

## LECTURE 31

I) The Four Fundamental Subspaces  
of a Matrix.

II) Row Rank = Column Rank

III) Pair wise Orthogonality

IV) Direct Sum Decomposition  
of Domain and Target space

V) Orthogonal Complementarity

31.1

### I) THE FOUR FUNDAMENTAL SUBSPACES OF A LINEAR TRANSFORMATION.

Consider an  $m \times n$  matrix

$$A: \mathbb{R}^m \rightarrow \mathbb{R}^n; A = \begin{bmatrix} & & \\ & & \\ & & \end{bmatrix} \begin{matrix} m \\ n \end{matrix}$$

and its transpose

$$A^T: \mathbb{R}^n \rightarrow \mathbb{R}^m$$

Thus one may have

$$A = \begin{bmatrix} \begin{matrix} m \\ n \end{matrix} \\ \begin{matrix} m \\ n \end{matrix} \end{bmatrix} \text{ and } A^T = \begin{bmatrix} \begin{matrix} m \\ n \end{matrix} \\ \begin{matrix} m \\ n \end{matrix} \end{bmatrix}$$

"skinny"                      "squat"

$$A = \begin{bmatrix} \begin{matrix} m \\ n \end{matrix} \\ \begin{matrix} m \\ n \end{matrix} \end{bmatrix} \text{ and } A^T = \begin{bmatrix} \begin{matrix} m \\ n \end{matrix} \\ \begin{matrix} m \\ n \end{matrix} \end{bmatrix}$$

"squat"                      "skinny"

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Such matrices are characterized by its four fundamental subspaces;

$$R(A) = (\text{column space of } A) \subset \mathbb{R}^n; \dim = r$$

$$N(A) = (\text{null space of } A) \subset \mathbb{R}^n; \dim = n - r$$

$$R(A^T) = (\text{row space of } A) \subset \mathbb{R}^m; \dim = r$$

$$N(A^T) = (\text{left null space of } A) \subset \mathbb{R}^m; \dim = m - r$$

### II) ROW RANK = COLUMN RANK

It is blindingly obvious that row space  $R(A)$  and the column space  $R(A)$  of  $A$

are in general two entirely different

vector spaces. Nevertheless one has the

following

Theorem (Row rank = column rank)

$$\dim R(A) = \dim R(A^T)$$

i.e. column rank of  $A =$  row rank of  $A$ .

The validity of this theorem is established by means of 4 Step line of reasoning;

Step 1 Row reduction to echelon form

$$A = \begin{bmatrix} a_{11} & \dots & a_{1n} \\ \vdots & & \vdots \\ a_{m1} & \dots & a_{mn} \end{bmatrix} \xrightarrow{\text{row reduction}} U = \begin{bmatrix} \boxed{1} & \dots & \dots & \dots & \dots \\ 0 & \dots & \dots & \dots & \dots \\ \vdots & & \vdots & & \vdots \\ 0 & \dots & \dots & \dots & \dots \\ \vdots & & \vdots & & \vdots \\ 0 & \dots & \dots & \dots & \dots \\ \vdots & & \vdots & & \vdots \\ 0 & \dots & \dots & \dots & \dots \\ \vdots & & \vdots & & \vdots \\ 0 & \dots & \dots & \dots & \dots \\ \vdots & & \vdots & & \vdots \\ 0 & \dots & \dots & \dots & \dots \end{bmatrix} \begin{matrix} \left. \begin{matrix} u_1 \\ \vdots \\ u_k \end{matrix} \right\} \\ \left. \begin{matrix} u_{k+1} \\ \vdots \\ u_m \end{matrix} \right\} \end{matrix}$$

Step 2

Noting that  $\{u_1, \dots, u_k\}$  is a basis for

$R(A^T)$ , we have

$$\boxed{\dim R(A^T) = k} \quad (= \text{"row rank"})$$

Step 3

Note that:

(i) The reversibility of every row operation guarantees that every row of  $A$  is a linear combination of the rows  $u_1, \dots, u_k$ . The coefficients of these combinations form a matrix  $[C_{ij}]$  which relates  $A$  to  $U$ .

