

## Step-1

Let  $A = QR$

Given that the orthogonal matrices have norm  $\|Q\| = 1$ .

We have to show that  $\|A\| \leq \|R\|$  and  $\|R\| \leq \|A\|$ .

## Step-2

If  $Q$  is an orthogonal matrix, then by definition, we have  $Q^{-1} = Q^T$  (1)

Also, we have by the definition of norm that  $\|Q\|^2 = \max_{x \neq 0} \frac{\|Qx\|^2}{\|x\|^2}$  (2)

$$\begin{aligned} &= \max_{x \neq 0} \frac{x^T Q^T Q x}{x^T x} \\ &= \max_{x \neq 0} \frac{x^T I x}{x^T x} \quad (\text{Since by (1)}) \end{aligned}$$

$$= 1$$

While norm is non negative, we get  $\|Q\| = 1$

## Step-3

By definition of norm, we have  $\|A\| = \max_{x \neq 0} \frac{\|Ax\|}{\|x\|}$

In other words,  $\|A\| \leq \|A\| \|x\|$  for every matrix  $x$

Now,  $A$  is decomposed into  $QR$  where  $Q$  is an orthogonal matrix.

Then  $\|A\| \leq \|(QR)x\|$

$$\leq \|Q\| \|Rx\|$$

$$\leq \|Q\| \|R\| \|x\|$$

## Step-4

Continuation to the above

$$\Rightarrow \frac{\|(QR)x\|}{\|x\|} \leq \|Q\|\|R\|$$

$$\Rightarrow \max \frac{\|(QR)x\|}{\|x\|} \leq \|Q\|\|R\|$$

$$\Rightarrow \|QR\| \leq \|Q\|\|R\| \quad (\text{By (2)})$$

$$\Rightarrow \|A\| \leq \|R\| \quad \hat{A} \in \hat{A} \quad (3)$$

## Step-5

On the other hand, we have  $\|R\| \leq \|XR\|$  for every non singular matrix  $X$ .

Since  $Q$  is the orthogonal matrix non singular, we can write  $\|R\| \leq \|QR\|$

Consequently  $\|R\| \leq \|A\| \quad \hat{A} \in \hat{A} \quad (4)$

Putting (3) and (4) together, we get  $\|A\| = \|R\|$

## Step-6

We have to find an example of  $A = LU$  with  $\|A\| < \|L\|\|U\|$

$$\text{Let } A = \begin{bmatrix} 2 & 1 \\ 3 & 2 \end{bmatrix}$$

$$\text{Suppose } L = \begin{bmatrix} a_{11} & 0 \\ a_{21} & a_{22} \end{bmatrix}, U = \begin{bmatrix} 1 & u_{12} \\ 0 & 1 \end{bmatrix} \text{ such that } LU = A$$

Then it follows that  $a_{11} = 2, a_{11}u_{12} = 1, a_{21} = 3, a_{21}u_{12} + a_{22} = 2$

From this, we get  $u_{12} = \frac{1}{2}$  and  $a_{22} = \frac{1}{2}$

$$\text{So, } L = \begin{bmatrix} 2 & 0 \\ 3 & \frac{1}{2} \end{bmatrix}, U = \begin{bmatrix} 1 & \frac{1}{2} \\ 0 & 1 \end{bmatrix}$$

## Step-7

Now

$$A^T A = \begin{bmatrix} 2 & 1 \\ 3 & 2 \end{bmatrix} \begin{bmatrix} 2 & 3 \\ 1 & 2 \end{bmatrix} \\ = \begin{bmatrix} 5 & 8 \\ 8 & 13 \end{bmatrix}$$

The characteristic equation of  $A^T A$  is

$$\begin{aligned} |A^T A - \lambda I| &= 0 \\ \Rightarrow \begin{vmatrix} 5-\lambda & 8 \\ 8 & 13-\lambda \end{vmatrix} &= 0 \\ \Rightarrow \lambda^2 - 18\lambda + 1 &= 0 \\ \Rightarrow \lambda &= \frac{18 + \sqrt{324 - 4}}{2}, \frac{18 - \sqrt{324 - 4}}{2} \\ &= 17.94, 0.0557 \end{aligned}$$

