

## 1. BASIC NOTIONS

**Definition 1.1.** A binary operation on a set  $S$  is a function  $f : S \times S \rightarrow S$ .

**Definition 1.2.** A *field* is a set  $\mathbb{F}$  together with two binary operations  $+$ , and  $\cdot$  called addition and multiplication (respectively) such that

1. For all  $a, b, c \in \mathbb{F}$  we have

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

and

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

2. For all  $a, b \in \mathbb{F}$  we have

$$a + b = b + a$$

and

$$a \cdot b = b \cdot a.$$

3. There exists an element  $0 \in \mathbb{F}$ , called an additive identity, such that for all  $a \in \mathbb{F}$  we have  $a + 0 = a$ .
4. There exists an element  $1 \in \mathbb{F}$ , called a multiplicative identity, such that for all  $a \in \mathbb{F}$  we have  $a \cdot 1 = a$ .
5. For all  $a \in \mathbb{F}$  there exists an element  $b \in \mathbb{F}$ , called an additive inverse, such that  $a + b = 0$ .
6. For all  $a \in \mathbb{F}$  such that  $a \neq 0$  there exists an element  $c \in \mathbb{F}$ , called a multiplicative inverse, such that  $a \cdot c = 1$ .
7. For all  $a, b, c \in \mathbb{F}$

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

**Note:** Fields have a unique additive and multiplicative identity denoted 0 and 1 respectively. Moreover, when the additive and multiplicative inverses exist they are unique.

**Some examples:** All of the following examples are with their standard operations.

1.  $\mathbb{Q}$  (rational numbers)
2.  $\mathbb{R}$  (real numbers)
3.  $\mathbb{C}$  (complex numbers)
4.  $\mathbb{Z}/p\mathbb{Z}$  for  $p$  prime (Integers modulo  $p$ )

**Non example:**  $\mathbb{Z}$  is not a field, it lacks multiplicative inverses.

**Definition 1.3.** A *vector space*  $V$  over a field  $\mathbb{F}$  is a set  $V$  with two operations called *vector addition* and *scalar multiplication* where vector addition is a function  $+: V \times V \rightarrow V$  and scalar multiplication is a function  $\cdot: \mathbb{F} \times V \rightarrow V$  such that

1. For all  $u, v \in V$  we have

$$u + v = v + u$$

2. For all  $u, v, w \in V$  and for all  $a, b \in \mathbb{F}$  we have

$$(u + v) + w = u + (v + w)$$

and

$$(ab) \cdot v = a \cdot (b \cdot v)$$

3. There exists a vector  $0 \in V$ , called an additive identity, such that for all  $v \in V$  we have

$$v + 0 = v$$

4. For all  $v \in V$  we have a vector  $w \in V$ , called an additive inverse, such that

$$v + w = 0$$

5. For all  $v \in V$  we have

$$1 \cdot v = v$$

6. For all  $a, b \in \mathbb{F}$  and for all  $u, v \in V$  we have

$$a \cdot (u + v) = a \cdot u + a \cdot v$$

**Some examples:** All of the following examples are with their standard operations.

1.  $\mathbb{F}^n = \left\{ \begin{bmatrix} a_1 \\ \vdots \\ a_n \end{bmatrix} : a_i \in \mathbb{F} \right\}$  where  $\mathbb{F}$  is a field.

2. Polynomials with coefficients in a field  $\mathbb{F}$ .

3. Polynomials (with coefficients in a field  $\mathbb{F}$ ) of degree  $\leq n$

4. Continuous functions  $f: X \rightarrow Y$ ,  $C(X, Y)$ , where  $X$  and  $Y$  are fields.

5. Functions from a field  $X$  into a field  $Y$ .

6.  $\mathbb{F}^\infty = \{(a_1, a_2, a_3, \dots) : a_i \in \mathbb{F}\}$ .

**Proposition 1.1.** *Every vector space  $V$  has a unique additive identity. The unique additive identity is denoted  $0$ .*

**Proposition 1.2.** *Every element  $v \in V$  has a unique additive inverse. For all  $v \in V$  its unique additive inverse is denoted  $-v$ .*

**Proposition 1.3.** *For all  $v \in V$  we have  $0 \cdot v = 0$ .*

**Proposition 1.4.** *For all  $a \in \mathbb{F}$  and  $0 \in V$  we have  $a \cdot 0 = 0$ .*

**Proposition 1.5.** *For every  $v \in V$  we have  $(-1) \cdot v = -v$*

## 2. BASIS FOR A VECTOR SPACE

**Definition 2.1.** A *linear combination* of a list of vectors  $v_1, \dots, v_m$  in  $V$  is a vector of the form

$$a_1v_1 + \dots + a_mv_m$$

where  $a_1, \dots, a_m \in \mathbb{F}$ .

