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



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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 4×4 example

$$\begin{bmatrix} \times \times \times \times \times \\ \times \times \times \times \\ \times \times \times \times \\ \times \times \times \times \end{bmatrix} \xrightarrow{\mathbf{L}_1} \begin{bmatrix} \times \times \times \times \\ \mathbf{0} \times \mathbf{x} \times \\ \mathbf{0} \times \mathbf{x} \times \\ \mathbf{0} \times \mathbf{x} \times \end{bmatrix} \xrightarrow{\mathbf{L}_2} \begin{bmatrix} \times \times \times \times \\ \times \times \times \\ \mathbf{0} \times \mathbf{x} \\ \mathbf{0} \times \mathbf{x} \end{bmatrix} \xrightarrow{\mathbf{L}_3} \begin{bmatrix} \times \times \times \times \\ \times \times \times \\ \mathbf{0} \times \mathbf{x} \\ \mathbf{0} \times \mathbf{x} \end{bmatrix}$$

$$\xrightarrow{\mathbf{A}} \begin{bmatrix} \mathbf{L}_1 \\ \mathbf{0} \times \mathbf{x} \times \\ \mathbf{0} \times \mathbf{x} \times \\ \mathbf{L}_2 \end{bmatrix} \xrightarrow{\mathbf{L}_2} \begin{bmatrix} \mathbf{L}_2 \\ \mathbf{0} \times \mathbf{x} \times \\ \mathbf{0} \times \mathbf{x} \\ \mathbf{0} \times \mathbf{x} \end{bmatrix} \xrightarrow{\mathbf{L}_3} \begin{bmatrix} \times \times \times \times \\ \times \times \times \\ \mathbf{0} \times \mathbf{x} \\ \mathbf{0} \times \mathbf{x} \end{bmatrix}$$

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

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

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

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

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

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

1.1. General formulas for LU factorization

- Let \mathbf{u}_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 u_{kk} . To do this we construct the matrix \mathbf{L}_k :

$$\mathbf{L}_k = egin{bmatrix} 1 & & & & & & & \\ & \ddots & & & & & & \\ & & 1 & & & & & \\ & & -\ell_{k+1,k} & 1 & & & \\ & & dots & & \ddots & & \\ & & -\ell_{mk} & & & 1 \end{bmatrix} = egin{bmatrix} \mathbf{I}_{k-1} & \mathbf{0} & \mathbf{0} & & \\ \mathbf{0} & 1 & \mathbf{0} & & & \\ \mathbf{0} & -oldsymbol{l}_k & \mathbf{I}_{m-k} \end{bmatrix},$$

where
$$\boldsymbol{l}_k = \begin{bmatrix} \ell_{k+1,k} & \ell_{k+2,k} & \cdots & \ell_{mk} \end{bmatrix}^{\top}$$
 with the multipliers $\ell_{jk} = \frac{u_{jk}}{u_{kk}}, \quad k+1 \leq j \leq m.$

Proposition 1

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

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

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

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 ℓ_{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 ℓ_{ik} are computed and stored directly into L.

1.2. LU factorization algorithm

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Algorithm: LU factorization \mathbf{A} = \mathbf{L}\mathbf{U}
\mathbf{U} = \mathbf{A}, \quad \mathbf{L} = \mathbf{I}
\mathbf{for} \ k = 1 \ \mathbf{to} \ m - 1
\mathbf{for} \ j = k + 1 \ \mathbf{to} \ m
\ell_{jk} = u_{jk}/u_{kk} \qquad \text{(compute multipliers)}
u_{j,k:m} = u_{j,k:m} - \ell_{jk}u_{k,k:m} \qquad \text{(elimination)}
\mathbf{end}
\mathbf{end}
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- 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

$$y_k = b_k - \sum_{j=1}^{k-1} \ell_{kj} y_j$$

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

$$A = R^*R$$
.

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

2.1. A 4×4 example

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

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

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

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

Step 3:
$$\begin{bmatrix} 4 & 2i \\ \times & 10 \end{bmatrix} = \begin{bmatrix} 2 \\ -1i & 1 \end{bmatrix} \begin{bmatrix} 1 \\ & 9 \end{bmatrix} \begin{bmatrix} 2 & 1i \\ & 1 \end{bmatrix}$$

Step 4:
$$9 = 3 \times 3$$

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

2.2. Algorithm for Cholesky factorization

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

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

end

ullet Exercise: Design an algorithm to compute ${f R}^*$ 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?)

3. Gaussian elimination with partial pivoting (GEPP)

• Partial pivoting: $|u_{ik}| = \max_{k \le j \le m} |u_{jk}|$, rows are interchanged.

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

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

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 ≤ 1 due to the property at step k (after pivot selection and row interchange)

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

3.1. A 4×4 Example

• Step 1. Interchange the first and third rows by P_1

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

First elimination by L_1

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

• Step 2. Interchange the second and fourth rows by P_2

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

Second elimination by L_2

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

• Step 3. Interchange the third and fourth rows by P_3

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

Final elimination by L_3

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

• $A = (L_3P_3L_2P_2L_1P_1)^{-1}U$

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

Is there a unit lower-triangular L? Yes! Let $P = P_3P_2P_1$. Then

$$\mathbf{L} = \mathbf{P} (\mathbf{L}_3 \mathbf{P}_3 \mathbf{L}_2 \mathbf{P}_2 \mathbf{L}_1 \mathbf{P}_1)^{-1} = \mathbf{P}_3 \mathbf{P}_2 \mathbf{L}_1^{-1} \mathbf{P}_2^{-1} \mathbf{P}_3^{-1} \mathbf{P}_3 \mathbf{L}_2^{-1} \mathbf{P}_3^{-1} \mathbf{L}_3^{-1}.$$

