

1

Matrices and Gaussian Elimination

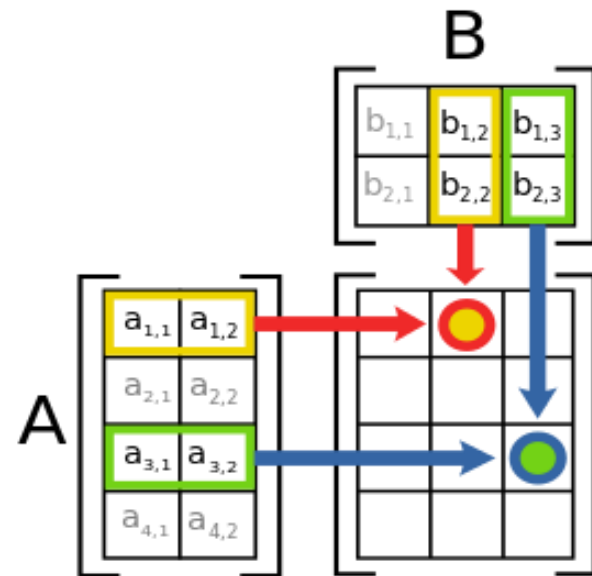
1.7

TRIANGULAR FACTORS AND
ROW EXCHANGES

(矩阵的三角分解和换行)

LU Factorization

Row Exchanges

*** Textbook: Section 1.5 + Section 1.6 (part)**

I. Triangular Factors (矩阵的LU分解)

Example 1 将矩阵

$$A = \begin{bmatrix} 2 & 1 & 0 & 0 \\ 1 & 2 & 1 & 0 \\ 0 & 1 & 2 & 1 \\ 0 & 0 & 1 & 2 \end{bmatrix}$$

分解成为主对角元为1的下三角矩阵 L 和上三角矩阵 U 的乘积, 即 $A=LU$ (称为**矩阵的LU分解** or **Triangular factorization** $A=LU$) .

解 利用倍加初等变换(Replacement)把 A 变为上三角矩阵:

$$E_{12}\left(-\frac{1}{2}\right)A = \begin{bmatrix} 2 & 1 & 0 & 0 \\ 0 & \frac{3}{2} & 1 & 0 \\ 0 & 1 & 2 & 1 \\ 0 & 0 & 1 & 2 \end{bmatrix}, \quad E_{23}\left(-\frac{2}{3}\right)E_{12}\left(-\frac{1}{2}\right)A = \begin{bmatrix} 2 & 1 & 0 & 0 \\ 0 & \frac{3}{2} & 1 & 0 \\ 0 & 0 & \frac{4}{3} & 1 \\ 0 & 0 & 1 & 2 \end{bmatrix},$$

$$E_{34}\left(-\frac{3}{4}\right)E_{23}\left(-\frac{2}{3}\right)E_{12}\left(-\frac{1}{2}\right)A = \begin{bmatrix} 2 & 1 & 0 & 0 \\ 0 & \frac{3}{2} & 1 & 0 \\ 0 & 0 & \frac{4}{3} & 1 \\ 0 & 0 & 0 & \frac{5}{4} \end{bmatrix} = U$$

$$A = E_{12}^{-1}\left(-\frac{1}{2}\right)E_{23}^{-1}\left(-\frac{2}{3}\right)E_{34}^{-1}\left(-\frac{3}{4}\right)U = E_{12}\left(\frac{1}{2}\right)E_{23}\left(\frac{2}{3}\right)E_{34}\left(\frac{3}{4}\right)U = LU$$

其中 $L = E_{12}\left(\frac{1}{2}\right)E_{23}\left(\frac{2}{3}\right)E_{34}\left(\frac{3}{4}\right).$

$$L = \begin{bmatrix} 1 & 0 & 0 & 0 \\ \frac{1}{2} & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & \frac{2}{3} & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & \frac{3}{4} & 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ \frac{1}{2} & 1 & 0 & 0 \\ 0 & \frac{2}{3} & 1 & 0 \\ 0 & 0 & \frac{3}{4} & 1 \end{bmatrix}$$

Remark: In Example 1, A ($n \times n$ matrix) is written in the form $A = LU$, where L is an $n \times n$ lower triangular matrix with 1's on the diagonal and U is an $n \times n$ upper triangular matrix.