**Definition 2.2.** The set of all linear combinations of a list of vectors  $v_1, \dots, v_m$  in  $V$  is called the *span* of  $v_1, \dots, v_m$  denoted by  $\text{span}\{v_1, \dots, v_m\}$ .

$$\text{span}\{v_1, \dots, v_m\} = \{a_1v_1 + \dots + a_mv_m \mid a_i \in \mathbb{F}\}$$

**Definition 2.3.** If  $V$  is a vector space and  $V = \text{span}\{v_1, \dots, v_m\}$  then we say that  $v_1, \dots, v_m$  span  $V$ .

**Definition 2.4.** We say that a vectors space is *finite dimensional* if there exists a finite list of vectors  $v_1, \dots, v_m$  such that

$$\text{span}\{v_1, \dots, v_m\} = V$$

Otherwise we say that  $V$  is *infinite dimensional*.

**Definition 2.5.** A list of vectors  $v_1, \dots, v_m$  in  $V$  is called *linearly independent* if the only choice of  $a_1, \dots, a_m \in \mathbb{F}$  such that

$$a_1v_1 + \dots + a_mv_m = 0$$

is  $a_1 = a_2 = \dots = a_m = 0$ . A list is called *linearly dependent* if it is not linearly independent.

**Lemma 2.6.** *Suppose that  $v_1, \dots, v_m$  is a linearly dependent list in  $V$ . There exists a  $j \in \{1, \dots, m\}$  such that*

- 1)  $v_j \in \text{span}\{v_1, \dots, v_{j-1}\}$
- 2)  $\text{span}\{v_1, \dots, v_{j-1}, v_j, v_{j+1}, \dots, v_m\} = \text{span}\{v_1, \dots, v_{j-1}, v_{j+1}, \dots, v_m\}$

**Proposition 2.1.** *In a finite dimensional vector space the length of every linearly independent list of vectors is less than or equal to the length of every spanning list of vectors.*

**Definition 2.7.** A basis for a vector space  $V$  is a list of vectors  $\{v_1, \dots, v_n\}$  such that

1.  $\{v_1, \dots, v_n\}$  is linearly independent
2.  $\text{span}\{v_1, \dots, v_n\} = V$ .

**Proposition 2.2.** *A list of vectors  $\{v_1, \dots, v_n\}$  in  $V$  is a basis for  $V$  if and only if every vector  $v \in V$  can be written uniquely in the form*

$$v = a_1v_1 + \dots + a_nv_n$$

*for some  $a_1, \dots, a_n \in \mathbb{F}$ .*

**Proposition 2.3.** *Every spanning list of vectors in  $V$  can be reduced down to a basis.*

**Proposition 2.4.** *Every linearly independent list of vectors in  $V$  can be extended to a basis.*

**Proposition 2.5.** *Any two basis of a finite dimensional vector space  $V$  have the same length.*

**Definition 2.8.** The dimension of a finite dimensional vector space is the length of any basis of the vector space. Denoted  $\dim(V)$ .

**Proposition 2.6.** *Suppose  $V$  is finite dimensional. Every list of linearly independent vectors whose length is equal to the dimension of  $V$  is a basis.*

**Proposition 2.7.** *Suppose  $V$  is finite dimensional. Every spanning list vectors whose length is equal to the dimension of  $V$  is a basis.*

### 3. SUBSPACES

**Definition 3.1.** A subspace of a vector space  $V$  is a subset  $H$  such that  $H$  is a vector space under the same binary relations and field as  $V$ .

**Proposition 3.1** (Subspace Test). *A subset  $H$  is a subspace of  $V$  if and only if*

1.  $0 \in H$ .
2. For all  $u, v \in H$  we have  $u + v \in H$
3. For all  $u \in H$  and  $a \in \mathbb{F}$  we have  $au \in H$ .

