Ian S.W. Ma

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1 Vector Spaces

1.1 Solution of Simultianeous Linear Equations

Assume $F = \mathbb{Q}, \mathbb{R}$ or \mathbb{C} , where $a_{ij}, b_i \in F$, then

$$\sum_{i=1}^{m} a_{ij} x_j = b_i \quad \forall i \in [1, n] : i \in \mathbb{Z}$$

is a system of linear Equations

- if all b's are 0 then the system is **homogenous**
- $L = \{x_1, ..., x_m\}$ is the **solution set** of Equations

1.2 Fields and Vector Spaces

DEFINITION 1.2.1.

- 1. A **field** F is a set with functions:
 - addition = $+: F \times F \to F; (\lambda, \mu) \mapsto \lambda + \mu$
 - multiplication = $\cdot : F \times F \to F; (\lambda, \mu) \mapsto \lambda \mu$

such that (F, +) ans $(F \setminus \{0\}, \cdot)$ are abelian groups, with:

$$\lambda(\mu + \nu) = \lambda\mu + \lambda\nu \in F \quad \forall \lambda, \mu, \nu \in F$$

The neutral elements are called $0_F, 1_F$, in particular for all $\lambda, \mu \in F$

- $\bullet \ \lambda + \mu = \mu + \lambda \in F$
- $\lambda \cdot \mu = \mu \cdot \lambda \in F$
- $\lambda + 0_F = \lambda \in F$
- $\lambda \cdot 1_F = \lambda \in F$

For all $\lambda \in F$ there exists $-\lambda \in F$ such that $\lambda + (-\lambda) = 0_F \in F$ For all $\lambda \neq 0 \in F$ tehre exists $\lambda^{-1} \neq 0 \in F$ such that $\lambda(\lambda^{-1}) = 1_F \in F$

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

$$F \times V \to V; (\lambda, \overrightarrow{v})$$

such that for all $\lambda, \mu \in F$ and $\overrightarrow{v}, \overrightarrow{w} \in V$ the following identities hold:

- $\lambda(\overrightarrow{v} + \overrightarrow{w}) = \lambda \overrightarrow{v} + \lambda \overrightarrow{w}$ Distributive Law
- $(\lambda + \mu)\overrightarrow{v} = \lambda \overrightarrow{v} + \mu \overrightarrow{v}$ Distributive Law
- $\lambda(\mu\overrightarrow{v}) = (\lambda\mu)\overrightarrow{v}$ Associativity Law
- $1_F \overrightarrow{v} = \overrightarrow{v}$

LEMMA 1.2.1. If V is a vector space then $\forall \overrightarrow{v} \in V$, $0\overrightarrow{v} = \overrightarrow{0}$

LEMMA 1.2.2. If V is a vector space then $\forall \overrightarrow{v} \in V$, $(-1)\overrightarrow{v} = -\overrightarrow{v}$

LEMMA 1.2.3. If V is a vector space over a field F then $\lambda \overrightarrow{0} = \overrightarrow{0} \quad \forall \lambda \in F$

1.3 Product of Sets and of Vector Spaces

• Cartesian product of sets: $X_1 \times ... \times X_n := \{(x_1, ..., x_n) : x_i \in X_i \text{ for } 1 \leq i \leq n\}$, an element of this product is known as a **product n-tuples**.

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 $\forall n, m \geq 0, X^n \times X^m \xrightarrow{\sim} X^{n+m}; ((x_1, ..., x_n), (x_{n+1}, ..., x_{n+m})) \mapsto (x_1, ..., x_n, x_{n+1}, x_{n+m})$

EXAMPLE 1.3.1 (Examples of tuples).

- Vector spaces of *n*-tubles over $F: V = F^n$
- Vector space as space around us

1.4 Vector Subspaces

DEFINITION 1.4.1. A subset U of a vector space V is called a **vector subspace** or **subspace** if U contains the zero vector $(\overrightarrow{0})$ and whenever \overrightarrow{u} , \overrightarrow{v} and $\lambda \in F$ we have $\overrightarrow{u} + \overrightarrow{v} \in U$ and $\lambda \overrightarrow{u} \in U$

PROPOSITION 1.4.1. Let T be a subset of vector space V over a field F. Then amongst all vector subspaces of V that include T there is a smallest vector subspace

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

DEFINITION 1.4.2. A subset S of a vector space V is called a **generating set** of a V if its span is all of the vector space. A vector space that has a finite generating set is **finitely generated**

$$S \subseteq V \land \operatorname{span}(S) = V \Rightarrow S$$
 is a generating set of V

DEFINITION 1.4.3. If X is a set, then the set of all subsets $\mathcal{P}(X) = \{U : U \subseteq X\}$ of X is called **power set** of X. We call the subset of $\mathcal{P}(X)$ a **system of subsets of** X. Given such a system $\mathcal{U} \subseteq \mathcal{P}(\mathcal{X})$ we can create two new subsets of X, the **union** and the **intersection** of the set of our system \mathcal{U} as follows:

$$\bigcup_{U \in \mathcal{U}} U = \{x \in X : \text{there is } U \in \mathcal{U} \text{ with } x \in U\}$$

$$\bigcap_{U \in \mathcal{U}} U = \{x \in X : x \in U \mid \forall U \in \mathcal{U}\}$$

1.5 Linear Independence and Bases

DEFINITION 1.5.1. A subset $L = \{\overrightarrow{v_1}, ..., \overrightarrow{v_n}\}$ of a vector subspace V is **linearly independent** if for all arbitrary scalars $\alpha_1, ..., \alpha_n \in F$:

$$\sum_{i=1}^{n} \alpha_i \overrightarrow{v_i} = 0 \to \alpha_1 = \dots = \alpha_n = 0$$

DEFINITION 1.5.2. A subset $L = \{\overrightarrow{v_1}, ..., \overrightarrow{v_n}\}$ of a vector subspace V is **linearly dependent** if it's not **linearly independent**, which mean there exist some $\alpha_i \in \{a_1, ..., a_n\}, \alpha_i \neq 0$ such that

$$\sum_{i=1}^{n} \alpha_i \overrightarrow{v_i} = 0$$

DEFINITION 1.5.3. A basis of a vector space B of a vector space V is a linearly independent generating set in V

 $\operatorname{span}(B) = V \wedge B$ is linearly independent $\Rightarrow B$ is a basis of V

DEFINITION 1.5.4. Let F be a field, V a vector space over F and $\overrightarrow{v_1}, ..., \overrightarrow{v_r} \in V$ vectors. The family $(\overrightarrow{v_i})_{1 \leq i \leq r}$ is a basis of V if and only if the following "evaluation"

$$\Phi: F^r \to V$$

$$(\alpha_1, ..., \alpha_r) \mapsto \alpha_1 \overrightarrow{v}_1 + ... + \alpha_r \overrightarrow{v}_r$$

is a bijection

DEFINITION 1.5.5. The following for a subset E of a vector space V are equivalent: An isomorphism of a vector space to itself is called **anautomorphism** of our vector space.

- Our subset E is a basis, ie. a linearly independent generating set;
- Our subset E is a **minimal** among all generating sets, meaning that $E \setminus \{\overrightarrow{v}\}$ does not generate V
- Our subset E is a **maximal** among all linearly independent subsets, meaning that $E \cup \{\overrightarrow{v}\}$ is not linearly independent for any $\overrightarrow{v} \in V$

COROLLARY 1.5.1 (The existence of a basis). Let V be a finitely generated vector space over a vield F, then V is a basis

THEOREM 1.5.1. Let V be a vector space.

