

# Matrix Computat<sup>n</sup> (Q in slides)

## Behind scenes

Slide 11:-

If  $\|\cdot\|_p$  is vector induced norm, then

$$\|I\|_F = \sup_{x \neq 0} \frac{\|x\|_F}{\|x\|_F} = 1 \quad \text{but } \|I\|_F \text{ by def}^n = \sqrt{n}$$

Slide 12:-

$$\|x\|_2 \leq \|x\|_1 \leq \sqrt{n} \|x\|_2$$

by squaring on both sides

AM  $\leq$  RMS

$$\text{i.e. } \frac{a+b+c+\dots}{n} \leq \sqrt{\frac{a^2+b^2+c^2+\dots}{n}}$$

2.) If

$$x = (x_1, x_2, \dots, x_n) \quad \text{let } |x_n| \geq |x_i| \forall i \text{ (}\therefore \|x\|_\infty = |x_n|\text{)}$$

$$\|x\|_2 = \sqrt{x_1^2 + x_2^2 + \dots + x_n^2} \leq \sqrt{n} |x_n| = \sqrt{n} \|x\|_\infty$$

$$\Downarrow$$

$$x_1^2 + x_2^2 + \dots + x_n^2 \leq n x_n^2$$

$$\|x\|_\infty \leq \|x\|_2$$

$$\Downarrow$$

$$|x_n| \leq \sqrt{x_1^2 + \dots + x_n^2}$$

true since  $x_i^2 \leq x_n^2 \forall i$

## 1) Cauchy Schwartz

$$\|A\|_1 = \sup_{x \neq 0} \frac{\|Ax\|_1}{\|x\|_1}$$

$$A = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{bmatrix}$$

$$= \frac{|a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n| + |a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n| + \dots + |a_{m1}x_1 + a_{m2}x_2 + \dots + a_{mn}x_n|}{|x_1| + |x_2| + \dots + |x_n|}$$

$$|a_i^T x| \leq \|a_i\|_1 \|x\|_1$$

From Cauchy Schwartz, for the equality to hold

$$a_i^T x = \|a_i\|_1 \|x\|_1$$

$$= \frac{|a_{11} + a_{12} \frac{x_2}{x_1} + \dots + a_{1n} \frac{x_n}{x_1}| + |a_{21} + a_{22} \frac{x_2}{x_1} + \dots + a_{2n} \frac{x_n}{x_1}|}{1 + \frac{|x_2|}{|x_1|} + \dots + \frac{|x_n|}{|x_1|}}$$

9)  $\|Ax\| \leq \|A\| \|x\|$  for a ~~matrix~~ vector norm induced matrix norm.  
 $\Leftarrow$  T equality holds for some  $x$ .

10)  $\|A\| = \sup_{x \neq 0} \frac{\|Ax\|}{\|x\|}$  so  $\exists$  a sequence  $(x_1, x_2, \dots, x_i)$  that

Let  $\frac{\|Ax_i\|}{\|x_i\|} = \|A\|$ . As  $\mathbb{R}^n$  is compact, for  $i \rightarrow \infty$   $x_i$ , equality holds.

See from Watkins Pg 1109

1)  $\|A\|_1 = \sup_{x \neq 0} \frac{\|Ax\|_1}{\|x\|_1}$  Let  $m$  be the column in  $A$  with the max column sum

$$= \frac{|a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n| + \dots}{\|x\|_1}$$

$$\leq \frac{|a_{11}x_1| + |a_{12}x_2| + \dots + |a_{1n}x_n| + \dots}{\|x\|_1}$$

$$= \frac{|a_{11}| + |a_{12}| + \dots + |a_{1n}|}{1} |x_1| + \frac{|a_{21}| + |a_{22}| + \dots + |a_{2n}|}{1} |x_2| + \dots$$

$$\leq \frac{\left( \sum_{i=1}^n |a_{im}| \right) |x_1| + \left( \sum_{i=1}^n |a_{im}| \right) |x_2| + \dots}{\|x\|_1}$$

$$= \sum_{i=1}^n |a_{im}|$$

And the equality satisfies for  $x = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 1 \\ \vdots \\ 0 \end{bmatrix}$  in  $m^{\text{th}}$  position

