CH1 Vector Space Textbook

May 2, 2022

Vector Space

Definition (Vector Space).

A vector space (or linear space) W over a $field^2$ 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 a in F and each element x in W there is a unique element x in W, such that the following conditions hold.

There are eight rules for calculating the vector space! Please do not forget!

<u>Definition</u> (Definition of Subspace).

A subset W of a vector space W over a field F is called a subspace of W if W is a vector space over F with the operations of addition and scalar multiplication defined on W.

Theorem (1.3).

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 $x \in W$.

Theorem (1.4).

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

Example

Question (Exercise 19):

Let W_1 and W_2 be subspaces of a vector space V. Prove that $W_1 \cup W_2$ is a subspace of W if and only if $W_1 \subseteq W_2$ or $W_2 \subseteq W_1$.

Solution.

- 1. Let W_1 and W_2 be subspace of W, $W_1 \cup W_2$ is a subspace
- 2. Suppose $W_1 \cup W_2$ is a subspace of W while $W_1 \not\subseteq W_2$ and $W_2 \not\subseteq W_1 \to \exists \ x \in W_1, x \notin W_2$, $\exists \ y \in W_2, y \notin W_1$ If $W_1 \cup W_2$ is a subspace, $x, y \in W_1 \cup W_2 \to x + y \in W_1 \cup W_2$
- 3. $x + y \in W_1 \cup W_2$ $\therefore x + y \in W_1$ or $x + y \in W_2$ Since W_1 and W_2 are subspaces of V $\therefore x \in W_1, y \in W_2$ or $x \in W_2, y \in W_1$ $\rightarrow \leftarrow$ wtih (1)

<u>Definition</u> (Definition of Linear Combination).

Let W be a vector space and S a nonempty subset of V. A vector $v \in V$ is called a linear combination of vectors of S if there exist a finite number of vectors $u_1, u_2, ..., u_n$ in S and scalars $a_1, a_2, ..., a_n$ in F such that $v = a_1u_1 + a_2u_2 + ... + a_nu_n$. In this case we also say that W is a linear combination of $u_1, u_2, ..., u_n$ and call $a_1, a_2, ..., a_n$ the coefficients of the linear combination.

Definition (Definition of Span).

Let S be a nonempty subset of a vector space V. The span of S, denoted span(S), is the set consisting of all linear combinations of the vectors in S. For convenience, we define $\text{span}(\phi) = \{0\}.$

Example
$$S = \{(0,0,1), (1,0,0)\}$$

$$Span(S) = \{a_1(0,0,1) + a_2(1,0,0) \mid a_1, a_2 \in F\}$$

Theorem (1.5).

The span of any subset S of a vector space W is a subspace of V. Moreover, any subspace of W that contains S must also contain the span of S.

<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

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

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

<u>Definition</u> (Definition of 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.

Theorem (1.6).

Let W 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. Let W be a vector space, and let $S_1 \subseteq S_2 \subseteq V$. If S_2 is linearly independent, then S_1 is linearly independent.

Theorem (1.7).

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 $v \in \text{span}(S)$.

<u>Definition</u> (Definition of Bases).

A basis β for a vector space V is a linearly independent subset of V that generates V. If β is a basis for V, we also say that the vectors of β form a basis for V.

Theorem (1.8).

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

$$\mathbf{v} = a_1 u_1 + a_2 u_2 + \dots + a_n u_n$$

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

Theorem (1.9).

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 (1.10 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 $m \le n$ and there exists a subset H of G containing exactly $n \mp m$ vectors such that $L \cup H$ generates V.

Theorem (Corollary 1.).

Let V be a vector space having a finite basis. Then every basis for V contains the same number of vectors.

<u>Definition</u> (Definition of Dimension).

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 .

Chapter 2

§ 2-1 Linear transformation

Definition (Definition of Linear Transformation).

Let V and W be vector spaces (over F). We call a function $T:V\to W$ a linear transformation from V to W if, for all $x,y\in V$ and $c\in F$, we have

- (a) T(x + y) = T(x) + T(y)
- (b) T(cx) = cT(x)

<u>Definition</u> (Definition of nullity and rank).

Let V and W be vector spaces and let $T: V \to W$ be linear. If N(T) and R(T) are finite-dimensional, then we define the nullity of T, denoted nullity(T), and the rank of T, denoted rank(T), to be the dimensions of N(T) and R(T), respectively.

Theorem.

Let V and W be vector spaces and T : V \rightarrow W be linear. Then N(T) and R(T) are subspaces of V and W, respectively.

Theorem.

Let V and W be vector spaces, and let $T: V \to W$ be linear. If $\beta = \{v_1, v_2, ..., v_n\}$ is a basis for V, then $R(T) = \text{span}(T(\beta)) = \text{span}(\{T(v_1), T(v_2), ..., T(v_n)\})$.

Example 11

Question:

Let $T: P_2(R) \to P_3(R)$ be the linear trans formation defined by

$$T(f(x)) = 2f'(x) + \int_0^x 3f(t)dt$$

Solution.