- If $L \subset V$ is a linearly independent subset and E is a minimal amongst all generating sets of our vector with $L \subseteq E$, then E is a basis.
- If $L \subseteq V$ is a generating set and if L is maximal amongst all linearly independent subsets of vector space with $L \subseteq E$, then L is a basis.

DEFINITION 1.5.6 (To infinity but not beyond). Let X be a set and F a field. The set $\operatorname{Maps}(X,F)$ od 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 with send almost all elements of X to 0 is a vector subspace

THEOREM 1.5.2. Let field F, F-vector space V and family of vectors $(\overrightarrow{v_i})_{i \in I}$ from V, The following are equivalent:

- The family $(\overrightarrow{v_i})_{i \in I}$ is a basis of V;
- For each vector $\overrightarrow{v} \in V$ there is percisely one family $(a_i)_{i \in I}$ of elements of field F, almost all of which are zero and such that:

$$\overrightarrow{v} = \sum_{i \in I} a_i \overrightarrow{v_i}$$

1.6 Dimension of a vector space

THEOREM 1.6.1 (Fundamental Estimate of Linear Algebra). No linearly independent subset of a given vector space has more elements then 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 1.6.2 (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

LEMMA 1.6.1 (Exchange Lemma). 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 $\overrightarrow{w} \in V \setminus M$ is a vector not belonging to M such that $M \cup \{\overrightarrow{w}\}$ is linearly independent, then there exists $\overrightarrow{e} \in E \setminus M$ such that $\{E \setminus \{\overrightarrow{e}\}\} \cup \{\overrightarrow{w}\}$ is a generating set for V

COROLLARY 1.6.1 (Cardinality of Bases). Let V be a finitely generated vector space.

- V has a finite basis
- V cannot have an infinite basis
- \bullet Any two bases of V have the same number of elements

DEFINITION 1.6.1. The cardinality of one (and by Cardinality of Bases each) basis of a finitely generated vector space V is called the **dimension** of V and will be denoted by $\dim(V)$. If the vector space is not finitely generated, then we write $\dim(V) = \infty$ and call V infinite dimensional. As usual, we will ignore the difference between infinities.

COROLLARY 1.6.2 (Cardinality Criterion for Bases). Let V be a finitely generated vector space.

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

COROLLARY 1.6.3 (Dimension Estimate for Vector Subspaces). A proper vector subspace of a finite dimensional vector space has itself a strictly smaller dimension.

THEOREM 1.6.3 (The Dimension Theorem). Let V be vector space containing vector subspaces $U, W \subseteq V$. Then

$$\dim(U+W) + \dim(U\cap W) = \dim(U) + \dim(W)$$

1.7 Linear Mappings

DEFINITION 1.7.1 (Definition od homomorphism, isomorphism, and automorphism). Let V, W be a vector spaces over a field F. A mapping $f: V \to W$ is called **linear** or more percisely **F-linearly** or even a **homomorphism of** F-vector spaces if for all $\overrightarrow{v}_1, \overrightarrow{v}_2 \in V$ and $\lambda \in F$ we have

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

• A bijective linear mapping is called as **isomorphism** of vector spaces. If there is an isomorphism bwetween two vector spaces we say them **isomorphic**.

- A **homomorphism** from one vector space to itself is called anendomorphism of our vector space.
- An isomorphism of a vector space to itself is called an **automorphism** of our vector space.

DEFINITION 1.7.2. A point that is sent to itself br a mapping is called a **fixed point** of the mapping. Given a mapping $f: X \to X$, we do note the set of fixed points by

$$X^f = \{x \in X : f(x) = x\}$$

DEFINITION 1.7.3. Two vector subspaces V_1, V_2 of a vector space V are called **complementary** of addition defines a bijection $V_1 \times V_2 \xrightarrow{\sim} V$

THEOREM 1.7.1 (The Classification of Vector Spaces by their Dimension). Let $n \in \mathbb{N}$. Then a vector space V over a field F is isomorphic to F^n i.f.f. $\dim(V) = n$

LEMMA 1.7.1 (Linear Mappings and Bases). Let V, W be vector spaces over F and let $B \subset V$ be a basis. Then restriction of a mapping gives a bijection

$$\operatorname{Hom}_F(V,W) \stackrel{\sim}{\to} \operatorname{Maps}(B,W)$$

 $f \mapsto f|_B$

PROPOSITION 1.7.1.

- Every injective linar mapping $f: V \hookrightarrow W$ has a **left inverse**, in other words \exists a linear mapping $g: W \to V$ such that $g \circ f = \mathrm{id}_V$
- Every surjective linear mapping $f: V \to W$ has a **right inverse**, in other words \exists a linear mapping $g: W \to V$ such that $g \circ f = \mathrm{id}_W$

1.8 Rank-Nullity Throrem

DEFINITION 1.8.1. The **image** of a linear mapping $f: V \to W$ is the subset $\operatorname{im}(f) = f(V) \subseteq W$. The **preimage** of the zero vector (**kernel**) of a linear mapping $f: V \to W$ is denoted by

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

The kernel is a vector subspace if V

LEMMA 1.8.1. A linear mapping $f: V \to W$ is injective if an only if it's kernel is zero.

THEOREM 1.8.1 (Rank-Nullity Theorem). Let $f: V \to W$ be a linear mapping between vector spaces, then $\dim(V) = \dim(\ker f) + \dim(\inf(f))$

2 Linear Mappings and Matrices

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

THEOREM 2.1.1 (Linear mappings $F^m \to F^n$ and Matrices). Let F be a field and let $m, n \in \mathbb{N}$ be neutral numbers. There is a nijection between the space of linear mappings $F^m \to F^n$ and the set of matrices woth n rows and m columns and entries in F:

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

$$f \mapsto [f]$$

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

$$[f] = (f(\overrightarrow{e}_1)|f(\overrightarrow{e}_2)|\dots|f(\overrightarrow{e}_m))$$

DEFINITION 2.1.1. Let $n, m, l \in \mathbb{N}$, F a field and let $A \in \operatorname{Mat}(n \times m; F)$ and $B \in \operatorname{Mat}(m \times l; F)$ be matrices. The **product** $A \circ B = AB \in \operatorname{Mat}(n \times l; F)$ is the matrix defined by

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

THEOREM 2.1.2 (Composition of Linear Mapping and Products of Matrices). Let $g: F^l \to F^m$ and $f: F^m \to F^n$ be linear mappings. Then $[f \circ g] = [f] \circ [g]$

2.2 Basic Properties of Matrices

DEFINITION 2.2.1. A matrix A is called **invertable** if there exist matrices such tht BA = I and AC = I

DEFINITION 2.2.2. will define an **elementary matrix** to be any square matrix that differs from the identity matrix in at most one entry.

THEOREM 2.2.1. Every square matrix with entries in a field can be written as a product if elementary matrices.

DEFINITION 2.2.3. Any matrix whose only non-zero entries lies on the diagonal, and which has first 1's along the diagonal and then 0's, is said to be in **Smith Normal Form**:

$$A_{ij} = \begin{cases} 0 \text{ if } i \neq j \\ 1 \text{ if } A_{(i+1)(j+1)} = 1 \\ 0 \text{ otherwise} \end{cases}$$

THEOREM 2.2.2 (Transformation of a Matrix into Smith Normal Form). For each matrix $A \in \text{Mat}(n \times m; F)$ there exist invertable matrices P, Q such that PAQ is a matrix in Smith Normal Form.

