

Chapter 2 Linear Transformations, Null Spaces, and Ranges

§ 2-1 Linear transformation

Definition (Definition of Linear Transformation).

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

$$(a) \quad T(x + y) = T(x) + T(y)$$

$$(b) \quad T(cx) = cT(x)$$

Definition (Definition of nullity and rank).

Let V and W be vector spaces and let $T : V \rightarrow W$ be linear. If $N(T)$ and $R(T)$ are finite-dimensional, then we define the nullity of T , denoted $\text{nullity}(T)$, and the rank of T , denoted $\text{rank}(T)$, 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 $T : V \rightarrow W$ be linear. If $\beta = \{v_1, v_2, \dots, v_n\}$ is a basis for V , then $R(T) = \text{span}(T(\beta)) = \text{span}(\{T(v_1), T(v_2), \dots, T(v_n)\})$.

Example 11

Question :

Let $T : P_2(R) \rightarrow P_3(R)$ be the linear transformation defined by

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

solution:.

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

Since $\{3x, 2 + \frac{3}{2}x^2, 4x + x^3\}$ is linearly independent, $\text{rank}(T) = 3$. Since $\dim(P_3(\mathbb{R})) = 4$, T is not onto. From the dimension theorem (Thm 2.3), $\text{nullity}(T) + 3 = 4$. So $\text{nullity}(T) = 1$, and therefore, $N(T) \neq \{0\}$. We conclude from Theorem 2.4 that T is not one-to-one.

Example 12

Question :

Let $T : F^2 \rightarrow F^2$ 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 $T : V \rightarrow W$ be linear.

If V is finite-dimensional, then

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

Theorem (2.4).

Let V and W be vector spaces, and let $T : V \rightarrow W$ 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 $T : V \rightarrow W$ be linear. Then the following are equivalent.

- (a) T is one-to-one.
- (b) T is onto.
- (c) $\text{rank}(T) = \dim(V)$.

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:.

- Claim : T is linear.

Let $x = (a_1, a_2)$, $y = (b_1, b_2)$

$$\begin{aligned} T(cx + y) &= T(c(a_1, a_2) + (b_1, b_2)) = T(ca_1 + b_1, ca_2 + b_2) \\ &= (ca_1 + b_1 + ca_2 + b_2, 0, 2ca_1 + 2b_1 - ca_2 - b_2) \\ &= (c(a_1 + a_2) + (b_1 + b_2), 0, c(2a_1 - a_2) + (2b_1 - b_2)) \\ &= c(a_1 + a_2, 0, 2a_1 - a_2) + (b_1 + b_2, 0, 2b_1 - b_2) \\ &= cT(x) + T(y) \end{aligned}$$

$\therefore T$ is linear

- 1-1 and onto

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

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

Clearly it is L.I. $\implies \text{rank}(T) = 2$, and apply the Dimension Theorem $\text{rank}(T) = 2 \neq 3 = \dim(\mathbb{R}^3)$, and nullity(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:.

$$\text{Calim : } \sum_{i=1}^n a_i v_i = 0 \implies a_1 = a_2 = a_3 = \dots = a_n = 0$$

$$\text{Let } \sum_{i=1}^n a_i v_i = 0 \text{ then } T\left(\sum_{i=1}^n a_i v_i\right) = 0$$

$$\text{Since } T \text{ is linear, } T\left(\sum_{i=1}^n a_i v_i\right) = \sum_{i=1}^n a_i T(v_i) = \sum_{i=1}^n a_i w_i = 0$$

$$\text{Since } S \text{ is L.I. } \implies a_1 = a_2 = \dots = a_n = 0$$

Theorem (2.6).

Let V and W be vector spaces over F , and suppose that $\{v_1, v_2, \dots, v_n\}$ is a basis for V . For w_1, w_2, \dots, w_n in W , there exists exactly one linear transformation $T : V \rightarrow W$ such that $T(v_i) = w_i$ for $i = 1, 2, \dots, n$.

§ 2-2 The Matrix Representation of a Linear transformation

Definition (Definition of an Ordered Basis).

Let V be a finite-dimensional vector space. An ordered basis for V is a basis for V endowed with a specific order; that is, an ordered basis for V is a finite sequence of linearly independent vectors in V that generates V .

Definition (Definition for the symbol $[x]_\beta$).

Let $\beta = \{u_1, u_2, \dots, u_n\}$ be an ordered basis for a finite-dimensional vector space V . For $x \in V$, let a_1, a_2, \dots, a_n be the unique scalars such that $x = \sum_{i=1}^n a_i u_i$. We define the coordinate vector of x relative to β , denoted $[x]_\beta$, by

$$[x]_\beta = \begin{bmatrix} a_1 \\ a_2 \\ \vdots \\ a_n \end{bmatrix}$$

Notice that $[u_i]_\beta = e_i$ in the preceding definition. It is left as an exercise to show that the correspondence $x \rightarrow [x]_\beta$ provides us with a linear transformation from V to F^n . We study this transformation in Section 2.4 in more detail.

Definition (Definition for the symbol $[T]_\beta^\gamma$).

Using the notation above, we call the $m \times n$ matrix A defined by $A_{ij} = a_{ij}$ the matrix representation of T in the ordered bases β and γ and write $A = [T]_\beta^\gamma$. If $V = W$ and $\beta = \gamma$, then we write $A = [T]_\beta$. Notice that the j th column of A is simply $[T(v_j)]_\gamma$. Also observe that if $U : V \rightarrow W$ is a linear transformation such that $[U]_\beta^\gamma = [T]_\beta^\gamma$, then $U = T$ by the corollary to Theorem 2.6 (p. 73).

Definition (The addition and multiplication for Linear Transformation).

Let $T, U : V \rightarrow W$ be arbitrary functions, where V and W are vector spaces over F , and let $a \in F$. We define $T + U : V \rightarrow W$ by $(T + U)(x) = T(x) + U(x)$ for all $x \in V$, and $aT : V \rightarrow W$ by $(aT)(x) = aT(x)$ for all $x \in V$.

Of course, these are just the usual definitions of addition and scalar multiplication of functions. We are fortunate, however, to have the result that both sums and scalar multiples of linear transformations are also linear.

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