

Lecture 10

L10①

As we mentioned earlier we will elaborate the study of matrices in the framework of linear spaces. We will basically view a matrix as a map from a linear space U to a linear space V . We will see how these additional structures on U & V play a role.

Linear transformation (or linear map): Let U and V be two linear spaces. A map $T: U \rightarrow V$ is called a linear map if

$$(i) T(u_1 + u_2) = T(u_1) + T(u_2) \text{ for } u_1, u_2 \in U$$

$$(ii) T(\alpha u) = \alpha T(u), \alpha \in \mathbb{R} \text{ and } u \in U.$$

From conditions (i) & (ii) we see that linear maps preserve algebraic operations. Note that T satisfies (i) & (ii) if

$$T(\alpha u_1 + \beta u_2) = \alpha T(u_1) + \beta T(u_2) \text{ for } \alpha, \beta \in \mathbb{R} \text{ and } u_1, u_2 \in U.$$

We will see that a matrix is a linear map.

Examples:

1. $T: \mathbb{R} \rightarrow \mathbb{R}^2$ defined by $T(x) = (x, 3x)$, $x \in \mathbb{R}$ is a linear map.

2. $T_i: \mathbb{R}^n \rightarrow \mathbb{R}$ defined by $T_i(x_1, x_2, \dots, x_n) = x_i$ is a linear map

3. For an $m \times n$ matrix, define $T_A: \mathbb{R}^n \rightarrow \mathbb{R}^m$ by $T_A(x) = Ax$. Then T_A is linear. (Here Ax means Ax^t).

4. $T: \mathbb{R}^2 \rightarrow \mathbb{R}^2$ defined by $T(x, y) = (xy, xy)$ is not linear

5. $T: \mathbb{R}^{n+1} \rightarrow \mathbb{P}_n$ defined by $T(a_0, a_1, \dots, a_n) = a_0 + a_1x + \dots + a_nx^n$ is linear

6. $\frac{d}{dx}: \mathbb{P}_n \rightarrow \mathbb{P}_n$, $\frac{d}{dx}(P_n(x)) = P_n'(x)$ is linear

7. $T: \mathbb{P}_n \rightarrow \mathbb{R}$, $T(P_n(x)) = \int_0^1 P_n(x)$ is linear.

Proposition: If T is a linear map then $T(0) = 0$, $T(-u) = -T(u)$ &
 $T(u_1 - u_2) = T(u_1) - T(u_2)$.

Proof: Exercise.

Remark: $T \equiv 0$ is a linear map.

Kernel and Range of a linear map: Let $T: V \rightarrow W$ be a linear

map. Then the kernel of T , denoted by $N(T)$, is defined by

$N(T) = \{v \in V : Tv = 0\}$ and the range of T , denoted $R(T)$ by $R(T)$ is defined by $R(T) = T(V) = \{w \in W : \exists v \in V \text{ s.t. } Tv = w\}$.

Exercise: Verify that $N(T)$ & $R(T)$ are subspaces of V & W resp.

Example: Find the range & kernel of $T: \mathbb{R}^3 \rightarrow \mathbb{R}^3$ defined by

$$T(x, y, z) = (x+z, x+y+2z, 2x+y+3z).$$

$$\begin{aligned} \text{Note that } N(T) &= \{(x, y, z) \in \mathbb{R}^3 : T(x, y, z) = (0, 0, 0)\} \\ &= \{(x, y, z) \in \mathbb{R}^3 : x+z=0, x+y+2z=0 \text{ \& } 2x+y+3z=0\} \\ &= \{(x, x, -x) : x \in \mathbb{R}\} = \text{span}\{(1, 1, -1)\}. \end{aligned}$$