DEFINITION 2.2.4. The **column rank** of a matrix $A \in \text{Mat}(n \times m; F)$ is the dimension of the subsequence of F^n generated by the columns of A. Simmilarly, the **row rank** of A is the dimension of the subspace of F^m generated by the rows of A.

THEOREM 2.2.3. The column rank and the row rank of any matrix are equal.

Let's now refer the column and row rank as **rank** for the sake of not losing any generality.

DEFINITION 2.2.5. When the rank is as big as possible, meaning that it's equal to either the number of rows or number of columns (whichever is smaller), then the matrix has **full rank**

$$rank(M) = min(\{rowrank(M), colrank(M)\}) \Rightarrow M$$
 has full rank

2.3 Abstract Linear Mappings and Matrices

THEOREM 2.3.1 (Abstract Linear Mappings and Matrices). Let F be a field, V, W vector spaces over F with ordered bases $\mathcal{A} = (\vec{v}_1, ..., \vec{v}_m)$ and $\mathcal{B} = (\vec{w}_1, ..., \vec{w}_n)$. Then to each linear mapping $f: V \to W$ we assosiate bases a **representing matrix** $_{\mathcal{B}}[f]_{\mathcal{A}}$ whose entried $a_i j$ are defined by the identity

$$f(\vec{v}_j) = \sum_{i=1}^n a_{ij} \vec{w}_i \in W$$

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

$$\mathbf{M}_{\mathcal{B}}^{\mathcal{A}}: \mathrm{Hom}_{F}(V, W) \xrightarrow{\sim} \mathrm{Mat}(n \times m; F)$$

$$f \mapsto_{\mathcal{B}} [f]_{\mathcal{A}}$$

We call $\mathbf{M}_{\mathcal{B}}^{\mathcal{A}}(f) =_{\mathcal{B}} [f]_{\mathcal{A}}$ the epresenting matrix of the mapping with respect to the bases \mathcal{A} and \mathcal{B}

THEOREM 2.3.2 (The Representing Matrix of a Composition of Linear Mappings). Let F be a field and U, V, W finite dimensional vector spaces over F 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 matrix of f and g:

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

DEFINITION 2.3.1. Let V be a finite dimensional vector space with an ordered basis $\mathcal{A} = (\vec{v}_1, ..., \vec{v}_m)$. We will denote the invese to the bijection of $\Phi_{\mathcal{A}} \stackrel{\sim}{\to} V, (\alpha_a, ..., \alpha_m)^T \mapsto \sum_{i=1}^m \alpha_i \vec{v}_i$ by

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

THEOREM 2.3.3 (Representation of the Image of a Vector). Let V, W be finite dimensional vector spaces over F with ordered bases \mathcal{A}, \mathcal{B} and let $f: V \to W$ be a linear mapping. The following holds for $\vec{v} \in V$:

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

2.4 Change of Matrix by Change of Basis

DEFINITION 2.4.1. Let $\mathcal{A} = (\vec{v}_1, ..., \vec{v}_n), \mathcal{B} = (\vec{w}_1, ..., \vec{w}_n)$ be ordered bses of the same F-vector space V. Then the matrix representing the identity mapping with respect to the bases $\mathcal{B}[\mathrm{id}_V]_{\mathcal{A}}$ is called a **change of basis matrix**. By definition, its entries are given by the equalities $\vec{v}_j = \sum_{i=1}^n a_{ij}\vec{w}_i$

THEOREM 2.4.1 (Change of Basis). Let V, W be finite dimensional vector spaces over F and let $f: V \to W$ br a linear mapping. Suppose that $\mathcal{A}, \mathcal{A}'$ are order bases of V and $\mathcal{B}, \mathcal{B}'$ are ordered bases of W. Then:

$$_{\mathcal{B}'}[f]_{\mathcal{A}'} =_{\mathcal{B}'} [\mathrm{id}_W]_{\mathcal{B}} \circ_{\mathcal{B}} [f]_{\mathcal{A}} \circ_{\mathcal{B}} [\mathrm{id}_V]_{\mathcal{A}'}$$

COROLLARY 2.4.1. Let V be a finite dimensional vector space and let $f:V\to V$ be an endomorphism of V. Suppose that $\mathcal{A}, \mathcal{A}'$ are ordered bases of V. Then

$$_{\mathcal{A}'}[f]_{\mathcal{A}'} =_{\mathcal{A}} [\mathrm{id}_V]_{\mathcal{A}'}^{-1} \circ_{\mathcal{A}} [f]_{\mathcal{A}} \circ_{\mathcal{A}} [\mathrm{id}_V]_{\mathcal{A}'}$$

THEOREM 2.4.2 (Smith Normal Form). Let $f: V \to W$ be a linear mapping between finite dimensional F-vector spaces. There exist an order basis \mathcal{A} of V and an ordered basis \mathcal{B} of W, such that the representing matrix $_{\mathcal{B}}[f]_{\mathcal{A}}$ has zero entries everywhere except possibly on one diagonal, and along the diagonal there are 1's first, followed by 0's

DEFINITION 2.4.2 (Trace). The trace of a square matrix is defined to be the sum of its diagonal entries:

$$\operatorname{tr}(A) = \sum_{i=1}^{n} a_{ii}$$

- Let $A \in \operatorname{Mat}(n \times m, F)$, $B \in \operatorname{Mat}(m \times n, F)$, then $\operatorname{tr}(AB) = \operatorname{tr}(BA)$
- Let $f: V \to W$ and $g: W \to V$ two linear mappings where V and W are both finite dimensional F-vector spaces, then $\operatorname{tr}(fg) = \operatorname{tr}(gf)$
- Let V be a finite dimensional F-vector space and let $f: V \to V$ be an idempotent, that is $f^2 = f$, then $\operatorname{tr}(f) = \dim(\operatorname{im} f)$
- Let V be a finite dimensional F-vector space and $f: v \to V$ a linear mapping, then $\operatorname{tr}(((f \circ) | \operatorname{End}_F(V)) = (\dim_F V) \operatorname{tr}(f | V))$

3 Rings and Modules

3.1 Rings

DEFINITION 3.1.1. A ring is a set with two operations $(R, +, \cdot)$ that satisfy:

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

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

A ring which element is commutative, that means $a \cdot b = b \cdot a \quad \forall a, b \in R$, is a **commutative** ring

PROPOSITION 3.1.1 (Divisbility by Sum). A natural number is divisible by 3(respectively by 9) percisely when the sum of its digits is divisible by 3 (or 9)

DEFINITION 3.1.2. 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 $a \cdot a^{-1} = a^{-1} \cdot a = 1$

PROPOSITION 3.1.2. Let m be a positive integer. The commutative ring $\mathbb{Z}/m\mathbb{Z}$ is a field i.f.f. m is prime.

3.2 Peoperties of Rings

LEMMA 3.2.1 (Mutiplying by zero and negatives). 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 3.2.1. Let $m \in \mathbb{Z}$. The m-th multiple ma of an element a in an abelian group R is: ma = a + a + ... + a if m > 0 (sum of m a's). Otherwise 0a = 0 and (-m)a = -(ma)

LEMMA 3.2.2 (Rules of multiples). Let R be a ring, let $a, b \in R$ and $m.n \in \mathbb{Z}$. 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 3.2.2. Let R be a ring. An element $a \in R$ is called a **unit** if it is **invertable** in R or in other words has a **multiplicative inverse** in R, meaning that $\exists a^{-1} \in R$ such that

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

PROPOSITION 3.2.1. The set R^{\times} of units in a ring R forms a group under multiplication.