**Proposition 3.2.** *If  $U$  is a subspace of a finite dimensional vector space  $V$  then  $\dim(U) \leq \dim(V)$ . Moreover,  $\dim(U) = \dim(V)$  if and only if  $V = U$ .*

**Definition 3.2.** Suppose  $U_1, \dots, U_m$  are subsets of  $V$ . The *sum* of  $U_1, \dots, U_m$  denoted  $U_1 + \dots + U_m$  is the set of all possible sums i.e.,

$$U_1 + \dots + U_m = \{u_1 + \dots + u_m \mid u_i \in U_i, i = 1, \dots, m\}$$

**Proposition 3.3.** *If  $U_1, \dots, U_m$  are subspaces then so is  $U_1 + \dots + U_m$ .*

**Definition 3.3.** Suppose  $U_1, \dots, U_m$  are subspaces of  $V$ . The sum  $U_1 + \dots + U_m$  is a *direct sum* if each element of  $U_1 + \dots + U_m$  can be written in only one way as a sum  $u_1 + \dots + u_m$  where  $u_i \in U_i, i = 1, \dots, m$ . The direct sum is denoted  $U_1 \oplus \dots \oplus U_m$ .

**Proposition 3.4.**  *$U_1 + \dots + U_m$  is a direct sum if and only if the only way to write 0 as a sum is by taking each  $u_i$  where  $i = 1, \dots, m$  to be 0.*

**Proposition 3.5.** *The sum of two subspaces  $U$  and  $W$  is a direct sum if and only if  $U \cap W = \{0\}$ .*

**Proposition 3.6.** *If  $V$  is a finite dimensional vector space and  $U$  is a subspace of  $V$  then there exists a  $W$  which is a subspace of  $V$  such that  $V = U \oplus W$ .*

### 3.1. Quotient Spaces.

**Definition 3.4.** Let  $V$  be a vectors space and  $U$  a subspace. For every  $v \in V$  define

$$v + U = \{v + u \mid u \in U\}$$

and

$$V/U = \{v + U \mid v \in V\}$$

**Proposition 3.7.** *Let  $V$  be a vectorspace,  $U$  a subspace and  $v, w \in V$ . The following are equivalent.*

- (a)  $v - w \in U$
- (b)  $v + U = w + U$
- (c)  $(v + U) \cap (w + U) \neq \emptyset$

**Proposition 3.8.** *Let  $V$  be a vectorspace,  $U$  a subspace,  $\lambda \in \mathbb{F}$ , and  $v, w \in V$  The set  $V/U$  is a vector space with the following operations:*

$$(v + U) + (w + U) = (v + w) + U$$

$$\lambda(v + U) = (\lambda v) + U$$

**Proposition 3.9.** *Let  $V$  be a finite dimensional vector space and  $U$  be a subspace.*

$$\dim(V/U) = \dim(V) - \dim(U)$$

## 4. LINEAR MAPS

**Definition 4.1.** A linear map from  $V(\mathbb{F})$  to  $W(\mathbb{F})$  is a function  $T : V \rightarrow W$  such that

$$T(\lambda x + y) = \lambda T(x) + T(y)$$

for every  $x, y \in V$  and  $\lambda \in \mathbb{F}$ . Denote the set of all linear maps from  $V$  to  $W$  as  $\mathcal{L}(V, W)$ .

**Proposition 4.1.** Suppose  $\{v_1, \dots, v_n\}$  is a basis of  $V$  and  $w_1, \dots, w_n \in W$ . There is a unique linear map  $T : V \rightarrow W$  such that

$$T(v_j) = w_j \quad j = 1, \dots, n.$$

**Proposition 4.2.** If  $T : V \rightarrow W$  is linear then

$$T(0_V) = 0_W.$$

**Definition 4.2.** Suppose  $S, T \in \mathcal{L}(V, W)$  and  $\lambda \in \mathbb{F}$  define

$$(S + T)(v) = S(v) + T(v) \text{ for all } v \in V$$

and

$$(\lambda T)(v) = \lambda T(v) \text{ for all } v \in V.$$

**Proposition 4.3.**  $\mathcal{L}(V, W)$  is a vector space with the above operations.

**Definition 4.3.** For  $T \in \mathcal{L}(V, W)$  define the *null space* of  $T$  (or kernel of  $T$ ) to be

$$\text{null}(T) = \{v \in V \mid Tv = 0\} \subseteq V$$

and define the *range* of  $T$  to be

$$\text{ran}(T) = \{Tv \mid v \in V\} \subseteq W.$$

**Theorem 4.4.** For  $T \in \mathcal{L}(V, W)$  both  $\text{null}(T)$  and  $\text{ran}(T)$  are subspaces. A linear transformation is injective if and only if

$$\text{null}(T) = \{0\}.$$

A linear transformation is surjective if and only if  $\text{ran}(T) = W$ .