$$\|A\|_\infty = \sup_{x \neq 0} \frac{\|Ax\|_\infty}{\|x\|_\infty}$$

$$= \max_i \frac{|a_{i1}x_1 + a_{i2}x_2 + \dots + a_{in}x_n|}{\max_j |x_j|}$$

$$\leq \max_i (|a_{i1}| + |a_{i2}| + \dots + |a_{in}|) \max_j |x_j|$$

Let the maximum row sum occur at  $m$

Equality holds when  $x =$

$$\leq \frac{|a_{m1}x_1 + a_{m2}x_2 + \dots + a_{mn}x_n|}{\max_i |x_i|}$$

$$\leq \frac{|a_{m1}x_1| + |a_{m2}x_2| + \dots + |a_{mn}x_n|}{\max_i |x_i|}$$

$$\leq \frac{(|a_{m1}| + |a_{m2}| + \dots + |a_{mn}|) \max_i |x_i|}{\max_i |x_i|}$$

and equality occurs when  $x = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}$   $x_i = \begin{cases} +1 & \text{if } a_{mi} \geq 0 \\ -1 & \text{if } a_{mi} < 0 \end{cases}$

$$3) \|A\|_1 \leq \|A\|_2 \leq \|A\|_\infty$$

$$2.1.28 \text{ b) } \|A\|_2 = \sqrt{\lambda_{\max}(A^*A)} = \sigma_{\max}(A)$$

$$\|A\|_F = \sqrt{\sum \sigma_i^2(A)}$$

2.1.32  $\star$  - Proving

$$\|A\|_1 \leq \sqrt{n} \|A\|_2 \leq n \|A\|_\infty$$

$$\|A\|_\infty \leq \sqrt{n} \|A\|_2 \leq n \|A\|_1$$

# GE-1

Slide 15:-

8) P-T  $M_k = I_n - \begin{bmatrix} 0 \\ \vdots \\ m_{k+1,k} \\ \vdots \\ m_{nk} \end{bmatrix} e_k^T$

$M_k^{-1} = I_n + \begin{bmatrix} 0 \\ \vdots \\ m_{k+1,k} \\ \vdots \\ m_{nk} \end{bmatrix} e_k^T$

PF:-

$\begin{pmatrix} \begin{bmatrix} 0 \\ \vdots \\ m_{k+1,k} \\ \vdots \\ m_{nk} \end{bmatrix} e_k^T & \begin{bmatrix} 0 \\ \vdots \\ m_{k+1,k} \\ \vdots \\ m_{nk} \end{bmatrix} e_k^T \end{pmatrix}$

$= \begin{bmatrix} \text{Rest all zeros} \\ \vdots \\ m_{k+1,k} \\ \vdots \\ m_{nk} \end{bmatrix} \begin{bmatrix} \text{Rest all zeros} \\ \vdots \\ m_{k+1,k} \\ \vdots \\ m_{nk} \end{bmatrix}$

$\begin{bmatrix} 0 & \dots & 0 \\ \vdots & & \vdots \\ 0 & \dots & 0 \end{bmatrix} = 0$

GE-all  
Slide-21

LU decomposition is unique

PF:-  $A = \begin{bmatrix} 1 & & & \\ l_{11} & 1 & & \\ & l_{12} & l_{22} & \\ & & & \ddots \\ & & & & 1 \end{bmatrix} \begin{bmatrix} u_{11} & u_{12} & \dots & \\ & u_{22} & \dots & \\ & & \ddots & \\ 0 & & & (A) \end{bmatrix}$

~~Find all~~ If u know  $1^{st}$  k columns of L &  $1^{st}$  k rows of U, u can find  $(k+1)^{th}$  of L &  $(k+1)^{th}$  of U

Slide-28

Let  $P_k = P_{k+1} \dots P_{n-1}$ . Then  $P_k = \left[ \begin{array}{c|c} I_k & \\ \hline \tilde{P}_{k+1} & \dots & \tilde{P}_{n-1} \end{array} \right]$  where

$\tilde{P}_i$  are transposes of size  $n-k \times n-k$ .



For  $P_{n-2} = P_{n-1} = \left[ \begin{array}{c|c} I_{n-2} & \tilde{P}_{n-1} \\ \hline & \end{array} \right]_2$

Remember  $P_k = \left[ \begin{array}{c|c} I_{k-1} & \tilde{P}_k \\ \hline & \end{array} \right]_{n-k+1}$

$\tilde{P}_k$  is a transposition

Suppose true for  $k+1$  T.S.T true for  $k$

$P_{k+1} = P_{k+2} \dots P_{n-1} = \left[ \begin{array}{c|c} I_{k+1} & \tilde{P}_{k+2} \dots \tilde{P}_{n-1} \\ \hline & \end{array} \right]$

