Then the Gram matrix defined by B is precisely the Hilbert matrix H_6 , which we have seen earlier.



² **Jorgen Gram** was a Danish mathematician who, at the end of the 19th century, developed computational techniques for inner product spaces in connection with his work for insurance companies.

(15.2) Proposition (Cauchy-Schwarz-Bunyakovsky³ Theorem): Let V be an inner product space. If $v, w \in V$, then $|\langle v, w \rangle|^2 \leq \langle v, v \rangle \langle w, w \rangle$.

Proof: If $v = 0_V$ or $w = 0_V$ then the result is immediate, and so we can assume that both vectors differ from 0_V . Let $a = -\langle w, v \rangle$ and $b = \langle v, v \rangle$. Then $\bar{a} = -\langle v, w \rangle$ and $\bar{b} = b$ so

$$\begin{aligned} 0 &\leq \langle av + bw, av + bw \rangle = a\bar{a} \langle v, v \rangle + ab \langle v, w \rangle + b\bar{a} \langle w, v \rangle + b^2 \langle w, w \rangle \\ &= a\bar{a}b - a\bar{a}b - ab\bar{a} + b^2 \langle w, w \rangle = b[-a\bar{a} + b \langle w, w \rangle]. \end{aligned}$$

Since $v \neq 0_V$, it follows that b is a positive real number and so $a\bar{a} \leq b \langle w, w \rangle$, which is what we want. \square

Example: If $a_1, \dots, a_n, b_1, \dots, b_n, c_1, \dots, c_n$ are real numbers, with $c_i > 0$ for all $1 \leq i \leq n$, then

$$\left| \sum_{i=1}^n c_i a_i b_i \right| \leq \left(\sqrt{\sum_{i=1}^n c_i a_i^2} \right) \left(\sqrt{\sum_{i=1}^n c_i b_i^2} \right).$$

Indeed, this is a consequence of the Cauchy-Schwarz-Bunyakovsky Theorem, using the weighted dot product

$$\left\langle \begin{bmatrix} a_1 \\ \vdots \\ a_n \end{bmatrix}, \begin{bmatrix} b_1 \\ \vdots \\ b_n \end{bmatrix} \right\rangle = \sum_{i=1}^n c_i a_i b_i$$

defined on \mathbb{R}^n .

In general, the Cauchy-Schwarz-Bunyakovsky Theorem is an extremely rich source of inequalities between real-valued functions of several real variables. For example, consider the vectors

$$v = \frac{1}{\sqrt{a+b+c}} \begin{bmatrix} \sqrt{a+b} \\ \sqrt{a+c} \\ \sqrt{b+c} \end{bmatrix} \quad \text{and} \quad w = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$



3

Herman Schwarz was a German mathematician who, in the late 19th century, studied spaces of functions and their structure as inner product spaces. **Viktor Yakovlevich Bunyakovsky** was a Russian student of Cauchy who proved this theorem a generation before Schwarz, but since his work was published in an obscure journal, it was not widely recognized until the 20th century.

in \mathbb{R}^3 , where a , b , and c are positive. Then, by the Cauchy-Schwarz-Bunyakovsky Theorem, we see that

$$\sqrt{\frac{a+b}{a+b+c}} + \sqrt{\frac{a+c}{a+b+c}} + \sqrt{\frac{b+c}{a+b+c}} = v \cdot w \leq \sqrt{\langle v, v \rangle \langle w, w \rangle} = \sqrt{6}.$$

Similarly, we note that the matrix $D = \begin{bmatrix} \sqrt{3} & 0 \\ 1 & \sqrt{2} \end{bmatrix} \in \mathcal{M}_{2 \times 2}(\mathbb{R})$ is nonsingular and so, by a previous example, we have an inner product μ on \mathbb{R}^2 defined by

$$\mu \left(\begin{bmatrix} a \\ b \end{bmatrix}, \begin{bmatrix} c \\ d \end{bmatrix} \right) = \begin{bmatrix} a \\ b \end{bmatrix}^T D D^T \begin{bmatrix} c \\ d \end{bmatrix} = 3(ac + bd) + (\sqrt{3})(ad + bc).$$

Applying the Cauchy-Schwarz-Bunyakovsky Theorem, we see that for all real numbers a , b , c , and d we have

$$\begin{aligned} & \left[3(ac + bd) + (\sqrt{3})(ad + bc) \right]^2 \\ & \leq \left[3(a^2 + b^2) + (2\sqrt{3})ab \right] \left[3(c^2 + d^2) + (2\sqrt{3})cd \right]. \end{aligned}$$

In particular, if we take $b = d = \sqrt{3}$, we see that

$$(ac + a + c + 3)^2 \leq (a^2 + 2a + 3)(c^2 + 2c + 3)$$

for all real numbers a and c .

Let V be an inner product space. The **norm** of a vector $v \in V$ is defined to be the scalar $\|v\| = \sqrt{\langle v, v \rangle}$. A vector v satisfying $\|v\| = 1$ is **normal**.

Example: Let $V = \mathbb{R}^n$, and endow V with the dot product. Then

$$\left\| \begin{bmatrix} a_1 \\ \vdots \\ a_n \end{bmatrix} \right\| = \sqrt{\sum_{i=1}^n a_i^2}. \text{ This norm is known as the } \mathbf{euclidean \ norm} \text{ on } V.$$

Example: Let $V = C(-\pi, \pi)$, on which we have defined the inner product $\langle f, g \rangle = \int_{-\pi}^{\pi} f(x)g(x)dx$. For each positive integer k , consider the function $f_k : x \mapsto \sin(kx)$. Then

$$\|f_k\| = \sqrt{\langle f_k, f_k \rangle} = \sqrt{\int_{-\pi}^{\pi} \sin^2(kx)dx} = \sqrt{\pi}$$

and so $g_k = \frac{1}{\sqrt{\pi}}f_k$ is a normal vector in this space.

We have seen how the vector space \mathbb{R}^3 , endowed the cross product \times , is a Lie algebra. It is easy to check that the cross product is related to the dot product on \mathbb{R}^3 by the relations

$$u \times (v \times w) = (u \cdot w)v - (u \cdot v)w \quad \text{and} \quad (u \times v) \times w = (u \cdot w)v - (v \cdot w)u$$

for all $u, v, w \in \mathbb{R}^3$. Moreover, we have the following identities:

- (1) $v \cdot (v \times w) = 0$ for all $v, w \in \mathbb{R}^3$;
- (2) (**Lagrange identity**) $\|v \times w\|^2 = \|v\|^2 \|w\|^2 - (v \cdot w)^2$.

There are only two possible anticommutative operations on \mathbb{R}^3 which turn it into an \mathbb{R} -algebra satisfying these two identities, namely \times and the operation \times' given by $v \times' w = -(v \times w)$. Furthermore, if $n > 3$ no such operation can be defined on \mathbb{R}^n , except for the case of $n = 7$. In that case, we can define an operation \times as follows: write elements of \mathbb{R}^7

in the form $\begin{bmatrix} v \\ a \\ v' \end{bmatrix}$, where $v, v' \in \mathbb{R}^3$ and $a \in \mathbb{R}$ and then set

$$\begin{bmatrix} v \\ a \\ v' \end{bmatrix} \times \begin{bmatrix} w \\ b \\ w' \end{bmatrix} = \begin{bmatrix} aw' - bv' + (v \times w) - (v' \times w') \\ -v \cdot w + v' \cdot w \\ bv - aw + (v \times w') - (v' \times w) \end{bmatrix}.$$

We also note that if $u = \begin{bmatrix} a_1 \\ a_2 \\ a_3 \end{bmatrix}$, $v = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix}$ and $w = \begin{bmatrix} c_1 \\ c_2 \\ c_3 \end{bmatrix}$ in

\mathbb{R}^3 then $u \cdot (v \times w) = \begin{vmatrix} a_1 & b_1 & c_1 \\ a_2 & b_2 & c_2 \\ a_3 & b_3 & c_3 \end{vmatrix}$. As an immediate consequence, we

observe that if $u, v, w \in \mathbb{R}^3$ then:

- (1) $u \cdot (v \times w) = v \cdot (w \times u) = w \cdot (u \times v)$;
- (2) $u \cdot (v \times w) = 0$ if and only if two of these vectors are equal or the set $\{u, v, w\}$ is linearly dependent.

The scalar value $u \cdot (v \times w)$ is often called the **scalar triple product** of the vectors u, v, w , to distinguish it from the **vector triple product** $u \times (v \times w)$.

(15.3) Proposition: Let V be an inner product space. If $v, w \in V$ and if a is a scalar, then:

- (1) $\|av\| = |a| \cdot \|v\|$;
- (2) $\|v\| \geq 0$, with equality if and only if $v = 0_V$;

- (3) (**Minkowski's⁴ inequality**): $\|v + w\| \leq \|v\| + \|w\|$;
 (4) (**Parallelogram law**):

$$\|v + w\|^2 + \|v - w\|^2 = 2(\|v\|^2 + \|w\|^2);$$

- (5) (**Triangle difference inequality**): $\|v - w\| \geq \left| \|v\| - \|w\| \right|$.

Proof: We see that $\|av\| = \sqrt{\langle av, av \rangle} = \sqrt{a\bar{a}\langle v, v \rangle} = |a| \cdot \|v\|$, proving (1). Inequality (2) follows immediately from the definition. As for (3), note that if $z = a + bi$ then $z + \bar{z} = 2a \leq 2|a| = 2\sqrt{a^2} \leq 2\sqrt{a^2 + b^2} = 2|z|$. From the Cauchy-Schwarz-Bunyakovsky Theorem, it follows that

$$|\langle v, w \rangle| = |\langle w, v \rangle| \leq \|v\| \cdot \|w\|$$

and so

$$\begin{aligned} \|v + w\|^2 &= \langle v + w, v + w \rangle = \langle v, v \rangle + \langle v, w \rangle + \langle w, v \rangle + \langle w, w \rangle \\ &\leq \|v\|^2 + 2\|v\| \cdot \|w\| + \|w\|^2 = (\|v\| + \|w\|)^2 \end{aligned}$$

and that proves (3). Moreover, we know that

$$\|v + w\|^2 = \langle v + w, v + w \rangle = \langle v, v \rangle + \langle v, w \rangle + \langle w, v \rangle + \langle w, w \rangle$$

and $\|v - w\|^2 = \langle v - w, v - w \rangle = \langle v, v \rangle - \langle v, w \rangle - \langle w, v \rangle + \langle w, w \rangle$. Adding these two gives us (4).

Finally, by (3) we have $\|w\| = \|w + (v - w)\| \leq \|w\| + \|v - w\|$, and so $\|v - w\| \geq \|v\| - \|w\|$. Interchanging the roles of v and w and using (1), gives us $\|v - w\| = \|w - v\| \geq \|w\| - \|v\|$, and so we have (5). \square

Note that by Proposition 15.3 we see that if $0_V \neq v \in V$ then $\frac{1}{\|v\|}v$ is a normal vector. Moreover, if v is normal and c is a scalar satisfying $|c| = 1$, then cv is again normal.

Example: Let V be an inner product space, and let Ω be a nonempty set. A function $f \in V^\Omega$ is **bounded** if and only if there exists a real



4

Hermann Minkowski, a German mathematician at the end of the 19th century, built an elegant mathematical framework for the theory of relativity, using four-dimensional noneuclidean geometry.

number b_f satisfying $\|f(i)\| \leq b_f$ for all $i \in \Omega$. If $f, g \in V^\Omega$ are bounded functions then, from Minkowski's inequality, we see that

$$\|(f+g)(i)\| \leq \|f(i)\| + \|g(i)\| \leq b_f + b_g$$

for all $i \in \Omega$. If c is a scalar then $\|(cf)(i)\| = |c| \cdot \|f(i)\| \leq |c|b_f$ for all $i \in \Omega$. Thus both $f+g$ and cf are both bounded, and we see that the set of all bounded elements of V^Ω is a subspace of V^Ω .

Example: We now return to a previous example. Let p be an integer greater than 1, not necessarily prime, and let $G = \mathbb{Z}/(p)$, on which we have an operation of addition as defined in Chapter 2. Let $V = \mathbb{C}^G$, which is a vector space of dimension p over \mathbb{C} . On this space, we can define an inner product by setting $\langle f, g \rangle = \sum_{n \in G} f(n)\overline{g(n)}$. Every element $n \in G$ defines a function $h_n : k \mapsto \cos\left(\frac{2\pi nk}{p}\right) + i \sin\left(\frac{2\pi nk}{p}\right)$ which belongs to V . Given a function $f \in V$, define a function $\hat{f} \in V$ as $\hat{f} : n \mapsto \langle f, h_n \rangle = \sum_{k \in G} f(k)h_n(-k)$. This function is called the **discrete Fourier transform** of f of order p . One can show that the function $f \mapsto \hat{f}$ is in fact an automorphism of V . Moreover, $f(n) = \frac{1}{p} \hat{\hat{f}}(-n)$ and $\|f\| = \frac{1}{\sqrt{p}} \|\hat{f}\|$ for all $f \in V$ and all $n \in G$.

Example: There are various generalizations of Theorem 15.2 which, as a rule, require more sophisticated methods of complex analysis to prove. For example, the contemporary Greek mathematicians Manolis Magiropoulos and Dimitri Karayannakis have shown that if V is an inner product space and if u, v , and w are distinct elements of V , then

$$2|\langle u, v \rangle| \cdot |\langle u, w \rangle| \leq \langle u, u \rangle [\|v\| \cdot \|w\| + |\langle v, w \rangle|].$$

In case the set $\{v, w\}$ is linearly dependent, it is clear that this reduces to the inequality in Proposition 15.2. Inequalities such as these allow us to get better bounds on inner products. For example, let $0 < a < b$ are real numbers and let $V = C(a, b)$, on which we have the inner product $\langle f, g \rangle = \int_a^b f(x)g(x)dx$. If $u, v, w \in V$ are given by

$$u : x \mapsto \frac{1}{x}, \quad v : x \mapsto \sin(x), \quad \text{and} \quad w : x \mapsto \cos(x)$$

then Proposition 15.2 gives us the bound

$$|\langle u, v \rangle| \cdot |\langle u, w \rangle| \leq \left(\int_a^b \frac{dx}{x^2} \right) \sqrt{\int_a^b \sin^2(x)dx} \sqrt{\int_a^b \cos^2(x)dx}$$

whereas this result gives us the better upper bound

$$\frac{1}{2} \left(\int_a^b \frac{dx}{x^2} \right) \left[\sqrt{\int_a^b \sin^2(x)dx} \sqrt{\int_a^b \cos^2(x)dx} + \left| \int_a^b \sin(x) \cos(x)dx \right| \right].$$

(15.4) Proposition: Let V be an inner product space and let $\alpha \in \text{End}(V)$ satisfy the condition that there exists a real number $0 < c < 1$ such that $\|\alpha(v)\| \leq c\|v\|$ for all $v \in V$. Then $\sigma_1 + \alpha$ is monic.

Proof: If $0_V \neq v \in V$ then, by Proposition 15.3,

$$\begin{aligned}\|v\| &= \|v + \alpha(v) - \alpha(v)\| \leq \|v + \alpha(v)\| + \|\alpha(v)\| \\ &= \|(\sigma_1 + \alpha)(v)\| + \|\alpha(v)\|\end{aligned}$$

and so $\|(\sigma_1 + \alpha)(v)\| \geq \|v\| - \|\alpha(v)\| \geq (1 - c)\|v\| > 0$, which shows that $v \notin \ker(\sigma_1 + \alpha)$. Thus $\sigma_1 + \alpha$ is monic. \square

In particular, if V is a finitely-generated inner product space and if $\alpha \in \text{End}(V)$ satisfies the condition that there exists a real number $0 < c < 1$ such that $\|\alpha(v)\| \leq c\|v\|$ for all $v \in V$, then $\sigma_1 + \alpha \in \text{Aut}(V)$. Let $\beta = (\sigma_1 + \alpha)^{-1}$. If $0_V \neq v \in V$ then

$$\begin{aligned}\|v\| &= \|(\sigma_1 + \alpha)\beta(v)\| = \|\beta(v) + \alpha\beta(v)\| \geq \|\beta(v)\| - \|\alpha\beta(v)\| \\ &\geq \|\beta(v)\| - c\|\beta(v)\| = (1 - c)\|\beta(v)\|.\end{aligned}$$

Similarly, $\|v\| \leq \|\beta(v)\| + \|\alpha\beta(v)\| \leq \|\beta(v)\| + c\|\beta(v)\| = (1 + c)\|\beta(v)\|$ and so $\frac{1}{1+c}\|v\| \leq \|\beta(v)\| \leq \frac{1}{1-c}\|v\|$ for all $v \in V$.

Sometimes, however, we need a bit more generality. If V is a vector space over \mathbb{R} or \mathbb{C} then, in general, a function $v \mapsto \|v\|$ satisfying conditions (1) - (3) of Proposition 15.3 is called a **norm** and a vector space on which a fixed norm is defined is called a **normed space** or, in a functional-analysis context, a **pre-Banach space**⁵. An immediate question is whether every norm defined on a vector space comes from an inner product. The answer is negative: if, for example, we define the

norm $\|\cdot\|_1$ on \mathbb{C}^n by setting $\left\| \begin{bmatrix} a_1 \\ \vdots \\ a_n \end{bmatrix} \right\|_1 = \sum_{i=1}^n |a_i|$, then this cannot



5

Normed spaces were first studied at the beginning of the 20th century by the Austrian mathematician **Hans Hahn**, and then by the American mathematician **Norbert Wiener** and the Polish mathematician **Stefan Banach**.

come from an inner product since the parallelogram law is not satisfied by this norm. In fact, satisfying the parallelogram law is necessary for a norm to come from an inner product in the following sense: let V be a vector space over \mathbb{R} or \mathbb{C} on which we have a norm $\psi : V \rightarrow \mathbb{R}$ satisfying $\psi(v+w)^2 + \psi(v-w)^2 = 2[\psi(v)^2 + \psi(w)^2]$ for all $v, w \in V$, and write $\lambda(v, w) = \frac{1}{4} [\psi(v+w)^2 - \psi(v-w)^2]$. Then it is possible to define an inner product on V relative to which the norm of a vector v is precisely $\psi(v)$. In the case the field of scalars is \mathbb{R} , then this inner product is defined by $\langle v, w \rangle = \lambda(v, w)$ and otherwise this inner product is defined by $\langle v, w \rangle = \lambda(v, w) + i\lambda(v, iw)$.

We should point out, however, that for vector spaces V finitely generated over \mathbb{R} or \mathbb{C} , all norms are equivalent, in the sense that if $\|\cdot\|_a$ and $\|\cdot\|_b$ are norms defined on V then there exist positive real numbers c and d such that $c\|v\|_a \leq \|v\|_b \leq d\|v\|_a$ for all $v \in V$. For vector spaces which are not finitely generated the situation is different, as the following example shows.

Example: Let $V = C(0, 1)$, which is a vector space over \mathbb{R} , and for each positive integer n , let $f_n \in V$ be the function defined by

$$f_n : x \mapsto \begin{cases} 1 - nx & \text{if } 0 \leq x \leq \frac{1}{n} \\ 0 & \text{otherwise} \end{cases}.$$

Let $\|\cdot\|$ be the norm defined on V by the inner product $\langle f, g \rangle = \int_0^1 f(x)g(x)dx$ and let $\|\cdot\|_\infty$ be the norm on V defined by $\|f\|_\infty = \sup\{|f(x)| \mid 0 \leq x \leq 1\}$. Then $\|f_n\| = \frac{1}{\sqrt{3n}}$ for all positive integers n , whereas $\|f_n\|_\infty = 1$ for all positive integers n . Thus there can be no real number c satisfying $\|f\|_\infty \leq c\|f\|$ for all $f \in V$.

Example: Let V and W be normed spaces over the same field of scalars F (which is either \mathbb{R} or \mathbb{C}). If $\alpha \in \text{Hom}(V, W)$, set

$$\|\alpha\| = \sup \left\{ \frac{\|\alpha(v)\|}{\|v\|} \mid 0_V \neq v \in V \right\}$$

where the norm in the numerator is the one defined on W and the norm in the denominator is the one defined on V . We claim that this is a norm defined on $\text{Hom}(V, W)$, called norm **induced** by the respective norms on V and W . Indeed, as an immediate consequence of the definition we see that $\|\alpha\| \geq 0$ for all $\alpha \in \text{Hom}(V, W)$, with equality happening only when α is the 0-function. We also note that if α is not the 0-function then $\|\alpha\|$ is the smallest positive real number c such that $\|\alpha(v)\| \leq c\|v\|$ for all $v \in V$.

Now let $\alpha \in \text{Hom}(V, W)$ and $a \in F$. Then

$$\begin{aligned}\|\alpha\alpha\| &= \sup \left\{ \frac{\|a\alpha(v)\|}{\|v\|} \mid 0_V \neq v \in V \right\} \\ &= \sup \left\{ \frac{|a| \cdot \|\alpha(v)\|}{\|v\|} \mid 0_V \neq v \in V \right\} = |a| \cdot \|\alpha\|.\end{aligned}$$

Finally, if $\alpha, \beta \in \text{Hom}(V, W)$ then

$$\begin{aligned}\|\alpha + \beta\| &= \sup \left\{ \frac{\|(\alpha + \beta)(v)\|}{\|v\|} \mid 0_V \neq v \in V \right\} \\ &= \sup \left\{ \frac{\|\alpha(v) + \beta(v)\|}{\|v\|} \mid 0_V \neq v \in V \right\} \\ &\leq \sup \left\{ \frac{\|\alpha(v)\| + \|\beta(v)\|}{\|v\|} \mid 0_V \neq v \in V \right\} \leq \|\alpha\| + \|\beta\|.\end{aligned}$$

If $V = F^n$ and $W = F^k$, endowed with respective dot products and the norms defined by them, then the induced norm on $\text{Hom}(V, W)$ is called the **spectral norm**. If $A \in \mathcal{M}_{k \times n}(F)$, then the **spectral norm** of A is defined to be the spectral norm of the homomorphism from F^n to F^k given by $v \mapsto Av$.

Example: If p is any positive integer, we can define a norm $\|\cdot\|_p$ on

$$\mathbb{C}^n \text{ by } \left\| \begin{bmatrix} a_1 \\ \vdots \\ a_n \end{bmatrix} \right\|_p = \sqrt[p]{\sum_{i=1}^n |a_i|^p}. \text{ For the case } p = 2, \text{ this of course reduces}$$

to the norm coming from the dot product. The proof that this is a norm in the general case relies on a generalization of Minkowski's inequality: $\|v + w\|_p \leq \|v\|_p + \|w\|_p$ for all $v, w \in \mathbb{C}^n$ and any positive integer p . This norm can be used to define a norm on $\text{Hom}(\mathbb{C}^n, \mathbb{C}^k)$ for positive integers k and n , by setting

$$\|\alpha\|_p = \sup \left\{ \frac{\|\alpha(v)\|_p}{\|v\|_p} \mid 0_V \neq v \in V \right\}$$

for any $\alpha \in \text{Hom}(\mathbb{C}^n, \mathbb{C}^k)$.

Example: For positive integers k and n , we define the **Frobenius norm** or **Hilbert-Schmidt norm** of a matrix $A = [a_{ij}] \in \mathcal{M}_{k \times n}(\mathbb{C})$ by setting

$$\|A\|_{\mathfrak{F}} = \sqrt{\sum_{i=1}^k \sum_{j=1}^n |a_{ij}|^2} = \sqrt{\text{tr}(AA^H)}.$$

Note that this is precisely the norm coming from the inner product on $\mathcal{M}_{k \times n}(\mathbb{C})$ given by $\langle A, B \rangle = \text{tr}(AB^H)$.

The norm $\| \cdot \|_1$ defined on \mathbb{C}^n is important in other contexts as well. Let n be a positive integer and let θ be the function from $\mathcal{M}_{n \times n}(\mathbb{C})$ to \mathbb{R} defined by $\theta : [a_{ij}] \mapsto \max \{ \sum_{i=1}^n |a_{ij}| \mid 1 \leq j \leq n \}$, which we have already seen when we defined condition numbers. Numerical algorithms that compute the eigenvalues of a matrix, as a rule, make roundoff errors on the order of $c\theta(A)$, where c is a constant determined by the precision of the computer on which the algorithm is running. Since the eigenvalues of similar matrices are identical, it is usually useful, given a square matrix A , to find a matrix B similar to A with $\theta(B)$ small. This can often be done by choosing B of the form PAP^{-1} , where P is a nonsingular diagonal matrix.

Example: If $A = \begin{bmatrix} 1 & 0 & 10^{-4} \\ 1 & 1 & 10^{-2} \\ 10^4 & 10^2 & 1 \end{bmatrix}$, then $\theta(A) = 1002$. However, if we choose $P = \begin{bmatrix} 10^2 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 10^{-2} \end{bmatrix}$, then $\theta(PAP^{-1}) = 3$.

Let α be the endomorphism of \mathbb{C}^n represented with respect to the canonical basis by a matrix $A \in \mathcal{M}_{n \times n}(\mathbb{C})$. Then for each $v \in \mathbb{C}^n$ we have $\theta(A) \|v\|_1 \geq \|\alpha(v)\|_1$. In particular, if c is an eigenvalue of α associated with an eigenvector v then $\theta(A) \|v\|_1 \geq \|\alpha(v)\|_1 = |c| \cdot \|v\|_1$ and so $\theta(A) \geq |c|$. Thus we see that $\theta(A) \geq \rho(A)$, where $\rho(A)$ is the spectral radius of A . This bound is called the **Gershgorin bound**⁶. In fact, we can sharpen this result.

(15.5) Proposition (Gershgorin's Theorem): Let α be the endomorphism of \mathbb{C}^n represented with respect to the canonical



6

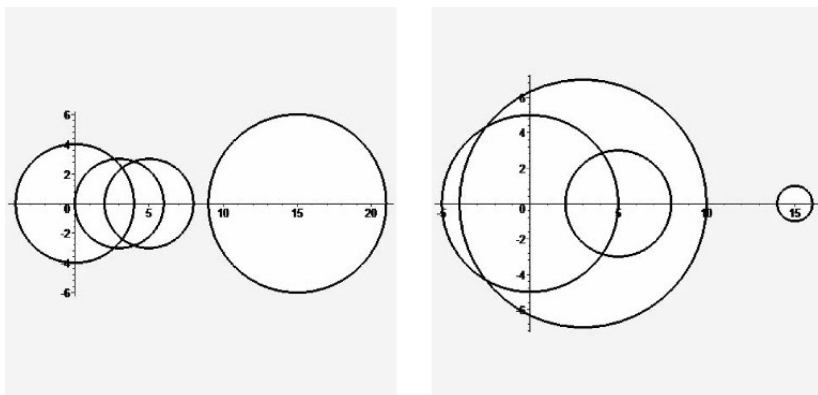
Semyon Aranovich Gershgorin was a 20th century Russian mathematician. Gershgorin's theorem was published in a Russian journal in 1931 and was generally ignored, until it was noticed and publicized by the Austrian-born American mathematician **Olga Taussky-Todd**, one of the most important researchers in the development of numerical linear algebra methods for computers after World War II. Similar results were obtained earlier by Minkowski and Hadamard.

basis by the matrix $A = [a_{ij}] \in M_{n \times n}(\mathbb{C})$ **and, for each** $1 \leq i \leq n$, **let** $r_i = \sum_{j \neq i} |a_{ij}|$. **Let** K_i **be the circle in the complex plane with radius** r_i **and center** a_{ii} . **Then** $\text{spec}(\alpha) \subseteq K = \bigcup_{i \neq j} K_i$.

Proof: Let c be an eigenvalue of α and let $v = \begin{bmatrix} b_1 \\ \vdots \\ b_n \end{bmatrix}$ be an eigenvector of α associated with c . Let h be an index satisfying $|b_h| \geq |b_i|$ for all $1 \leq i \leq n$. Then $b_h \neq 0$ and $Av = cv$ so $(c - a_{hh})b_h = \sum_{j \neq h} a_{hj}b_j$ and hence $|c - a_{hh}||b_h| \leq \sum_{j \neq h} |a_{hj}b_j| \leq |b_h|r_h$. Thus $|c - a_{hh}| \leq r_h$ and so $c \in K_h \subseteq K$, as desired. \square

Example: Let α be the endomorphism of \mathbb{C}^4 represented with respect to the canonical basis by the matrix $A = \begin{bmatrix} 3 & 1 & 2 & 0 \\ 4 & 15 & 0 & -2 \\ -3 & 0 & 0 & -1 \\ 0 & 0 & 3 & 5 \end{bmatrix}$. Then

$\text{spec}(A) = \{15.32, 4.49, 1.59 \pm 2.35i\}$. These numbers are found in the union K of the following circles in the complex plane: the circle of radius 3 around the point $(3, 0)$; the circle of radius 6 around the point $(15, 0)$, the circle of radius 4 around the point $(0, 0)$, and the circle of radius 3 around the point $(5, 0)$. We furthermore note that $\text{spec}(A) = \text{spec}(A^T)$ and so, by the same argument, we see that the eigenvalues of α lie in the union K' of the following circles in the complex plane: the circle of radius 7 around the point $(3, 0)$, the circle of radius 1 around the point $(15, 0)$, the circle of radius 5 around the point $(0, 0)$, and the circle of radius 3 around the point $(5, 0)$. These circles and the location of the eigenvalues can be seen in the following pictures.



Thus, the eigenvalues of α lie in $K \cap K'$.

Since any polynomial in $\mathbb{C}[X]$ is the characteristic polynomial of a matrix, we can use Gershgorin's Theorem to get a bound on the location of the zeros of any polynomial. However, there are more sophisticated methods available to get much better bounds.

There are several generalizations of Gershgorin's Theorem, the best-known of which is the following.

(15.6) Proposition (Brauer's Theorem): Let α be the endomorphism of \mathbb{C}^n represented with respect to the canonical basis by the matrix $A = [a_{ij}] \in M_{n \times n}(\mathbb{C})$ and, for each $1 \leq i \leq n$, let $r_i = \sum_{j \neq i} |a_{ij}|$. For each $1 \leq i \neq j \leq n$, let K_{ij} be the Cassini oval⁷ $\{z \in \mathbb{C} \mid |z - a_{ii}| |z - a_{jj}| \leq r_i r_j\}$ in the complex plane. Then $\text{spec}(\alpha) \subseteq K = \bigcup_{i \neq j} K_{ij}$.

Proof: Let c be an eigenvalue of α and let $v = \begin{bmatrix} b_1 \\ \vdots \\ b_n \end{bmatrix}$ be an eigenvector of α associated with c . Let h and k be indices such that $|b_h| \geq |b_k| \geq |b_i|$ for all $i \neq h, k$. We know that $b_h \neq 0$, and we can assume, as well, that $b_k \neq 0$ for otherwise we would have $a_{hh} = c$, in which case surely $c \in K$. Since $Av = cv$, we have $(c - a_{hh})b_h = \sum_{j \neq h} a_{hj}b_j$ and so

$$|c - a_{hh}| |b_h| = \left| \sum_{j \neq h} a_{hj}b_j \right| \leq \sum_{j \neq h} |a_{hj}| |b_j| \leq \sum_{j \neq h} |a_{hj}| |b_k| = r_h |b_k|.$$

In other words, $|c - a_{hh}| \leq r_h |b_k| |b_h|^{-1}$. In the same manner, we obtain $|c - a_{kk}| \leq r_k |b_h| |b_k|^{-1}$ and so, multiplying these two results together, we see that $|c - a_{hh}| |c - a_{kk}| \leq r_h r_k$, so $c \in K_{hk} \subseteq K$, as desired. \square

Note that Gershgorin's Theorem involves n circles, whereas Brauer's Theorem involves $\binom{n}{2} = \frac{1}{2}n(n-1)$ ovals.



⁷ Twentieth-century German mathematician **Alfred Brauer** emigrated to the United States in 1939; his research was primarily in matrix theory. **Giovanni Domenico Cassini** was a 17th-century Italian mathematician and astronomer.

Example: It is sometimes useful to consider norms on vector spaces V not over subfields of \mathbb{C} , namely functions $v \mapsto \|v\|$ from V to \mathbb{R} satisfying conditions (1) - (3) of Proposition 15.3. For example, let F be a finite field and let $V = F^n$ for some positive integer n . Define $\|v\|$ to be the number of nonzero entries in v , for each $v \in V$. This function is called the **Hamming⁸ norm** and is of extreme importance in algebraic coding theory, where one is interested in vector spaces over F in which every nonzero vector has a large Hamming norm. In an example at the beginning of Chapter 3, we showed a vector space of dimension 3 over $GF(2)$, every nonzero element of which has Hamming norm equal to 4.

If v and w are vectors in space V over which we have a norm defined, then the **distance** between v and w is defined as $d(v, w) = \|v - w\|$. When $V = \mathbb{R}^n$ on which we have the dot product, this just gives us the ordinary notion of euclidean distance. The ability to define the notion of distance in such spaces is important, since it allows us to measure the degree of error in algorithmic computations by measuring the distance between a computed value and the value predicted by theory. It also allows us to define the notion of convergence.

The following proposition shows that this abstract notion of distance indeed has the geometric properties that one would expect from a notion of distance.

Example: Let A be a finite set, and let V be the collection of all subsets of A , which is a vector space over $F = GF(2)$. We have a norm defined on V by letting $\|B\|$ be the number of elements in B . Then the distance between subsets B and C of A is $\|B + C\|$, namely the number of elements in their symmetric difference.

(15.7) Proposition: Let V be a normed space and let $v, w, y \in V$. Then:

- (1) $d(v, w) = d(w, v)$;
- (2) $d(v, w) \geq 0$, where equality exists if and only if $v = w$;
- (3) **(Triangle inequality)** $d(v, w) \leq d(v, y) + d(y, w)$.

Proof: This is an immediate consequence of Proposition 15.3. □



⁸ **Richard Hamming**, a 20th-century American mathematician and computer scientist, is best known for his development of the theory of error-detecting and error-correcting codes.

Let n be a positive integer. If $A = [a_{ij}] \in \mathcal{M}_{n \times n}(\mathbb{C})$, and if $k > 0$ is an integer, let us define the matrix $P(k) = [p_{ij}^{(k)}]$ to be $I + \sum_{h=1}^k \frac{1}{h!} A^h$. We claim that, for each fixed $1 \leq i, j \leq n$, the limit $\lim_{h \rightarrow \infty} p_{ij}^{(h)}$ exists in \mathbb{C} . Indeed, if $B = [b_{ij}] \in \mathcal{M}_{n \times n}(F)$, set $m(B) = \max_{1 \leq i, j \leq n} |b_{ij}|$. Then every entry in the matrix A^2 equals the sum of n products of pairs of entries of A and so, in absolute value, is equal to at most $m(A)^2 n$. Thus we see that $m(A^2) \leq m(A)^2 n$. Similarly, $m(A^3) \leq m(A^2)m(A)n \leq m(A)^3 n^2$ and so forth. Thus, in general,

$$m\left(\frac{1}{k!} A^k\right) \leq \frac{n^{k-1}}{k!} m(A)^k \leq \frac{1}{k!} [m(A)n]^k$$

and so, in particular, $m(P(k)) \leq \sum_{k=0}^{\infty} \frac{1}{k!} m(A)^k$ for all $k \geq 1$. But from calculus we know that the series $\sum_{h=0}^{\infty} \frac{1}{h!} r^h$ converges absolutely to e^r for each real number r . Therefore the limit we seek exists, and, at least by analogy, we are justified in denoting the matrix $\left[\lim_{h \rightarrow \infty} p_{ij}^{(h)}\right]$ by e^A .

(15.8) Proposition: If n is a positive integer $A = [a_{ij}] \in M_{n \times n}(F)$ is a diagonal matrix, where F is \mathbb{R} or \mathbb{C} , then $e^A = [b_{ij}]$ is a diagonal matrix, where

$$b_{ij} = \begin{cases} e^{a_{ii}} & \text{if } i = j \\ 0 & \text{otherwise} \end{cases}.$$

Proof: This is an immediate consequence of the definition. \square

In particular, this implies that if $B \in M_{n \times n}(F)$ is similar to a diagonal matrix then B and e^B have the same eigenvectors, while the eigenvalues of e^B are the exponentials of the eigenvalues of B .

Actually, we can do a bit better: if $A = \begin{bmatrix} A_1 & O & \dots & O \\ O & A_2 & \dots & O \\ & & \ddots & \\ O & O & \dots & A_m \end{bmatrix}$, where each A_h is a square matrix, then $e^A = \begin{bmatrix} e^{A_1} & O & \dots & O \\ O & e^{A_2} & \dots & O \\ & & \ddots & \\ O & O & \dots & e^{A_m} \end{bmatrix}$.

Example: If $O \neq A \in M_{n \times n}(F)$, where F is either \mathbb{R} or \mathbb{C} , is a matrix for which there exists a positive integer t such that $A^t = O$,

and if k is the least such integer, then $e^A = I + \sum_{h=1}^k \frac{1}{h!} A^h$. Thus, for

example, if $A = \begin{bmatrix} 0 & 1 & 2 \\ 0 & 0 & -1 \\ 0 & 0 & 0 \end{bmatrix}$, we have

$$e^A = I + A + \frac{1}{2}A^2 = \begin{bmatrix} 1 & 1 & \frac{3}{2} \\ 0 & 1 & -1 \\ 0 & 0 & 1 \end{bmatrix}.$$

Example: If $A = \begin{bmatrix} 0 & k \\ 0 & 0 \end{bmatrix} \in \mathcal{M}_{2 \times 2}(\mathbb{R})$, then $e^A = \begin{bmatrix} 1 & k \\ 0 & 1 \end{bmatrix}$; if $B = \begin{bmatrix} 0 & r \\ -r & 0 \end{bmatrix}$ then $e^B = \begin{bmatrix} \cos(r) & \sin(r) \\ -\sin(r) & \cos(r) \end{bmatrix}$.

Example: If $A = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{R})$, then

$$e^A = \begin{bmatrix} \frac{2}{3} + \frac{1}{3}e^3 & \frac{1}{3}e^3 - \frac{1}{3} & \frac{1}{3}e^3 - \frac{1}{3} \\ \frac{1}{3}e^3 - \frac{1}{3} & \frac{2}{3} + \frac{1}{3}e^3 & \frac{1}{3}e^3 - \frac{1}{3} \\ \frac{1}{3}e^3 - \frac{1}{3} & \frac{1}{3}e^3 - \frac{1}{3} & \frac{2}{3} + \frac{1}{3}e^3 \end{bmatrix}.$$

If $P \in \mathcal{M}_{n \times n}(F)$ is nonsingular, where F is \mathbb{R} or \mathbb{C} , then

$$P^{-1} \left[I + \sum_{h=1}^k \frac{1}{h!} A^h \right] P = I + \sum_{h=1}^k \frac{1}{h!} (P^{-1}AP)^h$$

for each k and so $P^{-1}e^AP = e^{P^{-1}AP}$. Thus we see that the exponentials of similar matrices are themselves similar. This is very important in calculations. In particular, if A is diagonalizable there exists a nonsingular matrix P such that $P^{-1}AP$ is a diagonal matrix $D = [d_{ij}]$ and so $P^{-1}e^AP = e^D$ is also a diagonal matrix. Thus e^A is diagonalizable whenever A is.

If (A, B) is a commuting pair of matrices in $\mathcal{M}_{n \times n}(F)$, then as a direct consequence of the definition we see that $e^A e^B = e^{A+B} = e^B e^A$. But this is not true in general, as the following example shows.

Example: If $A = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$ and $B = \begin{bmatrix} -1 & 0 \\ 0 & 0 \end{bmatrix}$, then $e^A e^B = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} e^{-1} & 0 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} e^{-1} & 1 \\ 0 & 1 \end{bmatrix} \neq \begin{bmatrix} e^{-1} & 1 - e^{-1} \\ 0 & 1 \end{bmatrix} = e^{A+B}$.

This fact is significant when it comes to calculating e^A in many cases. For example, suppose that A is an $n \times n$ matrix having a single eigenvalue c of multiplicity n . Then for each scalar t , the matrices ctI and $t(A - cI)$ commute and so $e^{tA} = e^{tcI}e^{t(A-cI)} = (e^{tc}I) \sum_{k=0}^{\infty} \frac{t^k}{k!} (A - cI)^k$ and, from the Cayley-Hamilton Theorem, we know that $(A - cI)^k = 0$ for all $k \geq n$. Thus we see that $e^{tA} = (e^{tc}I) \sum_{k=0}^{n-1} \frac{t^k}{k!} (A - cI)^k$ and so there exists a polynomial $p(X) \in F[X]$ satisfying $e^{tA} = p(A)$.

We also note that if A is nonsingular then A and A^{-1} commute so $e^A e^{-A} = e^{A-A} = e^O = I$, proving that e^A is nonsingular and $e^{-A} = (e^A)^{-1}$.

A similar proof can be used to show that if A has distinct eigenvalues $\{c_1, \dots, c_n\}$ and if $p_k(X) = \prod_{j \neq k} (c_k - c_j)^{-1} (X - c_j I)$ for all $1 \leq k \leq n$ then for any scalar t we have $e^{tA} = \sum_{k=1}^n e^{tc_k} p_k(A)$.

We have thus seen that e^A makes sense for any square matrix A over the real or complex numbers. Indeed, matrix exponentials play an important role in solutions of systems of linear differential equations.

What about, say, $\cos(A)$ and $\sin(A)$? We know that the cosine function has a Maclauren representation

$$\cos(x) = \sum_{i=0}^{\infty} \frac{(-1)^i}{(2i)!} x^{2i}.$$

For each natural number n , let us consider the polynomial

$$p_n(X) = \sum_{i=0}^n \frac{(-1)^i}{(2i)!} X^{2i}.$$

Then we can surely calculate $p_n(A)$ for each n and see whether the sequence of such matrices converges in some sense. However, there is another possibility. We know that for any real or complex number z we have $\cos(z) = \frac{1}{2}[e^{iz} + e^{-iz}]$ and so we can just define $\cos(A)$ to be the matrix $\frac{1}{2}[e^{iA} + e^{-iA}]$, which we know always exists.

Example: We see that $\cos(I) = \begin{bmatrix} \cos(1) & 0 & 0 \\ 0 & \cos(1) & 0 \\ 0 & 0 & \cos(1) \end{bmatrix}$ and

$$\cos \left(\begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix} \right) = \begin{bmatrix} \frac{2}{3} + \frac{1}{3} \cos(3) & \frac{1}{3} \cos(3) - \frac{1}{3} & \frac{1}{3} \cos(3) - \frac{1}{3} \\ \frac{1}{3} \cos(3) - \frac{1}{3} & \frac{2}{3} + \frac{1}{3} \cos(3) & \frac{1}{3} \cos(3) - \frac{1}{3} \\ \frac{1}{3} \cos(3) - \frac{1}{3} & \frac{1}{3} \cos(3) - \frac{1}{3} & \frac{2}{3} + \frac{1}{3} \cos(3) \end{bmatrix}.$$

Similarly, we know that $\sin(z) = \frac{-i}{2}[e^{iz} - e^{-iz}]$ and so we can define $\sin(A)$ to be $\frac{-i}{2}[e^{iA} - e^{-iA}]$.

Exercises

Exercise 851 Let $V = C(-1, 1)$ and let $a > -\frac{1}{2}$ be a real number. Is the function $\mu : V \times V \rightarrow \mathbb{R}$ defined by $\langle f, g \rangle = \int_{-1}^1 [1 - t^2]^{a-1/2} f(t)g(t)dt$ an inner product on V ?

Exercise 852 Is the function $\mu : \mathbb{R}^2 \times \mathbb{R}^2 \rightarrow \mathbb{R}$ defined by

$$\mu : \left(\begin{bmatrix} a_1 \\ a_2 \end{bmatrix}, \begin{bmatrix} b_1 \\ b_2 \end{bmatrix} \right) \mapsto a_1(b_1 + b_2) + a_2(b_1 + 2b_2)$$

an inner product on \mathbb{R}^2 ?

Exercise 853 Is the function $\mu : \mathbb{R}^2 \times \mathbb{R}^2 \rightarrow \mathbb{R}$ defined by

$$\mu : \left(\begin{bmatrix} a_1 \\ a_2 \end{bmatrix}, \begin{bmatrix} b_1 \\ b_2 \end{bmatrix} \right) \mapsto a_1b_1 - a_1b_2 - a_2b_1 + 4a_2b_2$$

an inner product on \mathbb{R}^2 ?

Exercise 854 Is the function $\mu : \mathbb{R}^3 \times \mathbb{R}^3 \rightarrow \mathbb{R}$ defined by

$$\mu : \left(\begin{bmatrix} a_1 \\ a_2 \\ a_3 \end{bmatrix}, \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix} \right) \mapsto a_1b_1 + 2a_2b_2 + 3a_3b_3 + a_1b_2 + a_2b_1$$

an inner product on \mathbb{R}^3 ?

Exercise 855 Verify whether the function $\mu : \mathbb{R}[X] \times \mathbb{R}[X] \rightarrow \mathbb{R}$ defined by $\mu : (f, g) \mapsto \deg(fg)$ is an inner product on $\mathbb{R}[X]$?

Exercise 856 Give an example of a function $\mu : \mathbb{R}^2 \times \mathbb{R}^2 \rightarrow \mathbb{R}$ which satisfies the first two conditions of an inner product, which does not satisfy

the third, but does satisfy $\mu \left(\begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \end{bmatrix} \right) = 1$.

Exercise 857 Is the function $\mu : \mathbb{R}[X] \times \mathbb{R}[X] \rightarrow \mathbb{R}$ defined by

$$\mu : \left(\sum_{i=0}^{\infty} a_i X^i, \sum_{j=0}^{\infty} b_j X^j \right) \mapsto \sum_{i=0}^{\infty} \sum_{j=0}^{\infty} \frac{1}{i+j+1} a_i b_j$$

an inner product on $\mathbb{R}[X]$?

Exercise 858 Let V be the vector space of all continuous functions from \mathbb{R} to itself. Let $\mu : V \times V \rightarrow \mathbb{R}$ be the function given by $\mu : (f, g) \mapsto \lim_{t \rightarrow \infty} \frac{1}{t} \int_{-t}^t f(s)g(s)ds$. Is μ an inner product?