$$A = \begin{bmatrix} 2 & 1 & 0 & 0 \\ 1 & 2 & 1 & 0 \\ 0 & 1 & 2 & 1 \\ 0 & 0 & 1 & 2 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ \frac{1}{2} & 1 & 0 & 0 \\ 0 & \frac{2}{3} & 1 & 0 \\ 0 & 0 & \frac{3}{4} & 1 \end{bmatrix} \begin{bmatrix} 2 & 1 & 0 & 0 \\ 0 & \frac{3}{2} & 1 & 0 \\ 0 & 0 & \frac{4}{3} & 1 \\ 0 & 0 & 0 & \frac{5}{4} \end{bmatrix}$$

*Divide out of U
a diagonal pivot
matrix D*

$$A = \begin{bmatrix} 1 & 0 & 0 & 0 \\ \frac{1}{2} & 1 & 0 & 0 \\ 0 & \frac{2}{3} & 1 & 0 \\ 0 & 0 & \frac{3}{4} & 1 \end{bmatrix} \begin{bmatrix} \frac{2}{1} & 0 & 0 & 0 \\ 0 & \frac{3}{2} & 0 & 0 \\ 0 & 0 & \frac{4}{3} & 0 \\ 0 & 0 & 0 & \frac{5}{4} \end{bmatrix} \begin{bmatrix} 1 & \frac{1}{2} & 0 & 0 \\ 0 & 1 & \frac{2}{3} & 0 \\ 0 & 0 & 1 & \frac{3}{4} \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

A is symmetric:

$$A = LDL^T.$$

L

D

V

习惯上仍记为 U
(此时为单位上三角矩阵)

The triangular factorization can be written $A = LDU$, where L and U have 1's on the diagonal and D is the diagonal matrix of pivots.

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

Remarks: 1. *The LDU factorization is **uniquely** determined by A if A is invertible.*

(Proof: P53, Problem Set 1.6, #17)

2. Some matrices *cannot* be factored into $A = LU$ or LDU .

For instance, $A = \begin{bmatrix} 0 & 2 \\ 3 & 4 \end{bmatrix}$.

练习 将矩阵

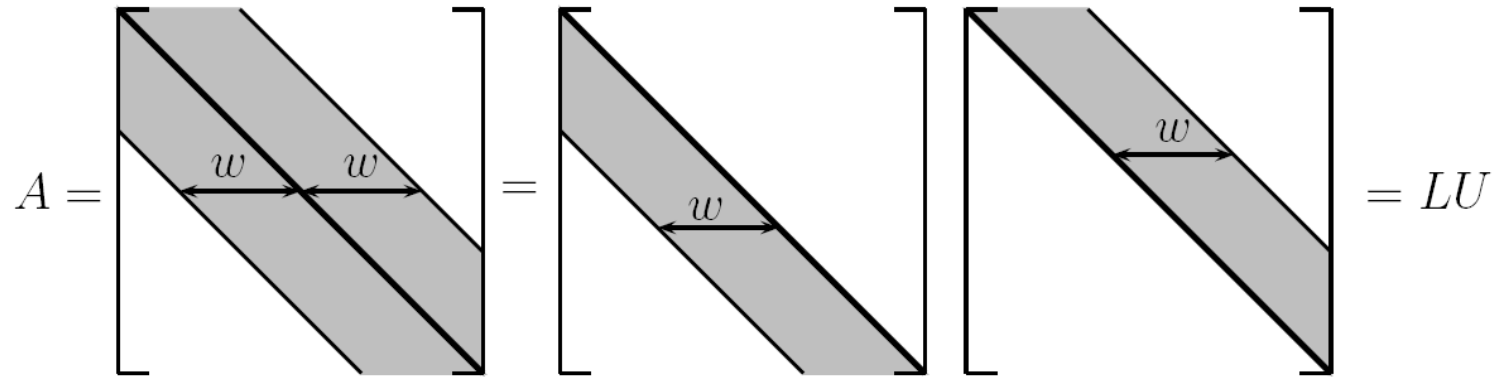
$$A = \begin{bmatrix} 1 & -1 & 0 & 0 \\ -1 & 2 & -1 & 0 \\ 0 & -1 & 2 & -1 \\ 0 & 0 & -1 & 2 \end{bmatrix}$$

分解成为主对角元为1的下三角矩阵 L (invertible, unit lower triangular matrix)和上三角矩阵 U (upper triangular matrix)的乘积, 即 $A=LU$.

解

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

Remark: band matrix (带状矩阵) (P61, Figure 1.8)



A band matrix A and its factors L and U .

