# Lecture 5: LU factorization, Cholesky factorization, Gaussian elimination with pivoting



School of Mathematical Sciences, Xiamen University

### 1. LU factorization

• Definition: Given  $\mathbf{A} \in \mathbb{C}^{m \times m}$ , an LU factorization (if it exists) of  $\mathbf{A}$  is a factorization

$$A = LU$$

where  $\mathbf{L} \in \mathbb{C}^{m \times m}$  is unit lower-triangular and  $\mathbf{U} \in \mathbb{C}^{m \times m}$  is upper-triangular.

ullet An approach: find a sequence of unit lower-triangular matrices  ${f L}_k$  such that

$$\mathbf{L}_{m-1}\cdots\mathbf{L}_2\mathbf{L}_1\mathbf{A}=\mathbf{U}$$

and set

$$\mathbf{L} = \mathbf{L}_1^{-1} \mathbf{L}_2^{-1} \cdots \mathbf{L}_{m-1}^{-1}.$$

• A  $4 \times 4$  example

$$\mathbf{A} = \begin{bmatrix} 2 & 1 & 1 & 0 \\ 4 & 3 & 3 & 1 \\ 8 & 7 & 9 & 5 \\ 6 & 7 & 9 & 8 \end{bmatrix}$$

$$\mathbf{L}_{1}\mathbf{A} = \begin{bmatrix} 1 & & & \\ -2 & 1 & & \\ -4 & & 1 & \\ -3 & & & 1 \end{bmatrix} \begin{bmatrix} 2 & 1 & 1 & 0 \\ 4 & 3 & 3 & 1 \\ 8 & 7 & 9 & 5 \\ 6 & 7 & 9 & 8 \end{bmatrix} = \begin{bmatrix} 2 & 1 & 1 & 0 \\ 1 & 1 & 1 \\ 3 & 5 & 5 \\ 4 & 6 & 8 \end{bmatrix}$$

$$\mathbf{L}_{2}\mathbf{L}_{1}\mathbf{A} = \begin{bmatrix} 1 & & & \\ & 1 & & \\ & -3 & 1 & \\ & -4 & & 1 \end{bmatrix} \begin{bmatrix} 2 & 1 & 1 & 0 \\ & 1 & 1 & 1 \\ & 3 & 5 & 5 \\ & 4 & 6 & 8 \end{bmatrix} = \begin{bmatrix} 2 & 1 & 1 & 0 \\ & 1 & 1 & 1 \\ & & 2 & 2 \\ & & 2 & 4 \end{bmatrix}$$

$$\mathbf{L}_{3}\mathbf{L}_{2}\mathbf{L}_{1}\mathbf{A} = \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & -1 & 1 \end{bmatrix} \begin{bmatrix} 2 & 1 & 1 & 0 \\ & 1 & 1 & 1 \\ & & 2 & 2 \\ & & 2 & 4 \end{bmatrix} = \begin{bmatrix} 2 & 1 & 1 & 0 \\ & 1 & 1 & 1 \\ & & 2 & 2 \\ & & & 2 \end{bmatrix} = \mathbf{U}.$$

$$\begin{bmatrix} 1 & & & \\ -2 & 1 & & \\ -4 & & 1 & \\ -3 & & & 1 \end{bmatrix}^{-1} = \begin{bmatrix} 1 & & & \\ 2 & 1 & & \\ 4 & & 1 & \\ 3 & & & 1 \end{bmatrix}$$

$$\begin{bmatrix} 2 & 1 & 1 & 0 \\ 4 & 3 & 3 & 1 \\ 8 & 7 & 9 & 5 \\ 6 & 7 & 9 & 8 \end{bmatrix} = \begin{bmatrix} 1 \\ 2 & 1 \\ 4 & 3 & 1 \\ 3 & 4 & 1 & 1 \end{bmatrix} \begin{bmatrix} 2 & 1 & 1 & 0 \\ 1 & 1 & 1 & 1 \\ 2 & 2 & 2 & 2 \end{bmatrix}$$