Exercise 859 Let n be a positive integer and let $v = \begin{bmatrix} c_1 \\ \vdots \\ c_n \end{bmatrix} \in \mathbb{R}^n$. Is

the function $\mu_v : \mathbb{R}[X_1, \dots, X_n] \times \mathbb{R}[X_1, \dots, X_n] \rightarrow \mathbb{R}$ defined by

$$\mu_v : (p, q) \mapsto p(c_1, \dots, c_n)q(c_1, \dots, c_n)$$

an inner product on $\mathbb{R}[X_1, \dots, X_n]$?

Exercise 860 Let $a < b$ be real numbers and let $V = C(a, b)$. Let $h_0 \in V$ be a function satisfying the condition that $h_0(t) > 0$ for all $a < t < b$. Show that the function $\mu : V \times V \rightarrow \mathbb{R}$ defined by $\mu : (f, g) \mapsto \int_a^b f(x)g(x)h_0(x)dx$ is an inner product on V .

Exercise 861 Let c and d be given real numbers. Find a necessary and sufficient condition that the function $\mu : \left(\begin{bmatrix} a_1 \\ a_2 \end{bmatrix}, \begin{bmatrix} b_1 \\ b_2 \end{bmatrix} \right) \mapsto ca_1b_1 + da_2b_2$ be an inner product on \mathbb{R}^2 .

Exercise 862 Is the function $\mu : \mathbb{R}^3 \times \mathbb{R}^3 \rightarrow \mathbb{R}$ defined by

$$\mu : \left(\begin{bmatrix} a_1 \\ a_2 \\ a_3 \end{bmatrix}, \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix} \right) \mapsto a_1^2b_2 + b_1^2a_2 + (a_3b_3)^2$$

an inner product on \mathbb{R}^3 ?

Exercise 863 Let V be an inner product space over \mathbb{R} and let $n > 1$ be an integer. For positive real numbers a_1, \dots, a_n , define the function

$$\mu : V^n \times V^n \rightarrow \mathbb{R} \text{ by } \mu : \left(\begin{bmatrix} v_1 \\ \vdots \\ v_n \end{bmatrix}, \begin{bmatrix} w_1 \\ \vdots \\ w_n \end{bmatrix} \right) \mapsto \sum_{i=1}^n a_i \langle v_i, w_i \rangle. \text{ Is } \mu$$

an inner product on V^n ?

Exercise 864 Let n be a positive integer and let V be the subspace of $\mathbb{R}[X]$ consisting of all polynomials of degree at most n . Is the function $\mu : V \times V \rightarrow \mathbb{R}$ defined by $\mu : (p, q) \mapsto \sum_{i=0}^n p\left(\frac{i}{n}\right)q\left(\frac{i}{n}\right)$ an inner product on V ?

Exercise 865 Let $0 < n \in \mathbb{Z}$. Is the function $\mu : \mathbb{C}^n \times \mathbb{C}^n \rightarrow \mathbb{C}$ defined

$$\text{by } \mu : \left(\begin{bmatrix} a_1 \\ \vdots \\ a_n \end{bmatrix}, \begin{bmatrix} b_1 \\ \vdots \\ b_n \end{bmatrix} \right) \mapsto \sum_{i=1}^n a_i \bar{b}_{n-i+1} \text{ an inner product?}$$

Exercise 866 Let $\{u, v\}$ be a linearly-dependent subset of an inner product space V , not containing 0_V . Show that $\|u\|^2 v = \langle v, u \rangle u$.

Exercise 867 Let $V = \mathbb{C}^2$ on which we have defined the dot product, and let $D = \{v \in V \mid \|v\| = 1\}$. Find $\{\langle Av, v \rangle \mid v \in D\}$, where $A = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} \in \mathcal{M}_{2 \times 2}(\mathbb{C})$.

Exercise 868 Let n be a positive integer and let $\{v_1, \dots, v_k\}$ be a set of vectors in \mathbb{R}^n satisfying $v_i \cdot v_j \leq 0$ for all $1 \leq i < j \leq k$. Show that $k \leq 2n$ and give an example in which equality holds.

Exercise 869 Let n be a positive integer and let $A \in \mathcal{M}_{n \times n}(\mathbb{C})$ satisfy $A^2 = A$. Is it necessarily true that $(A^H)^2 = A^H$?

Exercise 870 Let V be an inner product space and let $v \neq v'$ be vectors in V . Show that there exists a vector $w \in V$ satisfying $\langle v, w \rangle \neq \langle v', w \rangle$.

Exercise 871 Let V be an inner product space finitely generated over its field of scalars, and let $B = \{v_1, \dots, v_n\}$ be a basis of V . Show that there exists a basis $\{w_1, \dots, w_n\}$ of V satisfying the condition that

$$\langle v_i, w_j \rangle = \begin{cases} 1 & \text{if } i = j \\ 0 & \text{otherwise} \end{cases}.$$

Exercise 872 Let W be a subspace of a vector space V over \mathbb{R} and let Y be a complement of W in V . Define an inner product μ on W and an inner product ν on Y . Is the function from $V \times V \rightarrow \mathbb{R}$ defined by $(w + y, w' + y') \mapsto \mu(w, w') + \nu(y, y')$ an inner product on V ?

Exercise 873 Let $V = C(0, 1)$. Let $A = \{f_1, \dots, f_n\}$ be a linearly-independent subset of V and define a function $u : \mathbb{R} \times \mathbb{R} \rightarrow \mathbb{R}$ by $u : (a, b) \mapsto \sum_{j=1}^n f_j(a) \cos^j(b)$. Show that if $h \in V$ and if there exists a function $g \in V$ such that $h(x) = \int_0^1 u(x, y)g(y)dy$ for all $x \in \mathbb{R}$, then $h \in \mathbb{R}A$.

Exercise 874 Let V be an inner product space over \mathbb{R} . For each real number a , set $U(a) = \{v \in V \mid \langle v, v \rangle \leq a\}$. Given a real number a , find a real number b such that $\langle v + w, v + w \rangle \in U(b)$ for all $v, w \in U(a)$.

Exercise 875 Let V be an inner product space and let $\alpha \in \text{End}(V)$. Show that $\langle \alpha(v), v \rangle \langle v, \alpha(v) \rangle \leq \|\alpha(v)\|^2$ for every normal vector $v \in V$.

Exercise 876 For real numbers a_1, \dots, a_n , show that

$$\sum_{i=1}^n a_i \leq \left(\sqrt[n]{\sum_{i=1}^n |a_i|^{2/3}} \right) \left(\sqrt[n]{\sum_{i=1}^n |a_i|^{4/3}} \right).$$

Exercise 877 (Binet-Cauchy identity) For $u, v, w, y \in \mathbb{R}^3$, show that $(v \times w)(y \times u) = (v \cdot y)(w \cdot u) - (v \cdot u)(y \cdot w)$.

Exercise 878 Let v be a normal vector in \mathbb{R}^3 . Show that the function $\alpha_v : \mathbb{R}^3 \rightarrow \mathbb{R}^3$ defined by $\alpha_v : w \mapsto v \times (v \times w) + w$ is a projection in $\text{End}(\mathbb{R}^3)$.

Exercise 879 For $u, v, w, y \in \mathbb{R}^3$, show that

$$(u \times v) \cdot (w \times y) = \begin{vmatrix} u \cdot w & u \cdot y \\ v \cdot w & v \cdot y \end{vmatrix}.$$

Exercise 880 For nonnegative real numbers a , b , and c , show that

$$(a + b + c)\sqrt{2} \leq \sqrt{a^2 + b^2} + \sqrt{b^2 + c^2} + \sqrt{a^2 + c^2}.$$

Exercise 881 For real numbers $0 < a \leq b \leq c$, show that

$$\sqrt{b^2 + c^2} \leq (\sqrt{2})a \leq \sqrt{(b-a)^2 + (c-a)^2}.$$

Exercise 882 Let n be a positive integer and let $A \in \mathcal{M}_{n \times n}(\mathbb{R})$ be a matrix the n Gershgorin circles of which are mutually disjoint. Prove that all of the eigenvalues of A are real.

Exercise 883 Show that $\left[\int_0^1 f(x) dx \right]^2 \leq \int_0^1 f(x)^2 dx$ for any $f \in C(0, 1)$.

Exercise 884 Let $f : \mathbb{R} \rightarrow \mathbb{R}$ be the constant function $x \mapsto 1$. Calculate $\|f\|$ when f is considered as an element of $C(0, \frac{\pi}{2})$ and compare it to $\|f\|$, when f is considered as an element of $C(0, \pi)$.

Exercise 885 Let V be an inner product space over \mathbb{R} and let $v, w \in V$ satisfy $\|v + w\| = \|v\| + \|w\|$. Show that $\|av + bw\| = a\|v\| + b\|w\|$ for all $0 \leq a, b \in \mathbb{R}$.

Exercise 886 Let V be an inner product space over \mathbb{R} . Show that $\langle u, v \rangle = \frac{1}{4}\|u + v\|^2 - \frac{1}{4}\|u - v\|^2$ for all $u, v \in V$.

Exercise 887 Let $V = C(0, 1)$ on which we have defined the inner product $\langle f, g \rangle = \int_0^1 f(t)g(t)dt$. Let W be the subspace of V generated by the function $x \mapsto x^2$. Find all elements of W normal with respect to this inner product.

Exercise 888 Let V be an inner product space over \mathbb{R} and assume that $v, w \in V$ are nonzero vectors satisfying the condition $\langle v, w \rangle = \|v\| \cdot \|w\|$. Show that $\mathbb{R}v = \mathbb{R}w$.

Exercise 889 Let V be a vector space over \mathbb{R} on which we have two inner products, μ and μ' defined, which in turn define distance functions d and d' respectively. If $d = d'$, does it necessarily follow that $\mu = \mu'$?

Exercise 890 (Apollonius' identity)⁹ Let V be an inner product space. Show that $\|u - w\|^2 + \|v - w\|^2 = \frac{1}{2}\|u - v\|^2 + 2\left\|\frac{1}{2}(u + v) - w\right\|^2$ for all $u, v, w \in V$.

Exercise 891 Let V be an inner product space over \mathbb{R} and let $\theta : V \times V \times V \rightarrow \mathbb{R}$ be the function defined by

$$\theta(v, w, y) = \|v + w + y\|^2 + \|v + w + y\|^2 - \|v - w - y\|^2 - \|v - w + y\|^2.$$

Show that, for any $v, w, y \in V$, the value of $\theta(v, w, y)$ does not depend on y .

Exercise 892 Let V be an inner product space over \mathbb{R} and let $n > 2$.

Let $\theta : V^n \rightarrow V$ be the function defined by $\theta : \begin{bmatrix} v_1 \\ \vdots \\ v_n \end{bmatrix} \mapsto \frac{1}{n} \sum_{i=1}^n v_i$. Show that $\sum_{i=1}^n \left\| v_i - \theta \left(\begin{bmatrix} v_1 \\ \vdots \\ v_n \end{bmatrix} \right) \right\|^2 = \sum_{i=1}^n \|v_i\|^2 - n \left\| \begin{bmatrix} v_1 \\ \vdots \\ v_n \end{bmatrix} \right\|^2$.

Exercise 893 Let V be an inner product space over \mathbb{R} and let v and w be nonzero vectors in V . Show that $|\langle v, w \rangle|^2 = \langle v, v \rangle \langle w, w \rangle$ if and only if the set $\{v, w\}$ is linearly dependent.

Exercise 894 Let n be a positive integer and let $\|\cdot\|$ be a norm defined on \mathbb{C}^n . For each $A \in \mathcal{M}_{n \times n}(\mathbb{C})$, let $\|A\|$ be the spectral norm of A . If $A \in \mathcal{M}_{n \times n}(\mathbb{C})$ is nonsingular, show that every singular matrix $B \in \mathcal{M}_{n \times n}(\mathbb{C})$ satisfies $\|A - B\| \geq \|A^{-1}\|^{-1}$. Does there necessarily exist a singular matrix B for which equality holds?

Exercise 895 Let V be a vector space over \mathbb{R} and let $\|\cdot\|$ be a norm defined on V . Show that $|||v| - |w||| \leq \|v - w\|$ for all $v, w \in V$.

Exercise 896 Let V be an inner product space finitely generated over \mathbb{R} and let $\delta \in D(V)$. Pick $v_0 \in V$. Show that for each real number $e > 0$



⁹ Greek geometer **Apollonius of Perga**, who worked in Alexandria in the third century BC, in his famous book *Conics*, was the first to introduce the terms “hyperbola”, “parabola”, and “ellipse”.

there exists a real number $d > 0$ such that $|\delta(v) - \delta(v_0)| < \epsilon$ whenever $\|v - v_0\| < d$.

Exercise 897 Let V be an inner product space. For any $u, v, w \in V$,

$$\text{show that } \begin{vmatrix} 0 & 1 & 1 & 1 \\ 1 & 0 & d(u, v)^2 & d(u, w)^2 \\ 1 & d(u, v)^2 & 0 & d(v, w)^2 \\ 1 & d(u, w)^2 & d(v, w)^2 & 0 \end{vmatrix} \leq 0.$$

Exercise 898 Let $n > 1$ be an integer. Show that there is no norm $v \mapsto \|v\|$ defined on \mathbb{C}^n satisfying

$$\|A\|_{\mathfrak{F}} = \sup \left\{ \frac{\|Av\|}{\|v\|} \mid 0_V \neq v \in \mathbb{C}^n \right\}$$

for all $A \in \mathcal{M}_{n \times n}(\mathbb{C})$.

Exercise 899 Let $n > 1$ be an integer. For each $A \in \mathcal{M}_{n \times n}(\mathbb{C})$, let $\|A\| = \rho(A)$, the spectral radius of A . Does this turn $\mathcal{M}_{n \times n}(\mathbb{C})$ into a normed space?

Exercise 900 Let $p > 2$ be prime and let n be a positive integer. For each $1 \leq i \leq n$, define $w(i) = \min\{i - 1, p - i + 1\}$. Does the function

$$GF(p)^n \rightarrow \mathbb{R} \text{ defined by } \begin{bmatrix} a_1 \\ \vdots \\ a_n \end{bmatrix} \mapsto \sum_{i=1}^n w(i)a_i \text{ turn } GF(p)^n \text{ into a}$$

normed space?

Exercise 901 Let $0 < p < 1$ and let $f: \mathbb{R}^n \rightarrow \mathbb{R}$ be defined by

$$f: \begin{bmatrix} a_1 \\ \vdots \\ a_n \end{bmatrix} \rightarrow \left(\sum_{i=1}^n |a_i|^p \right)^{1/p}$$

is not a norm but does satisfy the inequality

$$f(v + w) \leq 2^{(1-p)/p} [f(v) + f(w)]$$

for all $v, w \in \mathbb{R}^n$.

Exercise 902 Let $n > 1$ and let $A \in \mathcal{M}_{n \times n}(\mathbb{C})$. Show that there are infinitely many other matrices in $\mathcal{M}_{n \times n}(\mathbb{C})$ having the same Gershgorin circles as A .

Exercise 903 Let $A = [a_{ij}] \in \mathcal{M}_{2 \times 2}(\mathbb{C})$ and let K be the Cassini oval defined by A . Show that every point on the boundary of K is an eigenvalue of a matrix $B \in \mathcal{M}_{2 \times 2}(\mathbb{C})$ defining the same Cassini oval.

16

Orthogonality

Let V be an inner product space and let $0_V \neq v, w \in V$. From Proposition 15.2 we see that

$$-1 \leq \frac{\langle v, w \rangle + \langle w, v \rangle}{2 \|v\| \cdot \|w\|} \leq 1$$

and so there exists a real number $0 \leq t \leq \pi$ satisfying

$$\cos(t) = \frac{\langle v, w \rangle + \langle w, v \rangle}{2 \|v\| \cdot \|w\|}.$$

This number t is the **angle** between v and w . Note that if we are working over \mathbb{R} , then

$$\cos(t) = \frac{\langle v, w \rangle}{\|v\| \cdot \|w\|}.$$

Example: If $V = \mathbb{R}^n$ is endowed with the dot product, and if $0_V \neq v, w \in V$ then, using analytic geometry, it is easy to show that the angle as defined here is indeed the angle between the straight line determined by v and the origin, and the straight line determined by w and the origin. If we define different inner products on V , we build in this manner various non-euclidean geometries in n -space.

Example: Let $V = C(0, 1)$, on which we have defined the inner product $\langle f, g \rangle = \int_0^1 f(x)g(x)dx$. In particular, consider the functions $f : x \mapsto 5x^2$

and $g : x \mapsto 3x$. Then $\|f\| = \sqrt{5}$ and $\|g\| = \sqrt{3}$, and the angle t between f and g satisfies $\cos(t) = \sqrt{\frac{1}{15}} \int_0^1 (5x^2)(3x)dx = \frac{1}{4}\sqrt{15}$.

Vectors v and w in an inner product space V are **orthogonal** if and only if $\langle v, w \rangle = 0$. In this case we write $v \perp w$. We note that if $v \perp w$ then $\|v + w\|^2 = \|v\|^2 + \langle v, w \rangle + \langle w, v \rangle + \|w\|^2 = \|v\|^2 + \|w\|^2$. A nonempty subset D of V is a set of **mutually orthogonal** vectors if $v \perp w$ whenever $v \neq w$ in D .

Example: We have already seen that if $v, w \in \mathbb{R}^3$, then $v \cdot (v \times w) = 0$. This says that a vector v is orthogonal to $v \times w$, for any vector w . The same is also true for w and $v \times w$ and so we see that if $\{v, w\}$ is a linearly-independent subset of \mathbb{R}^3 then the set $\{v, w, v \times w\}$ is linearly independent and so is a basis of \mathbb{R}^3 .

Moreover, as an immediate consequence of the Lagrange identity on \mathbb{R}^3 ,

we see that if $v \times w = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$ and $v \cdot w = 0$ then either $v = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$ or

$w = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$. If $v, w \in \mathbb{R}^3$, then the angle t between them satisfies the

condition that $v \cdot w = (\|v\| \cdot \|w\|) \cos(t)$. Using the Lagrange identity, we see that

$$\begin{aligned} \|v \times w\|^2 &= \|v\|^2 \|w\|^2 - (v \cdot w)^2 = \|v\|^2 \|w\|^2 [1 - \cos^2(t)] \\ &= \|v\|^2 \|w\|^2 \sin^2(t) \end{aligned}$$

and so $\|v \times w\| = (\|v\| \cdot \|w\|) |\sin(t)|$. Thus $|\cos(t)| = \frac{\|v \times w\|}{\|v\| \cdot \|w\|}$.

Example: If $V = \mathbb{C}^2$ on which we have the dot product, then it is easy to see that $\begin{bmatrix} 2+3i \\ -1+5i \end{bmatrix} \perp \begin{bmatrix} 1+i \\ -i \end{bmatrix}$.

Example: Let $V = C(-1, 1)$, on which we have defined the inner product $\langle f, g \rangle = \int_{-1}^1 f(x)g(x)dx$. For all $i \geq 0$, define the functions $p_i \in V$ as follows: $p_0 : x \mapsto 1$; $p_1 : x \mapsto x$; and

$$p_{h+1} : x \mapsto \left(\frac{2h+1}{h+1} \right) x p_h(x) - \left(\frac{h}{h+1} \right) p_{h-1}(x) \text{ whenever } h > 1.$$

These polynomial functions are known as **Legendre polynomials**. It is easy to verify that $p_i \perp p_h$ whenever $i \neq h$.

On the same space, we can define another inner product, namely

$$\langle f, g \rangle = \int_{-1}^1 \frac{f(x)g(x)}{\sqrt{1-x^2}} dx.$$

For each $i \geq 0$, define the function $q_i \in V$ by setting $q_0 : x \mapsto 1$; $q_1 : x \mapsto x$; and $q_{h+1} : x \mapsto 2xq_h(x) - q_{h-1}(x)$ whenever $h > 1$. These polynomial functions are known as **Chebyshev¹ polynomials**. It is again easy to verify that $q_i \perp q_h$ whenever $i \neq h$.

Both of these products are special instances of a more general construction. For any $-1 < r, s \in \mathbb{R}$, it is possible to define an inner product on $C(-1, 1)$ by setting $\langle f, g \rangle = \int_{-1}^1 f(x)g(x)(1-x)^r(1+x)^s dx$. The set of polynomial functions which are mutually orthogonal with respect to this inner product is called the set of **Jacobi polynomials** of type (r, s) . Such polynomials are important in many areas of numerical analysis, and in particular in numerical integration.

(16.1) Proposition: Let V be an inner product space over a field of scalars F .

- (1) If $v \in V$ satisfies $v \perp w$ for all $w \in V$, then $v = 0_V$.
- (2) If $\emptyset \neq A \subseteq V$ and if $v \in V$ satisfies the condition that $v \perp w$ for all $w \in A$, then $v \perp w$ for $w \in FA$.

Proof: (1) is an immediate consequence of the fact that if $v \neq 0_V$ then $\langle v, v \rangle \neq 0$. Now assume that $\emptyset \neq A \subseteq V$ and that $v \perp w$ for all $w \in A$. If $y \in FA$ then there exist elements $w_1, \dots, w_n \in A$ and scalars a_1, \dots, a_n such that $y = \sum_{i=1}^n a_i w_i$ and so $\langle v, y \rangle = \sum_{i=1}^n \bar{a}_i \langle v, w_i \rangle = 0$, whence $v \perp y$. \square

(16.2) Proposition: Let V be an inner product space and let A be a nonempty set of nonzero mutually-orthogonal vectors in V . Then A is linearly independent.



¹ **Adrien-Marie Legendre** was one of the first-rate mathematicians who worked in France during the time of the revolution and the generation after it. Among other things, he served on the committee that defined the metric system. **Pafnuty Lvovich Chebyshev**, a 19th-century Russian mathematician, made important contributions to both pure and applied mathematics.

Proof: Let $\{v_1, \dots, v_n\}$ be a finite subset of A and assume that there exist scalars c_1, \dots, c_n such that $\sum_{i=1}^n c_i v_i = 0_V$. Then, for $1 \leq h \leq n$, we have $c_h \langle v_h, v_h \rangle = \sum_{i=1}^n c_i \langle v_i, v_h \rangle = \langle \sum_{i=1}^n c_i v_i, v_h \rangle = \langle 0_V, v_h \rangle = 0$ and hence $c_h = 0$. Thus any finite subset of A is linearly independent, and therefore A is linearly independent. \square

If V is an inner product space then any vector $0_V \neq w \in V$ defines a function

$$\pi_w : v \mapsto \frac{\langle v, w \rangle}{\langle w, w \rangle} w$$

from V to itself, which is in fact a projection the image of which is the subspace of V generated by $\{w\}$. This easily-checked remark is the basis for the following theorem.

(16.3) Proposition (Gram-Schmidt Theorem): Any finitely-generated inner product space V has a basis composed of mutually-orthogonal vectors.

Proof: We will proceed by induction on $\dim(V)$. If $\dim(V) = 1$ the result is immediate. Therefore we can assume that the proposition is true for any inner product space of dimension k , and assume that $\dim(V) = k + 1$. Let W be a subspace of V of dimension k . By the induction hypothesis, there exists a basis $\{v_1, \dots, v_k\}$ of W composed of mutually-orthogonal vectors. Let $v \in V \setminus W$ and set $v_{k+1} = v - \sum_{i=1}^k \pi_{v_i}(v)$. This vector does not belong to W since $v \notin W$. Therefore $\{v_1, \dots, v_{k+1}\}$ is a generating set for V . Moreover, for $1 \leq j \leq k$, we have

$$\langle v_{k+1}, v_j \rangle = \langle v, v_j \rangle - \sum_{i=1}^k \frac{\langle v, v_i \rangle}{\langle v_i, v_i \rangle} \langle v_i, v_j \rangle = \langle v, v_j \rangle - \frac{\langle v, v_j \rangle}{\langle v_j, v_j \rangle} \langle v_j, v_j \rangle = 0$$

and so $v_{k+1} \perp v_j$ for all $1 \leq j \leq k$. By Proposition 5.3, it follows that the set $\{v_1, \dots, v_{k+1}\}$ is linearly independent and so is a basis for V . \square

We should note that the proof of Proposition 16.3 is an algorithm, called the **Gram-Schmidt process**, which is easy to implement by a computer program to create a basis composed of mutually-orthogonal vectors of V , when we are given a basis of any sort for the space.

Example: Let $v_1 = \begin{bmatrix} 3 \\ 0 \\ 0 \\ 0 \end{bmatrix}$, $v_2 = \begin{bmatrix} 0 \\ 1 \\ 2 \\ 1 \end{bmatrix}$, and $v_3 = \begin{bmatrix} 3 \\ -1 \\ 3 \\ 2 \end{bmatrix}$

be vectors in \mathbb{R}^4 , on which we have defined the dot product. The set

$\{v_1, v_2, v_3\}$ is linearly independent and so generates a three-dimensional subspace W of \mathbb{R}^4 . Let us use the Gram-Schmidt process to build a basis for W composed of mutually-orthogonal vectors. Indeed, we define

$$u_1 = v_1, \quad u_2 = v_2 - \pi_{u_1}(v_2) = \begin{bmatrix} 0 \\ 1 \\ 2 \\ 1 \end{bmatrix}, \quad \text{and} \quad u_3 = v_3 - \pi_{u_1}(v_3) - \pi_{u_2}(v_3) =$$

$$\frac{1}{6} \begin{bmatrix} 0 \\ -13 \\ 4 \\ 5 \end{bmatrix} \quad \text{then } \{u_1, u_2, u_3\} \text{ is a basis for } W, \text{ the vectors of which are}$$

mutually orthogonal.

Example: Let $V = C(-1, 1)$, on which we have defined the inner product $\langle f, g \rangle = \int_{-1}^1 f(x)g(x)dx$. For all $i \geq 0$, let f_i be the polynomial function $f_i : x \mapsto x^i$. Then for each $n > 0$, the set $\{f_0, \dots, f_n\}$ is linearly independent and so forms a basis for a subspace W of V . We now apply the Gram-Schmidt process to this basis, to obtain a basis $\{p_0, \dots, p_n\}$ of vectors in V which are mutually orthogonal, where the p_j are precisely the Legendre polynomials we introduced earlier.

Actually, the assumption that we have a basis in hand when initiating the Gram-Schmidt process is one of convenience rather than necessity. We could begin with an arbitrary generating set $\{v_1, \dots, v_n\}$ for the given space. In that case, at the h th stage of the process we would begin by checking whether v_h is a linear combination of the set of mutually-orthogonal vectors $\{u_1, \dots, u_{h-1}\}$ we have already created. If it is, we just discard it and go on to v_{h+1} .

We should point out that the Gram-Schmidt process is not considered computationally stable – small errors and roundoffs in the computational process accumulate rapidly and can lead at the end to a significant difference between the true solution and the computed solution. There are, fortunately, other more sophisticated methods of constructing a basis composed of mutually-orthogonal vectors from a given basis.

(16.4) Proposition (Hadamard inequality): Let n be a positive integer, let $A = [a_{ij}] \in M_{n \times n}(\mathbb{R})$ be a nonsingular matrix, and let $e = |A|$. Then $|e| \leq g^n \sqrt{n^n}$, where $g = \max\{|a_{ij}| \mid 1 \leq i, j \leq n\}$.

Proof: Denote the rows of A by v_1, \dots, v_n . Then $\{v_1, \dots, v_n\}$ is a basis for $V = \mathbb{R}^n$ and so, using the Gram-Schmidt method, we can find a new basis $\{u_1, \dots, u_n\}$ for V , on which we consider the dot product, composed of mutually-orthogonal vectors, and defined by setting $u_1 = v_1$ and $u_h = v_h - \sum_{j=1}^{h-1} c_{hj}u_j$, where $c_{hj} = (v_h \cdot u_j)(u_j \cdot u_j)^{-1}$.

If $B \in \mathcal{M}_{n \times n}(\mathbb{R})$ is the matrix the rows of which are u_1, \dots, u_n , then

$$A = CB, \text{ where } C = \begin{bmatrix} 1 & 0 & 0 & \dots & 0 \\ c_{21} & 1 & 0 & \dots & 0 \\ & & \dots & & \\ c_{n1} & c_{n2} & \dots & c_{n,n-1} & 1 \end{bmatrix}. \text{ Since } C \text{ is a}$$

lower-triangular matrix, its determinant is the product of the entries on its diagonal, namely 1. Therefore $e = |B|$.

By looking at the Gram-Schmidt method, we see that $\|u_i\| \leq \|v_i\|$ for all $1 \leq i \leq n$. Moreover, since the u_i are mutually orthogonal, we see that $BB^T = D$, where $D = [d_{ij}]$ is the diagonal matrix defined by $d_{ii} = \|u_i\|^2$ for all $1 \leq i \leq n$. Therefore $(e)^2 = |BB^T| = |D| = \prod_{i=1}^n \|u_i\|^2$. Now let $g = \max\{|a_{ij}| \mid 1 \leq i, j \leq n\}$. Then $\|u_i\| \leq \|v_i\| \leq g\sqrt{n}$ for all $1 \leq i \leq n$ and so $|e| \leq g^n \sqrt{n^n}$, as desired. \square

Let V be an inner product space having a subspace W . Let $W^\perp = \{v \in V \mid \langle v, w \rangle = 0 \text{ for all } w \in W\}$. By Proposition 16.1, we know that W^\perp is a subspace of V . Since $\langle v, v \rangle \neq 0$ for all $0_V \neq v \in V$, it is clear that W and W^\perp are disjoint. Also, again by Proposition 16.1, we see that $V^\perp = \{0_V\}$ and $\{0_V\}^\perp = V$. The space W^\perp is called the **orthogonal complement** of W in V , and this name is justified by the following result:

(16.5) Proposition: Let W be a subspace of a finitely-generated inner product space V . Then $V = W \oplus W^\perp$ and $W = (W^\perp)^\perp$.

Proof: By Proposition 16.3, we know that it is possible to find a basis $\{v_1, \dots, v_k\}$ of W which is composed of mutually-orthogonal vectors, and by the construction method used in the proof of this proposition, we see that this can be extended to a basis $\{v_1, \dots, v_n\}$ of V , the elements of which are still mutually orthogonal. Thus $v_i \in W^\perp$ for all $k < i \leq n$, proving that $V = W + W^\perp$. But we already know that W and W^\perp are disjoint and so we have $W \oplus W^\perp$. Moreover, $\{v_{k+1}, \dots, v_n\}$ is a basis for W^\perp and so $W = (W^\perp)^\perp$. \square

In particular, if V is an inner product space having a subspace W then we have a natural projection of $W \oplus W^\perp$ onto W , called the **orthogonal projection**. The image of a vector $v \in W \oplus W^\perp$ under this projection is the unique element of W closest to v , according to the distance function defined by the inner product on V , in the sense of the following theorem.

(16.6) Proposition: Let W be a subspace of an inner product space V and let $v = w + y$, where $w \in W$ and $y \in W^\perp$. Then $\|v - w'\| \geq \|v - w\|$ for all $w' \in W$, with equality holding if and only if $w' = w$.

Proof: If $w' \in W$ then

$$\begin{aligned}
 \|v - w'\|^2 &= \|w - w' + y\|^2 = \langle w - w' + y, w - w' + y \rangle \\
 &= \langle w - w', w - w' \rangle + \langle y, w - w' \rangle + \langle w - w', y \rangle + \langle y, y \rangle \\
 &= \langle w - w', w - w' \rangle + \langle y, y \rangle \\
 &= \|w - w'\|^2 + \|y\|^2 = \|w - w'\|^2 + \|v - w\|^2
 \end{aligned}$$

and from here the result follows immediately. \square

One of the important problems in computational algebra is the following: given an endomorphism α of a finitely-generated inner product space and a vector $0_V \neq v_0 \in V$, find an efficient procedure to define an orthogonal projection onto the Krylov subspace $F[\alpha]v_0$. One of the first of these is the **Arnoldi process**, a modification of the Gram-Schmidt process. Several variants of this procedure have been devised, depending on special properties of α . This process is not considered as computationally efficient as the Lanczos algorithm mentioned earlier. Arnoldi's process is also the basis for the **GMRES algorithm** (GMRES = generalized minimal residual) for solution of systems of linear equations, devised by Yousef Saad and Martin Schultz in 1986.

Note that Proposition 16.5 is not necessarily true if the space V is not finitely generated, as the following example shows.

Example: Let $V = \mathbb{R}^{(\infty)}$. For each $h \geq 0$, let v_h be the sequence in which the h th entry equals 1 and all other entries equal 0. Then $B = \{v_h \mid h \geq 0\}$ is a basis for V composed of mutually-orthogonal vectors. Let $W = \mathbb{R}\{v_0 - v_1, v_1 - v_2, \dots\}$. This subspace of V is proper since $v_0 \in V \setminus W$. If $0_V \neq y \in W^\perp$ then there exists a nonnegative integer n such that $y = \sum_{i=0}^n a_i v_i$ where the a_i are real numbers and $a_n \neq 0$. But then $a_n = \langle y, v_n - v_{n+1} \rangle = 0$, and that is a contradiction. Therefore we have shown that $W^\perp = \{0_V\}$, despite the fact that $W \neq V$. Moreover, in this case $V \neq W \oplus W^\perp$ and $(W^\perp)^\perp = V \neq W$.

Let V be an inner product space. A nonempty subset A of V is **orthonormal** if and only if the elements of A are mutually orthogonal, and each of them is normal. Thus, for example, the canonical basis of \mathbb{R}^n , equipped with the dot product, is orthonormal.

Example: Let $V = C(-\pi, \pi)$, on which we have an inner product defined by $\langle f, g \rangle = \frac{1}{\pi} \int_{-\pi}^{\pi} f(x)g(x)dx$. Then we have an orthonormal subset $\left\{ \frac{1}{\sqrt{2}} \right\} \cup \{\sin(nx) \mid n \geq 1\} \cup \{\cos(nx) \mid n \geq 1\}$ of V .

Example: Let V be the subspace of $\mathbb{R}^{\mathbb{R}}$ consisting of all functions f for which $\int_{-\infty}^{\infty} |f(x)|^2 dx$ is finite, where the norm is taken with respect

to the inner product $\langle f, g \rangle = \int_{-\infty}^{\infty} f(x)g(x)dx$ defined on V . Let $h \in V$ be the function defined by

$$h : x \mapsto \begin{cases} 1 & \text{for } 0 < x \leq \frac{1}{2} \\ -1 & \text{for } \frac{1}{2} < x \leq 1 \\ 0 & \text{otherwise} \end{cases}.$$

This function is known as the **Haar² wavelet**. For each $j, k \in \mathbb{N}$ define the function $h_k^j \in V$ by setting $h_k^j : x \mapsto 2^{j/2}h(2^jx - k)$. Then the subset $\{h_k^j \mid j, k \in \mathbb{N}\}$ of V is orthonormal. Haar wavelets have important applications in image compression.

(16.7) Proposition: Every finitely-generated inner product space V has an orthonormal basis.

Proof: By Proposition 16.3, we know that V has a basis $\{v_1, \dots, v_n\}$ the elements of which are mutually orthogonal. For each $1 \leq i \leq n$, let $w_i = \|v_i\|^{-1}v_i$. Then each w_i is normal and $\{w_1, \dots, w_n\}$ is a basis for V , the elements of which remain mutually orthogonal. \square

We can modify the Gram-Schmidt method to provide an algorithm for constructing an orthonormal basis from any given basis of a finitely-generated inner product space V , by normalizing each basis element as it is created. This has the added advantage of tending to reduce accumulated roundoff and truncation errors. The examples after Proposition 16.3 and Proposition 16.6 show that inner product spaces which are not finitely generated may have orthonormal bases as well, but this is not always true. Making use of the Hausdorff Maximum Principle, it is possible to show that every inner product space V has a maximal orthonormal set, which must be linearly independent by Proposition 16.2. Such a subset is called a **Hilbert subset** of V . Clearly, a subset A of V is a Hilbert subset if and only if for every $0_V \neq y \in V$ there exists a $v \in A$ satisfying $\langle v, y \rangle \neq 0$. If V is finitely generated then any Hilbert subset of V is a basis for V , but this is not necessarily true for inner product spaces which are not finitely generated.



² The twentieth-century Hungarian mathematician **Alfréd Haar** worked primarily in analysis.

Example: It is of course possible that a finitely-generated inner product space may have many different orthonormal bases. For example, the canonical basis of \mathbb{R}^4 is orthonormal, as is the basis

$$\left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ \frac{-1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 0 \\ 1 \end{bmatrix} \right\}.$$

(16.8) Proposition: Let V be a finitely-generated inner product space having an orthonormal basis $\{v_1, \dots, v_n\}$. If $x = \sum_{i=1}^n a_i v_i$ and $y = \sum_{i=1}^n b_i v_i$ are vectors in V , then $\langle x, y \rangle = \sum_{i=1}^n a_i \bar{b}_i$.

Proof: By the properties of the inner product, we have $\langle x, y \rangle = \langle \sum_{i=1}^n a_i v_i, \sum_{j=1}^n b_j v_j \rangle = \sum_{i=1}^n \sum_{j=1}^n a_i \bar{b}_j \langle v_i, v_j \rangle = \sum_{i=1}^n a_i \bar{b}_i$, as desired. \square

(16.9) Proposition: Let V be a finitely-generated inner product space having an orthonormal basis $\{v_1, \dots, v_n\}$. Then each $v \in V$ satisfies $v = \sum_{i=1}^n \langle v, v_i \rangle v_i$.

Proof: We know that $v = \sum_{i=1}^n a_i v_i$, for some scalars a_1, \dots, a_n . Then for each $1 \leq h \leq n$ we have $\langle v, v_h \rangle = \langle \sum_{i=1}^n a_i v_i, v_h \rangle = \sum_{i=1}^n a_i \langle v_i, v_h \rangle = a_h \langle v_h, v_h \rangle = a_h$, which yields the desired result. \square

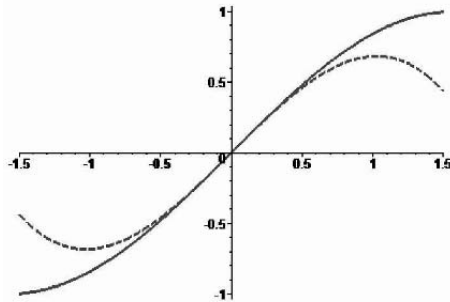
The coefficients $\langle v, v_i \rangle$ encountered in Proposition 16.9 are called the **Fourier coefficients** of the vector v with respect to the given orthonormal basis.

Example: Consider the vector space $C(-1, 1)$ over \mathbb{R} , on which we have the inner product $\langle f, g \rangle = \int_{-1}^1 f(x)g(x)dx$. We want to find a polynomial function of degree at most 3 which most closely approximates the function $f : x \mapsto \sin(x)$ on the interval $[-1, 1]$. To do so, consider the subspace V of $C(-1, 1)$ generated by the functions $p_i : x \mapsto x^i$ for $0 \leq i \leq 3$ and f . Apply the Gram-Schmidt process to the basis $\{p_0, \dots, p_3, f\}$ of V to get an orthonormal basis $\{q_0, \dots, q_3, g\}$, where $q_0 : x \mapsto \frac{1}{2}$; $q_1 : x \mapsto \left(\sqrt{\frac{3}{2}}\right)x$; $q_2 : x \mapsto \left(\sqrt{\frac{5}{2}}\right)\left(\frac{3}{2}x^2 - \frac{1}{2}\right)$; and $q_3 : x \mapsto \left(\sqrt{\frac{7}{2}}\right)\left(\frac{5}{2}x^3 - \frac{9}{6}x\right)$. By Proposition 16.6 and Proposition 16.9, we know that the polynomial function of degree at most 3 which most

closely approximates the function f is $\sum_{i=0}^3 \langle f, q_i \rangle q_i$, where the Fourier coefficients $\langle f, q_i \rangle$ are given by

$$\begin{aligned}\langle f, q_0 \rangle &= \int_{-1}^1 \frac{1}{2} \sin(x) dx = 0; \\ \langle f, q_1 \rangle &= \int_{-1}^1 \left(\sqrt{\frac{3}{2}} \right) \sin(x) x dx = \sqrt{6} (\sin(1) - \cos(1)) = 0.738; \\ \langle f, q_2 \rangle &= \int_{-1}^1 \left(\sqrt{\frac{5}{2}} \right) \left(\frac{3}{2} x^2 - \frac{1}{2} \right) \sin(x) dx = 0; \\ \langle f, q_3 \rangle &= \int_{-1}^1 \left(\sqrt{\frac{7}{2}} \right) \left(\frac{5}{2} x^3 - \frac{9}{6} x \right) \sin(x) dx \\ &= (14\sqrt{14}) \cos(1) - (9\sqrt{14}) \sin(1) = -0.034;\end{aligned}$$

Thus the polynomial function we seek is given by $u : x \mapsto -0.315x^3 + 0.998x$. The following diagram shows the graphs $\sin(x)$ and $u(x)$ over a slightly larger interval, in which $u(x)$ is represented by the dashed line and $\sin(x)$ by the solid line.



(16.10) Proposition: Let F be \mathbb{R} or \mathbb{C} and let k and n be positive integers. Let $A \in M_{k \times n}(F)$ be a matrix the columns of which are linearly independent in F^k . Then there exist matrices $Q \in M_{k \times n}(F)$ and $R \in M_{n \times n}(F)$ such that

- (1) $A = QR$;
- (2) The columns of Q are orthonormal with respect to the dot product on F^k ;
- (3) R is nonsingular and upper-triangular.

Proof: Let u_1, \dots, u_n be the columns of A . Apply the Gram-Schmidt process to the set $\{u_1, \dots, u_n\}$ and then normalize each of the resulting vectors to obtain an orthonormal set $\{v_1, \dots, v_n\}$ of vectors in F^k . Let $Q \in \mathcal{M}_{k \times n}(F)$ be the matrix having columns v_1, \dots, v_n . Then, by Proposition 16.9, we see that $u_i = \sum_{j=1}^n (u_i \cdot v_j) v_j$ for all $1 \leq i \leq n$, and so $A = QR$, where $R = [r_{ij}] \in \mathcal{M}_{n \times n}(F)$ is given by $r_{ij} = u_j \cdot v_i$ for all $1 \leq i, j \leq n$. This matrix is clearly nonsingular. Moreover, we note that the Gram-Schmidt process is such that v_j is orthogonal to u_1, \dots, u_{j-1} for all $2 \leq j \leq n$ and so $r_{ij} = 0$ when $i > j$. Therefore R is also upper-triangular. \square

A decomposition of a matrix in the form given by Proposition 16.10 is called a **QR-decomposition**. Such decompositions form a basis of many important numerical algorithms, and are widely used, for example, in computing eigenvalues of large matrices. The use is primarily iterative. If A is an $n \times n$ matrix over \mathbb{R} or \mathbb{C} the eigenvalues of which have distinct absolute values and if we can indefinitely perform the iteration

- (1) $A_1 = A$;
 - (2) If A_i has a QR-decomposition $A_i = Q_i R_i$ then set $A_{i+1} = R_i Q_i$;
- then, under rather mild conditions on A , the sequence A_1, A_2, \dots of matrices tends to an upper triangular matrix in which the eigenvalues of A appear in decreasing order of absolute value along the diagonal.³

One of the major advantages of QR-decompositions is that they are easy to update. If we are given a decomposition $A = QR$, and then the matrix A is altered slightly to obtain a matrix A' by changing a few of its entries, it is relatively easy to alter Q and R to get a QR-decomposition for A' . This is important since many applications of linear algebra involve solving successive systems of linear equations of the form $A^{(i)}X = w^{(i)}$, where $A^{(i+1)}$ and $w^{(i+1)}$ are obtained from $A^{(i)}$ and $w^{(i)}$ by relatively minor modifications, based on data from some external source which is periodically updated.

The following algorithm is used to compute a QR-decomposition of a matrix $A \in \mathcal{M}_{k \times n}(F)$ with columns u_1, \dots, u_n :



3

QR-decompositions were developed independently by Swiss computer scientist **Heinz Rutishauser**, one of the fathers of ALGOL, by Russian computer scientist **Vera Kublanovskaya**, and by **J. G. F. Francis** of the British computer manufacturer Ferranti Ltd.

For $i = 1$ to n do steps (1) - (3):

- (1) $v_i = u_i$;
- (2) For $j = 1$ to $i - 1$ set $r_{ji} = u_i \cdot v_j$ and $v_i = v_i - r_{ji}v_j$;
- (3) Set $r_{ii} = \|v_i\|$ and $v_i = r_{ii}^{-1}v_i$.

Then Q is the matrix with columns v_1, \dots, v_n and $R = [r_{ij}]$. Note that step (3) presupposes that we have already checked that the set of columns of A is linearly independent. If not, then we have to add an initial check to insure that r_{ii} is nonzero, before we attempt to invert it. As already noted, the Gram-Schmidt method is not numerically stable and hence neither is this algorithm for finding a QR-decomposition. It can be modified to produce a somewhat more stable algorithm by replacing the definition of r_{ji} in step (2) by $r_{ji} = v_i \cdot v_j$.

(16.11) Proposition: Let V be a finitely-generated inner product space having an orthonormal basis $\{v_1, \dots, v_n\}$. Then for all $v, w \in V$:

- (1) **(Parseval's identity)** $\langle v, w \rangle = \sum_{i=1}^n \overline{\langle v_i, v \rangle} \langle v_i, w \rangle$;
- (2) **(Bessel's identity⁴)** $\|v\|^2 = \sum_{i=1}^n |\langle v_i, v \rangle|^2$.

Proof: Parseval's identity follows from the calculation

$$\langle v, w \rangle = \left\langle \sum_{i=1}^n \langle v, v_i \rangle v_i, w \right\rangle = \sum_{i=1}^n \langle v, v_i \rangle \langle v_i, w \rangle = \sum_{i=1}^n \overline{\langle v_i, v \rangle} \langle v_i, w \rangle$$

and Bessel's identity derives from this in the special case $v = w$. □

The following results shows that orthogonality can be used to determine the relation between two different inner products defined on a vector space over \mathbb{R} .

(16.12) Proposition: Let V be a vector space over \mathbb{R} on which we have defined two inner products, μ_1 and μ_2 . For $i = 1, 2$, let



⁴

Wilhelm Bessel was a 19th century astronomer and friend of Gauss; his mathematical work came as a result of his research on planetary orbits. Little is known about the life of **Marc-Antoine Parseval**, a French mathematician who published only five short papers at the end of the 18th century, and no picture of him survives.

$Y_i = \{(v, w) \in V \times V \mid \mu_i(v, w) = 0\}$. Then the following conditions are equivalent:

- (1) There exists a positive real number c such that $\mu_2 = c^2 \mu_1$;
- (2) $Y_1 = Y_2$;
- (3) $Y_1 \subseteq Y_2$.

