

Existence of LU Decomposition

Theorem

A matrix $\mathbf{A} \in \mathbb{R}^{n \times n}$ has an LU decomposition if every (leading) principal submatrix $\mathbf{A}(1:k, 1:k)$ satisfies

$$\det(\mathbf{A}(1:k, 1:k)) \neq 0 \Rightarrow$$

all the pivots $a_{kk}^{(k-1)} \neq 0$

for $k = 1, 2, \dots, n-1$. If the LU decomposition of \mathbf{A} exists and \mathbf{A} is nonsingular, then the LU decomposition is unique and $\det(\mathbf{A}) = u_{11} \cdots u_{nn}$.

Proof

We first show that if $\det(A(1:k, 1:k)) \neq 0 \quad \forall k=1, \dots, n$, then $a_{kk}^{(k-1)} \neq 0$, so that LU decomposition exists.

Let $k=1, \dots, n-1$.

$$A^{(k-1)} = \underbrace{M_{k-1} \cdots M_1}_W A \quad (*)$$

Since each M_i is unit lower triangular, so is W . (fact 1)
We partition (*) as follows:

$$\left[\begin{array}{c|c} A^{(k-1)}(1:k, 1:k) & * \\ \hline * & * \end{array} \right] = \left[\begin{array}{c|c} W(1:k, 1:k) & 0 \\ \hline * & * \\ * & * \end{array} \right] \cdot \left[\begin{array}{c|c} A(1:k, 1:k) & * \\ \hline * & * \end{array} \right]$$

Proof (cont'd)

This gives $A^{(k-1)}(1:k, 1:k) = W(1:k, 1:k) A(1:k, 1:k)$ (#)

Note that $A^{(k-1)}(1:k, 1:k)$ is upper triangular.

$$\prod_{i=1}^k a_{ii}^{(k-1)} = \det(A^{(k-1)}(1:k, 1:k)) \stackrel{(\#)}{=} \det(W(1:k, 1:k)) \cdot \det(A(1:k, 1:k))$$

= 1
≠ 0

It follows that $a_{kk}^{(k-1)} \neq 0$.

\Rightarrow LU decomposition exists.

Proof (cont'd)

Let A be nonsingular, i.e., $\det(A) \neq 0$.

Assume to the contrary that A has two LU decompositions $L_1 U_1$ and $L_2 U_2$. Note that L_1, U_1, L_2, U_2 nonsingular.

Then,
$$\underbrace{L_2^{-1} L_1 U_1 U_1^{-1}} = \underbrace{L_2^{-1} L_2 U_2 U_1^{-1}}$$

Facts 1 & 3
$$\underbrace{L_2^{-1} L_1}_{\downarrow} = \underbrace{U_2 U_1^{-1}}_{\downarrow}$$

Unit lower triangular

upper triangular

Therefore, the above equation only holds for

$$L_2^{-1} L_1 = U_2 U_1^{-1} = I \Rightarrow L_1 = L_2, U_1 = U_2.$$

Finally, $\det(A) = \det(L) \det(U) = \det(U) = \prod_{i=1}^n u_{ii}$.

Matrix Computations

Chapter 2 Linear systems and LU decomposition

Section 2.2 Pivoting for LU Decomposition

Jie Lu
ShanghaiTech University

Pivoting

- Previously, we assume all the pivots are nonzero. What if some $a_{kk}^{(k-1)}$ happens to be zero?
- Gaussian elimination is known to be numerically unstable when a pivot is close to zero
 - Relatively small pivots can cause large entries in **L** and **U** and thus non-negligible error in solution due to round-off errors
- **Pivoting**: Find permutations of A with a proper LU decomposition
 - Partial pivoting, complete pivoting, rook pivoting, etc.