DEFINITION 3.2.3 (Zero-divisor). 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

$$a \neq 0 \in R, \exists b \neq 0 \in R \text{ s.t. } ab = 0 \cup ba = 0 \Rightarrow a \text{ is a zero divisor}$$

DEFINITION 3.2.4 (Integral domain). An **integral domain** is a non-zero commutative ring that has no zero-divisors, therefore if D is an integral domain then:

- 1. $ab = 0 \Rightarrow a = 0$ or b = 0, and
- $2. \ a,b \neq 0 \Rightarrow ab \neq 0$

PROPOSITION 3.2.2 (Cancellation Law for Integral Domains). Let R be an integral domain and let $a, b, c \in R$. If $ab = ac \land a \neq 0 \Rightarrow b = c$

PROPOSITION 3.2.3. Let m be a natural number. The $\mathbb{Z}/m\mathbb{Z}$ is an integral domain i.f.f. m is prime

THEOREM 3.2.1. Every **fnite** integral domain is a field

3.3 Polynomials

DEFINITION 3.3.1. 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 $m \in \mathbb{N} \setminus 0$ and elements $a_i \in R$ for $0 \le i \le m$. The set of all Polynomials over R is denoted by R[X]. Incase a_m is not zero, the polynomial P has a degree of m, written $\deg(P) = m$, where a_m is the leading coefficient.

When the leading coefficient is 1 the polynomial is a **monic** polynomial, linear for a_1 , quardratic for a_2 , then cubic for a_3 .

DEFINITION 3.3.2. Whith the definition in the set R[X] becomes a ring called the **ring of polynomials with coefficients in** R, **or over** R. The zero and the indentity of R[X] are the zero and identity of R resp.

LEMMA 3.3.1.

- 1. If a ring R with no zero-divisors, then R[X] has no zero-divisors and $\deg(PQ) = \deg(P) + \deg(Q)$ for all non-zero $P, Q \in R[X]$
- 2. If R is an integral domain then so is R[X], so if R^{\times} is an integral domain then so is $R[X]^{\times}$

THEOREM 3.3.1 (Division and Remainder). Let R be an integral domain and let $P, Q \in R[X]$ with Q monic. Then \exists unique $A, B \in R[X]$ such that P = AQ + B and $\deg(B) < \deg(Q)$ or B = 0

DEFINITION 3.3.3. 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 X in the polynomial P by the corresponding powers of X. In this way we have a mapping $R[X] \to \operatorname{Maps}(R,R)$

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

PROPOSITION 3.3.1. Let R be a commutative ring, let $\lambda \in R$ and $P(X) \in R[X]$. Then λ is a root of P(X) i.f.f. $(X - \lambda)$ divides P(X)

THEOREM 3.3.2. Let R be a field, or more generally an integral domain. Then a non-zero polynomial $P \in R[X] \setminus \{0\}$ has at most $\deg(P)$ roots in R

DEFINITION 3.3.4. A field F is algebraically closed of each non-constant polynomial $P \in F[X] \setminus F$ with coefficients in our field has a root in our field F

THEOREM 3.3.3 (Fundamental Theorem of Algebra). The field of complex numbers \mathbb{C} , is algebraically closed.

THEOREM 3.3.4. If F is an algebraically closed field, then every non-zero polynomial $P \in F[X] \setminus \{0\}$ decomposes into linear factors

$$P = c \prod_{i=1}^{n} (X - \lambda_i)$$

with $n \geq 0, c \in F^{\times}$ and $\lambda_1, ..., \lambda_n \in F$

3.4 Homomorpgism, Ideals and Subrings

DEFINITION 3.4.1. Let R, S be rings. A mapping $f : R \to S$ is a **ring homomorphism** if the following hold for all $x, y \in R$:

$$f(x+y) = f(x)f(y)$$
$$f(xy) = f(x)f(y)$$

LEMMA 3.4.1. Let R, S be rings and $f: R \to S$ a ring homomorphism. Then for all $x, y \in R, m \in \mathbb{Z}$:

- 1. $f(0_r) = 0_s$ Where $O_R, 0_S$ are the zeros of the resptive ring
- $2. \ f(-x) = -f(x)$
- 3. f(x y) = f(x) f(y)
- 4. f(mx) = mf(x)

DEFINITION 3.4.2. A subset I of a ring R is an **ideal**, written $I \subseteq R$, if the following hold:

- 1. $I \neq 0$
- 2. I is closed under subtraction
- 3. $\forall i \in I, r \in R : ir, ri \in I$

DEFINITION 3.4.3. Let R be a commutative ring and let $T \subset R$. Then the **ideal of** R **generated by** T is the set

$$_{R}\langle T\rangle = \left\{ \sum_{i=1}^{m} r_{i}t_{i} : t_{1}, ..., t_{m} \in T, r_{1}, ..., r_{m} \in R \right\}$$

EXAMPLE 3.4.1.

- Let $m \in \mathbb{Z}$. Then $\mathbb{Z}\langle m \rangle = m\mathbb{Z}$
- Let $P \in \mathbb{R}[X]$. Then $\mathbb{R}[X] \langle P \rangle = \{AP : A \in \mathbb{R}[X]\} = \{Q : P \text{ divides } Q \text{ in } \mathbb{R}[X]\}$

PROPOSITION 3.4.1. Let R be a commutative ring and let $T \subseteq R$. Then $R \langle T \rangle$ is the smallest ideal of R that contains T

DEFINITION 3.4.4. Let R be a commutative ring. An ideal I of R is called a **principal ideal** if $I = \langle t \rangle$ for some $t \in R$

DEFINITION 3.4.5. Let R, S be rings with zero elements $0_R, 0_S$ resp and let $f: R \to S$ be a ring homomorphism. Since f is in particular a group homomorphism from (R, +) to (S, +), ther kernel of f already has a meaning: $\ker(f) = \{r \in R : f(r) = 0_S\}$.

PROPOSITION 3.4.2. Let R and S be rings and $f: R \to S$ a ring homomorphism. Then $\ker(f)$ is an ideal of R.

LEMMA 3.4.2. f is injective i.f.f. $ker(f) = \{0\}$

LEMMA 3.4.3. The intersection of any collection of ideals of a ring R is an ideal of R

LEMMA 3.4.4. Let I, J be ideals of ring R. Then $I + J = \{a + b : a \in I, b \in J\}$ is an ideal of R

DEFINITION 3.4.6 (subring). Let R be a ring. A subset R' of R is a **subring** of R if R' itself is a ring under the operations of addition and multiplication defined in R

EXAMPLE 3.4.2. \forall ring R, $\{0\}$, R are always the subrings of R

PROPOSITION 3.4.3 (Test of a subring). Let R' be a subset of a ring R. Then R' is a subring i.f.f.

