# 1 Vector Spaces

# 1.1 Fields and Vector Spaces

### Definition 1.1.1: Definition of a field

A field F is a set with two functions

• Addition:  $+: F \times F \to F$ ,  $(\lambda, \mu) \mapsto \lambda + \mu$ 

• Multiplication:  $\cdot: F \times F$ ,  $(\lambda, \mu) \mapsto \lambda \mu$ 

which satisfy the following axioms:

1. (F, +) is an abelian group  $F^+$ , with identity  $0_F$ 

2.  $(F \setminus \{0_F\}, \cdot)$  is an abelian group  $F^{\times}$ , with identity  $1_F$ 

3. **Distributive law**: For all a, b, and c in F, we have

$$a(b+c) = ab + ac \in F$$

and the following lemmas:

1. The elements  $0_F$  and  $1_F$  of F are distinct

2. For all  $a \in F$ ,  $a \cdot 0_F = 0_F$  and  $0_F \cdot a = 0_F$ 

3. Multiplication in  ${\cal F}$  is associative, and  $1_{\cal F}$  is an identity element

A vector space V over a field F is a pair consisting of an abelian group  $V=(V,\dot+)$  and a mapping

$$F \times V \to V : (\lambda, \vec{v}) \mapsto \lambda \vec{v}$$

s.t. for all  $\lambda, \mu \in F$  and  $\vec{v}, \vec{w} \in V$  the following identities hold:

• Distributivity 1:  $\lambda(\vec{v} + \vec{w}) = \lambda \vec{v} + \lambda \vec{w}$ 

• Distributivity 2:  $(\lambda + \mu)\vec{v} = \lambda \vec{v} + \mu \vec{v}$ 

• Associativity:  $\lambda(\mu \vec{v}) = (\lambda \mu) \vec{v}$ 

• Identity:  $1\vec{v} = \vec{v}$ 

and so do the following lemmas:

1. If V is a vector space and  $\vec{v} \in V$ , then  $0\vec{v} = \vec{0}$ 

2. If V is a vector space and  $\vec{v} \in V$ , then  $(-1)\vec{v} = -\vec{v}$ 

3. If V is a vector space over a field F, then  $\lambda \vec{0} = \vec{0}$  for all  $\lambda \in F$ . Furthermore, if  $\lambda \vec{v} = \vec{0}$  then either  $\lambda = 0$  or  $\vec{v} = \vec{0}$ 

# 1.2 Working with Vector Spaces

### Definition 1.2.1: Cartesian Product of n sets

$$X_1 \times \cdots \times X_n := \{(x_1, \dots, x_n) : x_i \in X_i \text{ for } 1 \le i \le n\}$$

The elements of a product are called *n*-tuples. An individual entry  $x_i = (x_1, \ldots, x_n)$  is called a **component**.

There are special mappings called **projections** for a cartesian product:

$$\operatorname{pr}_i: X_1 \times \dots \times X_n \to X_i$$
  
 $(x_1, \dots, x_n) \mapsto x_i$ 

The cartesian product of n copies of a set X is written in short as:  $X^n$ 

#### Definition 1.2.2: Vector Subspace

A subset U of a vector space V is called a **vector subspace** or **subspace** if U contains the zero vector, and whenever  $\vec{u}, \vec{v} \in U$  and  $\lambda \in F$  we have  $\vec{u} + \vec{v} \in U$  and  $\lambda \vec{u} \in U$ 

### Definition 1.2.3: Spans and Linear Independence

Let  $T\subset V$  for some vector space V over a field F. Then amongus all subspaces of V that include T there is a smallest subspace

$$\langle T \rangle = \langle T \rangle_F \subseteq V$$

"the set of all vectors  $\alpha_1 \vec{v}_1 + \cdots + \alpha_r \vec{v}_r$  with  $\alpha_1, \ldots, \alpha_r \in F$  and  $\vec{v}_1, \ldots, \vec{v}_r \in T$ , together with the zero vector in the case  $T = \emptyset$ "

# Terminology Dump

- An expression of the form  $\alpha_1 \vec{v}_1 + \cdots + \alpha_r \vec{v}_r$  is called a **linear** combination of vectors  $\vec{v}_1, \dots, \vec{v}_r$
- The smallest vector subspace  $\langle T \rangle \subseteq V$  containing T is called the **vector subspace generated by** T or the vector subspace **spanned by** T or even the **span of** T
- If we allow the zero vector to be the "empty linear combination of r=0 vectors", then the span of T is exactly the set of all linear combinations of vectors from T
- A subset of a vector space that spans the entire space is called a **generating** or **spanning set**. A vector space that has a finite generating set is said to be **finitely generated**

#### Linear Independence

A subset L of a vector space V is called **linearly independent** if for all pairwise different vectors  $\vec{v}_1, \ldots, \vec{v}_r \in L$  and arbitrary scalars  $\alpha, \ldots, \alpha_r \in F$ ,

$$a_1 \vec{v}_1 + \dots + \alpha_r \vec{v}_r = \vec{0} \implies a_1 = \dots = \alpha_r = 0$$

A subset L of a vector space V is called **linearly dependent** if it is not linearly independent (duh..). This means there exists pairwise different vectors  $\vec{v}j_1,\ldots,\vec{v}_r\in L$  and scalars  $\alpha_1,\ldots,\alpha_r\in F$ , not all zero, such that  $\alpha_1\vec{v}_1+\cdots\alpha_r\vec{v}_r=\vec{0}$ 

# 1.3 Linear Independence and Bases

# Definition 1.3.1: Basis of a Vector Space

A basis of a vector space V is a linearly independent generating set in V

# Example 1.3.2: Standard Basis

Let F be a field and  $n\in\mathbb{N}.$  We consider the following vectors in  $F^n$ 

$$\vec{e}_i = (0, \dots, 0, 1, 0, \dots, 0)$$

with one 1 in the *i*-th place and zero everywhere else. Then  $\vec{e}_1, \ldots, \vec{e}_n$  form an ordered basis of  $F^n$ , the so-called **standard** basis of  $F^n$ 