### 1.1. General formulas for LU factorization

- Let  $\mathbf{x}_k$  denote the kth column of the matrix at the beginning of step k (which matrix?  $\mathbf{L}_{k-1} \cdots \mathbf{L}_2 \mathbf{L}_1 \mathbf{A}$ ).
- The purpose is to eliminate the entries below  $x_{kk}$ . To do this we construct the matrix  $\mathbf{L}_k$ :

where the *multiplier* 

$$\ell_{jk} = \frac{x_{jk}}{x_{kk}}, \quad k+1 \le j \le m.$$

# Proposition 1

The matrix  $\mathbf{L}_k$  can be inverted by negating its subdiagonal entries. We have

$$\mathbf{L}_k^{-1} = egin{bmatrix} 1 & & & & & & \\ & & 1 & & & & \\ & & \ell_{k+1,k} & 1 & & \\ & & \vdots & & \ddots & \\ & & \ell_{mk} & & & 1 \end{bmatrix} = egin{bmatrix} \mathbf{I}_{k-1} & \mathbf{0} & \mathbf{0} & \\ \mathbf{0} & 1 & \mathbf{0} & \\ \mathbf{0} & - \bigstar & \mathbf{I}_{m-k} \end{bmatrix}.$$

Proof. Define the vector

$$\boldsymbol{\ell}_k = \begin{bmatrix} 0 & \cdots & 0 & \ell_{k+1,k} & \cdots & \ell_{mk} \end{bmatrix}^\top$$
.

The matrix  $\mathbf{L}_k = \mathbf{I} - \boldsymbol{\ell}_k \mathbf{e}_k^*$ , where  $\mathbf{e}_k$  is the kth column of the identity matrix  $\mathbf{I}$ . Obviously,  $\mathbf{e}_k^* \boldsymbol{\ell}_k = 0$ . Therefore, the statement follows from

$$(\mathbf{I} - \boldsymbol{\ell}_k \mathbf{e}_k^*)(\mathbf{I} + \boldsymbol{\ell}_k \mathbf{e}_k^*) = \mathbf{I} - \boldsymbol{\ell}_k \mathbf{e}_k^* \boldsymbol{\ell}_k \mathbf{e}_k^* = \mathbf{I}.$$

# Proposition 2

The product  $\mathbf{L}_1^{-1}\mathbf{L}_2^{-1}\cdots\mathbf{L}_{m-1}^{-1}$ , i.e., the L factor **L**, can be formed by collecting the entries  $\ell_{jk}$  in the appropriate places. We have

$$\mathbf{L} = \begin{bmatrix} 1 & & & & \\ \ell_{21} & 1 & & & \\ \ell_{31} & \ell_{32} & 1 & & \\ \vdots & \vdots & \ddots & \ddots & \\ \ell_{m1} & \ell_{m2} & \cdots & \ell_{m,m-1} & 1 \end{bmatrix}.$$

Proof. It follows from  $\mathbf{L}_k^{-1} = \mathbf{I} + \boldsymbol{\ell}_k \mathbf{e}_k^*$  and  $\mathbf{e}_k^* \boldsymbol{\ell}_j = 0 \ (\forall j \geq k)$  that

$$\mathbf{L}_k^{-1}\mathbf{L}_{k+1}^{-1} = \mathbf{I} + \boldsymbol{\ell}_k\mathbf{e}_k^* + \boldsymbol{\ell}_{k+1}\mathbf{e}_{k+1}^*.$$

Therefore,

$$\mathbf{L} = \mathbf{L}_1^{-1} \mathbf{L}_2^{-1} \cdots \mathbf{L}_{m-1}^{-1} = \mathbf{I} + \ell_1 \mathbf{e}_1^* + \ell_2 \mathbf{e}_2^* + \cdots + \ell_{m-1} \mathbf{e}_{m-1}^*.$$

### Remark 3

- The matrices  $\mathbf{L}_k^{-1}$  are never formed and multiplied explicitly.
- The multipliers  $\ell_{jk}$  are computed and stored directly into **L**.