Permutation Matrix

A square matrix with exactly one entry of 1 in each row and each column and 0 elsewhere is a **permutation matrix**

Example: Let $\mathbf{\Pi}$ be a 4×4 permutation matrix and $\mathbf{A} \in \mathbb{R}^4$

$$\mathbf{\Pi} = \begin{bmatrix} 0 & 0 & 0 & 1 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 1 & 0 & 0 & 0 \end{bmatrix} \quad \mathbf{A} = [\mathbf{a}_1 \quad \mathbf{a}_2 \quad \mathbf{a}_3 \quad \mathbf{a}_4] = \begin{bmatrix} -\tilde{\mathbf{a}}_1^T & - \\ -\tilde{\mathbf{a}}_2^T & - \\ -\tilde{\mathbf{a}}_3^T & - \\ -\tilde{\mathbf{a}}_4^T & - \end{bmatrix}$$

- $\mathbf{\Pi A}$ is obtained by swapping row 1 and row 4 of \mathbf{A}
- $\mathbf{A \Pi}$ is obtained by swapping column 1 and column 4 of \mathbf{A}

$$\mathbf{\Pi A} = \begin{bmatrix} -\tilde{\mathbf{a}}_4^T & - \\ -\tilde{\mathbf{a}}_2^T & - \\ -\tilde{\mathbf{a}}_3^T & - \\ -\tilde{\mathbf{a}}_1^T & - \end{bmatrix} \quad \mathbf{A \Pi} = [\mathbf{a}_4 \quad \mathbf{a}_2 \quad \mathbf{a}_3 \quad \mathbf{a}_1]$$

Permutation Matrix

Example:

$$\mathbf{P} = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix}, \quad \mathbf{PA} = \begin{bmatrix} -\tilde{\mathbf{a}}_2^T \\ -\tilde{\mathbf{a}}_4^T \\ -\tilde{\mathbf{a}}_1^T \\ -\tilde{\mathbf{a}}_3^T \end{bmatrix}, \quad \mathbf{AP} = [\mathbf{a}_3 \quad \mathbf{a}_1 \quad \mathbf{a}_4 \quad \mathbf{a}_2]$$

$\times [a_2 \ a_4 \ a_1 \ a_3]$

Note that \mathbf{P} can be decomposed as

$$\mathbf{P} = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} = \begin{bmatrix} 0 & 0 & 0 & 1 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 1 & 0 & 0 & 0 \end{bmatrix} \cdot \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \end{bmatrix} \cdot \begin{bmatrix} 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

P_3

P_2

P_1

$$P_1 A = \begin{bmatrix} \tilde{\mathbf{a}}_3^T \\ \tilde{\mathbf{a}}_2^T \\ \tilde{\mathbf{a}}_4^T \\ \tilde{\mathbf{a}}_1^T \end{bmatrix}$$

\rightarrow

$$P_2 (P_1 A) =$$

$$\begin{bmatrix} \tilde{\mathbf{a}}_3^T \\ \tilde{\mathbf{a}}_4^T \\ \tilde{\mathbf{a}}_1^T \\ \tilde{\mathbf{a}}_2^T \end{bmatrix}$$

\rightarrow

$$P_3 (P_2 P_1 A)$$

$= \dots$

Interchange Permutations

Let $\Pi_k \in \mathbb{R}^{n \times n}$, $k = 1, \dots, m \leq n$ be the $n \times n$ identity matrix \mathbf{I} with row k and row $\text{piv}(k)$ swapped, which are called **interchange permutations**

Let $\mathbf{P} = \Pi_m \cdots \Pi_1$ $\in \{1, \dots, n\}$

- Π_k is symmetric (but \mathbf{P} may not be symmetric)
- $\mathbf{P}^T = \Pi_1 \cdots \Pi_m$
- If $\text{piv} = [1, \dots, m]^T$, then $\mathbf{P} = \mathbf{I}$ $\hookrightarrow \text{piv}(k) = k$

Π_k nonsingular
 $\Pi_k^{-1} = \Pi_k$

Computation of $\mathbf{P}\mathbf{x}$, $\mathbf{x} \in \mathbb{R}^n$

```
for k=1:m % overwrite x with Px
    x(k) ↔ x(piv(k)) % swap entry k and entry piv(k)
end
```

Computation of $\mathbf{P}^T\mathbf{x}$, $\mathbf{x} \in \mathbb{R}^n$

```
for k=m:-1:1
    x(k) ↔ x(piv(k))
end
```

No flops needed for permutation (but affect performance nontrivially)