#### Theorem 1.3.3: Linear combinations of basis elements

Let F be a field, V a vector space over F and  $\vec{v}_1,\ldots,\vec{v}_r\in V$  vectors. The family  $(\vec{v}_i)_{1\leq i\leq r}$  is a basis of V if and only if the following "evaluation" mapping

$$\psi: F^r \to V$$
$$(\alpha_1, \dots, a_r) \mapsto a_1 \vec{v}_1 + \dots + \alpha_r \vec{v}_r$$

is a bijection

If we label our ordered family by  $\mathcal{A} = (\vec{v}_1, \dots, \vec{v}_r)$ , then we done the above mapping by

$$\psi = \psi_{\mathcal{A}} : F^r \to V$$

#### Theorem 1.3.4: Characterisations of Bases

The following are equivalent for a subset E of a vector space V:

- 1. E is a basis, i.e. a linearly independent generating set
- 2. E is minimal among all generating sets, meaning that  $E \setminus \{\vec{v}\}$  does not generate V, for any  $\vec{v} \in E$
- 3. E is maximal among all linearly independent subsets, meaning that  $E \cup \{\vec{v}\}$  is linearly dependent for any  $\vec{v} \in V$

Corrollary: Let V be a finitely generated vector space over a field F. Then V has a finite basis

#### Basis Characterisation Variant

- 1. If  $L \subset V$  is a linearly independent subset and E is minimal amongst all generating sets of V with the property that  $L \subseteq E$ , then E is a basis.
- 2. If  $E \subseteq V$  is a generating set and if L is maximal amongst all linearly independent sets of V with the property  $L \subseteq E$ , then L is a basis.

### Definition 1.3.5: Free Vector Space

Let X be a set and F a field. The set  $\mathrm{Maps}(X,F)$  of all mappings  $f:X\to F$  becomes an F-vector space with the operations of pointwise addition and multiplication by a scalar. The subset of all mappings which send almost all elements of X to zero is a vector subspace

$$F\langle X \rangle \subseteq \operatorname{Maps}(X, F)$$

This subspace is called the free vector space on the set X

#### Theorem 1.3.6: Variant of Linear Combinations

Let F be a field, V be an F-vector space and  $(\vec{v}_i)_{i \in I}$  a family of vectors from the vector space V. The following are equivalent:

- 1. The family  $(\vec{v}_i)_{i \in I}$  is a basis for V
- 2. For each  $\vec{v} \in V$  there is precisely one family  $(a_i)_{i \in I}$  of elements of F, almost all which are zero and such that

$$\vec{v} = \sum_{i=I} a_i \vec{v}_i$$

### 1.4 Dimension of a Vector Space

### Theorem 1.4.1: Fundamental Estimate of LinAlg

No linearly independent subset of a given vector has more elements than a generating set. Thus if V is a vector space,  $L \subset V$  a linearly independent subset and  $E \subseteq V$  a generating set, then

$$|L| \leq |E|$$

# Theorem 1.4.2: Steinitz Exchange Theorem

Let V be a vector space,  $L \subset V$  a finite linearly independent subset and  $E \subseteq V$  a generating set. Then there is an injection  $\phi: L \hookrightarrow E$  such that  $(E \setminus \phi(L)) \cup L$  is also a generating set for V

Let V be a vector space,  $M \subseteq V$  a linearly independent subset, and  $E \subseteq V$  a generating subset, such that  $M \subseteq E$ . If  $\vec{w} \in V \setminus M$  is a vector  $\not \in M$  such that  $M \cup \{\vec{w}\}$  is linearly independent, then there exists  $\vec{e} \in E \setminus M$  such that  $(E \setminus \{\vec{e}\}) \cup \{\vec{w}\}$  is a generating set

## Theorem 1.4.3: Cardinality of Bases

Let V be a finitely generated vector space.

- 1. V has a finite basis
- 2. V cannot have an infinite basis
- 3. Any two bases of V have the same number of elements

#### Definition 1.4.4: Dimension of a Vector Space

The cardinality of a basis of a finitely generated vector space V is called the **dimension** of V, written  $\dim V$ . If F is a field, and we want to denote that we mean dimension as an F-vector space, then we write  $\dim_F V$ . If the vector space is not finitely generated, then we say  $\dim V = \infty$  and call V infinite dimensional.

#### Theorem 1.4.5: Dimension Theorems

#### Cardinality Criterion for Bases

- 1. Each linearly independent subset  $L \subset V$  has at most dim V elements, and if  $|L| = \dim V$  then L is a basis
- 2. Each generating set  $E\subseteq V$  has at least dim V elements, and if  $|E|=\dim V$  then E is a basis

**Dimension Estimate for Vector Subspaces**: A proper vector subspace of a finite dimensional vector space has itself a strictly smaller dimension

If  $U\subseteq V$  is a vector subspace of an arbitrary vector space, then we have  $\dim U \le \dim V$  and if we have  $\dim U = \dim V < \infty$  then it follows that U=V

# 1.5 Linear Mappings

### Definition 1.5.1: Linear Mappings

Let V, W be vector spaces over a field F. A mapping  $f: V \to W$  is called **linear**, or F-**linear**, or even a **homomorphism of** F-**vector spaces** if for all  $\vec{v}_1, \vec{v}_2 \in V$  and  $\lambda \in F$  we have

$$f(\vec{v}_1 + \vec{v}_2) = f(\vec{v}_1) + f(\vec{v}_2)$$
$$f(\lambda \vec{v}_1) = \lambda f(\vec{v}_1)$$

A bijective linear mapping is called an **isomorphism** of vector spaces. If there is an isomorphism between two vector spaces, we call them **isomorphic**. A homomorphism  $V \to V$  is called an **endomorphism** of V. An isomorphism  $V \to V$  is called an **automorphism** of V

