

$$5 \begin{bmatrix} 1 \\ 3 \end{bmatrix} + 6 \begin{bmatrix} 2 \\ 4 \end{bmatrix} = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \begin{bmatrix} 5 \\ 6 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \begin{bmatrix} 5 & 7 \\ 6 & 8 \end{bmatrix} = \begin{bmatrix} \uparrow & \rightarrow \\ | & \end{bmatrix} \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \begin{bmatrix} 7 \\ 8 \end{bmatrix} = 7 \begin{bmatrix} 1 \\ 3 \end{bmatrix} + 8 \begin{bmatrix} 2 \\ 4 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \begin{bmatrix} 5 & 7 \\ 6 & 8 \end{bmatrix} = \begin{bmatrix} \leftarrow & \\ \rightarrow & \end{bmatrix} \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \begin{bmatrix} 5 & 7 \\ 6 & 8 \end{bmatrix} = \begin{bmatrix} 5 & 7 \\ 6 & 8 \end{bmatrix} + 2 \begin{bmatrix} 6 & 8 \end{bmatrix}$$

$$= 3 \begin{bmatrix} 5 & 7 \\ 6 & 8 \end{bmatrix} + 4 \begin{bmatrix} 6 & 8 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \begin{bmatrix} 5 & 7 \\ 6 & 8 \end{bmatrix} = \begin{bmatrix} 1 \\ 3 \end{bmatrix} \begin{bmatrix} 5 & 6 \end{bmatrix} + \begin{bmatrix} 2 \\ 4 \end{bmatrix} \begin{bmatrix} 7 & 8 \end{bmatrix}$$

$$A = \begin{bmatrix} 1 & 3 & 8 \\ 1 & 2 & 6 \\ 0 & 1 & 2 \end{bmatrix} = \begin{bmatrix} 1 & 3 \\ 1 & 2 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 2 \\ 0 & 1 & 2 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \\ 0 \\ 1 \\ 2 \end{bmatrix}$$

A is $m \times n$

Column space

$\text{Col}(A)$

$= \text{span} \left\{ \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 3 \\ 2 \\ 1 \end{bmatrix} \right\}$ in \mathbb{R}^m

$\mathbb{C} \begin{bmatrix} 1 \\ 0 \end{bmatrix}$

$\mathbb{C} \begin{bmatrix} 0 \\ 1 \end{bmatrix}$

$\mathbb{C} \begin{bmatrix} 2 \\ 2 \end{bmatrix} = 2 \begin{bmatrix} 1 \\ 1 \end{bmatrix} + 2 \begin{bmatrix} 3 \\ 2 \\ 1 \end{bmatrix}$

$\text{Row}(A) = \mathbb{C}(A^T) = \text{span} \left\{ \begin{bmatrix} 1 \\ 0 \\ 2 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 2 \end{bmatrix} \right\}$ in \mathbb{R}^n

$\text{Null}(A) = \left\{ x \in \mathbb{F}^n : \begin{matrix} n \times n \\ m \times n \end{matrix} Ax = 0 \text{ in } \mathbb{F}^m \right\} / \mathbb{F} = \mathbb{R} \sim \mathbb{C}$

Example. $\begin{bmatrix} 1 & 3 & 8 & | & 0 \\ 1 & 2 & 6 & | & 0 \\ 0 & 1 & 2 & | & 0 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 0 & 2 & | & 0 \\ 0 & 1 & 2 & | & 0 \\ 0 & 0 & 0 & | & 0 \end{bmatrix}$ $\begin{bmatrix} -2t \\ -2t \\ t \end{bmatrix} = t \begin{bmatrix} -2 \\ -2 \\ 1 \end{bmatrix}$

$\text{Null}(A) = \text{span} \left\{ \begin{bmatrix} -2 \\ -2 \\ 1 \end{bmatrix} \right\}$

$1+2t=0$

Left null space of A .