**Definition 4.5.** The *rank* of a linear transformation is the dimension of its range. The *nullity* of a transformation is the dimension of its null space.

**Definition 4.6.** A vector space isomorphism from  $V$  onto  $W$  is a bijective linear map  $T : V \rightarrow W$ . If there is a vector space isomorphism from  $V$  onto  $W$  we say  $V$  is isomorphic to  $W$  and write  $V \cong W$ .

**Theorem 4.7.** Let  $V$  and  $W$  be two finite dimensional vector spaces.  $V$  is isomorphic to  $W$  if and only if  $\dim(V) = \dim(W)$ .

**Definition 4.8.** A linear map  $T : V \rightarrow W$  is *invertible* if there exists a map  $S : W \rightarrow V$  such that

$$S \circ T = \text{Id}_V$$

$$T \circ S = \text{Id}_W$$

**Theorem 4.9.** A map  $T : V \rightarrow W$  is invertible if and only if it is bijective.

**Proposition 4.4.** Suppose  $T \in \mathcal{L}(V, W)$  and define

$$\tilde{T} : V/\text{null}(T) \rightarrow W$$

by

$$\tilde{T}(v + \text{null}(T)) = Tv.$$

The following hold:

- a)  $\tilde{T}$  is linear
- b)  $\tilde{T}$  is injective
- c)  $\text{ran}(\tilde{T}) = \text{ran}(T)$
- d)  $V/\text{null}(T) \cong \text{ran}(T)$ .

**Theorem 4.10** (Fundamental Theorem of Linear Maps/ Rank-Nullity). Suppose  $V$  is a finite dimensional vector space and  $T \in \mathcal{L}(V, W)$ . We have  $\text{ran}(T)$  is finite dimensional and

$$\dim(V) = \dim(\text{ran}(T)) + \dim(\text{null}(T)).$$

**Proposition 4.5.** Suppose  $V$  and  $W$  are finite dimensional vector spaces. Let  $T \in \mathcal{L}(V, W)$ .

- a) If  $\dim(V) < \dim(W)$  then  $T$  is not surjective.
- b) If  $\dim(V) > \dim(W)$  then  $T$  is not injective.

**Definition 4.11.** A linear map  $T \in \mathcal{L}(V, V)$  is called an *operator*.

**Theorem 4.12.** Suppose  $T$  is an operator over a vector space  $V$ . If  $V$  is finite dimension the following are equivalent:

- a)  $T$  is injective
- b)  $T$  is surjective
- c)  $T$  is bijective

#### 4.1. The Matrix of a Linear Map and the Coordinate Transform.

**Definition 4.13.** Let  $V$  be an  $n$ -dimensional vector space over the field  $\mathbb{F}$ . Let  $\beta = \{v_1, \dots, v_n\}$  be an *ordered* basis for  $V$ . The coordinate transform  $\varphi_\beta : V \rightarrow \mathbb{F}^n$  is defined by

$$v = a_1v_1 + \dots + a_nv_n \xrightarrow{\varphi_\beta} \begin{bmatrix} a_1 \\ \vdots \\ a_n \end{bmatrix}$$

Denote the column vector  $\varphi_\beta(v)$  by  $[v]_\beta$ .

**Proposition 4.6.** *The coordinate transform from  $V$  onto  $\mathbb{F}^n$  is a vector space isomorphism.*

**Definition 4.14.** Suppose  $T : V \rightarrow W$  is a linear maps with  $\beta = \{v_1, \dots, v_n\}$  and  $\gamma = \{w_1, \dots, w_m\}$  as ordered basis for  $V$  and  $W$  respectively. Define the matrix of  $T$ , denoted by  $[T]_\beta^\gamma$ , by the following If

$$T(v_k) = a_{1,k}w_1 + \dots + a_{m,k}w_m$$

then the  $k$ -th column of  $[T]_\beta^\gamma$  is given by  $\begin{bmatrix} a_{1,k} \\ \vdots \\ a_{m,k} \end{bmatrix}$ .

**Theorem 4.15.** *The following diagram commutes*

$$\begin{array}{ccc} V & \xrightarrow{T} & W \\ \varphi_\beta \downarrow & & \downarrow \varphi_\gamma \\ \mathbb{F}^n & \xrightarrow{[T]_\beta^\gamma} & \mathbb{F}^m \end{array}$$

*More specifically,*

$$[T]_\beta^\gamma[v]_\beta = [T(v)]_\gamma.$$