# 1 Absstractions upon Abstractions

see you guys in UG4 category theory!

## Definition A: Rings and Fields

A ring (left) is a set with two operations  $(\mathbb{R},+,\cdot)$  that satisfies the following lemmas.

A **field** (right) is an extension of a ring where  $(\cdot)$  is a group

- 1. (R, +) is an abelian group with identity 0
- (R,·) is a monoid, i.e. it is a set with Associativity and Identity (written as 1)
- 3. **Distributive law**: For all a, b, and c in F, we have
  - $a \cdot (b+c) = (a \cdot b) + (a \cdot c)$  $(a+b) \cdot c = (a \cdot c) + (b \cdot c)$
- 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 they satisfy the following lemmas (for both):

- 1. 0a = 0 = a0
- 2. The elements 0 and 1 are distinct (only ring case is zero ring)

#### Field Specific Lemmas:

1. (·) in F is associative,  $1_F$  is an identity (it's an abelian group only in  $(F\backslash \{0_F\},\cdot))$ 

## Ring Specific Lemmas and Definitions:

- 1. The **null ring** or **zero ring** is defined as a ring where R is a single element i.e.  $\{0\}$  where 0+0=0 and  $0\times 0=0$
- 2. A **commutative ring** is one where  $a \cdot b = b \cdot a$  for all  $a, b \in R$ 
  - (-a)(b) = -(ab) = a(-b)
  - (-a)(-b) = ab
- m(ab) = (ma)b = a(mb)

• m(na) = (mn)a

- m(a+b) = ma + mb
- m(ab) = (ma)b = a(m
- (m+n)a = ma + na
- (ma)(nb) = (mn)(ab)

## Definition B: Modules and Vector Spaces

A left module M over a ring R (or an R-module) (left) is a pair consisting of an abelian group  $M=(M,\dot+)$  and a mapping A vector space V over a field F (right) is an extension of a module but over a field instead, and using vectors -  $V=(V,\dot+)$ 

$$R \times M \to M : (r, a) \mapsto ra$$

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

s.t.  $\forall r, s \in R$  and  $a, b \in M$ , the following axioms apply:

s.t.  $\forall \lambda, \mu \in F$  and  $\vec{v}, \vec{w} \in v$ , the following axioms apply:

$$r(a \dot{+} b) = (ra) \dot{+} (rb)$$
  

$$(r+s)a = (ra) \dot{+} (sa)$$
  

$$r(sa) = (rs)a$$
  

$$1_R a = a$$

Distributivity 1
Distributivity 2
Associativity
Identity

 $\begin{array}{l} \lambda(\vec{v}\dot{+}\vec{w}) = \lambda\vec{v}\dot{+}\lambda\vec{w} \\ (\lambda+\mu)\vec{v} = \lambda\vec{v}\dot{+}\mu\vec{v} \\ \lambda(\mu\vec{v}) = (\lambda\mu)\vec{v} \\ 1\vec{v} = \vec{v} \end{array}$ 

and they satisfy the following lemmas (for both):

- 1.  $0_R a = 0_M$  for all  $a \in M$  or  $0\vec{v} = \vec{0}$  for all  $\vec{v} \in V$
- 2.  $r0_M = 0_M$  for all  $r \in R$  or  $\lambda \vec{0} = \vec{0}$  for all  $\lambda \in F$
- 3. (-r)a = r(-a) = -(ra) for all  $r \in R$ ,  $a \in M$ 
  - $(-1)\vec{v} = -\vec{v}$  for all  $\vec{v} \in V$

## Definition C: Sub-things

A sub-thing is basically something that is a smaller but self-contained version of a thing

- Vector Subspace (left): A subset U of a vector space V
- Subring (centre): A subset R' of a ring R under the same operations of addition and multiplication defined in R
- Submodule (right): A subset M' of a module M under the same operations of the R-module M restricted to M

Subspace Criteron $\forall \vec{u}, \vec{v} \in U, \lambda \in F$	Subring Criteron $\forall a, b \in R'$	Submod. Criteron $\forall a, b \in M', r \in R$
1. $\vec{0} \in U$	1. R' has a multiplicative identity	$1. \ 0_M \in M'$
$2. \ \vec{u} + \vec{v} \in U$	2. $a - b \in R'$	$2. \ a-b \in M'$
3. $\lambda \vec{u} \in U$	$\begin{vmatrix} 2 & a & b \in R \\ 3 & a \cdot b \in R' \end{vmatrix}$	3. $ra \in M'$

#### Definition D: Homo no homo

Everything has its own homomorphism and they are all the exact same thing

- Linear Mapping (left): Homomorphism on a Vector Space
- Ring Homomorphism (centre): Homomorphism on a ring
- R-homomorphism (right): Homomorphism on a module

V. Space Criteron $\forall \vec{u}, \vec{v} \in U, \lambda \in F$		Module Criteron   $\forall a, b \in M', r \in R$
• $f(\vec{v}_1 + \vec{v}_2) = f(\vec{v}_1) + f(\vec{v}_2)$	f(x+y) = f(x) + f(y)	$\bullet f(a+b) = f(a) + f(b)$
• $f(\lambda \vec{v}_1) = \lambda f(\vec{v}_1)$	$\bullet \ f(xy) = f(x)f(y)$	f(ra) = rf(a)

- A bijective homomorphism is called a **isomorphism**
- Two objects with an iso. are called **isomorphic**, written  $A \cong B$
- A homomorphism  $V \to V$  is called an **endomorphism** of V
- An isomorphism  $V \to V$  is called an **automorphism** of V

