

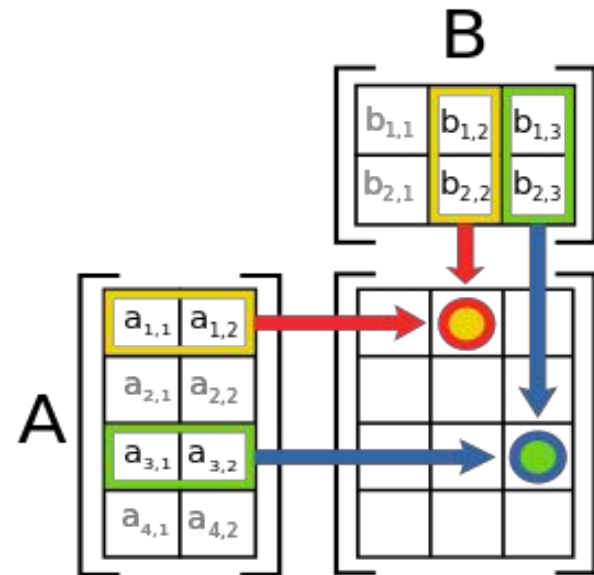
## 1

## 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  $A = \mathbf{L}\mathbf{D}\mathbf{U}$ , where  $\mathbf{L}$  and  $\mathbf{U}$  have 1's on the diagonal and  $\mathbf{D}$  is the diagonal matrix of pivots.

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

**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  $A = \mathbf{L}\mathbf{U}$  or  $\mathbf{L}\mathbf{D}\mathbf{U}$ .

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

练习 将矩阵

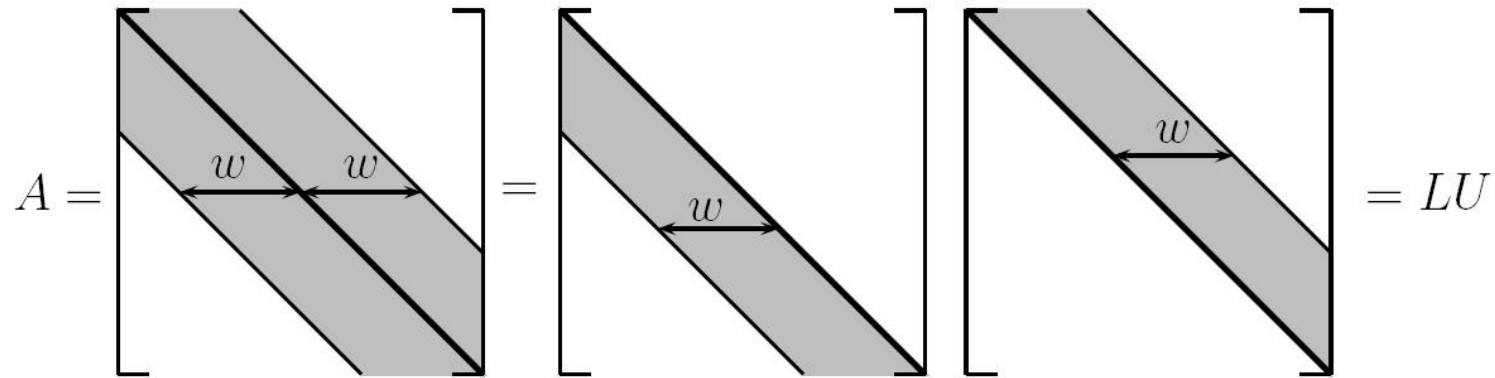
$$A = \begin{bmatrix} 1 & -1 & 0 & 0 \\ -1 & 2 & -1 & 0 \\ 0 & -1 & 2 & -1 \\ 0 & 0 & -1 & 2 \end{bmatrix}$$

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 $\underline{Ax = b}$

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

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

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

$\underline{Ax = b}$  splits into  $\underline{Lc = b}$  and  $\underline{Ux = c}$

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

$$\begin{array}{rrrrrcl} & & c_1 & & & = & 1 \\ Lc = b & & -c_1 & + & c_2 & = & 1 \\ & & & - & c_2 & + & c_3 = & 1 \\ \text{solved forward} & & & & - & c_3 & + & c_4 = & 1 \end{array}$$

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

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

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  $A$ .

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

cannot be factored into  $A = LU$ .

## II. Row Exchanges and Permutation Matrices

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

**Remedy:** *Exchange the two rows*

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


**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  $n = 2$ ?  
 $n = 3$ ? ...

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

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

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

*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.*

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

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

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

$d = 0 \implies$  no first pivot

$a = 0 \implies$  no second pivot

$c = 0 \implies$  no third pivot.

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

With the rows in the right order  $\mathbf{PA}$ , 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 )

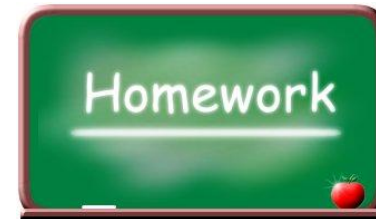
### Exchange rows

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

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

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