Now R(T) =
$$span(\{T(1), T(x), T(x^2)) = span(\{3x, 2 + \frac{3}{2}x^2, 4x + x^3\}).$$

Since $\{3x, 2+\frac{3}{2}x^2, 4x+x^3\}$ is linearly independent, $\operatorname{rank}(T)=3$. Since $2\dim(P3(R))=4$, T is not onto. From the dimension theorem (Thm 2.3), nullity(T) + 3 = 3. So nullity(T) = 0, and therefore, N(T) = 0. We conclude from Theorem 2.4 that T is one-to-one.

Example 12

Question:

Let $T: F^2 \to F^2$ be the linear transformation defined by

$$T(a_1, a_2) = (a_1 + a_2, a_1)$$

Solution.

It is easy to see that $N(T) = \{0\}$; so T is one-to-one. Hence Theorem 2.5 tells us that T must be onto.

Theorem (2.3 Dimension Theorem).

Let V and W be vector spaces, and let $T: V \to W$ be linear.

If V is finite-dimensional, then

$$\operatorname{nullity}(T) + \operatorname{rank}(T) = \dim(V).$$

Theorem (2.4).

Let V and W be vector spaces, and let $T:V\to W$ be linear. Then T is one-to-one if and only if N(T) = 0 .

Theorem (2.5).

Let V and W be vector spaces of equal (finite) dimension, and let $T: V \to W$ be linear. Then the following are equivalent.

- (a) T is one-to-one.
- (b) T is onto.
- (c) rank(T) = dim(V).

Example

Let $T:\mathbb{R}^2\longrightarrow\mathbb{R}^3$, $T(a_1,a_2)=(a_1+a_2,0,2a_1-a_2)$, determine that T is linear , 1-1 , onto or not.

Solution.

• Calim: T is linear.

$$\begin{aligned} \text{Let } x &= (a_1, a_2) \,, \, y = (b_1, b_2) \\ \text{T}(\text{cx} + \text{y}) &= \text{T} \left(\text{c} \left(\mathbf{a}_1, \mathbf{a}_2 \right) + (\mathbf{b}_1, \mathbf{b}_2) \right) = \text{T} \left(\mathbf{c} \mathbf{a}_1 + \mathbf{b}_1, \mathbf{c} \mathbf{a}_2 + \mathbf{b}_2 \right) \\ &= \left(\mathbf{c} \mathbf{a}_1 + \mathbf{b}_1 + \mathbf{c} \mathbf{a}_2 + \mathbf{b}_2, \mathbf{0}, 2\mathbf{c} \mathbf{a}_1 + 2\mathbf{b}_1 - \mathbf{c} \mathbf{a}_2 - \mathbf{b}_2 \right) \\ &= \left(\text{c} \left(\mathbf{a}_1 + \mathbf{a}_2 \right) + \left(\mathbf{b}_1 + \mathbf{b}_2 \right), \mathbf{0}, \text{c} \left(2\mathbf{a}_1 - \mathbf{a}_2 \right) + \left(2\mathbf{b}_1 - \mathbf{b}_2 \right) \right) \\ &= \text{c} \left(\mathbf{a}_1 + \mathbf{a}_2, \mathbf{0}, 2\mathbf{a}_1 - \mathbf{a}_2 \right) + \left(\mathbf{b}_1 + \mathbf{b}_2, \mathbf{0}, 2\mathbf{b}_1 - \mathbf{b}_2 \right) \\ &= \text{cT}(\mathbf{x}) + \text{T}(\mathbf{y}) \end{aligned}$$

∴ T is linear

• 1-1 and onto

By Thm 2.2 in text book choose a basis for \mathbb{R}^2 , $\beta = \{(1,0), (0,1)\}$

$$R(T) = \mathrm{span}\,(T(\beta)) = \mathrm{span}(\{T(1,0),T(0,1)\}) = \mathrm{span}\,((1,0,-1),(1,0,-2))$$

Clearly it is L.I. \implies rank(T) = 2, and apply the Dimension Theorem rank $(T) = 2 \neq 3 = \dim(\mathbb{R}^3)$, and nullitily(T) = 1, then it is not onto.

By Thm 2.4 it is not one to one.

Example

Let V and W be vector spaces, let T: V \rightarrow W be linear, and let $\{w_1, w_2, \dots, w_k\}$ be a linearly independent subset of R(T). Prove that if $S = \{v_1, v_2, \dots, v_k\}$ is chosen so that T $(v_i) = w_i$ for $i = 1, 2, \dots, k$, then S is linearly independent.

Solution. Calim:
$$\sum_{i=1}^n a_i v_i = 0 \implies a_1 = a_2 = a_3 = \dots = a_n = 0$$
 Let
$$\sum_{i=1}^n a_i v_i = 0 \text{ then } \mathbf{T} \left(\sum_{i=1}^n a_i v_i \right) = 0 \text{ Since T is linear , } \mathbf{T} \left(\sum_{i=1}^n a_i v_i \right) = \sum_{i=1}^n a_i \mathbf{T} \left(v_i \right) = \sum_{i=1}^n a_i w_i = 0 \text{ Since S is L.I.} \implies a_1 = a_2 = \dots = a_n = 0$$

Theorem (2.6).

Let V and W be vector spaces over F, and suppose that $\{v_1, v_2, ..., v_n\}$ is a basis for V.

For $w_1, w_2, ..., w_n$ in W, there exists exactly one linear transformation $T: V \to W$ such that $T(v_i) = w_i$ for i = 1, 2, ..., n.