Proof: Since it is clear that (1) implies (2), and (2) implies (3), all we have to prove is that (3) implies (1). Therefore, assume (3). First let us consider the case $\dim(V) = 1$, i.e. the case in which $V = \mathbb{R}$. Then, for $i = 1, 2$, the scalar $b_i = \mu_i(1, 1)$ is nonzero. Set $d = \mu_2(1, 1)/\mu_1(1, 1)$. If $a, b \in \mathbb{R}$, then $\mu_2(a, b) = ab\mu_2(1, 1) = abd\mu_1(1, 1) = d\mu_1(a, b)$ and so, taking $c = \sqrt{d}$, we have established (1). Thus we can assume that $\dim(V) \geq 2$. For each $i = 1, 2$, and each $v \in V$, let $\|v\|_i = \sqrt{\mu_i(v, v)} > 0$. Without loss of generality, we can assume that there exist elements $v, w \in V$ and positive real numbers $a < b$ such that $\|v\|_2 = a\|v\|_1$ and $\|w\|_2 = b\|w\|_1$, since otherwise we would immediately have (1). Suppose that $w = dv$ for some $0 \neq d \in \mathbb{R}$. Then $\|w\|_2 = |d| \cdot \|v\|_2 = |d|a \cdot \|v\|_1 a\|w\|_1$ and so $a = b$, which is contrary to our assumption that $a < b$. Therefore we conclude that the set $\{v, w\}$ is linearly independent. Normalizing v and w with respect to μ_1 if necessary, we can furthermore assume that $\|v\|_1 = 1 = \|w\|_1$.

We claim that $(v, w) \notin Y_1$. Indeed, assume otherwise. Then

$$\mu_1(v + w, v - w) = \mu_1(v, v) - \mu_1(w, w) = \|v\|_1^2 - \|w\|_1^2 = 0$$

and so $(v + w, v - w) \in Y_1$. Therefore, by (3), $(v + w, v - w) \in Y_2$. But $\mu_2(v + w, v - w) = \|v\|_2^2 - \|w\|_2^2 = a^2 - b^2 \in \mathbb{R} \setminus \{0\}$, and so we have a contradiction, establishing the claim. Set $y = v - \|v\|_1^2 \mu_1(v, w)^{-1} w$. Then $\mu_1(y, v) = \|v\|_1^2 - r\mu_1(v, w) = 0$, where $r = \|v\|_1^2 \mu_1(v, w)^{-1}$, and so $(y, v) \in Y_1 \subseteq Y_2$. Since the set $\{v, w\}$ is linearly independent, we know that $y \neq 0_V$. If we set $y' = \|y\|_1^{-1} y$, then $(y', v) \in Y_1 \subseteq Y_2$ and so, as before, $\mu_1(y' + v, y' - v) = \|y'\|_1^2 - \|v\|_1^2 = 0$. Hence $(y' + v, y' - v) \in Y_1$. Thus $\|y\|_1^2 + \|v\|_1^2 = \|-y + v\|_1^2 = \|rw\|_1^2 = \|v\|_1^4 |\mu_1(v, w)|^{-2} \|w\|_1^2$ and so $\|y\|_1^2 = \|v\|_1^4 |\mu_1(v, w)|^{-2} \|w\|_1^2 - \|v\|_1^2$. Since $\mu_2(y, v) = 0$, we see that

$$\|y\|_2^2 + \|v\|_2^2 = \|-y + v\|_2^2 = \|rw\|_2^2 = \|v\|_1^4 |\mu_1(v, w)|^{-2} \|w\|_2^2$$

and so

$$\begin{aligned} \|y\|_2^2 &= \|v\|_1^4 |\mu_1(v, w)|^{-2} \|w\|_2^2 - \|v\|_2^2 \\ &= \|v\|_1^4 |\mu_1(v, w)|^{-2} b^2 \|w\|_1^2 - a^2 \|v\|_1^2 \\ &> a^2 \left(\|v\|_1^4 |\mu_1(v, w)|^{-2} \|w\|_1^2 - \|v\|_1^2 \right) = a^2 \|y\|_1^2. \end{aligned}$$

Since $\mu_2(y', v) = 0$, this implies that

$$\mu_2(y' + v, y' - v) = \|y'\|_2^2 - \|v\|_2^2 > a^2 - a^2 = 0,$$

contradicting (3) and the fact that $\mu_1(y' + v, y' - v) = 0$. From this contradiction, we conclude that there can be no elements a and b as above, so there must exist a positive real number c such that $\|v\|_2^2 = c\|v\|_1^2$ for each nonzero vector $v \in V$. Then for each $v, w \in V$ we have

$$\begin{aligned}\mu_2(v, w) &= \frac{1}{4} [\|v + w\|_2 - \|v - w\|_2] \\ &= \frac{c^2}{4} [\|v + w\|_1 - \|v - w\|_1] = c^2 \mu_1(v, w),\end{aligned}$$

which proves (1). \square

Let V be an inner product space. We have already seen that, for each $v \in V$, the function from V to the field of scalars given by $v \mapsto \langle v, w \rangle$ belongs to $D(V)$. If V is finitely generated, we claim that every element of $D(V)$ is of this form. The following result is actually a special case of a much wider, and more complicated, theorem.

(16.13) Proposition (Riesz⁵ Representation Theorem): Let V be a finitely-generated inner product space. If $\delta \in D(V)$ then there exists a unique vector $y \in V$ satisfying $\delta(v) = \langle v, y \rangle$ for all $v \in V$.

Proof: Let $\{v_1, \dots, v_n\}$ be an orthonormal basis for V and let $y = \sum_{i=1}^n \overline{\delta(v_i)} v_i$. Then for all $1 \leq h \leq n$ we have

$$\langle v_h, y \rangle = \left\langle v_h, \sum_{i=1}^n \overline{\delta(v_i)} v_i \right\rangle = \sum_{i=1}^n \delta(v_i) \langle v_h, v_i \rangle = \delta(v_h)$$

and so $\langle v, y \rangle = \delta(v)$ for all $v \in V$. The vector y is unique since if $\langle v, x \rangle = \langle v, y \rangle$ for all $v \in V$ then $x = \sum_{i=1}^n \langle x, v_i \rangle v_i = \sum_{i=1}^n \overline{\delta(v_i)} v_i = y$, as desired. \square

Example: Let $n > 1$ be an integer and let V be the subspace of $\mathbb{R}^{\mathbb{R}}$ consisting of all polynomial functions of degree at most n , on which we have an inner product defined by $\langle f, g \rangle = \int_{-1}^1 f(t)g(t)dt$. Let $\delta \in D(V)$ be the linear functional defined by $\delta : f \mapsto f(0)$. By Proposition



5

The 20th century Hungarian mathematician **Frigyes Riesz** was one of the founders of functional analysis.

16.13, there exists a polynomial function $p \in V$ satisfying the condition $f(0) = \int_{-1}^1 f(t)p(t)dt$ for all $f \in V$. The function p is defined to be $\sum_{i=0}^n p_i(0)p_i$, where p_i is the i th Legendre polynomial.

(16.14) Proposition: Let V and W be finitely-generated inner product spaces, and let $\alpha : V \rightarrow W$ be a linear transformation. Then there exists a unique linear transformation $\alpha^* : W \rightarrow V$ satisfying the condition $\langle \alpha(v), w \rangle = \langle v, \alpha^*(w) \rangle$ for all $v \in V$ and all $w \in W$.

Proof: Let w be a given vector in W . It is easy to check that the function δ from V to F defined by $\delta : v \mapsto \langle \alpha(v), w \rangle$ is a linear functional. By Proposition 16.13, we know that there exists a unique vector $y_w \in V$ satisfying $\delta(v) = \langle v, y_w \rangle$ for all $v \in V$. Define the function $\alpha^* : W \rightarrow V$ by $\alpha^* : w \mapsto y_w$. We have to prove that this function is indeed a linear transformation. Indeed, if $w_1, w_2 \in W$ then

$$\begin{aligned} \langle v, \alpha^*(w_1 + w_2) \rangle &= \langle \alpha(v), w_1 + w_2 \rangle = \langle \alpha(v), w_1 \rangle + \langle \alpha(v), w_2 \rangle \\ &= \langle v, \alpha^*(w_1) \rangle + \langle v, \alpha^*(w_2) \rangle = \langle v, \alpha^*(w_1) + \alpha^*(w_2) \rangle \end{aligned}$$

and this is true for all $v \in V$, we have $\alpha^*(w_1 + w_2) = \alpha^*(w_1) + \alpha^*(w_2)$ for all $w_1, w_2 \in W$. If c is a scalar and if $w \in W$ then

$$\langle v, \alpha^*(cw) \rangle = \langle \alpha(v), cw \rangle = \bar{c} \langle \alpha(v), w \rangle = \bar{c} \langle v, \alpha^*(w) \rangle = \langle v, c\alpha^*(w) \rangle$$

for all $v \in V$, and hence $\alpha^*(cw) = c\alpha^*(w)$. Thus α^* is a linear transformation and, since y_w is uniquely defined, it is also unique. \square

Let V and W be inner product spaces and let $\alpha : V \rightarrow W$ be a linear transformation. A linear transformation $\alpha^* : W \rightarrow V$ satisfying the condition $\langle \alpha(v), w \rangle = \langle v, \alpha^*(w) \rangle$ for all $v \in V$ and $w \in W$ is called an **adjoint transformation** of α . By Proposition 16.14, we know that if V and W are finitely generated then every $\alpha \in \text{Hom}(V, W)$ has a unique adjoint.

(16.15) Proposition: Let V and W be finitely-generated inner product spaces, having orthonormal bases $B = \{v_1, \dots, v_n\}$ and $D = \{w_1, \dots, w_k\}$ respectively. Let $\alpha : V \rightarrow W$ be a linear transformation. Then $\Phi_{BD}(\alpha)$ is the matrix $A = [a_{ij}]$, where $a_{ji} = \langle \alpha(v_i), w_j \rangle$ and $\Phi_{DB}(\alpha^*) = A^H$.

Proof: For all $1 \leq i \leq n$, let $\alpha(v_i) = \sum_{h=1}^k a_{hj}w_h$. Then for all $1 \leq j \leq k$ we have $\langle \alpha(v_i), w_j \rangle = \left\langle \sum_{h=1}^k a_{hj}w_h, w_j \right\rangle = a_{ji}$ and also $\langle \alpha^*(w_j), v_i \rangle = \overline{\langle v_i, \alpha^*(w_j) \rangle} = \overline{\langle \alpha(v_i), w_j \rangle} = \bar{a}_{ji}$, which is what we needed. \square

Example: It is of course possible that a linear transformation between inner product spaces can have an adjoint even if the spaces are not finitely generated. For example, let $[a, b]$ be a closed interval on the real line and let V be the vector space of all differentiable functions from $[a, b]$ to \mathbb{R} . Define an inner product on V by setting $\langle f, g \rangle = \int_a^b f(x)g(x)dx$. This is an inner product space which is not finitely generated over \mathbb{R} . Let α be the endomorphism of V satisfying $\alpha(f) : x \mapsto \int_a^b e^{-tx} f(t)dt$. Then $\langle \alpha(f), g \rangle = \langle f, \alpha(g) \rangle$ for all $f, g \in V$, and so α^* exists, and equals α .

(16.16) Proposition: Let V , W , and Y be inner product spaces. Let α and β be linear transformations from V to W having adjoints, let ζ be a linear transformation from W to Y having an adjoint, and let c be a scalar. Then:

- (1) $(\alpha + \beta)^* = \alpha^* + \beta^*$;
- (2) $(c\alpha)^* = \bar{c}\alpha^*$;
- (3) $(\zeta\alpha)^* = \alpha^*\zeta^*$;
- (4) $\alpha^{**} = \alpha$.

Proof: (1) For all $v \in V$ and all $w \in W$, we have

$$\begin{aligned} \langle v, (\alpha + \beta)^*(w) \rangle &= \langle (\alpha + \beta)(v), w \rangle = \langle \alpha(v) + \beta(v), w \rangle \\ &= \langle \alpha(v), w \rangle + \langle \beta(v), w \rangle = \langle v, \alpha^*(w) \rangle + \langle v, \beta^*(w) \rangle \\ &= \langle v, (\alpha^* + \beta^*)(w) \rangle \end{aligned}$$

and so by the uniqueness of the adjoint we get $(\alpha + \beta)^* = \alpha^* + \beta^*$.

(2) For all $v \in V$ and all $w \in W$, we have

$$\begin{aligned} \langle v, (c\alpha)^*(w) \rangle &= \langle (c\alpha)(v), w \rangle = \langle c(\alpha(v)), w \rangle = c \langle \alpha(v), w \rangle \\ &= c \langle v, \alpha^*(w) \rangle = \langle v, \bar{c}\alpha^*(w) \rangle = \langle v, (\bar{c}\alpha^*)(w) \rangle \end{aligned}$$

and so $(c\alpha)^* = \bar{c}\alpha^*$.

(3) For all $v \in V$ and all $y \in Y$, we have

$$\langle v, (\zeta\alpha)^*(y) \rangle = \langle (\zeta\alpha)(v), y \rangle = \langle \alpha(v), \zeta^*(y) \rangle = \langle v, \alpha^*\zeta^*(y) \rangle$$

and so $(\zeta\alpha)^* = \alpha^*\zeta^*$.

(4) For all $v \in V$ and all $w \in W$, we have

$$\langle w, \alpha^{**}(v) \rangle = \langle \alpha^*(w), v \rangle = \overline{\langle v, \alpha^*(w) \rangle} = \overline{\langle \alpha(v), w \rangle} = \langle w, \alpha(v) \rangle$$

and so $\alpha^{**} = \alpha$. \square

For a finitely-generated inner product space V , we can consider the assignment $\alpha \mapsto \alpha^*$ as a function from $\text{End}(V)$ to itself which satisfies the following conditions:

- (1) $\sigma_1^* = \sigma_1$;

- (2) $(\alpha + \beta)^* = \alpha^* + \beta^*$ and $(\alpha\beta)^* = \beta^*\alpha^*$ for all $\alpha, \beta \in \text{End}(V)$;
 (3) $\alpha^{**} = \alpha$ for all $\alpha \in \text{End}(V)$.

A function satisfying these conditions is called an **involution** on the F -algebra $\text{End}(V)$. Another involution we have already seen is the function $A \mapsto A^T$ on the F -algebra $\mathcal{M}_{n \times n}(F)$, for any field F . Of course, in the case $F = \mathbb{R}$, the relation between these two involutions can be seen from Proposition 16.15. We have also seen that the function $A \mapsto A^H$ is an involution of $\mathcal{M}_{n \times n}(\mathbb{C})$, and its relation to the involution $\alpha \mapsto \alpha^*$ is also immediate from Proposition 16.15.

(16.17) Proposition: Let $\alpha : V \rightarrow W$ be a linear transformation between finitely-generated inner product spaces. Then:

- (1) $\ker(\alpha^*) = \text{im}(\alpha)^\perp$;
 (2) $\ker(\alpha) = \text{im}(\alpha^*)^\perp$;
 (3) $\text{im}(\alpha) = \ker(\alpha^*)^\perp$;
 (4) $\text{im}(\alpha^*) = \ker(\alpha)^\perp$.

Proof: (1) We note that

$$\begin{aligned} \ker(\alpha^*) &= \{w \in W \mid \alpha^*(w) = 0_V\} \\ &= \{w \in W \mid \langle v, \alpha^*(w) \rangle = 0 \text{ for all } v \in V\} \\ &= \{w \in W \mid \langle \alpha(v), w \rangle = 0 \text{ for all } v \in V\} = \text{im}(\alpha)^\perp. \end{aligned}$$

(2) This follows from the same argument as (1), replacing α by α^* .

(3) By (1) and Proposition 16.6, we have $\text{im}(\alpha) = (\text{im}(\alpha)^\perp)^\perp = \ker(\alpha^*)^\perp$.

(4) This follows from (2) in the same way (3) follows from (1).

□

(16.18) Proposition: If α is an endomorphism of a finitely-generated inner product space V then $\text{null}(\alpha) = \text{null}(\alpha^*)$.

Proof: By Proposition 6.10 and Proposition 16.17, we know that

$$\text{null}(\alpha) = \dim(\text{im}(\alpha^*)^\perp) = \dim(V) - \dim(\text{im}(\alpha^*)) = \text{null}(\alpha^*)$$

and so we are done. □

Example: Proposition 16.18 is not necessarily true for inner product spaces which are not finitely generated. For example, let $V = \mathbb{R}^{(\infty)}$ with the inner product $\langle [a_0, a_1, \dots], [b_0, b_1, \dots] \rangle = \sum_{i=0}^{\infty} a_i b_i$. Let $\alpha \in \text{End}(V)$ be given by $\alpha : [a_0, a_1, \dots] \mapsto [0, a_0, a_1, \dots]$. Then α^* exists and is given by $\alpha^* : [a_0, a_1, \dots] \mapsto [a_1, a_2, \dots]$. Clearly, $\ker(\alpha)$ is trivial but $\ker(\alpha^*)$ is not.

(16.19) Proposition: Let $\alpha : V \rightarrow W$ be a linear transformation between finitely-generated inner product spaces. Then

(1) If α is a monomorphism then $\alpha^*\alpha$ is an automorphism of V ;

(2) If α is an epimorphism then $\alpha\alpha^*$ is an automorphism of W .

Proof: (1) It suffices to prove that the linear transformation $\alpha^*\alpha$ is monic. And, indeed, if $v \in V$ satisfies $\alpha^*\alpha(v) = 0_V$ then $\langle \alpha(v), \alpha(v) \rangle = \langle \alpha^*\alpha(v), v \rangle = \langle 0_V, v \rangle = 0$ and so $\alpha(v) = 0_W$. Since α is a monomorphism, $v = 0_V$ and so we have shown that $\alpha^*\alpha$ is monic, as we needed.

(2) First of all, we will show that α^* is a monomorphism. Indeed, if $w_1, w_2 \in W$ are vectors satisfying $\alpha^*(w_1) = \alpha^*(w_2)$ then for all $v \in V$ we have $\langle \alpha(v), w_1 - w_2 \rangle = \langle v, \alpha^*(w_1) - \alpha^*(w_2) \rangle = 0$ and since α is an epimorphism, we conclude that $\langle w, w_1 - w_2 \rangle = 0$ for all $w \in W$. This implies that $w_1 - w_2 = 0_W$ and so $w_1 = w_2$, showing that α^* is indeed monic. Now we will show that $\alpha\alpha^*$ is also monic, which will suffice to prove (2). Indeed, if $\alpha\alpha^*(w) = 0_W$ then $\langle \alpha^*(w), \alpha^*(w) \rangle = \langle \alpha\alpha^*(w), w \rangle = 0$ and so $\alpha^*(w) = 0_V$, proving that $w = 0_W$. \square

(16.20) Proposition: Let $\alpha : V \rightarrow W$ be an isomorphism between finitely-generated inner product spaces. Then $(\alpha^*)^{-1} = (\alpha^{-1})^*$.

Proof: Let $\beta = (\alpha^{-1})^*$. Then for all $v_1, v_2 \in V$ we have $\langle v_1, v_2 \rangle = \langle \alpha^{-1}\alpha(v_1), v_2 \rangle = \langle \alpha(v_1), \beta(v_2) \rangle = \langle v_1, \alpha^*\beta(v_2) \rangle$ and so $\alpha^*\beta(v_2) = v_2$ for all $v_2 \in V$, which means that $\alpha^*\beta$ is the identity map on V . Thus $\beta = (\alpha^*)^{-1}$. \square

Finally, we note that in an inner product space over \mathbb{R} we can also project onto affine subsets and not just onto subspaces. Indeed, if V is an inner product space over \mathbb{R} then any element v of V defines a linear functional $\delta_v \in D(V)$ given by $\delta_v : w \mapsto \langle w, v \rangle$. If $0 \neq c \in \mathbb{R}$, then $\delta_c^{-1}(c)$ is an affine subset of V . Define a function $\theta_v : V \rightarrow V$ by setting

$$\theta_v : y \mapsto y + \left[\frac{c - \delta_v(y)}{\|v\|^2} \right] v.$$

Then for all $y \in V$ we have $\delta_v\theta_v(y) = c$ and so we see that $\theta_v(y) \in \delta_c^{-1}(c)$, so that $\text{im}(\theta_v) \subseteq \delta_c^{-1}(c)$. Moreover, $\theta_v^2 = \theta_v$. We call the function θ_v the **projection** on the affine set $\delta_c^{-1}(c)$. Such projections have many applications, such as the algebraic reconstruction technique (ART), which is very important in computerized imaging.

Exercises

Exercise 904 Let $A = \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix} \in \mathcal{M}_{2 \times 2}(\mathbb{R})$. Does there exist a matrix

$B \in \mathcal{M}_{2 \times 2}(\mathbb{R})$ of the form $\begin{bmatrix} \cos(t) & -\sin(t) \\ \sin(t) & \cos(t) \end{bmatrix}$ such that the columns of $A + B$ are orthogonal, when considered as elements of the space \mathbb{R}^2 , endowed with the dot product?

Exercise 905 Calculate the angle between the vectors $\begin{bmatrix} 2 \\ 1 \\ 1 \end{bmatrix}$ and $\begin{bmatrix} 3 \\ 1 \\ -2 \end{bmatrix}$ in the space \mathbb{R}^3 , endowed with the dot product.

Exercise 906 Calculate the angle between the vectors $\begin{bmatrix} 1 \\ 1 \\ 1 \\ 2 \\ 1 \end{bmatrix}$ and $\begin{bmatrix} 1 \\ -1 \\ 1 \\ -1 \\ 1 \end{bmatrix}$ in the space \mathbb{R}^5 , endowed with the dot product.

Exercise 907 Let A and B be nonempty subsets of \mathbb{R}^3 which satisfy the condition that $u \times v \in B$ whenever $u \in A$ and $v \in B$. Is it true that $u \times w \in B^\perp$ whenever $u \in A$ and $w \in B^\perp$.

Exercise 908 Let $V = \mathcal{M}_{2 \times 2}(\mathbb{R})$ and define an inner product on V by setting $\left\langle \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}, \begin{bmatrix} b_{11} & b_{12} \\ b_{21} & b_{22} \end{bmatrix} \right\rangle = \sum_{i=1}^2 \sum_{j=1}^2 a_{ij} b_{ij}$. Find the angle between the matrices $\begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}$ and $\begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix}$.

Exercise 909 Let $V = C(0, 1)$ on which we have defined the inner product $\langle f, g \rangle = \int_0^1 f(x)g(x)dx$. Calculate $\|\cos(t)\|$.

Exercise 910 Let V be the space \mathbb{R}^4 , together with the dot product. Find a normal vector in V which is orthogonal to each of the vectors

$$\begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ -1 \\ -1 \\ 1 \end{bmatrix}, \text{ and } \begin{bmatrix} 2 \\ 1 \\ 1 \\ 3 \end{bmatrix}.$$

Exercise 911 Let $V = \mathbb{R}^3$ on which some inner product is defined. Does there exist a vector $0_V \neq v \in V$ which is orthogonal to each of the vectors

$$\begin{bmatrix} 2 \\ 1 \\ 3 \end{bmatrix}, \quad \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix}, \quad \text{and} \quad \begin{bmatrix} 2 \\ 0 \\ 4 \end{bmatrix}?$$

Exercise 912 Let $f, g \in \mathbb{R}^{\mathbb{R}}$ be defined by $f: x \mapsto x$ and $g: x \mapsto x^2 - \frac{1}{2}$. Are f and g orthogonal as elements of $C(0, 1)$? Are they orthogonal as elements of $C(0, 2)$?

Exercise 913 Find a real number c such that $\|v - w\| = c$ for every orthonormal pair $\{v, w\}$ of vectors in \mathbb{R}^n , on which the dot product is defined.

Exercise 914 Let n be a positive integer and let c_1, \dots, c_n is a list of real numbers. Let $\{v_1, \dots, v_n\}$ be an orthonormal basis for \mathbb{R}^n , let $d = \min\{c_1, \dots, c_n\}$. For each $1 \leq i \leq n$, set $d_i = \sqrt{c_i - d}$ and let $w_i = dv_i$. Let $B \in \mathcal{M}_{n \times n}(\mathbb{R})$ be the matrix the columns of which are d_1, \dots, d_n and let $A = BB^T + dI$. For $1 \leq i \leq n$, show that v_i is an eigenvector of A associated with the eigenvalue c_i .

Exercise 915 Let $V = C(0, 1)$ on which we have defined the inner product $\langle f, g \rangle = \int_0^1 f(x)g(x)dx$, and let $W = \mathbb{R}\{e^x\} \subseteq V$. Find an infinite set of elements of W^\perp .

Exercise 916 Let $V = \mathbb{R}^4$ on which some inner product is defined. Find distinct vectors $v, w, y \in V$ such that $v \perp w$ and $w \perp y$, but not $v \perp y$.

Exercise 917 Let V be an inner product space over \mathbb{R} and let v and w be vectors in V . Show that $\|v\| = \|w\|$ if and only if $(v+w) \perp (v-w)$.

Exercise 918 Let $V = C(-1, 1)$ and define an inner product on V by setting $\langle f, g \rangle = \int_{-1}^1 f(x)g(x)dx$. Let W be the subspace of V composed of all even functions. Find W^\perp .

Exercise 919 Let n be a positive integer and let $V = \mathcal{M}_{n \times n}(\mathbb{C})$, on which we have an inner product defined by $\langle A, B \rangle = \text{tr}(A^T \overline{B})$. Let W be the subspace of V consisting of all those matrices $A \in V$ satisfying $\text{tr}(A) = 0$. Find W^\perp .

Exercise 920 Let V be an inner product space having subspaces W and Y . Show that $(W + Y)^\perp = W^\perp \cap Y^\perp$.

Exercise 921 Define an inner product on \mathbb{R}^2 with respect to which the vectors $\begin{bmatrix} -1 \\ 2 \end{bmatrix}$ and $\begin{bmatrix} 2 \\ 4 \end{bmatrix}$ are orthogonal.

Exercise 922 Make use of the Gram-Schmidt process to find an orthonormal basis for the space \mathbb{R}^3 together with the dot product, beginning with

$$\text{the initial basis } \left\{ \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ -2 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} \right\}.$$

Exercise 923 Let V be an inner product space and let A be an orthonormal subset of V . Show that A is a maximal orthonormal subset if and only if for every $0_V \neq y \in V$ there exists a $v \in A$ satisfying $\langle v, y \rangle \neq 0$.

Exercise 924 Let V be an inner product space of finite dimension n over its field of scalars. Show that there exists a subset $\{v_1, \dots, v_{2n}\}$ of V satisfying the conditions that $\langle v_i, v_j \rangle \leq 0$ for all $1 \leq i \neq j \leq 2n$.

Exercise 925 Let V be the space of all polynomial functions in $\mathbb{R}^{\mathbb{R}}$ of degree less than 3, with inner product $\langle p, q \rangle = \frac{1}{2} \int_{-1}^1 p(t)q(t)dt$. Find an orthonormal basis $\{p_0, p_1, p_2\}$ of V satisfying $\deg(p_h) = h$ for $h = 0, 1, 2$.

Exercise 926 Consider the function $\mu : \mathbb{R}^3 \times \mathbb{R}^3 \rightarrow \mathbb{R}$ given by

$$\mu : \left(\begin{bmatrix} a \\ b \\ c \end{bmatrix}, \begin{bmatrix} a' \\ b' \\ c' \end{bmatrix} \right) \mapsto 2aa' + ac' + ca' + bb' + cc'.$$

Show that μ is an inner product and find a basis of \mathbb{R}^3 orthonormal with respect to μ .

Exercise 927 Let V be an inner product space over \mathbb{R} and let W be a finitely-generated subspace of V with orthonormal basis $\{w_1, \dots, w_n\}$. Let $\alpha \in \text{Hom}(V, W)$ be defined by $\alpha : v \mapsto \sum_{i=1}^n \langle v, w_i \rangle w_i$. Show that $\alpha(v) - v \in W^\perp$ for all $v \in V$ and that $\|\alpha(v) - v\| < \|w - v\|$ for all $\alpha(v) \neq w \in W$.

Exercise 928 Let W be the subspace of \mathbb{R}^4 spanned by linearly-indepen-

$$\text{dent subset } \left\{ \begin{bmatrix} 1 \\ 2 \\ 0 \\ 3 \end{bmatrix}, \begin{bmatrix} 4 \\ 0 \\ 5 \\ 8 \end{bmatrix}, \begin{bmatrix} 8 \\ 1 \\ 5 \\ 6 \end{bmatrix} \right\}, \text{ which is an inner product space}$$

with respect to the dot product. Make use of the Gram-Schmidt process to find an orthonormal basis for W .

Exercise 929 Let n be a positive integer and let $A \in \mathcal{M}_{n \times n}(\mathbb{R})$. If the set of rows of A is orthonormal with respect to the dot product, is the same true for the set of columns of A ?

Exercise 930 Let $n > k$ be positive integers and let $A \in \mathcal{M}_{k \times n}(\mathbb{R})$ satisfy the condition that its set of rows is orthonormal with respect to the dot product. Show that $(A^T A)^2 = A^T A$.

Exercise 931 Let n be a positive integer and let $A \in \mathcal{M}_{n \times n}(\mathbb{R})$. Show that A is symmetric if and only if for some $k < n$ there exists a matrix $B \in \mathcal{M}_{n \times k}(\mathbb{R})$ and a real number r such that $A = BB^T + rI$ and the columns of B are mutually orthogonal.

Exercise 932 Let W be the subspace of \mathbb{R}^4 , which is an inner product space with respect to the dot product, generated by

$$\left\{ \begin{bmatrix} 2 \\ 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \\ 2 \end{bmatrix}, \begin{bmatrix} 4 \\ 2 \\ 2 \\ 4 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix} \right\}.$$

Find an orthonormal basis for W and an orthonormal basis for W^\perp .

Exercise 933 Consider \mathbb{R}^4 as an inner product space with respect to the

dot product. Add two vectors to the set $\left\{ \frac{1}{6} \begin{bmatrix} 1 \\ 1 \\ 3 \\ -5 \end{bmatrix}, \frac{1}{2} \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix} \right\}$ in order

to get an orthonormal basis for this space.

Exercise 934 Let n be a positive integer and let $V = \mathbb{R}^n$, which is an inner product space with respect to the dot product. Let $\{v_1, \dots, v_n\}$ be an orthonormal basis for V , let $a \in \mathbb{R}$, and let $1 \leq h \neq k \leq n$. Define vectors w_1, \dots, w_n in V by setting

$$w_i = \begin{cases} \cos(a)v_h - \sin(a)v_k & \text{if } i = h \\ \sin(a)v_h + \cos(a)v_k & \text{if } i = k \\ v_i & \text{otherwise} \end{cases}.$$

Is $\{w_1, \dots, w_n\}$ an orthonormal basis for V ?

Exercise 935 Consider \mathbb{R}^3 as an inner product space with respect to the dot product. Does there exist an integer k such that

$$\left\{ \begin{bmatrix} 4k \\ 4 \\ -k \end{bmatrix}, \begin{bmatrix} 0 \\ k \\ 4 \end{bmatrix}, \begin{bmatrix} -25 \\ 16k \\ -12k \end{bmatrix} \right\}$$

is an orthonormal basis for this space.

Exercise 936 Consider \mathbb{R}^4 as an inner product space with respect to the

dot product. Find an orthonormal basis for $\mathbb{R} \left\{ \begin{bmatrix} 1 \\ 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ 2 \\ 1 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 1 \\ 2 \end{bmatrix} \right\}$.

Exercise 937 Define an inner product on \mathbb{R}^2 by setting

$$\left\langle \begin{bmatrix} a \\ b \end{bmatrix}, \begin{bmatrix} c \\ d \end{bmatrix} \right\rangle = ac + \frac{1}{2}(ad + bc) + bd.$$

Find an orthonormal basis for this space.

Exercise 938 Consider \mathbb{R}^3 as an inner product space with respect to the dot product. Let a, b, c, d be nonzero real numbers satisfying the conditions that $a^2 + b^2 + c^2 = d^2$ and $ab + ac = bc$. Show that the subset

$$\left\{ \frac{1}{d} \begin{bmatrix} a \\ b \\ c \end{bmatrix}, \frac{1}{d} \begin{bmatrix} b \\ -c \\ a \end{bmatrix}, \frac{1}{d} \begin{bmatrix} c \\ a \\ -b \end{bmatrix} \right\} \text{ of } \mathbb{R}^3 \text{ is orthonormal.}$$

Exercise 939 Define an inner product on \mathbb{R}^3 by setting

$$\left\langle \begin{bmatrix} a_1 \\ a_2 \\ a_3 \end{bmatrix}, \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix} \right\rangle = a_1 b_1 + 2(a_2 b_2 + a_3 b_3) - (a_1 b_2 + a_2 b_1) - (a_2 b_3 + a_3 b_2).$$

Find an orthonormal basis for this space.

Exercise 940 Let m be a positive integer and let

$$W = \{f \in \mathbb{C}^{\mathbb{Z}} \mid f(i+m) = f(i) \text{ for all } i \in \mathbb{Z}\},$$

which is a subspace of the vector space $\mathbb{C}^{\mathbb{Z}}$ over \mathbb{C} . Define a function $\mu : W \times W \rightarrow \mathbb{C}$ by setting $\mu : (f, g) \mapsto \sum_{h=0}^{m-1} f(h)\overline{g(h)}$. For each $0 \leq j < m$, let $f_j \in W$ be the function defined by

$$f_j(h) = \begin{cases} 1 & \text{if } h \text{ is of the form } j + mi, \text{ for } i \in \mathbb{Z} \\ 0 & \text{otherwise} \end{cases}.$$

Show that μ is an inner product on W and that, with respect to that product, $\{f_0, \dots, f_{m-1}\}$ is an orthonormal basis for W .

Exercise 941 A function $f \in \mathbb{R}^{\mathbb{R}}$ has **bounded support** if and only if there exist real numbers $a \leq b$ such that $f(x) = 0$ for all x not in the interval $[a, b]$ on the real line. Let V be the set of all such functions and define a function $\mu : V \times V \rightarrow \mathbb{R}$ by setting $\mu(f, g) = \int_{-\infty}^{\infty} f(x)g(x)dx$. For each $k \in \mathbb{Z}$, let $f_k \in V$ be the function defined by

$$f_k : x \mapsto \begin{cases} 1 & \text{if } k \leq x \leq k+1 \\ 0 & \text{otherwise} \end{cases}.$$

Show that μ is an inner product on V and that the subset $\{f_k \mid k \in \mathbb{Z}\}$ of V is orthonormal with respect to this inner product.

Exercise 942 Let V be the subspace of $\mathbb{R}^{\mathbb{R}}$ consisting of all infinitely-differentiable functions f which are periodic of period $h > 0$. (In other words, $f(x+h) = f(x)$ for all $x \in \mathbb{R}$). Define an inner product on V by setting $\langle f, g \rangle = \int_{-h}^h f(x)g(x)dx$. Let α be the endomorphism of V which assigns to every element of V its derivative. Find α^* .

Exercise 943 Let V be an inner product space over \mathbb{R} and let $\{v_1, \dots, v_n\}$ be a set of mutually-orthogonal nonzero vectors in V . Let a_1, \dots, a_n be positive real numbers satisfying $\sum_{i=1}^n a_i = 1$, and let $w = \sum_{i=1}^n a_i v_i$. Suppose that $w \perp v_i - v_j$ for all $1 \leq i \neq j \leq n$. Then show that $\|w\|^{-2} = \sum_{i=1}^n \|v_i\|^{-2}$.

Exercise 944 Let V be a finitely-generated inner product space and let $\alpha, \beta_1, \beta_2 \in \text{End}(V)$ satisfy $\alpha^* \alpha \beta_1 = \alpha^* \alpha \beta_2$. Show that $\alpha \beta_1 = \alpha \beta_2$.

17

Selfadjoint Endomorphisms

Let V be an inner product space. An endomorphism α of V is **selfadjoint** if and only if $\langle \alpha(v), w \rangle = \langle v, \alpha(w) \rangle$ for all $v, w \in V$. Such endomorphisms always exist since σ_c is selfadjoint for any $c \in \mathbb{R}$. Selfadjoint endomorphisms have important applications in mathematical models in physics. For example, in mathematical models of quantum theory, selfadjoint operators on the state space of a system represent measurements which can be performed on the system. Note that if $\alpha \in \text{End}(V)$ is selfadjoint, then $\langle \alpha(v), v \rangle = \langle v, \alpha(v) \rangle = \overline{\langle \alpha(v), v \rangle}$ and so $\langle \alpha(v), v \rangle \in \mathbb{R}$ for all $v \in V$.

Example: Let $V = C(0, 1)$, which is an inner product space over \mathbb{R} in which $\langle f, g \rangle = \int_0^1 f(x)g(x)dx$. Then the endomorphism α of V defined by $\alpha(f) : x \mapsto \int_0^1 \cos(x - y)f(y)dy$ for all $f \in V$ is selfadjoint.

(17.1) Proposition: Let V be an inner product space. Then:

- (1) If $\alpha \in \text{End}(V)$ has an adjoint α^* , then $\alpha + \alpha^*$ is self-adjoint;
- (2) If $\alpha \in \text{End}(V)$ is selfadjoint, so is $c\alpha$ for each $c \in \mathbb{R}$;
- (3) If $\alpha \in \text{End}(V)$ is selfadjoint, so is α^n for each positive integer n ;
- (4) If $\alpha, \beta \in \text{End}(V)$ are selfadjoint so are $\alpha + \beta$ and $\alpha \bullet \beta$, where \bullet is the Jordan product in $\text{End}(V)$;
- (5) If $\alpha \in \text{End}(V)$ is selfadjoint and $\beta \in \text{End}(V)$ has an adjoint, then $\beta\alpha\beta^*$ is selfadjoint.

Proof: (1). If $v, w \in V$ then

$$\begin{aligned}\langle (\alpha + \alpha^*)(v), w \rangle &= \langle \alpha(v), w \rangle + \langle \alpha^*(v), w \rangle \\ &= \langle v, \alpha^*(w) \rangle + \langle v, \alpha(w) \rangle = \langle v, (\alpha + \alpha^*)(w) \rangle.\end{aligned}$$

(2) If $v, w \in V$ then

$$\langle (c\alpha)(v), w \rangle = c \langle \alpha(v), w \rangle = c \langle v, \alpha(w) \rangle = \langle v, (c\alpha)(w) \rangle.$$

(3) This follows by an easy induction argument, using Proposition 16.16(3).

(4) The selfadjointness of $\alpha + \beta$ is an immediate consequence of Proposition 16.16(1). Also, recall that $\alpha \bullet \beta = \frac{1}{2}(\alpha\beta + \beta\alpha)$ and so, if $v, w \in V$ then

$$\begin{aligned}\langle (\alpha \bullet \beta)(v), w \rangle &= \frac{1}{2} \langle \alpha\beta(v), w \rangle + \frac{1}{2} \langle \beta\alpha(v), w \rangle \\ &= \frac{1}{2} \langle v, \beta\alpha(w) \rangle + \frac{1}{2} \langle v, \alpha\beta(w) \rangle = \langle v, (\alpha \bullet \beta)(w) \rangle.\end{aligned}$$

(5) By Proposition 16.16, we see that $(\beta\alpha\beta^*)^* = \beta^{**}\alpha\beta^* = \beta\alpha\beta^*$.

□

In particular, if α is a selfadjoint endomorphism of an inner product space V , and if $p(X) \in \mathbb{R}[X]$, then $p(\alpha)$ is selfadjoint. The product of selfadjoint endomorphisms of V need not be selfadjoint, as we will see in the example after Proposition 17.4.

Example: Let V be an inner product space over \mathbb{R} and let $\alpha \in \text{End}(V)$ be selfadjoint. Let $a, b \in \mathbb{R}$ satisfy the condition that $a^2 < 4b$. Then, by the previous remark, we know that $\beta = \alpha^2 + a\alpha + b\sigma_1$ is again a selfadjoint endomorphism of V . Moreover, if $0_V \neq v \in V$ then

$$\begin{aligned}\langle \beta(v), v \rangle &= \langle \alpha^2(v), v \rangle + a \langle \alpha(v), v \rangle + b \langle v, v \rangle \\ &= \langle \alpha(v), \alpha(v) \rangle + a \langle \alpha(v), v \rangle + b \langle v, v \rangle \\ &= \|\alpha(v)\|^2 + a \langle \alpha(v), v \rangle + b \|v\|^2.\end{aligned}$$

By Proposition 15.2, we know that $|\langle \alpha(v), v \rangle| \leq \|\alpha(v)\| \cdot \|v\|$ and so

$$\begin{aligned}\langle \beta(v), v \rangle &\geq \|\alpha(v)\|^2 - |a| \cdot \|\alpha(v)\| \cdot \|v\| + b \|v\|^2 \\ &= \left(\|\alpha(v)\| - \frac{1}{2}|a| \cdot \|v\| \right)^2 + \left(b - \frac{1}{4}a^2 \right) \|v\|^2 > 0.\end{aligned}$$

Thus $\beta(v) \neq 0_V$ for each $0_V \neq v \in V$, showing that β is monic. In particular, if V is finitely generated, this in fact shows that β is an automorphism of V .

Let V be a finitely-generated inner product space having an orthonormal basis D . If $\alpha \in \text{End}(V)$ and if $\Phi_{DD}(\alpha) = [a_{ij}]$, then we know from Proposition 16.15 that $\Phi_{DD}(\alpha^*) = \Phi_{DD}(\alpha)^H = [\bar{a}_{ij}]^T$. Therefore, if α is selfadjoint we have $a_{ij} = \bar{a}_{ji}$ for all $1 \leq i, j \leq n$. In particular, $a_{ii} = \bar{a}_{ii}$ for all $1 \leq i \leq n$ and so the diagonal entries in $\Phi_{DD}(\alpha)$ belong to \mathbb{R} . Matrices A over \mathbb{C} satisfying the condition that $A = A^H$ are known as **Hermitian matrices**. When we are working over \mathbb{R} , these are of course just the symmetric matrices. It is clear that the sum of Hermitian matrices is again a Hermitian matrix, but the product of Hermitian matrices is not necessarily Hermitian, just as we have seen that the product of symmetric matrices is not necessarily symmetric. We do note, however, that if a matrix $A \in \mathcal{M}_{n \times n}(\mathbb{C})$ is Hermitian then so is A^2 . Indeed, if A and B are Hermitian matrices in $\mathcal{M}_{n \times n}(\mathbb{C})$, then their Jordan product $\frac{1}{2}(AB + BA)$ is a Hermitian matrix and so, in particular, the product of a commuting pair of Hermitian matrices is again Hermitian. Moreover, any matrix $D \in \mathcal{M}_{n \times n}(\mathbb{C})$ can be written in the form $A + iB$, where $A = \frac{1}{2}(D + D^H)$ and $B = -\frac{1}{2i}(D - D^H)$ are both Hermitian matrices. If $A \in \mathcal{M}_{n \times n}(\mathbb{C})$ is Hermitian, then so is cA for any $c \in \mathbb{R}$, and so the set of all Hermitian matrices in $\mathcal{M}_{n \times n}(\mathbb{C})$ is a subspace of $\mathcal{M}_{n \times n}(\mathbb{C})$, considered as a vector space over \mathbb{R} ; indeed, it is a subalgebra of the commutative Jordan \mathbb{R} -algebra $\mathcal{M}_{n \times n}(\mathbb{C})^+$. However, this set is not closed under multiplication by complex scalars, and so it is not a vector space over \mathbb{C} .

We have already seen that if V is an inner product space and if $\alpha \in \text{End}(V)$ is selfadjoint, then $\langle \alpha(v), v \rangle \in \mathbb{R}$ for all $v \in V$. If V is finitely generated, the reverse is also true, as follows immediately from the following result.

(17.2) Proposition: Let V be an inner product space over \mathbb{C} and let $\alpha \in \text{End}(V)$ have an adjoint. If $\langle \alpha(v), v \rangle \in \mathbb{R}$ for all $v \in V$, then α is selfadjoint.

Proof: For vectors $v, w \in V$, we have

$$\langle \alpha(v + w), v + w \rangle = \langle \alpha(v), v \rangle + \langle \alpha(v), w \rangle + \langle \alpha(w), v \rangle + \langle \alpha(w), w \rangle$$

and since, by assumption, we know that the scalars $\langle \alpha(v + w), v + w \rangle$, $\langle \alpha(v), v \rangle$, and $\langle \alpha(w), w \rangle$ are all real, we see that $\langle \alpha(v), w \rangle + \langle \alpha(w), v \rangle \in \mathbb{R}$ as well. This implies that

$$\langle \alpha(v), w \rangle + \langle \alpha(w), v \rangle = \langle w, \alpha(v) \rangle + \langle v, \alpha(w) \rangle$$

and so $i \langle \alpha(v), w \rangle + i \langle \alpha(w), v \rangle = i \langle w, \alpha(v) \rangle + i \langle v, \alpha(w) \rangle$. Similarly,

$$\langle \alpha(v + iw), v + iw \rangle = \langle \alpha(v), v \rangle - i \langle \alpha(v), w \rangle + i \langle \alpha(w), v \rangle + \langle \alpha(w), w \rangle$$

and so $-i \langle \alpha(v), w \rangle + i \langle \alpha(w), v \rangle \in \mathbb{R}$. This implies that

$$-i \langle \alpha(v), w \rangle + i \langle \alpha(w), v \rangle = i \langle w, \alpha(v) \rangle - i \langle v, \alpha(w) \rangle$$

and so, multiplying by i and adding it to the previous result, we get $2 \langle \alpha(v), w \rangle = 2 \langle w, \alpha(v) \rangle$, whence $\langle \alpha(v), w \rangle = \langle w, \alpha(v) \rangle$. Therefore $\alpha = \alpha^*$. \square

(17.3) Proposition: Let V be an inner product space and let $\sigma_0 \neq \alpha \in \text{End}(V)$ be selfadjoint. Then there exists a vector $v \in V$ satisfying $\langle \alpha(v), v \rangle \neq 0$.