# 1.2. LU factorization algorithm

Algorithm: LU factorization 
$$\mathbf{A} = \mathbf{L}\mathbf{U}$$

$$\mathbf{U} = \mathbf{A}, \quad \mathbf{L} = \mathbf{I}$$
for  $k = 1$  to  $m - 1$ 
for  $j = k + 1$  to  $m$ 

$$\ell_{jk} = u_{jk}/u_{kk}$$

$$u_{j,k:m} = u_{j,k:m} - \ell_{jk}u_{k,k:m}$$
end
end

- 1.3. Gaussian elimination for Ax = b
  - $\bullet \mathbf{A} = \mathbf{L}\mathbf{U}, \quad \mathbf{L}\mathbf{y} = \mathbf{b}, \quad \mathbf{U}\mathbf{x} = \mathbf{y}$

# **Algorithm**: Forward elimination solving Ly = b

for k = 1 to m

$$y_k = \left(b_k - \sum_{j=1}^{k-1} \ell_{kj} y_j\right) / \ell_{kk}$$

end

# **Algorithm**: Back substitution solving $\mathbf{U}\mathbf{x} = \mathbf{y}$

for k = m downto 1

$$x_k = \left(y_k - \sum_{j=k+1}^m u_{kj} x_j\right) / u_{kk}$$

end

# 2. Cholesky factorization

• Every Hermitian positive definite matrix **A** has a factorization

$$A = LDL^*$$

where **L** is the unit lower-triangular matrix in its LU factorization  $\mathbf{A} = \mathbf{L}\mathbf{U}$  and **D** is a diagonal matrix with diagonal entries  $d_{ii} > 0$ .

• Definition: Given  $\mathbf{A} \in \mathbb{C}^{m \times m}$ , a Cholesky factorization (if it exists) of  $\mathbf{A}$  is a factorization

$$\mathbf{A} = \mathbf{R}^* \mathbf{R}$$

where  $\mathbf{R} \in \mathbb{C}^{m \times m}$  is upper-triangular.

# Theorem 4

Every Hermitian positive definite matrix  $\mathbf{A} \in \mathbb{C}^{m \times m}$  has a unique Cholesky factorization

$$\mathbf{A} = \mathbf{R}^* \mathbf{R}.$$

where  $\mathbf{R} \in \mathbb{C}^{m \times m}$  is upper-triangular and  $r_{ij} > 0$ .

# Proof. (By induction on the dimension).

It is easy for the case of dimension 1. Assume it is true for the case of dimension m-1. We prove the case of dimension m. Let  $\alpha = \sqrt{a_{11}}$ . We have

$$\mathbf{A} = \begin{bmatrix} a_{11} & \mathbf{w}^* \\ \mathbf{w} & \mathbf{K} \end{bmatrix} = \begin{bmatrix} \alpha & \mathbf{0} \\ \mathbf{w}/\alpha & \mathbf{I} \end{bmatrix} \begin{bmatrix} 1 & \mathbf{0} \\ \mathbf{0} & \mathbf{K} - \mathbf{w}\mathbf{w}^*/a_{11} \end{bmatrix} \begin{bmatrix} \alpha & \mathbf{w}^*/\alpha \\ \mathbf{0} & \mathbf{I} \end{bmatrix}$$

$$= \begin{bmatrix} \alpha & \mathbf{0} \\ \mathbf{w}/\alpha & \mathbf{I} \end{bmatrix} \begin{bmatrix} 1 & \mathbf{0} \\ \mathbf{0} & \widehat{\mathbf{R}}^*\widehat{\mathbf{R}} \end{bmatrix} \begin{bmatrix} \alpha & \mathbf{w}^*/\alpha \\ \mathbf{0} & \mathbf{I} \end{bmatrix}$$

$$(\text{by } \mathbf{K} - \mathbf{w}\mathbf{w}^*/a_{11} \text{ is HPD and the induction hypothesis})$$

$$= \begin{bmatrix} \alpha & \mathbf{0} \\ \mathbf{w}/\alpha & \widehat{\mathbf{R}}^* \end{bmatrix} \begin{bmatrix} \alpha & \mathbf{w}^*/\alpha \\ \mathbf{0} & \widehat{\mathbf{R}} \end{bmatrix} = \mathbf{R}^*\mathbf{R}.$$

The first row of **R** is uniquely determined by  $r_{11} > 0$  and the factorization itself. The uniqueness of **R** follows from the induction hypothesis that  $\hat{\mathbf{R}}$  is unique.