$P_k = P_{k+1} P_{k+2} \dots P_{n-1} = \left[ \begin{array}{c|c} I_k & \tilde{P}_{k+2} \dots \tilde{P}_{n-1} \\ \hline & \end{array} \right] \left[ \begin{array}{c|c} I_{k+1} & \tilde{P}_{k+2} \dots \tilde{P}_{n-1} \\ \hline & \end{array} \right]$

$= \left[ \begin{array}{c|c} I_k & \tilde{P}_{k+1} \\ \hline & \end{array} \right] \left[ \begin{array}{c|c} I_{k+1} & \tilde{P}_{k+2} \dots \tilde{P}_{n-1} \\ \hline & \end{array} \right]$

Something is wrong in this

2.  $\tilde{M}_k = P_k^T \left[ \begin{array}{c|c} I_n & \begin{bmatrix} 0 \\ \vdots \\ 0 \\ m_{k+1,k} \\ \vdots \\ m_{n,k} \end{bmatrix} e_k^T \end{array} \right] P_k$

$= I_n - P_k^T \begin{bmatrix} 0 \\ \vdots \\ 0 \\ m_{k+1,k} \\ \vdots \\ m_{n,k} \end{bmatrix} \left[ k^{th} \text{ row of } P_k \right] = e_k^T \text{ (see 1. in slide)}$

$= I_n - \tilde{P}_{n-1} \dots \tilde{P}_{k+1} \begin{bmatrix} 0 \\ \vdots \\ 0 \\ m_{k+1,k} \\ \vdots \\ m_{n,k} \end{bmatrix} e_k^T$

~~Thm:  $PA = LU$   
 $A = P^T LU$~~

PF for slide 29:-

1) If  $A$  is non-singular:-

We need to show that  $\exists$  a  $P$   $\exists$   $PA$  is a matrix whose principal sub-matrices are non-singular. ~~Then using~~

Let  $P$  be as defined by partial pivoting  $P_1, P_2, \dots, P_n$ . We know we can do partial pivoting unless at  $k^{th}$  stage,  $a_{kj}$  for  $k \leq j \leq n$  are all zero i.e. it is of the form

$$P_{k-1} \dots P_1 A = \begin{pmatrix} B_{11} & B_{12} \\ 0 & B_{22} \end{pmatrix} \quad B_{11} \text{ is a upper triangular matrix}$$

$$= |B_{11}| |B_{22}|$$

So if  $1^{st}$  column of  $B_{22} = 0$ , then  $|A| = 0$ .

$\therefore PA$  has a LU decomposition

2) If  $A$  is singular:-

then while doing LU decomposition, u will encounter the case where  $a_{kj} = 0$  for  $k \leq j \leq n$ .

But if u have not encountered that case  $\Rightarrow$  u will have done LU decomposition (even without partial pivoting (theoretically))  $\Rightarrow$  there was a LU decomposition where the diagonal entries of  $U$  are non-zero

$\Rightarrow |A| = |U| \neq 0 \Rightarrow A$  is non singular

$A$  is non-singular  $\Rightarrow$  u have not encountered that case - has been shown above

but  $U$  is singular  $|A| = |U|$  - so a diagonal entry = 0

Then, u simply skip for that  $k$  & go to next one (illy in the permutat<sup>n</sup> matrix)

GE-all lectures Pg 64/67

$$U = \begin{bmatrix} u_{11} & u_{12} & \dots & u_{1n} \\ & u_{22} & & \\ & & \ddots & \\ & & & u_{nn} \end{bmatrix}$$

$$A = LU \\ = LDV$$

$$D = \begin{bmatrix} d_{11} & & 0 \\ & d_{22} & \\ 0 & & \ddots \\ & & & d_{nn} \end{bmatrix}$$

$$V = \begin{bmatrix} v_{12} & \dots & v_{1n} \\ & \ddots & \\ & & v_{nn} \\ & & & 1 \end{bmatrix}$$

$U$  (a unit upper  $\Delta$ lar matrix) has unique decomp  $U = DV$   
 $\rightarrow V$  is unit upper  $\Delta$ lar  
 $V^{-1}$  is also unit upper  $\Delta$ lar

$$DV = \begin{bmatrix} d_{11} & d_{11}v_{12} & \dots & d_{11}v_{1n} \\ & d_{22} & & \\ & & \ddots & \\ & & & d_{nn} \end{bmatrix}$$

Clearly  $U = DV \Rightarrow$

$$d_{ii} = u_{ii} \\ d_{ij} = \frac{u_{ij}}{v_{ij}} \quad (i > j)$$

$\therefore U = DV$  is also unique

Suppose two decompositions exists.