Partial Pivoting

Recall Upper Triangularization in Section 2.1

Given $\mathbf{A}^{(k-1)}$, $k = 1, \dots, n-1$,

1. Find $piv(k) = \arg \max_{j \in [k, n]} |\mathbf{A}^{(k-1)}(j, k)|$
2. Let $\Pi_k \in \mathbb{R}^{n \times n}$ be the interchange permutation that swaps row k and row $piv(k)$ of \mathbf{I}
3. Determine the Gauss Transformation $\mathbf{M}_k = \mathbf{I}_n - \boldsymbol{\tau}^{(k)} \mathbf{e}_k^T$, where

$$\boldsymbol{\tau}^{(k)} = \begin{bmatrix} \mathbf{0}_k \\ (\Pi_k \mathbf{A}^{(k-1)})(k+1:n, k) / (\Pi_k \mathbf{A}^{(k-1)})(k, k) \end{bmatrix}$$

4. $\mathbf{A}^{(k)} = \mathbf{M}_k(\Pi_k \mathbf{A}^{(k-1)})$ (which satisfies $\mathbf{A}^{(k)}(k+1:n, k) = \mathbf{0}$)

Upon completing the above process, we have

$$\mathbf{M}_{n-1} \Pi_{n-1} \cdots \mathbf{M}_1 \Pi_1 \mathbf{A} = \mathbf{U}$$

Note that all the elements in $\boldsymbol{\tau}^{(k)}(k+1:n)$ are ≤ 1 in absolute value

Partial Pivoting (cont'd)

Example: $A = \begin{bmatrix} 2 & 1 & 1 & 0 \\ 4 & 3 & 3 & 1 \\ 8 & 7 & 9 & 5 \\ 6 & 7 & 9 & 8 \end{bmatrix} = A^{(0)}$

$k=1$: $\text{piv}(1)=3$ $\pi_1 = \begin{bmatrix} 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$

$\pi_1 A = \begin{bmatrix} 8 & 7 & 9 & 5 \\ 4 & 3 & 3 & 1 \\ 2 & 1 & 1 & 0 \\ 6 & 7 & 9 & 8 \end{bmatrix}$

$M_1 = \begin{bmatrix} 1 & & & \\ -\frac{1}{2} & 1 & & \\ -\frac{1}{4} & & 1 & \\ -\frac{3}{4} & & & 1 \end{bmatrix}$, $A^{(1)} = M_1 \pi_1 A^{(0)} = \begin{bmatrix} 8 & 7 & 9 & 5 \\ 0 & -\frac{1}{2} & -\frac{3}{2} & -\frac{3}{2} \\ 0 & -\frac{3}{4} & -\frac{5}{4} & -\frac{5}{4} \\ 0 & \frac{7}{4} & \frac{9}{4} & \frac{17}{4} \end{bmatrix}$

Partial Pivoting (cont'd)

$$k=2: \text{piv}(2)=4. \quad T_{12} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \end{bmatrix}$$

$$T_{12} A^{(1)} = \begin{bmatrix} 8 & 7 & 9 & 5 \\ 0 & \frac{7}{4} & \frac{9}{4} & \frac{17}{4} \\ 0 & -\frac{3}{4} & -\frac{5}{4} & -\frac{5}{4} \\ 0 & -\frac{1}{2} & -\frac{3}{2} & -\frac{3}{2} \end{bmatrix}$$

$$M_2 = \begin{bmatrix} 1 & & & \\ & \frac{1}{\frac{7}{4}} & & \\ & & 1 & \\ & & & 1 \end{bmatrix}$$

$$A^{(2)} = M_2 T_{12} A^{(1)} = \begin{bmatrix} 8 & 7 & 9 & 5 \\ 0 & \frac{7}{4} & \frac{9}{4} & \frac{17}{4} \\ 0 & 0 & -\frac{2}{7} & \frac{8}{7} \\ 0 & 0 & -\frac{6}{7} & \frac{2}{7} \end{bmatrix}$$

Partial Pivoting (cont'd)

$$k=3: \quad \text{piv}(3) = 4$$

$$T_3 = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 \end{bmatrix}$$

$$T_3 A^{(2)} = \begin{bmatrix} 8 & 7 & 9 & 5 \\ 0 & \frac{7}{4} & \frac{9}{4} & \frac{17}{4} \\ 0 & 0 & -\frac{6}{7} & -\frac{2}{7} \\ 0 & 0 & -\frac{2}{7} & \frac{4}{7} \end{bmatrix}$$

$$M_3 = \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & -\frac{1}{3} \end{bmatrix}$$

$$A^{(3)} = M_3 T_3 A^{(2)} = \begin{bmatrix} 8 & 7 & 9 & 5 \\ 0 & \frac{7}{4} & \frac{9}{4} & \frac{17}{4} \\ 0 & 0 & -\frac{6}{7} & -\frac{2}{7} \\ 0 & 0 & 0 & \frac{2}{3} \end{bmatrix}$$

