

Lecture 2

Theorem:- (1) Every vector space has a basis.

(2) Any two bases of a vector space have the same cardinality. This number is called the dimension of the vector space over F .

Def:- Let V be a vector space, let $B \subset V$ be a basis of V . Then $\dim(V) := |B|$.

Example:- (1) $\{ (1,0,0), (0,1,0), (0,0,1) \}$ is a basis of \mathbb{R}^3 .

(2) $\{ 1, x, x^2, \dots \}$ is a basis of $\mathbb{R}[x]$.

Characterization of basis

Thm:- Let V be a vector sp. over a field F .

Let $B \subset V$ be a subset of V . Then the following are equivalent:

- (i) B is a basis of V
- (ii) B is a maximal l.i. subset of V
- (iii) B is a minimal spanning subset of V .

Def:- Let A be an $m \times n$ matrix.

A matrix obtained from A by deleting some rows & some columns of A is

is called a submatrix of A .

Let $A = [a_{ij}]_{m \times n}$.

A submatrix denoted $A(i_1, \dots, i_r | j_1, \dots, j_s)$ of A is the matrix of size $r \times s$ where entries are as

$$A(i_1, \dots, i_r | j_1, \dots, j_s) = \begin{bmatrix} a_{i_1 j_1} & a_{i_1 j_2} & \dots & a_{i_1 j_s} \\ a_{i_2 j_1} & a_{i_2 j_2} & \dots & a_{i_2 j_s} \\ \vdots & \vdots & \ddots & \vdots \\ a_{i_r j_1} & a_{i_r j_2} & \dots & a_{i_r j_s} \end{bmatrix}_{r \times s}$$

where

$$1 \leq i_1 < \dots < i_r \leq m$$

$$1 \leq j_1 < \dots < j_s \leq n.$$

Def:- The determinants of square submatrices of A are called the minors of A .

Example ① $A = \begin{bmatrix} 1 & 2 & -1 & 4 \\ 0 & 1 & 3 & 5 \\ -1 & 0 & 1 & -1 \end{bmatrix}$

2x2 minors: $\begin{vmatrix} 1 & 2 \\ 0 & 1 \end{vmatrix}, \begin{vmatrix} 1 & -1 \\ 0 & 3 \end{vmatrix}, \begin{vmatrix} 3 & 5 \\ 1 & -1 \end{vmatrix}, \begin{vmatrix} 2 & 4 \\ 0 & -1 \end{vmatrix}$

3x3 minors: $|A(1,2,3 | 1,2,3)| = \begin{vmatrix} 1 & 2 & -1 \\ 0 & 1 & 3 \\ -1 & 0 & 1 \end{vmatrix}$
 $|A(1,2,3 | 1,3,4)| = \begin{vmatrix} 1 & -1 & 4 \\ 0 & 3 & 5 \\ -1 & 1 & -1 \end{vmatrix}$

• $\begin{bmatrix} 1 & 3 \\ -1 & 1 \end{bmatrix}$ is not a submatrix of A .

Defn Let $A_{m \times n}$ be an $m \times n$ matrix, say $A \neq \underline{0}$.

Then the rank of A is defined as the integer $r > 0$ such that there exists a $r \times r$ non-zero minor₁^{of A} and all minors of A of size $> r$ are all zeros.

i.e.) Rank of A is the largest integer r such that there exists an $r \times r$ minor which is non-zero.

Def:- The determinants of the submatrices of the form $A(i_1, \dots, i_r | i_1, \dots, i_r)$ are called the principal minors of $A_{m \times n}$ where $1 \leq i_1 < \dots < i_r \leq \min\{m, n\}$

Def:- The leading principal minors of $A_{n \times n}$ are determinants of $A(1|1)$, $A(1, 2|1, 2)$, \dots , $A(1, 2, \dots, n|1, 2, \dots, n)$.