$\text{Null}(A^T) = \left\{ \begin{matrix} n \times m \\ m \times n \end{matrix} y \in \mathbb{R}^m : A^T y = 0 \right\} \rightarrow n \times 1$
 $= \left\{ y \in \mathbb{R}^m : y^T A = 0 \right\}$ in $1 \times n$

Example. $\text{Null}(A^T)$:

$\begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 2 & 2 \end{bmatrix} \begin{bmatrix} 1 & 1 & 0 \\ 3 & 2 & 1 \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix} = \begin{bmatrix} 1 & 1 & 0 \\ 3 & 2 & 1 \\ 8 & 6 & 2 \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$

In $\mathbb{R}^{m \times n}$
Remark: $A = \begin{bmatrix} u_1^T \\ \vdots \\ u_m^T \end{bmatrix}$

$Ax = 0$ means.

x is orthogonal to the u_1, \dots, u_m

$$\langle x, y \rangle = \left\langle \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix}, \begin{pmatrix} y_1 \\ \vdots \\ y_n \end{pmatrix} \right\rangle = x_1 y_1 + \dots + x_n y_n$$

and $\langle x, y \rangle = 0$ means / called
 x and y are orthogonal.

i.e., x is orthogonal to v_1, \dots, v_r

where $\text{Row}(A) = C(A^T) = \text{span}\{v_1, \dots, v_r\}$

Matrix factorization.

$$A = \begin{bmatrix} a & 1 \\ 1 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ x & x \end{bmatrix} \begin{bmatrix} a & x \\ 0 & x \end{bmatrix}$$

$A = LU$, $A = QR$, $A = RDR^{-1}S$ $A = Q\Lambda Q^T$ if A is symmetric, $A = U\Sigma V^T$.

1. LU factorization for invertible $A \in M_n$.

To solve $Ax = b$ for many different vectors $b \in \mathbb{F}^n$.

$$Ax = b_1, Ax = b_2, \dots, Ax = b_2 \dots$$

Write $A = LU$ where L is lower triangular, and U as upper triangular.

Then solve $Ax = LUx = b$ by solving $Ly = b$ and $Ux = y$.

$$\begin{bmatrix} u & 10 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \end{bmatrix} \begin{bmatrix} x_1^T \\ A_2^T \\ A_3^T \end{bmatrix}$$

How to write $LU = A$? We will see this in Section 1.4

One can do an exchange of rows by a permutation matrix P to get $PA = LU$ if A is invertible.

$$= \begin{bmatrix} A_2^T \\ A_3^T \\ A_1^T \end{bmatrix}$$

Question What if A is not invertible or A is not a square matrix?

2. QR Decomposition for $A \in M_{m,n}$ with $m \geq n$.

Recall the inner product on \mathbb{F}^n defined by $\langle x, y \rangle = y^*x = \sum_{j=1}^n x_j y_j^*$, and the inner product norm $\|x\| = \langle x, x \rangle^{1/2}$.

We are interested in orthonormal set/basis $\{v_1, \dots, v_n\}$ such that $V^*V = I_n$ if $V = [v_1 | \dots | v_n]$, i.e., $\langle v_i, v_j \rangle = \delta_{ij}$.

To solve $Ax = b$ with b in the column space of A , we can solve $Rx = Q^*Ax = Q^*b$.

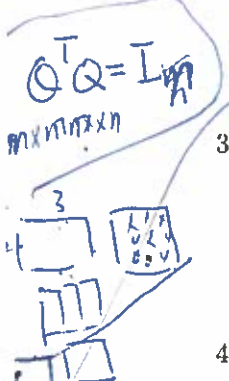
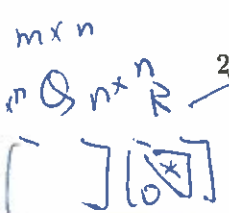
V^T real
Complex $\rightarrow Q^T$ real
Complex $\rightarrow Q^*$ complex

3. $A = RDR^{-1} = \sum_{j=1}^n \lambda_j x_j y_j^T$ where R has columns x_1, \dots, x_n and R^{-1} has rows y_1^T, \dots, y_n^T . Then $A^k = \sum_{j=1}^n \lambda_j^k x_j y_j^T$.