- 1. R' has a multiplicative identity, and
- 2. R' is closed under subtraction: $a, b \in R' \rightarrow a b \in R'$, and
- 3. R' is closed under multiplication

PROPOSITION 3.4.4. Let R and S be subrings and $f: R \to S$ a ring homomorphism

- 1. If R' is a subring of R then f(R') is a subring of S. In particular, im(f) is a subring of S
- 2. Assume that $f(1_R) = 1_S$. Then if x is a unit in R, f(x) is a unit in S and $(f(x))^{-1} = f(x^{-1})$. In this case f restricts to a group homomorphism $f|_{R^\times} : R^\times \to S^\times$

3.5 Equivalece Relations

DEFINITION 3.5.1. A **relation** R on a set X is a subset $R \subseteq X \times X$. In this context, and only in this context, we write xRy instead of $(x,y) \in R$. R is an **equivalence relation** on X when for all $x, y, z \in X$ the following holds:

- 1. Reflexivity: xRx
- 2. Symmetry: $xRy \Leftrightarrow yRx$
- 3. Transitivity: $(xRy \cap yRx) \rightarrow xRz$

DEFINITION 3.5.2. Suppose that \sim is an equivalence relation on a set X. For $x \in X$ the set $E(x) := \{z \in X : z \sim x\}$ is called the **equivalence class of** x. A subset $E \subseteq X$ is called an **equivalence class** for our equivalence relation if there is an $x \in X$ for which E = E(x). An element of an equivalence relation is called a **representative** of the class. A subset $Z \subseteq X$ contains percisely one element from each equivalence relation is called a **system of representatives** for the equivalence relation.

DEFINITION 3.5.3. Given an equivalence relation \sim on the set X we will denote the set of equivalence classes, which is a subset of the power set $\mathcal{P}(X)$, by

$$(X/\sim) := \{E(x) : x \in X\}$$

There is a canonical mapping can: $X \to (X/\sim), x \mapsto E(x)$. It is obviously a surjection.

EXAMPLE 3.5.1. Let \equiv be the equivalence relation of " $a \equiv b \pmod{m}$ ". Then $(\mathbb{Z}/\equiv) = \mathbb{Z}/m\mathbb{Z}$

DEFINITION 3.5.4. $g:(X/\sim)\to Z$ is **well-defined** if \exists a mapping $f:X\to Z$ such that f has the peoperty $x\sim y\to f(x)=f(y)$ and $g=\bar{f}$

3.6 Factor Rings and the First Isomorphism Theorem

DEFINITION 3.6.1 (Coset). Let R be a ring, $I \subseteq R$ be an ideal in ring R. The set

$$x + I = \{x + i : i \in I\} \subseteq R$$

is a coset of I in R or the coset of x with respect to I in R

DEFINITION 3.6.2. Let R be a ring, $I \subseteq R$ an ideal, and \sim the equivalence relation defined by $x \sim y \Leftrightarrow x - y \in I$. Then R/I, the factor ring of R by I or the quotient of R by I, is the set (R/\sim) of cosets of I in R.

THEOREM 3.6.1. Let R be a ring and $I \subseteq R$ and ideal. Then R/I is a ring, where the operation of additionis defined by

$$(x+I) + (y+I) = (x+y) + I \quad \forall x, y \in R$$

and multiplication is defined by

$$(x+I)\dot{(y+I)} = xy+I \quad \forall x,y \in R$$

THEOREM 3.6.2 (The Universal Property of Factor Rings). Let R be a ring and I an ideal of R:

- The mapping can: $R \to R/I$ sending r to r+I for all $r \in R$ is a surjective ring homomorphism with kernel I
- If $f: R \to S$ is a ring homomorphism with $f(I) = \{0_S\}$, so that $I \subseteq \ker(f)$, then there is a unique ring homomorphism $\bar{f}: R/I \to S$ such that $f = \bar{f} \circ \mathsf{can}$

THEOREM 3.6.3 (First Isomorphism Theorem for Rings). Let R and S be rings. Then every ring homomorphism $f: R \to S$ induces a ring isomorphism

$$\bar{f}: R/\ker f \stackrel{\sim}{\to} \operatorname{im} f$$

3.7 Modules and All That

DEFINITION 3.7.1. A (left) module M over a ring \mathbf{R} is a pair consisting if an abelian group M = (M, +) and a mapping

$$R \times M \to M$$

 $(r, a) \mapsto ra$

such that for all $r, s \in R$ and $a, b \in M$ the following indentities hold:

$$r(a + b) = ra + rb$$
$$(r + s)a = ra + sa$$
$$r(sa) = (rs)a$$
$$1_{R}a = a$$

LEMMA 3.7.1. Let R be a ring and M an R-module:

- 1. $0_R a = 0_M \quad \forall a \in M$
- $2. r0_M = 0_M \quad \forall r \in R$
- 3. $(-r)a = r(-a) = -(ra) \quad \forall r \in R, a \in M$

DEFINITION 3.7.2. Let R be a ring and let M, N be R-modules. A mapping $f: M \to N$ is an R-homomorphism or homomorphism if the following hold for all $a, b \in M$ and $r \in R$

$$f(a+b) = f(a+b)$$
$$f(ra) = rf(a)$$

The kernel of f is $\ker f = \{a \in M : f(a) = 0_N\} \subseteq M$ and the image of f is $\operatorname{im} f = \{f(a) : a \in M\} \subseteq N$. If f is a bijection then it is an R-module isomorphism or isomorphism $(M \simeq N)$

DEFINITION 3.7.3. A non-empty subset M' of an R-module M is a submodule if M' is an R-module with respect to the operations of the R-module M restricted to M'

PROPOSITION 3.7.1 (Test for a submodule). Let R be a ring and M an R-module. A subset M' of M is a submodule i.f.f.

- 1. $0_M \in M'$
- $2. \ a, b \in M' \Rightarrow a b \in M'$
- 3. $r \in R, ai \in M' \Rightarrow ra \in M'$

LEMMA 3.7.2. Let $f: M \to N$ be an R-homomorphism. The ker f is a submodule of M and im f is a submodule of N.

LEMMA 3.7.3. Let R be a ring, M, N be R-modules and let $f: M \to N$ be an R-homomorphism. Then f is injective i.f.f. $\ker f = \{0_M\}$

DEFINITION 3.7.4. Let R be a ring, M and R-module and let $T \subseteq M$. The the submodule of M generated by T is the set

$$_{R}\langle T \rangle = \{r_{1}t_{1} + ... + r_{m}t_{m} : t_{1}, ..., t_{m} \in T, r_{1}, ..., r_{m} \in R\}$$

together with the zero element in the case $T = \emptyset$. If $T = t_1, ..., t_n$ a finite set, we can write ${}_R\langle T\rangle$ as ${}_R\langle t_1, ..., t_n\rangle$. The module M is **finitely generated** if it is generated bt a finite set: $M = {}_R\langle t_1, ..., t_n\rangle$. It is called **cyclic** if it is generated by a singleton: $M = {}_R\langle t\rangle$

EXAMPLE 3.7.1.

- A cyclic group = a cyclic \mathbb{Z} -module
- Let R be a commutative ring. The the ideal generated by $T \subseteq R = \text{submodule of } R$ generated by T.
- A principal ideal of R = cyclic submodule of R
- $\{0_M\}$ is always a cyclic submodule of an R-module M, generated by the element 0_M : ${}_R\langle 0_M\rangle = \{0_M\}$

LEMMA 3.7.4. Let $T \subseteq M$. The ${}_R\langle T \rangle$ is the smallest submodule of M that contains T

LEMMA 3.7.5. The intersection of any collection of submodules of M is a submodule of M.

LEMMA 3.7.6. Let M_1, M_2 be submodules of M, Then $M_1 + M_2 = \{a + b : a \in M_1, b \in M_2\}$ is a submodule of M

DEFINITION 3.7.5. Let R be a ring, M an R-module and N a submodule of M. For each $a \in M$ the coset of a with respect to N in M is

$$a + N = \{a + b : b \in N\}$$

THEOREM 3.7.1 (The Universal Property of Factor Modules). Let R be a ring, let L and M be R-modules, and N a submodule of M.