Recall:- ① Let $B_{m \times n}$ be a matrix which row equivalent to $A_{m \times n}$. Then

$$\text{rank}[A] = \text{rank}[B].$$

Defn Let V, W be vector spaces over a field F .
A map $T: V \rightarrow W$ is said to be a
linear transformation (l.t), if

$$(i) \quad T(\underline{v}_1 + \underline{v}_2) = T(\underline{v}_1) + T(\underline{v}_2) \quad \forall \underline{v}_1, \underline{v}_2 \in V.$$

$$(ii) \quad T(\lambda \underline{v}) = \lambda T(\underline{v}) \quad \forall \underline{v} \in V, \quad \forall \lambda \in F.$$

Examples:- ① Let $A_{m \times n}$ be an $m \times n$ real matrix.
Then define a ~~map~~ map $T: \mathbb{R}^n \rightarrow \mathbb{R}^m$ as

$$T(\underline{v}) = A\underline{v} \quad \forall \underline{v} \in \mathbb{R}^n.$$

check that T is a l.t.

$$\textcircled{2} \quad T: \mathbb{R}^3 \rightarrow \mathbb{R}^2, \quad T(x, y, z) = (x+y, y+z)$$

l.t.

Convention:- All vectors in \mathbb{R}^n we write as
column vectors (column matrix)

$$\begin{aligned} \text{i.e., } \mathbb{R}^n &= \left\{ \begin{pmatrix} a_1 \\ \vdots \\ a_n \end{pmatrix} \mid a_1, \dots, a_n \in \mathbb{R} \right\} \\ &= \left\{ (a_1, \dots, a_n) \mid a_1, \dots, a_n \in \mathbb{R} \right\} \end{aligned}$$

③ $T: V \rightarrow W, \quad T(\underline{v}) = \underline{0} \quad \forall \underline{v} \in V.$
is a l.t called zero transformation.

Defn Let $T: V \rightarrow W$ be a L.T. Then the nullspace of T or kernel of T is

defined as $N(T) := \{v \in V \mid T(v) = 0\}$

check that $N(T) \subseteq V$ is a subspace of V

& its dimension is called the nullity of T .

i.e., $\text{nullity}(T) := \dim(N(T))$.

Defn Let $T: V \rightarrow W$ be a L.T. Then the range of T or image of T is defined

as $R(T) := \{T(v) \mid v \in V\}$

$= \{w \in W \mid w = T(v) \text{ for some } v \in V\}$

check that $R(T)$ is a subspace of W .

& its dimension is called the rank of T .

i.e., $\text{rank}(T) = \dim(R(T))$.

Theorem (Rank-nullity thm)

Let V, W be vector space over a field F . Let $T: V \rightarrow W$ be a L.T. Then

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

i.e., $\dim(R(T)) + \dim(N(T)) = \dim(V)$.

Proposition:— Let $A_{m \times n}$, $B_{n \times p}$ be matrices.

Then
$$\text{rank}(AB) \leq \min \{ \text{rank}(A), \text{rank}(B) \}.$$

Proof:—

We will show: $\text{rank}(AB) \leq \text{rank}(A)$
& $\text{rank}(AB) \leq \text{rank}(B).$

claim 1: $N(B) \subseteq N(AB).$

Pf of claim 1:— Let $\underline{v} \in N(B).$ $\left\{ \begin{array}{l} B: \mathbb{R}^p \rightarrow \mathbb{R}^n \\ B(\underline{v}) = B\underline{v} \\ \forall \underline{v} \in \mathbb{R}^p. \end{array} \right.$

$$\Rightarrow B\underline{v} = \underline{0}.$$

$$\Rightarrow A(B\underline{v}) = A\underline{0} = \underline{0}$$

$$\Rightarrow (AB)\underline{v} = \underline{0}$$

$$\Rightarrow \underline{v} \in N(AB)$$

$\therefore N(B) \subseteq N(AB) \subseteq \mathbb{R}^p$ subspaces.

$$\Rightarrow \dim(N(B)) \leq \dim(N(AB))$$

(\because if $W \subseteq V$ subsp. then $\dim(W) \leq \dim(V$)

$$\Rightarrow \underbrace{p - \dim(N(B))}_{\substack{\parallel \\ \text{rank}(B)}} \geq \underbrace{p - \dim(N(AB))}_{\substack{\parallel \\ \text{rank}(AB)}}$$

by rank-nullity Thm.