Two vector subspaces  $V_1, V_2$  of a vector space V are called **complementary** if addition defines a bijection

$$V_1 \times V_2 \xrightarrow{\sim} V$$

something about direct sums

### Theorem 1.5.2: Classifying VecSpaces by Dimension

Let n be a natural number. Then a vector space over a field F is isomorphic to  $F^n$  iff it has dimension n

### Theorem 1.5.3: Linear Mapping and Bases

Let V, W be vector spaces over a field F. The set of all homomorphisms from V to W is denoted by

$$\operatorname{Hom}_F(V, W) = \operatorname{Hom}(V, W) \subseteq \operatorname{Maps}(V, W)$$

Let  $B \subset V$  be a basis. Then restriction of a mapping gives a bijection

$$\operatorname{Hom}_F(V,W) \xrightarrow{\sim} \operatorname{Maps}(B,W)$$
  
 $f \mapsto f|_B$ 

# Theorem 1.5.4: Inverse Mappings

- 1. Every injective linear mapping  $f: V \hookrightarrow W$  has a **left inverse**, or a linear mapping  $g: W \to V$  s.t.  $g \circ f = \mathrm{id}_V$
- 2. Every surjective linear mapping  $f:V \twoheadrightarrow W$  has a **right** inverse, or a linear mapping  $G:W \to V$  s.t.  $f \circ g = \mathrm{id}_W$

# Definition 1.5.5: Image and Kernel of a map

The **image** of a linear mapping  $f:V\to W$  is the subset  $\operatorname{im}(f)=f(V)\subseteq W$ . It is a vector subspace of W. The preimage of the zero vector of a linear mapping  $f:V\to W$  is denoted by:

$$\ker(f) := f^{-1}(0) = \{ v \in V : f(v) = 0 \}$$

and is called the  ${\bf kernel}$  of the linear mapping f. The kernel is a subspace of V

Mini lemma: A linear mapping is injective iff its kernel is zero

# Theorem 1.5.6: Rank-Nullity / Dimension Theorem

Let  $f: V \to W$  be a linear mapping between vector spaces. Then:  $\dim V = \dim(\ker f) + \dim(\operatorname{im} f)$ 

Dimension of im f= rank of f, dimension of ker f = nullity of f

Let V be a vector space, and  $U,W\subseteq V$  vector subspaces. Then  $\dim(U+W)+\dim(U\cap W)=\dim U+\dim W$ 

# 2 Linear Mappings and Matrices

# 2.1 Linear Mappings $F^m \to F^n$ and Matrices

# Theorem 2.1.1: Linear Maps $F^m \to F^n$ and Matrices

Let F be a field and let  $m, n \in \mathbb{N}$ . There is a bijection between the space of linear mappings  $F^m \to F^n$  and the set of matrices with n rows, m columns, and entries in F:

$$M: \operatorname{Hom}_F(F^m, F^n) \xrightarrow{\sim} \operatorname{Mat}(n \times m; F)$$
  
 $f \mapsto [f]$ 

This attaches to each linear mapping f its **representing matrix** M(f) := [f]. The columns of this matrix are the images under f of the standard basis elements of  $F^m$ 

$$[f] := (f(\vec{e}_1)|f(\vec{e}_2)| | \cdots | f(\vec{e}_m))$$

# $\ \, \textbf{Definition 2.1.2: Matrix Multiplication} \\$

Let  $n, m, \ell \in \mathbb{N}$ , F a field, and let  $A \in \operatorname{Mat}(n \times m; F)$  and  $B \in \operatorname{Mat}(m \times \ell; F)$  be matrices. The **product**  $A \circ B = AB \in \operatorname{Mat}(n \times \ell; F)$  is the matrix defined by

$$(AB)_{ik} = \sum_{j=1}^{m} A_{ij} B_{jk}$$

# Theorem 2.1.3: Composition of maps to products

Let  $g:F^\ell\to F^m$  and  $f:F^m\to F^n$  be linear mappings. The representing matrix of their composition is the product of their representing matrices:

$$[f\circ g]=[f]\circ [g]$$

# Theorem 2.1.4: Calculating with Matrices

$$\bullet (A + A')B = AB + A'B$$

• 
$$AI = A$$

$$\bullet A(B + B') + AB + AB'$$

• 
$$(AB)C = A(BC)$$

## 2.2 Matrix Definitions

#### Definition 2.2.1: Big def-thm pairs

**Def**: A matrix A is called **invertible** if there exists matrices B and C such that BA = I and AC = I

#### Thm: Invertible Equivalence

- 1. There exists a square matrix B such that BA = I
- 2. There exists a square matrix C such that AC = I
- 3. The square matrix A is invertible

**Def**: An **elementary matrix** is any square matrix that differs from the identity matrix in at least one entry

Thm: Every square matrix with entries in a field can be written as a product of elementary matrices

**Def**: Any matrix whose only non-zero entries lie on the diagonal, and which has first 1's along the diagonal and then 0's, is said to be in **Smith Normal Form** 

**Thm**: For each matrix  $A \in \operatorname{Mat}(n \times m; F)$  there exist invertible matrices P and Q such that PAQ is a matrix in Smith Normal Form

**Def**: The **column rank** of a matrix  $A \in \text{Mat}(n \times m; F)$  is the dimension of the subspace of  $F^n$  generated by the columns of A. Similarly, the **row rank** of A is the dimension of the subspace of  $F^m$  generated by the rows of A.