$$A = L_1 D_1 V_1^{-1} = L_2 D_2 V_2^{-1}$$

Since  $A$  has unique  $LU$  decomposition,

$$\Rightarrow D_1^{-1} D_2 = V_1 V_2^{-1} \Rightarrow (V_2^{-1}) \text{ is also upper } \Delta \text{lar}$$

Lemma:-

$V_2$  is upper  $\Delta$ lar unit  $\Rightarrow V_2^{-1}$  is upper  $\Delta$ lar unit

$$\begin{bmatrix} v_{12} & \dots & v_{1n} \\ & \ddots & \\ & & v_{nn} \\ & & & 1 \end{bmatrix} \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & & & a_{2n} \\ \vdots & & \ddots & \vdots \\ a_{n1} & & & a_{nn} \end{bmatrix} = I$$

product of 2 unit upper  $\Delta$ lar also unit upper  $\Delta$ lar  $\Rightarrow V_1 V_2^{-1}$  also unit upper  $\Delta$ lar  $\Rightarrow D_1^{-1} D_2$  also unit upper  $\Delta$ lar  $\Rightarrow D_1^{-1} D_2 = I (D_1 = D_2) \Rightarrow V_1 = V_2$

$$\begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & & & a_{2n} \\ \vdots & & \ddots & \vdots \\ a_{n1} & & & a_{nn} \end{bmatrix} \begin{bmatrix} 1 & v_{12} & \dots & v_{1n} \\ & 1 & & \\ & & \ddots & \\ & & & 1 \end{bmatrix} = I$$

$$\begin{aligned} a_{11} &= 1 \\ a_{11} \cdot v_{12} + a_{12} &= 1 \\ a_{21} &= 0 \Rightarrow a_{21} = 0 \end{aligned}$$

Multiply  $i$ th row,  $j$ th col then  $(i+1)$ th row,  $j$ th col. (Finish all rows) and then other cols. This gives  $A = V_2^{-1}$  is upper  $\Delta$ lar.



2.  $A = LDL^T$

$A^T = \cancel{L} V^T D L^T = L^T D V$

Since  $LDL^T$  decomposition is unique,  $V = L^T$

3.  $D$  has positive diagonal entries  $\leftarrow x^T A x > 0 \quad \forall x$   
( $A = LDL^T$ )

$x^T A x > 0 \quad \forall x$

$x^T L D L^T x > 0$

Take  $x = L^T e_i \quad x = (L^T)^{-1} e_i$

$\therefore e_i^T D e_i > 0 \Rightarrow d_{ii} > 0$

$\Rightarrow x^T L D L^T x = y^T D y$  (where  $y = L^T x$ )

$[y_1 \ y_2 \ \dots \ y_n] \begin{bmatrix} d_1 & & 0 \\ & \ddots & \\ 0 & & d_n \end{bmatrix} \begin{bmatrix} y_1 \\ \vdots \\ y_n \end{bmatrix}$

$= [d_1 y_1 \ d_2 y_2 \ \dots \ d_n y_n] \begin{bmatrix} y_1 \\ \vdots \\ y_n \end{bmatrix} = d_1 y_1^2 + \dots > 0 \quad \forall x$   
 $\therefore d_i > 0$

#### 4. Cholesky

Slide 9

$\begin{bmatrix} x_1 & x_2 \end{bmatrix} \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} > 0$

$\begin{bmatrix} x_1 a_{11} + x_2 a_{21} & x_1 a_{12} + x_2 a_{22} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} > 0$

$a_{11} x_1^2 + 2 x_1 x_2 (a_{21} + a_{12}) + a_{22} x_2^2 > 0$

$\therefore A = \begin{bmatrix} 3 & 1 \\ 2 & 5 \end{bmatrix}$  there is no  $2 \times 2$  PD matrix

$\therefore A = \begin{bmatrix} 3 & -1 \\ +1 & 5 \end{bmatrix}$  is a PD matrix



$$x^* A x = (A^* x)^* x \xrightarrow{\text{since } x^* A x \in \mathbb{R}} (x^* A x)^* = x^* A^* x$$

$$\therefore x^* (A - A^*) x = 0 \quad \forall x \rightarrow 0$$

$$(x+y)^* (A - A^*) (x+y) = 0 \quad \leftarrow \text{Put } x+y \text{ in place of } x \text{ in } 0$$

$$(x+iy)^* (A - A^*) (x+iy) = 0 \quad \leftarrow \text{Put } x+iy \text{ in place of } x \text{ in } 0$$

Pf: Put  $x+ky$  in place of  $x$  in  $x^* B x$

$$(x+ky)^* B (x+ky) = x^* B x + \bar{k} y^* B x + x^* B k y + \bar{k} y^* B k y$$

Put  $k=1$  &  $k=i$

$$\Rightarrow y^* B x = 0 \quad \forall y \in \mathbb{C}$$

$$\Rightarrow B x = 0 \quad \forall x$$

Slide 5

If  $A$  is PD, its leading principal submatrices are PD.

