# Chapter 1

# Vector Space

**Definition** (Vector Space). A vector space (or linear space) W over a Field  $\mathbb{F}$  consists of a set on which two operations (called addition and scalar multiplication, respectively) are defined so that for each pair of elements x, y, in W there is a unique element x+y in W, and for each element x in y there is a unique element y in y, such that the following conditions hold.

### § Subspace

<u>Definition</u> (Subspace). A subset W of a vector space W over a field  $\mathbb{F}$  is called a subspace of W if W is a vector space over  $\mathbb{F}$  with the operations of addition and scalar multiplication defined on W.

**Remark.** Trivial subspaces of a vector space V, namely V itself and  $\{0\}$ . Note that empty set  $\phi$  is not a vector space, since it does not contains a zero vector.

**Theorem 1.1.** Let V be a vector space and W a subset of V. Then W is a subspace of V if and only if the following three conditions hold for the operations defined in V.

- (a)  $0 \in W$ .
- (b)  $x + y \in W$  whenever  $x \in W$  and  $y \in W$ .
- (c)  $cx \in W$  whenever  $c \in F$  and  $x \in W$ .

**Corollary 1.1.1.** Let W be a subset of vector space V. W is a subspace of V if and only if  $0 \in W$  and  $ax + y \in W$  whenever  $a \in F$  and,  $x, y \in W$ .

**Theorem 1.2.** Any intersection of subspaces of a vector space W is a subspace of V.

**Theorem 1.3.** Let  $W_1$  and  $W_2$  be subspaces of a vector space V, then  $W_1 \cup W_2$  is a subspace of V if and only if  $W_1 \subseteq W_2$  or  $W_2 \subseteq W_1$ .

**<u>Definition.</u>** If  $S_1$  and  $S_2$  are nonempty subsets of a vector space V, then the sum of  $S_1$  and  $S_2$ , denoted  $S_1 + S_2$ , is the set  $\{x + y : x \in S_1 \text{ and } y \in S_2\}$ .

**Theorem 1.4.** Let  $W_1$  and  $W_2$  be subspaces of a vector space V.

- (a)  $W_1 + W_2$  is a subspace of V that contains both  $W_1$  and  $W_2$ .
- (b) Any subspace of V that contains both  $W_1$  and  $W_2$  must also contain  $W_1 + W_2$ .

**Definition** (Direct Sum). A vector space V is called the direct sum of  $W_1$  and  $W_2$  if  $W_1$  and  $W_2$  are subspaces of V such that  $W_1 \cap W_2 = \{0\}$  and  $W_1 + W_2 = V$ . We denote that V is the direct sum of  $W_1$  and  $W_2$  by writing  $V = W_1 \oplus W_2$ .

**Theorem 1.5.** Let  $W_1$  and  $W_2$  be subspaces of a vector space V. V is the direct sum of  $W_1$  and  $W_2$  if and only if each vector in V can be uniquely written as  $x_1 + x_2$ , where  $x_1 \in W_1$  and  $x_2 \in W_2$ 

#### § Linear Combinations and Bases

<u>Definition</u> (Linearly Dependent). A subset S of a vector space W is called linearly dependent if there exist a finite number of distinct vectors  $u_1, u_2, ..., u_n$  in S and scalars  $a_1, a_2, ..., a_n$ , not all zero, such that

$$a_1u_1 + a_2u_2 + \cdots + a_nu_n = 0.$$

In this case we also say that the vectors of S are linearly dependent.

<u>Definition</u> (Linearly Independent). A subset S of a vector space that is not linearly dependent is called linearly independent. As before, we also say that the vectors of S are linearly independent.

#### Remark.

- 1. The empty set is linearly independent, for linearly dependent sets must be nonempty.
- 2. A set consisting of a single nonzero vector is linearly independent. For if  $\{u\}$  is linearly dependent, then au=0 for some nonzero scalar a. Thus

$$u = a^{-1}(au) = a^{-1}0 = 0$$

3. A set is linearly independent if and only if the only representations of 0 as linear combinations of its vectors are trivial representations.

