

# Linear Algebra

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## 1. Vector Spaces

### Definition 1.1: Vector Space

A triple  $(V, +, \cdot)$  consisting of a set  $V$ , an addition map

$$+ : V \times V \rightarrow V, \quad (x, y) \mapsto x + y$$

and a scalar multiplication map

$$\cdot : \mathbb{F} \times V \rightarrow V, \quad (\lambda, x) \mapsto \lambda x$$

is called a real vector space if for all  $x, y, z \in V$  and  $\lambda, \mu \in \mathbb{R}$ , the following axioms hold:

1.  $(x + y) + z = x + (y + z)$
2.  $x + y = y + x$
3.  $x + 0 = x$
4.  $x + (-x) = 0$
5.  $\lambda(\mu x) = (\lambda\mu)x$
6.  $1x = x$
7.  $\lambda(x + y) = \lambda x + \lambda y$
8.  $(\lambda + \mu)x = \lambda x + \mu x$

**Note:** 0 and 1 here denote the additive and multiplicative identities in  $\mathbb{R}$ .

**Definition 1.2: Subspace**

A subset  $U$  of  $V$  is a subspace if:

- $0 \in U$
- $u, w \in U \Rightarrow u + w \in U$
- $a \in \mathbb{F}, u \in U \Rightarrow au \in U$

**Definition 1.3: Sum of Subspaces**

If  $V_1, V_2, \dots, V_m$  are subspaces of  $V$ , their sum is

$$V_1 + V_2 + \dots + V_m = \{v_1 + v_2 + \dots + v_m : v_i \in V_i\}$$

**Theorem 1.1**

The sum  $V_1 + \dots + V_m$  is the smallest subspace of  $V$  containing all  $V_i$ .

*Proof.* Because  $0$  is in  $V_1, \dots, V_m$ , and subspaces are closed under addition, we get  $0 + 0 + \dots + 0 = 0 \in V_1 + V_2 + \dots + V_m$ . Then if  $x$  is in any  $V_i$  we can always put  $0 + \dots + x + 0 + \dots = x \in V_1 + V_2 + \dots + V_m$ . Therefore,  $V_1, V_2 + \dots + V_m \in V_1 + V_2 + \dots + V_m$ . Now we want to prove that it's the smallest subspace. Suppose that there is a subspace of  $V$ , namely  $u$  which containing  $V_1, \dots, V_m$  and  $U \subset V_1 + V_2 + \dots + V_m$ . Let  $x$  be a vector such that  $x \in V_1 + V_2 + \dots + V_m$  but not in  $u$ . For all  $i$ , there exists an  $v_i \in V_i$  such that

$$u = v_1 + \dots + v_m$$

And since  $U$  contains  $V_1, V_2, \dots, V_m$  then

$$\begin{aligned} v_1 + v_2 &\in U \\ (v_1 + v_2) + v_3 &\in U \\ &\dots \\ (v_1 + v_2 + \dots + v_{n-1}) + v_n &\in U \end{aligned}$$

Hence,  $u \in U$ , which contradicts our assumption that  $x$  is not in  $U$ . Therefore  $V_1 + \dots + V_m$  is the smallest subspace of  $V$  containing  $V_1, \dots, V_m$ .  $\square$

**Definition 1.4: Direct Sum**

The sum  $V_1 + \dots + V_m$  is a *direct sum*, denoted  $V_1 \oplus \dots \oplus V_m$ , if each element of the sum can be written uniquely as  $v_1 + \dots + v_m$  with  $v_i \in V_i$ .

**Theorem 1.2**

The sum  $V_1 + \cdots + V_m$  is a direct sum if and only if the only way to write 0 as  $v_1 + \cdots + v_m$  with  $v_i \in V_i$  is when all  $v_i = 0$ .

*Proof.* It suffices to prove the converse, that is if

$$v_1 + v_2 + \cdots + v_m = 0$$

then  $v_i = 0$ , where each  $v_i \in V_i$ . Now we suppose that there is an element in the sum can be written in two different ways. Suppose that a vector  $v$  can be written as

$$v = x_1 + \cdots + x_m = y_1 + \cdots + y_m$$

where each  $x_i, y_i \in V_i$ . By subtracting the two expressions we get

$$(x_1 - y_1) + \cdots + (y_1 - y_m) = 0$$

Let  $v_i = (x_i - y_i)$  we get  $v_1 + \cdots + v_m = 0$ . By the hypothesis, this implies  $v_i$  must be equal to zero, which means  $x_i = y_i$  for all  $i$ . This contradicts the assumption that the representations are distinct, that is  $(x_1, \dots, x_m)$  is different from  $(y_1, \dots, y_m)$

Therefore, every element in  $V_1 + \cdots + V_m$  has a unique representation. Hence,  $V_1 + \cdots + V_m$  is a direct sum.  $\square$