# **2.1.** A $4 \times 4$ example

$$\mathbf{A} = \begin{bmatrix} 4 & 4\mathrm{i} & 6 & 2 \\ -4\mathrm{i} & 5 & -4\mathrm{i} & 5 - 2\mathrm{i} \\ 6 & 4\mathrm{i} & 17 & 3 - 8\mathrm{i} \\ 2 & 5 + 2\mathrm{i} & 3 + 8\mathrm{i} & 36 \end{bmatrix}, \mathbf{w} = \begin{bmatrix} -4\mathrm{i} \\ 6 \\ 2 \end{bmatrix}, \mathbf{K} = \begin{bmatrix} 5 & -4\mathrm{i} & 5 - 2\mathrm{i} \\ 4\mathrm{i} & 17 & 3 - 8\mathrm{i} \\ 5 + 2\mathrm{i} & 3 + 8\mathrm{i} & 36 \end{bmatrix}$$

• Compute the upper triangular matrix **R** row by row

Row 1: 
$$\begin{bmatrix} 2 & & & \\ -2i & 1 & & \\ 3 & & 1 \\ 1 & & & 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ & 1 & 2i & 5 \\ & -2i & 8 & -8i \\ & 5 & 8i & 35 \end{bmatrix} \begin{bmatrix} 2 & 2i & 3 & 1 \\ & 1 & & \\ & & 1 & \\ & & & 1 \end{bmatrix}$$

Row 2: 
$$\begin{bmatrix} 1 & 2i & 5 \\ -2i & 8 & -8i \\ 5 & 8i & 35 \end{bmatrix} = \begin{bmatrix} 1 \\ -2i & 1 \\ 5 & & 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ & 4 & 2i \\ & -2i & 10 \end{bmatrix} \begin{bmatrix} 1 & 2i & 5 \\ & 1 & \\ & & 1 \end{bmatrix}$$

$$\text{Row 3: } \begin{bmatrix} 4 & 2\mathbf{i} \\ -2\mathbf{i} & 10 \end{bmatrix} = \begin{bmatrix} 2 \\ -1\mathbf{i} & 1 \end{bmatrix} \begin{bmatrix} 1 \\ & 9 \end{bmatrix} \begin{bmatrix} 2 & 1\mathbf{i} \\ & 1 \end{bmatrix}$$

Row 4:  $9 = 3 \times 1 \times 3$ 

• The Cholesky factor 
$$\mathbf{R} = \begin{bmatrix} 2 & 2i & 3 & 1 \\ & 1 & 2i & 5 \\ & & 2 & 1i \\ & & & 3 \end{bmatrix}$$
.

# 2.2. Algorithm for Cholesky factorization

# Algorithm: Cholesky factorization R=triu(A) for k = 1 to mfor j = k + 1 to m $r_{j,j:m} = r_{j,j:m} - r_{k,j:m} \overline{r}_{kj} / r_{kk}$ end

 $r_{k,k:m} = r_{k,k:m} / \sqrt{r_{kk}}$ 

• Exercise: Design an algorithm to compute  $\mathbb{R}^*$  column by column.

### 2.3. Other factorization of HPD matrix

• For any HPD matrix **A**, there exists a unique HPD matrix **B** satisfying

$$\mathbf{A} = \mathbf{B}^2$$
.

**B** is called the *square root* of **A**. (Proof? HPSD case?)

end

# 3. Gaussian elimination with partial pivoting (GEPP)

• Only rows are interchanged (partial pivoting = row pivoting).

• After m-1 steps, **A** becomes an upper-triangular matrix **U**:

$$\mathbf{L}_{m-1}\mathbf{P}_{m-1}\cdots\mathbf{L}_{2}\mathbf{P}_{2}\mathbf{L}_{1}\mathbf{P}_{1}\mathbf{A}=\mathbf{U},$$

where  $\mathbf{P}_k$  is an elementary permutation matrix  $(\mathbf{P}_k = \mathbf{P}_k^{\top} = \mathbf{P}_k^{-1})$ .

