

# Linear Algebra 1 (TB2) Notes

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*An important note, these notes are absolutely **NOT** guaranteed to be correct, representative of the course, or rigorous. Any result of this is not the author's fault.*

# 1 Vector Spaces, Fields, and Maps

## 1.1 Groups

A group is a *non-empty* set  $(G)$  paired with a *binary group operation*  $(*)$  denoted by  $(G, *)$ . The following properties hold for all groups:

*Let  $(G, *)$  be a group with elements  $f, g, h$ ,*

- **Associativity:**  $f * (g * h) = (f * g) * h$
- **Identity:**  $\exists e \in G : e * f = f * e = f$
- **Inverse:**  $\exists x \in G : x * f = f * x = e$

*A note, for a group  $(G, *)$  with  $g * h = h * g$  for all  $g, h \in G$ , this group is called **commutative** or **abelian**. However, it should be textitased that this is **not** a necessary condition for a group.*

## 1.2 A note on the invertibility of matrices

For a matrix  $A \in M_{m,n}(\mathbb{F})$ , the following are all **equivalent** statements:

- $A$  is **invertible**
- $\det A \neq 0$
- The **rows** of  $A$  are **linearly independent**
- The **columns** of  $A$  are **linearly independent**
- The **reduced row echelon form** of  $A$  is the **identity**
- For all  $\mathbf{b} \in \mathbb{F}^n$ ,  $A\mathbf{x} = \mathbf{b}$  has a **unique solution**

### 1.3 Fields

A field is a set  $(F)$  defined under multiplication and division with the following properties:

- **Associativity** under multiplication and division
- **Commutativity** under multiplication and division
- $F$  contains an **identity** under multiplication and division
- All elements in  $F$  contain an **inverse** under addition and multiplication (except 0 under multiplication)
- The defined multiplication is **distributive** across the defined addition

### 1.4 Vector Spaces

A group  $(V, +_V)$  ( $+_V$  denotes addition defined with respect to the set  $V$  as it can be ambiguous in some cases) is a vector space over the field  $(\mathbb{F})$  if the following holds:

Let  $v, w \in V, \lambda, \mu \in \mathbb{F}$ ,

- $(V, +_V)$  is **abelian**
- $V$  is **closed under multiplication** with elements in  $\mathbb{F}$
- $\lambda(v +_V w) = \lambda v + \lambda w$
- $(\lambda + \mu)v = \lambda v +_V \mu v$
- $(\lambda\mu)v = \lambda(\mu v)$
- $fv = v$  where  $f$  is the **multiplicative identity** of  $\mathbb{F}$

## 1.5 Subspaces

Let  $V$  be a vector space over  $\mathbb{F}$ ,  $U \subseteq V$  is a subspace if the following properties hold:

- $U$  is **non-empty**
- $U$  is **closed** under the **addition** defined by  $V$
- $U$  is **closed** under the **multiplication** defined by  $V$

*Some notes on subspaces:*

- Subspaces are vector spaces
- The intersection of subspaces is a subspace
- The span of any non-empty subset of a given vector space is a subspace

## 1.6 Linear Maps

For  $V, W$  vector spaces over  $\mathbb{F}$ , the map  $T : V \rightarrow W$  is called linear if the following properties hold:

*Let  $u, w \in V$ ,  $\lambda \in \mathbb{F}$ ,*

- $T(u + v) = T(u) + T(v)$
- $T(\lambda u) = \lambda T(u)$

*A note, for a linear map ( $T : V \rightarrow W$ ), if  $V = W$ ,  $T$  is sometimes referred to as a linear **operator**. Also, composed linear maps are also linear maps*

## 1.7 The Kernel and Image

For a linear map ( $T : V \rightarrow W$ ), the kernel is defined as follows:

$$\text{Ker } T = \{v \in V : T(v) = 0\}$$

The image is defined as follows:

$$\text{Im } T = \{w \in W : \exists v \in V \text{ with } T(v) = w\}$$

*Some notes on linear maps (let  $T : V \rightarrow W$  be a linear map):*

- The kernel and image of  $T$  are subspaces of  $V$  and  $W$  respectively
- For  $U \subseteq V$ ,  $T(U)$  is also a subspace (but of  $W$  instead of  $V$ )

## 1.8 Bases and Dimension

For  $V$  a vector space, with  $S \subseteq V$ ,

Let  $s_1, s_2, \dots \in S$ ,

- $S$  is linearly independent if  $\sum_{n=1}^{|S|} \lambda_n s_n = 0 \iff \lambda_i = 0 \ \forall i$
- $S$  is linearly dependent if it's not linearly independent

*A note, if  $S$  is linearly dependent, there's a vector in  $S$  such that it can be written as the sum of other vectors in  $S$ .*

A result of linear dependence is that for a linear dependent set  $S$ , there exists  $s \in S$  such that  $\text{span}(S) = \text{span}(S \setminus \{s\})$