#### Image and Kernel

The image and kernel of a mapping  $f: M \to N$  are as follows:

- Image: im  $f = \{f(a) : a \in M\} \subseteq N$
- Kernel:  $\ker f = \{a \in M : f(a) = 0_N\} \subseteq M$

#### Theorem E: Universal Properties and First Iso Thm

# Thm: Universal Properties

Let A be an object of type  $\sigma$ , and I be an ideal-ish  $\sigma$  object

- The mapping can :  $A \to A/I$  sending a to a+I for all  $a \in A$  is a surjective  $\sigma$ -homomorphism with kernel I
- If  $f:A\to B$  is an  $\sigma$ -homomorphism with  $f(I)=\{0_B\}$ , so that  $I\subseteq \ker f$ , then there is a unique  $\sigma$ -homomorphism  $f:A/I\to B$  such that  $f=\overline{f}\circ \operatorname{can}$

#### Thm: First Isomorphism Theorem

Every  $\sigma$  homomorphism  $f: A \to B$  induces an  $\sigma$ -homomorphism

$$\overline{f}: A/\ker f \xrightarrow{\sim} \operatorname{im} f$$

This can be applied to pretty much everything!

- Factor Rings:  $\sigma$  are rings (so A is a ring), and I is an ideal
- Factor Modules:  $\sigma$  are R-modules, and I is a submodule
- Groups:  $\sigma$  are groups, and I is a normal subgroup

# 2 Rings and Modules

#### Example 3.1.4: Modulo Rings

Let  $m \in \mathbb{Z}$ . Then the set of **integers modulo** m is a ring, written

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

The elements of  $\mathbb{Z}/m\mathbb{Z}$  consist of **congruence classes** of integers modulo m, written  $\overline{a}$ , - i.e. "the subsets T of  $\mathbb{Z}$  of the form  $T=a+m\mathbb{Z}$  with  $a\in\mathbb{Z}$ ", or "set of integers that have the same remainder when you divide them by m".  $\overline{a}=\overline{b}$  is the same as  $a-b\in m\mathbb{Z}$ . and often I'll write

$$a \equiv b \mod m$$

Thm 3.1.11 - Prime Property for Fields: Let  $m \in \mathbb{N}$ . The commutative ring  $\mathbb{Z}/m\mathbb{Z}$  is a field if and only if m is prime

## Definition 3.2.3: Multiples of an abelian group

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

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

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

## Definition 3.2: Units and Field Construction

**Def 3.2.6**: Let R be a ring. An element  $a \in R$  is called a **unit** if it is invertible in R, i.e. there exists  $r^{-1} \in R$  such that

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

**Prop 3.2.9**: The set of  $R^{\times}$  units in a ring R forms a group under multiplication

**Definition 3.1.8:** A field is a non-zero commutative ring F in which every non-zero element  $a \in F$  is a unit.

#### Definition 3.2.11: 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.

#### Definition 3.2.12: Integral Domain

An  $\bf integral\ domain$  is a non-zero commutative ring that has no zero-divisors. The following two laws hold:

- 1.  $ab = 0 \implies a = 0 \text{ or } b = 0$
- 2.  $a \neq 0$  and  $b \neq 0 \implies ab \neq 0$

#### Theorem 3.2: Integral Domain Properties

- **3.2.15 (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
- **3.2.16** Let m be a natural number. Then  $\mathbb{Z}/m\mathbb{Z}$  is an integral domain if and only if m is prime.
- 3.2.17 Every finite integral domain is a field.

#### Definition 3.1.1: 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  $m \in \mathbb{Z}$  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<sub>m</sub> is non-zero, the polynomial P has degree m, (written deg(P)), and a<sub>m</sub> is its leading coefficient
- When the leading coefficient is 1 the polynomial is a monic polynomial
- A polynomial of degree one is called linear, degree two is called quadractic, and degree three is called cubic.

Thm 3.3.2: 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: Properties of a Polynomial Ring

- **3.3.3:** 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]
- **3.3.4**: 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 3.3.6: Evaluating a Function

Let R be a commutative ring and  $P \in R[X]$  a polynomial. P can be **evaluated** at  $\lambda \in R$  to make  $P(\lambda)$  by replacing the powers of X in 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 definition of thinking of a polynomial as a function. An element  $\lambda \in R$  is a **root** of P if  $P(\lambda)=0$ 

**Thm 3.3.9:** Let R be a commutative ring, let  $\lambda \in R$  and  $P(X) \in R[X]$ . Then  $\lambda$  is a root of P(X) iff  $(X - \lambda)$  divides P(X)

#### Theorem 3.3.10: Degrees of Polynomial Roots

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 3.3.11: Algebraically closed fields

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

Thm 3.3.13 (Fundamental Thm of Algebra): The field of complex numbers  $\mathbb C$  is algebraically closed.