Let $e_1 = (1, 0, 0)$, $e_2 = (0, 1, 0)$ & $e_3 = (0, 0, 1)$. Then

$$\begin{aligned} R(T) &= \{Tu : u \in \mathbb{R}^3\} = \{T(\alpha e_1 + \beta e_2 + \gamma e_3) : \alpha, \beta, \gamma \in \mathbb{R}\} \\ &= \{\alpha T e_1 + \beta T e_2 + \gamma T e_3 : \alpha, \beta, \gamma \in \mathbb{R}\} \\ &= \text{span}\{T e_1, T e_2, T e_3\} \\ &= \text{span}\{(1, 1, 2), (0, 1, 1), (1, 2, 3)\} \\ &= \text{span}\{(1, 1, 2), (0, 1, 1)\} \end{aligned}$$

Rank & Nullity: Rank of $T = \dim(R(T))$ & Nullity of $T = \dim N(T)$.

Remark: In the previous example, we have seen that

$$\text{Rank}(T) + \text{Nullity}(T) = 3 = \dim(\mathbb{R}^3).$$

The relation between Rank & Nullity is given in the following theorem.

Rank-Nullity Theorem: Let V & W be f.d.v.s. & $T: V \rightarrow W$ be a linear map. Then $\text{Rank}(T) + \text{Nullity}(T) = \dim(V)$.

Proof: Let $\{u_1, u_2, \dots, u_k\}$ be a basis of $N(T)$. Extend this set to a basis of V , say $\{u_1, u_2, \dots, u_k, v_1, v_2, \dots, v_n\}$.

claim: $\{TV_1, \dots, TV_n\}$ is a basis of $R(T)$.

It is easy to verify that $\text{span}\{TV_1, \dots, TV_n\} = R(T)$.

To show the set is L.I., let $\sum_1^n \alpha_i TV_i = 0$. Then

$$T\left(\sum_1^n \alpha_i V_i\right) = 0 \Rightarrow \sum_1^n \alpha_i V_i \in N(T)$$

$$\Rightarrow \sum_1^n \alpha_i V_i = \sum_1^k \beta_j U_j \text{ for some } \beta_j$$

$$\Rightarrow \sum_1^n \alpha_i V_i - \sum_1^k \beta_j U_j = 0$$

$$\Rightarrow \alpha_i = 0 \text{ \& } \beta_j = 0, \quad i=1, \dots, n \text{ \& } j=1, 2, \dots, k$$

(because $\{U_1, \dots, U_k, V_1, \dots, V_n\}$ is L.I.) \square

Applications: 1. Let V be a f.d.v.s of dimension n & $T: V \rightarrow W$ be a linear map. Then T is one-one $\iff T$ is onto.

Proof: T is one-one $\iff N(T) = \{0\} \iff \text{Nullity}(T) = 0 \iff$
 $\text{Rank}(T) = n \iff R(T) = W \iff T$ is onto

2. Let V be a f.d.v.s & $T: V \rightarrow V$ be linear. $\iff T$ is one-one or onto then T is invertible.

* Rank of a matrix: Let $A = (a_{ij})$ be an $m \times n$ matrix. Suppose

$A = (C_1 \ C_2 \ \dots \ C_n)$ where C_i 's are columns of A . Then

$A: \mathbb{R}^n \rightarrow \mathbb{R}^m$ and $\text{Rank}(A) = \dim(R(A)) \leq m$.

Note that $R(A) = \text{span}\{Ae_1, Ae_2, \dots, Ae_n\} = \text{span}\{C_1, \dots, C_n\}$.
 This space is ^{sometimes} called column space of A because it is spanned by column vectors of A .

Similarly a row space of A is defined as follows:

Row space of $A = \text{span}\{R_1, \dots, R_m\}$, span of row

vectors of A .

* 3. Let A be an $m \times n$ matrix. Then $Ax = 0$ has a nontrivial solution \iff the number of unknowns are more than the number of equations, because $A: \mathbb{R}^n \rightarrow \mathbb{R}^m$, $m < n$ & $\text{nullity}(A) \neq \{0\}$