$$\therefore \boxed{\text{rank}(AB) \leq \text{rank}(B).}$$

claim 2: $\text{rank}(AB) \leq \text{rank}(A)$

proof of lem 2:

Let $\underline{w} \in R(AB)$

$$\Rightarrow \underline{w} = (AB)(\underline{v}) \text{ for some } \underline{v}.$$

$$\Rightarrow \underline{w} = A(\underbrace{B\underline{v}}_{\substack{\in R(A) \\ \parallel \\ B\underline{v}}}) = A(\underbrace{\underline{u}}_{\substack{\parallel \\ B\underline{v}}}) \text{ for some } \underline{u}.$$

$$\therefore \underline{w} \in R(A)$$

$$\therefore R(AB) \subseteq R(A)$$

$$\Rightarrow \dim(R(AB)) \leq \dim(R(A))$$

$\parallel \qquad \qquad \parallel$
 $\text{rank}(AB) \qquad \text{rank}(A)$

$$\therefore \boxed{\text{rank}(AB) \leq \text{rank}(A)}$$

Eg:- Let A be any $m \times n$ real matrix.

① Define a L.T $T_A: \mathbb{R}^n \rightarrow \mathbb{R}^m$

$$T_A(\underline{v}) = A\underline{v} \quad \forall \underline{v} \in \mathbb{R}^n.$$

$$A \mapsto (T_A)$$

Corollary (consequence):-

Let $A_{n \times n}$ be a matrix &

Suppose $P_{n \times n}$ is an invertible matrix.

$$\text{Then } \text{rank}(A) = \text{rank}(PA) = \text{rank}(AP)$$

proof:- $\text{rank}(A) = \text{rank}(P(PA))$

$$\leq \text{rank}(PA) \\ \leq \text{rank}(A).$$

$$\Rightarrow \text{rank}(A) = \text{rank}(PA).$$

Let $A_{m \times n}$ be an $m \times n$ matrix.

$$A = [\underline{c}_1 \ \underline{c}_2 \ \dots \ \underline{c}_n]$$

where $\underline{c}_1, \underline{c}_2, \dots, \underline{c}_n$ are column vectors of A which lies in \mathbb{R}^m .

By $A = \begin{bmatrix} \underline{r}_1 \\ \underline{r}_2 \\ \vdots \\ \underline{r}_m \end{bmatrix}$ where $\underline{r}_1, \dots, \underline{r}_m$ are the row vectors of A which lies in \mathbb{R}^n .

Blocks of matrices.

Definition: Let A be an $m \times n$ matrix.

Then the row space of A is defined as the subspace spanned by the row vectors of A in \mathbb{R}^n .

$$\text{i.e. } \text{row space}(A) = \text{span}\{\underline{r}_1, \dots, \underline{r}_m\} \subseteq \mathbb{R}^n.$$

Where $\underline{r}_1, \dots, \underline{r}_m$ are the rows of A .

Also the column space of A is defined as

Def: The column space of A is defined as the subspace of \mathbb{R}^m spanned by the column vectors of A .

i.e., $\text{ColumnSpace}(A) = \text{Span}(\{\underline{c}_1, \dots, \underline{c}_n\}) \subseteq \mathbb{R}^m$
where $\underline{c}_1, \dots, \underline{c}_n$ are the columns of A .

Def: The dimension of row space of $A_{m \times n}$ is called the rowrank of A .

By the dimension of column space of $A_{m \times n}$ is called the Columnrank of A .