A band matrix A has $a_{ij} = 0$ except in the band $|i - j| < w$.

w : “half bandwidth”

$w = 1$: a diagonal matrix,

$w = 2$: a tridiagonal matrix (三对角矩阵),

$w = n$: a full matrix.

$$\begin{bmatrix} 2 & 1 & 0 & 0 \\ 1 & 2 & 1 & 0 \\ 0 & 1 & 2 & 1 \\ 0 & 0 & 1 & 2 \end{bmatrix}$$

Example 2 Solve $Ax = b$

$$\begin{array}{rrrrrrcl}
 x_1 & - & x_2 & & & & = & 1 \\
 -x_1 & + & 2x_2 & - & x_3 & & = & 1 \\
 & & -x_2 & + & 2x_3 & - & x_4 & = & 1 \\
 & & & & -x_3 & + & 2x_4 & = & 1
 \end{array}$$

This is the previous matrix A with a right-hand side

$$b = (1, 1, 1, 1)^T.$$

$Ax = b$ splits into $Lc = b$ and $Ux = c$

$$(LU)x = b \Rightarrow Ux = c \text{ \& } Lc = b$$

$$\begin{array}{rrrrrrcl}
 & & c_1 & & & & = & 1 \\
 Lc = b & & -c_1 & + & c_2 & & = & 1 \\
 & & & - & c_2 & + & c_3 & = & 1 \\
 & & & & & - & c_3 & + & c_4 & = & 1
 \end{array}$$

solved forward

gives $c = \begin{bmatrix} 1 \\ 2 \\ 3 \\ 4 \end{bmatrix}.$

$$\begin{array}{rrrrrrcl}
 & & x_1 & - & x_2 & & = & 1 \\
 & & & & x_2 & - & x_3 & = & 2 \\
 Ux = c & & & & & & x_3 & - & x_4 & = & 3 \\
 & & & & & & & & x_4 & = & 4
 \end{array}$$

solved backward

gives $x = \begin{bmatrix} 10 \\ 9 \\ 7 \\ 4 \end{bmatrix}.$

One Linear System = Two Triangular Systems

Splitting of $Ax = b$

First $Lc = b$ and then $Ux = c$.

1. *Factor* (from A find its factors L and U).
2. *Solve* (from L and U and b find the solution x).

$$\begin{array}{ccc}
 A = \begin{bmatrix} 1 & & & \\ * & 1 & & \\ * & * & 1 & \\ * & * & * & 1 \end{bmatrix} & \begin{bmatrix} \blacksquare & * & * & * \\ & \blacksquare & * & * \\ & & \blacksquare & * \\ & & & \blacksquare \end{bmatrix} \\
 n \times n & n \times n & n \times n
 \end{array}$$

What if A is an $m \times n$ matrix ?

LU factorization

$$A = \underbrace{\begin{bmatrix} 1 & 0 & 0 & 0 \\ * & 1 & 0 & 0 \\ * & * & 1 & 0 \\ * & * & * & 1 \end{bmatrix}}_L \underbrace{\begin{bmatrix} \blacksquare & * & * & * & * \\ 0 & \blacksquare & * & * & * \\ 0 & 0 & 0 & \blacksquare & * \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}}_U$$

Notes: Assume that A is an $m \times n$ matrix that can be row reduced to echelon form, *without row interchanges*.

Then A can be written in the form $A = LU$, where L is an $m \times m$ lower triangular matrix with 1's on the diagonal and U is an $m \times n$ echelon form of A .

$$A = \begin{bmatrix} 0 & 2 \\ 3 & 4 \end{bmatrix}, \quad \text{which needs a row exchange,}$$

cannot be factored into $A = LU$.

II. Row Exchanges and Permutation Matrices

$$\mathbf{A} = \begin{bmatrix} 0 & 2 \\ 3 & 4 \end{bmatrix}, \quad \text{cannot be factored into } \mathbf{A} = \mathbf{LU}.$$

Remedy: *Exchange the two rows*

$$\mathbf{P}_{12} = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}, \quad \mathbf{P}_{12}\mathbf{A} = \begin{bmatrix} 3 & 4 \\ 0 & 2 \end{bmatrix}.$$


Permutation matrix (置换矩阵)

A permutation matrix has the same rows as the identity matrix but in some order.

There is a single “1” in every row and every column.

How many permutation matrices do we have for $n = 2$?
 $n = 3$? ...

$$n = 2 \quad \mathbf{I} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \quad \mathbf{P}_{12} = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}.$$

$$n = 3 \quad \mathbf{I} = \begin{bmatrix} 1 & & \\ & 1 & \\ & & 1 \end{bmatrix}, \quad \mathbf{P}_{21} = \begin{bmatrix} & 1 & \\ 1 & & \\ & & 1 \end{bmatrix}, \quad \mathbf{P}_{32}\mathbf{P}_{21} = \begin{bmatrix} & 1 & \\ & & 1 \\ 1 & & \end{bmatrix},$$

$$\mathbf{P}_{31} = \begin{bmatrix} & & 1 \\ & 1 & \\ 1 & & \end{bmatrix}, \quad \mathbf{P}_{32} = \begin{bmatrix} 1 & & \\ & & 1 \\ & 1 & \end{bmatrix}, \quad \mathbf{P}_{21}\mathbf{P}_{32} = \begin{bmatrix} & & 1 \\ 1 & & \\ & 1 & \end{bmatrix}.$$

There are $n! = n(n - 1) \dots (1)$ permutations of size n .