**Thm**: The column and row rank of any matrix are equal

**Def**: Since they are both the same, "column" and "row" can be omitted for the **rank of a matrix**, written as rk A. If the rank is equal to the no. of rows/columns, then the matrix has **full rank** 

# 2.3 Abstract Linear Mappings and Matrices

### Theorem 2.3.1: Representing Matrices

Let F be a field, V and W vector spaces over F with ordered bases  $\mathcal{A} = (\vec{v}_1, \dots, \vec{v}_m)$  and  $\mathcal{B} = (\vec{w}_1, \dots, \vec{w}_n)$ . Then to each linear mapping  $f: V \to W$  we associate a **representing matrix**  $\mathcal{B}[f]\mathcal{A}$  whose entries  $a_{ij}$  are defined by the identity

$$f(\vec{v}_j) = a_{1j}\vec{w}_1 + \dots + a_{nj}\vec{w}_n \in W$$

This produces a bijection, which is even an isomorphism of vector spaces:

$$M_{\mathcal{B}}^{\mathcal{A}}: \operatorname{Hom}_{F}(V, W) \xrightarrow{\sim} \operatorname{Mat}(n \times m; F)$$
  
$$f \mapsto_{\mathcal{B}} [f]_{\mathcal{A}}$$

### Theorem 2.3.2: Repr. Mat of Compositions

Let F be a field and U, V, W finite dimensional vector spaces over kF with ordered bases  $\mathcal{A}, \mathcal{B}, \mathcal{C}$ . If  $f: U \to V$  and  $g: V \to W$  are linear mappings, then the representing matrix of the composition  $g \circ f: U \to W$  is the matrix product of the representing matrices of f and g:

$$_{\mathcal{C}}[g \circ f]_{\mathcal{A}} =_{\mathcal{C}} [g]_{\mathcal{B}} \circ_{\mathcal{B}} [f]_{\mathcal{A}}$$

#### Definition 2.3.3: Representation of a vector

Let V be a finite dimensional vector space with an ordered basis  $\mathcal{A}=(\vec{v}_1,\ldots,\vec{v}_m)$ . We'll denote the inverse to the bijection in 1.3.3 " $\Phi_{\mathcal{A}}: F^m \xrightarrow{\sim} V, (\alpha_1,\ldots,\alpha_m)^T \mapsto \alpha_1 \vec{v}_1 + \cdots + \alpha_m \vec{v}_m$ " by

$$\vec{v}\mapsto_{\mathcal{A}} [\vec{v}]$$

The column vector  $_{\mathcal{A}}[\vec{v}]$  is called the **representation of the vector**  $\vec{v}$  with respect to the basis  $\mathcal{A}$ 

Thm: Representation of the Image of a Vector: Let V, W be finite dim. vector spaces over F with ordered bases A, B and let  $f: V \to W$  be a linear mapping. The following holds for  $\vec{v} \in V$ :

$$_{\mathcal{B}}[f(\vec{v})] =_{\mathcal{B}} [f]_{\mathcal{A}} \circ_{\mathcal{A}} [\vec{v}]$$

# 3 Rings

I can't be bothered doing changes of basis and stuff, time for something more interesting :D

# 3.1 Ring basics

# Definition 3.1.1: Definition of a Ring

A **ring** is a set with two operations  $(\mathbb{R}, +, \cdot)$  that satisfy:

- 1. (R, +) is an abelian group
- 2.  $(R, \cdot)$  is a **monoid** this means that the second operation  $\cdot : R \times R \to R$  is associative and that there is an **identity element**  $1 = 1_R \in R$ , often just called the identity, with the property that  $1 \cdot a = a \cdot 1 = a$  for all  $a \in R$ .
- 3. The distributive laws hold, meaning that for all  $a, b, c \in R$ ,

$$a \cdot (b+c) = (a \cdot b) + (a \cdot c))$$

$$(a+b)\cdot c = (a\cdot c) + (b\cdot c)$$

The two operations are called **addition** and **multiplication** in our ring. A ring in which multiplication, that is  $a\cdot b=b\cdot a$  for all  $a,b\in R$ , is a **commutative ring** 

**Note:** We'll call the element  $1 \in R$  as the identity element of the monoid  $(R, \cdot)$ , and we call the additive identity of (R, +) zero, written as  $0_R$  or 0

**Example:** We can define the **null ring** or **zero ring** as a ring where R is a single ement set, e.g.  $\{0\}$ , with the operations 0 + 0 = 0 and  $0 \times 0 = 0$ . We will call any ring that isn't the zero ring a **non-zero ring** 

# Example 3.1.2: Modulo Rings

Let  $m \in \mathbb{Z}$  be an integer. Then the set of **integers modulo** m, written

$$\mathbb{Z}/m\mathbb{Z}$$

is a ring. The elements of  $\mathbb{Z}/m\mathbb{Z}$  consist of **congruence classes** of integers modulo m - that is the elements are the subsets T of  $\mathbb{Z}$  of the form  $T=a+m\mathbb{Z}$  with  $a\in\mathbb{Z}$ . Think of these as the set of integers that have the same remainder when you divide them by m. I denote the above congruence class by  $\overline{a}$ . Obviously  $\overline{a}=\overline{b}$  is the same as  $a-b\in m\mathbb{Z}$ , and often I'll write

$$a \equiv b \mod m$$

# 3.2 Linking Rings to Fields and Further Properties

# Definition 3.2.1: Ring definition of a field

A field is a non-zero commutative ring F in which every non-zero element  $a \in F$  has an inverse  $a^{-1} \in F$ , that is an element  $a^{-1}$  with the property that  $a \cdot a^{-1} = a^{-1} \cdot a = 1$ 

# Theorem 3.2.2: Prime property of fields

Let m be a positive integer. The commutative ring  $\mathbb{Z}/m\mathbb{Z}$  is a field if and only if m is prime.