# Remark 5

Absolute values of all the entries of  $\mathbf{L}_k$  in GEPP are  $\leq 1$  due to the property at step k (after pivoting)

$$|x_{kk}| = \max_{k \le j \le m} |x_{jk}|.$$

# 3.1. A $4 \times 4$ Example

• 1: Interchange the first and third rows by  $P_1$ 

$$\begin{bmatrix} & & 1 \\ & 1 & & \\ 1 & & & \\ & & & 1 \end{bmatrix} \begin{bmatrix} 2 & 1 & 1 & 0 \\ 4 & 3 & 3 & 1 \\ 8 & 7 & 9 & 5 \\ 6 & 7 & 9 & 8 \end{bmatrix} = \begin{bmatrix} 8 & 7 & 9 & 5 \\ 4 & 3 & 3 & 1 \\ 2 & 1 & 1 & 0 \\ 6 & 7 & 9 & 8 \end{bmatrix}$$

• 2: First elimination by  $L_1$ 

$$\begin{bmatrix} 1 & & & \\ -\frac{1}{2} & 1 & & \\ -\frac{1}{4} & & 1 & \\ -\frac{3}{4} & & & 1 \end{bmatrix} \begin{bmatrix} 8 & 7 & 9 & 5 \\ 4 & 3 & 3 & 1 \\ 2 & 1 & 1 & 0 \\ 6 & 7 & 9 & 8 \end{bmatrix} = \begin{bmatrix} 8 & 7 & 9 & 5 \\ -\frac{1}{2} & -\frac{3}{2} & -\frac{3}{2} \\ -\frac{3}{4} & -\frac{5}{4} & -\frac{5}{4} \\ \frac{7}{4} & \frac{9}{4} & \frac{17}{4} \end{bmatrix}$$

• 3: Interchange the second and fourth rows by  $P_2$ 

$$\begin{bmatrix} 1 & & & & \\ & & & 1 \\ & & & 1 \\ & & 1 & \end{bmatrix} \begin{bmatrix} 8 & 7 & 9 & 5 \\ -\frac{1}{2} & -\frac{3}{2} & -\frac{3}{2} \\ -\frac{3}{4} & -\frac{5}{4} & -\frac{5}{4} \\ & \frac{7}{4} & \frac{9}{4} & \frac{17}{4} \end{bmatrix} = \begin{bmatrix} 8 & 7 & 9 & 5 \\ \frac{7}{4} & \frac{9}{4} & \frac{17}{4} \\ -\frac{3}{4} & -\frac{5}{4} & -\frac{5}{4} \\ -\frac{1}{2} & -\frac{3}{2} & -\frac{3}{2} \end{bmatrix}$$

• 4: Second elimination by  $L_2$ 

$$\begin{bmatrix} 1 & & & \\ & 1 & & \\ & \frac{3}{7} & 1 & \\ & \frac{2}{7} & & 1 \end{bmatrix} \begin{bmatrix} 8 & 7 & 9 & 5 \\ & \frac{7}{4} & \frac{9}{4} & \frac{17}{4} \\ & -\frac{3}{4} & -\frac{5}{4} & -\frac{5}{4} \\ & -\frac{1}{2} & -\frac{3}{2} & -\frac{3}{2} \end{bmatrix} = \begin{bmatrix} 8 & 7 & 9 & 5 \\ & \frac{7}{4} & \frac{9}{4} & \frac{17}{4} \\ & & -\frac{2}{7} & \frac{4}{7} \\ & & -\frac{6}{7} & -\frac{2}{7} \end{bmatrix}$$