Thm 3.3.14 (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

## Theorem 3.4.5: Properties of Ring Homomorphisms

Let R and S be rings and  $f: R \to S$  a ring homomorphism. Then for all  $x, y \in R$  and  $m \in \mathbb{Z}$  (where  $0_R$  and  $0_S$  are the zeros of R and S):

1.  $f(0_R) = 0_S$ 

- 4. f(mx) = mf(x)
- 2. f(-x) = -f(x)
- 5.  $f(x^n) = (f(x))^n$  for all
- 3. f(x y) = f(x) f(y)

 $x \in R$  and  $n \in \mathbb{N}$ 

## Definition 3.4: All about Ideals

**Def 3.4.7**:  $I \subseteq R$  is an **ideal**,  $I \subseteq R$ , if the following hold:

- 1.  $I \neq \emptyset$
- 2. I is closed under subtraction
- 3. for all  $i \in I$  and  $r \in R$  we have  $ri, ir \in I$

**Def 3.4.11:** R be a commutative ring and let  $T \subset R$ . Then the **ideal of** R **generated by** T is the set

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

**Thm 3.4.14**: 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

**Def 3.4.15**: Let R be a commutative ring. An ideal I of R is called a **principal ideal** if  $I=\langle t \rangle$  for some  $t\in R$ 

#### Theorem 3.4: Kernels as Ideals

- **3.4.18** Let R and S be rings and  $f:R\to S$  a ring homomorphism. Then  $\ker f$  is an ideal of R.
- **3.4.20** f is injective if and only if ker  $f = \{0\}$
- ${\bf 3.4.21}\,$  The intersection of any collection of ideals of a ring R is an ideal of R
- **3.4.22** Let I and J be ideals of a ring R. Then

$$I + J = \{a + b : a \in I, b \in J\}$$

is an ideal of R

## Definition 3.5.1: Equivalence Relations

A **relation** R on a set X is a subset  $R \subseteq X \times X$ . In the context of relations, it's written xRy instead of  $(x,y) \in R$ . R is an **equivalence relation on** X when for all elements  $x, y, z \in X$  the following hold:

- 1. Reflexivity: xRx
- 2. Symmetry:  $xRy \iff yRx$
- 3. Transivity: xRy and  $yRz \implies xRz$

Suppose that is an equivalence relation on a set X.

- Equivalence class of x:  $E(x) := z \in X : z \sim x$  for  $x \in X$
- Equivalence class for  $\sim$ :  $E \subseteq X$ , if  $\exists x \in X$  s.t. E = E(x)
- Representative: Element of an equivalence class
- System of representatives for  $\sim$ : A subset  $Z \subseteq X$  containing precisely one element from each equivalence class

Given an equivalence relation  $\sim$  on the set X I 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 :  $X \to (X/\sim), \ x \mapsto E(x)$  (surjection)

#### Definition 3.6.1: Coset

Let  $I \triangleleft R$  be an ideal in a 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 w.r.t I in R

Let R be a ring,  $I \subseteq R$  be an ideal, and  $\sim$  the equivalence relation defined by  $x \sim y \iff x - y \in I$ . 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

**Thm 3.6.4**: Let R be a ring and  $I \subseteq R$  an ideal. Then R/I is a ring, where the operation of addition and multiplication is defined by

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

## Theorem 3.7: Submodule lemmas

- **3.7.21** Let  $f:M\to N$  be an R-homomorphism. Then  $\ker f$  is a submodule of M and im f is a submodule of N
- **2.7.22** Let R be a ring, M an R-homomorphism. Then f is injective if and only if  $\ker f = \{0_M\}$

## Definition 3.7.23: Generated Submodules

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

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

together with the zero element in the case  $T=\emptyset$ . If  $T=\{t_1,\ldots,t_n\}$ , a finite set, we write  ${}_R\langle t_1,\ldots,t_n\rangle$  instead of  ${}_R\langle \{t_1,\ldots,t_n\}\rangle$ . M is **finitely generated** if it's generated by a finite set  $M={}_R\langle t_1,\ldots,t_n\rangle$ . M is **cyclic** if it's generated by a singleton  $M={}_R\langle T\rangle$ 

- **3.7.28** Let  $T\subseteq M.$  Then  ${}_R\langle T\rangle$  is the smallest submodule of M that contains T
- **3.7.29** The intersection of any collection of submodules of M is a submodule of M.
- **3.7.30** Let  $M_1$  and  $M_2$  be submodules of a M. Then

$$M_1 + M_2 = \{a + b : a \in M_1, b \in M_2\}$$

is a submodule of M

#### Definition 3.7.31: Submodule Cosets

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

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

It is a coset of N in the abelian group M and so is an equivalence class for the equivalence relation  $a \sim b \iff a - b \in N$ .

Let M/N, the factor of N by N or the quotient of M by N to be the set  $(M/\sim)$  of all cosets of N in M. This becomes an R-module by introducing the operations of addition and multiplication:

$$(a+N)\dot{+}(b+N) = (a+b) + N$$
$$r(a+N) = ra + N$$

for all  $a, b \in M, r \in R$ .

The zero of M/N is the coset  $0_{M/N}=0_M+N$ . The negative of  $a+N\in M/N$  is the coset -(a+N)=(-a)+NThe R-module M/N is the **factor module** of M by the submod. N

## 3 Linear algebra ew

## Definition 1.4.9: Power sets

The set of all subsets  $\mathcal{P}(X) = \{U : U \subseteq X\}$  of X is the **power set** of X,  $\mathcal{P}(X)$  is referred to as a **system of subsets of** X. We can now define 2 new subsets - the **union** and **intersection** 

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

$$\bigcap_{U \in \mathcal{U}} U = \{x \in X : x \in U \text{ for all } U \in \mathcal{U}\}$$

#### Definition 1.4.5: 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  $\alpha_1 \vec{v}_1 + \dots + \alpha_r \vec{v}_r$  with  $\alpha_1, \dots, \alpha_r \in F$  and  $\vec{v}_1, \dots, \vec{v}_r \in T$ , 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  $\vec{v}_1, \ldots, \vec{v}_r$
- 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
- 1.4.7: 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