So the eigenvalues of  $A^T A$  are 0.0557, 17.94

## Step-8

Now

$$\begin{aligned} \sqrt{\lambda_{\max}(A^T A)} &= \sqrt{17.94} \\ &= 4.24 \\ &= \|A\| \quad \text{â€œâ€œ (5)} \end{aligned}$$

## Step-9

$$L^T L = \begin{bmatrix} 2 & 3 \\ 0 & \frac{1}{2} \end{bmatrix} \begin{bmatrix} 2 & 0 \\ 3 & \frac{1}{2} \end{bmatrix}$$

On the other hand,

$$= \begin{bmatrix} 13 & \frac{3}{2} \\ \frac{3}{2} & \frac{1}{4} \end{bmatrix}$$

The characteristic equation of  $L^T L$  is

$$\begin{aligned}
& |L^T L - \lambda I| = 0 \\
& \Rightarrow \begin{vmatrix} 13 - \lambda & \frac{3}{2} \\ \frac{3}{2} & \frac{1}{4} - \lambda \end{vmatrix} = 0 \\
& \Rightarrow (13 - \lambda) \left( \frac{1}{4} - \lambda \right) - \frac{9}{4} = 0
\end{aligned}$$

## Step-10

Continuation to the above

$$\begin{aligned}
& \Rightarrow \lambda^2 - \frac{53}{4} \lambda + 1 = 0 \\
& \lambda = \frac{(53/4) + \sqrt{(53/4)^2 - 4}}{2}, \frac{(53/4) - \sqrt{(53/4)^2 - 4}}{2} \\
& = 13.174, 0.076
\end{aligned}$$

So

$$\begin{aligned}
\|L\| &= \sqrt{\lambda_{\max}(L^T L)} \\
&= \sqrt{13.174} \\
&\approx 3.627 \quad \text{â€œâ€œ} (6)
\end{aligned}$$

## Step-11

Now

$$\begin{aligned}
U^T U &= \begin{bmatrix} 1 & 0 \\ \frac{1}{2} & 1 \end{bmatrix} \begin{bmatrix} 1 & \frac{1}{2} \\ 0 & 1 \end{bmatrix} \\
&= \begin{bmatrix} 1 & 0.5 \\ 0.5 & 1 \end{bmatrix}
\end{aligned}$$

The characteristic equation of  $U^T U$  is

$$\begin{aligned}
& |U^T U - \lambda I| = 0 \\
& \Rightarrow \begin{vmatrix} 1 - \lambda & 0.5 \\ 0.5 & 1 - \lambda \end{vmatrix} = 0 \\
& \Rightarrow (1 - \lambda)(1 - \lambda) - 0.25 = 0
\end{aligned}$$

## Step-12

Continuation to the above

$$\Rightarrow \lambda^2 - 2\lambda + 0.75 = 0$$

$$\Rightarrow \lambda = \frac{2 + \sqrt{4-1}}{2}, \frac{2 - \sqrt{4-1}}{2}$$

$$= 1.866, 0.134$$

Therefore,

$$\begin{aligned}\|U\| &= \sqrt{\lambda_{\max}(U^T U)} \\ &= \sqrt{1.866} \\ &\approx 1.366\end{aligned}\quad \text{â€ˆâ€ˆâ€ˆ} (7)$$

## Step-13

Therefore,

$$\begin{aligned}\|A\| &= 4.24 < 4.95 \\ &< 3.627 \times 1.366 \\ &< \|L\| \|U\|\end{aligned}$$

Hence the required matrices are  $A = \begin{bmatrix} 2 & 1 \\ 3 & 2 \end{bmatrix}, L = \begin{bmatrix} 2 & 0 \\ 3 & \frac{1}{2} \end{bmatrix}$  and  $U = \begin{bmatrix} 1 & \frac{1}{2} \\ 0 & 1 \end{bmatrix}$ .