**Theorem 1.3**

For subspaces  $U, W \subset V$ ,

$$U + W \text{ is a direct sum} \Leftrightarrow U \cap W = \{0\}$$

*Proof.* We begin by proving the forward direction, then proceed to the converse. Suppose first that  $U + W$  is a direct sum and  $x$  is a non-zero element such that  $x \in U \cap W$  and  $x \in U \oplus W$ . According to the properties of direct sum, there exists a unique pair  $(u, w)$  with  $u \in U$  and  $w \in W$  such that

$$x = u + w$$

Using the properties of vector subspaces, we have  $x, u \in U$  implies  $x - u = w \in U$ .  $x, w \in W$  implies  $x - w = u \in W$ .  $x, w \in U$  implies  $x + u = u + w + w \in U$ . However,  $x$  can be rewritten as

$$x = (u + w + w) - u$$

for  $u + w + w \in U$  and  $-u \in W$ . This contradicts the assumption that the representations are distinct. Hence,  $x = 0$  and  $U \cap W = \{0\}$ . Secondly, suppose that  $U \cap W = \{0\}$  and  $x \in U + W$

such that  $x$  can be written as

$$x = x_1 + y_1 = x_2 + y_2$$

which  $(x_1, y_1) \neq (x_2, y_2)$  and  $x_1, x_2 \in U, y_1, y_2 \in W$ . Notice that we have  $x_1, x_2$  implies  $x_1 - x_2 \in U$   $y_1, y_2$  implies  $y_2 - y_1 \in W$  By subtracting the two above expression we get

$$0 = x - x = (x_1 - x_2) + (y_2 - y_1)$$

This implies  $x_1 - x_2 = y_2 - y_1 = t$ , which means  $t \in U$  and  $t \in W$ , or  $t \in U \cap W$ . However, 0 is the only element that belongs to  $U \cap W$ , then  $t = 0$  and in particular  $x_1 = x_2$  and  $y_1 = y_2$  implies  $(x_1, y_1) = (x_2, y_2)$ , this contradicts the hypothesis that  $(x_1, y_1) \neq (x_2, y_2)$ . Hence, every element in  $U + W$  can only be written as a unique representation, which means  $U + W$  is a direct sum.  $\square$   $\square$

## 2. Finite-Dimensional Vector Spaces

- A linear combination of a list  $v_1, \dots, v_m$  of vector in  $V$  is a vector of the form

$$a_1 v_1 + \dots + a_m v_m$$

- Set of all linear combinations of a list of vectors is called

$$\text{span}(v_1 + \dots + v_m) = \{a_1 v_1 + \dots + a_m v_m : a_1, \dots, a_m \in \mathbb{F}\}$$

- The span of a list of vectors in  $V$  is the smallest subspace of  $V$  containing all vectors in the list - If  $\text{span}(v_1, \dots, v_m) = V$  we say that list spans  $V$ . - a list of vectors in  $V$  is called linearly independent if the only choice of  $a_1, \dots, a_m \in \mathbb{F}$  that makes

$$a_1 v_1 + \dots + a_m v_m = 0$$

is  $a_1 = \dots = a_m = 0$ .

### Lemma 2.1: Linear dependence removal

Suppose  $v_1, \dots, v_m$  is a linearly dependent list in  $V$ . Then there exists  $k \in \{1, \dots, m\}$  such that

$$v_k \in \text{span}(v_1, \dots, v_k)$$

Furthermore, if  $k$  satisfies the condition above and the  $k^{\text{th}}$  term is removed from  $v_1, \dots, v_m$  then the span of remaining list equals  $\text{span}(v_1, \dots, v_m)$

**Theorem 2.1**

In a finite-dimensional vector space, every linearly independent list has length  $\leq$  any spanning list.

*Proof.* Let  $A = (v_1, \dots, v_m)$  be a linearly independent list, and let  $B = (w_1, \dots, w_n)$  be a spanning list of  $V$ . We aim to prove that  $m \leq n$ .

Since  $B$  spans  $V$ , each  $v_i$  can be written as a linear combination of vectors from  $B$ . In particular,  $v_1$  is a linear combination of  $w_1, \dots, w_n$ . Therefore, the list  $(v_1, w_1, \dots, w_n)$  is linearly dependent.

Because  $A$  is linearly independent,  $v_1 \neq 0$ . Hence, there must exist some  $k > 0$  such that  $w_k$  is a linear combination of  $v_1, w_1, \dots, w_{k-1}$ . Removing  $w_k$  from  $B$ , we still have a spanning list.

