# Honours Algebra Quick Notes

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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

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

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  $\lambda, \mu \in F$ 

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

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 and  $\overrightarrow{v} \in V$  then  $0\overrightarrow{v} = \overrightarrow{0}$ 

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

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

## 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

$$\operatorname{pr}_i: X_i \times \ldots \times X_n \to X_i$$
$$(x_i, ..., x_n) \mapsto x_i$$

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

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

## 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 od 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** 

**DEFINITION 1.4.3.** Check Definition 1.4.9 on Iain's (Not me I am Ian) note I dun think it's English

## 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  $\alpha_j \in \{a_1, ..., a_n\}, \alpha_j \neq 0$  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

**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 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.** 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 amongstall linearly independent subsets of vector space with  $L \subseteq E$ , then L is a basis.

**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  $(a_i)_{i \in I}$  od 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, adn  $E \subseteq V$  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
- 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.** 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 **anautomorphism** 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 over a field F is isomorphic to  $F^n$  i.f.f it has dimension 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 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 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  $\dim(V) = \dim(\ker f) + \dim(\operatorname{im}(f))$ 

# 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  $m, nin\mathbb{N}$  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:

$$\mathbf{M}: \mathrm{Hom}_F(F^m, F^n) \overset{\sim}{\to} \mathrm{Mat}(n \times m; F)$$
$$f \mapsto [f]$$

This attaches to each linear mapping f its **representing matrix**  $\mathbf{M}(f) := [f]$ . 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  $f: F^m \to F^n$ belinearmappings. 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** 

#### 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  $\mathcal{A} = (\overrightarrow{v}_1, ..., \overrightarrow{v}_m)$  and  $\mathcal{B} = (\overrightarrow{w}_1, ..., \overrightarrow{w}_n)$ . 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

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

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  $\mathcal{A} = (\overrightarrow{v}_1, ..., \overrightarrow{v}_m)$ . We will denote the invese to the bijection of  $\Phi_{\mathcal{A}} \xrightarrow{\sim} V, (\alpha_a, ..., \alpha_m)^T \mapsto \sum_{i=1}^m \alpha_i \overrightarrow{v}_i$  by

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

**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  $\overrightarrow{v} \in V$ :

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

#### 2.4 Change of Matrix by Change of Basis

**DEFINITION 2.4.1.** Let  $\mathcal{A} = (\overrightarrow{v}_1, ..., \overrightarrow{v}_n), \mathcal{B} = (\overrightarrow{w}_1, ..., \overrightarrow{w}_n)$  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  $\overrightarrow{v}_i = \sum_{i=1}^n a_{ij} \overrightarrow{w}_i$ 

**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 diagonaltherer 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:

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