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

Let A and I be sets. A mapping  $I \to A$  is referred to as a **family of elements of** A **indexed by** I, using the notation  $(a_i)_{i \in I}$ 

## Theorem 1.5.11: Linear combination 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

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

$$(\alpha_1, \dots, \alpha_r) \mapsto a_1 \vec{v}_1 + \dots + \alpha_r \vec{v}_r$$

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

$$\psi = \psi_A : F^r \to V$$

#### Theorem 1.5.12: 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  $\vec{v} \in V$

Crl 1.5.13: Let V be a finitely generated vector space over a field F. Then V has a finite basis

#### Thm 1.5.14: Basis Characterisation Variant

- 1. If  $L \subset V$  is a linearly indep. subset and E is minimal over all generating sets of V where  $L \subseteq E$ , then E is a basis.
- 2. If  $E \subseteq V$  is a generating set and if L is maximal amongst all linearly indep. sets of V where  $L \subseteq E$ , then L is a basis.

### Definition 1.5.15: 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 called the **free vector space on the set** X

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

#### Theorem 1.5.16: 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

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

## Theorem 1.6.1: 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

$$|L| \le |E|$$

## Theorem 1.6: Steinitz Exchange Theorem

**1.6.2**: Let V be a vector space,  $L \subset V$  a finite linearly indep. 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

**1.6.3:** Let V be a vector space,  $M \subseteq V$  a linearly indep. subset, and  $E \subseteq V$  a generating subset, such that  $M \subseteq E$ . If  $\vec{w} \in V \setminus M$  is a vector  $\notin M$  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 1.6.4: Cardinality of Bases

Let V be a finitely generated vector space. V has a finite basis, and any two bases of V also have the same number of elements

**Def 1.6.5**: The cardinality of a basis of a finitely generated vector space V is called the **dimension** of V, written dim V.

#### Theorem 1.6: Dimension Theorems

#### 1.6.7: 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

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

**1.6.9**: If  $U \subseteq V$  is a subspace of an arbitrary vector space, then we have dim  $U < \dim V$ , and if dim  $U = \dim V < \infty$  then U = V

**1.6.10 (The Dimension Theorem)**: Let V be a vector space containing vector subspaces  $U, W \subseteq V$ . Then

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

## Definition 1.7.1: Linear Mappings

**1.7.6**: 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 1.7.7: 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 1.7.8: 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

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

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 1.7.9: 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.  $f \circ q = \mathrm{id}_W$

#### Definition 1.8.1: Image and Kernel of a map

Lemma 1.8.2: A linear mapping is injective iff its kernel is zero

### Theorem 1.8.4: 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  $f = \mathbf{rank}$  of f, dimension of ker  $f = \mathbf{nullity}$  of f

## 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:

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

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  $F^m$ 

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

## Theorem 2.1.8: 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:

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

## Definition 2.2: Big def-thm pairs

Thm 2.2.3: Every square matrix with entries in a field can be written as a product of elementary matrices

**Def 2.2.4:** 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 2.2.5**: 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

**Thm 2.4.5:** Let  $f: V \to W$  be a linear map between finite dim. F-vector spaces. There exists two ordered bases  $\mathcal{A}$  of V, and  $\mathcal{B}$  of W s.t. the representing matrix  $\mathcal{B}[f]_{\mathcal{A}}$  is in Smith Normal Form

**Def 2.2.9**: The **rank** of a matrix  $A \in \operatorname{Mat}(n \times m; F)$ , written  $\operatorname{rk} A$ , is the dim. of the subspace of  $F^n$  generated by the columns of A, or same with the row. The row/column rank are the same. If the rank is equal to the no. of rows/columns, then the matrix has **full rank** 

**Def 2.4.6**: The **trace** of a square matrix is defined to be the sum of its diagonal entries, denoted by tr(A)

## 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

$$f(\vec{v}_i) = a_{1i}\vec{w}_1 + \dots + a_{ni}\vec{w}_n \in W$$

This makes a bijection, which is 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 2.3.4: Representation of a vector image

Let V be a finite dimensional vector space with an ordered basis  $\mathcal{A} = (\vec{v}_1, \dots, \vec{v}_m)$ . We'll denote the inverse to the bijection in 3 " $\Phi_A : F^m \xrightarrow{\sim} V, (\alpha_1, \dots, \alpha_m)^T \mapsto \alpha_1 \vec{v}_1 + \dots + \alpha_m \vec{v}_m$ " by

$$\vec{v} \mapsto \Delta [\vec{v}]$$

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  $\mathcal{A}, \mathcal{B}$  and let  $f: V \to W$  be a linear mapping. The following holds for  $\vec{v} \in V$ :

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

#### Definition 2.4.1: Change of Basis Matrix

Let  $\mathcal{A} = (\vec{v}_1, \dots, \vec{v}_n)$  and  $\mathcal{B} = (\vec{w}_1, \dots, \vec{w}_n)$  be ordered basics of the same F-vector space V. Then the matrix representing the identity mapping w.r.t. these 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 2.4.3: Change of Basis

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

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

**Crl 2.4.4**: Let V be a finite dimensional vector space and let  $f:V\to V$  be an endomorphim of V. Suppose that  $\mathcal{A},\mathcal{A}'$  are ordered bases of V. Then