$$A = \begin{bmatrix} R \end{bmatrix} \begin{bmatrix} \lambda_1 & 0 \\ 0 & \lambda_n \end{bmatrix} R^{-1}$$

If $H = H^*$ is Hermitian, then $H = Q\Lambda Q^* = \sum_{j=1}^n \lambda_j v_j v_j^*$ with $\lambda_1, \dots, \lambda_n \in \mathbb{R}$ so that $H^k = \sum_{j=1}^n \lambda_j^k v_j v_j^*$.

4. $A = U\Sigma V^*$. Then $A = \sum_{j=1}^r s_j u_j v_j^*$.



Remark

Solving $Ax = b$ by $x = A^{-1}b$ is not a good way

because A^{-1} may be bad, A is ill-conditioned.

Example $\begin{bmatrix} 1 & 1 \\ 1 & 1.00000001 \end{bmatrix}$

$$A = \begin{bmatrix} 1 & 0 \\ x & 1 \end{bmatrix} \begin{bmatrix} 1 & x \\ 0 & 1 \end{bmatrix}$$

$$Dx = b$$

$$\begin{pmatrix} d_1 & 0 \\ 0 & d_n \end{pmatrix} \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} b_1 \\ \vdots \\ b_n \end{bmatrix}$$

$$\begin{bmatrix} 1 & 1 \\ 1 & 1.00000001 \end{bmatrix}^{-1} \begin{bmatrix} 1 \\ 2 \end{bmatrix} = \begin{bmatrix} 1 \\ 2+0.01 \end{bmatrix}$$

$A : m \times n = \begin{matrix} \boxed{} \\ \boxed{} \\ \boxed{} \end{matrix} = \begin{matrix} \boxed{} \\ \boxed{} \\ \boxed{} \end{matrix} \begin{matrix} \boxed{} \\ \boxed{} \\ \boxed{} \\ \boxed{} \end{matrix}$

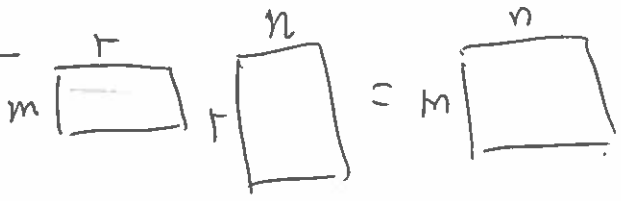
$A = \begin{bmatrix} 1 & 4 & 7 & 10 \\ 2 & 5 & 8 & 11 \\ 3 & 6 & 9 & 12 \end{bmatrix}$

I.3 The Four Fundamental Subspaces

- Let $A \in M_{m,n}$. We have
- the column space $C(A)$ in \mathbb{F}^m contains all combination of columns of A ,
 - the null space $N(A)$ contains x in \mathbb{F}^n such that $Ax = 0$.
 - the row space $C(A^T)$ contains all combination of columns of A^T ,
 - the left null space $N(A^T)$ contains y in \mathbb{F}^m such that $A^T y = 0$.

Proposition Let $A \in M_{m,n}$. Then $C(A)$ and $C(A^T)$ have the same dimension r ; $N(A)$ has dimension $n - r$, $N(A^T)$ has dimension $m - r$.

Let A, B be matrices
~~Theorem Let $A \in M_{m,r}, B \in M_{r,n}$.~~



- $\text{rank}(AB) \leq \min\{\text{rank}(A), \text{rank}(B)\}$.
- $\text{rank}(A+B) \leq \text{rank}(A) + \text{rank}(B)$.
- $\text{rank}(A^*A) = \text{rank}(AA^*) = \text{rank}(A) = \text{rank}(A^*)$.
- $\text{rank}(AB) = r$ if $\text{rank}(A) = \text{rank}(B) = r$.