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$$\begin{bmatrix} a_{11} & \dots & a_{1n} \\ \vdots & & \vdots \\ a_{m1} & \dots & a_{mn} \end{bmatrix} = \begin{bmatrix} c_{11} & \dots & c_{1k} & \dots & 0 & \dots & 0 \\ c_{21} & \dots & c_{2k} & \dots & 0 & \dots & 0 \\ \vdots & & \vdots & & \vdots & & \vdots \\ c_{m1} & \dots & c_{mk} & \dots & 0 & \dots & 0 \end{bmatrix} \begin{bmatrix} u_1 \\ \vdots \\ u_k \\ u_{k+1} \\ \vdots \\ u_m \end{bmatrix}$$

$$A = CU$$

The matrix  $C$  is a unique expression of the reversible row reduction process  $A \leftrightarrow U$

Indeed, we have

Step 4

$$i^{\text{th}} \text{ row of } A = \sum_{j=1}^k c_{ij} [u_j]$$

On the other hand,  $CU$  is a family of linear combinations of the columns of  $C$ , each linear combination being a column of  $A$ .

$$j^{\text{th}} \text{ column of } A = \sum_{i=1}^m c_{ij} [u_i]$$

Thus

$$R(A) = \text{sp} \{c_1, \dots, c_k\}$$

and so

$$\dim R(A) \leq k = \dim R(A^T)$$

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By reversing the roles of  $A$  and  $A^T$  one obtains

$$\dim R(A^T) \leq \dim R(A)$$

and hence

$$\boxed{\dim R(A^T) = \dim R(A)} = \text{"rank of } A \text{"}$$

Thus, although  $R(A^T)$  and  $R(A)$  are

different vector spaces, their dimensions

are the same.

### III) PAIRWISE ORTHOGONALITY

The second important property of 31.6

the four fundamental subspaces is their

pairwise orthogonality,

$$\text{In } \mathbb{R}^m \quad R(A) \perp N(A^T)$$

$$\text{In } \mathbb{R}^n \quad R(A^T) \perp N(A)$$

This property is based on the

Definition

Two subspaces  $V$  and  $W$  of  $\mathbb{R}^n$  are said to be orthogonal ( $V \perp W$ ) if every  $v \in V$  is orthogonal to every  $w \in W$ ,

and on the

Theorem

For any  $m \times n$  matrix  $A$ :  $\mathbb{R}^m \rightarrow \mathbb{R}^m$  on  $\mathbb{R}^m$  has

$$(a) \quad N(A) \perp R(A^T) \quad \text{in } \mathbb{R}^m$$

$$(b) \quad N(A^T) \perp R(A) \quad \text{in } \mathbb{R}^n$$

The proof is by inspecting the elements of the two nullspaces. 3.1.7

(A)  $w \in \mathcal{N}(A) \Rightarrow Aw = 0$

$$\begin{bmatrix} \text{row } 1 \\ \vdots \\ \text{row } m \end{bmatrix} \begin{bmatrix} w_1 \\ \vdots \\ w_m \end{bmatrix} = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$$

m equations

(i<sup>th</sup> column of  $A^T$ )<sup>T</sup>  $w = 0$

$\mathcal{R}(A^T) \perp \mathcal{N}(A) \quad \forall w \in \mathcal{N}(A)$   
 $\therefore \mathcal{R}(A^T) \perp \mathcal{N}(A)$

(B)  $y \in \mathcal{N}(A^T) \Rightarrow (y^T A = 0 \Leftrightarrow A^T y = 0)$

$$\begin{bmatrix} y_1 \dots y_m \end{bmatrix} \begin{bmatrix} | & & | \\ \text{col } 1 & \dots & \text{col } m \\ | & & | \end{bmatrix} = \begin{bmatrix} 0 & \dots & 0 \end{bmatrix}$$

m eq'ns

$y \perp \mathcal{R}(A) \quad \forall y \in \mathcal{N}(A^T)$

$\therefore \mathcal{N}(A^T) \perp \mathcal{R}(A)$

Comment: These orthogonalities imply that  $A$  induces a unique decomposition of the domain space  $\mathbb{R}^m$  and of the target space  $\mathbb{R}^n$  as follows

a) domain  $A = \mathbb{R}^m = \mathcal{N}(A) \oplus \mathcal{R}(A^T)$

b) domain  $A^T = \mathbb{R}^n = \mathcal{N}(A^T) \oplus \mathcal{R}(A)$

31.9

Comment:

The orthogonalities of the subspaces of  $A$  constitutes the Fundamental Theorem of Linear Algebra.

It is stated most succinctly in terms of orthogonal complements.

Definition

Given a subspace  $W \subset \mathbb{R}^n$ , the space of vectors orthogonal to  $W$  is called the orthogonal complement of  $W$  and is denoted by  $W^\perp$ .

FUNDAMENTAL THEOREM OF LINEAR ALGEBRA

$$N(A) = (R(A^T))^\perp \quad R(A^T) = (N(A))^\perp$$

$$N(A^T) = (R(A))^\perp \quad R(A) = (N(A^T))^\perp$$