Proof: First assume that the field of scalars is \mathbb{C} . Then it is easy to check that if $v, w \in V$ then

$$\begin{aligned} \langle \alpha(v), w \rangle &= \frac{1}{4} [\langle \alpha(v+w), v+w \rangle - \langle \alpha(v-w), v-w \rangle] \\ &\quad + \frac{i}{4} [\langle \alpha(v+iw), v+iw \rangle - \langle \alpha(v-iw), v-iw \rangle]. \end{aligned}$$

Moreover, each term on the right-hand side of this equality is of the form $\langle \alpha(y), y \rangle$ for some $y \in V$, so if all of these were equal to 0 we would see that $\alpha(v) \perp w$ for all $v, w \in V$, which means that $\alpha(v) = 0_V$ for all $v \in V$, contradicting the hypothesis that $\sigma_0 \neq \alpha$. Thus the desired result must hold.

Now assume that the field of scalars is \mathbb{R} . Then for all $v, w \in V$ we have $\langle \alpha(w), v \rangle = \langle w, \alpha(v) \rangle = \langle \alpha(v), w \rangle$ and so

$$\langle \alpha(v), w \rangle = \frac{1}{4} [\langle \alpha(v+w), v+w \rangle - \langle \alpha(v-w), v-w \rangle].$$

Again, each term on the right-hand side of this equality is of the form $\langle \alpha(y), y \rangle$ for some $y \in V$ so if all of these were all equal to 0 we would have $\alpha(v) \perp w$ for all $v, w \in V$ which, as we have seen in the previous case, leads to a contradiction. \square

We now return to a new variant of a question we have already posed: if α is an endomorphism of an inner product space V , when does there exist an orthonormal basis of V composed of eigenvectors of α ? If such a basis exists, we say that α is **orthogonally diagonalizable**.

(17.4) Proposition: Let V be an inner product space and let $\alpha \in \text{End}(V)$ be selfadjoint. Then $\text{spec}(\alpha) \subseteq \mathbb{R}$ and eigenvectors of α associated with distinct eigenvalues are orthogonal.

Proof: Let c be an eigenvalue of α and let v be an eigenvector of α associated with c . Then $c \langle v, v \rangle = \langle cv, v \rangle = \langle \alpha(v), v \rangle = \langle v, \alpha(v) \rangle =$

$\langle v, cv \rangle = \bar{c} \langle v, v \rangle$ and so, since $v \neq 0_V$, we see that $c = \bar{c}$, proving that $c \in \mathbb{R}$. Thus we have shown the first assertion.

If c and d are distinct eigenvalues of α associated with eigenvectors v and w respectively, then $c \langle v, w \rangle = \langle cv, w \rangle = \langle \alpha(v), w \rangle = \langle v, \alpha(w) \rangle = \langle v, dw \rangle = d \langle v, w \rangle$. Since $c \neq d$, this implies that $\langle v, w \rangle = 0$. \square

Example: The endomorphisms α and β of \mathbb{C}^2 which are given by $\alpha : \begin{bmatrix} a \\ b \end{bmatrix} \mapsto \begin{bmatrix} b \\ a \end{bmatrix}$ and $\beta : \begin{bmatrix} a \\ b \end{bmatrix} \mapsto \begin{bmatrix} a \\ -b \end{bmatrix}$ can easily be seen to be selfadjoint. However, $\text{spec}(\beta\alpha) = \{i, -i\}$ and so, by Proposition 17.4, $\beta\alpha$ is not selfadjoint.

We note the following consequence of Proposition 17.4: if the matrix $A \in \mathcal{M}_{n \times n}(\mathbb{R})$ is symmetric, then all eigenvalues of A are real, and so the characteristic polynomial of A is completely reducible in $\mathbb{R}[X]$.

By Proposition 17.4 we see that if V is an inner-product space of finite dimension n over \mathbb{C} and if $\alpha \in \text{End}(V)$ is selfadjoint, then the eigenvalues of α can be written uniquely as an n -tuple $(c_1(\alpha), \dots, c_n(\alpha))$ of real numbers, the entries of which form a nonincreasing sequence. Weyl showed that if $\alpha, \beta \in \text{End}(V)$, if $1 \leq k \leq i \leq n$, and $1 \leq j \leq n - i + 1$, then $c_{i+j-1}(\alpha) + c_{n-j+1}(\beta) \leq c_i(\alpha + \beta) \leq c_{i-k+1}(\alpha) + c_k(\beta)$ and so, if $j = k = 1$, we have $c_i(\alpha) + c_n(\beta) \leq c_i(\alpha + \beta) \leq c_i(\alpha) + c_1(\beta)$ for each $1 \leq i \leq n$. In particular, $c_1(\alpha + \beta) \leq c_1(\alpha) + c_1(\beta)$ and $c_n(\alpha) + c_n(\beta) \leq c_n(\alpha + \beta)$.

We now turn to the problem of finding the eigenvalues of a selfadjoint endomorphism of a finitely-generated inner product space. This problem arises in many important applications. For example, let Γ be a (nondirected) graph with vertex set $\{1, \dots, n\}$. We associate to this graph a symmetric matrix, called the **adjacency matrix** $[a_{ij}]$, the entries of which are nonnegative integers, by setting a_{ij} to be the number of edges in Γ connecting vertex i to vertex j . The matrix represents a selfadjoint endomorphism of \mathbb{R}^n with respect to some basis and its spectrum can be used to derive important information about Γ . This technique has important applications in the analysis of computer networks and in such areas as chemistry, where it is used to make rough estimates of the electron density distribution of molecules.

(17.5) Proposition: If V is a nontrivial finitely-generated inner product space, then $\text{spec}(\alpha) \neq \emptyset$ for any selfadjoint endomorphism α of V .

Proof: Let α be a selfadjoint endomorphism of V . Choose an orthonormal basis $B = \{v_1, \dots, v_n\}$ for V and let $A = \Phi_{BB}(\alpha)$. Since

α is selfadjoint, we know that $A = A^H$. Let $W = \mathbb{C}^n$ on which we have defined the dot product. Then the endomorphism β of W defined by $\beta : w \mapsto Aw$ is selfadjoint. The degree of the characteristic polynomial $|XI - A|$ of β is $n > 0$ and so, by the Fundamental Theorem of Algebra, it has a root $c \in \mathbb{C}$. Thus the matrix $cI - A$ is singular and so there exists a nonzero vector $w \in W$ satisfying $Aw = cw$. In other words, $c \in \text{spec}(\beta)$. By Proposition 17.4, this implies that $c \in \mathbb{R}$ and so $c \in \text{spec}(\alpha)$, even if V is an inner product space over \mathbb{R} . \square

In particular, we learn from Proposition 17.5 that every symmetric matrix over \mathbb{R} has an eigenvalue. Compare this to the example we have already seen of a symmetric matrix in $\mathcal{M}_{2 \times 2}(GF(2))$ having no eigenvalues.

Let V be an inner product space finitely generated over \mathbb{C} and let α be a selfadjoint endomorphism of V . We know, by Proposition 17.4, that the eigenvalues of α are all real and that eigenvectors of α associated with distinct eigenvalues are orthogonal. Let us denote the eigenvalues of α by c_1, \dots, c_n where the indices are so chosen that $c_1 \geq \dots \geq c_n$. An important result known as the Courant-Fischer Minimax Theorem states that, for each $1 \leq k \leq n$, we have $c_k = \sup \{ \inf \{ \langle \alpha(w), w \rangle \mid w \in W \text{ and } \|w\| = 1 \} \}$, where the supremum runs over all subspaces W of V having dimension k .

Let us look at this from a different perspective. The function which assigns to each $0_V \neq v \in V$ the scalar $R_\alpha(v) = \langle v, \alpha(v) \rangle \|v\|^{-2}$ is called the **Rayleigh¹ quotient function**. Note that the projection π_v defined in connection with the Gram-Schmidt theorem satisfies the condition that $\pi_v : \alpha(v) \mapsto R_\alpha(v)v$. By what we have already seen, the image D of this function is contained in \mathbb{R} . Moreover, if v is an eigenvector of α with associated eigenvalue c , then $R_\alpha(v) = c$, and so $\emptyset \neq \text{spec}(\alpha) \subseteq D$. On the other hand, it is possible to show – though we will not do it here – that D is contained in the closed interval $[c_n, c_1]$ bounded by the largest



¹ Twentieth-century German mathematicians **Ernst Fischer** and **Richard Courant** studied spaces of functions. Courant, who headed the Mathematics Institute at the University of Göttingen, fled Germany in 1933 and founded a similar institute in New York City, which now bears his name. **John William Strutt, Lord Rayleigh**, was a 19th-century British physicist and applied mathematician, who made important contributions to mathematical physics and who won the Nobel prize in 1904 for his discovery of the inert gas argon.

and the smallest eigenvalues of α , both endpoints of which in fact belong to D . This observation can be used to define the **Rayleigh quotient iterative scheme** to find eigenvalues of a selfadjoint endomorphism α :

As an initial guess, choose a normal vector v_0 and let $d_0 = R_\alpha(v_0)$.

For $k = 0, 1, 2, \dots$ repeat the following steps:

(1) If $\alpha - d_k \sigma_1 \notin \text{Aut}(V)$, then d_k is an eigenvalue of α , and we are done;

(2) Otherwise, $\alpha - d_k \sigma_1 \in \text{Aut}(V)$. Set $y = (\alpha - d_k \sigma_1)^{-1}(v_k)$ and then compute $v_{k+1} = \|y\|^{-2} y$ and $d_{k+1} = R_\alpha(v_{k+1})$.

This scheme will indeed produce an eigenvalue of α for all guesses of v_0 except those in a set of measure 0, and when it converges, the convergence is very rapid. Its main disadvantage is the time and effort needed in step (1) of the iteration, to decide if $\alpha - d_k \sigma_1$ is an automorphism of V or not (usually, if the matrix representing this endomorphism is nonsingular or not) and, if it is, to compute its inverse; the algorithm is therefore worthwhile only if this can be done without major computational effort.

Example: Let α be the endomorphism of \mathbb{R}^3 represented with respect

to the canonical basis by the symmetric matrix $A = \begin{bmatrix} 2 & 1 & 1 \\ 1 & 3 & 1 \\ 1 & 1 & 4 \end{bmatrix}$. Then

α is selfadjoint. Choose $v_0 = \frac{1}{\sqrt{3}} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$. Using the above algorithm, we

see that $d_0 = R_\alpha(v_0) = 5$, which is not an eigenvalue of α . Moreover,

$$\begin{aligned} v_1 &= \begin{bmatrix} 0.3841106399\dots \\ 0.5121475201\dots \\ 0.7682212801\dots \end{bmatrix} & \text{and } d_1 &= 5.213114754\dots \\ v_2 &= \begin{bmatrix} 0.3971170680\dots \\ 0.5206615990\dots \\ 0.7557840528\dots \end{bmatrix} & \text{and } d_2 &= 5.214319743\dots \end{aligned}$$

The actual value of an eigenvalue of α is $5.214319744\dots$, so we see that convergence was very rapid indeed.

(17.6) Proposition: Let V be a finitely-generated inner product space and let $\alpha \in \text{End}(V)$. If W is a subspace of V invariant under α , then W^\perp is invariant under α^* .

Proof: If $w \in W$ and $y \in W^\perp$. Then $\alpha(w) \in W$ and so $\langle w, \alpha^*(y) \rangle = \langle \alpha(w), y \rangle = 0$, whence $\alpha^*(y) \in W^\perp$. \square

If V is a nontrivial inner product space finitely generated over \mathbb{R} and assume that $\alpha \in \text{End}(V)$ is orthogonally diagonalizable. Then there exists an orthonormal basis $B = \{v_1, \dots, v_n\}$ composed of eigenvectors of α . Thus $\Phi_{BB}(\alpha)$ is a diagonal matrix and so symmetric. In particular, $\Phi_{BB}(\alpha^*) = \Phi_{BB}(\alpha)^T = \Phi_{BB}(\alpha)$, which proves that $\alpha = \alpha^*$ and so α is selfadjoint. The converse of this result follows from the following Proposition.

(17.7) Proposition: Let V be a nontrivial finitely-generated inner product space and let $\alpha \in \text{End}(V)$ be selfadjoint. Then α is orthogonally diagonalizable.

Proof: We will prove the result by induction on $n = \dim(V)$. For $n = 1$, we know by Proposition 17.5 that α has an eigenvector $v \in V$, and so $\{v_1\}$ is the desired basis, where $v_1 = \|v\|^{-1}v$. Now assume that $n > 1$ and that the proposition has been established for all spaces of dimension less than n . Pick v_1 as before and let W be the subspace of V generated by $\{v_1\}$. Then $V = W \oplus W^\perp$ and, by Proposition 17.6, we know that W^\perp is invariant under $\alpha^* = \alpha$. Moreover, W^\perp is an inner product space of dimension $n - 1$ and the restriction of α to W^\perp is selfadjoint. Therefore, by the induction hypothesis, there exists an orthonormal basis $\{v_2, \dots, v_n\}$ of W^\perp composed of eigenvectors of α . Since v_1 is orthogonal to each of the vectors in this basis, we see that $\{v_1, \dots, v_n\}$ is an orthonormal basis of V . \square

Let V be an inner product space. An endomorphism $\alpha \in \text{End}(V)$ is **positive definite** if and only if it is selfadjoint and satisfies the condition that $\langle \alpha(v), v \rangle$ is a positive real number for all $0_V \neq v \in V$. Thus we see that σ_c is positive definite for any positive real number c . We also note that a positive-definite endomorphism must be monic since if $\alpha(v) = 0_V$ implies that $\langle \alpha(v), v \rangle = 0$ and so $v = 0_V$. Therefore every positive-definite endomorphism of a finitely-generated inner product space is in fact an automorphism. Positive definite endomorphisms have important applications in optimization and linear programming. Note that if α is positive definite and if $0_V \neq v \in V$ then $0 < \langle \alpha(v), v \rangle = \|\alpha(v)\| \cdot \|v\| \cos(t)$, where t is the angle between $\alpha(v)$ and v , showing that $0 \leq t < \frac{\pi}{2}$.

Example: Let $V = \mathbb{R}^n$ on which we have the dot product defined, and let B be the canonical basis. An endomorphism α of V is positive definite if and only if $A = \Phi_{BB}(\alpha)$ is a symmetric matrix satisfying the condition that $v^T A v > 0$ for all nonzero vectors $v \in V$. Such matrices have nice properties. For example, it can be shown that if A is of this

form then the Gauss-Seidel method applied to an equation $AX = w$ will converge to the unique solution v , for any initial guess v_0 chosen.

Example: Even if $\sigma_0 \neq \alpha \in \text{End}(V)$ is selfadjoint, it may be the case that neither α nor $-\alpha$ is positive definite. For example, if $V = \mathbb{R}^2$ and if $\alpha \in \text{End}(V)$ is defined by $\alpha : \begin{bmatrix} a \\ b \end{bmatrix} \mapsto \begin{bmatrix} -a \\ b \end{bmatrix}$, then

$$\alpha \left(\begin{bmatrix} 1 \\ 0 \end{bmatrix} \right) \cdot \begin{bmatrix} 1 \\ 0 \end{bmatrix} = -1 = (-\alpha) \left(\begin{bmatrix} 0 \\ 1 \end{bmatrix} \right) \cdot \begin{bmatrix} 0 \\ 1 \end{bmatrix}.$$

Let V an inner product space. If $\alpha, \beta \in \text{End}(V)$ are selfadjoint, then $\alpha - \beta$ is also selfadjoint. We will write $\alpha > \beta$ whenever $\alpha - \beta$ is positive definite. Thus, α is positive definite if and only if $\alpha > \sigma_0$. We will write $\alpha \geq \beta$ if and only if $\alpha > \beta$ or $\alpha = \beta$. We claim that this is a partial-order relation on the set of all selfadjoint endomorphisms of V . Indeed, it is sure that $\alpha \geq \alpha$ for all such endomorphisms α . Suppose that α_1, α_2 , and α_3 are selfadjoint endomorphisms of V satisfying $\alpha_1 \geq \alpha_2 \geq \alpha_3$. If $\alpha_1 = \alpha_2$ or $\alpha_2 = \alpha_3$ then it is clear that $\alpha_1 \geq \alpha_3$. Let us therefore assume that $\alpha_1 > \alpha_2 > \alpha_3$. Then for all $v \in V$ we see that

$$\begin{aligned} \langle (\alpha_1 - \alpha_3)(v), v \rangle &= \langle \alpha_1(v) - \alpha_2(v) + \alpha_2(v) - \alpha_3(v), v \rangle \\ &= \langle \alpha_1(v) - \alpha_2(v), v \rangle + \langle \alpha_2(v) - \alpha_3(v), v \rangle > 0 \end{aligned}$$

and so $\alpha_1 > \alpha_3$. Finally, assume that $\alpha_1 \geq \alpha_2$ and $\alpha_2 \geq \alpha_1$ but $\alpha_1 \neq \alpha_2$. Then $\alpha_1 > \alpha_2 > \alpha_1$ and so, as we have seen, $\alpha_1 > \alpha_1$, which is a contradiction. Thus we have a partial order on the set of all selfadjoint endomorphisms of V , called the **Loewner² partial order**.

(17.8) Proposition: Let V be an inner product space and let $\alpha \in \text{End}(V)$ be an endomorphism for which α^* exists. Then α is positive definite if and only if the function $\mu : (v, w) \mapsto \langle \alpha(v), w \rangle$ from $V \times V$ to the field of scalars is also an inner product.



² **Karl Loewner** was a Czech mathematician who emigrated to the United States in 1933. His research concentrated in complex function theory and spaces of functions.

Proof: First, let us assume that α is a positive-definite endomorphism of V . If $v_1, v_2, w \in V$ then $\mu(v_1 + v_2, w) = \langle \alpha(v_1 + v_2), w \rangle = \langle \alpha(v_1), w \rangle + \langle \alpha(v_2), w \rangle = \mu(v_1, w) + \mu(v_2, w)$ and, similarly, we show that $\mu(cv, w) = c\mu(v, w)$ for all scalars c . We also see that $\mu(v, w) = \langle \alpha(v), w \rangle = \langle v, \alpha^*(w) \rangle = \langle v, \alpha(w) \rangle = \langle \alpha(w), v \rangle = \mu(w, v)$. If $0_V \neq v \in V$ then, by the assumption of positive definiteness, we see that $\mu(v, v) = \langle \alpha(v), v \rangle$ is a positive real number, and it is clear that $\mu(0_V, 0_V) = 0$. Thus μ is an inner product on V .

Conversely, assume that μ is an inner product on V . Then for all $v, w \in V$ we have $\langle v, \alpha^*(w) \rangle = \langle \alpha(v), w \rangle = \mu(v, w) = \overline{\mu(w, v)} = \overline{\langle \alpha(w), v \rangle} = \langle v, \alpha(w) \rangle$ and so $\alpha(w) = \alpha^*(w)$ for all $w \in V$, proving that α is selfadjoint. Moreover, for all $v \in V$ we have $\langle \alpha(v), v \rangle = \mu(v, v)$ for all $0_V \neq v \in V$ and so α is positive definite. \square

(17.9) Proposition: Let V be an inner product space, with given inner product $(v, w) \mapsto \langle v, w \rangle$, and let μ be another inner product defined on V . Then there exists a unique positive-definite endomorphism α of V satisfying the condition that $\mu(v, w) = \langle \alpha(v), w \rangle$ for all $v, w \in V$.

Proof: Fix a vector $w \in V$. The function $v \mapsto \mu(v, w)$ belongs to $D(V)$ and so there exists a unique vector $y_w \in V$ satisfying $\mu(v, w) = \langle v, y_w \rangle$ for all $v \in V$. Define a function $\alpha : V \rightarrow V$ by $\alpha : w \mapsto y_w$. Then $\langle \alpha(v), w \rangle = \overline{\langle w, \alpha(v) \rangle} = \overline{\mu(w, v)} = \mu(v, w)$ for all $v, w \in V$. We claim that $\alpha \in \text{End}(V)$. Indeed, if $w_1, w_2 \in V$ then for all $y \in V$ we have

$$\begin{aligned} \langle \alpha(w_1 + w_2), y \rangle &= \mu(w_1 + w_2, y) = \mu(w_1, y) + \mu(w_2, y) \\ &= \langle \alpha(w_1), y \rangle + \langle \alpha(w_2), y \rangle = \langle \alpha(w_1) + \alpha(w_2), y \rangle \end{aligned}$$

and so $\alpha(w_1 + w_2) = \alpha(w_1) + \alpha(w_2)$. Similarly we can show that $\alpha(cw) = c\alpha(w)$ for all $w \in V$ and all scalars c . Thus we see that α is indeed an endomorphism of V satisfying the condition $\mu(v, w) = \langle \alpha(v), w \rangle$ for all $v, w \in V$, and so it is positive definite.

Finally, α has to be unique since if $\mu(v, w) = \langle \beta(v), w \rangle$ for all $v, w \in V$, then $\langle (\alpha - \beta)(v), w \rangle = \langle \alpha(v) - \beta(v), w \rangle = \langle \alpha(v), w \rangle - \langle \beta(v), w \rangle = 0$ for all $v, w \in V$, which implies that $(\alpha - \beta)(v) = 0_V$ for all $v \in V$, showing that $\alpha = \beta$. \square

(17.10) Proposition: Let V be a finitely-generated inner product space and let $\alpha \in \text{End}(V)$. Then α is positive definite if and only if there exists an automorphism β of V satisfying $\alpha = \beta^*\beta$.

Proof: Assume that there exists an automorphism β of V satisfying $\alpha = \beta^*\beta$. Then, as previously noted, α is selfadjoint. Moreover, for all $0_V \neq v \in V$ we have $\langle \alpha(v), v \rangle = \langle \beta^*\beta(v), v \rangle = \langle \beta(v), \beta^{**}(v) \rangle =$

$\langle \beta(v), \beta(v) \rangle > 0$ since β is an automorphism and hence $\beta(v) \neq 0_V$. Therefore α is positive definite.

Conversely, assume that α is a positive-definite endomorphism of V . Then the function $\mu : (v, w) \mapsto \langle \alpha(v), w \rangle$ is an inner product on V . Let $\{v_1, \dots, v_n\}$ be a basis for V which is orthonormal with respect to the original inner product on V and let $\{w_1, \dots, w_n\}$ be a basis for V which is orthonormal with respect to μ . By Proposition 6.2, we know that there exists a unique endomorphism β of V satisfying $\beta(w_i) = v_i$ for all $1 \leq i \leq n$. Then β is an epimorphism since its image contains a basis for V and so, since V is finitely-generated, it is an automorphism of V . Therefore, if $v = \sum_{i=1}^n a_i w_i$ and $w = \sum_{j=1}^n b_j w_j$ are vectors in V we see that $\langle \alpha(v), w \rangle = \mu(v, w) = \mu\left(\sum_{i=1}^n a_i w_i, \sum_{j=1}^n b_j w_j\right) = \sum_{i=1}^n \sum_{j=1}^n a_i \bar{b}_j \mu(w_i, w_j) = \sum_{i=1}^n a_i \bar{b}_i$ and similarly $\langle \beta^* \beta(v), w \rangle = \langle \beta(v), \beta(w) \rangle = \left\langle \beta\left(\sum_{i=1}^n a_i w_i\right), \beta\left(\sum_{j=1}^n b_j w_j\right) \right\rangle = \sum_{i=1}^n \sum_{j=1}^n a_i \bar{b}_j \langle w_i, w_j \rangle = \sum_{i=1}^n a_i \bar{b}_i$, and so we see that $\langle \beta^* \beta(v), w \rangle = \langle \alpha(v), w \rangle$ for all $v, w \in V$, which shows that $\alpha = \beta^* \beta$. \square

Example: In order for an endomorphism to be positive definite, it is not enough for it to be represented by a matrix all of the entries of which are positive real numbers. For example, if $V = \mathbb{R}^2$ with the dot product, and if α is the endomorphism of V represented with respect to the

canonical basis by the matrix $\begin{bmatrix} 2 & 2 \\ 2 & 1 \end{bmatrix}$, then $\alpha\left(\begin{bmatrix} -1 \\ 1 \end{bmatrix}\right) \cdot \begin{bmatrix} -1 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ -1 \end{bmatrix} \cdot \begin{bmatrix} -1 \\ 1 \end{bmatrix} = -1$, so α is not positive definite.

From Proposition 17.10 we know that if $A = [a_{ij}] \in \mathcal{M}_{n \times n}(\mathbb{C})$ is a matrix representing a positive-definite endomorphism of \mathbb{C}^n , namely if it is a Hermitian matrix satisfying the condition that $v \cdot Av > 0$ for all nonzero vectors $v \in \mathbb{C}^n$, then there exists a nonsingular matrix B such that $A = B^H B$. Indeed, we can choose B to be upper triangular, so that it is a form of LU-decomposition, though it takes only half as many arithmetic operations to perform. This decomposition is known as a **Cholesky**³



³ Major **André-Louis Cholesky** was a cartographer in the French army, who used this method in connection with the mapping of the island of Crete before World War I. It had previously been used by other cartographers, including **Myrick H. Doolittle**, of the computing division of the US Coast and Geodetic Survey, in 1878.

decomposition of A . This decomposition need not be unique. Cholesky decompositions are widely used in building economic and financial models. Because of this wide usage, there are many algorithms available to efficiently calculate Cholesky decompositions of general matrices or of matrices in special forms. Indeed, one of the computational advantages of the Cholesky decomposition is that it is numerically stable, even with no pivoting.

The following algorithm calculates a Cholesky decomposition for real symmetric matrices.

For $k = 1, \dots, n$ perform the following steps:

- (1) For each $1 \leq i < k$ define $b_{ik} = b_{ii}^{-1} \left[a_{ik} - \sum_{j=1}^{i-1} b_{ji} b_{jk} \right]$;
- (2) Set $b_{kk} = \sqrt{a_{kk} - \sum_{j=1}^{k-1} b_{jk}^2}$;
- (3) For each $k < i \leq n$ set $b_{ik} = 0$.

Note that if the matrix A did not satisfy $v \cdot Av > 0$ for all nonzero vectors v , the algorithm would hang up at some stage, trying to take the square root of a negative number. Indeed, attempting a Cholesky decomposition is often used as a test to see whether a given matrix represents a positive-definite endomorphism or not.

Example: Let $A = \begin{bmatrix} 5 & 2 & 3 \\ 2 & 1 & 1 \\ 3 & 1 & 4 \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{R})$. This is a symmet-

ric matrix satisfying the condition that $v \cdot Av > 0$ for all nonzero vectors $v \in \mathbb{R}^3$ and having a Cholesky decomposition $B^T B$, where $B =$

$$\frac{1}{5} \begin{bmatrix} 5\sqrt{5} & 2\sqrt{5} & 3\sqrt{5} \\ 0 & \sqrt{5} & -\sqrt{5} \\ 0 & 0 & 5\sqrt{2} \end{bmatrix}.$$

Notice that the Proposition 17.10 extends our ongoing analogy between the operation $*$ and the conjugate operation on \mathbb{C} , just as the notion of “positive definite” is the analog of positivity of complex numbers: a complex number z is (real and) positive if and only if there exists a complex number y such that $z = \bar{y}y$.

(17.11) Proposition: Let V be an inner product space. If $\alpha \in \text{End}(V)$ is positive definite, then every eigenvalue of α is a positive real number. The converse holds if α is orthogonally diagonalizable.

Proof: Assume that α is positive definite. By Proposition 17.4, the eigenvalues of α are real numbers. If $c \in \text{spec}(\alpha)$ is an eigenvalue of α associated with an eigenvector v , then $0 < \langle \alpha(v), v \rangle = \langle cv, v \rangle = c \langle v, v \rangle$ and so $c > 0$, since we know that $\langle v, v \rangle > 0$. Conversely, assume that every eigenvalue of α is positive and that there exists an orthonormal basis B of V composed of eigenvectors of α . Let $v = \sum_{i=1}^n a_i v_i$, where $\{v_1, \dots, v_n\} \subseteq B$. For each $1 \leq i \leq n$, let c_i be an eigenvalue of α associated with v_i . We can assume that the v_i are arranged in such a manner that $0 < c_1 \leq c_2 \leq \dots \leq c_n$. Then

$$\begin{aligned} \langle \alpha(v), v \rangle &= \left\langle \sum_{i=1}^n \alpha(a_i v_i), \sum_{j=1}^n a_j v_j \right\rangle = \sum_{i=1}^n \sum_{j=1}^n a_i \bar{a}_j \langle c_i v_i, v_j \rangle \\ &= \sum_{i=1}^n \sum_{j=1}^n c_i a_i \bar{a}_j \langle v_i, v_j \rangle = \sum_{i=1}^n c_i |a_i|^2 \geq c_1 \sum_{i=1}^n |a_i|^2 > 0 \end{aligned}$$

and so α is positive definite. \square

From Propositions 17.11 and 17.7, we see that if V is an finitely-generated inner product space over \mathbb{R} or \mathbb{C} and if $\alpha \in \text{End}(V)$ is positive definite, then there exists a basis of V relative to which α is represented by a diagonal matrix in which the entries of the diagonal are positive real numbers. Such a matrix is, of course, nonsingular.

(17.12) Proposition: Let V and W be inner-product spaces finitely-generated over \mathbb{R} and let $\alpha \in \text{Hom}(V, W)$. Then $\|\alpha\| = \sqrt{c}$, where c is the largest eigenvalue of $\alpha^* \alpha \in \text{End}(V)$, and where $\|\alpha\|$ is the norm induced by the respective inner products on V and W .

Proof: If c is an eigenvalue of $\beta = \alpha^* \alpha$ then there exists a nonzero vector v such that $\beta(v) = cv$ and so $c \|v\|^2 = \langle v, cv \rangle = \langle v, \alpha^* \alpha(v) \rangle = \langle \alpha(v), \alpha(v) \rangle \geq 0$, and so $c \geq 0$. By Proposition 17.7, we know that there exists a basis $\{v_1, \dots, v_n\}$ of V composed of orthonormal eigenvectors of β . For each $1 \leq i \leq n$, let c_i be an eigenvalue of β associated with v_i . After renumbering, we can assume that $0 \leq c_1 \leq \dots \leq c_n$. If $v = \sum_{i=1}^n a_i v_i \in V$, then

$$\begin{aligned} \|\alpha(v)\|^2 &= \langle v, \beta(v) \rangle = \langle v, \alpha^* \alpha(v) \rangle = \sum_{j=1}^n \sum_{i=1}^n a_j c_i a_i \langle v_j, v_i \rangle \\ &= \sum_{i=1}^n c_i a_i^2 \leq c_n \left(\sum_{i=1}^n a_i^2 \right) = c_n \|v\|^2 \end{aligned}$$

and so $\|\alpha(v)\|^2 / \|v\|^2 \leq c_n$. Therefore, by definition of the induced norm, $\|\alpha\| \leq \sqrt{c_n}$. But one easily sees that $\|\alpha(v_n)\|^2 / \|v_n\|^2 = c_n$ and so $\sqrt{c_n} \leq \|\alpha\|$, proving equality. \square

Example: Let $\alpha : \mathbb{R}^3 \rightarrow \mathbb{R}^2$ be the linear transformation defined by $\alpha : v \mapsto Av$, where $A = \begin{bmatrix} 1 & -2 & 1 \\ 3 & 0 & -1 \end{bmatrix}$. Then $\alpha^* \alpha$ is the endomorphism of \mathbb{R}^3 given by $v \mapsto \begin{bmatrix} 10 & -2 & -2 \\ -2 & 4 & -2 \\ -2 & -2 & 2 \end{bmatrix} v$. The eigenvalues of this endomorphism are $0 \leq 8-2\sqrt{2} \leq 8+2\sqrt{2}$ and so $\|\alpha\| = \sqrt{8+2\sqrt{2}}$, which is approximately equal to 3.291.

Let V and W be inner product spaces. A linear transformation $\alpha : V \rightarrow W$ preserves inner products if and only if $\langle v_1, v_2 \rangle = \langle \alpha(v_1), \alpha(v_2) \rangle$ for all $v_1, v_2 \in V$. Notice that any linear transformation which preserves inner products also preserves distances: $\|v_1 - v_2\| = \|\alpha(v_1 - v_2)\| = \|\alpha(v_1) - \alpha(v_2)\|$ for all $v_1, v_2 \in V$. Also, as a direct consequence of the definition, such a linear transformation preserves the angles between vectors. Conversely, we have already noted that from the norm defined by an inner product we can recover the inner product itself, so that any linear transformation $\alpha : V \rightarrow W$ satisfying $\|v_1 - v_2\| = \|\alpha(v_1) - \alpha(v_2)\|$ for all v_1, v_2 also preserves inner products. Such a linear transformation is called an **isometry**.

We also note that if α is an endomorphism of an inner product space V which is an isometry, and if c is an eigenvalue of α with associated eigenvector v , then $\|v\|^2 = \|cv\|^2 = |c|^2 \|v\|^2$ and so $|c| = 1$. Thus, if V is an inner product space over \mathbb{C} and if $\alpha \in \text{End}(V)$ is an isometry, then the eigenvalues of α lie on the unit circle $\{z \in \mathbb{C} \mid |z| = 1\}$.

Example: Let V be an inner product space over \mathbb{R} and let $0_V \neq y \in V$. This vector y defines an endomorphism α_y of V by setting

$$\alpha_y : v \mapsto -v + 2 \frac{\langle v, y \rangle}{\langle y, y \rangle} y.$$

This endomorphism is an isometry which satisfies $\alpha_y^2 = \sigma_1$, and y is a fixed point of α_y .

(17.13) Proposition: Let V and W be finitely-generated inner product spaces and having equal dimensions. Then the following conditions on a linear transformation $\alpha : V \rightarrow W$ are equivalent:

- (1) α is an isometry;
- (2) α is an isomorphism which is an isometry;

(3) If $\{v_1, \dots, v_n\}$ is an orthonormal basis of V then the set $\{\alpha(v_1), \dots, \alpha(v_n)\}$ is an orthonormal basis of W .

Proof: (1) \Rightarrow (2): If $0_V \neq v \in V$ then $\langle v, v \rangle = \langle \alpha(v), \alpha(v) \rangle$ and so $\alpha(v) \neq 0_W$. Thus we see that $\ker(\alpha) = \{0_V\}$ and so α is an isomorphism, since V and W have the same finite dimension.

(2) \Rightarrow (3): If $\{v_1, \dots, v_n\}$ is an orthonormal basis of V then, since α is an isomorphism, we see that $\{\alpha(v_1), \dots, \alpha(v_n)\}$ is a basis for W . Moreover, for all $1 \leq i, j \leq n$ we know that

$$\langle \alpha(v_i), \alpha(v_j) \rangle = \langle v_i, v_j \rangle = \begin{cases} 1 & \text{when } i = j \\ 0 & \text{otherwise} \end{cases}$$

and so this basis is orthonormal.

(3) \Rightarrow (1): Let $\{v_1, \dots, v_n\}$ be an orthonormal basis of V . If $v = \sum_{i=1}^n a_i v_i$ and $y = \sum_{j=1}^n b_j v_j$, then $\langle v, y \rangle = \sum_{i=1}^n a_i \bar{b}_i$. Moreover,

$$\begin{aligned} \langle \alpha(v), \alpha(y) \rangle &= \left\langle \alpha \left(\sum_{i=1}^n a_i v_i \right), \alpha \left(\sum_{j=1}^n b_j v_j \right) \right\rangle \\ &= \sum_{i=1}^n \sum_{j=1}^n a_i \bar{b}_j \langle \alpha(v_i), \alpha(v_j) \rangle = \sum_{i=1}^n a_i \bar{b}_i = \langle v, y \rangle \end{aligned}$$

and this proves (1). \square

In particular, if V and W are finitely-generated inner product spaces having equal dimensions, then every isometry $\alpha : V \rightarrow W$ is an isomorphism. If $w_1, w_2 \in W$, then $\langle w_1, w_2 \rangle = \langle \alpha\alpha^{-1}(w_1), \alpha\alpha^{-1}(w_2) \rangle = \langle \alpha^{-1}(w_1), \alpha^{-1}(w_2) \rangle$ and so we see that α^{-1} is also an isometry. Moreover, there is always at least one isometry α from V to W . Just pick orthonormal bases $\{v_1, \dots, v_n\}$ for V and $\{w_1, \dots, w_n\}$ for W and define α by $\alpha : \sum_{i=1}^n a_i v_i \mapsto \sum_{i=1}^n a_i w_i$.

Example: Let W be the set of all matrices $A \in \mathcal{M}_{3 \times 3}(\mathbb{R})$ satisfying $A^T = -A$, which is a subspace of $\mathcal{M}_{3 \times 3}(\mathbb{R})$ of dimension 3. Define an inner product on W as follows: if $A, B \in W$ then $\langle A, B \rangle = \frac{1}{2} \text{tr}(AB^T)$. Let $V = \mathbb{R}^3$, which is an inner product space with respect to the dot product. Define a linear transformation $\alpha : V \rightarrow W$ by

$$\begin{aligned} \text{setting } \alpha : \begin{bmatrix} a \\ b \\ c \end{bmatrix} &\mapsto \begin{bmatrix} 0 & -c & b \\ c & 0 & -a \\ -b & a & 0 \end{bmatrix}. \text{ If } A = \begin{bmatrix} 0 & -c & b \\ c & 0 & -a \\ -b & a & 0 \end{bmatrix} \text{ and} \\ B = \begin{bmatrix} 0 & -f & e \\ f & 0 & -d \\ -e & d & 0 \end{bmatrix} &\text{ then } AB^T = \begin{bmatrix} cf + be & -bd & -dc \\ ea & cf + ad & ec \\ -af & fb & be + ad \end{bmatrix} \end{aligned}$$

and so we can check that $\langle A, B \rangle = \begin{bmatrix} a \\ b \\ c \end{bmatrix} \cdot \begin{bmatrix} d \\ e \\ f \end{bmatrix}$ and thus α is an isometry and hence is an isomorphism.

Example: Proposition 17.13 is no longer true if we remove the condition that the spaces are finitely generated. Indeed, let $V = C(0, 1)$, on which we have the inner product $\mu(f, g) = \int_0^1 f(x)g(x)x^2 dx$, and let W be the same space on which we have the inner product $\langle f, g \rangle = \int_0^1 f(x)g(x)dx$. Let $\alpha : V \rightarrow W$ be the linear transformation defined by $\alpha : f(x) \mapsto xf(x)$. Then $\mu(f, g) = \langle \alpha(f), \alpha(g) \rangle$ and so α is an isometry. But α is not an isomorphism since the function $x \mapsto x^2 + 1$ does not belong to the image of α .

Let us now return to the case of inner product spaces the dimensions of which are not necessarily equal.

(17.14) Proposition: Let V and W be inner product spaces finitely-generated over \mathbb{R} and let $\alpha \in \text{Hom}(V, W)$. Then α is an isometry if and only if $\alpha^* \alpha = \sigma_1 \in \text{End}(V)$.

Proof: By Proposition 16.14, α^* exists. If $\alpha^* \alpha = \sigma_1 \in \text{End}(V)$, and if $v_1, v_2 \in V$ then

$$\begin{aligned} \|v_1 - v_2\|^2 &= \langle v_1 - v_2, v_1 - v_2 \rangle = \langle v_1 - v_2, \alpha^* \alpha(v_1 - v_2) \rangle \\ &= \langle \alpha(v_1 - v_2), \alpha(v_1 - v_2) \rangle = \|\alpha(v_1) - \alpha(v_2)\|^2 \end{aligned}$$

and so $\|v_1 - v_2\| = \|\alpha(v_1) - \alpha(v_2)\|$, proving that α is an isometry. Conversely, if α is an isometry and if $v_1, v_2 \in V$ then $\langle \alpha^* \alpha(v_1), v_2 \rangle = \langle \alpha(v_1), \alpha(v_2) \rangle = \langle v_1, v_2 \rangle$. Therefore, by Proposition 16.13, we see that $\alpha^* \alpha(v_1) = v_1$ for all $v_1 \in V$, proving that $\alpha^* \alpha = \sigma_1 \in \text{End}(V)$. \square

Exercises

Exercise 945 Let $V = \mathbb{C}[X]$ and define an inner product on V by setting $\langle \sum_{i=0}^{\infty} a_i X^i, \sum_{i=0}^{\infty} b_i X^i \rangle = \sum_{i=0}^{\infty} a_i \bar{b}_i$. Let α be the endomorphism of V defined by $\alpha : p(X) \mapsto (X+1)p(X)$. Calculate α^* , or show that it does not exist.

Exercise 946 Let $V = \mathbb{C}[X]$ and define an inner product on V by setting $\langle \sum_{i=0}^{\infty} a_i X^i, \sum_{i=0}^{\infty} b_i X^i \rangle = \sum_{i=0}^{\infty} a_i \bar{b}_i$. Let β be the endomorphism of V defined by $\beta : p(X) \mapsto p(X+1)$. Calculate β^* , or show that it does not exist.

Exercise 947 Let $p > 1$ be an integer, let $G = \mathbb{Z}/(p)$, and let $V = \mathbb{C}^G$, which is an inner product space over \mathbb{C} with inner product defined by $\langle f, g \rangle = \sum_{n \in G} f(n) \overline{g(n)}$. Let α be the endomorphism of V defined by $\alpha(f) : n \mapsto f(n+1) + f(n-1)$. Is α selfadjoint?

Exercise 948 Let V be a vector space over \mathbb{R} . A nonempty subset K of V is **convex** if and only if $cv + (1-c)w \in K$ whenever $v, w \in K$ and $0 \leq c \leq 1$. Is the set of all selfadjoint endomorphisms of an inner product space Y over \mathbb{R} necessarily a convex subset of the vector space $\text{End}(Y)$?

Exercise 949 Let V be an inner product space and let α be an endomorphism of V . Is the endomorphism $\alpha^* \alpha - \sigma_1$ of V selfadjoint?

Exercise 950 Let n be a positive integer and let V be the space of all polynomial functions in $\mathbb{R}^{\mathbb{R}}$ of degree at most n . Define an inner product on V by setting $\langle f, g \rangle = \int_{-1}^1 f(t)g(t)dt$. Let $\alpha \in \text{End}(V)$ be defined by $\alpha(f) : x \mapsto (1-x^2)f''(x) - 2xf'(x)$. Show that α is self-adjoint.

Exercise 951 Let V be an inner product space finitely generated over \mathbb{C} and let α be an endomorphism of V satisfying $\alpha\alpha^* = \alpha^2$. Show that α is selfadjoint.

Exercise 952 Let V be an inner product space finitely generated over \mathbb{C} and let α and β be selfadjoint endomorphisms of V satisfying the condition that $\alpha\beta$ is a projection. Is $\beta\alpha$ necessarily also a projection?

Exercise 953 Give an example of nonzero Hermitian matrices A and B satisfying $AB = O = BA$, or show no such matrices exist.

Exercise 954 Let $A \in \mathcal{M}_{2 \times 2}(\mathbb{C})$ be Hermitian. Find real numbers w , x , y , and z satisfying $|A| = w^2 - x^2 - y^2 - z^2$.

Exercise 955 Let n be a positive integer and let $A \in \mathcal{M}_{n \times n}(\mathbb{C})$. Show that A is a Hermitian matrix if and only if the matrix $B = iA$ satisfies $B = -B^H$.

Exercise 956 Let $O \neq A \in \mathcal{M}_{3 \times 3}(\mathbb{C})$ be a Hermitian matrix. Show that $A^k \neq O$ for all positive integers k .

Exercise 957 Find complex numbers a and b such that
$$\begin{bmatrix} a & 0 & b \\ 0 & 2a & a \\ i & 1 & a \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{C})$$
 is a Hermitian matrix.

Exercise 958 A matrix $A \in \mathcal{M}_{n \times n}(\mathbb{C})$ is **anti-Hermitian** if and only if $A^H = -A$. Show that A is anti-Hermitian if and only if iA is Hermitian.

Exercise 959 If matrices $A, B \in \mathcal{M}_{n \times n}(\mathbb{C})$ are anti-Hermitian, show that the Lie product of A and B is also anti-Hermitian.

Exercise 960 Let n be a positive integer and let $A \in \mathcal{M}_{n \times n}(\mathbb{C})$. Show that every eigenvalue of $A^H A$ is a positive real number.

Exercise 961 Let V be an inner product space and let $\alpha \in \text{End}(V)$ be selfadjoint. Show that $\ker(\alpha) = \ker(\alpha^h)$ for all $h \geq 1$.

Exercise 962 Let V be a nontrivial finitely-generated inner product space and let $\alpha \in \text{End}(V)$ be orthogonally diagonalizable and satisfy the condition that each of its eigenvalues is real. Is α necessarily selfadjoint?

Exercise 963 Let $V = \mathbb{R}^2$, endowed with the dot product. Is the endomorphism α of V defined by $\alpha: \begin{bmatrix} a \\ b \end{bmatrix} \mapsto \begin{bmatrix} a+b \\ a+b \end{bmatrix}$ positive definite?

Exercise 964 Let $\alpha \in \text{End}(\mathbb{R}^3)$ be represented with respect to the canonical basis by the matrix $\begin{bmatrix} 1 & 2 & 3 \\ 2 & a & 4 \\ 3 & 4 & 5 \end{bmatrix}$. For which values of a is α positive definite?

Exercise 965 For each complex number z , let α_z be the endomorphism of \mathbb{C}^3 represented with respect to the canonical basis by $\begin{bmatrix} 1 & 1 & -1 \\ 1 & 1 & z \\ -1 & \bar{z} & 1 \end{bmatrix}$.

Does there exist a z for which this endomorphism is positive definite?

Exercise 966 Let V be an inner product space and let $\alpha \in \text{End}(V)$ be positive definite. Is α^2 necessarily positive definite?

Exercise 967 Let α be a positive definite automorphism of an inner product space V . Is α^{-1} necessarily positive definite?

Exercise 968 Let k and n be positive integers. A symmetric matrix in $\mathcal{M}_{k+n, k+n}(\mathbb{R})$ is **quasidefinite** when it is of the form $\begin{bmatrix} -B & A^T \\ A & C \end{bmatrix}$ where B is a matrix representing a positive-definite endomorphism of \mathbb{R}^k with respect to the canonical basis, and C is a matrix representing a positive-definite endomorphism of \mathbb{R}^n with respect to the canonical basis. Show that a quasidefinite matrix is nonsingular, and that its inverse is again quasidefinite.

Exercise 969 Let $V = \mathbb{R}^2$ together with the dot product. Find positive-definite endomorphisms α and β of V satisfying the condition that their Jordan product is not positive definite.

Exercise 970 Let V be an inner product space over \mathbb{R} and let $\alpha \in \text{End}(V)$. Show that α is positive definite if and only if $\alpha + \alpha^*$ is positive definite.