- 1. The mapping $g: M \to M/N$ sending a to a+N for all $a \in M$ is serjective R-homomorphism with kernel N: ker g=N
- 2. If $f: M \to L$ is an R-homomorphism with $f(N) = \{0_L\}$, so that $N \subseteq \ker f$, then there is a unique homomorphism $\bar{f}: M/N \to L$ such that $f = \bar{f} \circ \operatorname{can}$

THEOREM 3.7.2 (First Isomorphism Theorem for Modules). Let R be a ring and let M, N be R-modules. Then every R-homomorphism $f: M/\ker f \xrightarrow{\sim} \operatorname{im} f$

4 Determinants and Eigenvalues Redux

4.1 The sign of a permutation

DEFINITION 4.1.1. The group of all permutations of the set $\{1, 2, ..., n\}$, also known as bijections from $\{1, 2, ..., n\}$ to itself, is denoted by $\mathfrak{S}_{\mathfrak{n}}$ and called the n-th symmetric group. It is a group under composition and it has n! elements.

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

DEFINITION 4.1.2. An inversion of a permutation $\sigma \in \mathfrak{S}_n$ is apair (i, j) such that $1 \le i < j \le n$ and $\sigma(i) > \sigma(j)$. The number of inversions of the permutation σ is called the length of σ ans written $l(\sigma)$. In formulas:

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

The sign of σ is defined to be the parity of the permutations of σ . In formulas:

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

EXAMPLE 4.1.1. consider $(1\,2\,4\,5\,3) = (1\,2\,4\,5\,3)(6) \in \mathfrak{S}_6$, the inversions are (1,3), (2,3), (2,5), (4,5). therefore $l((1\,2\,4\,5\,3)) = 4$

LEMMA 4.1.1 (Multiplicity of the sign). For each $n \in \mathbb{N}$ the sign of a permutation produces a group homomorphism $\operatorname{sgn}: \mathfrak{S}_n \to \{+1, -1\}$ from the symmetric group ton the two-element group of signs. In formulas:

$$\mathbf{sgn}(\sigma\tau) = \mathbf{sgn}(\sigma)\mathbf{sgn}(\tau) \quad \forall \sigma, \tau \in \mathfrak{S}_n$$

DEFINITION 4.1.3. For $n \in \mathbb{N}$, the set of even permutations in \mathfrak{S}_n forms a subgroup of \mathfrak{S}_n because it is the kernel of the group homomorphism $\operatorname{sgn}: \mathfrak{S}_n \to \{+1, -1\}$. This group is the alternating group and is denoted A_n

4.2 Determinants and What They Mean

DEFINITION 4.2.1. Let R be a commutative ring and $n \in \mathbb{N}$. The determinant is a mapping det: $Mat(n; R) \to R$ from square matrices with coefficients in R to the ring R from square matrices with coefficient in R of the ring R that is given by the following formula (**Leibniz formula**):

$$A = \begin{pmatrix} a_{11} & \dots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{n1} & \dots & a_{nn} \end{pmatrix} \mapsto \det(A) = \sum_{\sigma \in \mathfrak{S}_n} \left(\mathbf{sgn}(\sigma) \prod_{i=1}^n a_{i\sigma(i)} \right)$$

There is a really good demo in Example 4.2.3 in the Iain's note

EXAMPLE 4.2.1.

- An $(n \times n)$ matrix $A = (a_{ij})$ is upper triangular if $a_{ij} = 0 \quad \forall i > j$. $\det(A) = \prod_{i=1}^{n} a_{ii}$
- The simmilar holds for the lower triangle

4.3 Characterising the Determinant

DEFINITION 4.3.1. 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 $\lambda \in F$:

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

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

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

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

A bilinear form is **symmetric** if U = V and $H(u, v) = H(v, u) \quad \forall u, v \in U = V$ A bilinear form is **altering** or **antisymmetric** if U = V and $H(u, u) = 0 \quad \forall u \in U = V$

DEFINITION 4.3.2. Let $V_1, ..., V_n, W$ be F-vector spaces. A mapping $H: V_1 \times v_2 \times ... \times V_n \to W$ is a **multilinear form** aka. **multilinear** if for each j the mapping V_j defined by $v_j \mapsto H(v_1, ..., v_j, ..., v_n)$ with the $v_i \in V_i$ arbitrary fixed vectors of V_i for $i \neq j$, is linear. In the case n = 2, this is exactly the definition of a linear mapping.

DEFINITION 4.3.3. Let V and W be F-vector spaces. A multilinear form $H: V \times ... \times V \to W$ is **alternating** if it vanishes on every n-tuple of elements of V that has at least two entries equal, in order words:

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

In the case n=2, this is exactly the definition of an alternating or antisymmetric bilinear mapping.

THEOREM 4.3.1 (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 Caluculating with Determinants

THEOREM 4.4.1 (Multiplicy for Calculating with Determinants). Let R be a commutative right and $A, B \in \text{Mat}(n, R)$. Then $\det(AB) = \det(A) \det(B)$

THEOREM 4.4.2 (Determinantal Criterion for invertibility). The determinant of a square matrix with entries in a field F is a non-zero i.f.f. the matrix is invertable.

DEFINITION 4.4.1. Let $A \in \operatorname{Mat}(n; R)$ fir some commutative ring R and natural number n. Let i and j be integer between 1 and n. Then the (i, j) cofactor of A is $C_{ij} = (-1)^{i+j} \det(A\langle i, j \rangle)$ where $A\langle i, j \rangle$ is the matrix I obtain from A deleting the i-th row and the j-th column.

THEOREM 4.4.3 (Laplace's Expansion of the Determinant). Let $A = (a_{ij})$ be an $n \times n$ -matrix with entries form a commutative ring R. For a fixed i the i-th row expansion of the determinant is

$$\det(A) = \sum_{i=1}^{n} a_{ij} C_{ij} = \sum_{i=1}^{n} a_{ij} (-1)^{i+j} \det(A\langle i, j \rangle)$$

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

$$\det(A) = \sum_{i=1}^{n} a_{ij} C_{ij} = \sum_{i=1}^{n} a_{ij} (-1)^{i+j} \det(A\langle i, j \rangle)$$

DEFINITION 4.4.2 (Adjugate Matrix). Let $A \in \text{Mat}(n; R)$ where R is a commutative ring. The **adjugate matrix** adj(A) is the $(n \times n)$ -matrix whose entries are $adj(A)_{ij} = C_{ij}$ where C_{ij} is the (i, j)-cofactor.

THEOREM 4.4.4 (Cramer's Rule). Let $A \in Mat(n; R)$ where R is a commutative ring. Then

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

COROLLARY 4.4.1 (invertibility of Matrices). A square matrix with entries in a commutative ring R is invertible i.f.f. its determinant is a unit in R. That is, $A \in Mat(n; R)$ is invertible i.f.f. $det(A) \in R^{\times}$.

4.5 Eigenvalues and Eigenvectors

DEFINITION 4.5.1. Let $f: V \to V$ an endomorphism of an F-vector space V. A scalar $\lambda \in F$ is an eigenvalue of f i.f.f. $\exists \vec{v} \in V \quad \vec{v} \neq \vec{0}$ sych that $f(\vec{v}) = \lambda \vec{v}$. Each such vector is called an eigenvector of f with eigenvalue λ . For any $\lambda \in F$, the eigenspace of f with eigenvalue λ is

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

THEOREM 4.5.1 (Existance of Eigenvalues). Each endomorphism of a non-zero finite dimensional vector space over an algebraically closed field has an eigenvalue.