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

## Definition 4.1.1: Symmetric Groups

The group of all permutations of the set  $\{1, 2, \ldots, n\}$ , also known as bijections from  $\{1, 2, \ldots, n\}$  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  $1 \le i \le j \le n$  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 4.1.5: Multiplicativity of the sign

For each  $n \in \mathbb{N}$  the sign of a permutation produces a group homomorphism 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 4.1.6: 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  $A_n$ 

#### 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  $\lambda \in F$ :

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

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

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

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

A bilinear form H is **symmetric** is U = V and

$$H(u, v) = H(v, u)$$
 for all  $u, v \in U$ 

while it is antisymmetric or alternating if U = V and

$$H(u, u) = 0$$
 for all  $u \in U$ 

- antisymmetric  $\implies H(u,v) = -H(v,u)$
- $H(u,v) = -H(v,u) \implies \text{antisymmetric iff } 1_F + 1_F \neq 0_F$

#### Definition 4.3.3: 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 multi**linear** if for each j, the mapping  $V_i \to W$  defined by  $v_i \mapsto H(v_1, \ldots, v_i, \ldots, v_n)$ , with the  $v_i \in V_i$  arbitrary fixed vectors of  $V_i$  for  $i \neq j$  is linear.

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

#### Theorem 4.3.6: 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

#### Theorem 4.4: Determinant Theorem Bank

**4.4.1**: Let R be a commutative ring,  $A, B \in Mat(n; R)$ . Then

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

**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**: - If A is invertible then  $det(A^{-1}) = det(A)^{-1}$ 

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

**4.4.4**: For all  $A \in Mat(n; R)$  with R a commutative ring,

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

#### Definition 4.4.6: Cofactors of a Matrix

Let  $A \in Mat(n; R)$  for some commutative ring R and  $n \in \mathbb{N}$ . Let  $i, j \in \mathbb{Z}$  between 1 and n. Then the (i, j) cofactor of A is  $C_{ij} = (-1)^{i+j} \det(A(i,j))$  where A(i,j) 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 4.4.7: 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

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

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

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

## Definition 4.4.8: 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 4.4.9: Cramer's Rule

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

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

## Theorem 4.4.11: 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,  $A \in Mat(n; R)$  is invertible if and only if  $det(A) \in \mathbb{R}^{\times}$ 

#### Theorem 4.4.14: Jacobi's Formula

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

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

## Theorem 4.5.4: Existence of Eigenvalues

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

## Definition 4.5.6: Characteristic Polynomial

Let R be a commutative ring and let  $A \in Mat(n; R)$  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

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

**Thm:** 4.5.8: Let F be a field and  $A \in Mat(n; F)$  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

#### Theorem 4.5.9: Eigenvalue Remarks

• Square matrices  $A, B \in Mat(n; R)$  of same size are **conjugate** if

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

for an invertible  $P \in GL(n; R)$ 

- Conjugacy is an equivalence relation on Mat(n; R)
- The char. polynomials for two conjugate matrices are the same
- We can define the char. polynomials of an endomorphism  $f: V \to V$  of an n-dim vector space over a field F to be

$$\chi_f(x) = \chi_{\mathcal{A}}(x) \in F[x]$$

with  $A = A[f]_A \in Mat(n; R)$  the matrix of f w.r.t any basis A for V. The E.V.s of f are exactly the roots of  $\chi_f$ 

#### Theorem 4.5.10: Extending Bases

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:

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

The char. poly. of f is the product of the char. poly.s of g and h

#### 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)

## Theorem 4.6.1 - 4.6.3

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]

Finding ordered bases - Choose from the following subspaces

- 1.  $W = \{\mu \vec{v}_1 \mid \mu \in F\} \subseteq V$
- 2.  $W' = \ker(f \lambda 1_V)$ . This has a basis of E.Vs  $\{\vec{v}_1, \dots, \vec{v}_r\}$
- 3.  $W'' = im(\lambda 1_V f)$

Then extend the basis to another ordered basis  $\mathcal{B}$  for V(the full space) where  $\operatorname{can}(\vec{v}_j) = \vec{u}_j$  forms a basis for V/W.  $\mathcal{B}[f]\mathcal{B}$  is upper triangular.

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

#### Definition 4.6.6: Diagonalisability

An endomorphism  $f: V \to V$  of an F-vector space V is **diagonalisable** iff 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  $\mathcal{B} = \{\vec{v}_1, \ldots, \vec{v}_n\}$  where  $\mathcal{B}[f]_{\mathcal{B}} = \operatorname{diag}(\lambda_1, \ldots, \lambda_n)$ . 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** iff A is conjugate to a diagonal matrix, i.e. there exists  $P \in \operatorname{GL}(n;F)$  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 4.6.9: 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 4.6.10: 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.

#### Definition 4.7.5: Markov Matrix

A matrix M whose entires are non-negative and s.t. the sum of the entries of each column equals 1 is a **Markov matrix** or a **stochastic matrix** 

**4.7.6:** Suppose  $M \in \operatorname{Mat}(n; \mathbb{R})$  is a M.M. Then  $\lambda = 1$  is an e.v.

#### Theorem 4.7.10: Perron-Frobenius Theorem

If  $M\in \operatorname{Mat}(n;\mathbb{R})$  is a Markov matrix with positive values, then the eigenspace E(1,M) is one-dimensional. There exists a unique basis vector  $\vec{v}\in E(1,M)$  with positive real entries s.t. the sum of its entries is 1

# 4 Inner Product Spaces

#### Definition 5.1.1: Inner Product

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 for all  $\vec{x}, \vec{y}, \vec{z} \in V$  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.  $(\vec{x}, \vec{x}) > 0$ , with equality iff  $\vec{x} = \vec{0}$

A **real inner product space** is a real vector space equipped with an inner product. **Note**: basically a generalisation of dot prod.