A zero in the pivot location raises two possibilities:
The trouble may be easy to fix, or it may be serious.

$$A = \begin{bmatrix} 0 & a & b \\ 0 & 0 & c \\ d & e & f \end{bmatrix},$$

$$P_{13} = \begin{bmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{bmatrix}, \quad P_{23} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix},$$

$$\boxed{P_{23}P_{13}}A = \begin{bmatrix} d & e & f \\ 0 & a & b \\ 0 & 0 & c \end{bmatrix}$$

$$\mathbf{P} = P_{23}P_{13} = \begin{bmatrix} 0 & 0 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}$$

$d = 0 \implies$ no first pivot

$a = 0 \implies$ no second pivot

$c = 0 \implies$ no third pivot.

With the rows in the right order $\mathbf{P}A$, any nonsingular matrix is ready for elimination.

$$PA = LU$$

Example:

$$A = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 3 \\ 2 & 5 & 8 \end{bmatrix} \longrightarrow \begin{bmatrix} 1 & 1 & 1 \\ 0 & 0 & 2 \\ 0 & 3 & 6 \end{bmatrix} \longrightarrow \begin{bmatrix} 1 & 1 & 1 \\ 0 & 3 & 6 \\ 0 & 0 & 2 \end{bmatrix}$$

$$P = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix}, PA = \begin{bmatrix} 1 & 1 & 1 \\ 2 & 5 & 8 \\ 1 & 1 & 3 \end{bmatrix} \longrightarrow \begin{bmatrix} 1 & 1 & 1 \\ 0 & 3 & 6 \\ 0 & 0 & 2 \end{bmatrix} = U$$

P

$$PA = \begin{bmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ 1 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 \\ 0 & 3 & 6 \\ 0 & 0 & 2 \end{bmatrix}$$

L U

Elimination in a Nutshell: $PA = LU$

In the *nonsingular* case, there is a permutation matrix P that reorders the rows of A to avoid zeros in the pivot positions. Then $A\mathbf{x} = \mathbf{b}$ has a *unique solution*.

With the rows reordered in advance, PA can be factored into LU .

In the *singular* case, no P can produce a full set of pivots: elimination fails.

Remark:

In practice, we also consider a row exchange when the original pivot is *near* zero — even if it is not exactly zero. Choosing a larger pivot reduces the roundoff error. (*partial pivoting*)

Partial pivoting (部分主元法) : Numerical note (P62)Elimination with small pivot

$$\begin{cases} 0.0001u + v = 1 \\ u + v = 2 \end{cases} \longrightarrow \begin{cases} 0.0001u + v = 1 \\ -9999v = -9998 \end{cases}$$

Correct result

$$\begin{cases} 0.0001u + v = 1 \\ v = 0.9999 \end{cases} \longrightarrow u = 1$$

Wrong result

$$\begin{cases} 0.0001u + v = 1 \\ v = 1 \end{cases} \longrightarrow u = 0$$

The small pivot 0.0001 brought instability, and the remedy is clear – *exchange rows*.

Partial pivoting (部分主元法)

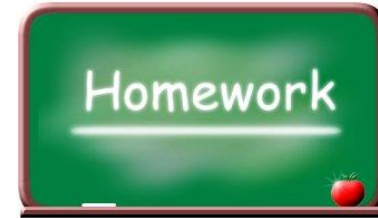
Exchange rows

$$\begin{cases} 0.0001u + v = 1 \\ u + v = 2 \end{cases} \xrightarrow{r_1 \longleftrightarrow r_2} \begin{cases} u + v = 2 \\ 0.0001u + v = 1 \end{cases}$$

$$\begin{cases} u + v = 2 \\ 0.9999v = 0.9998 \end{cases} \longrightarrow v = 1, \quad u = 1. \quad \text{Correct result}$$

A small pivot forces a practical change in elimination. Normally we compare each pivot with all possible pivots in the same column. Exchanging rows to obtain the **largest** possible pivot (having the largest absolute value) is called **partial pivoting**.

Homework



- See Blackboard announcement
- ***Hardcover* textbook + Supplementary problems**

Deadline (DDL):

- Next tutorial class