## Theorem 3.2.3: Lemmas for multiplying

Let R be a ring and let  $a, b \in R$ . Then

- 1. 0a = 0 = a0
- 2. (-a)b = -(ab) = a(-b)
- 3. (-a)(-b) = ab

## Definition 3.2.4: Multiples of an abelian group

Let  $m \in \mathbb{Z}$ . The m-th multiple ma of an element ain an abelian group R is:

$$ma = \underbrace{a + a + \dots + a}_{m \text{ terms}} \quad \text{if } m > 0$$

0a = 0 and negative multiples are defined by (-m)a = -(ma)

### Theorem 3.2.5: Lemmas for multiples

Let R be a ring, let  $a, b \in R$  and let  $m, n \in \mathbb{Z}$ . Then:

- 1. m(a + b) = ma + mb
- 2. (m+n)a = ma + na
- 3. m(na) = (mn)a
- 4. m(ab) = (ma)b = a(mb)
- 5. (ma)(nb) = (mn)(ab)

#### Definition 3.2.6: Unit of a ring

Let R be a ring. An element  $a \in R$  is called a **unit** if it is *invertible* in R or in other words has a multiplicative inverse in R, meaning that there exists  $a^{-1} \in R$  such that

$$aa^{-1} = 1 = a^{-1}a$$

# Theorem 3.2.7

The set  $R^{\times}$  of units in a ring R forms a group under multiplication

#### Definition 3.2.8: zero-divisors of a ring

In a ring R, a non-zero element a is called a **zero-divisor** or **divisor of zero** if there exists a non-zero element b such that either ab = 0 or ba = 0.

#### Theorem 3.2.9: Cancellation Law

Let R be an integral domain and let  $a,b,c\in R$ . If ab=ac and  $a\neq 0$  then b=c

## Theorem 3.2.10: Prime Property for Integral Domains

Let m be a natural number. Then  $\mathbb{Z}/m\mathbb{Z}$  is an integral domain if and only if m is prime.

#### Theorem 3.2.11

Every **finite** integral domain is a field.

# 3.3 Polynomials

# Definition 3.3.1: Polynomial

Let R be a ring. A **polynomial over** R is an expression of the form

$$P = a_0 + a_1 X + a_2 X^2 + \dots + a_m X^m$$

for some non-negative integer m and elements  $a_i \in R$  for  $0 \le i \le m$ . The set of all polynomials over R is denoted by R[X]. In the case where  $a_m$  is non-zero, the polynomial P has **degree** m, (written  $\deg(P)$ ), and  $a_m$  is its **leading coefficient**. When the leading coefficient is 1 the polynomial is a **monic polynomial**. A polynomial of degree one is called **linear**, a polynomial of degree two is called **quadractic**, and a polynomial of degree three is called **quadractic**.

# Definition 3.3.2: Ring of Polynomials

The set R[X] becomes a ring called the **ring of polynomials** with coefficients in R, or over R. The zero and the identity of R[X] are the zero and identity of R, respectively.

## Theorem 3.3.3: Zero-Divisors of a Polynomial Ring

If R is a ring with no zero-divisors, then R[X] has no zero-divisors and  $\deg(PQ) = \deg(P) + \deg(Q)$  for non-zero  $P, Q \in R[X]$ .

If R is an integral domain, then so is R[X]

#### Theorem 3.3.4: Division and Remainder

Let R be an integral domain and let  $P,Q\in R[X]$  with Q monic. Then there exists unique  $A,B\in R[X]$  such that P=AQ+B and  $\deg(B)<\deg(Q)$  or B=0

#### Definition 3.3.5: Formal definition of a function

Let R be a commutative ring and  $P \in R[X]$  a polynomial. Then the polynomial P can be **evaluated** at the element  $\lambda \in R$  to produce  $P(\lambda)$  by replacing the powers of X in the polynomial P by the corresponding powers of  $\lambda$ . In this way we have a mapping

$$R[X] \to \operatorname{Maps}(R, R)$$

This is the precise mathematical description of thinking of a polynomial as a function. An element  $\lambda \in R$  is a **root** of P is  $P(\lambda) = 0$ 

# Theorem 3.3.6: Roots of a Polynomial

Let R be a commutative ring, let  $\lambda \in R$  and  $P(X) \in R[X]$ . Then  $\lambda$  is a root of P(X) if and only if  $(X - \lambda)$  divides P(X)

### Theorem 3.3.7: Degrees of Polynomial Roots

Let R be a field, or more generally an integral domain. Then a non-zero polynomial  $P\in R[X]\backslash\{0\}$  has at most  $\deg(P)$  roots in R

#### Definition 3.3.8: Algebraically closed fields

A field F is **algebraically closed** if each non-constant polynomial  $P \in F[X] \backslash F$  with coefficients in our field has a root in our field F

#### Theorem 3.3.9: Fundamental Theorem of Algebra

The field of complex numbers  $\mathbb C$  is algebraically closed.

#### Theorem 3.3.10: Linear Factors of Closed Fields

If F is an algebraically closed field, then every non-zero polynomial  $P \in F[X] \setminus \{0\}$  decomposes into linear factors

$$P = c(X - \lambda_1) \cdots (X - \lambda_n)$$

with  $n \geq 0$ ,  $c \in F^{\times}$  and  $\lambda_1, \ldots, \lambda_n \in F$ . This decomposition is unique up to reordering the factors

# 4 Determinants and Eigenvalues Redux

# 4.1 Symmetric Groups

### **Definition 4.1.1: Symmetric Groups**

The group of all permutations of the set  $\{1, 2, ..., n\}$ , also known as bijections from  $\{1, 2, ..., n\}$  to itself is denoted by  $\mathfrak{S}_n$  (but i will just write  $S_n$  because icba) and called the n-th symmetric group. It is a group under composition and has n! elements.

A **tranposition** is a permutation that swaps two elements of the set and leaves all the others unchanged.