Exercise 971 Let V be an inner product space finitely generated over \mathbb{C} and let α and β be positive-definite endomorphisms of V satisfying $\alpha\beta = \sigma_0$. Is it necessarily true that $\alpha = \sigma_0$ or $\beta = \sigma_0$?

Exercise 972 Let $V = \left\{ \begin{bmatrix} a \\ b \\ b \\ c \end{bmatrix} \mid a, b, c \in \mathbb{R} \right\}$ and let $W = \mathbb{R}^3$, both

of which together with the dot product, are inner product spaces of dimension 3 over \mathbb{R} . Find an isomorphism $\alpha : V \rightarrow W$ which is also an isometry.

Exercise 973 Let V be an inner product space and let α be an endomorphism of V which is an isometry. Does α also preserves angles between vectors?

Exercise 974 Let α be a positive-definite endomorphism of a finite-dimensional inner product space V represented with respect to some fixed basis by an $n \times n$ matrix $A = [a_{ij}]$. Show that $|A| \leq \prod_{i=1}^n a_{ii}$.

Exercise 975 Let n be a positive integer and let α be a positive-definite endomorphism of \mathbb{C}^n represented with respect to the canonical basis by the matrix $A = [a_{ij}] \in \mathcal{M}_{n \times n}(\mathbb{C})$. Show that a_{ii} is a positive real number for all $1 \leq i \leq n$.

Exercise 976 Let $\alpha : \mathbb{R}^3 \rightarrow \mathbb{R}^2$ be the linear transformation defined by

$$\alpha : \begin{bmatrix} a \\ b \\ c \end{bmatrix} \mapsto \begin{bmatrix} b - 2c \\ a + c \end{bmatrix}. \text{ Calculate } \text{spec}(\alpha\alpha^*) \text{ and } \text{spec}(\alpha^*\alpha).$$

Exercise 977 Let n be a positive integer and let $V = \mathbb{R}^n$, on which we have defined the dot product. Let α be a positive-definite endomorphism of V represented with respect to the canonical basis by the matrix $A = [a_{ij}] \in \mathcal{M}_{n \times n}(\mathbb{R})$. Show that $|A| > 0$. Is it necessarily true that $\text{tr}(A) > 0$?

Exercise 978 Let $V = \mathbb{C}^2$. Find endomorphisms $\alpha, \beta \in \text{End}(V)$ satisfying $\alpha > \beta$ (in the sense of Loewner) but not $\alpha^2 > \beta^2$.

Exercise 979 Let V be an inner product space and let $\alpha, \beta \in \text{End}(V)$ be positive definite. Is it necessarily true that $\alpha + \beta > \beta$?

Exercise 980 Let V and W be inner product spaces finitely generated over \mathbb{C} . Let $\alpha \in \text{Hom}(V, W)$, $\theta \in \text{Aut}(V)$, and $\varphi \in \text{Aut}(W)$, where the automorphisms θ and φ are positive definite. Show that the automorphism $\theta - \alpha^* \varphi \alpha$ of V is positive definite if and only if the automorphism $\varphi - \alpha \theta \alpha^*$ of W is positive definite.

Exercise 981 Let V be the vector space of all infinitely-differentiable functions in $\mathbb{R}^{\mathbb{R}}$, on which we define the inner product $\langle f, g \rangle = \int_0^\pi f(x)g(x)dx$. Let W be the subspace of all functions $f \in V$ satisfying $f(0) = f(\pi) = 0$. Show that the endomorphism of W defined by $f \mapsto f''$ is selfadjoint.

Exercise 982 Let V be a finitely-generated inner product space and let $\alpha \in \text{End}(V)$ be selfadjoint. Show that $\|\alpha(v)\| \leq \|v\|$ for all $v \in V$.

Exercise 983 Let V be an inner product space finitely generated over \mathbb{R} of dimension greater than 1, and let α be a selfadjoint endomorphism of V . Show that there are eigenvalues $c < d$ of α satisfying $c\|v\|^2 \leq \langle \alpha(v), v \rangle \leq d\|v\|^2$ for all $v \in V$.

Exercise 984 Let V be an inner product space finitely generated over \mathbb{R} and let $\alpha, \beta \in \text{End}(V)$ be selfadjoint. Assume that the eigenvalues of α all lie in the interval $[a, b]$ on the real line and that the eigenvalues of β all lie in the interval $[c, d]$ on the real line. Show that the eigenvalues of $\alpha + \beta$ all lie in the interval $[a + c, b + d]$ of the real line.

Exercise 985 Let V be an inner product space finitely generated over \mathbb{C} and let α be a positive-definite selfadjoint automorphism of V . Show that $\langle (\alpha + \alpha^{-1})(v), v \rangle \geq 2\langle v, v \rangle$ for all $v \in V$.

Exercise 986 Let V be an inner product space finitely generated over \mathbb{R} and let α be a positive-definite selfadjoint automorphism of V . Show that $\langle \alpha^{-1}(v), v \rangle = \max\{2\langle v, w \rangle - \langle \alpha(w), w \rangle \mid w \in W\}$ for all $v \in V$.

Exercise 987 Let V be a vector space finitely-generated over \mathbb{R} and let $\alpha \in \text{End}(V)$ be selfadjoint. Show that at least one of the values $\pm \|\alpha\|$ is an eigenvalue of α and any eigenvalue c of α satisfies $-\|\alpha\| \leq c \leq \|\alpha\|$.

Exercise 988 Let $A = \begin{bmatrix} a & b \\ \bar{b} & c \end{bmatrix} \in \mathcal{M}_{2 \times 2}(\mathbb{C})$ be Hermitian and let $r \geq s$ be the (necessarily real) eigenvalues of A . Show that $|b| \leq \frac{1}{2}(r - s)$.

18

Unitary and Normal endomorphisms

Let V be an inner product space. An automorphism of V which is an isometry is called a **unitary automorphism**. It is easy to see that if α and β are unitary automorphisms of V then $\alpha\beta$ and α^{-1} are also unitary automorphisms of V . It is also clear that σ_1 is unitary. Therefore, the set of all unitary automorphisms of V is a group of automorphisms.

(18.1) Proposition: Let V be an inner product space and let $\alpha \in \text{Aut}(V)$ have an adjoint. Then α is unitary if and only if $\alpha^* = \alpha^{-1}$.

Proof: If α is unitary then $\langle \alpha(v), w \rangle = \langle \alpha(v), \alpha\alpha^{-1}(w) \rangle = \langle v, \alpha^{-1}(w) \rangle$ for all $v, w \in V$ and so $\alpha^* = \alpha^{-1}$. Conversely, if $\alpha^* = \alpha^{-1}$ then $\langle \alpha(v), \alpha(w) \rangle = \langle v, \alpha^*\alpha(w) \rangle = \langle v, w \rangle$ for all $v, w \in V$ and so α is unitary. \square

As a direct consequence of Proposition 17.13, we see that if V is an inner product space finitely generated over its field of scalars then for $\alpha \in \text{End}(V)$ the following conditions are equivalent:

- (1) α is an isometry;
- (2) α is unitary;
- (3) α maps an orthonormal basis of V to an orthonormal basis of V .

If V is an inner product space finitely generated over its field of scalars F , and if α is a unitary automorphism of V represented by a matrix

$A = [a_{ij}] \in \mathcal{M}_{n \times n}(F)$ with respect to a given orthonormal basis. Then we see that $A^{-1} = A^H \in \mathcal{M}_{n \times n}(F)$. A matrix of this form over F is called a **unitary matrix**. If A is a unitary matrix then so is A^{-1} since $(A^{-1})^H = (A^H)^{-1}$. Also, if A and B are unitary matrices then $(AB)^{-1} = B^{-1}A^{-1} = B^H A^H = (AB)^H$ so AB is also unitary. The

converse is false. For example, the matrix $A = \begin{bmatrix} -1 & 1 \\ 0 & 1 \end{bmatrix} \in \mathcal{M}_{2 \times 2}(\mathbb{R})$ is not unitary, but $A^2 = I$ is.

Thus we see that the set of unitary matrices in $\mathcal{M}_{n \times n}(F)$ define a group of automorphisms of F^n and so an equivalence relation \sim defined by the condition that $A \sim B$ if and only if there exists a unitary matrix P such that $A = P^{-1}BP$. Matrices equivalent in this sense are **unitarily similar**. As an immediate consequence of the definition, we see that A is unitary if and only if the set of columns [resp. rows] of A is an orthonormal basis of F^n [resp. $\mathcal{M}_{1 \times n}(F)$] endowed with the dot product.

(18.2) Proposition: Let n be a positive integer and let $A = [a_{ij}]$ and $B = [b_{ij}]$ be unitarily-similar matrices in $\mathcal{M}_{n \times n}(\mathbb{C})$. Then $\sum_{i=1}^n \sum_{j=1}^n |a_{ij}|^2 = \sum_{i=1}^n \sum_{j=1}^n |b_{ij}|^2$.

Proof: We note that $\sum_{i=1}^n \sum_{j=1}^n |a_{ij}|^2 = \text{tr}(A^H A)$. If P is a unitary matrix satisfying $B = P^{-1}AP$ then $\text{tr}(B^H B) = \text{tr}(P^{-1}A^H P P^{-1}AP) = \text{tr}(P^{-1}A^H AP) = \text{tr}(A^H AP^{-1}P) = \text{tr}(A^H A)$, and we are done. \square

Example: If $c, d \in \mathbb{C}$ satisfy the condition that $|c|^2 + |d|^2 = 1$, then the matrix $\begin{bmatrix} c & d \\ -\bar{d} & \bar{c} \end{bmatrix} \in \mathcal{M}_{2 \times 2}(\mathbb{C})$ is unitary. A matrix of this form is

known as a **Givens¹ rotation matrix**. More generally, if $n > 3$ then a matrix $A = [a_{ij}] \in \mathcal{M}_{n \times n}(\mathbb{C})$ is a Givens rotation matrix if and only if there exist integers $1 \leq h < k \leq n$ and nonzero complex numbers c and



1

James Wallace Givens, a former assistant to Von Neumann, is considered to be one of the fathers of 20th-century American numerical analysis, who made major contributions to numerical matrix computation.

d satisfying $|c|^2 + |d|^2 = 1$ such that

$$a_{ij} = \begin{cases} c & \text{if } i = j \in \{h, k\} \\ 1 & \text{if } i = j \notin \{h, k\} \\ d & \text{if } i = h \text{ and } j = k \\ -\bar{d} & \text{if } i = k \text{ and } j = h \\ 0 & \text{otherwise} \end{cases}.$$

These matrices play important roles in numerical algorithms.

Example: The matrix $A = \frac{1}{2} \begin{bmatrix} 1-i & 1+i \\ 1+i & 1-i \end{bmatrix} \in \mathcal{M}_{2 \times 2}(\mathbb{C})$ is unitary. This matrix has important applications in the modeling of quantum computing, where it is often denoted by \sqrt{NOT} , since $A^2 = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$ represents the negation operator in this context.

Example: It is easy to show that $A_b = \begin{bmatrix} \sqrt{1+b^2} & bi \\ -bi & \sqrt{1+b^2} \end{bmatrix} \in \mathcal{M}_{2 \times 2}(\mathbb{C})$ satisfies $A_b A_b^T = I$ for any real number b , but, except for the case of $b = 0$, it is not unitary.

Unitarily-similar matrices are surely similar, but the converse is not true.

Example: The matrices $\begin{bmatrix} 3 & 1 \\ -2 & 0 \end{bmatrix}$ and $\begin{bmatrix} 1 & 1 \\ 0 & 2 \end{bmatrix}$ in $\mathcal{M}_{2 \times 2}(\mathbb{R})$ are similar but not unitarily similar, as we can see from Proposition 18.2.

(18.3) Proposition (Schur's Theorem²): If n is a positive integer, then every matrix in $\mathcal{M}_{n \times n}(\mathbb{C})$ is unitarily similar to an upper-triangular matrix.

Proof: We will proceed by induction on n . For $n = 1$, the result is trivial since every 1×1 matrix is upper triangular. Assume now that $n > 1$ and that the result has been established for $\mathcal{M}_{(n-1) \times (n-1)}(\mathbb{C})$. Let $A = [a_{ij}] \in \mathcal{M}_{n \times n}(\mathbb{C})$. Since we are working over \mathbb{C} , we know that the



² **Issai Schur** was a 20th-century German mathematician who is known primarily for his work in group theory.

characteristic polynomial of A is completely reducible, and so A has an eigenvalue, call it c_1 . Corresponding to that eigenvalue, we have a normal

eigenvector $v_1 = \begin{bmatrix} d_1 \\ \vdots \\ d_n \end{bmatrix}$ in which we can assume that $d_1 \in \mathbb{R}$. We now

are able to construct a basis $\{v_1, \dots, v_n\}$ for \mathbb{C}^n to which we can apply the Gram-Schmidt procedure, and thus assume that it is in fact an orthonormal basis (the vector v_1 doesn't change, since it was assumed to be normal to begin with). The matrix P_1 , the columns of which are these vectors, is therefore unitary. Now set $A_1 = P_1^{-1}AP_1$. It is easy to see that the first

column of A_1 is of the form $\begin{bmatrix} c_1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$ so we can write A_1 in block form as

$\begin{bmatrix} c_1 & x \\ O & A_2 \end{bmatrix}$, where $A_2 \in \mathcal{M}_{(n-1) \times (n-1)}(\mathbb{C})$. By the induction hypothesis, there is a unitary matrix $Q \in \mathcal{M}_{(n-1) \times (n-1)}(\mathbb{C})$ such that $Q^{-1}A_2Q$ is an upper-triangular matrix. Now set $P_2 = \begin{bmatrix} 1 & O \\ O & Q \end{bmatrix}$. Then P_2 is a unitary matrix in $\mathcal{M}_{n \times n}(\mathbb{C})$ and $P_2^{-1}P_1^{-1}AP_1P_2 = \begin{bmatrix} c_1 & y \\ O & Q^{-1}A_2Q \end{bmatrix}$ is an upper triangular matrix in $\mathcal{M}_{n \times n}(\mathbb{C})$. Since P_1P_2 is again unitary, we are done. \square

If we are working over \mathbb{R} , then a matrix A representing a unitary automorphism of \mathbb{R}^n satisfies $A^{-1} = A^T$. Such a matrix is called an **orthogonal matrix**. It is clear that the matrix I is orthogonal and that A^{-1} is orthogonal whenever A is orthogonal. If A and B are orthogonal matrices then $(AB)^{-1} = B^{-1}A^{-1} = B^TA^T = (AB)^T$ and so AB is also orthogonal. As an immediate consequence of the definition, we see that A is orthogonal if and only if the set of columns [resp. rows] of A is an orthonormal basis of \mathbb{R}^n [resp. $\mathcal{M}_{1 \times n}(\mathbb{R})$] endowed with the dot product. It is also clear that A is orthogonal if and only if A^T is orthogonal.

Example: Permutation matrices, which we considered earlier, are clearly orthogonal.

Example: The matrices $\begin{bmatrix} \cos(t) & \sin(t) \\ -\sin(t) & \cos(t) \end{bmatrix}$ and $\begin{bmatrix} \cos(t) & \sin(t) \\ \sin(t) & -\cos(t) \end{bmatrix}$ are orthogonal for every $t \in \mathbb{R}$, and one can show that these are the only orthogonal matrices in $\mathcal{M}_{2 \times 2}(\mathbb{R})$. Indeed, suppose that the matrix

$\begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \in \mathcal{M}_{2 \times 2}(\mathbb{R})$ is orthogonal. Then $a_{11}^2 + a_{12}^2 = 1 = a_{21}^2 + a_{22}^2$ so $-1 \leq a_{11} \leq 1$. Hence there exists a real number t such that $a_{11} = \cos(t)$. Then $a_{12}^2 = 1 - a_{11}^2 = 1 - \cos^2(t) = \sin^2(t)$ and so $a_{12} = \pm \sin(t)$. Also, $a_{11} = \cos(-t)$ and $\sin(-t) = -\sin(t)$. Thus, replacing t by $-t$ if necessary we can assume that $a_{11} = \cos(t)$ and $a_{12} = \sin(t)$. Similarly, there exists an angle s such that $a_{22} = \cos(s)$ and $a_{21} = \sin(s)$.

Since $0 = a_{11}a_{21} + a_{12}a_{22} = \cos(t)\sin(s) + \sin(t)\cos(s) = \sin(t+s)$, we see that $t+s=0$ or $t+s=\pi$. If $t+s=0$, we obtain

$$A = \begin{bmatrix} \cos(t) & \sin(t) \\ -\sin(t) & \cos(t) \end{bmatrix}.$$

If $t+s=\pi$, then $s=\pi-t$ and so $A = \begin{bmatrix} \cos(t) & \sin(t) \\ \sin(t) & -\cos(t) \end{bmatrix}$ since $\sin(t) = \sin(\pi-t)$ and $-\cos(t) = \cos(\pi-t)$.

One can also show that every orthogonal matrix in $\mathcal{M}_{3 \times 3}(\mathbb{R})$ is similar

to a matrix of the form $\begin{bmatrix} \cos(t) & -\sin(t) & 0 \\ \sin(t) & \cos(t) & 0 \\ 0 & 0 & \pm 1 \end{bmatrix}$ for some $t \in \mathbb{R}$.

Example: Let n be a positive integer. If $0 \leq c \leq 1$ is a real number, the matrix $\begin{bmatrix} (\sqrt{c})I & (-\sqrt{1-c})I \\ (\sqrt{1-c})I & (\sqrt{c})I \end{bmatrix} \in \mathcal{M}_{2n \times 2n}(\mathbb{R})$ is orthogonal, where I denotes the identity matrix in $\mathcal{M}_{n \times n}(\mathbb{R})$.

Example: Let n be a positive integer and let $V = \mathbb{R}^n$, on which we have defined the dot product. If $v \in V$ is a normal vector, then the matrix $A = I - 2(v \wedge v)$ is a **Householder**³ **matrix**. These matrices are clearly symmetric. Moreover, if $A = I - 2(v \wedge v)$ then

$$A^T A = A^2 = (I - 2[v \wedge v])^2 = I - 4v(v^T v)v^T + 4v(v^T v)v^T = I$$

and so A is orthogonal. Householder matrices have important uses in numerical analysis. We should also mention that if $u \neq v$ are vectors



3

Alston Householder, a 20th-century American mathematician, was among the pioneer researchers of the numerical analysis of matrices using computers, who developed many of the basic algorithms used in this field.

in V satisfying $\|u\| = \|v\|$, then the vector $w = \|v - u\|^{-1}(v - u)$ defines a Householder matrix $A = I - 2(w \wedge w)$ satisfying $Au = Av$. Since a Householder matrix is totally determined by one vector, it is easy to store in a computer. One of the important uses of Householder matrices is to compute QR-decompositions of matrices in a manner far more stable numerically than via the use of the Gram-Schmidt method.

The complex analog of Householder matrices are matrices of the form $I - 2ww^H$, where $w \in \mathbb{C}^n$. Such matrices are Hermitian and unitary and, too, have an important role in numerical computation.

Example: A general method for the construction of orthogonal matrices, due to contemporary American mathematician George W. Soules, is given as follows: Let $n > 1$ be an integer and let $w_1 \in \mathbb{R}^n$ be a normal vector all of the entries of which are all positive. Let $1 \leq k < n$

and write $w_1 = \begin{bmatrix} u \\ v \end{bmatrix}$, where $u \in \mathbb{R}^k$ and $v \in \mathbb{R}^{n-k}$. Set $a = \frac{\|v\|}{\|u\|}$

and $w_2 = \begin{bmatrix} au \\ -a^{-1}v \end{bmatrix}$. Then it is easy to see that w_2 is normal and

orthogonal to w_1 . Moreover, by further partitioning the vectors au and $-a^{-1}v$, we can eventually construct a mutually-orthogonal normal vectors w_1, w_2, \dots, w_n . The matrix with these vectors as columns is then orthogonal.

Notice that if F is either \mathbb{R} or \mathbb{C} , and if $A \in \mathcal{M}_{n \times n}(F)$ is a unitary matrix the columns of which are v_1, \dots, v_n , then the identity $AA^H = I$ implies that $\{v_1, \dots, v_n\}$ is an orthonormal set of vectors in F^n , on which we have the dot product, and hence it is a basis for this space. Conversely, if $\{v_1, \dots, v_n\}$ is an orthonormal basis of F^n then the matrix the columns of which are these vectors is unitary. Similarly, a matrix in $\mathcal{M}_{n \times n}(\mathbb{R})$ is orthogonal if and only if the set of its columns forms an orthonormal basis for \mathbb{R}^n with the dot product. Another way of putting this is that a matrix in $\mathcal{M}_{n \times n}(\mathbb{R})$ the columns of which are v_1, \dots, v_n is orthogonal if and only if $\sum_{i=1}^n v_i \wedge v_i = I$.

(18.4) Proposition: Let V be an inner product space of finite dimension n over \mathbb{R} . Let α be a unitary automorphism of V , which is represented by a matrix $A \in \mathcal{M}_{n \times n}(\mathbb{R})$ with respect to a given orthonormal basis of V . Then $|A| = \pm 1$.

Proof: We know that if α is represented by $A = [a_{ij}]$ with respect to the given basis, then α^* is represented by A^T . From Proposition 18.1, we deduce that $AA^T = I$ and so $|A|^2 = |A| \cdot |A^T| = |I| = 1$, and so $|A| = \pm 1$. \square

The orthogonal matrices in $\mathcal{M}_{n \times n}(\mathbb{R})$ having determinant equal to 1 are known as the **special orthogonal matrices**, and the set of all such matrices is denoted by $SO(n)$. This subset of $\mathcal{M}_{n \times n}(\mathbb{R})$ is clearly closed under taking products as well as taking inverses, since if $A \in SO(n)$ then $|A^{-1}| = |A^T| = |A| = 1$. If $A \in \mathcal{M}_{n \times n}(\mathbb{R})$ is a special orthogonal matrix, where n is an odd integer, then $1 \in \text{spec}(A)$. To see this, we note that

$$|A - 1I| = |A - I| = |A - AA^T| = |A| \cdot |I - A^T| = |I - A^T| = |I - A|$$

and, since n is odd, $|I - A| = (-1)^n |A - I|$. Thus we must have $|A - I| = 0$, and so $1 \in \text{spec}(A)$.

Example: We have already noted that the only orthogonal matrices in

$$\mathcal{M}_{2 \times 2}(\mathbb{R}) \text{ are of the form } \begin{bmatrix} \cos(t) & \sin(t) \\ -\sin(t) & \cos(t) \end{bmatrix} \text{ or } \begin{bmatrix} \cos(t) & \sin(t) \\ \sin(t) & -\cos(t) \end{bmatrix}$$

for some $t \in \mathbb{R}$. Matrices of the first type are special, whereas matrices of the second type are not.

Let V be an inner product space. An endomorphism $\alpha \in \text{End}(V)$ is **normal** if and only if α^* exists and satisfies $\alpha^* \alpha = \alpha \alpha^*$. From this definition, it is clear that α is normal if and only if α^* is normal. Clearly selfadjoint endomorphisms of V are normal, as are unitary automorphisms.

Example: If $a, b \in \mathbb{R}$ satisfy $b \neq 0$ and $a^2 + b^2 \neq 1$, then the automorphism α of \mathbb{R}^2 defined by $\begin{bmatrix} c \\ d \end{bmatrix} \mapsto \begin{bmatrix} ac + bd \\ ad - bc \end{bmatrix}$ is normal but neither unitary nor selfadjoint.

Example: If $0 \neq a, b \in \mathbb{R}$ then the automorphism α of \mathbb{C}^2 defined by $\begin{bmatrix} c \\ d \end{bmatrix} \mapsto \begin{bmatrix} c + id \\ c - id \end{bmatrix}$ is normal but neither unitary and nor selfadjoint.

(18.5) Proposition: Let V be an inner product space. An endomorphism $\alpha \in \text{End}(V)$ for which α^* exists is normal if and only if $\|\alpha(v)\| = \|\alpha^*(v)\|$ for all $v \in V$.

Proof: If α is normal and $v \in V$, then $\|\alpha(v)\|^2 = \langle \alpha(v), \alpha(v) \rangle = \langle v, \alpha^* \alpha(v) \rangle = \langle v, \alpha \alpha^*(v) \rangle = \langle v, \alpha^{**} \alpha^*(v) \rangle = \langle \alpha^*(v), \alpha^*(v) \rangle = \|\alpha^*(v)\|^2$ and so $\|\alpha(v)\| = \|\alpha^*(v)\|$. Conversely, assume that this condition holds. Then for each $v \in V$ we have

$$\begin{aligned} \langle (\alpha \alpha^* - \alpha^* \alpha)(v), v \rangle &= \langle \alpha \alpha^*(v), v \rangle - \langle \alpha^* \alpha(v), v \rangle \\ &= \langle \alpha^*(v), \alpha^*(v) \rangle - \langle \alpha(v), \alpha(v) \rangle = 0. \end{aligned}$$

But $\alpha\alpha^* - \alpha^*\alpha$ is selfadjoint and so, by Proposition 17.3 we see that $\alpha\alpha^* - \alpha^*\alpha = \sigma_0$, and so $\alpha\alpha^* = \alpha^*\alpha$. \square

We now take a short look at the extensive theory of eigenvalues of normal endomorphisms of inner product spaces. We will restrict our attention to finite-dimensional spaces, since the theory for infinite-dimensional spaces requires additional topological assumptions.⁴

(18.6) Proposition: Let V be an inner product space and let $\alpha \in \text{End}(V)$ be normal. Then every eigenvector of α is also an eigenvector of α^* and if c is an eigenvalue of α then \bar{c} is an eigenvalue of α^* .

Proof: If $v \in V$ then, as we have noted, $\|\alpha(v)\| = \|\alpha^*(v)\|$. For a scalar c , we see that

$$\begin{aligned} (\alpha - c\sigma_1)^*(\alpha - c\sigma_1) &= (\alpha^* - \bar{c}\sigma_1)(\alpha - c\sigma_1) = (\alpha - c\sigma_1)(\alpha^* - \bar{c}\sigma_1) \\ &= (\alpha - c\sigma_1)(\alpha - c\sigma_1)^* \end{aligned}$$

and so $\alpha - c\sigma_1$ is also normal. Thus, $\|(\alpha - c\sigma_1)(v)\| = \|(\alpha^* - \bar{c}\sigma_1)(v)\|$ for $v \in V$ and so, in particular, we see that $v \in \ker(\alpha - c\sigma_1)$ if and only if $v \in \ker(\alpha^* - \bar{c}\sigma_1)$. In other words, v is an eigenvector of α associated with the eigenvalue c if and only if it is an eigenvector of α^* associated with the eigenvalue \bar{c} . \square

Since $\alpha^{**} = \alpha$ for any endomorphism α of V , we see from Proposition 18.6 that if α is normal then a scalar c is an eigenvalue of α if and only if \bar{c} is an eigenvalue of α^* .

Another interesting consequence of Proposition 18.6 is the following: let V be a finitely-generated inner product space and let $\alpha \in \text{Aut}(V)$ be unitary. Then α is surely normal. If $c \in \text{spec}(\alpha)$ then $c \neq 0$ since α is an automorphism. If v is an eigenvector associated with c then $v = (\alpha^*\alpha)(v) = \alpha^*(cv) = c\alpha^*(v)$ and so $\alpha^*(v) = c^{-1}v$. This shows that c^{-1} is an eigenvalue of α^* and hence, by Proposition 18.6, $c^{-1} \in \text{spec}(\alpha)$.



⁴ The study of eigenvalues of normal and selfadjoint endomorphisms of inner product spaces was developed simultaneously by the American mathematician **Marshall Stone** and by John von Neumann, inspired by problems in quantum theory.

Example: In Proposition 17.5 we saw that if α is a selfadjoint endomorphism of an inner product space V finitely generated over \mathbb{R} , then $\text{spec}(\alpha) \neq \emptyset$. This is not necessarily true for normal endomorphisms of inner product spaces which are not selfadjoint. For example, let $V = \mathbb{R}^2$ together with the dot product, and if $\alpha \in \text{End}(V)$ is defined by

$$\alpha : \begin{bmatrix} a \\ b \end{bmatrix} \mapsto \begin{bmatrix} -b \\ a \end{bmatrix},$$

then we have already seen that $\text{spec}(\alpha) = \emptyset$. One can easily check that α is normal but not selfadjoint.

(18.7) Proposition: Let V be an inner product space finitely generated over \mathbb{C} and let $\alpha \in \text{End}(V)$. Then α is normal if and only if it is orthogonally diagonalizable.

Proof: Assume that α is normal. We will proceed by induction on $n = \dim(V)$. First assume that $n = 1$. Since we are working over \mathbb{C} , we know that $\text{spec}(\alpha) \neq \emptyset$ and so there exists a normal eigenvector v_1 of α . Then $V = \mathbb{C}v_1$ and we are done. Now assume that $n > 1$ and that the result has been proven for subspaces of dimension $n - 1$. Again, there exists a normal eigenvector v_1 of α . Set $W = \mathbb{C}v_1$. The subspace W of V is invariant under α , and so, by Proposition 18.6, it is also invariant under α^* . Therefore W^\perp is invariant under $\alpha^{**} = \alpha$. The restriction of α to W^\perp is a normal endomorphism, the adjoint of which is the restriction of α^* to W^\perp . By induction, we know that there exists an orthonormal basis $\{v_2, \dots, v_n\}$ composed of eigenvectors of α , and so $\{v_1, \dots, v_n\}$ is the basis of V we are seeking.

Conversely, assume that there exists an orthonormal basis $B = \{v_1, \dots, v_n\}$ composed of eigenvectors of α . Then $\Phi_{BB}(\alpha) = [a_{ij}]$ is a diagonal matrix satisfying the condition that each a_{ii} is an eigenvalue of α . Moreover, $\Phi_{BB}(\alpha^*) = \Phi_{BB}(\alpha)^H$ and this too is a diagonal matrix. Since diagonal matrices commute, we see that $\alpha\alpha^* = \alpha^*\alpha$ and so α is normal. \square

Note that Proposition 18.7 does not imply that if V is an inner product space finitely generated over \mathbb{C} and if $\alpha \in \text{End}(V)$ is normal, then every basis B of V composed of eigenvectors of α is necessarily orthonormal or that its elements are even necessarily mutually orthogonal, merely that one such basis exists.

Example: Let α be the endomorphism of \mathbb{C}^4 represented with respect

to the canonical basis by the matrix $A = \begin{bmatrix} 1 & 2 & 0 & 0 \\ -2 & 1 & 0 & 0 \\ 0 & 0 & 3 & -1 \\ 0 & 0 & 1 & 3 \end{bmatrix}$. One easily

checks that $AA^H = A^HA$ and so α is a normal automorphism of \mathbb{C}^4 . The characteristic polynomial of A is

$$X^4 - 8X^3 + 27X^2 - 50X + 50 = (X^2 - 6X + 10)(X^2 - 2X + 5)$$

and so $\text{spec}(\alpha) = \{3 \pm i, 1 \pm 2i\}$. The set

$$\left\{ \frac{1}{\sqrt{2}} \begin{bmatrix} -i \\ 1 \\ 0 \\ 0 \end{bmatrix}, \frac{1}{\sqrt{2}} \begin{bmatrix} i \\ 1 \\ 0 \\ 0 \end{bmatrix}, \frac{1}{\sqrt{2}} \begin{bmatrix} 0 \\ 0 \\ 1 \\ -i \end{bmatrix}, \frac{1}{\sqrt{2}} \begin{bmatrix} 0 \\ 0 \\ 1 \\ i \end{bmatrix} \right\}$$

is an orthonormal basis for \mathbb{C}^4 composed of eigenvectors of α .

(18.8) Proposition: Let V be a finitely-generated inner product space. Then the following conditions on a projection $\alpha \in \text{End}(V)$ are equivalent:

- (1) α is normal;
- (2) α is selfadjoint;
- (3) $\ker(\alpha) = \text{im}(\alpha)^\perp$.

Proof: (1) \Rightarrow (2): From (1) we know that $\|\alpha(v)\| = \|\alpha^*(v)\|$ for all $v \in V$. In particular, $\alpha(v) = 0_V$ if and only if $\alpha^*(v) = 0_V$ so that $\ker(\alpha) = \ker(\alpha^*)$. If $v \in V$ and $w = v - \alpha(v)$ then $\alpha(w) = \alpha(v) - \alpha^2(v) = \alpha(v) - \alpha(v) = 0_V$ and so $\alpha^*(w) = 0_V$. Therefore $\alpha^*(v) = \alpha^*\alpha(v)$ for all $v \in V$, whence $\alpha^* = \alpha^*\alpha$. This implies that $\alpha = \alpha^{**} = (\alpha^*\alpha)^* = \alpha^*\alpha^{**} = \alpha^*\alpha = \alpha^*$, which proves (2).

(2) \Rightarrow (3): If $v, w \in V$ then, from (2), we see that $\langle \alpha(v), w \rangle = \langle v, \alpha(w) \rangle$. In particular, if $v \in \ker(\alpha)$ then $\langle v, \alpha(w) \rangle = 0$ for all $w \in V$, which is to say that $v \in \text{im}(\alpha)^\perp$. Conversely, if $v \in \text{im}(\alpha)^\perp$ then $\langle v, \alpha(w) \rangle = 0$ for all $w \in V$, which implies that $\alpha(v)$ is orthogonal to every element of V . Therefore $\alpha(v) = 0_V$ and so $v \in \ker(\alpha)$. This proves (3).

(3) \Rightarrow (1): Let $v, w \in V$. Since α is a projection, we have $v - \alpha(v) \in \ker(\alpha)$. It is also clear that $\alpha(w) \in \text{im}(\alpha)$. Therefore

$$0 = \langle v - \alpha(v), \alpha(w) \rangle = \langle v, \alpha(w) \rangle - \langle \alpha(v), \alpha(w) \rangle = \langle v, \alpha(w) \rangle - \langle v, \alpha^*\alpha(w) \rangle$$

and since this is true for all $v, w \in V$, we have $\alpha = \alpha^*\alpha$. This implies that $\alpha = \alpha^*$ is selfadjoint and therefore surely normal, proving (1). \square

We note that if V is a finitely-generated inner product space and if $\alpha \in \text{End}(V)$ is normal, then, by Propositions 16.5, 16.7. and 18.5, we have $V = \ker(\alpha) \oplus \text{im}(\alpha)$, and, in particular, $\{\text{im}(\alpha), \ker(\alpha)\}$ is an independent set of subspaces of V . Moreover, $v \perp v'$ for all $v \in \ker(\alpha)$ and $v' \in \text{im}(\alpha)$. While the direct-sum decomposition is valid for any projection, it is the normality which ensures the orthogonality.

(18.9) Proposition: Let V be a finitely-generated inner product space. Let W_1, \dots, W_n be subspaces of V and, for each $1 \leq i \leq n$, let α_i be the projection of V onto the subspace W_i coming from the decomposition $V = W_i \oplus W_i^\perp$. Then the following conditions are equivalent:

- (1) $V = \bigoplus_{i=1}^n W_i$ and $W_h^\perp = \bigoplus_{j \neq h} W_j$ for all $1 \leq h \leq n$;
- (2) $\alpha_1 + \dots + \alpha_n = \sigma_1$ and $\alpha_i \alpha_j = \sigma_0$ for all $i \neq j$;
- (3) If B_i is an orthonormal basis of W_i for each i , then $B = \bigcup_{i=1}^n B_i$ is an orthonormal basis of V .

Proof: This has essentially already been established when we talked about the decomposition of a vector space into a direct sum of subspaces. \square

(18.10) Proposition: Let F be either \mathbb{R} or \mathbb{C} and let V be a finitely-generated inner product space over F . If $p(X) \in F[X]$ and if α is a normal endomorphism of V , then $p(\alpha)$ is a normal endomorphism of V .

Proof: If $p(X) = \sum_{i=0}^n a_i X^i$. Then $p(\alpha) = \sum_{i=0}^n a_i \alpha^i$ and $p(\alpha)^* = \sum_{i=0}^n \bar{a}_i (\alpha^*)^i$. Since $\alpha \alpha^* = \alpha^* \alpha$, it follows from the definition of the product that $p(\alpha) p(\alpha)^* = p(\alpha)^* p(\alpha)$. Therefore $p(\alpha)$ is a normal endomorphism of V . \square

(18.11) Proposition: Let V be a finitely-generated inner product space and let α be a normal endomorphism of V . If the minimal polynomial of α is completely reducible, then it does not have multiple roots.

Proof: Let $p(X)$ be the minimal polynomial of α , which we assume is completely reducible. Assume that there exists a scalar c and a polynomial $q(X)$ such that $p(X) = (X - c)^2 q(X)$. Since $p(\alpha) = \sigma_0$, we have $(\alpha - c\sigma_1)^2 q(\alpha) = \sigma_0$ and so $\ker((\alpha - c\sigma_1)^2 q(\alpha)) = V$. By Proposition 18.10, we know that $\beta = \alpha - c\sigma_1$ is a normal endomorphism of V . Let $v \in V$ and let $w = q(\alpha)(v)$. Then $\beta^2(w) = 0_V$ and so $\beta(w) \in \text{im}(\beta) \cap \ker(\beta) = \{0_V\}$. Thus we see that $\beta q(\alpha)(v) = 0_V$ for all $v \in V$ and hence α annihilates the polynomial $(X - c)q(X)$, contradicting the minimality of $p(X)$. \square

(18.12) Proposition (Spectral Decomposition Theorem): Let V be an inner product space finitely generated over \mathbb{C} and let α be a normal endomorphism of V . Then there exist scalars c_1, \dots, c_n and projections $\alpha_1, \dots, \alpha_n$ of V satisfying:

- (1) $\alpha = c_1 \alpha_1 + \dots + c_n \alpha_n$;

- (2) $\sigma_1 = \alpha_1 + \dots + \alpha_n$;
 (3) $\alpha_h \alpha_j = \sigma_0$ for all $h \neq j$.

Moreover, these c_j and α_j are unique. The c_j are precisely the distinct eigenvalues of α and each α_j is the projection of V onto the eigenspace W_j associated with c_j coming from the decomposition $V = W_j \oplus W_j^\perp$.

Proof: Let $p(X)$ be the minimal polynomial of α , which we will write in the form $p(X) = \prod_{i=1}^n (X - c_i)$, where the c_i are complex numbers which, by Proposition 18.11, are distinct. For each $1 \leq j \leq n$, let $p_j(X)$ be the j th Lagrange interpolation polynomial determined by the c_i .

Let $f(X)$ be a polynomial of degree at most $n - 1$. Then the polynomial $f(X) - \sum_{i=1}^n f(c_i)p_i(X)$ is of degree at most $n - 1$ and has n distinct roots c_1, \dots, c_n . Thus it must be the 0-polynomial and so $f(X) = \sum_{i=1}^n f(c_i)p_i(X)$. In particular, we see that $1 = \sum_{i=1}^n p_i(X)$ and $X = \sum_{i=1}^n c_i p_i(X)$. Set $\alpha_j = p_j(\alpha)$. Then $\sigma_1 = \sum_{i=1}^n \alpha_i$ and $\alpha = \sum_{i=1}^n c_i \alpha_i$. Note that $\alpha_j \neq \sigma_0$ since $\alpha_j = p_j(\alpha)$ and the degree of $p_j(X)$ is less than the degree of the minimal polynomial of α . Moreover, if $h \neq j$ then there exists a polynomial $u(X) \in \mathbb{C}[X]$ satisfying $\alpha_h \alpha_j = u(\alpha)p(\alpha) = u(\alpha)\sigma_0 = \sigma_0$. Thus we see that for all $1 \leq j \leq n$ we have $\alpha_j = \alpha_j \sigma_1 = \sum_{i=1}^n \alpha_j \alpha_i = \alpha_j^2$ and so each α_j is a projection. Thus we see that $\{\text{im}(\alpha_j) \mid 1 \leq j \leq n\}$ is an independent set of subspaces of V .

Since the minimal polynomial and the characteristic polynomial of α have the same roots, we know that $\text{spec}(\alpha) = \{c_1, \dots, c_n\}$. To show that $W_h = \text{im}(\alpha_h)$, we have to prove that a vector v belongs to $\text{im}(\alpha_h)$ if and only if $\alpha(v) = c_h v$. Indeed, if $\alpha(v) = c_h v$ then $c_h \left[\sum_{j=1}^n \alpha_j(v) \right] = c_h v = \alpha(v) = \sum_{j=1}^n (c_j \alpha_j)(v)$ and so $\sum_{j=1}^n [(c_h - c_j) \alpha_j](v) = 0_V$. Thus, for all $j \neq h$, we have $\alpha_j(v) = 0_V$ and so $v = \alpha_h(v) \in \text{im}(\alpha_h)$.

Finally, we note that α_h is the projection coming from the decomposition $V = W_j \oplus W_j^\perp$ since α_h is a polynomial in α and hence normal and so the result follows from the remark after Proposition 18.8. \square

Note that we could have deduced Proposition 18.12 directly from Proposition 18.7. What is important in the above proof is the explicit construction of the projection maps as polynomials in α .

If $\alpha = \sum_{i=1}^n c_i \alpha_i$ is as in Proposition 18.12, then $\alpha^k = (\sum_{i=1}^n c_i \alpha_i)^k = \sum_{i=1}^n c_i^k \alpha_i$ for any positive integer k , and from this we see that if $p(X) \in \mathbb{C}[X]$ then $p(\alpha) = \sum_{i=1}^n p(c_i) \alpha_i$.

(18.13) Proposition: Let V be an inner product space finitely generated over \mathbb{C} . A normal endomorphism α of V is positive definite if and only if each of its eigenvalues is positive.

Proof: If α is positive definite then, by Proposition 17.11, each of its eigenvalues is positive. Conversely, assume each of the eigenvalues of α is positive. By Proposition 18.12, we write $\alpha = \sum_{i=1}^n c_i \alpha_i$, where the c_i are the eigenvalues of α and the α_i are projections in $\text{End}(V)$ satisfying $\alpha_i \alpha_j = \sigma_0$ for $i \neq j$. If $0_V \neq v \in V$ then $\langle \alpha(v), v \rangle = \sum_{i=1}^n \sum_{j=1}^n c_i \langle \alpha_i(v), \alpha_j(v) \rangle = \sum_{i=1}^n c_i \|\alpha_i(v)\|^2 > 0$ and so α is positive definite. \square

Example: Let $V = \mathbb{R}^3$. For each $a \in \mathbb{R}$, let $\alpha_a \in \text{End}(V)$ be the normal endomorphism of V represented with respect to the canonical

basis by the matrix $\begin{bmatrix} 1 & a & a \\ a & 1 & a \\ a & a & 1 \end{bmatrix}$. Then $\text{spec}(\alpha) = \{2a+1, 1-a\}$ and so,

by Proposition 18.13, α is positive definite precisely when $-1 < 2a < 2$.

As a consequence of Proposition 18.13 and the comments before it, we see that if α is a positive-definite endomorphism of a finitely-generated inner product space V over \mathbb{C} then there exists an endomorphism $\sqrt{\alpha}$ of V satisfying $(\sqrt{\alpha})^2 = \alpha$. This endomorphism is defined by $\sqrt{\alpha} = \sum_{i=1}^n (\sqrt{c_i}) \alpha_i$, where the c_i are the eigenvalues of α , and where the α_i are defined as in Proposition 18.12. In particular, if β is an automorphism of V then, by Proposition 17.10, we can talk about $\sqrt{\beta^* \beta}$, which is also positive definite by Proposition 18.13.

(18.14) Proposition: Let V be an inner product space finitely generated over \mathbb{C} and let $\alpha \in \text{Aut}(V)$. Then there exists a unique positive-definite automorphism θ of V and a unique unitary automorphism ψ of V satisfying $\alpha = \psi\theta$.

Proof: By Proposition 17.10, we know that the automorphism $\alpha^* \alpha$ of V is positive definite and so we can set $\theta = \sqrt{\alpha^* \alpha}$. Let $\varphi = \theta \alpha^{-1}$. Then $\varphi^* = (\alpha^{-1})^* \theta^* = (\alpha^*)^{-1} \theta$ so $\varphi^* \varphi = (\alpha^*)^{-1} \theta \theta \alpha^{-1} = (\alpha^*)^{-1} \alpha^* \alpha \alpha^{-1} = \sigma_1$, proving that φ is unitary by Proposition 18.1, and hence belongs to $\text{Aut}(V)$. If we now define $\psi = \varphi^{-1}$, we see that $\alpha = \psi\theta$. Moreover, we note that $\theta \in \text{Aut}(V)$ since $\theta = \varphi \alpha$. To prove uniqueness, assume that $\psi\theta = \psi'\theta'$, where ψ and ψ' are unitary automorphisms of V and where θ and θ' are positive-definite automorphisms of V . Then $\psi^2 = \psi\theta^* \theta \psi = \psi'(\theta')^* \theta' \psi' = (\psi')^2$. Since ψ is positive definite, this implies that $\psi = \psi'$ and so, since ψ is an automorphism, we have $\theta = \psi^{-1} \psi\theta = \psi^{-1} \psi\theta' = \theta'$. \square

The representation of an automorphism α of an inner product space finitely generated over \mathbb{C} in the form given in Proposition 18.14 is sometimes

called the **polar decomposition** of α . If we move over to matrices, we see that the **polar decomposition** of a nonsingular matrix $A \in \mathcal{M}_{n \times n}(\mathbb{C})$ is of the form $A = UM$, where U is a unitary matrix and M is a Hermitian matrix satisfying the condition that $\bar{v}Mv > 0$ for all nonzero $v \in \mathbb{C}^n$. Similarly, there exists a unitary matrix U' and a Hermitian matrix M' satisfying $A^H = U'M'$ and so $A = M'(U')^H$, where $(U')^H$ is again unitary.

In the case we are working over \mathbb{R} , the matrix U is orthogonal, and M is symmetric and satisfies $v^T M v > 0$ for all nonzero $v \in \mathbb{R}^n$. Because polar decompositions are important in applications, several iterative algorithms exist to compute them.

