

## 6.5. Matrix factorization.

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Q. Can we present  $A=LU$ , where  $L$  is lower triangular and  $U$  is upper triangular?

Motivation: Solving  $A\vec{x}=\vec{b}$  requires  $O(\frac{n^3}{3})$  using Gaussian elimination. However, an upper-triangular system requires only backsubstitution, which takes  $O(n^2)$  operations. A similar number of operations is needed to solve a lower-triangular system.

Assume  $A=LU$  and one needs to solve  $A\vec{x}=\vec{b}$ .

Step 1. Let  $\vec{y}=U\vec{x}$ . Solve  $L\vec{y}=\vec{b}$  for  $\vec{y}$ . This requires  $O(n^2)$  operations.

Step 2. Solve  $U\vec{x}=\vec{y}$ . This requires  $O(n^2)$  operations.

Both steps require  $O(2n^2)$  operations as opposed to  $O(\frac{n^3}{3})$  when  $A$  is in a non-factored form.

Determining  $L$  and  $U$  requires  $O(\frac{n^3}{3})$  operations.

Assume  $L$  and  $U$  can be obtained by Gaussian elimination without row interchanges, which is equivalent to  $a_{ii}^{(i)} \neq 0$  ( $i=1, \dots, n$ ).

Step 1 of Gaussian elimination

$$(E_j - m_{j,1} E_1) \rightarrow (E_j) \quad \text{where } m_{j,1} = \frac{a_{j,1}^{(1)}}{a_{1,1}^{(1)}} \quad (6.8).$$

$$A = \begin{bmatrix} a_{11} & \dots & a_{1n} \\ \vdots & & \vdots \\ a_{n1} & \dots & a_{nn} \end{bmatrix}$$

Introduce

$$M^{(1)} = \begin{bmatrix} 1 & 0 & \cdots & 0 \\ -m_{21} & 1 & & \\ \vdots & & \ddots & \\ -m_{n1} & 0 & & 1 \end{bmatrix}$$

Denote  $A^{(1)} = A$ . Calculate

$$A^{(2)} \equiv M^{(1)}A = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ 0 & -m_{21}a_{21} + a_{22} & & -m_{21}a_{n1} + a_{n2} \\ \vdots & & \ddots & \\ 0 & -m_{n1}a_{21} + a_{n2} & & -m_{n1}a_{n1} + a_{nn} \end{bmatrix} \quad \text{which is exactly (6.8).}$$

Denote  $M^{(2)}\vec{b} = \vec{b}^{(2)}$ . Then a solution to  $A\vec{x} = \vec{b}$  satisfies

$$A^{(2)}\vec{x} = M^{(1)}A\vec{x} = M^{(1)}\vec{b} = \vec{b}^{(2)}.$$

Step 2. Construct  $M^{(2)} = \begin{bmatrix} 1 & 0 & & 0 \\ 0 & 1 & & 0 \\ & & \ddots & \\ 0 & & -m_{n2} & 1 \end{bmatrix}$

$$m_{j2} = \frac{a_{j2}^{(2)}}{a_{22}^{(2)}}$$

$A^{(3)} \equiv M^{(2)}A^{(2)}$ ,  $\vec{b}^{(3)} = M^{(2)}\vec{b}^{(2)}$ . Then

$$A^{(3)}\vec{x} = M^{(2)}A^{(2)}\vec{x} = M^{(2)}\vec{b}^{(2)} = \vec{b}^{(3)}$$

In general,

$$M^{(k)} = \begin{bmatrix} 1 & 0 & \cdots & 0 \\ \vdots & \ddots & \ddots & \vdots \\ 0 & \cdots & 1 & 0 \\ \vdots & \ddots & \vdots & \ddots \\ 0 & \cdots & -m_{k+1,k} & 1 \\ \vdots & \ddots & \vdots & \vdots \\ 0 & \cdots & -m_{n,k} & \vdots & 1 \end{bmatrix}$$

$$A^{(k+1)} \vec{x} = \vec{b}^{(k+1)}, \text{ or}$$

$$M^{(k)} \cdots M^{(1)} A \vec{x} = M^{(k)} \cdots M^{(1)} \vec{b} \quad (6.9).$$

$A^{(n)}$  will be upper-triangular;  $A^{(n)} = U$

$$A^{(n)} = M^{(n-1)} M^{(n-2)} \cdots M^{(1)} A. \quad (*)$$

(\*) forms the upper-triangular portion of  $A = LU$ .  
Now we will identify the complementary  $L$ .