• 5: Interchange the third and fourth rows by  $P_3$ 

$$\begin{bmatrix} 1 & & & \\ & 1 & & \\ & & & 1 \\ & & & 1 \end{bmatrix} \begin{bmatrix} 8 & 7 & 9 & 5 \\ & \frac{7}{4} & \frac{9}{4} & \frac{17}{4} \\ & & -\frac{2}{7} & \frac{4}{7} \\ & & -\frac{6}{7} & -\frac{2}{7} \end{bmatrix} = \begin{bmatrix} 8 & 7 & 9 & 5 \\ & \frac{7}{4} & \frac{9}{4} & \frac{17}{4} \\ & & -\frac{6}{7} & -\frac{2}{7} \\ & & -\frac{2}{7} & \frac{4}{7} \end{bmatrix}$$

• 6: Final elimination by  $L_3$ 

$$\begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & -\frac{1}{3} & 1 \end{bmatrix} \begin{bmatrix} 8 & 7 & 9 & 5 \\ & \frac{7}{4} & \frac{9}{4} & \frac{17}{4} \\ & & -\frac{6}{7} & -\frac{2}{7} \\ & & & -\frac{2}{7} & \frac{4}{7} \end{bmatrix} = \begin{bmatrix} 8 & 7 & 9 & 5 \\ & \frac{7}{4} & \frac{9}{4} & \frac{17}{4} \\ & & -\frac{6}{7} & -\frac{2}{7} \\ & & & \frac{2}{3} \end{bmatrix}$$

 $\bullet \ \mathbf{A} = \mathbf{P}_1^{-1} \mathbf{L}_1^{-1} \mathbf{P}_2^{-1} \mathbf{L}_2^{-1} \mathbf{P}_3^{-1} \mathbf{L}_3^{-1} \mathbf{U}$ 

$$\begin{bmatrix} 2 & 1 & 1 & 0 \\ 4 & 3 & 3 & 1 \\ 8 & 7 & 9 & 5 \\ 6 & 7 & 9 & 8 \end{bmatrix} = \begin{bmatrix} \frac{1}{4} & -\frac{3}{7} & \frac{1}{3} & 1 \\ \frac{1}{2} & -\frac{2}{7} & 1 \\ 1 & & & \\ \frac{3}{4} & 1 & \end{bmatrix} \begin{bmatrix} 8 & 7 & 9 & 5 \\ & \frac{7}{4} & \frac{9}{4} & \frac{17}{4} \\ & & -\frac{6}{7} & -\frac{2}{7} \\ & & & \frac{2}{3} \end{bmatrix}$$

 $\bullet \ \mathbf{PA} = \mathbf{LU} \ \mathrm{with} \ \mathbf{P} = \mathbf{P}_3 \mathbf{P}_2 \mathbf{P}_1 \ \mathrm{and} \ \mathbf{L} = \mathbf{P}_3 \mathbf{P}_2 \mathbf{L}_1^{-1} \mathbf{P}_2^{-1} \mathbf{L}_2^{-1} \mathbf{P}_3^{-1} \mathbf{L}_3^{-1}$ 

$$\begin{bmatrix} & 1 \\ & & 1 \\ 1 \\ 1 & & \end{bmatrix} \begin{bmatrix} 2 & 1 & 1 & 0 \\ 4 & 3 & 3 & 1 \\ 8 & 7 & 9 & 5 \\ 6 & 7 & 9 & 8 \end{bmatrix} = \begin{bmatrix} 1 \\ \frac{3}{4} & 1 \\ \frac{1}{2} & -\frac{2}{7} & 1 \\ \frac{1}{4} & -\frac{3}{7} & \frac{1}{3} & 1 \end{bmatrix} \begin{bmatrix} 8 & 7 & 9 & 5 \\ \frac{7}{4} & \frac{9}{4} & \frac{17}{4} \\ & -\frac{6}{7} & -\frac{2}{7} \\ & & & \frac{2}{3} \end{bmatrix}$$

$$\mathbf{P} \qquad \mathbf{A} \qquad \mathbf{L} \qquad \mathbf{U}$$

### 3.2. General formulas for PA = LU

• The matrix  $\mathbf{L}_{m-1}\mathbf{P}_{m-1}\cdots\mathbf{L}_{2}\mathbf{P}_{2}\mathbf{L}_{1}\mathbf{P}_{1}$  can be rewritten in the form