#### Definition 4.1.2: Inversions of a permutation

An **inversion** of a permutation  $\sigma \in S_n$  is a pair (i, j) such that  $1 \leq i < j \leq n$  and  $\sigma(i) > \sigma(j)$ . The number of inversions of the permutation  $\sigma$  is called the **length of**  $\sigma$  and written  $\ell(\sigma)$ . In formulas:

$$\ell(\sigma) = |\{(i,j) : i < j \text{ but } \sigma(i) > \sigma(j)\}|$$

The sign of  $\sigma$  is defined to be the parity of the number of inversions of  $\sigma$ . In formulas:

$$sgn(\sigma) = (-1)^{\ell(\sigma)}$$

## Theorem 4.1.3: Multiplicativity of the sign

For each  $n \in \mathbb{N}$  the sign of a permutation produces a group homomorphism  $\operatorname{sgn}: S_n \to \{+1, -1\}$  from the symmetric group to the two-element group of signs. In formulas:

$$\operatorname{sgn}(\sigma\tau) = \operatorname{sgn}(\sigma)\operatorname{sgn}(\tau) \quad \forall \sigma, \tau \in S_n$$

### Definition 4.1.4: Alternating Group of a Permutation

For  $n \in \mathbb{N}$ , the set of even permutations in  $S_n$  forms a subgroup of  $S_n$  because it is the kernel of the group homomorphism  $\operatorname{sgn}: S_n \to \{+1, -1\}$ . This group is the **alternating group** and is denoted  $A_n$ 

#### 4.2 Determinants

#### Definition 4.2.1: Determinants

Let R be a commutative ring and  $n \in \mathbb{N}$ . The **determinant** is a mapping det:  $\operatorname{Mat}(n;R) \to R$  from square matrices with coefficients in R to the ring R that is given by the following formula

$$A = \begin{pmatrix} a_{11} & \cdots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{n1} & \cdots & a_{nn} \end{pmatrix} \mapsto \det(A) = \sum_{\sigma \in S_n} \operatorname{sgn}(\sigma) a_{1\sigma(1)} \dots a_{n\sigma(n)}$$

# 4.3 Characterising the Determinant

#### Definition 4.3.1: Bilinear Forms

Let U, V, W be F-vector spaces. A **bilinear form on**  $U \times V$  **with values in** W is a mapping  $H: U \times V \to W$  which is a linear mapping in both of its entries. This means that it must satisfy the following properties for all  $u_1, u_2 \in U$  and  $v_1, v_2 \in V$  and all  $\lambda \in F$ :

$$H(u_1 + u_2, v_2) = H(u_1, v_1) + H(u_2, v_1)$$

$$H(\lambda u_1, v_1) = \lambda H(u_1, v_1)$$

$$H(u_1, v_2 + u_2) = H(u_1, v_1) + H(u_2, v_1)$$

$$H(u_1, \lambda v_1) = \lambda H(u_1, v_1)$$

### Definition 4.3.2: Multilinear Forms

Let  $V_1,\ldots,V_n,W$  be F-vector spaces. A mapping  $H:V_1\times V_2\times\cdots\times V_n\to W$  is a **multilinear form** or just **multilinear** if for each j, the mapping  $V_j\to W$  defined by  $v_j\mapsto H(v_1,\ldots,v_j,\ldots,v_n)$ , with the  $v_i\in V_i$  arbitrary fixed vectors of  $V_i$  for  $i\neq j$  is linear.

### Definition 4.3.3: Alternating Multilinear Forms

Let V and W be F-vector spaces. A multilinear form H:  $V \times \cdots \times V \to W$  is **alternating** if it vanishes on every n-tuple of elements of V that has at least two entries equal, in other words if:

$$(\exists i \neq j \text{ with } v_i = v_j) \rightarrow H(v_1, \dots, v_i, \dots, v_j, \dots, v_n) = 0$$

#### Theorem 4.3.4: Characterisation of the Determinant

Let F be a field. The mapping

$$\det: \operatorname{Mat}(n; F) \to F$$

is the unique alternating multilinear form on n-tuples of column vectors with values in F that takes the value  $1_F$  on the identity matrix

# 4.4 Rules for Calculating with Determinants

#### Theorem 4.4.1: Multiplicativity of the Determinant

Let R be a commutative ring and let  $A, B \in \text{Mat}(n; R)$ . Then

$$\det(AB) = \det(A)\det(B)$$

#### Theorem 4.4.2

The determinant of a square matrix with entries in a field F is non-zero if and only if the matrix is invertible

#### 4.4.3 Consequences of determinant rules

- If A is invertible then  $det(A^{-1}) = det(A)^{-1}$
- If B is a square matrix then  $det(A^{-1}BA) = det(B)$

# Theorem 4.4.4: Determinants of a Transpose Matrix

The determinant of a square matrix and of the transpose of the square matrix are equal, that is for all  $A \in \text{Mat}(n; R)$  with R a commutative ring,

$$\det(A^T) = \det(A)$$

#### Definition 4.4.5: Cofactors of a Matrix

Let  $A \in \operatorname{Mat}(n;R)$  for some commutative ring R and natural number n. Let i and j be integers between 1 and n. Then the (i,) **cofactor of** A is  $C_{ij} = (-1)^{i+j} \det(A\langle i,j\rangle)$  where  $A\langle i,j\rangle$  is the matrix obtained from A by deleting the i-th row and j-th column.

$$C_{23} = (-1)^{2+3} \det \begin{pmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{pmatrix} = -a_{11}a_{32} + a_{31}a_{12}$$

# Theorem 4.4.6: Laplace's Expansion

Let  $A = (a_{ij})$  be an  $(n \times n)$ -matrix with entries from a commutative ring R. For a fixed i, the i-th row expansion of the determinant is

$$\det(A) = \sum_{j=1}^{n} a_{ij} C_{ij}$$

and for a fixed j, the j-th column expansion of the determinant is

$$\det(A) = \sum_{i=1}^{n} a_{ij} C_{ij}$$

# Definition 4.4.7: Adjugate Matrix

Let A be a  $(n \times n)$ -matrix with entries in a commutative ring R. The **adjugate matrix** adj(A) is the  $(n \times n)$ -matrix whose entries are  $adj(A)_{ij} = C_{ji}$  where  $C_{ji}$  is the (j,i)-cofactor