A complex inner product space is a complex vector space equipped with an inner product. This is the exact same, but condition 2 uses  $(\vec{x}, \vec{y}) = \overline{(\vec{y}, \vec{x})}$  where  $\overline{z}$  is the complex conjugate

## Definition 5.1.5: Norm

In a real or complex inner product space, the **length** or **inner product norm** or **norm**  $\|\vec{v}\| \in \mathbb{R}$  of a vector  $\vec{v}$  is defined as the non-negative square root

$$\|\vec{v}\| = \sqrt{(\vec{v}, \vec{v})}$$

Vectors whose length are 1 are called **units**. Two vectors  $\vec{v}$ ,  $\vec{w}$  are **orthogonal**, written  $\vec{v} \perp \vec{w}$ , iff  $(\vec{v}, \vec{w}) = 0$ 

The norm  $\|\cdot\|$  on an inner product spaces V satisfies, for any  $\vec{v},\vec{w}\in V$  and scalar  $\lambda\colon$ 

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

## Definition 5.1.7: Orthonormal Family

A family  $(\vec{v_i})_{i\in I}$  for vectors from an inner product space is an **orthonormal family** if all the vectors  $\vec{v_i}$  have length 1 and if they are pairwise orthogonal to each other, which, if  $\delta_{i,j}$  is the **Kronecker delta** defined by

$$\delta_{i,j} = \begin{cases} 1 & i = j \\ 0 & \text{otherwise} \end{cases}$$

means that  $(\vec{v}_i, \vec{v}_j) = \delta_{ij}$ .

An orthonormal family that has a basis is an orthonormal basis

Thm 5.1.10: Every finite dimensional inner product space has an orthonormal basis

#### Definition 5.2.1: Orthogonals to a Subset

Let V be an inner product space and let  $T\subseteq V$  be an arbitrary subset. Define

$$T^{\perp} = \{ \vec{v} \in V : \vec{v} \perp \vec{t} \, \forall \vec{t} \in T \}$$

calling this set the orthogonal to T

## Theorem 5.2.2: Complementary Othorgonals

Let V be an inner product space and let U be a finite dimensional subspace of V. Then U and  $U^{\perp}$  are complementary in the sense of 3. i.e.  $V=U\oplus U^{\perp}$ 

#### Definition 5.2.3: Orthogonal Projection

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 from** V **onto** U is the map

$$\pi_U: V \to V$$

that sends  $\vec{v} = \vec{p} + \vec{r}$  to  $\vec{p}$ 

**Prop 5.2.4**: Let U be a finite dimensional subspace of an inner product space V and let  $\pi_U$  be the orthogonal projection from V onto U

- 1.  $\pi_U$  is a linear mapping with  $\operatorname{im}(\pi_U) = U$  and  $\ker(\pi_U) = U^{\perp}$
- 2. If  $\{\vec{v}_1,\dots,\vec{v}_n\}$  is an orthonormal basis of U, then  $\pi_U$  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,  $\pi_U$  is an idempotent

## Theorem 5.2.5: Cauchy-Shwarz Inequality

Let  $\vec{v}, \vec{w}$  be vectors in an inner product space. Then

$$|(\vec{v}, \vec{w})| < ||\vec{v}|| ||\vec{w}||$$

with equality if and only if  $\vec{v}$  and  $\vec{w}$  are linearly dependent

#### Theorem 5.2.7: Gram-Shmidt Process

Let  $\vec{v}_1, \ldots, \vec{v}_k$  be linearly independent vectors in an inner product space V. Then there exists an orthonormal family  $\vec{w}_1, \ldots, \vec{w}_k$  with the property that for all  $1 \leq i \leq k$ ,

$$\vec{w}_i \in \mathbb{R}_{>0} \vec{v}_i + \langle \vec{v}_{i-1}, \dots, \vec{v}_1 \rangle$$

TODO: write how to actually do the gram-shmidt process

## Definition 5.3.1: Adjoints

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

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

In this case I will write  $S = T^*$  and call S the **adjoint** of T

Remark 5.3.2: Any endomorphism has at most one adjoint.

# Theorem 5.3.4

Let V be a finite dimensional inner product space. Let  $T:V\to V$  be an endomorphism. Then  $T^*$  exists. That is, there is 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 5.3.5: Self Adjoints

An endomorphism of an inner product space  $T: V \to V$  is **self-adjoint** if it equals its own adjoint, i.e. if  $T^* = T$ 

## Theorem 5.3.7: Self-Adjoint Theorem bank

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  $\lambda$  and  $\mu$  are distinct eigenvalues of T with corresponding eigenvectors  $\vec{v}$  and  $\vec{w}$ , then  $(\vec{v}, \vec{w}) = 0$
- 3. T has an eigenvalue

#### Definition 5.3.11: Orthogonal Matrices

An **Orthogonal matrix** is an  $(n \times n)$ -matrix P with real entries such that  $P^T P = I_n$ , or in other words such that  $P^{-1} = P^T$ 

## Definition 5.3.14: Complex Matrices

A **hermitian matrix** is one that is self-adjoint in  $\mathbb{C},$  or in other words one where  $A=\overline{A}^T$  holds