DEFINITION 4.5.2. Let R be a commutative ring and let $A \in \operatorname{Mat}(n; R)$ be a square matrix with entries in R. The polynomial $\det(A - xI_n) \in R[x]$ is called the **characteristic polynomial** of the matrix A. It is denoted by $\chi_A(x) := \det(A - xI_n)$.

Where χ stands for χ aracteristic (Okay Iain you're a good dad for sure...).

THEOREM 4.5.2 (Eigenvalues and Characteristic Polynomials). Let F be a field and $A \in \operatorname{Mat}(n; F)$ with entries in F. The eigenvalues of the linear mapping $A : F^n \to F^n$ are exactly the roots of the characteristic polynomial χ_A .

4.6 Triangularisable, Diagonalisable, and the Cayley-Hamilton Theorem

Iain (not Ian just to be clear) said he's not examing triangularisable. (Will FC)

PROPOSITION 4.6.1 (Triangularisability). Let $f: V \to V$ be an endomorphism of a finite dimensional F-vector space V. The following two statements are equivalent:

1. The vector space V has an ordered basis $\mathcal{B} = (\vec{v_1}, \vec{v_2}, ..., \vec{v_n})$ such that

$$f(\vec{v_1}) = a_{11}\vec{v_1}$$

$$f(\vec{v_2}) = a_{12}\vec{v_1} + a_{22}\vec{v_2}$$

$$\vdots$$

$$f(\vec{v_n}) = \sum_{i=1}^n a_{in} \vec{v_i} \in V$$

such that the $n \times n$ matrix $\mathfrak{B}[f]_{\mathfrak{B}} = (a_{ij})$ representing f with respect to \mathfrak{B} is upper triangular

$$A = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ 0 & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & a_{nn} \end{pmatrix}$$

When this happens, we say that f is **triangularisable**.

2. The characteristic polynomial $\chi_f(x)$ of f decomposes into linear factors in F[x].

DEFINITION 4.6.1 (Diagonalisibility). As endomorphism $f: V \to V$ of an F-vector space V is **diagonalisable** i.f.f. \exists basis of V consisting of eigenvectors of f. If V is finite dimensional then this is the same as saying that \exists an order basis $\mathcal{B} = \{\vec{v_1}, ..., \vec{v_n}\}$ such that corresponding matrix representing f is diagonal, that is $\mathcal{B}[f]_{\mathcal{B}} = \operatorname{diag}(\lambda_1, ..., \lambda_b)$. In this case, $f(\vec{v_1}) = \lambda_1 \vec{v_1}$.

A square matrix is **diagonalisable** i.f.f. the corresponding linear mapping $F^n \to F^n$ given by left multiplication by A is diagonalisable. This means \exists invertable matrix $P \in GL(n; F)$ such that $P^{-1}AP = \operatorname{diag}(\lambda_1, ..., \lambda_n)$

LEMMA 4.6.1 (Linear Independence of Eigenvectors). Let $f: V \to V$ be an endomorphism of a vector space V and let $\vec{v_1}, ..., \vec{v_n}$ be eigenvectors of f with pairwise different eigenvalues $\lambda_1, ..., \lambda_n$. Then the vectors are linearly independent.

THEOREM 4.6.1 (Cayley-Hamilton Theorem). Let $A \in \text{Mat}(n; R)$ be a square matrix with entries in a commutative ring R. The evaluating its characteristic polynomial $\chi_A(x) \in R[x]$ at the matrix A gives zero.

4.7 Google's PageRank Algorithm (Markov matrix/stochastic matrix)

DEFINITION 4.7.1. A matrix *M* whose entries are non-zero and such that the sum of the entries of each column equal 1 is a **Markov matrix** or a **stochastic matrix**.

LEMMA 4.7.1. Suppose $M \in \mathrm{Mat}(n;\mathbb{R})$ is a Markov matrix. Then 1 is an eigenvalue of M

THEOREM 4.7.1 (Perron, 1907). If $M \in \operatorname{Mat}(n; \mathbb{R})$ is a Markov matrix all whose entries are positive, then eigenspace E(1, M) is one dimensional. There exists a unique basis vector $\vec{v} \in E(1, M)$ all of whose entries are positive real numbers, $v_i > 0$ for all i, and such that the sum of it entries is $1, \sum_i v_i = 1$

5 Inner Product Spaces

5.1 Inner Product Space: Definitions

DEFINITION 5.1.1. Let V be a vector space over \mathbb{R} . An **inner product** on V is a mapping $(-,-): V \times V \to \mathbb{R}$ that satisfies the following $\forall \vec{x}, \vec{y}, \vec{z} \in V$ and $\lambda, \mu \in \mathbb{R}$:

- 1. $(\lambda \vec{x} + \mu \vec{y}, z) = \lambda(\vec{x}, \vec{z}) + \mu(\vec{y}, \vec{z})$
- 2. $(\vec{x}, \vec{y}) = (\vec{y}, \vec{x})$
- 3. $(\vec{x}, \vec{x}) \ge 0$, with $\vec{x} = \vec{0} \Leftrightarrow (\vec{x}, \vec{x}) = 0$

A real innter space product space is a real vector space endowed with an inner product.

EXAMPLE 5.1.1.

• A dot product is a statdard inner product

DEFINITION 5.1.2. Let V be a vector space over \mathbb{C} . An **inner product** on V is a mapping $(-,-): V \times V \to \mathbb{C}$ that satisfies the following $\forall \vec{x}, \vec{y}, \vec{z} \in V$ amdd $\lambda, \mu \in \mathbb{C}$:

- 1. $(\lambda \vec{x} + \mu \vec{y}, z) = \lambda(\vec{x}, \vec{z}) + \mu(\vec{y}, \vec{z})$
- 2. $(\vec{x}, \vec{y}) = \overline{(\vec{y}, \vec{x})}$
- 3. $(\vec{x}, \vec{x}) \ge 0$

Where \overline{z} denotes the complex conjugate of z A complex innter space product space is a complex vector space endowed with an inner product.

EXAMPLE 5.1.2.

• Standard inner product: $(\vec{v}, \vec{w}) = \sum_{i=1}^{n} (v_i \overline{w_i})$

DEFINITION 5.1.3. In a real or complex innter product space the **length** or **inner product norm** or **norm** $\|\vec{v}\| = \sqrt{(\vec{v}, \vec{v})}$.