Pf:-

$$A = \begin{bmatrix} \hat{A}_{11} & \hat{A}_{12} \\ \hat{A}_{21} & \hat{A}_{22} \end{bmatrix}$$

If  $x = [x_1, \dots, x_k, 0, 0, \dots]^T$

$$x^* A x = \bar{x}^* \hat{A}_{11} x \quad (\text{where } \bar{x} = [\alpha_1, \dots, \alpha_k])$$

6-Finite Precision systems

$$\begin{pmatrix} p & p & E_{\min} \\ 2 & 3 & -1 & 1 \end{pmatrix} E_{\max}$$

b)  $E = (1.01)_2 - 1$   $N_{\min} = (1.00)_2 + 2^{-1}$

$u = E/2$

$N_{\max} = (1.11)_2 + 2^{-1}$

c)  $x = \frac{1}{3} \quad y = \frac{1}{2} \quad z = \frac{1}{3}$   $f(\frac{1}{6}) = 0.11$   $\alpha = 0.10 \rightarrow \frac{1}{2} - \frac{5}{6} = -\frac{1}{6}$

$f(x)$

$$f(x+y+z) = f(f(x+y)+z) = f((0.11)_2 + \frac{1}{3}) = f(\frac{1}{2} + \frac{1}{4} + \frac{1}{6})$$

$$f(f(z+x)+y) = f((0.11)_2 + \frac{1}{2}) = (1.01)_2$$

$$f(\frac{2}{3}) = (0.11)_{\frac{1}{2}}$$

$$= (1.00)_2$$

Slide 10

Q1)  $S^\perp$  is always a subspace of  $\mathbb{F}^n$

PF:-  $x \in S^\perp, y \in S^\perp$  then  $\langle x, c \rangle = 0 \quad \langle y, c \rangle = 0 \quad \forall c \in S$

$$\langle ax+by, c \rangle = a\langle x, c \rangle + b\langle y, c \rangle = 0$$

$$\therefore ax+by \in S^\perp$$

2)  $S$

Ex:-

$$N(A)^\perp = S \quad \langle y, x \rangle = 0$$

Some properties:-

- 1) If  $x \in N(A^T A)$  then  $Ax$  is in both  $R(A)$  &  $N(A^T)$
- 2)  $N(A^T A) = N(A)$  ( $N(A) \subseteq N(A^T A) \rightarrow$  obv,  $x \in N(A^T A) \Rightarrow A^T A x = 0$   
 $\Downarrow$   
 $A x = 0 \Leftarrow x^T A^T A x = 0 \Rightarrow (Ax)^T Ax = 0$ )
- 3)  $A$  &  $A^T A$  have same rank

Use rank-Nullity Thm,

$$\text{rank}(A) = \dim(R(A)) \quad (\text{definit}^n)$$

$$\text{rank}(A) + \text{nullity}(A) = \dim(A) = n \quad (\text{rank-nullity thm})$$

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

From 2), as null spaces are same, nullities are same  
 $\text{rank}(A) = \text{rank}(A^T A)$

Ex:-  ~~$A^T A y = 0 \Rightarrow y = 0$~~

4) If  $A$  has l.i columns,  $A^T A$  is non-singular  
 As  $\text{rank}(A) = n$ ,  $\text{rank}(A^T A) = n$  & hence non-singular

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

PF:-  $Ax = 0 \Rightarrow x^T A^T = 0 \Rightarrow x$  is orthogonal to the columns of  $A^T$   
 $S^\perp = \text{span}(S^\perp) = R(A^T)^\perp \Leftarrow N(A) = S^\perp$  where  $S$  is set of rows of  $A$  (columns of  $A^T$ )