(18.15) Proposition (Singular Value Decomposition Theorem⁵): Let V and W be inner product spaces of finite dimensions k and n respectively, and let $\alpha \in \text{Hom}(V, W)$. Then there exists an integer $t \leq \min\{k, n\}$, together with positive real numbers $c_1 \geq c_2 \geq \dots \geq c_t$ and with orthonormal bases $\{v_1, \dots, v_k\}$ of V and $\{w_1, \dots, w_n\}$ of W satisfying

$$\alpha(v_i) = \begin{cases} c_i w_i & \text{if } 1 \leq i \leq t \\ 0_W & \text{otherwise} \end{cases}$$

and

$$\alpha^*(w_i) = \begin{cases} c_i v_i & \text{if } 1 \leq i \leq t \\ 0_V & \text{otherwise} \end{cases}.$$

Proof: If α is the 0-map, then the result is immediate, so assume that that is not the case. We note that $\beta = \alpha^* \alpha$ is a selfadjoint endomorphism of V and so, by Proposition 17.7, it is orthogonally diagonalizable. Hence V has an orthonormal basis $\{v_1, \dots, v_k\}$ composed of eigenvectors of β , where each v_i is associated with an eigenvalue b_i . By Proposition 17.10, we know that each b_i belongs to \mathbb{R} . Moreover, for each i we note that $b_i = b_i \langle v_i, v_i \rangle = \langle b_i v_i, v_i \rangle = \langle \beta(v_i), v_i \rangle = \langle \alpha(v_i), \alpha(v_i) \rangle \geq 0$. Indeed, renumbering if necessary, we can assume that there exists an integer $t \leq k$ such that $b_1 \geq b_2 \geq \dots \geq b_t > 0$ while $b_{t+1} = \dots = b_k = 0$. For each $1 \leq i \leq t$, set $c_i = \sqrt{b_i}$ and let $w_i = c_i^{-1} \alpha(v_i) \in W$. If



5

The first version of the Singular Value Decomposition Theorem was proven by the 19th-century Italian mathematician **Eugenio Beltrami**. It was subsequently extended by many others, including Camille Jordan and Sylvester.

$i \neq j$ then $\langle w_i, w_j \rangle = (c_i c_j)^{-1} \langle \alpha(v_i), \alpha(v_j) \rangle = (c_i c_j)^{-1} \langle \beta(v_i), v_j \rangle = (c_i c_j)^{-1} b_i \langle v_i, v_j \rangle = 0$ while, for each $1 \leq i \leq t$, we have $\langle w_i, w_i \rangle = c_i^{-2} \langle \alpha(v_i), \alpha(v_i) \rangle = c_i^{-2} \langle \beta(v_i), v_i \rangle = c_i^{-2} \langle b_i v_i, v_i \rangle = \langle v_i, v_i \rangle = 1$. Thus we see that the set $\{w_1, \dots, w_t\}$ is orthonormal. Moreover, for each $1 \leq i \leq t$ we have $\|\alpha(v_i)\|^2 = b_i$ so $\|\alpha(v_i)\| = c_i$ and $\alpha^*(w_i) = \alpha^*(c_i^{-1} \alpha(v_i)) = c_i^{-1} \alpha^* \alpha(v_i) = c_i^{-1} \beta(v_i) = c_i^{-1} b_i v_i = c_i v_i$. For $t+1 \leq i \leq k$ we have $\alpha^* \alpha(v_i) = \beta(v_i) = 0_V$ and so $0 = \langle \beta(v_i), v_i \rangle = \langle \alpha(v_i), \alpha(v_i) \rangle$, which implies that $\alpha(v_i) = 0_W$. Thus $v_i \in \ker(\alpha)$ for each $t+1 \leq i \leq k$.

We are therefore left with the matter of defining w_{t+1}, \dots, w_n in the case $t < n$. By Proposition 16.17, we know that $\ker(\alpha^*) = \text{im}(\alpha)^\perp$ and so, if we pick an orthonormal basis $\{w_{t+1}, \dots, w_n\}$ for $\ker(\alpha^*)$ we see that $\{w_1, \dots, w_n\}$ is an orthonormal basis for W having the desired properties. \square

The scalars $c_1 \geq c_2 \geq \dots \geq c_t$ given in the Proposition 18.15 are called the **singular values** of the linear transformation α . The number c_1/c_t , called the **spectral condition number**, is used as a measure of the numerical instability of the matrix representing $\alpha^* \alpha \in \text{End}(V)$ with respect to the given basis.

If we consider the special case of a linear transformation $\alpha : \mathbb{C}^k \rightarrow \mathbb{C}^n$ represented with respect to the canonical bases by a matrix $A \in \mathcal{M}_{n \times k}(\mathbb{C})$, the Singular Value Decomposition Theorem says that there exist unitary matrices $P \in \mathcal{M}_{n \times n}(\mathbb{C})$ and $Q \in \mathcal{M}_{k \times k}(\mathbb{C})$ such that A can be written as $P \begin{bmatrix} D & O \\ O & O \end{bmatrix} Q^H$, where $D \in \mathcal{M}_{t \times t}(\mathbb{R})$ is a diagonal matrix having the singular values of α on the diagonal. These singular values are precisely the square roots of the eigenvalues of $A^H A$. The columns of Q form an orthonormal basis for \mathbb{C}^k consisting of eigenvectors of $A^H A$, and the columns of P form an orthonormal basis for \mathbb{C}^n .

If $\alpha : \mathbb{R}^k \rightarrow \mathbb{R}^n$ then of course the matrices P and Q are orthogonal and $A = P \begin{bmatrix} D & O \\ O & O \end{bmatrix} Q^T$.

Example: The matrix $A = \frac{1}{10} \begin{bmatrix} 20 & 20 & -20 & 20 \\ 1 & 17 & 1 & -17 \\ 18 & 6 & 18 & -6 \end{bmatrix}$ can be written as a product $P \begin{bmatrix} 4 & 0 & 0 & 0 \\ 0 & 3 & 0 & 0 \\ 0 & 0 & 2 & 0 \end{bmatrix} Q$, where $P = \frac{1}{5} \begin{bmatrix} 5 & 0 & 0 \\ 0 & 3 & -4 \\ 0 & 4 & 3 \end{bmatrix}$

and $Q = \frac{1}{2} \begin{bmatrix} 1 & 1 & -1 & 1 \\ 1 & 1 & 1 & -1 \\ 1 & -1 & 1 & 1 \\ 1 & -1 & -1 & -1 \end{bmatrix}$ are orthogonal and where the singular

values of A are 4, 3, 2.

Singular value decompositions have many applications, and play important roles in the mathematics of optimization, of data compression, and of image processing. They are especially useful since accurate and relatively-efficient algorithms for computing these decompositions are readily available in many common linear-algebra software packages.

Exercises

Exercise 989 Let $A \in \mathcal{M}_{n \times n}(\mathbb{C})$ be similar to a unitary matrix. Is A^{-1} necessarily similar to A^H ?

Exercise 990 Let $n > 1$ be an integer and let V be the subspace of $\mathbb{C}[X]$ consisting of all polynomials of degree at most n . Let $0 \neq c \in \mathbb{C}$ and let α be the endomorphism of V defined by $\alpha : p(X) \mapsto p(X + c)$. Is it possible to define an inner product on V relative to which α is normal?

Exercise 991 Let $a, b, c \in \mathbb{C}$. Find the set of all triples (x, y, z) of

complex numbers satisfying the condition that the matrix
$$\begin{bmatrix} a & x & y \\ 0 & b & z \\ 0 & 0 & c \end{bmatrix}$$

represents a normal endomorphism of \mathbb{C}^3 , endowed with the dot product, with respect to the canonical basis.

Exercise 992 Let n be a positive integer. A matrix $A \in \mathcal{M}_{n \times n}(\mathbb{C})$ is **normal** if and only if $A^H A = A A^H$. Show that every normal upper-triangular matrix is a diagonal matrix.

Exercise 993 Let $V = \mathbb{R}^2$, together with the dot product. Show that a matrix in $\mathcal{M}_{2 \times 2}(\mathbb{R})$ is of the form $\Phi_{BB}(\alpha)$ for some normal endomorphism α of V which is not selfadjoint if and only if it is of the form

$$\begin{bmatrix} a & -b \\ b & a \end{bmatrix} \quad \text{for real numbers } a \text{ and } b \neq 0.$$

Exercise 994 Let V be an inner product space finitely generated over \mathbb{R} and let S be the set of all isometries V . Is S an \mathbb{R} -subalgebra of $\text{End}(V)$?

Exercise 995 Let n be a positive integer and let $V = \mathbb{C}^n$ on which we have the dot product. If $\alpha \in \text{End}(V)$, let $G(\alpha) = \{\langle \alpha(v), v \rangle \mid \|v\| = 1\}$. For the special case $n = 2$, find $G(\alpha)$ and $G(\beta)$, where α is represented

with respect to the canonical basis by the matrix $\begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$ and β is represented with respect to the canonical basis by the matrix $\begin{bmatrix} 0 & 2 \\ 0 & 0 \end{bmatrix}$.

Exercise 996 Let $V = \mathbb{R}^3$ on which we have the dot product, and let W be the space of all polynomial functions in $\mathbb{R}^{\mathbb{R}}$ of degree at most 2, on which we define the inner product $\langle f, g \rangle = \int_0^1 f(x)g(x)dx$. Let

$$\alpha \in \text{Hom}(V, W) \text{ be defined by } \alpha \left(\begin{bmatrix} a \\ b \\ c \end{bmatrix} \right) : x \mapsto 1 + \frac{b}{2} + \frac{c}{6} + (b-c)x + cx^2.$$

Is this linear transformation an isometry?

Exercise 997 Let V be an inner product space and let α be an endomorphism of V satisfying the condition that $\alpha^* \alpha = \sigma_0$. Show that $\alpha = \sigma_0$.

Exercise 998 Let $V = \mathbb{R}^3$ with the dot product, and let α be the automorphism of V defined by $\alpha : \begin{bmatrix} a \\ b \\ c \end{bmatrix} \mapsto \begin{bmatrix} -c \\ -b \\ -a \end{bmatrix}$. Is α unitary?

Exercise 999 Is the matrix $\begin{bmatrix} 0 & 0 & 0 & i \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ i & 0 & 0 & 0 \end{bmatrix} \in \mathcal{M}_{4 \times 4}(\mathbb{C})$ unitary?

Exercise 1000 Find a real number a satisfying the condition that the

matrix $a \begin{bmatrix} -9 + 8i & -10 - 4i & -16 - 18i \\ -2 - 24i & 1 + 12i & -10 - 4i \\ 4 - 10i & -2 - 24i & -9 + 8i \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{C})$ is unitary.

Exercise 1001 Find a real number a satisfying the condition that the

matrix $\frac{1}{24} \begin{bmatrix} 12 & 6 - 12i & 12 + 6i & 6 - 6i \\ 6 + 12i & a & 5i & 3 + i \\ 12 - 6i & -5i & a & 1 - 3i \\ 6 + 6i & 3 - i & 1 + 3i & -22 \end{bmatrix} \in \mathcal{M}_{4 \times 4}(\mathbb{C})$ is unitary.

Exercise 1002 Given a real number a , check if the matrix

$$\begin{bmatrix} -\sin^2(a) + i \cos^2(a) & (1 + i) \sin(a) \cos(a) \\ (1 + i) \sin(a) \cos(a) & -\cos^2(a) + i \sin^2(a) \end{bmatrix} \in \mathcal{M}_{2 \times 2}(\mathbb{C})$$

is unitary.

Exercise 1003 Find all possible triples (a, b, c) of real numbers, if any

exist, such that the matrix $\frac{1}{3} \begin{bmatrix} 1 & -2 & 2 \\ 2 & -1 & 2 \\ a & b & c \end{bmatrix}$ is orthogonal.

Exercise 1004 Is the matrix $\begin{bmatrix} 0.5 & 0.5 & 0.5 & 0.5 \\ 0.5 & 0.5 & -0.5 & -0.5 \\ 0.5 & -0.5 & 0.5 & -0.5 \\ 0.5 & -0.5 & -0.5 & 0.5 \end{bmatrix} \in \mathcal{M}_{4 \times 4}(\mathbb{R})$ orthogonal?

Exercise 1005 If $v = \begin{bmatrix} c \\ d \end{bmatrix} \in \mathbb{R}^2$, show that there exists an orthogonal matrix $A \in \mathcal{M}_{2 \times 2}(\mathbb{R})$ and a real number b satisfying the condition that $Av = \begin{bmatrix} b \\ 0 \end{bmatrix}$.

Exercise 1006 Let a and b be real numbers, not both equal to 0.

Show that the matrix $\frac{1}{a^2 + ab + b^2} \begin{bmatrix} ab & a(a+b) & b(a+b) \\ a(a+b) & -b(a+b) & ab \\ b(a+b) & ab & -a(a+b) \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{R})$ is orthogonal.

Exercise 1007 Find all $a \in \mathbb{R}$ such that the matrix $\begin{bmatrix} 2a & -2a & a \\ -2a & -a & 2a \\ a & 2a & 2a \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{R})$ is orthogonal.

Exercise 1008 Let $A = \begin{bmatrix} 4 & -1 & 1 \\ -1 & 4 & -1 \\ 1 & -1 & 4 \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{R})$. Find an orthogonal matrix P such that $P^T A P$ is a diagonal matrix.

Exercise 1009 $A = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \in \mathcal{M}_{4 \times 4}(\mathbb{R})$. Find an orthogonal matrix P such that $P^T A P$ is a diagonal matrix.

Exercise 1010 Let n be a positive integer and let A and B be orthogonal matrices in $\mathcal{M}_{n \times n}(\mathbb{R})$ satisfying $|A| + |B| = 0$. Show that $|A + B| = 0$.

Exercise 1011 Let $A = \frac{1}{2} \begin{bmatrix} 1 & -1 & 2\sqrt{2} \\ 2\sqrt{2} & 2\sqrt{2} & 0 \\ -1 & 1 & 2\sqrt{2} \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{R})$. Find an

infinite number of pairs (P, Q) of orthogonal matrices such that PAQ is a diagonal matrix.

Exercise 1012 Find an $a \in \mathbb{R}$ such that the matrix $\begin{bmatrix} a & -\frac{4}{5} & 0 \\ \frac{4}{5} & a & 0 \\ 0 & 0 & 1 \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{R})$ is orthogonal.

Exercise 1013 Let $A, B \in \mathcal{M}_{k \times n}(\mathbb{R})$ be matrices such that the columns of each form orthonormal bases for the same subspace W of \mathbb{R}^k . Show that $AA^T = BB^T$.

Exercise 1014 Let $A, B \in \mathcal{M}_{n \times n}(\mathbb{R})$ be orthogonal matrices. Is the matrix $\begin{bmatrix} A & O \\ O & B \end{bmatrix} \in \mathcal{M}_{2n \times 2n}(\mathbb{R})$ necessarily orthogonal?

Exercise 1015 Let n be a positive integer and let $A \in \mathcal{M}_{n \times n}(\mathbb{R})$ be a skew-symmetric matrix. Show that $(A - I)^{-1}(A + I)$ is an orthogonal matrix which does not have 1 as an eigenvalue.

Exercise 1016 Find two distinct functions $f_1, f_2 : \mathbb{R}^4 \setminus \left\{ \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \end{bmatrix} \right\} \rightarrow \mathbb{R}$ satisfying the condition that

$$f_i \left(\begin{bmatrix} a \\ b \\ c \\ d \end{bmatrix} \right) \begin{bmatrix} a^2 + b^2 - c^2 - d^2 & 2(bc - da) & 2(bd - ca) \\ 2(bcda) & a^2 + b^2 - c^2 - d^2 & 2(cd - ba) \\ 2(bd - ca) & 2(cd - ba) & a^2 + b^2 - c^2 - d^2 \end{bmatrix}$$

is always an orthogonal matrix.

Exercise 1017 Let $O \neq A \in \mathcal{M}_{3 \times 3}(\mathbb{C})$ be a matrix satisfying $\text{adj}(A) = A^H$. Show that A is a unitary matrix having determinant 1.

Exercise 1018 Let n be a positive integer and let α be the endomorphism of \mathbb{C}^n defined by $\alpha : v \mapsto iv$. Is α normal?

Exercise 1019 Let V be an inner product space and let $\alpha, \beta \in \text{End}(V)$ be normal. Is $\beta\alpha$ necessarily normal?

Exercise 1020 Let V be an inner product space finitely-generated over \mathbb{C} and let $\alpha \in \text{End}(V)$ satisfy the condition that every eigenvector of $\beta = \alpha + \alpha^*$ is also an eigenvector of $\gamma = \alpha - \alpha^*$. Prove that α is normal.

Exercise 1021 Let α be the endomorphism of \mathbb{C}^2 represented with respect to the canonical basis by the matrix $A = \begin{bmatrix} 2 & i \\ i & 2 \end{bmatrix}$. Is α normal?

Exercise 1022 Let V be an inner product space over \mathbb{C} and let $\alpha \in \text{End}(V)$ be normal. If $c \in \mathbb{C}$, is the endomorphism $\alpha - c\sigma_1$ necessarily normal?

Exercise 1023 Let V be an inner product space finitely generated over \mathbb{C} and let $\sigma_0 \neq \alpha \in \text{End}(V)$ be normal. Show that α is not nilpotent.

Exercise 1024 Let $a, b \in \mathbb{R}$ let $\alpha \in \text{End}(\mathbb{R}^3)$ be represented with respect to the canonical basis by $\begin{bmatrix} a & -2 & b \\ b & a & -2 \\ -2 & 3 & a \end{bmatrix}$. For which values of a and b is this endomorphism normal?

Exercise 1025 Let $\alpha \in \text{End}(\mathbb{R}^3)$ be represented with respect to the canonical basis by the matrix $\frac{1}{3} \begin{bmatrix} 14 & 2 & 14 \\ 2 & -1 & -16 \\ 14 & -16 & 5 \end{bmatrix}$. Show that α is selfadjoint and find an orthonormal basis of \mathbb{R}^3 composed of eigenvectors of α .

Exercise 1026 Let α be the endomorphism of \mathbb{C}^3 represented with respect to the canonical basis by the matrix $\begin{bmatrix} 6 & -2 & 3 \\ 3 & 6 & -2 \\ -2 & 3 & 6 \end{bmatrix}$. Show that α is normal and find an orthonormal basis of \mathbb{C}^3 composed of eigenvectors of α .

Exercise 1027 Let V be an inner product space and let $\sigma_0 \neq \alpha \in \text{End}(V)$ be a normal projection. Show that $\|\alpha(v)\| \leq \|v\|$ for all $v \in V$, with equality whenever $v \in \text{im}(\alpha)$. Give an example where this does not hold for α which is not normal.

Exercise 1028 Let $\alpha \in \text{End}(\mathbb{R}^4)$ be defined by $\alpha : \begin{bmatrix} a_1 \\ a_2 \\ a_3 \\ a_4 \end{bmatrix} \mapsto \begin{bmatrix} a_1 \\ a_2 \\ a_3 + a_4 \\ a_4 - a_3 \end{bmatrix}$.

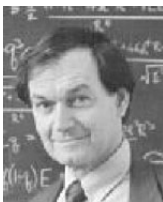
Show that α is normal but not selfadjoint.

19

Moore-Penrose pseudoinverses

Let V and W be inner product spaces, and let $\alpha : V \rightarrow W$ be a linear transformation. We know that there exists a linear transformation $\beta : W \rightarrow V$ satisfying the condition that $\beta\alpha$ is the identity function on V and $\alpha\beta$ is the identity function on W if and only if α is an isomorphism; in this case, $\beta = \alpha^{-1}$. If both spaces are finitely generated, we also know that such an isomorphism can exist only when $\dim(V) = \dim(W)$. If α is not an isomorphism, it is possible to weaken the notion of the the inverse of a function. Given a linear transformation $\alpha : V \rightarrow W$, we say that a linear transformation $\beta : W \rightarrow V$ is a **Moore-Penrose¹ pseudoinverse** of α if and only if the following conditions are satisfied:

- (1) $\alpha\beta\alpha = \alpha$ and $\beta\alpha\beta = \beta$;
- (2) The endomorphisms $\beta\alpha \in \text{End}(V)$ and $\alpha\beta \in \text{End}(W)$ are selfadjoint.



¹ Eliakim Hastings Moore developed this construction in 1922, but it did not receive much attention at the time; it was rediscovered independently in 1955 by **Sir Roger Penrose**, a contemporary British applied mathematician, best known for his collaboration with the physicist Stephen Hawking.

Example: The two parts of condition (2) in the definition of the Moore-Penrose pseudoinverse are independent. To see this, consider the linear

transformation $\alpha : \mathbb{R}^3 \rightarrow \mathbb{R}^2$ defined by $\alpha : v \mapsto \begin{bmatrix} 1 & 2 & 3 \\ 0 & 1 & 0 \end{bmatrix} v$. For

any $c, d \in \mathbb{R}$, let $\beta : \mathbb{R}^2 \rightarrow \mathbb{R}^3$ be the linear transformation defined

by $\beta : w \mapsto \begin{bmatrix} 1-3c & -2-3d \\ 0 & 1 \\ c & d \end{bmatrix} w$. Then one can check that $\alpha\beta\alpha = \alpha$

and $\beta\alpha\beta = \beta$ and that $\alpha\beta = \sigma_1$ in $\text{End}(\mathbb{R}^2)$. On the other hand,

$\beta\alpha : v \mapsto \begin{bmatrix} 1-3c & -6c-3d & 3-9c \\ 0 & 1 & 0 \\ c & 2c+d & 3c \end{bmatrix} v$, and it is easy enough to choose

c and d so that this matrix is not symmetric and hence $\beta\alpha$ is not selfadjoint.

We will denote the Moore-Penrose pseudoinverse of α by α^+ . Of course, in order to justify this notation we have to show that β exists and is unique, which we will do for the case that V and W are finitely generated. We will begin with uniqueness.

(19.1) Proposition: Let V and W be inner product spaces and let $\alpha : V \rightarrow W$ be a linear transformation. If α has a Moore-Penrose pseudoinverse, it must be unique.

Proof: Suppose that $\beta, \gamma \in \text{Hom}(W, V)$ are Moore-Penrose pseudoinverses of α . Then $\beta = \beta\alpha\beta = (\beta\alpha)^*\beta = \alpha^*\beta^*\beta = (\alpha\gamma\alpha)^*\beta^*\beta = (\gamma\alpha)^*\alpha^*\beta^*\beta = \gamma\alpha\alpha^*\beta^*\beta = \gamma\alpha(\beta\alpha)^*\beta = \gamma\alpha\beta\alpha\beta = \gamma\alpha\beta = \gamma\alpha\gamma\alpha\beta = \gamma(\alpha\gamma)^*\alpha\beta = \gamma\gamma^*\alpha^*\alpha\beta = \gamma\gamma^*\alpha^*(\alpha\beta)^* = \gamma\gamma^*(\alpha\beta\alpha)^* = \gamma\gamma^*\alpha^* = \gamma(\alpha\gamma)^* = \gamma\alpha\gamma = \gamma$ and so we have proven uniqueness. \square

In particular, if $\alpha : V \rightarrow W$ is an isomorphism, then, by Proposition 19.1, we have $\alpha^+ = \alpha^{-1}$. If α is the 0-function then so is α^+ .

(19.2) Proposition: Let V and W be finitely-generated inner product spaces and let $\alpha : V \rightarrow W$ be a linear transformation.

(1) If α is a monomorphism, then α^+ exists and equals $(\alpha^*\alpha)^{-1}\alpha^*$. Moreover, $\alpha^+\alpha$ is the identity function on V ;

(2) If α is an epimorphism, then α^+ exists and equals $\alpha^*(\alpha\alpha^*)^{-1}$. Moreover, $\alpha\alpha^+$ is the identity function on W .

Proof: (1) From Proposition 16.19, we see that if α is a monomorphism then $\alpha^*\alpha \in \text{Aut}(V)$ and so $(\alpha^*\alpha)^{-1}$ exists. Set $\beta = (\alpha^*\alpha)^{-1}\alpha^*$. Then

$\beta\alpha$ is the identity function on V and so $\beta\alpha$ is a selfadjoint endomorphism of V which satisfies $\alpha\beta\alpha = \alpha$ and $\beta\alpha\beta = \beta$. Finally, $(\alpha\beta)^* = [\alpha(\alpha^*\alpha)^{-1}\alpha^*]^* = \alpha[(\alpha^*\alpha)^{-1}]^* \alpha^* = \alpha[(\alpha^*\alpha)^*]^{-1} \alpha^* = \alpha(\alpha^*\alpha)^{-1} \alpha^* = \alpha\beta$ and so $\alpha\beta$ is also selfadjoint. Thus $\beta = \alpha^+$.

(2) From Proposition 16.19, we see that if α is an epimorphism then $\alpha\alpha^* \in \text{Aut}(W)$ and so $(\alpha\alpha^*)^{-1}$ exists. As in (1), we see that $\alpha^*(\alpha\alpha^*)^{-1} = \alpha^+$. \square

Example: Let $\alpha : \mathbb{R}^2 \rightarrow \mathbb{R}^3$ be the linear transformation represented

with respect to the canonical bases by the matrix $\begin{bmatrix} 1 & 2 \\ -1 & 3 \\ 2 & 4 \end{bmatrix}$. This is a

monomorphism and so, by Proposition 19.2, α^+ exists and is represented

with respect to the canonical bases by the matrix $\frac{1}{25} \begin{bmatrix} 3 & -10 & 6 \\ 1 & 5 & 2 \end{bmatrix}$.

(19.3) Proposition: Let V and W be finitely-generated inner product spaces. Then every $\alpha \in \text{Hom}(V, W)$ has a Moore-Penrose pseudoinverse $\alpha^+ \in \text{Hom}(W, V)$.

Proof: Let $Y = \text{im}(\alpha)$, and write $\alpha = \mu\beta$, where $\beta : V \rightarrow Y$ is an epimorphism given by $\beta : v \mapsto \alpha(v)$, and $\mu : Y \rightarrow W$ is the inclusion monomorphism. By Proposition 19.2, we know that $\beta^+ \in \text{Hom}(Y, V)$ and $\mu^+ \in \text{Hom}(W, Y)$ exist and satisfy the conditions that $\beta\beta^+$ and $\mu^+\mu$ are equal to the identity function on Y . Therefore we see that $(\mu\beta)(\beta^+\mu^+)(\mu\beta) = \mu\beta$ and $(\beta^+\mu^+)(\mu\beta)(\beta^+\mu^+) = \beta^+\mu^+$ and we see that $(\beta^+\mu^+)(\mu\beta) = \beta^+\beta$ and $(\mu\beta)(\beta^+\mu^+) = \mu\mu^+$ are selfadjoint. Thus α^+ exists and equals $\beta^+\mu^+$. \square

As an immediate consequence of this, we note that if α is an endomorphism of a finitely-generated inner product space V then, by Proposition 6.11, we see that $\text{rk}(\alpha\alpha^+) \leq \text{rk}(\alpha)$ and $\text{rk}(\alpha) = \text{rk}(\alpha\alpha^+\alpha) \leq \text{rk}(\alpha\alpha^+)$ and so $\text{rk}(\alpha\alpha^+) = \text{rk}(\alpha)$. Similarly, $\text{rk}(\alpha^+\alpha) = \text{rk}(\alpha)$.

If F is either \mathbb{R} or \mathbb{C} , and if we are given a linear transformation $\alpha : F^k \rightarrow F^n$ which is represented with respect to the canonical bases by the matrix $A = [a_{ij}]$, then we will denote the matrix representing α^+ with respect to these bases by A^+ . Thus the matrix A^+ has the following properties:

- (1) $AA^+A = A$ and $A^+AA^+ = A^+$;
- (2) The matrices AA^+ and A^+A are symmetric (in the case $F = \mathbb{R}$) or Hermitian (in the case $F = \mathbb{C}$).

Example: Let $A = \begin{bmatrix} 1 & -1 & 2 \\ 2 & 1 & -2 \\ 3 & 0 & 0 \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{R})$. This matrix is clearly singular and hence A^{-1} does not exist. However, we can check that $A^+ = \frac{1}{45} \begin{bmatrix} 5 & 5 & 10 \\ -5 & 4 & -1 \\ 10 & -8 & 2 \end{bmatrix}$.

For nonsingular square matrices of the same size A and B , we know that $(AB)^{-1} = B^{-1}A^{-1}$. A similar equality does not hold for the Moore-Penrose pseudoinverse, as the following example shows.

Example: If $A = \begin{bmatrix} 2 & 6 \\ 1 & 3 \end{bmatrix}$ and $B = \begin{bmatrix} 1 & 2 \\ 2 & 4 \end{bmatrix}$ in $\mathcal{M}_{2 \times 2}(\mathbb{R})$, then $AB = \begin{bmatrix} 14 & 28 \\ 7 & 14 \end{bmatrix}$. Then $A^+ = \frac{1}{50} \begin{bmatrix} 2 & 1 \\ 6 & 3 \end{bmatrix}$ and $B^+ = \frac{1}{25} \begin{bmatrix} 1 & 2 \\ 2 & 4 \end{bmatrix}$ so $B^+A^+ = \frac{7}{1250} \begin{bmatrix} 2 & 1 \\ 4 & 2 \end{bmatrix}$, while $(AB)^+ = \frac{1}{175} \begin{bmatrix} 2 & 1 \\ 4 & 2 \end{bmatrix}$.

Example: If $A(t) = \begin{bmatrix} 1 & 0 \\ 0 & t \end{bmatrix}$ for all real numbers t then we see that $A(t)^+ = \begin{bmatrix} 1 & 0 \\ 0 & t^{-1} \end{bmatrix}$ when $t \neq 0$, but is equal to $\begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$ for $t = 0$. Thus we see that not only is $\lim_{t \rightarrow 0} A(t)$ not equal to $A(0)$, but indeed that the value of $A(t)$ moves farther and farther away from $A(0)$ as t approaches 0.

Thus we see that the Moore-Penrose pseudoinverse is not computationally stable. This means that one has to be very careful in actual applications. Because of the importance and utility of Moore-Penrose pseudoinverses, there exists a considerable literature on techniques for computing A^+ or A^+A , given a matrix A . One of the methods used in practice for computing the Moore-Penrose pseudoinverse over \mathbb{R} is a recursive one, known as **Greville's method**², which is based on the following result: if



² In the 1970's, American mathematician **Thomas N. E. Greville** and American/Israeli mathematician **Adi Ben-Israel** popularized and reinvented the use of the Moore-Penrose pseudoinverse as a computational tool.

$A \in \mathcal{M}_{k \times n}(\mathbb{R})$, and if we write $A = \begin{bmatrix} B & v \end{bmatrix}$, where $B \in \mathcal{M}_{k \times (n-1)}(\mathbb{R})$, then $A^+ = \begin{bmatrix} B^+(I - v \wedge w) \\ w \end{bmatrix}$, where

$$w = \begin{cases} (\|(I - BB^+)v\|)^{-2} (I - BB^+)v & \text{if } \|(I - BB^+)v\| \neq 0 \\ (1 + \|B^+v\|^2)^{-1} (B^+)^T B^+v & \text{otherwise} \end{cases}.$$

Another technique is to break A up into blocks, if possible. Indeed, if $A = \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix}$, where A_{11} is a nonsingular square matrix the rank of which equals the rank of A , then, by **Zlobec's formula**, we have $A^+ = \begin{bmatrix} A_{11} & A_{12} \end{bmatrix}^* B^* \begin{bmatrix} A_{11} \\ A_{21} \end{bmatrix}^*$, where $B = \left(\begin{bmatrix} A_{11} & A_{12} \end{bmatrix} A^* \begin{bmatrix} A_{11} \\ A_{21} \end{bmatrix} \right)^{-1}$.

One can also use convergence methods to compute the Moore-Penrose pseudoinverse of a matrix. If $A \in \mathcal{M}_{k \times n}(\mathbb{C})$ then, by Proposition 17.4, we know that the eigenvalues of A^*A are real. Let c be the largest such eigenvalue and pick a real number b satisfying $0 < bc < 2$. For each integer $p \geq 2$, define the sequence Y_0, Y_1, \dots of matrices in $\mathcal{M}_{n \times k}(\mathbb{C})$ as follows:

- (1) $Y_0 = bA^*$;
- (2) If $k \geq 0$ and Y_k has already been defined, set $T_k = I - Y_k A$ and set $Y_{k+1} = Y_k + \sum_{i=1}^{p-1} T_k^i Y_k$. Then the sequence Y_0, Y_1, \dots converges to A^+ .

Another method is the following: if $A \in \mathcal{M}_{n \times n}(\mathbb{R})$ is an arbitrary symmetric matrix we can define matrices A_0, A_1, \dots by setting $A_0 = A$ and $A_{k+1} = [I + (I - A_k)(I + A_k)^{-1}] A_k$ for all $k \geq 0$. Also, we can define real numbers c_0, c_1, \dots by setting $c_0 = 1$ and $c_{i+1} = 2c_i + 1$ for each $i \geq 0$. Then the **Kovarik algorithm** states that if none of the numbers $-c_i^{-1}$ is an eigenvalue of A , the sequence A_0, A_1, \dots converges to A^+A .

Let F be either \mathbb{R} or \mathbb{C} , let k and n be positive integers, and let $A \in \mathcal{M}_{k \times n}(F)$. We now look at what the matrix A^+ says about a solution (if any) to a system of linear equations of the form $AX = w$, where $w \in F^k$. First of all, we note that in general

(19.4) Proposition: Let F be either \mathbb{R} or \mathbb{C} , let k and n be positive integers, let $A \in \mathcal{M}_{k \times n}(F)$, and let $w \in F^k$. The system of linear equations $AX = w$ has a solution if and only if $(AA^+)w = w$.

Proof: If there is a vector $v \in F^n$ satisfying $Av = w$ then $(AA^+)w = (AA^+)(Av) = (AA^+A)v = Av = w$. Conversely, if $(AA^+)w = w$ then $A(A^+w) = w$ and so $Av = w$, where $v = A^+w$. \square

We also note that, in the situation above, if $y \in F^n$ then $A(I - A^+A)y = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$ and so we also see that $A^+w + (I - A^+A)y$ is also a solution to

the system $AX = w$, assuming that the system has any solutions at all. Conversely, any solution to this system is of the form $A^+w + u$, where

$$Au = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}, \text{ and so } (I - A^+A)u = u.$$

(19.5) Proposition: Let F be either \mathbb{R} or \mathbb{C} , let k and n be positive integers, and let $A \in M_{k \times n}(F)$. Let $w \in F^k$. If the system $AX = w$ has a solution then in the set of all solutions to this system of linear equations there is precisely one having a minimal norm, and it is A^+w .

Proof: If u is a solution to this system, then we have already seen that it is of the form $A^+w + (I - A^+A)y$. But we note that

$$\begin{aligned} \langle A^+w, (I - A^+A)y \rangle &= \langle A^+AA^+w, (I - A^+A)y \rangle \\ &= \langle A^+w, (A^+A)(I - A^+A)y \rangle \\ &= \langle A^+w, (A^+A - A^+AA^+A)y \rangle \\ &= \left\langle A^+w, \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix} \right\rangle = 0 \end{aligned}$$

so

$$\begin{aligned} \|u\|^2 &= \langle u, u \rangle = \langle A^+w + (I - A^+A)y, A^+w + (I - A^+A)y \rangle \\ &= \langle A^+w, A^+w \rangle + \langle (I - A^+A)y, (I - A^+A)y \rangle \\ &= \|A^+w\|^2 + \|(I - A^+A)y\|^2, \end{aligned}$$

which implies that $\|u\| \geq \|A^+w\|$. \square

Example: Let $A = \begin{bmatrix} 2 & -1 & 1 \\ 1 & 1 & 1 \end{bmatrix}$ and $w = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$. Then $A^+ = \frac{1}{14} \begin{bmatrix} 4 & 2 \\ -5 & 8 \\ 1 & 4 \end{bmatrix}$ and so the solution to the system $AX = w$ having minimal norm is $A^+w = \frac{1}{14} \begin{bmatrix} 8 \\ 11 \\ 9 \end{bmatrix}$. Its norm is $\frac{1}{14}\sqrt{266}$.

But what happens if the system $AX = w$ doesn't have a solution? Suppose that F is either \mathbb{R} or \mathbb{C} and that $A \in \mathcal{M}_{k \times n}(F)$, and $w \in F^k$, where k and n be positive integers. Then the system $(A^+A)X = A^+w$ always has a solution, namely A^+w , and by Proposition 19.5 this is in fact the solution of minimal norm of this equation, which is the **best approximation** to a solution of $AX = w$.

In order to emphasize the use of Proposition 19.5, we briefly consider the **least squares method**, which is an important tool in many areas of applied mathematics and statistics. This method was developed at the beginning of the nineteenth century by Gauss and Legendre and, independently, by the American mathematical pioneer Robert Adrain³. Suppose that we have before us the results of several observations, which, depending on values t_1, \dots, t_n of a real parameter, give us real values c_1, \dots, c_n . Our theory tells us that the set of points $\{(t_i, c_i) \mid 1 \leq i \leq n\}$ in the euclidean plane should lie on a straight line. However, because of measuring and/or computational errors, this doesn't quite work out. So we want to find the equation of the line in the plane which best fits our observed data. In other words, we want to find a solution of minimal norm to the system of linear equations $\{X_1 + t_i X_2 = c_i \mid 1 \leq i \leq n\}$, which can be

written as $\begin{bmatrix} 1 & t_1 \\ 1 & t_2 \\ \vdots & \vdots \\ 1 & t_n \end{bmatrix} \begin{bmatrix} X_1 \\ X_2 \end{bmatrix} = \begin{bmatrix} c_1 \\ c_2 \\ \vdots \\ c_n \end{bmatrix}$. As we have seen, the solution



³ Irish-born **Robert Adrain** emigrated to the United States in 1798. He published his own mathematics journal but his work received no international attention at the time.

of minimal norm, if it exists, is
$$\begin{bmatrix} 1 & t_1 \\ 1 & t_2 \\ \vdots & \vdots \\ 1 & t_n \end{bmatrix}^+ \begin{bmatrix} c_1 \\ c_2 \\ \vdots \\ c_n \end{bmatrix}.$$
 Otherwise, this is the best approximation to the solution the system.

Example: To find the equation of the line in the euclidean plane which best fits the set of points $\{(1, 3), (2, 7), (3, 8), (4, 11)\}$, we calculate

$$\begin{bmatrix} 1 & 1 \\ 1 & 2 \\ 1 & 3 \\ 1 & 4 \end{bmatrix}^+ \begin{bmatrix} 3 \\ 7 \\ 8 \\ 11 \end{bmatrix} = \left(\frac{1}{20} \begin{bmatrix} 20 & 10 & 0 & -10 \\ -6 & -2 & 2 & 6 \end{bmatrix} \right) \begin{bmatrix} 3 \\ 7 \\ 8 \\ 11 \end{bmatrix} = \frac{1}{2} \begin{bmatrix} 2 \\ 5 \end{bmatrix}$$

so the line we want is given by $\{(t, 1 + \frac{5}{2}t) \mid t \in \mathbb{R}\}$.

We can use the same method to find the best fit of any polynomial of a higher degree to a set of points. For example, if we wish to find a parabola which best fits the set of points $\{(t_i, c_i) \mid 1 \leq i \leq n\}$ in the euclidean plane, we have to find a best approximation to a solution of the system of linear equations $\{X_1 + t_i X_2 + t_i^2 = c_i \mid 1 \leq i \leq n\}$, which we know is

$$\begin{bmatrix} 1 & t_1 & t_1^2 \\ 1 & t_2 & t_2^2 \\ \vdots & \vdots & \vdots \\ 1 & t_n & t_n^2 \end{bmatrix}^+ \begin{bmatrix} c_1 \\ c_2 \\ \vdots \\ c_n \end{bmatrix}.$$

Example: To find the equation of the parabola in the euclidean plane which best fits the set of points $\{(1, 3), (2, 7), (3, 8), (4, 11)\}$, we calculate

$$\begin{aligned} \begin{bmatrix} 1 & 1 & 1 \\ 1 & 2 & 4 \\ 1 & 3 & 9 \\ 1 & 4 & 16 \end{bmatrix}^+ \begin{bmatrix} 3 \\ 7 \\ 8 \\ 11 \end{bmatrix} &= \left(\frac{1}{20} \begin{bmatrix} 45 & -15 & -25 & 15 \\ -31 & 23 & 27 & -19 \\ 5 & -5 & -5 & 5 \end{bmatrix} \right) \begin{bmatrix} 3 \\ 7 \\ 8 \\ 11 \end{bmatrix} \\ &= \frac{1}{4} \begin{bmatrix} -1 \\ 15 \\ -1 \end{bmatrix}. \end{aligned}$$

and so the parabola we want is given by $\{(t, -\frac{1}{4} + \frac{15}{4}t - \frac{1}{4}t^2) \mid t \in \mathbb{R}\}$.

Needless to say, we can also consider a much more general context. Suppose that W is a finitely-generated subspace of \mathbb{R}^A . Given a set of observations $\{(t_i, c_i) \mid 1 \leq i \leq n\} \subseteq A \times \mathbb{R}$, we want to find the function $g \in W$ which best approximates these observations.

To do this, we pick a basis $\{f_1, \dots, f_k\}$ for W . Then we want to find a best approximation to a solution of the system of linear equations

$$\{X_1 f_1(t_i) + \dots + X_k f_k(t_i) = c_i \mid 1 \leq i \leq n\},$$

which can be written as
$$\begin{bmatrix} f_1(t_1) & \dots & f_k(t_1) \\ \vdots & & \vdots \\ f_1(t_n) & \dots & f_k(t_n) \end{bmatrix} \begin{bmatrix} X_1 \\ \vdots \\ X_k \end{bmatrix} = \begin{bmatrix} c_1 \\ \vdots \\ c_n \end{bmatrix}.$$
 As

we have seen, this is
$$\begin{bmatrix} f_1(t_1) & \dots & f_k(t_1) \\ \vdots & & \vdots \\ f_1(t_n) & \dots & f_k(t_n) \end{bmatrix}^+ \begin{bmatrix} c_1 \\ \vdots \\ c_n \end{bmatrix}.$$

Least-squares approximations are often used to find best-fit solutions to very large systems of linear equations of the form $AX = w$ which, in theory, have an exact solution but in practice that solution cannot be found because of errors in measurement of the data and computational errors. Indeed, Gauss developed this method for finding solutions to the very large systems of linear equations which resulted from laying down a triangulation grid for a geodetic survey of the state of Hanover he conducted in 1818. In 1978, the American National Geodetic Survey used it to solve a system of over 2.5 million linear equations in 400,000 unknowns which resulted from the updating of the triangulation grid for the continental United States.

Exercises

Exercise 1029 Let $A = \begin{bmatrix} 1 & 1 \\ -1 & 1 \\ 2 & 3 \end{bmatrix} \in \mathcal{M}_{3 \times 2}(\mathbb{R})$. Calculate A^+ .

Exercise 1030 Let $A = \begin{bmatrix} 5 & 0 & 0 \end{bmatrix} \in \mathcal{M}_{1 \times 2}(\mathbb{Q})$. Calculate A^+ .

Exercise 1031 Let $A = \begin{bmatrix} a_1 & \dots & a_n \end{bmatrix} \in \mathcal{M}_{1 \times n}(\mathbb{C})$, where n is a positive integer. Show that $A^+ = (AA^H)^{-1}A^H$.

Exercise 1032 Let $A = \begin{bmatrix} 2 & 2 & 0 \\ 1 & 2 & 1 \\ 1 & 2 & 1 \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{R})$. Calculate A^+ .

Exercise 1033 Let V and W be finitely-generated inner product spaces, and let $\alpha \in \text{Hom}(V, W)$. For any nonzero scalar c , show that $(c\alpha)^+ = \frac{1}{c}\alpha^+$.

Exercise 1034 Let n be a positive integer and let $A \in \mathcal{M}_{n \times n}(\mathbb{R})$ be a diagonal matrix. Calculate A^+ .

Exercise 1035 Let V and W be finitely-generated inner product spaces, and let $\alpha \in \text{Hom}(V, W)$. Show that $(\alpha^*)^+ = (\alpha^+)^*$.

Exercise 1036 Let $V = \mathbb{R}^2$, which is endowed with the dot product and let $\alpha : V \rightarrow \mathbb{R}$ be the linear functional defined by $\alpha : \begin{bmatrix} a \\ b \end{bmatrix} \mapsto a$. Let

$\beta : \mathbb{R} \rightarrow V$ be the linear transformation defined by $\beta : a \mapsto \begin{bmatrix} a \\ a \end{bmatrix}$. Show that $(\alpha\beta)^+ \neq \beta^+\alpha^+$.

Exercise 1037 Let n be a positive integer and let $A = [a_{ij}] \in \mathcal{M}_{n \times n}(\mathbb{R})$ be the matrix all entries of which are equal to 1. Show that $A^+ = n^{-2}A$.

Exercise 1038 Let $A \in \mathcal{M}_{k \times n}(\mathbb{R})$ be a matrix of the form $\begin{bmatrix} C & O \\ O & O \end{bmatrix}$, where C is a $t \times t$ nonsingular diagonal matrix. Show that $A^+ = \begin{bmatrix} D & O \\ O & O \end{bmatrix}$, where $D = C^{-1}$.

Exercise 1039 Let $V = \mathbb{R}^n$ on which we have the dot product defined, and let $\alpha \in \text{End}(V)$ satisfy the condition that $\ker(\alpha) = \text{im}(\alpha)^\perp$. Show that the restriction β of α to $\text{im}(\alpha)$ is an automorphism of $\text{im}(\alpha)$ and that the restriction of α^+ to $\text{im}(\alpha)$ equals β^{-1} .

Exercise 1040 Let n be a positive integer and let $A, B \in \mathcal{M}_{n \times n}(\mathbb{R})$ be matrices satisfying the conditions $ABA = A$, $BAB = B$, and $A^2 = A$. Is it necessarily true that $B^2 = B$?

Exercise 1041 Let h, k, m , and n be positive integers, let $A \in \mathcal{M}_{h \times k}(\mathbb{R})$, let $B \in \mathcal{M}_{m \times n}(\mathbb{R})$, and let $C \in \mathcal{M}_{h \times n}(\mathbb{R})$. Show that there exists a matrix $X \in \mathcal{M}_{k \times m}(\mathbb{R})$ satisfying $AXB = C$ if and only if $AA^+CB^+B = C$.

Exercise 1042 Let k and n be positive integers and let $B \in \mathcal{M}_{k \times k}(\mathbb{R})$ and $C \in \mathcal{M}_{n \times n}(\mathbb{R})$ be orthogonal matrices. For $A \in \mathcal{M}_{k \times n}(\mathbb{R})$, show that $(BAC)^+ = C^T A^+ B^T$.