#### Theorem 4.4.8: Cramer's Rule

Let A be a  $(n \times n)$ -matrix with entries in a commutative ring R. Then

$$A \cdot \operatorname{adj}(A) = (\det A)I_n$$

#### 4.4.9 Alternative Definition of Cramer's

In many sources, such as Wikipedia, Cramer's Rule means the formula

$$x_i = \frac{\det(a_{*1} \mid \dots \mid b_* \mid \dots \mid a_{*n})}{\det(a_{*1} \mid \dots \mid a_{*i} \mid \dots \mid a_{*n})}$$

for solving a field F the system  $A\vec{x}=\vec{b}$  of n linear equations in n unknowns, provided that a unique solution exists. A unique solution exists if and only if A is invertible. So, instead of applying the Gaussian algorithm, you can calculate lots of determinants, replacing the i-th column of A by the given solution vector  $\vec{b}$ . It turns out that if you implement this rule on a computer, it has the same efficiency as the Gaussian algorithm. The relationship between this version of Cramer's rule and the above theorem is got by successively taking the vector  $\vec{b}$  in the system of linear equations to be the standard basis elements  $\vec{e_i}$  with  $1 \leq i \leq n$ .

# Theorem 4.4.10: Invertibility of Matrices

A square matrix with entries in a commutative ring R is invertible if and only if its determinant is a unit in R. That is,  $A \in \operatorname{Mat}(n;R)$  is invertible if and only if  $\det(A) \in R^{\times}$ 

So for instance, an integral matrix  $A \in \operatorname{Mat}(n; \mathbb{Z})$  is invertible if and only if  $\det(A)$  is 1 or -1, since  $\mathbb{Z}^{\times} = \{\pm 1\}$ . On the other hand, a matrix  $A \in \operatorname{Mat}(n; F)$  with entries in a field F is invertible if and only if  $\det(A) \neq 0$  since  $F^{\times}$  consists of the non-zero elements of F.

#### Theorem 4.4.11: Jacobi's Formula

Let  $A = (a_{ij})$  where the coefficients  $a_{ij} = a_{ij}(t)$  are functions of t. Then

$$\frac{d}{dt}\det A = \text{TrAdj}A\frac{dA}{dt}$$

# 4.5 Eigenvalues and Eigenvectors

# Definition 4.5.1: Eigenvalues and Eigenvectors

Let  $f:V\to V$  be an endomorphism of an F-vector space V. A scalar  $\lambda\in F$  is an **eigenvalue of** f if and only if there exists a non-zero vector  $\vec{v}\in V$  such that  $f(\vec{v})=\lambda\vec{v}$ . Each such vector is called an **eigenvector of** f **with eigenvalue**  $\lambda$ . For any  $\lambda\in F$ , the **eigenspace of** f **with eigenvalue**  $\lambda$  is

$$E(\lambda, f) = \{ \vec{v} \in V : f(\vec{v}) = \lambda \vec{v} \}$$

# Theorem 4.5.2: Existence of Eigenvalues

Each endomorphism of a non-zero finite dimensional vector space over an algebraically closed field has an eigenvalue

# Definition 4.5.3: Characteristic Polynomial

Let R be a commutative ring and let  $A \in \operatorname{Mat}(n;R)$  be a square matrix with entries in R. The polynomial  $\det(xI_n-A)\in R[x]$  is called the **characteristic polynomial of the matrix** A. It is denoted by

$$\chi_A(x) := \det(xI_n - A)$$

(where  $\chi$  stands for  $\chi$ aracteristic, lol)

#### Theorem 4.5.4: EVs and Characteristic Polynomials

Let F be a field and  $A \in \operatorname{Mat}(n;F)$  a square matrix with entries in F. The eigenvalues of the linear mapping  $A:F^n \to F^n$  are exactly the roots of the characteristic polynomial  $\chi_A$ 

#### 4.5.5 Eigenvalue remarks

1. Square matrices  $A,\,B\in \mathrm{Mat}(n;R)$  of the same size are conjugate if

$$B = P^{-1}AP \in Mat(n; R)$$

for an invertible  $P \in \mathrm{GL}(n;R)$ . Conjugacy is an equivalence relation on  $\mathrm{Mat}(n;R)$ . (The definition makes sense for any commutative ring R, although we will mainly be concerned with the case of a field)

2. The motivation for conjugacy comes from the various matrix representations for an endomorphism  $f:V\to V$  of an n-dimensional vector space V over a field F. Let

$$A = (a_{ij}) = {}_{\mathcal{A}}[f]_{\mathcal{A}}, B = (b_{ij}) = {}_{\mathcal{B}}[f]_{\mathcal{B}} \in \operatorname{Mat}(n; f)]$$

be the matrices of f with respect to bases  $\mathcal{A}=(\vec{v_1},\vec{v_2},\ldots,\vec{v_n}),~\mathcal{B}=(\vec{w_1},\vec{w_2},\ldots,\vec{w_n})$  for V

$$f(\vec{v_j}) = \sum_{i=1}^n a_{ij}\vec{v_i}, f(\vec{w_j}) = \sum_{i=1}^n b_{ij}\vec{w_i} \in V$$

The change of basis matrix  $P = (p_{ij}) = {}_{\mathcal{A}}[\mathrm{id}_V]_{\mathcal{B}} \in \mathrm{Mat}(n;F)$  is invertible, with

$$\vec{w_j} = \sum_{i=1}^n p_{ij} \vec{v_i} \in V$$

We have the identity

$$B = P^{-1}AP \in Mat(n; F)$$

so A, B are conjugate

3. Key observation: the characteristic polynomials of conjugate  $A,B\in \mathrm{Mat}(n,R)$  are the same

$$\chi_B(x) = \det(xI_n - B) = \det(xI_n = P^{-1}AP)$$

$$= \det(P^{-1}(xI_n - A)P) = \det(P)^{-1}\det(xI_n - A)\det(P)$$

$$= \det(xI_n - A) = \chi_A(x) \in R[x]$$

4. In view of 2 and 3 we can define the characteristic polynomial of an endomorphism  $f:V\to V$  of an n-dimensional vector space over a field F to be

$$\chi_f(x) = \chi_A(x) \in F[x]$$

with  $A = \mathcal{A}[f]_{\mathcal{A}} \in \operatorname{Mat}(n;R)$  the matrix of f with respect to any basis  $\mathcal{A}$  for V. Thanks to 4.5.4, the eigenvalues of f are exactly the roots of  $\chi_f$ , the characteristic polynomial of f