An **unitary matrix** is an  $(n \times n)$ -matrix P with complex entries such that  $\overline{P}^T P = I_n$ , or such that  $P^{-1} = \overline{P}^T$ 

## Theorem 5.3.9: Spectral Theorems

**5.3.9**: 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 eigenvalues of T.

**5.3.11:** The Spectral Theorem for Real Symmetric Matrices Let A be a real  $(n \times n)$ -symmetric matrix. Then there is an  $(n \times n)$ -orthogonal matrix P such that

$$P^T A P = P^{-1} A P = \operatorname{diag}(\lambda_1, \dots, \lambda_n)$$

where  $\lambda_1, \ldots, \lambda_n$  are the (necessarily real) eigenvalues of A, repeated according to their multiplicity as roots of  $\chi_A$ 

**5.3.15**: 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

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

where  $\lambda_1, \ldots, \lambda_n$  are the (necessarily real) eigenvalues of A, repeated according to their multiplicity as roots of  $\gamma_A$ 

# 5 Jordan Normal Form

#### Definition 6.2.1: Jordan Blocks

Given an integer  $r \geq 1$  define an  $(r \times r)$ -matrix J(r) called the **nilpotent Jordan block of size** r, by the rule  $J(r)_{ij} = 1$  for j = i + 1 AND  $J(r)_{ij} = 0$  otherwise In particular, J(1) is a  $(1 \times 1)$ -matrix whose only entry is zero.

Given an integer  $r\geq 1$  and a scalar  $\lambda\in F$ , define an  $(r\times r)$ -matrix  $J(r,\lambda)$  called the **Jordan block of size** r **and eigenvalue**  $\lambda$  by the rule

$$J(r, \lambda) = \lambda I_r + J(r) = D + N$$

with  $\lambda I_r=\mathrm{diag}(\lambda,\lambda,\dots,\lambda)=D$  diagonal and J(r)=N nilpotent such that DN=ND

#### Theorem 6.2.2: Jordan Normal Form

Let F be an algebraically closed field. Let V be a finite dimensional vector space and let  $\phi:V\to V$  be an endomorphism of V with characteristic polynomial

$$\chi_{\phi}(x) = (x - \lambda_1)^{a_1} (x - \lambda_2)^{a_2} ... (x - \lambda_s)^{a_s} \in F[x], a_i \ge 1, \sum_{i=1}^s a_i = n$$

For distinct  $\lambda_1, \lambda_2, \ldots, \lambda_s \in F$ . Then there exists an ordered basis  $\mathcal{B}$  of V such that the matrix of  $\phi$  with respect to the block  $\mathcal{B}$  is block diagonal with Jordan blocks on the diagonal,  $_{\mathcal{B}}[\phi]_{\mathcal{B}}$ 

= diag
$$(J(r_{11}, \lambda_1), \dots, J(r_{1m_1}, \lambda_1), J(r_{21}, \lambda_2), \dots, J(r_{sm_s}, \lambda_s))$$

with  $r_{11}, \ldots, r_{1m_1}, r_{21, \ldots, r_{sm_s}} \ge 1$  such that

$$a_i = r_{i_1} + r_{i_2} + \dots + r_{i_{m_i}} \quad (1 \le i \le s)$$

## Theorem 6.3.1: Bézout's identity for polynomials

For a characteristic polynomial

$$\chi_{\phi}(x) = \prod_{i=1}^{s} (x - \lambda_i)^{a_i} \in F[x]$$

where each  $a_i$  is a positive integer,  $\lambda_i \neq \lambda_j$  for  $i \neq j$ , and  $\lambda_i$  are e.v.s of  $\phi$ . For each  $1 \leq j \leq s$  define

$$P_j(x) = \prod_{\substack{i=1\\i\neq j}}^s (x - \lambda_i)^{a_i}$$

There exists polynomials  $Q_j(x) \in F[x]$  such that

$$\sum_{j=1}^{s} P_j(x)Q_j(x) = 1$$

# Definition 6.3.2: Generalised Eigenspace

The generalised eigenspace of  $\phi$  with eigenvalue  $\lambda_i$ ,  $E^{\text{gen}}(\lambda_i, \phi)$  is the following subspace of V:

$$E^{\text{gen}}(\lambda_i, \phi) = \{ \vec{v} \in V \mid (\phi - \lambda_i \operatorname{id}_V)^{a_i}(\vec{v}) = \vec{0} \}$$

The dimension of  $E^{\mathrm{gen}}(\lambda_i,\phi)$  is called the **algebraic multiplicity** of  $\phi$  with eigenvalue  $\lambda_i$  while the dimension of the eigenspace  $E(\lambda_i,\phi)$  is called the **geometric multiplicity** of  $\phi$  with eigenvalue  $\lambda$ 

Remark 6.3.4: The actual eigenspace is defined by

$$E(\lambda_i, \phi) = \{ \vec{v} \in V \mid (\phi - \lambda_i \operatorname{id}_V)(\vec{v}) = \vec{0} \}$$

 $E^{\mathrm{gen}}(\lambda_i, \phi) \subseteq E^{\mathrm{gen}}(\lambda_i, \phi)$ , or the algebraic multiplicity of any e.v. must be greater or equal to the corresponding geometric multiplicity

#### Definition 6.3.4: Stable subsets

Let  $f: X \to X$  be a mapping from a set X to itself. A subset  $Y \subseteq X$  is **stable under** f precisely when  $f(Y) \subseteq Y$ , that is if  $y \in Y$  then  $f(y) \in Y$ .