We continue this process: successively add each  $v_i$  into the list and eliminate one redundant vector in  $B$  each time to maintain a spanning list. Eventually, we obtain a list containing all of  $v_1, \dots, v_m$  and still spanning  $V$ . Since we remove one vector from  $B$  for each  $v_i$  added, we cannot perform this more than  $n$  times. Therefore,  $m \leq n$ .  $\square$

**Theorem 2.2**

Every subspace of a finite-dimensional vector space is finite-dimensional.

*Proof.* Let  $V$  be a finite-dimensional vector space, and let  $U \subseteq V$  be a subspace. We aim to show that  $U$  is finite-dimensional.

If  $U = \{0\}$ , then  $U$  is clearly finite-dimensional. Otherwise, there exists a nonzero vector  $u_1 \in U$ . Let  $A = \{u_1\}$ .

Proceed inductively: suppose we have constructed a linearly independent list  $A = \{u_1, u_2, \dots, u_{k-1}\} \subseteq U$ . If  $U \subseteq \text{span}(A)$ , then  $U$  is spanned by finitely many vectors, and we are done.

Otherwise, there exists  $u_k \in U$  such that

$$u_k \notin \text{span}(u_1, u_2, \dots, u_{k-1}).$$

Add  $u_k$  to  $A$  and repeat the process.

Since  $V$  is finite-dimensional, any linearly independent set in  $V$  contains at most  $\dim V$  vectors. Therefore, this process must terminate after finitely many steps. At termination, we have a finite linearly independent set  $A \subseteq U$  such that  $U \subseteq \text{span}(A)$ , i.e.,  $U = \text{span}(A)$ .

Hence,  $U$  is finite-dimensional.  $\square$

**Definition 2.1: Basis**

A basis of  $V$  is a linearly independent list that spans  $V$ .

Every  $v \in V$  has a unique representation in a basis:

$$v = a_1 v_1 + \cdots + a_m v_m$$

Every spanning list can be reduced to a basis.

Every finite-dimensional space has a basis.

Every linearly independent list can be extended to a basis.

### Theorem 2.3

Suppose  $V$  is finite-dimensional and  $U$  is a subspace of  $V$ . Then there is a subspace  $W$  of  $V$  such that  $V = U \oplus W$ .

*Proof.* Let  $u_1, \dots, u_n$  be a basis of  $U$ . This basis can be extended to  $u_1, \dots, u_n, w_1, \dots, w_m$  which is a basis of  $V$ , and of course spans  $V$ . Now let  $W = \text{span}(w_1, \dots, w_m)$ . We are processed to prove  $V = U \oplus W$  and  $U \cap W = \{0\}$ . First of all let  $v \in V$ , by the above hypothesis

$$v = (a_1 w_1 + \cdots + a_n w_n) + (b_1 u_1 + \cdots + b_m u_m) = x + y$$

Indeed,  $x \in U, y \in W$  and  $u_1, \dots, u_n, w_1, \dots, w_m$  is linearly independent. Therefore, there is unique pair  $(x, y) \in U \times W$  such that  $v = x + y$ . This means  $v \in U \oplus W$ , hence  $V = U \oplus W$ .

□

### Definition 2.2: Dimension, $\dim V$

- The dimension of a finite-dimensional vector space is the length of any basis of the vector space.
- The dimension of a finite-dimensional vector space  $V$  is denoted by  $\dim V$ .

### Theorem 2.4: Basis length does not depend on basis

Any two bases of a finite-dimensional vector space have the same length

*Proof.* Let  $X$  and  $Y$  be the basis set of  $V$ . Now we consider that  $X$  is linearly independent and  $Y$  is spanning list of  $V$ . By the lemma 2.1  $|X| \leq |Y|$ . Conversely,  $X$  is spanning list of  $V$  and  $Y$  is linearly independent, then  $|X| \geq |Y|$ . Hence  $|X| = |Y|$ . □

### Definition 2.3

We consider these properties

- If  $V$  is finite-dimensional and  $U$  is a subspace of  $V$ , then  $\dim U \leq \dim V$ .
- Every linearly independent list of vectors in  $V$  of length  $\dim V$  is a basis of  $V$ .

- If  $U$  is subspace of  $V$  and  $\dim U = \dim V$ , then  $U = V$ .
- Every spanning list of vectors in  $V$  of length  $\dim V$  is a basis of  $V$ .

**Theorem 2.5: Dimension of a sum**

If  $V_1$  and  $V_2$  are subspaces of a finite-dimensional vector space, then

$$\dim(V_1 + V_2) = \dim V_1 + \dim V_2 - \dim(V_1 \cap V_2)$$

*Proof.* Let  $v_1, \dots, v_m$  be a basis of  $V_1 \cap V_2$ . This basis can be extended to  $v_1, \dots, v_m, a_1, \dots, a_k$  which is basis of  $V_1$ . We also have  $v_1, \dots, v_m, b_1, \dots, b_l$  is a basis of  $V_2$ .  $\square$