Computation of \mathbf{L} with Partial Pivoting

Define $\mathbf{P} = \mathbf{\Pi}_{n-1} \cdots \mathbf{\Pi}_1$ and for each $k = 1, \dots, n-1$,

When $k = n-1$

$$\tilde{\mathbf{M}}_k = (\mathbf{\Pi}_{n-1} \cdots \mathbf{\Pi}_{k+1}) \mathbf{M}_k (\mathbf{\Pi}_{k+1} \cdots \mathbf{\Pi}_{n-1})$$

$$\tilde{\mathbf{M}}_k = \mathbf{M}_k$$

Note: $\tilde{\mathbf{M}}_k$ is a Gauss transformation

$$\tilde{\mathbf{M}}_k = (\mathbf{\Pi}_{n-1} \cdots \mathbf{\Pi}_{k+1}) \cdot (\mathbf{I} - \boldsymbol{\tau}^{(k)} \mathbf{e}_k^T) \cdot (\mathbf{\Pi}_{k+1} \cdots \mathbf{\Pi}_{n-1}) = \mathbf{I} - \tilde{\boldsymbol{\tau}}^{(k)} \mathbf{e}_k^T$$

with $\tilde{\boldsymbol{\tau}}^{(k)} = \mathbf{\Pi}_{n-1} \cdots \mathbf{\Pi}_{k+1} \boldsymbol{\tau}^{(k)}$ (Why?)

$$\tilde{\mathbf{M}}_k = \mathbf{\Pi}_{n-1} \cdots \mathbf{\Pi}_{k+1} \mathbf{I} \mathbf{\Pi}_{k+1} \cdots \mathbf{\Pi}_{n-1} - \underbrace{\mathbf{\Pi}_{n-1} \cdots \mathbf{\Pi}_{k+1}}_{\tilde{\boldsymbol{\tau}}^{(k)}} \mathbf{e}_k^T \underbrace{\mathbf{\Pi}_{k+1} \cdots \mathbf{\Pi}_{n-1}}_{\text{interchange permutation only for } k+1:n \text{ columns rows}}$$

$$\mathbf{e}_k^T = [0 \cdots 0 \mid 1 \mid 0 \cdots 0]$$

↑
k-th

Computation of \mathbf{L} with Partial Pivoting (cont'd)

Example: Let $n = 4$

$$\begin{aligned}\tilde{\mathbf{M}}_3 \tilde{\mathbf{M}}_2 \tilde{\mathbf{M}}_1 \mathbf{P} \mathbf{A} &= \mathbf{M}_3 \cdot (\Pi_3 \mathbf{M}_2 \Pi_3) \cdot (\Pi_3 \Pi_2 \mathbf{M}_1 \Pi_2 \Pi_3) \cdot (\Pi_3 \Pi_2 \Pi_1) \mathbf{A} \\ &= \mathbf{M}_3 \Pi_3 \mathbf{M}_2 \Pi_2 \mathbf{M}_1 \Pi_1 \mathbf{A} = \mathbf{U}\end{aligned}$$

We can easily extend this to general n and obtain

$$\tilde{\mathbf{M}}_{n-1} \cdots \tilde{\mathbf{M}}_1 \mathbf{P} \mathbf{A} = \mathbf{U}$$

In addition, let

$$\mathbf{L} = \tilde{\mathbf{M}}_1^{-1} \cdots \tilde{\mathbf{M}}_{n-1}^{-1} = (\mathbf{I} + \tilde{\mathbf{r}}^{(1)} \mathbf{e}_1^T) \cdots (\mathbf{I} + \tilde{\mathbf{r}}^{(n-1)} \mathbf{e}_{n-1}^T) = \mathbf{I} + \sum_{k=1}^{n-1} \tilde{\mathbf{r}}^{(k)} \mathbf{e}_k^T$$

where each entry of \mathbf{L} is ≤ 1 (Why?)

absolute value

Therefore, LU decomposition with pivoting is equivalent to

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