#### Theorem 6.3.5: Direct Sum Composition

For each  $1 \le i \le s$ , let

$$\mathcal{B}_i = \{ \vec{v}_{ij} \in V \mid 1 \le j \le a_i \}$$

be a basis of  $E^{\text{gen}}(\lambda_i, \phi)$ , where  $a_i$  is the algebraic multiplicity of  $\phi$  with eigenvalue  $\lambda_i$  s.t.  $\sum_{i=1}^s a_i = n$  is the dimension of V.

- 1. Each  $E^{\text{gen}}(\lambda_i, \phi)$  is stable under  $\phi$
- 2. For each  $\vec{v} \in V$  there exist unique  $\vec{v}_i \in E^{\mathrm{gen}}(\lambda_i, \phi)$  such that  $\vec{v} = \sum_{i=1}^s \vec{v}_i$ . In other words, there is a direct sum decomposition

$$V = \bigoplus_{i=1}^{s} E^{\text{gen}}(\lambda_i, \phi)$$

with  $\phi$  restricting to endomorphisms of the summands

$$\phi_i = \phi \mid : E^{\text{gen}}(\lambda_i, \phi) \to E^{\text{gen}}(\lambda_i, \phi)$$

3. Then

$$\mathcal{B} = \mathcal{B}_1 \cup \mathcal{B}_2 \cup \cdots \cup \mathcal{B}_s = \{\vec{v}_{ij} \mid 1 < i < s, 1 < j < a_i\}$$

is a basis of V. The matrix of the endomorphism  $\phi$  w.r.t. this basis is given by the block diagonal matrix

$$_{\mathcal{B}}[\phi]_{\mathcal{B}} = \begin{pmatrix} \begin{array}{c|ccc} B_1 & 0 & 0 & 0 \\ \hline 0 & B_2 & 0 & 0 \\ \hline & & . & \\ 0 & 0 & \ddots & 0 \\ \hline 0 & 0 & 0 & B_s \\ \end{array} \end{pmatrix} \in \operatorname{Mat}(n; F)$$

with  $B_i = \mathcal{B}_i[\phi_i]_{\mathcal{B}_i} \in \operatorname{Mat}(a_i; F)$ 

#### Theorem 6.3: JNF Theorem Bank

**6.3.6**: For each i, define a linear mapping

$$\psi_i: \frac{W_i}{W_{i-1}} \to \frac{W_{i-1}}{W_{i-2}}$$

by  $\psi_i(\vec{w} + W_{i-1}) = \psi(\vec{w}) + W_{i-2}$  for  $\vec{w} \in W_i$ . Then  $\psi_i$  is well-defined and injective

- **6.3.7:** Let  $f: X \to Y$  be an injective linear mapping between the F-vector spaces X and Y. If  $\{\vec{x}_1, \ldots, \vec{x}_t\}$  is a linearly independent set in X, then  $\{f(\vec{x}_1, \ldots, \vec{x}_t)\}$  is a linearly independent set in Y
- **6.3.8**: The set of elements  $\{\vec{v}_{j,k}: 1\leq j\leq m, 1\leq k\leq d_j\}$  constructed in the next algorithm is a basis for W
- **6.3.9**: Let  $\mathcal B$  be the ordered basis of W  $\{\vec v_{j,k}: 1\leq j\leq m, 1\leq k\leq d_j\}$ . Then  $_{\mathcal B}[\psi]_{\mathcal B}=$

diag 
$$\underbrace{J(m),..,J(m)}_{d_m \text{ times}},\underbrace{J(m-1),..,J(m-1)}_{d_m-1-d_m \text{ times}},..,\underbrace{J(1),..,J(1)}_{d_1-d_2 \text{ times}}$$

where J(r) denotes the nilpotent Jordan block of size r

# Theorem 6.3: JNF Basis Algorithm

Algorithm to construct a basis for each  $W_i/W_{i-1}$ :

- Choose an arbitrary basis for  $W_m/W_{m-1}$ , say  $\{v_{m,1} + W_{m-1}, \vec{v}_{m,2} + W_{m-1}, \dots, \vec{v}_m, d_m + W_{m-1}\}$
- Since  $\psi_m: W_m/W_{m-1} \to W_{m-1}/W_{m-2}$  is injective by 6.3.6, 6.3.7 proves that

$$\begin{split} \{\psi(\vec{v}_{m,1}) + W_{m-2}, \psi(\vec{v}_m, 2) + W_{m-2}, .., \psi(\vec{v}_m, d_m + W_{m-2})\} \\ \text{is a linearly independent set in } W_{m-1}/W_{m-2}. \text{ Set} \\ \vec{v}_{m-1,i} &= \psi(\vec{v}_{m,i}) \text{ for } 1 \leq i \leq d_m \end{split}$$

- Choose vectors  $\{\vec{v}_{m-1,i}:d_m+1\leq i\leq d_{m-1}\}$  so that  $\{\vec{v}_{m-1,i}+W_{m-i-1}:1\leq k\leq d_{m-i}\}$  is a basis of  $W_{m-1}/W_{m-2}$
- Repeat!

## 5.1 PageRank, again

#### Theorem 6.5.1

If  $M \in \operatorname{Mat}(n; \mathbb{R})$  is a Markov matrix with all positive entries, consider M as a complex matrix whose entries just happen to be real. If  $\lambda \in \mathbb{C}$  is an eigenvalue of M then either  $\lambda = 1$  or  $|\lambda| < 1$ 

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