

Lecture 10: Linear transformation & Rank - Nullity theorem

Consider a matrix A of order $m \times n$.
The matrix A induces a map $T_A: \mathbb{R}^n \rightarrow \mathbb{R}^m$
as follows: $T_A(x) = Ax \quad \forall x \in \mathbb{R}^n$, here
 x is treated as $n \times 1$ matrix (or
column vector)

T_A satisfies the following properties:

$$\star \quad T_A(x + y) = T_A(x) + T_A(y)$$

$$\star \quad T_A(\lambda x) = \lambda T_A(x)$$

$$\forall x, y \in \mathbb{R}^n, \lambda \in \mathbb{R}.$$

These properties leads to notion of
linear transformation.

Linear transformation: Let V and W be
two vector spaces over K ($K = \mathbb{R}$ or \mathbb{C})

A map $T: V \rightarrow W$ is said to be linear
transformation if

$$(i) \quad T(x + y) = T(x) + T(y), \quad \forall x, y \in V$$

$$(ii) \quad T(\lambda x) = \lambda T(x) \quad \forall \lambda \in K, x \in V.$$

Examples :- (1) As discussed earlier, for a

matrix A of order $m \times n$, the map

$T_A : K^n \rightarrow K^m$ given by

$$T_A(x) = Ax, \quad x \in K^n \text{ is a}$$

linear transformation, where $K = \mathbb{R} \text{ or } \mathbb{C}$.

For instance, let us take $A = \begin{pmatrix} \cos \theta & \sin \theta \\ -\sin \theta & \cos \theta \end{pmatrix}$

$$\text{f } K = \mathbb{R}, \quad m, n = 2, \quad T_A \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} x \cos \theta + y \sin \theta \\ -x \sin \theta + y \cos \theta \end{pmatrix}$$

or written as $T_A(x, y) = (x \cos \theta + y \sin \theta, -x \sin \theta + y \cos \theta)$

(2) Let $P(x)$ be the vector space of all polynomial of single variable with coefficients from \mathbb{R} .

Consider the derivative map

$$\frac{d}{dx} : P(x) \rightarrow P(x)$$

$$\frac{d}{dx} (f(x)) = f'(x)$$

Check that $\frac{d}{dx}$ is linear.

(iii) The map $T: \mathbb{R}^{n+1} \rightarrow \mathcal{P}(x)$
 $T(a_0, a_1, \dots, a_n) = a_0 + a_1x + \dots + a_nx^n$
is linear.

(iv) The map $T: \mathcal{P}(x) \rightarrow \mathcal{P}(x)$
 $T(f(x)) = \int_0^x f(x) dx$
is linear.

Non-Example: (i) The map $T: \mathbb{R} \rightarrow \mathbb{R}$
 $T(x) = x^2$
is not linear.

(ii) $T: \mathbb{R}^2 \rightarrow \mathbb{R}^2$ defined by
 $T(x, y) = (x, y^3)$
is not linear.

Proposition: Let $T: V \rightarrow W$ be a linear map then $T(0) = 0$

Proof: $T(0) = T(0+0) = T(0) + T(0)$
 $\Rightarrow T(0) = 0 \quad \square$

Problem: Every linear transformation
 $T: \mathbb{R} \rightarrow \mathbb{R}$ is of the form
 $T(x) = \lambda x$ for some $\lambda \in \mathbb{R}$

Solution: Here \mathbb{R} is vector space over itself.

$\{1\}$ is a basis of \mathbb{R} , any element $x \in \mathbb{R}$ can be written as $x = x \cdot 1$

$$T(x) = T(x \cdot 1) = x \cdot T(1)$$

$$\text{Let } \lambda = T(1), \quad T(x) = x \cdot \lambda = \lambda x$$

Kernel of a linear transformation

Let $T: V \rightarrow W$ be a linear transformation.

Kernel of T is defined as $\{u \in V : T(u) = 0\}$.
It is denoted as $\ker(T)$.

Proposition: $\ker(T)$ is a subspace of V .

Proof: Left as an exercise

Example: (1) Consider the linear map

$T: \mathbb{R}^2 \rightarrow \mathbb{R}$ defined by

$$T(x, y) = x - y.$$

$$\ker(T) = \{ (x, y) \in \mathbb{R}^2 : T(x, y) = 0 \}$$

$$= \{ (x, y) \in \mathbb{R}^2 : y = x \}$$

= (line passing through origin with slope 1)

(2) $T: \mathbb{R}^2 \rightarrow \mathbb{R}^2$ defined by $T(x, y) = (x+y, y)$ is linear & $\ker(T) = \{(x, y) \in \mathbb{R}^2: T(x, y) = (0, 0)\}$
 $= \{(x, y) \in \mathbb{R}^2: (x+y, y) = (0, 0)\}$
 $= \{(0, 0)\}$

(3) Consider the linear map $T_A: \mathbb{R}^n \rightarrow \mathbb{R}^m$
 $T_A(x) = Ax$ where A is a matrix of order $m \times n$.

$$\ker(T_A) = \{x \in \mathbb{R}^n: Ax = 0\}$$

Finding a point in $\ker(T_A)$ means finding a solution of the homogenous system of linear equations $Ax = 0$.