Vectors whose length is 1 are called **units**. Two vectors \vec{v}, \vec{w} are **orthogonal** can be denoted as $\vec{v} \perp \vec{w}$ i.f.f. $\vec{v}, \vec{w} = 0$

DEFINITION 5.1.4. A family $(\vec{v_i})_{i \in I}$ for vectors from an inner product space is an **orthogonal** family if all the vectors $\vec{v_i}$ have length 1 and if they are pairwise orthogonal to each other, which, using the *Kronecker delta symbol* defined in Example 2.1.2(Check full note), means $(\vec{v_i}, \vec{v_j}) = \delta_{ij}$ An orthogonal family that is a basis is an **orthogonal basis**

THEOREM 5.1.1. Every finite dimensional inner product space has an orthonormal basis

5.2 Orthogonal Complements and Orthogonal Projections

DEFINITION 5.2.1. Let V be an inner product space and let $T \subseteq V$ be an arbitrary subset. Define $T^{\perp} = \{\vec{v} \in V : \vec{v} \perp \vec{t} \mid \forall \vec{t} \in T\}$ calling this set the **orthogonal** to T

PROPOSITION 5.2.1. Let V be an inner product space and let U be a finite dimensional subspace of V. Then U and Y^{\perp} are complementary in the sense of definition 1.7.6 (Check full note). In other words $V = U \oplus U^{\perp}$

DEFINITION 5.2.2. Let U be a finite dimensional subspace of an inner product space V. The space U^{\perp} is the **orthogonal Complement to** U. The **orthogonal projection frome** V **to** U is the mapping $\pi_U: V \to V$ that sends $\vec{v} = \vec{p} + \vec{r}$ to \vec{p}

PROPOSITION 5.2.2. Let U be a finite dimensional subspace of an inner product space V and let π_U be the orthogonal projection form V onto U.

- 1. π_U is a linear mapping with $\operatorname{im}(\pi_U) = U$ and $\ker(\pi_U) = U^{\perp}$
- 2. If $\{\vec{v_1},...,\vec{v_n}\}$ is an orthogonal basis of U, then π_U is given by the following formula for all $\vec{v} \in V$

$$\pi_U(\vec{v}) = \sum_{i=1}^n (\vec{v}, \vec{v_i}) \vec{v_i}$$

3. $\pi_U^2 = \pi_U$, that is π_U is an **idempotent**

THEOREM 5.2.1 (Cauchy-Schwarz Inequality). Let \vec{v}, \vec{w} be vectors in an inner product space. Then $|(\vec{v}, \vec{w})| \leq ||\vec{v}|| ||\vec{w}||$ with equality i.f.f. \vec{v}, \vec{w} are linearly dependent.

COROLLARY 5.2.1. The norm $\|\cdot\|$ on an inner product space V satisfies, for any $\vec{v}, \vec{w} \in V$ and scalar λ :

- 1. $\|\vec{v}\| > 0$ with equality i.f.f. $\vec{v} = \vec{0}$
- $2. \|\lambda \vec{v}\| = |\lambda| \|\vec{v}\|$
- 3. $\|\vec{v} + \vec{w}\| \le \|\vec{v}\| + \|\vec{w}\|$ (triangle inequality)

THEOREM 5.2.2 (Gram-Schmit Process). Let $\vec{v_1}, ..., \vec{v_k}$ be a linearly independent vectors in an inner product space V. Then there exists an orthonormal family $\vec{w_1}, ... \vec{w_k}$ with the property for all $1 \le i \le k$:

$$\vec{w_i} \in \mathbb{R}_{>0} \vec{v_i} + \langle v_i - \vec{1, ..., v_1} \rangle$$

Therefore, $\forall i \in [1, k] \in \mathbb{N} \exists \vec{\lambda} \text{ s.t. } \exists \vec{w_i} = \vec{v_i} - \sum_{j=1}^{i-1} \lambda_j \vec{v_j}$

5.3 Adjoints and Self-Adjoints

DEFINITION 5.3.1. Let V be an inner product space. Then two endomorphism $T, S : V \to V$ are called **adjoint** to the other if the following holds for all $\vec{v}, \vec{w} \in V$:

$$(T\vec{v}, \vec{w}) = (\vec{v}, S\vec{w})$$

In this say we can express $S = T^*$ and call S the adjoint of T. If $S = T^*$ then $T = S^*$

THEOREM 5.3.1. Let V be a finite dimensional inner product space. Let $T:V\to V$ be an endomorphism. Then T* exists. That is, \exists a unique linear mapping $T^*:V\to V$ such that for all $\vec{v},\vec{w}\in V$

$$(T\vec{v}, \vec{w}) = (\vec{v}, T^*\vec{w})$$

DEFINITION 5.3.2. An endomorphism of an inner product space $T:V\to V$ is **self adjoint** if $T^*=T$

THEOREM 5.3.2. Let $T: V \to V$ be a self-adjoint linear mapping on an inner product space V.

- 1. Every eigenvalue of T is real.
- 2. If λ and μ are distinct eigenvalues of T with corresponding eigenvectors \vec{v}, \vec{w} , then $(\vec{v}, \vec{w}) = 0$
- 3. T has an eigenvalue

THEOREM 5.3.3 (The Spectral Theorem for Self-Adjoint Endomorphisms). Let V be a finite dimensional inner product space and let $T:V\to V$ be a self-adjoint linear mapping. Then V has an orthonormal basis consisting of eigenvectors of T

DEFINITION 5.3.3 (Orthogonal Matrix). An **orthogonal matrix** is a square matrix P with real entries such that $P^{-1} = P^T$

DEFINITION 5.3.4. The **orthogonal group** is defined $O(n) = \{P \in Mat(n; \mathbb{R}) : P^{\mathsf{T}}P\}$

COROLLARY 5.3.1 (The Spectral Theorem for Real Symmetric Matrices). Let A be a real $(n \times n)$ -symmetrical matrix. Then \exists an $(n \times n)$ -orthogonal matrix P such that

$$P^{\mathsf{T}}AP = P^{-1}AP = \operatorname{diag}(\lambda_1, ..., \lambda_n)$$

Where $\lambda_1, ..., \lambda_n$ are the (necessarily real) eigenvalues of A, repeated according to their multiplicity as roots of the characteristic polynomial of A

DEFINITION 5.3.5 (Unitary Matrix). An **unitary matrix** is an $(n \times n)$ -matrix P with complex entries such that $\overline{P}^T P = I_n$. In other words, $P^{-1} = \overline{P}^T$

COROLLARY 5.3.2 (The Spectral Theorem for Hermitian Matrices). Let A be a $(n \times n)$ -hermitian matrix. Then there is an $(n \times n)$ -unitary matrix P such that

$$\overline{P}^T AP = P^{-1}AP = \operatorname{diag}(\lambda_1, ..., \lambda_n)$$

Where $\lambda_1, ..., \lambda_n$ are the (necessarily real) eigenvalues of A, repeated according to their multiplicity as roots of the characteristic polynomial of A

6 Jordan Normal Form

Will not be incuded in Exam yay

Quick References

• **Abelian Group**: For group in with operation (A, *) to be abelian group, the following has to be fulfilled:

1. Closure: $\forall a, b \in A, a * b \in A$

2. Associativity: $\forall a, b, c \in A, (a * b) * c = a * (b * c)$

3. Identity: $\exists e \in A \text{ s.t. } e * a = a * e = a \forall a \in A$

4. Inverse: $\forall a \in A \exists a^{-1} \text{ s.t. } a^{-1} * a = a * a^{-1} = e$

5. Communitativity: $\forall a, b \in A, a * b = b * a$

• Change of Basis: for standard basis $A = \{\vec{v_1}, ..., \vec{v_n}\}$ and basis $B = \{(\sum_{i=1}^n a_{i1}\vec{w_i}), ..., (\sum_{i=1}^n a_{in}\vec{w_i})\}$,

$$_A[id]_B = (a_{ij})$$

The blah-morphism table

blah-morphism	Linear?	Bijective?	$f: V \to V$?
Homomorpgism	Yes	Not-required	Not-required
Isomorphism	Yes	Yes	Not-required
Endomorphism	Yes	Not-required	Yes
Automorphism	Yes	Yes	Yes