**Remark**: Let  $f: V \to V$  be an endomorphism of an n-dimensional vector space V over a field F. Suppose given an m-dimensional subspace  $W \subseteq V$  such that  $f(W) \subseteq W$ , so that there are defined endomorphisms of the subspace and the quotient space

$$g: W \to W; \vec{w} \mapsto f(\vec{w})$$
  
 $h: V/W \to V/W; W + \vec{v} \mapsto W + f(\vec{v})$ 

Any ordered basis  $\mathcal{A}=(\vec{w_1},\vec{w_2},\ldots,\vec{w_m})$  for W can be extended to an ordered basis for V

$$\mathcal{B} = (\vec{w_1}, \vec{w_2}, \dots, \vec{w_m}, \vec{v_{m+1}}, \vec{v_{m+2}}, \dots, \vec{v_n})$$

The images of the  $\vec{v_j}$ 's under the canonical projection can :  $V \to V/W$  are then an ordered basis for V/W

$$\mathcal{C} = (\operatorname{can}(v_{m+1}), \operatorname{can}(v_{m+2}), \dots, \operatorname{can}(\vec{v}_n))$$

Let  $a_{ij}, b_{ik}, c_{ik} \in F$  be the coefficients in the linear combinations

$$f(\vec{w}_j) = \sum_{i=1}^m a_{ij} \vec{w}_i \in W, f(\vec{v}_k) = \sum_{j=m+1}^n b_{jk} \vec{v}_j + \sum_{i=1}^n c_{ik} \vec{w}_i \in V$$

[WIP SO MUCH WRITING OMG]

# 4.6 Triangularisable, Diagonalisable, and Cayley-Hamilton

#### Definition 4.6.1: Triangularisability

Let  $f: V \to V$  be an endomorphism of a finite dimensional F-vector space V. f is **triangularisable** if the vector space V has an ordered basis  $\mathcal{B} = (\vec{v}_1, \vec{v}_2, \dots, \vec{v}_n)$  such that

$$f(\vec{v}_1) = a_{11}\vec{v}_1,$$

$$f(\vec{v}_2) = a_{12}\vec{v}_1 + a_{22}\vec{v}_2,$$

$$\vdots$$

$$f(\vec{v}_n) = a_{1n}\vec{v}_1 + a_{2n}\vec{v}_2 + \dots + a_{nn}\vec{v}_n \in V$$

(so that the first basis vector  $\vec{v}_1$  is an eigenvector, with eigenvalue  $a_{11}$ ) or equivalently such that the  $n \times n$  matrix  $_{\mathcal{B}}[f]_{\mathcal{B}} = (a_{ij})$  representing f with respect to  $\mathcal{B}$  is upper triangular (or any other triangular)

$$A = \begin{pmatrix} a_{11} & a_{12} & a_{13} & \cdots & a_{1n} \\ 0 & a_{22} & a_{23} & \cdots & a_{2n} \\ 0 & 0 & a_{33} & \cdots & a_{3n} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \cdots & a_{nn} \end{pmatrix}$$

#### Theorem 4.6.2

Let  $f: V \to V$  be an endomorphism of a finite dimensional F-vector space V. Then f is triangularisable iff the characteristic polynomial  $\chi_f$  decomposes into linear factors in F[x]

# Theorem 4.6.3: Triangularisability and Conjugacy

An endomorphism  $A: F^n \to F^n$  is triangularisable if and only if  $A=(a_{ij})$  is conjugate to an upper triangular matrix  $B=(b_{ij})(b_{ij}=0 \text{ for } i>j)$ , with  $P^{-1}AP=B$  for an invertible matrix P

# Definition 4.6.4: Diagonalisability

An endomorphism  $f: V \to V$  of an F-vector space V is **diagonalisable** if and only if there exists a basis of V consisting of eigenvectors of f. If V is finite dimensional then this is the same as saying that there exists an ordered basis  $\mathcal{B} = \{\vec{v}_1, \ldots, \vec{v}_n\}$  such that corresponding matrix representing f is diagonal, that is  $\mathcal{B}[f]_{\mathcal{B}} = \operatorname{diag}(\lambda_1, \ldots, \lambda_n)$ . In this case, of course,  $f(\vec{v}_i) = \lambda_i \vec{v}_i$ . A square matrix  $A \in \operatorname{Mat}(n; F)$  is **diagonalisable** if and only if the corresponding linear mapping  $F^n \to F^n$  given by left multiplication by A is diagonalisable. Thanks to [something] this just means that A is conjugate to a diagonal matrix, there exists an invertible matrix  $P \in \operatorname{GL}_J(n; F)$  such that  $P^{-1}AP = \operatorname{diag}(\lambda_1, \ldots, \lambda_n)$ . In this case the columns P are the vectors of a basis of  $F^n$  consisting of eigenvectors of A with eigenvalues  $\lambda_1, \ldots, \lambda_n$ 

### Theorem 4.6.5: Linear Independence of Eigenvectors

Let  $f: V \to V$  be an endomorphism of a vector space V and let  $\vec{v}_1, \ldots, \vec{v}_n$  be eigenvectors of f with pairwise different eigenvalues  $\lambda_1, \ldots, \lambda_n$ . Then the vectors  $\vec{v}_1, \ldots, \vec{v}_n$  are linearly independent

#### Theorem 4.6.6: Cayley-Hamilton Theorem

Let  $A \in \operatorname{Mat}(n;R)$  be a square matrix with entries in a commutative ring R. Then evaluating its characteristic polynomial  $\chi_A(x) \in R[x]$  at the matrix A gives zero.

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