Proposition: A linear map $T: V \rightarrow W$ is injective if and only if $\ker(T) = \{0\}$

Proof: Let T be injective. Let $x \in \ker(T)$
 $\Rightarrow T(x) = 0 = T(0) \Rightarrow x = 0$ (as T is injective)

(Conversely, let $\ker(T) = \{0\}$

$$\text{Let } T(x) = T(y) \Rightarrow T(x-y) = 0$$

$$\Rightarrow x-y \in \ker T = \{0\}$$

$$\Rightarrow x = y.$$

□

Range of a linear map :- Let $T: V \rightarrow W$ be a linear transformation. Range of T , denoted by $R(T)$, is the set $\{T(u) : u \in V\}$.

Proposition: $R(T)$ is a subspace of W .

Proof: Left as an exercise.

Example: Consider the linear map $T: \mathbb{R}^3 \rightarrow \mathbb{R}^3$ defined by

$$T(x, y, z) = (x - y, x - 2y + z, 2x - y - z)$$

$$T(x, y, z) = (0, 0, 0) \Leftrightarrow x = y = z$$

$$\therefore \text{Ker}(T) = \{(x, x, x) : x \in \mathbb{R}\}$$

Its basis is $\{(1, 1, 1)\}$ & $\dim(\text{Ker}(T)) = 1$.

$$R(T) = \{(x - y, x - 2y + z, 2x - y - z) : x, y, z \in \mathbb{R}\}$$

$$(x - y, x - 2y + z, 2x - y - z)$$

$$= x(1, 1, 2) - y(1, 2, 1) + z(0, 1, -1)$$

$$= x(1, 1, 2) - y\{(1, 1, 2) + (0, 1, -1)\} + z(0, 1, -1)$$

$$= (x - y)(1, 1, 2) + (z - y)(0, 1, -1)$$

$$R(T) = L(\{(1, 1, 2), (0, 1, -1)\})$$

& $\{(1, 1, 2), (0, 1, -1)\}$ is L.I.

$$\dim(R(T)) = 2. \quad \text{So,}$$

$$\dim(\ker(T)) + \dim(R(T)) = 3 = \dim(\mathbb{R}^3).$$

This formula is true in general & it is called
Rank-Nullity theorem

Theorem: Let $T: V \rightarrow W$ be a linear transformation between finite dimensional vector spaces V & W . Then

$$\dim(\ker(T)) + \dim(R(T)) = \dim(V).$$

Proof: Let $n = \dim(V)$ & $\{u_1, \dots, u_m\}$

be a basis of $\ker(T)$. Extend this basis to a basis $\{u_1, \dots, u_m, u_{m+1}, \dots, u_n\}$ of V . We will prove that $\{T(u_{m+1}), \dots, T(u_n)\}$ is a basis of $R(T)$.

Let $y \in R(T) \Rightarrow \exists x \in V$ such that
 $y = T(x), \quad x = \sum_{i=1}^n \alpha_i u_i$

$$y = T(x) = \sum_{i=1}^n \alpha_i T(u_i) = \sum_{i=m+1}^n \alpha_i T(u_i)$$

as $T(u_i) = 0 \quad \forall 1 \leq i \leq m.$

$$\Rightarrow R(T) = L(\{T(u_{m+1}), \dots, T(u_n)\})$$

$$\sum_{i=1}^n \lambda_i T(u_{m+i}) = 0 \Rightarrow T\left(\sum_{i=1}^n \lambda_i u_{m+i}\right) = 0$$

$$\Rightarrow \sum_{i=1}^n \lambda_i u_{m+i} \in \ker(T) = L(\{u_1, \dots, u_m\})$$

$$\Rightarrow \sum_{i=1}^n \lambda_i u_{m+i} = \sum_{j=1}^m \beta_j u_j$$

$$\Rightarrow \beta_1 u_1 + \dots + \beta_m u_m - \lambda_1 u_{m+1} - \dots - \lambda_n u_{m+n} = 0$$

$$\Rightarrow \beta_1 = \dots = \beta_m = \lambda_1 = \dots = \lambda_n = 0$$

as $\{u_1, \dots, u_n\}$ is L.I.

$$\text{So, } \lambda_1 = \dots = \lambda_n = 0$$

So $\{T(u_{m+1}), \dots, T(u_n)\}$ is a basis of $R(T)$

$$\Rightarrow \dim(R(T)) = n - m.$$

$$\therefore \dim(\ker(T)) + \dim(R(T)) = m + (n - m)$$

$$= n$$

$$= \dim(V) \quad \square$$

Rank of a linear transformation Let $T: V \rightarrow W$ be

a linear transformation between finite dimensional vector spaces V & W . By rank of T we mean $\dim(R(T))$. Thus, the terminology 'Rank-Nullity' theorem makes sense.