Exercise 1043 Let k and n be positive integers and let $A \in \mathcal{M}_{k \times k}(\mathbb{R})$ and $B \in \mathcal{M}_{k \times n}(\mathbb{R})$. Let $C \in \mathcal{M}_{n \times n}(\mathbb{R})$ be nonsingular. Prove that

$$\begin{bmatrix} A & AB \\ O & C \end{bmatrix}^+ = \begin{bmatrix} A^+ & -A^+ABC^{-1} \\ O & C^{-1} \end{bmatrix}.$$

20

Bilinear transformations and forms

Let V , W , and Y be vector spaces over a field F . We say that a function $f : V \times W \rightarrow Y$ is a **bilinear transformation** if and only if the function $v \mapsto f(v, w_0)$ belongs to $\text{Hom}(V, Y)$ for any given vector $w_0 \in W$ and the function $w \mapsto f(v_0, w)$ belongs to $\text{Hom}(W, Y)$ for any given vector $v_0 \in V$. The set of all bilinear transformations from $V \times W$ to Y will be denoted by $\text{Bil}(V \times W, Y)$. If $f, g \in \text{Bil}(V \times W, Y)$ and if $c \in F$ then $f + g$ and cf also belong to $\text{Bil}(V \times W, Y)$, and so $\text{Bil}(V \times W, Y)$ is a subspace of the vector space $Y^{V \times W}$ over F . Also, any bilinear transformation $f : V \times W \rightarrow Y$ defines a bilinear transformation $f^{op} : W \times V \rightarrow Y$, called the **opposite transformation** of f , by setting $f^{op} : (w, v) \mapsto f(v, w)$. It is clear that the function

$$(\)^{op} : \text{Bil}(V \times W, Y) \rightarrow \text{Bil}(W \times V, Y)$$

is an isomorphism of vector spaces. We say that a bilinear transformation $f \in \text{Bil}(V \times V, Y)$ is **symmetric** if and only if $f = f^{op}$.

In particular, if we consider a single vector space V over a field F , then we note that $f \in \text{Bil}(V \times V, V)$ if and only if the operation \bullet on V defined by $v \bullet w = f(v, w)$ turns V into an F -algebra. This algebra is commutative if and only if f is symmetric.

Example: Let F be a field and let k, n , and t be positive integers. Set $V = \mathcal{M}_{k \times n}(F)$, $W = \mathcal{M}_{t \times n}(F)$, and $Y = \mathcal{M}_{k \times t}(F)$. Then there

exists a bilinear transformation $V \times W \rightarrow Y$ defined by $(A, B) \mapsto AB^T$. In particular, we have a bilinear transformation $F^n \times F^n \rightarrow \mathcal{M}_{n \times n}(F)$ given by $(v, w) \mapsto v \wedge w$. More generally, if V , W , and Y are as mentioned, every matrix $C \in \mathcal{M}_{n \times n}(F)$ defines a bilinear transformation $V \times W \rightarrow Y$ by setting $(A, B) \mapsto ACB^T$.

Example: For vector spaces V and W over a field F , the function $\text{Hom}(V, W) \times V \rightarrow W$ given by $(\alpha, v) \mapsto \alpha(v)$ is a bilinear transformation.

Let V , W , and Y be vector spaces over a field F . The image of a bilinear transformation $f \in \text{Bil}(V \times W, Y)$ is not necessarily a subspace of Y , as the following example shows.

Example: Consider the bilinear transformation $f : \mathbb{R}^2 \times \mathbb{R}^2 \rightarrow \mathcal{M}_{2 \times 2}(\mathbb{R})$ defined by $f : (v, w) \mapsto v \wedge w$. The image of f contains $\begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$ and $\begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}$, but is not a subspace since $\begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \notin \text{im}(f)$.

As with linear transformations, bilinear transformations are totally determined by their behavior on bases. That is to say, let V and W be vector spaces over a field F , and let $B = \{v_i \mid i \in \Omega\}$ and $D = \{w_j \mid j \in \Lambda\}$ be bases of V and W respectively. Let Y be a vector space over F and let $f_0 : B \times D \rightarrow Y$ be a function. Then there exists a unique bilinear transformation $f \in \text{Bil}(V \times W, Y)$ satisfying $f(v_i, w_j) = f_0(v_i, w_j)$ for all i and j , namely the function defined by $f : \left(\sum_{i \in \Omega} a_i v_i, \sum_{j \in \Lambda} b_j w_j \right) \mapsto \sum_{i \in \Omega} \sum_{j \in \Lambda} a_i b_j f_0(v_i, w_j)$. In the case that $V = W = Y$, we have already noted this fact in Proposition 5.5.

(20.1) Proposition: If V , W , and Y are vector spaces over a field F , then $\text{Bil}(V \times W, Y)$ is isomorphic to $\text{Hom}(V, \text{Hom}(W, Y))$.

Proof: Define a function $\theta : \text{Bil}(V \times W, Y) \rightarrow \text{Hom}(V, \text{Hom}(W, Y))$ as follows: given a bilinear transformation $f \in \text{Bil}(V \times W, Y)$ and a vector $v \in V$, then $\theta(f)(v) : w \mapsto f(v, w)$. It is straightforward to check that indeed $\theta(f)(v) \in \text{Hom}(W, Y)$ for all $f \in \text{Bil}(V \times W, Y)$ and all $v \in V$. Moreover, $\theta(f)(v_1 + v_2) = \theta(f)(v_1) + \theta(f)(v_2)$ and $\theta(f)(cv) = c\theta(f)(v)$ for all $v, v_1, v_2 \in V$ and all $c \in F$, so $\theta(f) \in \text{Hom}(V, \text{Hom}(W, Y))$ for all $f \in \text{Bil}(V \times W, Y)$. Finally, $\theta(f + g) = \theta(f) + \theta(g)$ and $\theta(cf) = c\theta(f)$ for all $f, g \in \text{Bil}(V \times W, Y)$ and all $c \in F$, and so we have shown that θ is a linear transformation.

It is also possible to define a function

$$\varphi : \text{Hom}(V, \text{Hom}(W, Y)) \rightarrow \text{Bil}(V \times W, Y)$$

by setting $\varphi(\alpha) : (v, w) \mapsto \alpha(v)(w)$ for all $v \in V$ and $w \in W$, and again it is easy to show that this is a linear transformation. If $\alpha \in \text{Hom}(V, \text{Hom}(W, Y))$ and $v \in V$, then $\theta\varphi(\alpha)(v) : w \mapsto \varphi(\alpha)(v)(w) = \alpha(v)(w)$ and so $\theta\varphi(\alpha)(v) = \alpha(v)$ for all $v \in V$. Thus $\theta\varphi(\alpha) = \alpha$ for all $\alpha \in \text{Hom}(V, \text{Hom}(W, Y))$, and so $\theta\varphi$ is the identity function on $\text{Hom}(V, \text{Hom}(W, Y))$. Conversely, if $f \in \text{Bil}(V \times W, Y)$ then

$$\varphi\theta(f) : (v, w) \mapsto \theta(f)(v)(w) = f(v, w)$$

for all $v \in V$ and $w \in W$ and so $\varphi\theta(f) = f$ for all $f \in \text{Bil}(V \times W, Y)$, proving that $\varphi\theta$ is the identity function on $\text{Bil}(V \times W, Y)$. Thus we have established that θ is an isomorphism, with $\theta^{-1} = \varphi$. \square

Let V and W be vector spaces over a field F . A bilinear transformation $f : V \times W \rightarrow F$ is called a **bilinear form**. We will denote the set of all such bilinear forms by $\text{Bil}(V \times W)$, instead of $\text{Bil}(V \times W, F)$. By what we have seen above, $\text{Bil}(V \times W)$ is a subspace of $F^{V \times W}$ which is isomorphic to $\text{Hom}(V, D(W))$. If V and W are vector spaces over a field F , then a bilinear form $f \in \text{Bil}(V, W)$ is **nondegenerate** if and only if for each $0_V \neq v \in V$ there exists a $w \in W$ satisfying $f(v, w) \neq 0$ and for each $0_W \neq w \in W$ there exists a $v \in V$ satisfying $f(v, w) \neq 0$.

Example: If V is an inner product space over \mathbb{R} , then the function $(v, w) \mapsto \langle v, w \rangle$ belongs to $\text{Bil}(V \times V)$. This is not true, of course, if our field of scalars is \mathbb{C} .

Example: If F is a field and $V = F^n$ for some positive integer n , then the function $(v, w) \mapsto v \odot w$ belongs to $\text{Bil}(F^n \times F^n)$. This function is particularly useful in the case $F = GF(2)$. Indeed, if $v \in GF(2)^n$, then

$$v \odot v = \begin{cases} 0 & \text{if an even number of entries in } v \text{ are equal to } 1 \\ 1 & \text{if an odd number of entries in } v \text{ are equal to } 1 \end{cases}.$$

This value is known as the **parity** of v .

More generally, let A be a finite set and let V be the collection of all subsets of A . Define a bilinear form $f : V \times V \rightarrow GF(2)$ by setting

$$f(A, B) = \begin{cases} 0 & \text{if } A \cap B \text{ has an even number of elements} \\ 1 & \text{if } A \cap B \text{ has an odd number of elements} \end{cases}.$$

Then $f \in \text{Bil}(V, V)$.

Example: If V is a vector space over a field F , we have a nondegenerate bilinear form in $\text{Bil}(D(V) \times V)$ given by $(\delta, v) \mapsto \delta(v)$. Similarly, if $\delta_1, \delta_2 \in D(V)$, we have a bilinear form in $\text{Bil}(V \times V)$ given by $(v, w) \mapsto \delta_1(v)\delta_2(w)$, which is nondegenerate if δ_1 and δ_2 are not the 0-functional.

Example: If F is a field and if k and n are positive integers, then each matrix $A \in \mathcal{M}_{k \times n}(F)$ defines a bilinear form in $\text{Bil}(F^k \times F^n)$ by $(v, w) \mapsto v \odot Aw$.

If V and W are vector spaces of finite dimension k and n respectively over F , then any bilinear form on $V \times W$ can be represented as in the previous example. Indeed if we fix bases $B = \{v_1, \dots, v_k\}$ for V and $D = \{w_1, \dots, w_n\}$ for W , then for any $f \in \text{Bil}(V \times W)$ we define the matrix $T_{BD}(f) = [f(v_i, w_j)] \in \mathcal{M}_{k \times n}(F)$ and check that if $v = \sum_{i=1}^k a_i v_i$ and $w = \sum_{j=1}^n b_j w_j$, then $f(v, w) = v \odot T_{BD}(f)w$. Indeed, for fixed B and D , the function $f \mapsto T_{BD}(f)$ is an isomorphism from $\text{Bil}(V \times W)$ to $\mathcal{M}_{k \times n}(F)$.

Example: Let F be a field and let $V = F^2$. Consider the bases $B = \left\{ \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \end{bmatrix} \right\}$ and $D = \left\{ \begin{bmatrix} 1 \\ -1 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \end{bmatrix} \right\}$ of V . If $f \in \text{Bil}(V, V)$ is given by $f : \left(\begin{bmatrix} a \\ b \end{bmatrix}, \begin{bmatrix} c \\ d \end{bmatrix} \right) \mapsto (a+b)(c+d)$, then it is easy to verify that $T_{BB}(f) = \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}$ and $T_{DD}(f) = \begin{bmatrix} 0 & 0 \\ 0 & 4 \end{bmatrix}$.

(20.2) Proposition: Let V and W be vector spaces finitely generated over a field F and having bases $B = \{v_1, \dots, v_k\}$ and $D = \{w_1, \dots, w_n\}$ respectively. Let $C = \{x_1, \dots, x_k\}$ and $E = \{y_1, \dots, y_n\}$ also be bases for V and W respectively, and let $P = [p_{ir}] \in \mathcal{M}_{k \times k}(F)$ and $Q = [q_{js}] \in \mathcal{M}_{n \times n}(F)$ be nonsingular

matrices satisfying $\begin{bmatrix} x_1 \\ \vdots \\ x_k \end{bmatrix} = P \begin{bmatrix} v_1 \\ \vdots \\ v_k \end{bmatrix}$ and $\begin{bmatrix} y_1 \\ \vdots \\ y_n \end{bmatrix} = Q \begin{bmatrix} w_1 \\ \vdots \\ w_n \end{bmatrix}$.

Then for $f \in \text{Bil}(V \times W)$ we see that $T_{CD}(f) = PT_{BD}(f)Q^T$.

Proof: As a direct consequence of the definitions, we see that

$$f(x_i, y_j) = \left(\sum_{r=1}^k p_{ir} v_r, \sum_{s=1}^n q_{js} w_s \right) = \sum_{r=1}^k \sum_{s=1}^n p_{ir} f(v_r, w_s) q_{js},$$

and this is precisely the (i, j) entry of $PT_{BD}(f)Q^T$. \square

In particular, we see that if $f \in \text{Bil}(V \times V)$, where V is a vector space of finite dimension n over a field F , and if B and D are bases of V , then there exists a nonsingular matrix $P \in \mathcal{M}_{n \times n}(F)$ satisfying

$T_{DD}(f) = PT_{BB}(f)P^T$. In general, matrices A and C in $\mathcal{M}_{n \times n}(F)$ are **congruent** if and only if there exists a nonsingular matrix $P \in \mathcal{M}_{n \times n}(F)$ satisfying $C = PAP^T$. Congruence is easily checked to be an equivalence relation on $\mathcal{M}_{n \times n}(F)$, which joins the relations of equivalence and similarity, that we have already defined. Congruent matrices clearly have the same rank, so that the rank of a matrix of the form $T_{BB}(f)$ depends only on f and not on the choice of basis B . Therefore we call this the **rank** of the bilinear form f . Thus, for example, the bilinear forms in $\text{Bil}(V \times V)$ of rank 1 are precisely those of the form $(v, w) \mapsto \alpha(v)\beta(w)$, where $\alpha, \beta \in D(V)$.

A matrix congruent to a symmetric matrix is again symmetric. Indeed, if $A \in \mathcal{M}_{n \times n}(F)$ is symmetric, then for any nonsingular matrix P we have $(PAP^T)^T = P^{TT}A^TP^T = PAP^T$.

Example: The matrix $A = \begin{bmatrix} 1 & -6 & -6 \\ -6 & 40 & 39 \\ -6 & 39 & 39 \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{R})$ is congruent to I , since $PAP^T = I$, where $P = \begin{bmatrix} 1 & 0 & 0 \\ 3 & \frac{1}{2} & 0 \\ \sqrt{3} & -\frac{\sqrt{3}}{2} & \frac{2\sqrt{3}}{3} \end{bmatrix}$.

As was the case with inner products, we can define orthogonality with respect to an arbitrary bilinear form. This concept has important applications when we are working over fields other than \mathbb{R} or \mathbb{C} , and especially in areas such as algebraic coding theory, where all of the work is done over finite fields. Let V be a vector space over a field F and let $f \in \text{Bil}(V \times V)$. Vectors $v, w \in V$ are **f -orthogonal** if and only if $f(v, w) = 0$. In this case, we will write $v \perp_f w$. (One has to be careful here, it may be true that $v \perp_f w$ but false that $w \perp_f v$; this will not happen, of course, if f is symmetric.) If A is a nonempty subset of V , then we can talk about the **right f -orthogonal complement** of A to be the set $A^{\perp_f} = \{w \in V \mid v \perp_f w \text{ for all } v \in A\}$. Complements of this form may behave very differently than complements defined by inner products, as the following example shows.

Example: Let $F = GF(2)$ and let $V = F^4$. Define $f \in \text{Bil}(V \times V)$ by setting $f(v, w) = v \odot w$. Then $W = \left\{ \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \\ 1 \end{bmatrix} \right\}$ is a subspace of V which satisfies $W^{\perp_f} = W$.

We note that V^{\perp_f} is trivial if and only if for any $0_V \neq w \in V$ there exists a vector $v \in V$ satisfying $f(v, w) \neq 0$. This condition is not a

consequence of our definitions, and we must explicitly state it when we need it. It holds, of course, if f is nondegenerate.

Example: Let $V = \mathbb{R} \left\{ \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ -1 \\ -1 \end{bmatrix} \right\} \subseteq \mathbb{R}^4$. If $f \in \text{Bil}(V \times V)$ is defined by $f : \left(\begin{bmatrix} a_1 \\ a_2 \\ a_3 \\ a_4 \end{bmatrix}, \begin{bmatrix} b_1 \\ b_2 \\ b_3 \\ b_4 \end{bmatrix} \right) \mapsto a_1b_1 + a_2b_2 + a_3b_3 - a_4b_4$, then $\begin{bmatrix} 0 \\ 0 \\ 1 \\ 1 \end{bmatrix} \in V^{\perp_f}$ and, indeed, $V^{\perp_f} = \mathbb{R} \left\{ \begin{bmatrix} 1 \\ -1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \\ 1 \end{bmatrix} \right\}$.

(20.3) Proposition: Let V be a vector space finitely generated over a field F and let $f \in \text{Bil}(V \times V)$ satisfy the condition that V^{\perp_f} is trivial. Then each subspace W of V satisfies the following conditions:

- (1) If $\delta \in D(W)$ there exists a $v \in V$ such that $\delta(w) = f(v, w)$ for all $w \in W$;
- (2) $\dim(W) + \dim(W^{\perp_f}) = \dim(V)$.

Proof: (1) Every vector $v \in V$ defines a linear functional $\delta_v \in D(V)$ by setting $\delta_v : y \mapsto f(y, v)$. Moreover, the function $v \mapsto \delta_v$ from V to $D(V)$ is a linear transformation, which is a monomorphism as a result of the condition that V^{\perp_f} is trivial. But $\dim(V) = \dim(D(V))$ since V is finitely generated, and hence this is an isomorphism. Now let $\delta \in D(W)$ and let Y be a complement of W in V . Then the function from V to F given by $w + y \mapsto \delta(w)$ belongs to $D(V)$ and so there exists a vector $v \in V$ such that it equals δ_v . In particular, $\delta(w) = f(v, w)$ for all $w \in W$, proving (1).

(2) The function from V to $D(W)$ which assigns to each $v \in V$ the restriction of δ_v to W is a linear transformation which, by (1), is an epimorphism. The kernel of this epimorphism consists of all vectors $v \in V$ satisfying $f(w, v) = 0$ for all $w \in W$, and that is precisely W^{\perp_f} . Therefore, by Proposition 6.10, we have (2). \square

In particular, we see from Proposition 20.3, that a necessary and sufficient condition for us to have $V = W \oplus W^{\perp_f}$ is that W and W^{\perp_f} be disjoint.

(20.4) Proposition: Let V and W be vector spaces finitely generated over a field F and let $f \in \text{Bil}(V \times W)$ be a bilinear form

which is not the 0-function. Then there exist bases $\{v_1, \dots, v_k\}$ and $\{w_1, \dots, w_n\}$ of V and W respectively and there exists a positive integer $1 \leq t \leq \min\{k, n\}$ such that

$$f(v_i, w_j) = \begin{cases} 1 & \text{if } i = j \leq t \\ 0 & \text{otherwise} \end{cases}.$$

Proof: Since f is not the 0-function, there exist vectors $v_1 \in V$ and $y_1 \in W$ such that $f(v_1, y_1) \neq 0$. Therefore, if we set $w_1 = f(v_1, y_1)^{-1}y_1$, we have $f(v_1, w_1) = 1$. Let $V_1 = Fv_1$ and $W_1 = Fw_1$. If we set $W_2 = \{w \in W \mid f(v_1, w) = 0\}$, then $W_1 \cap W_2 = \{0_W\}$ since $cw_1 \notin W_2$ for all $0 \neq c \in F$. We claim that $W = W_1 \oplus W_2$. Indeed, if $w \in W$ and if $c = f(v_1, w)$ then we see that

$$f(v_1, w - cw_1) = f(v_1, w) - cf(v_1, w_1) = c - c = 0$$

and so $w - cw_1 \in W_2$, which proves the claim. In a similar way, we have $V = V_1 \oplus V_2$, where $V_2 = \{v \in V \mid f(v, w_1) = 0\}$. Thus we see that $f(v, w) = 0$ whenever $(v, w) \in [V_1 \times W_2] \cup [V_2 \times W_1]$.

By passing to the oppose form if necessary, we can assume without loss of generality that $k \leq n$. If $k = 1$, we choose $\{v_1\}$ as a basis for V and $\{w_1, \dots, w_n\}$ as a basis for W , where $\{w_2, \dots, w_n\}$ is an arbitrary basis for W_2 . This proves the proposition, with $t = 1$. Now assume that $k > 1$ (which implies $n > 1$) and that the proposition has been proven whenever $\dim(V) < k$. In particular, we will look at the restriction of f to $V_2 \times W_2$. By the induction hypothesis, there exist bases $\{v_2, \dots, v_k\}$ of V_2 and $\{w_2, \dots, w_n\}$ of W_2 such that

$$f(v_i, w_j) = \begin{cases} 1 & \text{if } 2 \leq i = j \leq t \\ 0 & \text{otherwise} \end{cases}.$$

Then $\{v_1, \dots, v_k\}$ and $\{w_1, \dots, w_n\}$ are the bases we want. \square

We see that if V and W are vector spaces finitely generated over a field F and if $f \in \text{Bil}(V \times W)$, then Proposition 20.4 says that there exist bases of V and W with respect to which f is represented by a

matrix of the form $\begin{bmatrix} I & O \\ O & O \end{bmatrix}$.

We will be particularly interested in symmetric bilinear forms. As an immediate consequence of the definition, we see that if V is a vector space finitely generated over a field F and if B is a given basis for V , then a bilinear form $f \in \text{Bil}(V \times V)$ is symmetric if and only if the matrix $T_{BB}(f)$ is symmetric. Moreover, every symmetric matrix is $T_{BB}(f)$ for some symmetric bilinear form $f \in \text{Bil}(V \times V)$.

Example: Let B be the canonical basis of \mathbb{R}^3 and let $A = \begin{bmatrix} 1 & -5 & 3 \\ -5 & 1 & 7 \\ 3 & 7 & 4 \end{bmatrix}$. Then $A = T_{BB}(f)$, where $f \in \text{Bil}(\mathbb{R}^3 \times \mathbb{R}^3)$ is defined by

$$f\left(\begin{bmatrix} a_1 \\ a_2 \\ a_3 \end{bmatrix}, \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix}\right) = a_1b_1 + a_2b_2 - 5(a_1b_2 + a_2b_1) \\ + 3(a_1b_3 + a_3b_1) + 7(a_2b_3 + a_3b_2) + 4a_3b_3.$$

(20.5) Proposition: Let F be a set of characteristic other than 2 and let V be a vector space finitely-generated over F . Let $f \in \text{Bil}(V \times V)$ be symmetric. Then there exists a basis $B = \{v_1, \dots, v_n\}$ of V such that $T_{BB}(f)$ is a diagonal matrix.

Proof: The proposition is trivially true if f is the 0-function, and so we can assume that that is not the case. We will proceed by induction on $n = \dim(V)$. For $n = 1$, the result is again immediate, and so we can assume that $n > 1$ and that the result has been established for all spaces having dimension less than n . We first claim is that there exists a vector $v \in V$ satisfying $f(v, v) \neq 0$. Indeed, assume that this is not the case. Then if v and w are arbitrary vectors in V we have

$$0 = f(v + w, v + w) = f(v, v) + 2f(v, w) + f(w, w) = 2f(v, w)$$

and since the characteristic of F is not 2, this implies that $f(v, w) = 0$, contradicting our assumption that f is not the 0-function. Hence we can select a vector $v_1 \in V$ satisfying $f(v_1, v_1) \neq 0$.

Let $V_1 = Fv_1$ and let $V_2 = V_1^{\perp f}$. From the definition of V_1 it is clear that V_1 and V_2 are disjoint, and from Proposition 20.3 it follows that $V = V_1 \oplus V_2$. In particular, $\dim(V_2) = n - 1$ and so, by the induction hypothesis, there exists a basis $C = \{v_2, \dots, v_n\}$ of V_2 , such that, if f_2 is the restriction of f to V_2 , then $T_{CC}(f_2)$ is a diagonal matrix. Since $f(v_1, v_i) = 0$ for all $2 \leq i \leq n$, it follows that $B = \{v_1, \dots, v_n\}$ does indeed give us the desired result. \square

Thus we see that every symmetric matrix over a field of characteristic other than 2 is congruent to a diagonal matrix.

(20.6) Proposition: Let V be a vector space finitely-generated over \mathbb{C} and let $f \in \text{Bil}(V \times V)$ be a symmetric bilinear form of

rank r . Then there exists a basis $B = \{v_1, \dots, v_n\}$ of V satisfying the following conditions:

(1) $T_{BB}(f)$ is a diagonal matrix;

(2) $f(v_i, v_i) = \begin{cases} 1 & \text{if } 1 \leq i \leq r \\ 0 & \text{otherwise} \end{cases}$.

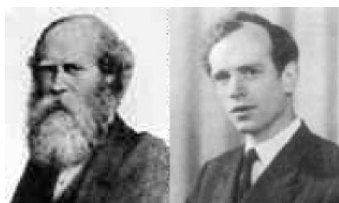
Proof: By Proposition 20.5, we know that there is a basis $B = \{v_1, \dots, v_n\}$ of V satisfying the condition that $T_{BB}(f)$ is a diagonal matrix. This matrix is of rank r and so, renumbering the basis elements if necessary, we can assume that $f(v_i, v_i) \neq 0$ when and only when $1 \leq i \leq r$. For each $1 \leq i \leq r$, define $c_i = f(v_i, v_i)^{-1/2} \in \mathbb{C}$, and replace v_i by $c_i v_i$ to get a basis satisfying (2) as well. \square

Let V be a vector space finitely-generated over a field F of characteristic other than 2 and let $f \in \text{Bil}(V \times V)$ be a bilinear form. The function $q : V \rightarrow F$ defined by $q : v \mapsto f(v, v)$ is called the **quadratic form** defined by f . Note that if $a \in F$ and $v \in V$ then $q(av) = f(av, av) = a^2 f(v, v) = a^2 q(v)$. Moreover, if $f \in \text{Bil}(V \times V)$ and if $g \in \text{Bil}(V \times V)$ is the symmetric bilinear form given by

$$g : (v, w) \mapsto \frac{1}{2} [f(v, w) + f(w, v)]$$

then the quadratic forms defined by f and g are the same. Therefore, without loss of generality, we will always assume that all quadratic forms over such fields are defined by symmetric bilinear forms. We further see that different symmetric bilinear forms define different quadratic forms, since, for any $v, w \in V$, we have $f(v, w) = \frac{1}{2} [q(v+w) - q(v) - q(w)]$. The classification of quadratic forms is of great importance in analytic geometry and in number theory.¹

Example: If V is an inner product space over \mathbb{R} , then we have already noted that the function $f : (v, w) \mapsto \langle v, w \rangle$ is a symmetric bilinear form. The quadratic form defined by f is given by $v \mapsto \|v\|^2$.



¹

The theory of quadratic forms over \mathbb{R} was developed by Gauss and his student Eisenstein, and the need to study such forms was one of the factors which led to the development of determinant theory. Their work was extended to quadratic forms over \mathbb{C} by the 19th-century British mathematician **Henry Smith**. The fundamental development in the theory of symmetric bilinear forms on vector spaces over fields of characteristic other than 2 is due to the 20th-century German mathematician **Ernst Witt**.

Example: If V is the vector space of all polynomial functions from \mathbb{R} to itself, then we have a symmetric bilinear form from $V \times V$ to \mathbb{R} defined by $(f, g) \mapsto \int_0^1 f(t)g(t)dt$, which in turn defines the quadratic form $f \mapsto \int_0^1 f(t)^2 dt$.

Example: Let $V = \mathbb{R}^4$ and let $f \in \text{Bil}(V \times V)$ be the symmetric bilinear form defined by

$$\left(\begin{bmatrix} a_1 \\ a_2 \\ a_3 \\ a_4 \end{bmatrix}, \begin{bmatrix} b_1 \\ b_2 \\ b_3 \\ b_4 \end{bmatrix} \right) \mapsto a_1 b_1 + a_2 b_2 + a_3 b_3 - a_4 b_4,$$

which lies at the center of Minkowski's mathematical formulation of Einstein's relativity theory. The quadratic form defined by this bilinear form

$$\text{is } \begin{bmatrix} a_1 \\ a_2 \\ a_3 \\ a_4 \end{bmatrix} \mapsto a_1^2 + a_2^2 + a_3^2 - a_4^2. \text{ A similar symmetric bilinear form is}$$

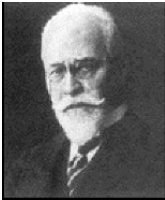
$$\text{the Lorentz}^2 \text{ form } \left(\begin{bmatrix} a_1 \\ a_2 \\ a_3 \\ a_4 \end{bmatrix}, \begin{bmatrix} b_1 \\ b_2 \\ b_3 \\ b_4 \end{bmatrix} \right) \mapsto a_1 b_1 + a_2 b_2 + a_3 b_3 - c^2 a_4 b_4,$$

where c is the speed of light. The quadratic form defined by this bilinear

$$\text{form is } \begin{bmatrix} a_1 \\ a_2 \\ a_3 \\ a_4 \end{bmatrix} \mapsto a_1^2 + a_2^2 + a_3^2 - c^2 a_4^2.$$

A more general result, also based on the work of Lorentz and Minkowski, gives a fascinating "reversal" of the Cauchy-Schwarz-Bunyakovski inequality. Let n be a positive integer and consider the subset (not subspace)

U of \mathbb{R}^{n+1} consisting of all vectors of the form $\begin{bmatrix} a \\ v \end{bmatrix}$, where a is a nonnegative real number and $v \in \mathbb{R}^n$ satisfies $\|v\| \leq a$. For $u = \begin{bmatrix} a \\ v \end{bmatrix}$



² Dutch physicist **Hendrick Antoon Lorentz**, the first to conceive of the notion of the electron, won a Nobel prize in 1902. His work formed a basis for much of Einstein's theory.

and $y = \begin{bmatrix} b \\ w \end{bmatrix}$ in U , let us define $u \boxdot y$ to be $ab - v \cdot w$. By our assumption on U , we note that $u \boxdot u \geq 0$ for every $u \in U$. Then one can show that $u \boxdot y \geq \left[\sqrt{u \boxdot u} \right] \left[\sqrt{y \boxdot y} \right]$. This inequality is often known as the **lightcone inequality**, because of its applications in physics.

By Proposition 20.6 we see that if V is a vector space finitely generated over \mathbb{C} and if $f \in \text{Bil}(V \times V)$ is symmetric and has rank r , we can find a basis $\{v_1, \dots, v_n\}$ of V such that the quadratic form q defined by f is given by $q : \sum_{i=1}^n a_i v_i \mapsto \sum_{i=1}^r a_i^2$.

Example: Let F be either \mathbb{R} or \mathbb{C} . Let n be a positive integer and let $A \in \mathcal{M}_{n \times n}(F)$ be symmetric. Let $f \in \text{Bil}(F^n, F^n)$ be the symmetric bilinear form given by $f : (v, w) = v^T A w$, and let q be the quadratic form defined by f . The set $\{q(v) \mid \|v\| = 1\}$ (here the norm is the one defined by the dot product on F^n) is called the **numerical range** of the matrix A . In the case $F = \mathbb{C}$, this is always a bounded convex subset which contains all of the eigenvalues of A . For the special case $n = 2$, this set is an ellipse with its foci at the eigenvalues of A , assuming that they are distinct, or a circle with center at the sole eigenvalue of A , assuming that A has only one eigenvalue of multiplicity 2. For $n > 2$, the characterization of the numerical range is much more complicated.

(20.7) Proposition: Let n be a positive integer and let $A \in \mathcal{M}_{n \times n}(\mathbb{R})$ be symmetric. Let $f \in \text{Bil}(\mathbb{R}^n \times \mathbb{R}^n)$ be the symmetric bilinear form given by $f : (v, w) = v^T A w$, and let q be the quadratic form defined by f . Let $c_1 \geq c_2 \geq \dots \geq c_n$ be the eigenvalues of A . Then the numerical range of A lies in the closed interval $[c_n, c_1]$. Moreover, both endpoints of this interval belong to the numerical range of A .

Proof: By Proposition 17.7, we know that there exists an orthonormal basis $B = \{v_1, \dots, v_n\}$ of V consisting of eigenvectors of A . Moreover, if $v \in V$ then $v = \sum_{i=1}^n \langle v, v_i \rangle v_i$ by Proposition 17.9, and so $1 = \|v\|^2 = \langle v, v \rangle = \sum_{i=1}^n \langle v, v_i \rangle^2$. We also see that $Av = \sum_{i=1}^n \langle v, v_i \rangle A(v_i) = \sum_{i=1}^n c_i \langle v, v_i \rangle v_i$. Thus $v^T Av = \langle v, Av \rangle = \sum_{i=1}^n c_i \langle v, v_i \rangle^2$. Since

$$c_1 = c_1 \left(\sum_{i=1}^n \langle v, v_i \rangle^2 \right) \geq \sum_{i=1}^n c_i \langle v, v_i \rangle^2 \geq c_n \left(\sum_{i=1}^n \langle v, v_i \rangle^2 \right) = c_n.$$

Therefore the numerical range of A lies in the closed interval $[c_n, c_1]$.

If v is a normal eigenvector of A corresponding to c_n , then $v^T Av = \langle v, Av \rangle = \langle v, c_n v \rangle = c_n \langle v, v \rangle = c_n$, and similarly for the case of an eigenvector of A satisfying $\|v\| = 1$ and corresponding to c_1 . \square

In order to see the geometric significance of quadratic forms, let us recall that a **general quadratic equation** in three unknowns over \mathbb{R} is one of the form

$$(a_{11}X_1^2 + a_{22}X_2^2 + a_{33}X_3^2) + 2(a_{12}X_1X_2 + a_{13}X_1X_3 + a_{23}X_2X_3) + b_1X_1 + b_2X_2 + b_3X_3 + c = 0$$

in which not all of the a_{ij} are equal to 0. Such an equation can be written in the form $f(v, v) + w \cdot v + c = 0$, where $f \in \text{Bil}(\mathbb{R}^3, \mathbb{R}^3)$ is the symmetric bilinear form defined with respect to the canonical basis by

the matrix $A = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{12} & a_{22} & a_{23} \\ a_{13} & a_{23} & a_{33} \end{bmatrix}$, where $w = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix}$, and where

$v = \begin{bmatrix} X_1 \\ X_2 \\ X_3 \end{bmatrix}$. The graph of such an equation is a **quadratic surface**.

The various quadratic surfaces in \mathbb{R}^3 can then be classified by considering congruence classes of the matrices A , a task very important in analytic geometry.

We will now return to the general case of bilinear transformations. Let F be a field, let V and W be vector spaces over F , and let $G = F^{(V \times W)}$. Then G is a subspace of $F^{V \times W}$ having a basis $\{g_{v,w} \mid (v,w) \in V \times W\}$, where

$$g_{v,w} : (v', w') \mapsto \begin{cases} 1 & \text{if } (v', w') = (v, w) \\ 0 & \text{otherwise} \end{cases}.$$

Let H be the subspace of G generated by all functions of the form

$$g_{v_1+v_2, w} - g_{v_1, w} - g_{v_2, w}, \quad g_{v, w_1+w_2} - g_{v, w_1} - g_{v, w_2}, \quad g_{av, w} - ag_{v, w}, \\ \text{or } g_{v, aw} - ag_{v, w}$$

for all $v, v_1, v_2 \in V$, $w, w_1, w_2 \in W$, and $a \in F$. Let us pick a complement of H in G , and call it $V \otimes W$. By Proposition 7.8, we know that $V \otimes W$ is unique up to isomorphism. Let α be the projection of G with image $V \otimes W$ coming from the decomposition $G = H \oplus (V \otimes W)$ and, for all $v \in V$ and $w \in W$, denote $\alpha(g_{v,w})$ by $v \otimes w$. Then $B = \{v \otimes w \mid (v, w) \in V \times W\}$ is a generating set for $V \otimes W$. It is important to emphasize that the elements of $V \otimes W$ are linear combinations of elements of B . In quantum physics, elements of $V \otimes W \setminus B$, for suitable spaces V and W , are known as **entangled tensors** and these have important physical interpretations. Elements of B are known as **simple tensors**.

If $v_1, v_2 \in V$ and $w \in W$, then

$$[v_1 + v_2] \otimes w - (v_1 \otimes w) - (v_2 \otimes w) = \alpha(g_{v_1+v_2, w} - g_{v_1, w} - g_{v_2, w}) = 0_G$$

and so $[v_1 + v_2] \otimes w = (v_1 \otimes w) + (v_2 \otimes w)$. Similarly, if $v \in V$ and $w_1, w_2 \in W$ then $v \otimes [w_1 + w_2] = v \otimes w_1 + v \otimes w_2$. We also see that if $v \in V$, $w \in W$ and $c \in F$, then $cv \otimes w = c(v \otimes w) = v \otimes cw$. The vector space $V \otimes W$ is called the **tensor product**³ of V and W .

From the definition of the tensor product, we see that the function t_{VW} from $V \times W$ to $V \otimes W$ given by $(v, w) \mapsto v \otimes w$ is a bilinear transformation. This transformation has a very special significance, due to the following theorem, which allows us to move from bilinear transformations to linear transformations.

(20.8) Proposition: Let V , W , and Y be vector spaces over a field F . For each bilinear transformation $f \in \text{Bil}(V \times W, Y)$ there exists a unique linear transformation $\alpha \in \text{Hom}(V \otimes W, Y)$ satisfying $f = \alpha t_{VW}$.

Proof: Given $f \in \text{Bil}(V \times W, Y)$, there exists a linear transformation $\beta \in \text{Hom}(G, Y)$ defined on the elements of a basis of the space G defined above, given by the condition that $\beta : g_{v,w} \mapsto f(v, w)$. Since f is a bilinear transformation, $H \subseteq \ker(\beta)$ and so we can define the linear transformation $\alpha \in \text{Hom}(V \otimes W, Y)$ by setting

$$\alpha : \sum_{i=1}^n a_i [v_i \otimes w_i] \mapsto \sum_{i=1}^n a_i f(v_i, w_i).$$

This function is well-defined since if $\sum_{i=1}^n a_i [v_i \otimes w_i] = \sum_{i=1}^n b_i [v_i \otimes w_i]$ in $V \otimes W$ then $\sum_{i=1}^n (a_i - b_i) g_{v_i, w_i} \in H \subseteq \ker(\beta)$. Therefore

$$\alpha \left(\sum_{i=1}^n a_i [v_i \otimes w_i] \right) = \alpha \left(\sum_{i=1}^n b_i [v_i \otimes w_i] \right)$$

Clearly α is a linear transformation and satisfies $f = \alpha t_{VW}$.



There are many equivalent definitions of the tensor product. The definition given here is due to the 20th-century French mathematician, **Claude Chevalley**. The notion of a tensor was first introduced in differential calculus by the 19th-century Italian mathematicians **Gregorio Ricci-Curbastro** and **Tullio Levi-Civita** and became a central tool in relativity theory.

We are left to prove uniqueness. Suppose that $\gamma \in \text{Hom}(V \otimes W, Y)$ satisfies $f = \gamma t_{VW}$. In particular, $\alpha(v \otimes w) = \gamma(v \otimes w)$ for all $(v, w) \in V \times W$. That is to say, α and γ act identically on a generating set for $V \otimes W$ and so, in particular, on a basis for $V \otimes W$ contained in this generating set. Therefore, by Proposition 6.2, it follows that $\alpha = \gamma$. \square

The following proposition is very important, and is often used as a basis for the definition of the tensor product.

(20.9) Proposition: If V , W , and Y are vector spaces over a field F then the vector spaces $\text{Hom}(V \otimes W, Y)$ and $\text{Hom}(V, \text{Hom}(W, Y))$ are isomorphic.

Proof: It is easy to see that the function

$$\text{Hom}(V \otimes W, Y) \rightarrow \text{Bil}(V \times W, Y)$$

defined by $\beta \mapsto \beta t_{VW}$ is a linear transformation and from Proposition 20.8 it follows that this is an isomorphism. Therefore the result follows from Proposition 20.1. \square

Example: Let V and W be vector spaces over a field F and let $\delta_1 \in D(V)$ and $\delta_2 \in D(W)$ be linear functionals. Then there exists a bilinear form in $\text{Bil}(V \times W)$ defined by $(v, w) \mapsto \delta_1(v)\delta_2(w)$. From Proposition 20.8, it follows that there exists a linear functional $\delta_1 \otimes \delta_2 \in D(V \otimes W)$ satisfying $\delta_1 \otimes \delta_2 : \sum_{i=1}^n a_i [v_i \otimes w_i] \mapsto \sum_{i=1}^n a_i \delta_1(v_i)\delta_2(w_i)$.

Example: More generally, let V and W be vector spaces over a field F , let α be an endomorphism of V , and let β be an endomorphism of W . The function from $V \times W$ to $V \otimes W$ defined by

$$(v, w) \mapsto \alpha(v) \otimes \beta(w)$$

is a bilinear transformation and so defines an endomorphism $\alpha \otimes \beta$ of $V \otimes W$ satisfying $\alpha \otimes \beta : \sum_{i=1}^n a_i [v_i \otimes w_i] \mapsto \sum_{i=1}^n a_i [\alpha(v_i) \otimes \beta(w_i)]$.

By Proposition 5.13, we know that if $V \oplus W$ is a vector space finitely-generated over a field F , then $\dim(V \oplus W) = \dim(V) + \dim(W)$. We now prove the “multiplicative” analog of this assertion for tensor products.

(20.10) Proposition: Let V and W be vector spaces finitely generated over a field F . Then $V \otimes W$ is also finitely generated, and $\dim(V \otimes W) = \dim(V) \dim(W)$.

Proof: Let us choose bases $\{v_1, \dots, v_k\}$ of V and $\{w_1, \dots, w_n\}$ of W . Then for any $v = \sum_{i=1}^k a_i v_i \in V$ and $w = \sum_{j=1}^n b_j w_j \in$

W , we see that $v \otimes w = \sum_{i=1}^k \sum_{j=1}^n a_i b_j (v_i \otimes w_j)$. Thus we see that $\{v_i \otimes w_j \mid 1 \leq i \leq k \text{ and } 1 \leq j \leq n\}$ is a generating set for $V \otimes W$, showing that $V \otimes W$ is finitely-generated. Moreover, by Proposition 20.9 and Proposition 14.6, we see that the dimension of $V \otimes W$ is equal to the dimension of $D(V \otimes W)$ and hence to the dimension of $\text{Hom}(V, D(W))$, and this is equal to the dimension of $\text{Hom}(V, W)$, which is precisely $\dim(V) \dim(W)$. \square

In particular, we see that in the context of Proposition 20.10, the set $\{v_i \otimes w_j \mid 1 \leq i \leq k \text{ and } 1 \leq j \leq n\}$ is in fact a basis of $V \otimes W$.

Example: Let F be a field and let k and n be positive integers. Then, by Proposition 20.10, we know that $\dim(F^k \otimes F^n) = kn = \dim(\mathcal{M}_{k \times n}(F))$, and so the vector spaces $F^k \otimes F^n$ and $\mathcal{M}_{k \times n}(F)$ are isomorphic. Indeed, if we choose bases $\{v_1, \dots, v_k\}$ of V and $\{w_1, \dots, w_n\}$ of W , then the function $v_i \otimes w_j \mapsto v_i \wedge w_j$ extends to an isomorphism between these two spaces.

Example: Let F be a field, let n be a positive integer, and let V be a vector space finitely generated over F and having a basis $\{v_1, \dots, v_k\}$. The dimension of the vector space $\mathcal{M}_{n \times n}(V)$ over F is $n^2 k$. Consider the bilinear transformation $f: \mathcal{M}_{n \times n}(F) \times V \rightarrow \mathcal{M}_{n \times n}(V)$ defined by $([a_{ij}], v) \mapsto [a_{ij}v]$. By Proposition 20.8, we know that this bilinear transformation defines a linear transformation $\alpha: \mathcal{M}_{n \times n}(F) \otimes V \rightarrow \mathcal{M}_{n \times n}(V)$ and it is clear that this is an epimorphism. But, by Proposition 20.10, we see that the dimension of $\mathcal{M}_{n \times n}(F) \otimes V$ is also equal to $n^2 k$ and so α must be an isomorphism.

Example: Let F be a field and let k, n, s , and t be positive integers. Let $f: \mathcal{M}_{k \times n}(F) \times \mathcal{M}_{s \times t}(F) \rightarrow \mathcal{M}_{ks \times nt}(F)$ be the function defined by

$$f: (A, B) \mapsto \begin{bmatrix} a_{11}B & \dots & a_{1n}B \\ \vdots & & \vdots \\ a_{k1}B & \dots & a_{kn}B \end{bmatrix}. \quad \text{This is a bilinear transformation}$$

of vector spaces over F and so by Proposition 20.8, it defines a linear transformation $\alpha: \mathcal{M}_{k \times n}(F) \otimes \mathcal{M}_{s \times t}(F) \rightarrow \mathcal{M}_{ks \times nt}(F)$ which, again, can be shown to be an isomorphism. In the literature, it is usual to write $A \otimes B$ instead of $f(A, B)$. This matrix is called the **Kronecker product** of the matrices A and B . Kronecker products are very important in matrix theory and its applications. It is easy to see that for all such matrices A and B we have $(A \otimes B)^T = A^T \otimes B^T$. Moreover, if $k = n$ and $s = t$ and if A and B are nonsingular, then $A \otimes B$ is nonsingular, and $(A \otimes B)^{-1} = A^{-1} \otimes B^{-1}$. We also note that if A and B are symmetric

then so is $A \otimes B$. Furthermore, Cholesky or QR-factorizations of $A \otimes B$ come immediately from the corresponding factorizations of A and B .

Let V , V' , W and W' be vector spaces over a field F . If $\alpha \in \text{Hom}(V, V')$ and $\beta \in \text{Hom}(W, W')$ then we have a bilinear transformation $V \times W \rightarrow V' \otimes W'$ defined by $(v, w) \mapsto \alpha(v) \otimes \beta(w)$ and so, by Proposition 20.8, there exists a linear transformation from $V \otimes W$ to $V' \otimes W'$ satisfying $v \otimes w \mapsto \alpha(v) \otimes \beta(w)$. We will denote this linear transformation by $\alpha \otimes \beta$.

(20.11) Proposition: Let V , V' , W , and W' be vector spaces finitely generated over a field F . Any element of the space $\text{Hom}(V \otimes W, V' \otimes W')$ is of the form $\sum_{i=1}^n \alpha_i \otimes \beta_i$, where $\alpha_i \in \text{Hom}(V, V')$ and $\beta_i \in \text{Hom}(W, W')$ for each $1 \leq i \leq n$.