$A B_i = \begin{bmatrix} b_{1i} \\ \vdots \\ b_{ri} \end{bmatrix} = b_{1i}u_1 + \dots + b_{ri}u_r$

$A = [u_1 \dots u_r]$

Proof ① Col space of $AB = [AB_1 \dots AB_n]$

$= \begin{bmatrix} A_1^T B \\ \vdots \\ A_m^T B \end{bmatrix}$

Col(AB) in
 the col(A)

$\text{Row}(AB) = \text{Col}(AB)^T$

$\therefore \dim \text{Col}(AB) \leq \dim \text{Col}(A)$

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

Also $\text{Row}(AB) \subseteq \text{Row}(B)$

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

$\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \begin{bmatrix} 5 & 7 \\ 6 & 8 \end{bmatrix}$

$= \begin{bmatrix} A_1^T \\ A_2^T \end{bmatrix} A \begin{bmatrix} 5 \\ 7 \end{bmatrix}$

$= \left(5 \cdot \begin{bmatrix} 1 \\ 3 \end{bmatrix} + 7 \cdot \begin{bmatrix} 2 \\ 4 \end{bmatrix} \right)$

$$\textcircled{2} \quad \text{col}(A+B) \stackrel{\text{Span}}{\subseteq} \{\text{col}(A) \cup \text{col}(B)\}$$

Let

$$A = [A_1 | \dots | A_n] \quad B = [B_1 | \dots | B_n]$$

$$A+B = [A_1+B_1 | \dots | A_n+B_n]$$

$$\therefore \text{col}(A+B) \subseteq \text{Span}\{A_1, B_1, A_2, B_2, \dots, A_n, B_n\}$$

$$\subseteq \text{col} \begin{bmatrix} A_1 & B_1 & \dots & A_n & B_n \end{bmatrix}$$

$$\therefore \dim \text{col} [A_1 \dots A_n | B_1 \dots B_n] \leq \text{rank}(A) + \text{rank}(B) \quad \square$$

Remark: $\text{rank}(A+B) = \text{rank}(A) + \text{rank}(B)$

$$- \dim(\text{col}(A) \cap \text{col}(B))$$

$\textcircled{3}$

A is $m \times n$

$$\text{rank}(A^T A) = \text{rank}(A A^T)$$

$$= \text{rank}(A) = \text{rank}(A^T)$$

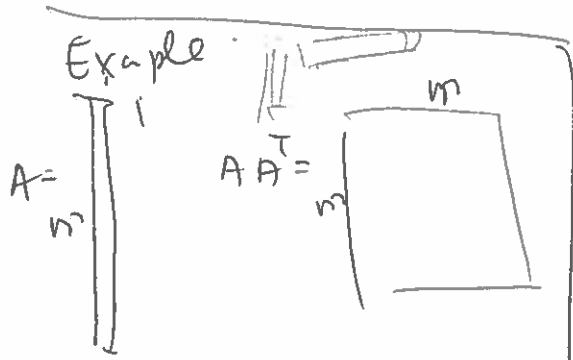
$$\text{rank}(A) + \dim \text{null}(A) = n$$

$$\text{rank}(A^T A) + \dim \text{null}(A^T A) = n$$

(Claim: $\text{null}(A) = \text{null}(A^T A)$)

(\subseteq) $x \in \text{null}(A) \Rightarrow Ax=0, \therefore A^T Ax=0 \Rightarrow x \in \text{null}(A^T A)$

(\supseteq) $x \in \text{null}(A^T A) \Rightarrow A^T Ax=0, \therefore x^T A^T Ax=0$
 i.e., $Ax = \begin{bmatrix} b_1 \\ \vdots \\ b_n \end{bmatrix}$ with $b_1^2 + \dots + b_n^2 = 0 \Rightarrow b_1 = \dots = b_n = 0$
 $\therefore x \in \text{null}(A)$



$$A^T A = [\cdot]$$

④

A is $m \times r$ B is $r \times n$

$\text{rank}(AB) = r$ if and only if $\text{rank}(A) = r$ and $\text{rank}(B) = r$

(\Rightarrow) If $\text{rank}(AB) = r$ then $\text{col space}(AB) = r$
and $\text{row space}(AB) = r$.