**Theorem 1.6.** Let V be a vector space, and let  $S_1 \subseteq S_2 \subseteq V$ . If  $S_1$  is linearly dependent, then  $S_2$  is linearly dependent.

**Corollary 1.6.1.** Let S be a linearly independent subset of a vector space V, and let v be a vector in V that is not in S. Then  $S \cup \{v\}$  is linearly dependent if and only if  $v \in span(S)$ .

**<u>Definition</u>** (Basis). A basis  $\beta$  for a vector space V is a linearly independent subset of V that generates V. If  $\beta$  is a basis for V, we also say that the vectors of  $\beta$  form a basis for V.

**Theorem 1.7.** Let V be a vector space and  $\beta = \{u_1, u_2, ..., u_n\}$  be a subset of V. Then  $\beta$  is a basis for V if and only if each  $v \in V$  can be uniquely expressed as a linear combination of vectors of  $\beta$ , that is, can be expressed in the form

$$V = a_1 u_1 + a_2 u_2 + \dots + a_n u_n$$

for unique scalars  $a_1, a_2, \cdots, a_n$ .

**Theorem 1.8.** If a vector space V is generated by a finite set S, then some subset of S is a basis for V. Hence V has a finite basis.

**Theorem 1.9** (Replacement Theorem). Let V be a vector space that is generated by a set G containing exactly n vectors, and let L be a linearly independent subset of V containing exactly m vectors. Then  $m \leq n$  and there exists a subset H of G containing exactly n-m vectors such that  $L \cup H$  generates V.

**Definition** (Finite-Dimensional). A vector space is called finite-dimensional if it has a basis consisting of a finite number of vectors. The unique number of vectors in each basis for V is called the dimension of V and is denoted by dim(V). A vector space that is not finite-dimensional is called infinite-dimensional.

Corollary 1.9.1. Let V be a vector space with dimension n.

- 1. Any finite generating set for V contains at least n vectors, and a generating set for V that contains exactly n vectors is a basis for V.
- 2. Any linearly independent subset of V that contains exactly n vectors is a basis for V.
- 3. Every linearly independent subset of V can be extended to a basis for V.

**Theorem 1.10.** Let W be a subspace of a finite-dimensional vector space V. Then W is finite-dimensional and  $\dim(W) \leq \dim(V)$ . Moreover, if  $\dim(W) = \dim(V)$ , then V = W.

**Propersition 1.11.** Let  $W_1$  and  $W_2$  be subspaces of a finite-dimensional vector space V.  $W_1 \subseteq W_2$  if and only if  $\dim(W_1 \cap W_2) = \dim(W_1)$ 

**Theorem 1.12.** Let  $v_1, v_2, \dots, v_k, v$  be vectors in a vector space V, and define  $W_1 = span(\{v_1, v_2, \dots, v_k\})$ , and  $W_2 = span(\{v_1, v_2, \dots, v_k, v\})$ . Then  $v \in span(W_1)$  if and only if  $dim(W_1) = dim(W_2)$ .

**Remark.** We may give an example for satisfying the conditions on above but  $\dim(W_1) \neq \dim(W_2)$ .

**Theorem 1.13.** Let  $W_1$  and  $W_2$  be finite-dimensional subspaces of a vector space V.

(a) Then the subspace  $W_1 + W_2$  is finite-dimensional, and

$$\dim(W_1 + W_2) = \dim(W_1) + \dim(W_2) - \dim(W_1 \cap W_2)$$

(b) Let  $V = W_1 + W_2$ . Deduce that V is the direct sum of  $W_1$  and  $W_2$  if and only if

$$\dim(V) = \dim(W_1) + \dim(W_2)$$

**Theorem 1.14.** Let  $W_1$  and  $W_2$  be subspaces of a vector space V such that  $V = W_1 \oplus W_2$  if and only if there exist base  $\beta_1$ ,  $\beta_2$  of  $W_1$ ,  $W_2$ , respectively such that  $\beta_1 \cup \beta_2$  is a basis for V.

### Theorem 1.15.

If  $W_1$  is any subspace of vector space of V, then there exists a subspace  $W_2$  of V such that

$$V = W_1 \oplus W_2$$