# Step-1

Let A = QR

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

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

# Step-2

If Q is an orthogonal matrix, then by definition, we have  $Q^{-1} = Q^{T}$   $\hat{a} \in \hat{a} \in \hat{a} \in \hat{a}$ 

 $\|Q\|^2 = \max \frac{\|Qx\|^2}{\|x\|^2}$  Also, we have by the definition of norm that

 $= \max \frac{x^T Q^T Q x}{x^T x}$ 

 $= \max \frac{x^T I x}{x^T x}$  (Since by (1))

= 1

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

## Step-3

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

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

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

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

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

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

## Step-4

Continuation to the above

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

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

$$\Rightarrow \|QR\| \le \|Q\| \|R\|$$

$$\Rightarrow \|A\| \le \|R\| \|\hat{a}\varepsilon|\hat{a}\varepsilon| (3)$$

#### Step-5

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

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

Consequently  $||R|| \le ||A|| \hat{a} \in |\hat{a} \in (4)$ 

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

#### Step-6

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

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

Suppose 
$$L = \begin{bmatrix} a_{11} & 0 \\ a_{21} & a_{22} \end{bmatrix}, U = \begin{bmatrix} 1 & u_{12} \\ 0 & 1 \end{bmatrix}$$
 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  $a_{22} = \frac{1}{2}$ 

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

#### 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

$$\begin{vmatrix} A^T A - \lambda I \end{vmatrix} = 0$$

$$\Rightarrow \begin{vmatrix} 5 - \lambda & 8 \\ 8 & 13 - \lambda \end{vmatrix} = 0$$

$$\Rightarrow \lambda^2 - 18\lambda + 1 = 0$$

$$\Rightarrow \lambda = \frac{18 + \sqrt{324 - 4}}{2}, \frac{18 - \sqrt{324 - 4}}{2}$$

$$= 17.94, 0.0557$$

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

# Step-8

Now

$$\sqrt{\lambda_{\max} \left( A^T A \right)} = \sqrt{17.94}$$

$$= 4.24$$

$$= ||A|| \quad \hat{a} \in |\hat{a} \in [1, \infty)$$

# 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  $L^TL$  is

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

# Step-10

Continuation to the above

$$\Rightarrow \lambda^2 - \frac{53}{4}\lambda + 1 = 0$$

$$\lambda = \frac{\left(53/4\right) + \sqrt{\left(53/4\right)^2 - 4}}{2}, \frac{\left(53/4\right) + \sqrt{\left(53/4\right)^2 - 4}}{2}$$

=13.174, 0.076

So

$$||L|| = \sqrt{\lambda_{\max} \left( L^T L \right)}$$

$$= \sqrt{13.174}$$

$$\approx 3.627 \qquad \hat{a} \in |\hat{a} \in |\hat{b}| (6)$$

#### Step-11

Now

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

The characteristic equation of  $U^TU$  is

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

## 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,

$$||U|| = \sqrt{\lambda_{\text{max}} \left( U^T U \right)}$$

$$= \sqrt{1.866}$$

$$\approx 1.366 \quad \hat{a} \in |\hat{a} \in |(7)|$$

## Step-13

Therefore,

$$||A|| = 4.24 < 4.95$$
  
 $< 3.627 \times 1.366$   
 $< ||L||||U||$ 

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

Hence the required matrices are