$$\mathbf{L}_{m-1}\mathbf{P}_{m-1}\cdots\mathbf{L}_{2}\mathbf{P}_{2}\mathbf{L}_{1}\mathbf{P}_{1}=\widehat{\mathbf{L}}_{m-1}\cdots\widehat{\mathbf{L}}_{2}\widehat{\mathbf{L}}_{1}\mathbf{P}_{m-1}\cdots\mathbf{P}_{2}\mathbf{P}_{1},$$

where 
$$\hat{\mathbf{L}}_k = \mathbf{P}_{m-1} \cdots \mathbf{P}_{k+2} \mathbf{P}_{k+1} \mathbf{L}_k \mathbf{P}_{k+1}^{-1} \mathbf{P}_{k+2}^{-1} \cdots \mathbf{P}_{m-1}^{-1}$$
.

### Remark 6

The elementary permutation matrix  $\mathbf{P}_k$  in GEPP has the form

$$\mathbf{P}_k = egin{bmatrix} \mathbf{I}_{k-1} & \mathbf{0} \\ \mathbf{0} & \widehat{\mathbf{P}}_k \end{bmatrix},$$

where  $\widehat{\mathbf{P}}_k \in \mathbb{R}^{(m-k+1)\times(m-k+1)}$  is an elementary permutation matrix.

## Remark 7

The unit lower triangular matrix  $\hat{\mathbf{L}}_k$  in GEPP has the same sparsity pattern as that of  $\mathbf{L}_k$ . The sparsity pattern is

$$egin{bmatrix} \mathbf{I}_{k-1} & \mathbf{0} & \mathbf{0} \ \mathbf{0} & 1 & \mathbf{0} \ \mathbf{0} & \bigstar & \mathbf{I}_{m-k} \end{bmatrix} = egin{bmatrix} \mathbf{0} & \mathbf{0} & \mathbf{0} \ \mathbf{0} & \mathbf{0} & \mathbf{0} \ \mathbf{0} & \bigstar & \mathbf{0} \end{bmatrix} + \mathbf{I}.$$

The matrix  $\hat{\mathbf{L}}_k$  is equal to  $\mathbf{L}_k$  but with the  $\bigstar$ 's entries permuted.

### Remark 8

By Proposition 1,  $\widehat{\mathbf{L}}_k^{-1}$  has the same sparsity pattern as that of  $\widehat{\mathbf{L}}_k$ . Thus, the product  $\widehat{\mathbf{L}}_1^{-1}\widehat{\mathbf{L}}_2^{-1}\cdots\widehat{\mathbf{L}}_{m-1}^{-1}$  is unit lower triangular.

# Remark 9

GEPP has the LU factorization PA = LU where

$$\mathbf{P} = \mathbf{P}_{m-1} \cdots \mathbf{P}_2 \mathbf{P}_1, \quad \mathbf{U} = \mathbf{L}_{m-1} \mathbf{P}_{m-1} \cdots \mathbf{L}_2 \mathbf{P}_2 \mathbf{L}_1 \mathbf{P}_1 \mathbf{A},$$

$$\mathbf{L} = \widehat{\mathbf{L}}_{1}^{-1} \widehat{\mathbf{L}}_{2}^{-1} \cdots \widehat{\mathbf{L}}_{m-1}^{-1} = \mathbf{P}_{m-1} \cdots \mathbf{P}_{3} \mathbf{P}_{2} \mathbf{L}_{1}^{-1} \mathbf{P}_{2}^{-1} \mathbf{L}_{2}^{-1} \mathbf{P}_{3}^{-1} \cdots \mathbf{P}_{m-1}^{-1} \mathbf{L}_{m-1}^{-1}.$$

# Remark 10

The matrices  $\widehat{\mathbf{L}}_k^{-1}$  are never formed and multiplied explicitly. The multipliers  $\ell_{jk}$  are computed and stored in the appropriate places.

# Remark 11

The permutation matrix **P** is not known ahead of time.