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

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

3.2. General formulas for PA = LU

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

 $\mathbf{L} = \mathbf{P} (\mathbf{L}_{m-1} \mathbf{P}_{m-1} \cdots \mathbf{L}_2 \mathbf{P}_2 \mathbf{L}_1 \mathbf{P}_1)^{-1} = \widehat{\mathbf{L}}_1^{-1} \widehat{\mathbf{L}}_2^{-1} \cdots \widehat{\mathbf{L}}_{m-1}^{-1},$
 $\widehat{\mathbf{L}}_k^{-1} = \mathbf{P}_{m-1} \cdots \mathbf{P}_{k+2} \mathbf{P}_{k+1} \mathbf{L}_k^{-1} \mathbf{P}_{k+1}^{-1} \mathbf{P}_{k+2}^{-1} \cdots \mathbf{P}_{m-1}^{-1}.$

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 $\widehat{\mathbf{L}}_k^{-1}$ in GEPP has the same sparsity pattern as that of \mathbf{L}_k^{-1} . 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 $\hat{\mathbf{L}}_k^{-1}$ is equal to \mathbf{L}_k^{-1} but with the \bigstar 's entries permuted.

Remark 8

By Proposition 2, the product $\widehat{\mathbf{L}}_1^{-1}\widehat{\mathbf{L}}_2^{-1}\cdots\widehat{\mathbf{L}}_{m-1}^{-1}$ is unit lower triangular, and the matrices $\widehat{\mathbf{L}}_k^{-1}$ are never formed and multiplied explicitly.

Remark 9

The multipliers ℓ_{jk} are computed and stored in the appropriate places. It is not a one-step process, but a dynamic adjustment one.

Remark 10

We have the LU factorization of $\mathbf{PA} = \mathbf{LU}$ (note that the permutation matrix \mathbf{P} is not known ahead of time) where $\mathbf{P} = \mathbf{P}_{m-1} \cdots \mathbf{P}_2 \mathbf{P}_1$,

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

$$\mathbf{L} = \mathbf{P} (\mathbf{L}_{m-1} \mathbf{P}_{m-1} \cdots \mathbf{L}_2 \mathbf{P}_2 \mathbf{L}_1 \mathbf{P}_1)^{-1} = \widehat{\mathbf{L}}_1^{-1} \widehat{\mathbf{L}}_2^{-1} \cdots \widehat{\mathbf{L}}_{m-1}^{-1},$$

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

3.3. GEPP for Ax = b

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

Algorithm: LU factorization PA = LU in GEPP for Ax = bU = A, L = I, P = I**for** k = 1 **to** m - 1Select i > k to maximize $|u_{ik}|$ (select pivot) (update **P**) $p_{k..} \leftrightarrow p_{i..}$ (update $\widehat{\mathbf{L}}_1^{-1}, \widehat{\mathbf{L}}_2^{-1}, \dots, \widehat{\mathbf{L}}_{i-1}^{-1}$) $\ell_{k,1:k-1} \leftrightarrow \ell_{i,1:k-1}$ (interchange two rows) $u_{k,k:m} \leftrightarrow u_{i,k:m}$ for j = k + 1 to m $\ell_{ik} = u_{ik}/u_{kk}$ (compute multipliers) $u_{j,k:m} = u_{j,k:m} - \ell_{jk} u_{k,k:m}$ (elimination) end end

• Exercise. Add one line in the above algorithm to obtain Pb.

3.4. Growth factor of GEPP

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

Proposition 11

The growth factor ρ of GEPP applied to any matrix $\mathbf{A} \in \mathbb{C}^{m \times m}$ satisfies $\rho \leq 2^{m-1}$.

Proof. See Exercise 22.1.

• The upper bound in the above proposition is sharp. Consider the 5 × 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 $\rho = 2^{5-1} = 16$.

It is easy to construct an $m \times m$ matrix such that $\rho = 2^{m-1}$.

- 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 12

We have the LU factorization (note that the permutation matrices **P** and **Q** are not known ahead of time)

$$PAQ = LU$$

where
$$\mathbf{P} = \mathbf{P}_{m-1} \cdots \mathbf{P}_2 \mathbf{P}_1$$
, $\mathbf{Q} = \mathbf{Q}_1 \mathbf{Q}_2 \cdots \mathbf{Q}_{m-1}$,
$$\mathbf{U} = \mathbf{L}_{m-1} \mathbf{P}_{m-1} \cdots \mathbf{L}_2 \mathbf{P}_2 \mathbf{L}_1 \mathbf{P}_1 \mathbf{A} \mathbf{Q}_1 \mathbf{Q}_2 \cdots \mathbf{Q}_{m-1}$$
,
$$\mathbf{L} = \mathbf{P} (\mathbf{L}_{m-1} \mathbf{P}_{m-1} \cdots \mathbf{L}_2 \mathbf{P}_2 \mathbf{L}_1 \mathbf{P}_1)^{-1} = \widehat{\mathbf{L}}_1^{-1} \widehat{\mathbf{L}}_2^{-1} \cdots \widehat{\mathbf{L}}_{m-1}^{-1}$$
,
$$\widehat{\mathbf{L}}_k^{-1} = \mathbf{P}_{m-1} \cdots \mathbf{P}_{k+2} \mathbf{P}_{k+1} \mathbf{L}_k^{-1} \mathbf{P}_{k+1}^{-1} \mathbf{P}_{k+2}^{-1} \cdots \mathbf{P}_{m-1}^{-1}$$
.

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