Thm: Let  $U$  &  $W$  be 2 subspaces of  $F^n \Rightarrow F^n = U+W$ . Then

$$\textcircled{1} \dim U + \dim W - \dim(U \cap W) = n$$

$$\text{span}(U+W) = U+W \rightarrow \textcircled{2}$$

pf:  $\textcircled{1}$  - Obv

Let  $B = \{u_1, \dots, u_r\}$  be a basis of  $U \cap W$   
Extend this to get

$$B_1 = \{u_1, \dots, u_r, v_1, \dots, v_s\} \rightarrow \text{basis of } U$$

$$B_2 = \{u_1, \dots, u_r, w_1, \dots, w_t\} \rightarrow \text{basis of } W$$

T.S.T  $B = \{u_1, \dots, u_r, v_1, \dots, v_s, w_1, \dots, w_t\}$  is a basis of  $U+W$

$$\sum_{i=1}^r a_i u_i + \sum_{j=1}^s b_j v_j + \sum_{k=1}^t c_k w_k = 0 \rightarrow \textcircled{4}$$

$$\sum_{i=1}^r a_i u_i + \sum_{j=1}^s b_j v_j = - \sum_{k=1}^t c_k w_k$$

$\underbrace{\hspace{10em}}_{\in U} \quad \underbrace{\hspace{10em}}_{\in W}$

$\therefore \text{LHS} = \text{RHS} \in U \cap W$

$$- \sum_{k=1}^t c_k w_k = \sum_{i=1}^r d_i u_i \quad (\because \text{LHS} \in U \cap W)$$

$$\sum_{k=1}^t c_k w_k + \sum_{i=1}^r d_i w_k = 0 \Rightarrow d_i = c_i = 0 \quad \textcircled{3}$$

( $\because B_2$  is l.i.)

$$(\text{Ifly } b_i = 0) \rightarrow \textcircled{2}$$

After  $\textcircled{2}$  &  $\textcircled{3}$  in  $\textcircled{4}$ , as  $B$  is basis,  $a_i = 0$

Obvs that  $\text{span}(B) = U_1 + U_2$ .

$$\therefore \dim(U+W) = r+s+t$$



Thm: Suppose  $U, W$  are subspaces of  $F^n \ni F^n = U + W$ . Then  
 $F^n = U \oplus W \iff u + w = 0$  for  $u \in U, w \in W$  implies  $u = w = 0$

PF:  ~~$u_1 + w_1 = u_2 + w_2$~~

$\rightarrow$  obv

$\leftarrow$  T.S.T

$$u_1 + w_1 = u_2 + w_2 \Rightarrow u_1 - u_2 = w_2 - w_1$$

$$\underbrace{(u_1 - u_2)}_{\in U} + \underbrace{(w_1 - w_2)}_{\in W} = 0$$

$$\therefore u_1 - u_2 = 0 \quad w_1 - w_2 = 0$$

$$(N(P))^{\perp} = R(P^T) \quad R(P) = N(P^T)^{\perp}$$

$$N(P) = N(P^T)$$

$$Px = 0$$

When  $P_{n \times n}^2 = P$

1)  $N(P) = R(I - P)$  As  $I - P$  is also a projxn,

$$N(I - P) = R(I - (I - P)) = R(P)$$

$$2) (I - P)^2 = I + P^2 - 2P = I - P$$

$$R(I - P) \oplus R(P)$$

$$x = (I - P)k_1$$

$$y = Pk_2$$

I have to Prove

$$N(P) + R(P) = F^n \text{ first}$$

Let  $x \in F^n$

$$x = (Px) + (x - Px)$$

$$Px \in R(P)$$

$$x - Px \in N(P) \quad (\because P(x - Px) = 0)$$

$$x \in N(P) \Rightarrow Px = 0$$

$$\text{T.S.T } \exists y \ni (I - P)y = x$$

$$\text{Let } y = (I - P)x$$

$$(I - P)y = (I + P^2 - 2P)x \quad (\because P^2 = P)$$

$$= x$$

$$x \in R(I - P) \Rightarrow (I - P)y = x$$

$$\text{T.S.T } Px = 0$$

$$\therefore Px = P(I - P)y = 0$$

T.I.T  $F^n = N(P) \oplus R(P)$  (given  $F^n = N(P) + R(P)$ )

IF  $u \in N(P), v \in R(P)$   
 $= R(I-P)$   $u+v=0$

$$u = (I-P)k_1$$

$$v = \cancel{(I-P)} Pk_2$$

$$\Rightarrow (I-P)k_1 + Pk_2 = 0 \rightarrow \textcircled{1}$$

$$P(I-P)k_1 + P^2 k_2 = 0$$

$$Pk_2 = 0$$

$$v = 0 \rightarrow \textcircled{2}$$

$$\textcircled{1} \times (I-P) \Rightarrow$$

$$(I-P)k_1 = 0$$

$$u = 0 \rightarrow \textcircled{3}$$

From  $\textcircled{2}$  &  $\textcircled{3}$   $\Rightarrow$  Thus in prev slide,  $\textcircled{2}$  &  $\textcircled{3}$  mean  $\oplus$  sum is direct sum