~~row space~~ $\rightarrow r \times n$

$\text{Row}(AB) \subseteq \text{Row}(B)$, must have dim r
 $\text{Col}(AB) \subseteq \text{Col}(A)$, must have dim r
 $\nwarrow m \times r$

(\Leftarrow) If $\text{rank}(A) = r$, $\text{rank}(B) = r$,

then $\text{col}(AB) = r$. by the CR decomposition interpretation

I.4 LU factorization

Recall that we want to write $A = LU$ for an invertible $A \in M_n$.

We can use elementary operations $E_r \dots E_1[A|I_n] = [U|R]$, $R = E_r \dots E_1$.

Then $RA = U$ so that $R = L^{-1} = E_1^{-1} \dots E_r^{-1}$.

A better way: $A - u_1 v_1^T = [0] \oplus A_1$ if we let $u_1 = (a_{11}, \dots, a_{n1})^T / a_{11}$ and $v_1^T = (a_{11}, \dots, a_{1n})$.

Then use induction on A_1 to get $A = \sum_{j=1}^n u_j v_j^T$.

Example

$$A = \begin{bmatrix} 1 & 2 & 3 \\ 2 & 5 & 7 \\ 2 & 7 & 8 \end{bmatrix} \xrightarrow{E_1 = \begin{bmatrix} 1 & 0 & 0 \\ -2 & 1 & 0 \\ -2 & 0 & 1 \end{bmatrix}} \begin{bmatrix} 1 & 2 & 3 \\ 0 & 1 & 1 \\ 0 & 3 & 2 \end{bmatrix} \xrightarrow{E_2 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & -3 & 1 \end{bmatrix}} \begin{bmatrix} 1 & 2 & 3 \\ 0 & 1 & 1 \\ 0 & 0 & -1 \end{bmatrix} = U.$$

i.e., $(E_2 E_1) A = U \quad \therefore A = E_1^{-1} E_2^{-1} U = L U$

$$L = \begin{bmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ 2 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & +3 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ 2 & 3 & 1 \end{bmatrix}$$

Remark We can apply the procedure as long as the (1,1) entry of A_k is nonzero in each step.

Else, one can apply a permutation P to A so that $PA = LU$.

Remark

$$\begin{bmatrix} 1 & & & \\ x & \triangle & & \\ & 0 & \triangle & \\ & & & 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ -x & \triangle & & \\ & 0 & \triangle & \\ & & & 1 \end{bmatrix} = \begin{bmatrix} 1 & & & \\ 0 & \triangle & & \\ 0 & & \triangle & \\ 0 & & & I_{n-1} \end{bmatrix}$$

$$\begin{bmatrix} 1 & 0 & \dots & 0 \\ x & 1 & \dots & 0 \\ & & \ddots & \\ & & & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 \\ & & \ddots & \\ 0 & & & C_{ij} \triangle & 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & \dots & 0 \\ x & 1 & \dots & 0 \\ & & \ddots & \\ & & & C_{ij} \triangle & 1 \end{bmatrix}$$

$$A \leftarrow \begin{bmatrix} 1 \\ 2 \\ 2 \end{bmatrix} \begin{bmatrix} 1 & 2 & 3 \end{bmatrix} = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 5 & 7 \\ 0 & 7 & 8 \end{bmatrix}$$

$$\begin{pmatrix} 0 & 0 & 0 \\ 0 & 5 & 7 \\ 0 & 7 & 8 \end{pmatrix} - \begin{bmatrix} 0 \\ 5/5 \\ 7/5 \end{bmatrix} \begin{bmatrix} 0 & 5 & 7 \end{bmatrix} = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 8 \end{bmatrix}$$

$$A - \begin{bmatrix} 1 \\ x \\ x \end{bmatrix} \begin{bmatrix} 1 & 2 & 3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \begin{bmatrix} 0 & 0 & 8 \end{bmatrix}$$