### 3.3. GEPP for Ax = b

 $\bullet \mathbf{PA} = \mathbf{LU}, \quad \mathbf{Ly} = \mathbf{Pb}, \quad \mathbf{Ux} = \mathbf{y}$ 

Algorithm: LU factorization 
$$\mathbf{PA} = \mathbf{LU}$$
 in GEPP  $\mathbf{U} = \mathbf{A}, \ \mathbf{L} = \mathbf{I}, \ \mathbf{P} = \mathbf{I}$  for  $k = 1$  to  $m - 1$  Select  $i \geq k$  to maximize  $|u_{ik}|$   $u_{k,k:m} \leftrightarrow u_{i,k:m}$  (interchange two rows)  $\ell_{k,1:k-1} \leftrightarrow \ell_{i,1:k-1}$   $p_{k,:} \leftrightarrow p_{i,:}$  for  $j = k + 1$  to  $m$   $\ell_{jk} = u_{jk}/u_{kk}$   $u_{j,k:m} = u_{j,k:m} - \ell_{jk}u_{k,k:m}$  end end

### 3.4. Growth factor

• Define the growth factor for **A** as the ratio  $\rho = \frac{\max_{ij} |u_{ij}|}{\max_{ij} |a_{ij}|}$ .

# Proposition 12

The growth factor  $\rho$  of Gaussian elimination with partial pivoting applied to any matrix  $\mathbf{A} \in \mathbb{C}^{m \times m}$  satisfies  $\rho \leq 2^{m-1}$ .

Proof. Exercise 22.1.

• Worst case of  $\rho$ : Consider the  $5 \times 5$  matrix **A**:

The L and U factors are given by

and

$$\mathbf{U} = \begin{bmatrix} 1 & & & 1 \\ & 1 & & & 2 \\ & & 1 & & 4 \\ & & & 1 & 8 \\ & & & & 16 \end{bmatrix}.$$

The growth factor  $\rho = 2^{m-1} = 16$ .

# 4. Gaussian elimination with complete pivoting (GECP)

- Both rows and columns are interchanged
- After m-1 steps, **A** becomes an upper-triangular matrix **U**:

$$\mathbf{L}_{m-1}\mathbf{P}_{m-1}\cdots\mathbf{L}_{2}\mathbf{P}_{2}\mathbf{L}_{1}\mathbf{P}_{1}\mathbf{A}\mathbf{Q}_{1}\mathbf{Q}_{2}\cdots\mathbf{Q}_{m-1}=\mathbf{U}.$$

### Remark 13

 $\operatorname{GE}$  with complete pivoting has the LU factorization

$$PAQ = LU$$
,

where 
$$\mathbf{P} = \mathbf{P}_{m-1} \cdots \mathbf{P}_2 \mathbf{P}_1$$
,  $\mathbf{Q} = \mathbf{Q}_1 \mathbf{Q}_2 \cdots \mathbf{Q}_{m-1}$ , and

$$\mathbf{L} = \widehat{\mathbf{L}}_1^{-1} \widehat{\mathbf{L}}_2^{-1} \cdots \widehat{\mathbf{L}}_{m-1}^{-1} = \mathbf{P}_{m-1} \cdots \mathbf{P}_3 \mathbf{P}_2 \mathbf{L}_1^{-1} \mathbf{P}_2^{-1} \mathbf{L}_2^{-1} \mathbf{P}_3^{-1} \cdots \mathbf{P}_{m-1}^{-1} \mathbf{L}_{m-1}^{-1}.$$

# Remark 14

The permutation matrices P and Q are not known ahead of time.

### 4.1. GECP for Ax = b

 $\bullet \ \mathbf{PAQ} = \mathbf{LU}, \quad \mathbf{Ly} = \mathbf{Pb}, \quad \mathbf{Uz} = \mathbf{y}, \quad \mathbf{x} = \mathbf{Qz}$ 

# **Algorithm**: LU factorization PAQ = LU in GECP

The details are left as an exercise.

### • Exercise:

Modify the pseudocode of the algorithms in this lecture to save storage.

# • Further reading:

Shufang Xu, Li Gao, and Pingwen Zhang Numerical Linear Algebra.

Second Edition, Peking University Press, 2013