2)  $R(P) \subseteq U$  (obv)

T.I.T  $R(P) \supseteq U$

Let  $u \in U$

$$u = (u - Pu) + \cancel{Pu}$$

$$Pk = \cancel{(I-P)u} = (I-P)u$$

Let  $v \in V$ . As  $F^n = U \oplus V$ ,

$Pv = 0$  ( $\because$  there is only one way of writing  $v \in F^n$  in terms of  $U$  &  $V$   
 $v = 0 + v$ )

There is only one way of writing  $u \in F^n$  in terms of  $U$  &  $V$   
 i.e.  $u = u + 0$

$$\therefore Pu = u \Rightarrow u \in R(P)$$

CGS  $\equiv$  Condensed QR (QR decomposition 1 - slide 18/18)

$$\underbrace{[v_1 \dots v_m]}_V = \underbrace{[q_1 \dots q_m]}_Q \underbrace{\begin{bmatrix} r_{11} & r_{12} & \dots & r_{1m} \\ r_{22} & & & r_{2m} \\ & \ddots & & \vdots \\ & & r_{mm} & \end{bmatrix}}_R$$

$$= \begin{bmatrix} r_{11}q_1 & r_{12}q_1 + r_{22}q_2 & r_{13}q_1 + r_{23}q_2 + r_{33}q_3 & \dots \end{bmatrix}$$

$$v_j = \sum_{k=1}^n r_{kj} q_{0k}$$

Take dot product with  $q_i$  on both sides

$$v_j^T q_i = r_{1j} q_i^T q_i + 0 + 0 + \dots$$

$$= r_{ij}$$

For  $i=j$ ,  $r_{ii} q_i = v_i - \sum_{k=1}^{i-1} (r_{ki} q_k + r_{(i+1)i} q_{i+1} + \dots + r_{ni} q_n)$

Take 2 norm on both sides

$$r_{ii} = \|q_i\|_2$$

Thm 3.2.46:-  $A \in \mathbb{R}^{n \times n}$  be non-singular. Then  $\exists Q, R$  s.t.  $Q$  is orthogonal,  $R$  is upper  $\Delta$ lar with positive-main diagonal entries  $u$ ,  $A = QR$ .

PF:- Uniqueness:- (Pf)  $A = Q_1 R_1 = Q_2 R_2$

$$A^T A = R_1^T R_1 = R_2^T R_2$$

$\therefore R_1$  &  $R_2$  are Cholesky decomposition of  $A^T A$ .

PF 2):-  $B^T B = I$   $\therefore R_1 = R_2$

$$\begin{pmatrix} b_{11} & & & \\ 0 & b_{22} & & \\ & & \ddots & \\ 0 & & & b_{nn} \end{pmatrix}$$

$b_{11} = \pm 1$  (diagonal matrix whole entries are  $\pm 1$ )

$$b_{11} b_{12} + 0 = 0$$

$$b_{12} = 0 \text{ (illy } \neq b_{12})$$

~~$Q_2 = Q_1$~~   
Let  $Q_1 R_1 = Q_2 R_2$

$$Q_2 = Q_1 R_1 R_2^{-1}$$

$$Q_1^{-1} Q_2 = (R_1 R_2^{-1}) = K$$

$K$  is orthogonal as well as upper  $\Delta$ lar.  
 $\therefore K$  is a diagonal matrix with  $\pm 1$  entries  
 $K = D$

$$R_2 = D R_1, R_1 = D R_2$$

$$Q_2 = Q_1 D$$

$$Q_1 = Q_2 D^{-1} = Q_2 D$$



Ex: - 3.2.68)

$$Gx = y$$

$$(I - uv^T)x = y$$

$$x - y = uv^T x$$

$$= (v^T x)u$$

$\therefore u$  is a multiple of  $x - y$

3.2.69)