Observations:

$$\begin{aligned} \textcircled{1} \text{rowSpace}(A_{m \times n}) &= \left\{ \lambda_1 \underline{x}_1 + \dots + \lambda_n \underline{x}_n \mid \lambda_1, \dots, \lambda_n \in \mathbb{R} \right\} \\ &= \left\{ [\underline{x}_1 \dots \underline{x}_n] \begin{pmatrix} \lambda_1 \\ \vdots \\ \lambda_n \end{pmatrix} \mid \lambda_i \in \mathbb{R} \forall i \right\} \\ &= \left\{ A^t \begin{bmatrix} \lambda_1 \\ \vdots \\ \lambda_n \end{bmatrix} \mid \lambda_i \in \mathbb{R} \forall i \right\} \\ &= R(A^t). \end{aligned}$$

$$\textcircled{2} \text{ColumnSpace}(A_{m \times n}) = \left\{ \lambda_1 \underline{c}_1 + \dots + \lambda_n \underline{c}_n \mid \lambda_i \in \mathbb{R} \right\}$$

$$\begin{aligned}
 &= \left\{ \left[c_1 \ c_2 \ \dots \ c_n \right] \begin{bmatrix} \lambda_1 \\ \vdots \\ \lambda_n \end{bmatrix} \mid \lambda_i \in \mathbb{R} \ \forall i \right\} \\
 &= \left\{ A \begin{bmatrix} \lambda_1 \\ \vdots \\ \lambda_n \end{bmatrix} \mid \lambda_i \in \mathbb{R} \ \forall i \right\} \\
 &= R(A).
 \end{aligned}$$

Proposition:— let $A_{m \times n}$ be an $m \times n$ matrix.
 Then $\text{rank}(A) = \text{rank}(A^t)$.

proof:

claim:— $A\underline{x} = \underline{0} \iff A^t A \underline{x} = \underline{0}$
 $\forall \underline{x} \in \mathbb{R}^n$.

Pf of claim:

\Rightarrow : Assume $A\underline{x} = \underline{0}$.

$$\Rightarrow A^t(A\underline{x}) = A^t \underline{0} = \underline{0}$$

$$\Rightarrow (A^t A) \underline{x} = \underline{0}.$$

\Leftarrow : Assume $A^t A \underline{x} = \underline{0}$.

To show: $A\underline{x} = \underline{0}$.

We have $A^t A \underline{x} = \underline{0}$

$$\Rightarrow \underbrace{x^t}_{1 \times n} \underbrace{A^t}_{n \times m} \underbrace{A}_{m \times n} \underbrace{x}_{n \times 1} = \underbrace{x^t}_0 = 0$$

$$\Rightarrow (Ax)^t (Ax) = 0 \quad \left(\because (AB)^t = B^t A^t \right)$$

$$\Rightarrow |Ax|^2 = 0$$

$$\Rightarrow |Ax| = 0$$

$$\Rightarrow Ax = 0.$$

as required

$$\left| \begin{array}{l} \underline{y} = \begin{pmatrix} y_1 \\ \vdots \\ y_m \end{pmatrix} \\ |\underline{y}| = \sqrt{y_1^2 + \dots + y_m^2} \\ \underline{y}^t \underline{y} = y_1^2 + \dots + y_m^2 \\ \quad = |\underline{y}|^2 \end{array} \right.$$

Thus the claim is true.

$$\begin{aligned} \text{claim } \Rightarrow N(A) &= \left\{ x \in \mathbb{R}^n \mid Ax = \underline{0} \right\} \\ &= \left\{ x \in \mathbb{R}^n \mid A^t Ax = 0 \right\} \\ &= N(A^t A). \end{aligned}$$

$$\therefore \boxed{N(A) = N(A^t A)}$$

$$\Rightarrow \text{nullity}(A) = \text{nullity}(A^t A)$$

$$\Rightarrow n - \text{nullity}(A) = n - \text{nullity}(A^t A)$$

$$\Rightarrow \text{rank}(A) = \text{rank}(A^t A)$$

(By rank-nullity theorem).

$$\leq \text{rank}(A^t).$$

$$(\because \text{rk}(AB) \leq \text{rk}(A))$$

$$\therefore \text{rank}(A) \leq \text{rank}(A^t).$$

If we can prove $\text{rank}(A^t) \leq \text{rank}(A)$
by considering A^t instead of A
in the above proof. (EXERCISE)

Thus

$$\boxed{\text{rank}(A) = \text{rank}(A^t)}.$$