Proof: The function $(\alpha, \beta) \mapsto \alpha \otimes \beta$ from $\text{Hom}(V, V') \times \text{Hom}(W, W')$ to $\text{Hom}(V \otimes W, V' \otimes W')$ is bilinear and so defines a linear transformation $\varphi : \text{Hom}(V, V') \otimes \text{Hom}(W, W') \rightarrow \text{Hom}(V \otimes W, V' \otimes W')$. We are done if we can show that φ is an isomorphism. By Propositions 8.1 and 20.10, we know that

$$\begin{aligned} & \dim(\text{Hom}(V, V') \otimes \text{Hom}(W, W')) \\ &= \dim(\text{Hom}(V, V')) \dim(\text{Hom}(W, W')) \\ &= \dim(V) \dim(V') \dim(W) \dim(W') \\ &= \dim(V \otimes W) \dim(V' \otimes W') \\ &= \dim(\text{Hom}(V \otimes W, V' \otimes W')) \end{aligned}$$

and so it suffices to prove that φ is a monomorphism.

Indeed, assume that $\sum_{i=1}^n \alpha_i \otimes \beta_i \in \ker(\varphi)$, where the set $\{\beta_1, \dots, \beta_n\}$ is linearly independent, and where none of the α_i is the 0-function. Then $\sum_{i=1}^n \alpha_i(v) \otimes \beta_i(w) = 0_{V' \otimes W'}$ for all $v \in V$ and all $w \in W$. Pick $v \in V$ satisfying $\alpha_1(v) \neq 0_{V'}$. By renumbering if necessary, we can assume that $\{\alpha_1(v), \dots, \alpha_k(v)\}$ is a maximal linearly-independent subset of $\{\alpha_1(v), \dots, \alpha_n(v)\}$. Therefore, for each $k < h \leq n$ there exists a scalar b_{hj} , not all of them being equal to 0, such that $\alpha_h(v) = \sum_{j=1}^k b_{hj} \alpha_j(v)$ and so

$$\begin{aligned} 0_{V' \otimes W'} &= \sum_{i=1}^k \alpha_i(v) \otimes \beta_i(w) + \sum_{h=k+1}^m \left(\sum_{j=1}^k b_{hj} \alpha_j(v) \right) \otimes \beta_h(w) \\ &= \sum_{i=1}^k \alpha_i(v) \otimes \beta_i(w) + \sum_{j=1}^k \alpha_j(v) \otimes \left(\sum_{h=k+1}^m b_{hj} \beta_h(w) \right) \\ &= \sum_{i=1}^k \alpha_i(v) \otimes \left(\beta_i(w) + \sum_{h=k+1}^m b_{hi} \beta_h(w) \right). \end{aligned}$$

Since the set $\{\alpha_1(v), \dots, \alpha_k(v)\}$ is linearly independent, we must have $\beta_i(w) + \sum_{h=k+1}^m b_{hj}\beta_h(w) = 0_{W'}$ for all $1 \leq i \leq k$ and all $w \in W$. Hence $\beta_i + \sum_{h=k+1}^m b_{hj}\beta_h$ is the 0-function for all $1 \leq i \leq k$, contradicting the assumption that the set $\{\beta_1, \dots, \beta_n\}$ is linearly independent. We therefore conclude that $\ker(\varphi)$ is trivial, which is what we needed to prove. \square

(20.12) Proposition: If U , V and W are vector spaces over a field F , then $U \otimes (V \otimes W) \cong (U \otimes V) \otimes W$.

Proof: The bilinear transformation $U \times (V \otimes W) \rightarrow (U \otimes V) \otimes W$ defined by $(u, v \otimes w) \mapsto (u \otimes v) \otimes w$ induces a linear transformation $\alpha : U \otimes (V \otimes W) \rightarrow (U \otimes V) \otimes W$ which satisfies $u \otimes (v \otimes w) \mapsto (u \otimes v) \otimes w$. Similarly we have a linear transformation $\beta : (U \otimes V) \otimes W \rightarrow U \otimes (V \otimes W)$ which satisfies $(u \otimes v) \otimes w \mapsto u \otimes (v \otimes w)$. Since $\alpha\beta$ and $\beta\alpha$ are clearly the respective identity maps, we see that α must be the isomorphism we seek. \square

(20.13) Proposition: If V and W are vector spaces over a field F , then $V \otimes W \cong W \otimes V$.

Proof: The bilinear transformation $V \times W \rightarrow W \otimes V$ defined by $(v, w) \mapsto w \otimes v$ induces a linear transformation α from $V \otimes W$ to $W \otimes V$ satisfying $\alpha : v \otimes w \mapsto w \otimes v$. Similarly, there exists a linear transformation $\beta : W \otimes V \rightarrow V \otimes W$ satisfying $\beta : w \otimes v \mapsto v \otimes w$. Since $\alpha\beta$ and $\beta\alpha$ are clearly the respective identity maps, we see that α must be the isomorphism we seek. \square

Finally, let us briefly mention two algebras built on the notion of the tensor product. The study of these algebras is beyond the scope of this book. However, the reader should be aware of them and will find it fruitful to explore them further. In what ensues, V is an arbitrary vector space over a field F .

(I) For each nonnegative integer k , we define the vector space $V^{\otimes k}$ over F by setting $V^{\otimes 0} = V$ and $V^{\otimes k} = V^{\otimes(k-1)} \otimes V$ if $k > 0$. Let $T(V) = \coprod_{k=0}^{\infty} V^{\otimes k}$. We can define a product \bullet on $T(V)$ by setting

$$(v_1 \otimes \dots \otimes v_k) \bullet (v_{k+1} \otimes \dots \otimes v_m) = v_1 \otimes \dots \otimes v_{k+m}$$

for all $v_1, \dots, v_{k+m} \in V$ and extend linearly. This is an F -algebra, known as the **tensor algebra** of V over F . The tensor algebra has several important properties, one of which is that if K is any algebra over F then any linear transformation $\alpha : V \rightarrow K$ can be uniquely extended to a homomorphism of F -algebras from $T(V)$ to K . Moreover, if W is a vector space over F then any linear transformation $\alpha : V \rightarrow W$

can be uniquely extended to a homomorphism of F -algebras from $T(V)$ to $T(W)$. (In the language of category theory, this says that $T(\)$ is a functor from the category of vector spaces over F to the category of F -algebras.)

(II) Let Y be the subspace of $V \otimes V$ generated by $\{v \otimes v \mid v \in V\}$. Then a complement of Y in $V \otimes V$ is called an **exterior square** of V and is denoted by $V \wedge V$. This space is unique up to isomorphism. If α is the projection of $V \otimes V$ with image $V \wedge V$ and kernel Y , denote $\alpha(v \otimes w)$ by $v \wedge w$. Since

$$(v + w) \otimes (v + w) = v \otimes v + v \otimes w + w \otimes v + w \otimes w$$

for all $v, w \in V$, we see that $v \wedge w = -w \wedge v$ for all $v, w \in V$. Therefore, if V is finitely-generated over F with basis $\{v_1, \dots, v_n\}$, we see that $\{v_i \wedge v_j \mid 1 \leq i < j \leq n\}$ is a basis for $V \wedge V$, and hence $\dim(V \wedge V) = \binom{n}{2} = \frac{1}{2}n(n-1)$. This construction can be iterated to more than two factors. If $k > 0$ is an integer, we can consider the subspace Y of $V^{\otimes k}$ generated by all expressions of the form $v_1 \otimes \dots \otimes v_k$ in which $v_i = v_j$ for some $i \neq j$. A complement of Y is denoted by $\wedge^k V$ and is called the **k th exterior power** of V . If V has finite dimension n , then $\dim(\wedge^k V) = \binom{n}{k}$. In particular, we note that $\wedge^k V$ is trivial when $k > n$. The subspace $\wedge(V) = \coprod_{k=0}^n (\wedge^k V)$ of $T(V)$ is known as the **exterior algebra** of V , and has important applications in geometry and cohomology theory. One can show that if (K, \bullet) is a unital F -algebra and if $\alpha : V \rightarrow K$ is a linear transformation satisfying the condition that $\alpha(v) \bullet \alpha(v) = 0_K$ for all $v \in V$, then α can be uniquely extended to a homomorphism of unital F -algebras from $\wedge(V)$ to K .

Exercises

Exercise 1044 Find a real number a such that the matrices

$$\begin{bmatrix} 1 & a & a \\ 0 & 1 & a \\ 0 & 0 & 1 \end{bmatrix} \quad \text{and} \quad \begin{bmatrix} 1 & -1 & -1 \\ -1 & 1 & -1 \\ -1 & -1 & 1 \end{bmatrix}$$

define the same bilinear form in $\text{Bil}(\mathbb{R}^3, \mathbb{R}^3)$.

Exercise 1045 Let $V = \mathbb{Q}^{\mathbb{Q}}$. Is the function from $V \times V$ to \mathbb{Q} given by $(f, g) \mapsto (f + g)(\frac{1}{2})(f - g)(2)$ a bilinear form?

Exercise 1046 Let F be a field and let $u : \mathbb{N} \times \mathbb{N} \rightarrow F$ be an arbitrary function. Is the function $f_u : F[X] \times F[X] \rightarrow F[X]$ defined by

$$f_u : \left(\sum_{i=0}^{\infty} a_i X^i, \sum_{j=0}^{\infty} b_j X^j \right) \mapsto \sum_{k=0}^{\infty} \left(\sum_{i+j=k} u(i, j) a_i b_j X^k \right)$$

a bilinear transformation?

Exercise 1047 Let B be the canonical basis for the vector space $V = \mathbb{R}^2$. Find a bilinear form $f \in \text{Bil}(V \times V)$ satisfying the condition

$$T_{BB}(f) = \begin{bmatrix} 2 & 2 \\ 4 & -1 \end{bmatrix}.$$

Exercise 1048 Let B be the canonical basis for \mathbb{R}^3 . Find $T_{BB}(f)$, where $f : \mathbb{R}^3 \times \mathbb{R}^3 \rightarrow \mathbb{R}$ is the bilinear form defined by

$$f : \left(\begin{bmatrix} a \\ b \\ c \end{bmatrix}, \begin{bmatrix} a' \\ b' \\ c' \end{bmatrix} \right) \mapsto aa' + 2bc' + cc' + 2cb' - ab' + bb' - ba'.$$

Exercise 1049 Let $f : \mathbb{R}^3 \times \mathbb{R}^3 \rightarrow \mathbb{R}$ be the bilinear form defined by

$$f : (v, w) \mapsto v \cdot \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 2 \\ 0 & 1 & 1 \end{bmatrix} w. \text{ Find the matrix representing } f \text{ with}$$

respect to the basis $\left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} \right\}$ of \mathbb{R}^3 .

Exercise 1050 Let V and W be vector spaces over a field F and let $\alpha \in \text{Hom}(V, W)$. For each $g \in \text{Bil}(W \times W)$, let us define the bilinear form $g_\alpha \in \text{Bil}(V \times V)$ by setting $g_\alpha : (v, v') \mapsto g(\alpha(v), \alpha(v'))$. Is the function $g \mapsto g_\alpha$ a linear transformation?

Exercise 1051 Let F be a field of characteristic other than 2 and let V be a vector space over F . Let $f \in \text{Bil}(V \times V)$. Show that $f(v, v) \neq 0$ for all $0_V \neq v \in V$ if and only if for every nontrivial subspace W of V and for every $0_V \neq w \in W$ there exists a vector $w' \in W$ satisfying $f(w, w') \neq 0$.

Exercise 1052 Show that if V and W are vector spaces finitely generated over a field F of unequal dimensions, then there is no nondegenerate $f \in \text{Bil}(V, W)$.

Exercise 1053 Let F be a field of characteristic 0 and let the bilinear form $f \in \text{Bil}(F^3 \times F^3)$ be defined by

$$f : \left(\begin{bmatrix} a \\ b \\ c \end{bmatrix}, \begin{bmatrix} a' \\ b' \\ c' \end{bmatrix} \right) \mapsto aa' + bb' - cc'.$$

Is there a nontrivial subspace W of V satisfying $f(w, w') = 0$ for all $w, w' \in W$?

Exercise 1054 Let $f \in \text{Bil}(\mathbb{R}^4 \times \mathbb{R}^4)$ be defined by $f : (v, w) \mapsto v \cdot (Aw)$,

$$\text{where } A = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & -1 \end{bmatrix}. \quad \text{Find a basis } \{v_1, v_2, v_3, v_4\} \text{ of } \mathbb{R}^4$$

satisfying the condition that $f(v_i, v_i) = 0$ for all $1 \leq i \leq 4$.

Exercise 1055 Let $f : \mathcal{M}_{n \times n}(F) \times \mathcal{M}_{n \times n}(F) \rightarrow F$ be the function defined by $f : (A, B) \mapsto \text{tr}(AB)$, where F is a field and n is a positive integer. Is f a bilinear form?

Exercise 1056 Let n be a positive integer and let

$$f : \mathcal{M}_{n \times n}(\mathbb{C}) \times \mathcal{M}_{n \times n}(\mathbb{C}) \rightarrow \mathbb{C}$$

be the function defined by $f : (A, B) \mapsto n \cdot \text{tr}(AB) - \text{tr}(A)\text{tr}(B)$. Show that f is a symmetric bilinear form.

Exercise 1057 Let V be a vector space over a field F and let $f \in \text{Bil}(V, V)$ be a symmetric bilinear form. Let $Y = F \times V$ and define an operation \bullet on Y by setting $(a, v) \bullet (b, w) = (ab + f(v, w), aw + bv)$ for all $a, b \in F$ and all $v, w \in V$. Show that (Y, \bullet) is a Jordan algebra.

Exercise 1058 Let V be a vector space over \mathbb{Q} . Is the function $V \times V \rightarrow V$ defined by $(v, v') \mapsto v + v'$ a bilinear transformation?

Exercise 1059 Are the matrices $\begin{bmatrix} 0 & 0 & 0 \\ 1 & 1 & 0 \\ 1 & 1 & 1 \end{bmatrix}$ and $\frac{1}{25} \begin{bmatrix} 25 & -5 & 35 \\ 0 & -3 & 21 \\ 0 & -4 & 28 \end{bmatrix}$

in $\mathcal{M}_{3 \times 3}(\mathbb{R})$ congruent?

Exercise 1060 Find an upper-triangular matrix in $\mathcal{M}_{3 \times 3}(\mathbb{R})$ congruent

to $\begin{bmatrix} 1 & 0 & -2 \\ -1 & 1 & 0 \\ 0 & -2 & 4 \end{bmatrix}$ or show that there is no such matrix.

Exercise 1061 Let F be a field and let n be a positive integer. A matrix $A = [a_{ij}] \in \mathcal{M}_{n \times n}(F)$ is an **upper Hessenberg matrix** if and only if $a_{ij} = 0$ whenever $i - j \geq 2$. Is every matrix in $\mathcal{M}_{n \times n}(\mathbb{R})$ necessarily congruent to an upper Hessenberg matrix?

Exercise 1062 Let F be a field. Show that every upper triangular matrix in $\mathcal{M}_{3 \times 3}(F)$ is congruent to a lower triangular matrix.

Exercise 1063 Let n be a positive integer and let A be a nonsingular symmetric matrix in $\mathcal{M}_{n \times n}(\mathbb{C})$. Show that A is congruent to A^{-1} .

Exercise 1064 Find a matrix $P \in \mathcal{M}_{3 \times 3}(\mathbb{R})$ such that the matrix $P \begin{bmatrix} 2 & 1 & 3 \\ 1 & 0 & 1 \\ 3 & 1 & 3 \end{bmatrix} P^T$ is diagonal.

Exercise 1065 Let $A = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \end{bmatrix} \in \mathcal{M}_{4 \times 4}(\mathbb{R})$. Find a nonsingular matrix $P \in \mathcal{M}_{4 \times 4}(\mathbb{R})$ such that PAP^T is diagonal.

Exercise 1066 Find a diagonal matrix in $\mathcal{M}_{4 \times 4}(\mathbb{R})$ congruent to the

$$\text{matrix } \begin{bmatrix} 1 & 2 & 3 & 2 \\ 2 & 3 & 5 & 8 \\ 3 & 5 & 8 & 10 \\ 2 & 8 & 10 & -8 \end{bmatrix}.$$

Exercise 1067 Is the matrix $\begin{bmatrix} 1 & i & 1+i \\ i & 0 & 2-i \\ 1+i & 2-i & 10+2i \end{bmatrix} \in \mathcal{M}_{3 \times 3}(\mathbb{C})$ congruent to I ?

Exercise 1068 Let n be a positive integer and let α be a positive-definite endomorphism of \mathbb{R}^n represented with respect to the canonical basis by the matrix A . If A' is a matrix congruent to A , does it too represent a positive-definite endomorphism of \mathbb{R}^n with respect to the canonical basis?

Exercise 1069 Let V be a vector space finitely generated over a field F of characteristic other than 2. If $f \in \text{Bil}(V, V)$ is symmetric and not the 0-function, show that there exists a vector $v \in V$ satisfying $f(v, v) \neq 0$.

Exercise 1070 Let V be a vector space finitely generated over a field F of characteristic other than 2. If $f \in \text{Bil}(V, V)$, show that $f(v, v) = 0$ for all $v \in V$ if and only if $f(v, w) = -f(w, v)$ for all $v, w \in V$.

Exercise 1071 Let n be a positive integer, let F be a field, and let $A \in \mathcal{M}_{n \times n}(F)$. Show that there exists a symmetric matrix $B \in \mathcal{M}_{n \times n}(F)$ satisfying $v \cdot Av = v \cdot Bv$ for all $v \in F^n$.

Exercise 1072 Find a bilinear form $f \in \text{Bil}(\mathbb{R}^3, \mathbb{R}^3)$ which defines the

$$\text{quadratic form } \begin{bmatrix} a \\ b \\ c \end{bmatrix} \mapsto a^2 - 2ab + 4ac - 2bc + 2c^2.$$

Exercise 1073 Let $f \in \text{Bil}(\mathbb{R}^3, \mathbb{R}^3)$ be the symmetric bilinear form de-

$$\text{fined by the matrix } \begin{bmatrix} -3 & 1 & 0 \\ 1 & -6 & 1 \\ 0 & 1 & 7 \end{bmatrix}. \text{ Find the quadratic form defined by } f.$$

Exercise 1074 Let $f \in \text{Bil}(\mathbb{R}^3, \mathbb{R}^3)$ be the symmetric bilinear form de-

$$\text{fined by the matrix } \begin{bmatrix} 2 & -1 & 5 \\ -1 & 0 & \frac{1}{3} \\ 5 & \frac{1}{3} & -3 \end{bmatrix}. \text{ Find the quadratic form defined by } f.$$

Exercise 1075 Find a symmetric bilinear form $f \in \text{Bil}(\mathbb{R}^3, \mathbb{R}^3)$ which

$$\text{defines the quadratic form } \begin{bmatrix} a \\ b \\ c \end{bmatrix} \mapsto 2ab + 4ac + 6bc.$$

Exercise 1076 Let F be a field of characteristic other than 2, and let V be a vector space over F . Let $q : V \rightarrow F$ be a function satisfying the condition that $q(v+w) + q(v-w) = 2q(v) + 2q(w)$ for all $v, w \in V$. Show that the function $f : V \times V \rightarrow F$ defined by

$$f : (v, w) \mapsto \frac{1}{4} [q(v+w) - q(v-w)]$$

is a symmetric bilinear form.

Exercise 1077 Let V be a vector space over a field F of characteristic other than 2, and let $f \in \text{Bil}(V, V)$ be a symmetric bilinear form which defines a quadratic form $q : V \rightarrow F$. Show that

$$q(u+v+w) = q(u+v) + q(u+w) + q(v+w) - q(u) - q(v) - q(w)$$

for all $u, v, w \in V$.

Exercise 1078 Find the numerical range of the quadratic form $q : \mathbb{R}^2 \rightarrow$

$$\mathbb{R} \text{ defined by } q : v \mapsto v^T \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} v.$$

Exercise 1079 Let V be a vector space over a field F . Show that $V \cong F \otimes V$.

Exercise 1080 Let V and W be vector spaces over a field F . Let $x \in V \otimes W$ be written in the form $x = \sum_{i=1}^n v_i \otimes w_i$, where n is minimal in the sense that there is no way to express x in the form $\sum_{i=1}^k v'_i \otimes w'_i$ for any $k < n$. Show that $\{v_1, \dots, v_n\}$ is a linearly-independent subset of V and that $\{w_1, \dots, w_n\}$ is a linearly-independent subset of W .

Exercise 1081 Let K be a field containing F as a subfield. If V is a vector space over F , show that $K \otimes V$ is a vector space over K .

Exercise 1082 Let V be a vector space of finite dimension n over a field F and let Y be the subspace of $V \otimes V$ generated by all elements of the form $v \otimes v' - v' \otimes v$, where $v, v' \in V$. Find the dimension of Y .

Exercise 1083 Let V and W be finite dimensional vector spaces over a field F . Let $v, v' \in V$ and $w, w' \in W$ be vectors satisfying the condition $v \otimes w = v' \otimes w'$ and this is not the identity element of $V \otimes W$ with respect to addition. Show that there exists a scalar $c \in F$ such that $v = cv'$ and $w' = cw$.

Exercise 1084 Let F be a field and, for all $A, B \in \mathcal{M}_{2 \times 2}(F)$, denote the Kronecker product of A and B by $A \otimes B$. If $\{H_1, \dots, H_4\}$ is the canonical basis for $\mathcal{M}_{2 \times 2}(F)$, is $\{H_i \otimes H_j \mid 1 \leq i, j \leq 4\}$ a basis for $\mathcal{M}_{4 \times 4}(F)$.

Exercise 1085 Let n be a positive integer and let F be a field. If $A \in \mathcal{M}_{n \times n}(F)$ is a magic matrix, is the same true for $A \otimes A \in \mathcal{M}_{2n \times 2n}(F)$?

Exercise 1086 Let F be a field and let k and n be positive integers. If matrices $A \in \mathcal{M}_{k \times k}(F)$ and $B \in \mathcal{M}_{n \times n}(F)$ have eigenvalues a and b respectively, show that ab is an eigenvalue of $A \otimes B$.

Exercise 1087 Let F be a field and let k and n be positive integers. If matrices $A \in \mathcal{M}_{k \times k}(F)$ and $B \in \mathcal{M}_{n \times n}(F)$ have eigenvalues a and b respectively, find a matrix $C \in \mathcal{M}_{kn \times kn}(F)$ with eigenvalue $a + b$.

Exercise 1088 Let F be a field of characteristic other than 2 and let V be a vector space over F . Find the minimal polynomial of the endomorphism α of $V \otimes V$ defined by

$$\alpha : \sum_{i=1}^n a_i(v_i \otimes w_i) \mapsto \sum_{i=1}^n a_i(w_i \otimes v_i).$$

Exercise 1089 Let F be a field, let k, n, s , and t be positive integers, and consider matrices $A \in \mathcal{M}_{k \times n}(F)$ and $B \in \mathcal{M}_{s \times t}(F)$. Is the rank of $A \otimes B$ necessarily equal to the product of the ranks of A and B ?

Exercise 1090 Let V, V', W, W' be vector spaces over a field F and let $\alpha : V \rightarrow V'$ and $\beta : W \rightarrow W'$ be monic linear transformations. Let $\alpha \otimes \beta$ be the linear transformation from $V \otimes V'$ to $W \otimes W'$

$$\alpha \otimes \beta : \sum_{i=1}^n a_i(v_i \otimes v'_i) \mapsto \sum_{i=1}^n a_i[\alpha(v_i) \otimes \beta(v'_i)].$$

Is $\alpha \otimes \beta$ necessarily monic?

Exercise 1091 Let F be a field and let (K, \bullet) and $(L, *)$ be F -algebras. Define an operation \diamond on $V \otimes W$ by setting

$$(v \otimes w) \diamond (v' \otimes w') = (v \bullet v') \otimes (w * w').$$

for all $v, v' \in K$ and $w, w' \in L$. Is $(K \otimes L, \diamond)$ an F -algebra?

Exercise 1092 Let $V = \mathbb{R}^2$ and let $W = V \otimes V$. If $w \in W$ is normal, do there necessarily exist normal vectors $v, v' \in V$ such that $w = v \otimes v'$?

Exercise 1093 Let V be an inner product space over \mathbb{R} having a basis $\{v_i \mid i \in \Omega\}$ and let W be an inner product space over \mathbb{R} having a basis $\{w_j \mid j \in \Lambda\}$. Define a function $\mu : (V \otimes W) \times (V \otimes W) \rightarrow \mathbb{R}$ by setting

$$\begin{aligned} \mu : \left(\sum_{i \in \Omega} \sum_{j \in \Lambda} a_{ij}(v_i \otimes w_j), \sum_{i \in \Omega} \sum_{j \in \Lambda} b_{ij}(v'_i \otimes w'_j) \right) \mapsto \\ \sum_{i \in \Omega} \sum_{j \in \Lambda} a_{ij}b_{ij} [\langle v_i, v'_i \rangle + \langle w_j, w'_j \rangle]. \end{aligned}$$

Is μ an inner product on $V \otimes W$?

Exercise 1094 Let V be a vector space over a field F and let $\alpha \in \text{End}(V)$. Is the function $V \wedge V \rightarrow V \wedge V$ defined by

$$\sum_{i=1}^n c_i(v_i \wedge w_i) \mapsto \sum_{i=1}^n c_i(\alpha(v_i) \wedge \alpha(w_i))$$

a linear transformation?

Appendix A

Summary of Notation

$\operatorname{Re}(z)$, 7
 $\operatorname{im}(z)$, 7
 \overline{z} , 8
 $\mathbb{Q}(\sqrt{p})$, 9
 $\mathbb{Z}/(p)$, 9
 $GF(p)$, 9
 $\prod_{i \in \Omega} V_i$, 19
 $\mathcal{M}_{k \times n}(V)$, 20
 V^Ω , 19
 χ_B , 21
 $\coprod_{i \in \Omega} V_i$, 22
 $V^{(\Omega)}$, 22
 Fv , 23
 FD , 24
 $\sum_{i \in \Omega} W_i$, 27
 K^- , 36
 $v \times w$, 36
 K^+ , 37
 $\deg(f)$, 38
 $F[X]$, 38
 $F[g(X)]$, 38
 $\Phi_q(X)$, 42

$\mu(d)$, 42
 $F[X_1, \dots, X_n]$, 44
 $\dim(V)$, 64
 $U \oplus W$, 66
 $\bigoplus_{i \in \Omega} W_i$, 67
 $gr(f)$, 81
 $Hom(V, W)$, 82
 $\ker(\alpha)$, 84
 $\text{im}(\alpha)$, 85
 A^T , 85
 $Aff(V, W)$, 87
 \cong , 87
 $\text{rk}(\alpha)$, 89
 $\text{null}(\alpha)$, 89
 V/W , 98
 $End(V)$, 99
 σ_c , 99
 $Aut(V)$, 101
 ε_{hk} , 102
 $\varepsilon_{h;c}$, 102
 $\varepsilon_{hk;c}$, 102
 Φ_{BD} , 118
 $v \odot w$, 121
 $v \wedge w$, 121
 E_{hk} , 137
 $E_{h;c}$, 137
 $E_{hk;c}$, 137
 S_n , 202
 $sgn(\pi)$, 203
 $|A|$, 203
 $\text{adj}(A)$, 213
 $\text{spec}(\alpha)$, 229
 $\rho(A)$, 233
 $\text{comp}(p)$, 240
 $A \sim B$, 241
 $\text{Ann}(v)$, 247
 $m_A(X)$, 248
 $F[\alpha]v_0$, 267
 $LR(V)$, 268
 $D(V)$, 285
 $\text{tr}(A)$, 286
 $\langle v, w \rangle$, 300
 $v \cdot w$, 300
 D^H , 301
 $\|v\|$, 305

e^A , 316
 $\cos(A)$, 15
 $\sin(A)$, 15
 $v \perp w$, 326
 W^\perp , 330
 α^* , 339
 R_α , 354
 $SO(n)$, 375
 $\sqrt{\alpha}$, 381
 α^+ , 390
 $Bil(V \times W, Y)$, 399
 f^{op} , 399
 $Bil(V \times W)$, 401
 $T_{BD}(f)$, 402
 $v \perp_f w$, 403
 A^{\perp_f} , 403
 $q(v)$, 407
 $V \otimes W$, 410
 $v \otimes w$, 410
 $A \otimes B$, 413
 $V \wedge V$, 416
 $V^{\otimes k}$, 415
 $\wedge^k V$, 416
 $\wedge(V)$, 416

Index

- addition
 - in a field, 6
 - vector, 17
- adjacency matrix, 353
- adjoint matrix, 213
- adjoint transformation, 339
- affine subset, 86
- affine transformation, 86
- algebra, 33
 - anticommutative, 34
 - associative, 33
 - Cayley, 302
 - commutative, 34
 - division, 45
 - entire, 38
 - exterior, 416
 - Jordan, 37
 - Lie, 35
 - optimization, 10
 - quaternion, 58
 - tensor, 415
 - unital, 33
- algebraic, 65
- algebraic multiplicity, 244
- algebraically closed, 43
- alphabet, 20
- angle, 325
- annihilate, 247
- anti-Hermitian matrix, 366
- anticommutative algebra, 34
- antitrace, 296
- Apollonius' identity, 323
- Arnoldi process, 331
- associative algebra, 33
- automorphism, 101
 - elementary, 102
 - of algebras, 101
 - unitary, 369
- Axiom of Choice, 3
- band matrix, 133
- basis, 55
 - canonical, 57, 271
 - dual, 290
 - Hamel, 61
- Bernstein function, 94
- Bessel's identity, 336
- best approximation, 395
- bijective, 2
- bilinear form, 401

- nondegenerate, 401
- bilinear transformation, 399
 - symmetric, 399
- Binet-Cauchy identity, 322
- binomial formula, 16
- block form, 123
- bounded, 60, 307
- bounded support, 347
- bra-ket product, 121
- Brauer's Theorem, 314
- canonical basis, 57, 271
- Cantor set, 61
- cartesian product, 2
- Cassini oval, 314
- Cauchy-Schwarz-Bunyakovsky Theorem, 304
- Cayley algebra, 302
- Cayley-Hamilton Theorem, 250
- chain, 53
- chain subset, 53
- change-of-basis matrix, 146
- characteristic, 10
- characteristic function, 21
- characteristic polynomial, 239, 268
- characteristic value, 229
- characteristic vector, 229
- Chebyshev polynomial, 327
- Cholesky decomposition, 360
- circulant matrix, 156
- co-independent hyperplanes, 298
- coefficient, 38
 - Fourier, 333
 - leading, 38
 - Taylor, 104
- coefficient matrix, 171
- column equivalent, 145
- column space, 179
- combination
 - linear, 24
- commutative algebra, 34
- commuting pair, 36
- companion matrix, 240
- complement, 68
 - orthogonal, 330
- completely reducible, 43
- complex conjugate, 8
- complex number, 7
- condition number, 190
 - spectral, 383
- congruent matrices, 403
- conjugate transpose, 301
- conjugate,
 - complex, 8
- continuant, 221
- convex subset, 365
- coproduct
 - direct, 22
- Courant-Fischer Minimax Theorem, 354
- Cramer's Theorem, 215
- cross product, 36
- Crout's algorithm, 153
- cyclic, 104
- cyclotomic polynomial, 42
- decomposition
 - Cholesky, 360
 - direct sum, 67
 - LU, 153
 - polar, 382
 - QR, 335
 - singular value, 382
 - spectral, 379
- defective eigenvalue, 244
- degree
 - of a generalized eigenvector, 275
 - of a polynomial, 38
 - of a polynomial function, 41
- derivation, 100
- derogatory endomorphism, 244
- determinant, 205
- determinant function, 199
- diagonal matrix, 132
- diagonalizable endomorphism, 236
- difference set, 2
- differential, 80
- differential operator, 100
- dimension, 63

- Dirac functional, 289
- direct coproduct, 22
- direct product, 19
- direct sum, 66
- direct sum decomposition, 67
- discrete cosine transform, 137
- discrete Fourier transform, 136, 308
- disjoint subspaces, 23
- distance, 315
- distribution, 286
- division algebra, 45
- Division Algorithm, 40
- domain, 2
 - integral, 10
- dot product, 300
 - weighted, 301
- dual basis, 290
- dual space, 285
 - weak, 290
- dyadic product, 130
- eigenspace, 231, 233
 - generalized, 276
- eigenvalue, 229, 232
 - defective, 244
 - semisimple, 244
 - simple, 244
- eigenvector, 229, 232
 - generalized, 275
- Eisenstein's criterion, 42
- elementary automorphism, 102
- elementary matrix, 137
- elementary operation, 143
- endomorphism, 99
 - bounded, 112
 - derogatory, 244
 - diagonalizable, 236
 - nilpotent, 272
 - normal, 375
 - orthogonally diagonalizable, 352
 - positive definite, 356
 - selfadjoint, 349
- entangled tensors, 410
- entire, 38
- epic, 2
- epimorphism, 85
- equal functions, 2
- equivalence relation, 106
- equivalent matrices, 145
- euclidean
 - norm, 305
 - subfield, 299
- evaluation functional, 291
- even function, 69
- even permutation, 203
- Exchange Property, 26
- extended coefficient matrix, 171
- exterior algebra, 416
- exterior power, 416
- exterior product, 121
- exterior square, 416
- factor space, 98
- fan, 168
- fast Fourier transform, 136
- Fibonacci sequence, 269
- field, 6
 - algebraically closed, 43
 - formally-real, 300
 - Galois, 9
 - orderable, 16
- field of algebraic numbers, 65
- finite dimensional, 63
- finitely generated, 26
- fixed point, 231
- fixed space, 231
- form
 - bilinear, 401
 - Lorentz, 408
 - quadratic, 407
- formal differentiation, 104
- formally real field, 300
- Fourier coefficient, 333
- Fredholm alternative, 293
- Frobenius norm, 311
- full pivoting, 152
- function, 2
 - characteristic, 21
 - determinant, 199

- even, 69
- inverse, 2
- odd, 69
- periodic, 62
- piecewise constant, 30
- spline, 48
- functional
 - Dirac, 289
 - evaluation, 291
 - linear, 285
 - zero, 285
- Fundamental Theorem of Algebra, 43
- Galois field, 9
- Gauss-Jordan method, 175
- Gauss-Seidel iteration method, 186
- Gaussian elimination, 175
- general Lie algebra, 131
- general quadratic equation, 410
- generalized eigenspace, 276
- generalized eigenvector, 275
- generating set, 25
- geometric multiplicity, 244
- Gershgorin bound, 312
- Gershgorin's Theorem, 312
- Givens rotation matrix, 370
- GMRES algorithm, 331
- golden ratio, 269
- Google matrix, 254
- Gram matrix, 303
- Gram-Schmidt process, 328
- Gram-Schmidt Theorem, 328
- graph, 81
- Grassmann's Theorem, 69
- Greville's method, 392
- group of automorphisms, 107
- Guttman's Theorem, 138
- Haar wavelet, 332
- Hadamard inequality, 329
- Hadamard matrix, 207
- Hadamard product, 157
- Hamel basis, 61
- Hamming norm, 315
- Hankel matrix, 219
- Hausdorff Maximum Principle, 60
- Hermitian matrix, 351
- Hermitian transpose, 301
- Hilbert subset, 332
- Hilbert matrix, 144
- Hilbert-Schmidt norm, 311
- homogeneous system of linear equations, 170
- homomorphism, 79
 - of algebras, 79
 - of unital algebras, 79
- Householder matrix, 373
- hyperplane, 292
- ill-conditioned, 189
- image, 85
- imaginary part, 7
- improper subspace, 22
- independent subspaces, 67
- indeterminate, 38
- index of nilpotence, 272
- induced norm, 310
- infinite dimensional, 63
- inner product, 299
- inner product space, 300
- integral domain, 10
- interior product, 121
- interpolation problem, 169
- intersection, 2
- invariant subspace, 103
- inverse function, 2
- inversion, 202
- involution, 341
- irreducible, 41
- isometry, 362
- isomorphic vector spaces, 87
- isomorphism, 85
 - of algebras, 85
 - of unital algebras, 85
- iteration method
 - Gauss-Seidel, 186
 - Jacobi, 186
- Jacobi identity, 36

- Jacobi iteration method, 186
- Jacobi overrelaxation method, 188
- Jacobi polynomial, 327
- JOR, 188
- Jordan algebra, 37
- Jordan canonical form, 275, 278
- Jordan identity, 37
- Jordan product, 37

- Karatsuba's algorithm, 39
- kernel, 84
- ket-bra product, 121
- Kovarik algorithm, 393
- Kronecker product, 413
- Krylov algorithm, 271
- Krylov subspace, 267

- Lagrange identity, 306
- Lagrange interpolation polynomial, 146
- Lanczos algorithm, 271
- leading coefficient, 38
- leading entry, 174
- least squares method, 395
- Legendre polynomial, 326
- Lie algebra, 35
 - general, 131
 - special, 286
- Lie product, 36
- lightcone inequality, 409
- linear combination, 24
- linear functional, 285
- linear transformation, 79
 - adjoint of, 339
- linearly dependent, 49
 - locally, 82
- linearly independent, 49
- linearly recurrent sequence, 268
- list, 1
- locally linearly dependent, 82
- Loewner partial order, 357
- Lorentz form, 408
- lower-triangular matrix, 134
- LU-decomposition, 153

- Möbius function, 42

- magic matrix, 259
- Markov matrix, 135
- matrices
 - column-equivalent, 145
 - congruent, 403
 - equivalent, 145
 - row-equivalent, 145
 - similar, 241
 - unitarily similar, 370
- matrix, 20
 - adjacency, 353
 - adjoint, 213
 - anti-Hermitian, 366
 - band, 133
 - change-of-basis, 146
 - circulant, 156
 - coefficient, 171
 - companion, 240
 - determinant of, 205
 - diagonal, 132
 - elementary, 137
 - extended coefficient, 171
 - Givens rotation, 370
 - Google, 254
 - Gram, 303
 - Hadamard, 207
 - Hankel, 219
 - Hermitian, 351
 - Hilbert, 144
 - Householder, 373
 - in block form, 123
 - in reduced row echelon form, 174
 - in row echelon form, 173
 - lower-triangular, 134
 - magic, 259
 - Markov, 135
 - Nievergelt's, 144
 - nonsingular, 135
 - normal, 384
 - orthogonal, 372
 - permutation, 142
 - quasidefinite, 366
 - scalar, 132
 - singular, 135

- skew-symmetric, 134
- sparse, 187
- special orthogonal, 375
- stochastic, 135
- symmetric, 134
- symmetric Toeplitz, 183
- transpose, 85
- tridiagonal, 133
- unitary, 370
- upper Hessenberg, 419
- upper-triangular, 133
- Vandermonde, 147
- zero, 21
- maximal, 53
- maximal subspace, 292
- minimal, 53
- minimal polynomial, 248, 268
- Minkowski's inequality, 307
- minor, 209
- Modular Law, 28
- monic
 - function, 2
 - polynomial, 38
- monomorphism, 85
- Moore-Penrose pseudoinverse, 389
- multiplication
 - in a field, 6
 - scalar, 17
- multiplication table, 58
- multiplicity
 - algebraic, 244
 - geometric, 244
- mutually orthogonal, 326
- Nievergelt's matrix, 144
- nilpotent, 272
- nondegenerate bilinear form, 401
- nonhomogeneous system of linear
 - equations, 170
- nonsingular matrix, 135
- nontrivial subspace, 22
- norm, 305, 309
 - euclidean, 305
 - Frobenius, 311
 - Hamming, 315
 - Hilbert-Schmidt, 311
 - induced, 310
 - spectral, 311
- normal endomorphism, 375
- normal matrix, 384
- normal vector, 305
- normed space, 309
- nullity, 89
- number
 - complex, 7
 - rational, 5
 - real, 5
- numerical range, 409
- odd function, 69
- odd permutation, 203
- operation
 - elementary, 143
- opposite transformation, 399
- optimization algebra, 10
- order of recurrence, 268
- orderable field, 16
- orthogonal, 326
- orthogonal complement, 330
 - right, 403
- orthogonal matrix, 372
- orthogonal projection, 330
- orthogonality with respect to a bi-
 - linear form, 403
- orthogonally diagonalizable, 352
- orthonormal, 331
- Padé approximant, 216
- pairwise disjoint, 24
- Parallelogram law, 307
- parity, 401
- Parseval's identity, 336
- partial order, 53
 - Loewner, 357
- partial pivoting, 152
- partially-ordered set, 53
- periodic function, 62
- permutation, 2
 - even, 203
 - odd, 203

- permutation matrix, 142
- Pfaffian, 206
- piecewise constant, 30
- pivot, 152
- pivoting
 - full, 152
 - partial, 152
- Poincaré-Birkhoff-Witt Theorem, 36
- polar decomposition, 382
- polynomial, 38
 - characteristic, 239, 268
 - Chebyshev, 327
 - completely reducible, 43
 - cyclotomic, 42
 - in several indeterminates, 44
 - irreducible, 41
 - Jacobi, 327
 - Lagrange interpolation, 146
 - Legendre, 326
 - minimal, 248, 268
 - monic, 38
 - reducible, 41
 - zero, 38
- polynomial function, 40
- positive definite, 356
- power
 - exterior, 416
- pre-Banach space, 309
- pre-Hilbert space, 300
- primitive root of unity, 136
- process
 - Arnoldi, 331
 - Gram-Schmidt, 328
- product
 - bra-ket, 121
 - cartesian, 2
 - cross, 36
 - direct, 19
 - dot, 300
 - dyadic, 130
 - exterior, 121
 - Hadamard, 157
 - inner, 299
 - interior, 121
 - Jordan, 37
 - ket-bra, 121
 - Kronecker, 413
 - Lie, 36
 - scalar triple, 306
 - Schur, 157
 - tensor, 411
 - vector triple, 306
- projection, 104
 - onto an affine set, 342
 - orthogonal, 330
- proper subspace, 22
- pseudoinverse
 - Moore-Penrose, 389
- QR-decomposition, 335
- quadratic form, 407
- quadratic surface, 410
- quasidefinite matrix, 366
- quaternion
 - algebra, 58
 - real, 58
- range, 2
 - numerical, 409
- rank, 89, 179, 403
- Rational Decomposition Theorem, 274
- rational number, 5
- Rayleigh quotient
 - function, 354
 - iteration scheme, 355
- real euclidean, 299
- real number, 5
- real part, 7
- real quaternion, 58
- reduced row echelon form, 174
- reducible, 41
- relation
 - equivalence, 106
 - partial order, 53
- relaxation method, 188
- restriction, 2
- Riesz Representation Theorem, 338
- right orthogonal complement, 403

- row echelon form, 173
- row equivalent, 145
- row space, 179
- scalar, 18
- scalar matrix, 132
- scalar multiplication, 17
- scalar triple product, 306
- Schur product, 157
- Schur's Theorem, 371
- selfadjoint, 349
- semifield, 10
- semisimple eigenvalue, 244
- sequence, 2
 - Fibonacci, 269
 - linearly recurrent, 268
- set
 - difference, 2
 - generating, 25
 - partially-ordered, 53
 - spanning, 25
- Sherman-Morrison-Woodbury Theorem, 138
- signum, 203
- similar matrices, 241
- simple eigenvalue, 244
- simple tensors, 410
- singular matrix, 135
- singular value, 383
- Singular Value Decomposition Theorem, 382
- skew symmetric matrix, 134
- solution set, 171
- solution space, 171
- SOR, 188
- space
 - dual, 285
 - inner product, 300
 - normed, 309
 - pre-Banach, 309
 - pre-Hilbert, 300
 - solution, 171
- spanning set, 25
- sparse matrix, 187
- special Lie algebra, 286
- special orthogonal matrix, 375
- spectral condition number, 383
- Spectral Decomposition Theorem, 379
- spectral norm, 311
- spectral radius, 233
- spectrum, 229
- spline function, 48
- Steinitz Replacement Property, 52
- stochastic matrix, 135
- Strassen-Winograd algorithm, 150
- subalgebra, 35
 - unital, 35
- subfield, 7
 - euclidean, 299
 - real euclidean, 299
- subset
 - affine, 86
 - bounded, 60
 - chain, 53
 - convex, 365
 - Hilbert, 332
 - orthonormal, 331
 - underlying, 2
- subspace, 22
 - cycle, 104
 - generated by, 25
 - improper, 22
 - invariant, 103
 - Krylov, 267
 - maximal, 292
 - nontrivial, 22
 - proper, 22
 - spanned by, 25
 - trivial, 22
- subspaces
 - disjoint, 23
 - independent, 67
 - pairwise disjoint, 24
- successive overrelaxation method, 188
- Sylvester's Theorem, 89
- symmetric bilinear transformation, 399
- symmetric difference, 21

- symmetric matrix, 134
- symmetric Toeplitz matrix, 183
- system of linear equations, 169
 - homogeneous, 170
 - nonhomogeneous, 170
- Taylor coefficient, 104
- tensor algebra, 415
- tensor product, 411
- tensors
 - entangled, 410
 - simple, 410
- trace, 286
- transcendental, 65
- transform
 - discrete cosine, 137
 - discrete Fourier, 136, 308
 - fast Fourier, 136
- transformation
 - affine, 86
 - bilinear, 399
 - linear, 79
 - opposite, 399
- transpose, 85
 - conjugate, 301
 - Hermitian, 301
- Triangle difference inequality, 307
- Triangle inequality, 315
- tridiagonal matrix, 133
- trivial subspace, 22
- underlying subset, 2
- union, 2
- unit, 34
- unital algebra, 33
- unital subalgebra, 35
- unitarily similar matrices, 370
- unitary automorphism, 369
- unitary matrix, 370
- upper Hessenberg matrix, 419
- upper-triangular matrix, 133
- Vandermonde matrix, 147
- vector, 18
 - normal, 305
- vector addition, 17
- vector space, 18
 - finite dimensional, 63
 - finitely generated, 26
 - infinite dimensional, 63
- vector triple product, 306
- vectors
 - orthogonal, 326
 - orthogonal with respect to a bilinear form, 403
- wavelet
 - Haar, 332
- weak dual space, 290
- weighted dot product, 301
- Well Ordering Principle, 53
- width of a band matrix, 133
- word, 20
- Wronskian, 207
- Zlobec's formula, 393
- Zorn's Lemma, 60