a) b)  $Gx = 0 \Rightarrow x$  is a multiple of  $u$   
 $(I - uv^T)x = 0$

$$x = (v^T x)u \quad x \text{ is a multiple of } u$$

$$x \text{ is a multiple of } u \Rightarrow Gx = 0$$

$$G(\lambda u) = \lambda(u - (v^T u)u)$$

Lemma:  $G$  is singular  $\Leftrightarrow v^T u = 1$

$G$  is singular  $\Rightarrow Ga = 0$  for  $a \neq 0$

$$(I - uv^T)a = 0 \rightarrow \textcircled{1}$$

$$v^T a - (v^T u)(v^T a) = 0$$

$$(v^T a)(1 - v^T u) = 0$$

$$v^T a = 0 \quad \text{or} \quad v^T u = 1$$

But if  $v^T a = 0$ , then  $\textcircled{1}$  becomes

$$a - 0 = 0 \Rightarrow a = 0$$

$$\therefore v^T u = 1$$

$$v^T u = 1 \Rightarrow G \text{ is singular}$$

$$Gu = u - (v^T u)u$$

$$Gu = u - (v^T u)u = 0 \quad \text{if } v^T u = 1$$

$$\text{As } v^T u = 1$$

$$G(\lambda u) = 0$$

$$(I - uv^T)G^{-1} = I$$

$$G^{-1}(I - uv^T) = I$$

$$G^{-1} = I + G^{-1}uv^T$$

$$G^{-1} = I + G^{-1}uv^T$$

$$G^{-1}u = I + G^{-1}u(v^T u)$$

$$G^{-1}u(1 - v^T u) = I$$

$$(I - uv^T)(I - \beta uv^T) = I$$

$$I - uv^T(\beta + 1) + \beta(uv^T)^2 = I$$

$$uv^T uv^T u$$

$$u - ku(\beta + 1) + \beta k^2 u = 0$$

$$1 - k(\beta + 1) + \beta k^2 = 0$$

c)  $G = I - uv^T$  is non-singular  $\Rightarrow$   
 $G^{-1} = I - \beta uv^T$

Suppose  $G^{-1} = I - \beta uv^T$

$$G^{-1}G = I$$

$$(I - \beta uv^T)(I - uv^T) = I$$

$$- \beta uv^T - uv^T + \beta uv^T uv^T = 0 \quad \text{Let } v^T u = k = u^T v$$

MOBS with  $u$  on right

$$- \beta k u - k u + \beta k^2 u = 0$$

$$- \beta - 1 + \beta k = 0$$

$$\beta(k-1) = 1$$

As we know  $G^{-1}$  ~~know~~ now,  $G^{-1}$  has the form  $I - \beta uv^T$

d)  $G^T = G^{-1}$

$$I - vu^T = I - \beta uv^T$$

$$vu^T = \beta uv^T \quad \leftarrow \text{MOBS with } u \text{ on the right}$$

$$(u^T u)v = \beta (v^T u)u \Rightarrow v = \beta u$$

Now  $u^T u = 1$

$$v = \left( \frac{v^T u}{v^T u - 1} \right) u = \left( \frac{\beta u^T u}{\beta u^T u - 1} \right) u = \left( \frac{\beta}{\beta - 1} \right) u = \beta u$$

$$\therefore \beta = 2$$

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3.4.35

$$\begin{aligned} [\tilde{V} \ v_k] &= [\tilde{Q} \ | \ v_k] \begin{bmatrix} \tilde{R} & | & s \\ \hline 0 & | & r_{kk} \end{bmatrix} \\ &= [\tilde{Q} \tilde{R} \quad \tilde{Q}s + v_k r_{kk}] \end{aligned}$$

$$v_k = \tilde{Q}s + v_k r_{kk}$$

$$v_k^T v_k = r_{kk}$$

~~$$v_k^T = s^T \tilde{Q}^T + r_{kk} v_k^T$$~~

~~$$v_k^T v_k = r_{kk}$$~~

~~$$v_k^T v_k = r_{kk}$$~~

$$\begin{aligned} \tilde{Q}^T \tilde{Q} &= I \\ (\text{not } \tilde{Q} \tilde{Q}^T &= I) \end{aligned}$$

$$v_k - v_k r_{kk} = \tilde{Q}s$$

$$\tilde{Q}^T v_k - \tilde{Q}^T v_k r_{kk} = s$$

3.4.36)

$$V = [v_1 \ \tilde{V}] \quad Q = [q_1 \ \tilde{Q}]$$

~~$$[v_1 \ \tilde{V}]$$~~ 
$$[v_1 \ | \ \tilde{V}] = [q_1 \ | \ \tilde{Q}] \begin{bmatrix} r_{11} & | & r^T \\ \hline 0 & | & \tilde{R} \end{bmatrix}$$

$$= [q_1 r_{11} \quad q_1 r^T + \tilde{Q} \tilde{R}]$$

$$v_1 = q_1 r_{11}$$

$$\tilde{V} = q_1 r^T + \tilde{Q} \tilde{R}$$

$$\boxed{q_1^T \tilde{V} = r^T}$$