Examine  $[M^{(k)}]^{-1}$ :

det

$$L^{(k)} = \begin{bmatrix} 1 & 0 & \dots & 0 \\ 0 & \ddots & & \\ \vdots & & \ddots & \\ 0 & 0 & \dots & m_{n,k} \end{bmatrix}$$

Diagram illustrating the structure of  $L^{(k)}$ . The matrix is upper triangular with 1s on the diagonal. The first row has a 1 in the first column and 0s elsewhere. The second row has a 0 in the first column and a 1 in the second column. The third row has a 0 in the first column, a 0 in the second column, and a 1 in the third column. The last row has a 0 in the first column, a 0 in the second column, and a 1 in the last column. The element  $m_{k+1,k}$  is shown in the  $(k+1, k)$  position, and  $m_{n,k}$  is shown in the  $(n, k)$  position.

Compute  $C^{(k)} = M^{(k)} \cdot L^{(k)}$ :

$$M^{(k)} \cdot L^{(k)} = \begin{bmatrix} 1 & 0 & \dots & 0 \\ 0 & \ddots & & \\ \vdots & & \ddots & \\ 0 & 0 & \dots & 1 \end{bmatrix}$$

Diagram illustrating the structure of  $M^{(k)} \cdot L^{(k)}$ . The matrix is upper triangular with 1s on the diagonal. The first row has a 1 in the first column and 0s elsewhere. The second row has a 0 in the first column and a 1 in the second column. The third row has a 0 in the first column, a 0 in the second column, and a 1 in the third column. The last row has a 0 in the first column, a 0 in the second column, and a 1 in the last column.

$$C_{k+p,j}^{(k)} = -m_{k+p,k} \cdot l_{k,j} + 1 \cdot l_{k+p,j} = \begin{cases} 0, & j < k \\ -m_{k+p,k} \cdot 1 + m_{k+p,k} = 0, & j = k \\ -m_{k+p,k} \cdot l_{k,k+p} + 1 \cdot 1 = 1, & j = k+p \\ 0, & j > k+p \end{cases}$$

Thus,  $C^{(k)} = I$  meaning that

$$L^{(k)} = [M^{(k)}]^{-1}$$

Introduce

$$L = L^{(1)} L^{(2)} \dots L^{(n-1)}$$

Then

$$LU = L^{(1)}L^{(2)} \dots L^{(n-1)} \cdot M^{(n-1)}M^{(n-2)} \dots M^{(2)}M^{(1)}A = A.$$

Thm 6.19. If Gaussian elimination can be performed on the linear system  $A\vec{x} = \vec{b}$  without row interchanges, then the matrix  $A$  can be factored into the product of a lower-triangular matrix  $L$  and an upper-triangular matrix  $U$ , that is,  $A = LU$ ,

where  $m_{ji} = \frac{a_{ji}^{(i)}}{a_{ii}^{(i)}}$ ,

$$U = \begin{bmatrix} a_{11}^{(1)} & a_{12}^{(1)} & \dots & a_{1n}^{(1)} \\ 0 & a_{22}^{(2)} & & \\ \vdots & & \ddots & \\ 0 & & & a_{nn}^{(n)} \end{bmatrix}$$

$$\text{and } L = \begin{bmatrix} 1 & 0 & \dots & 0 \\ m_{21} & 1 & & \\ \vdots & & \ddots & \\ m_{n1} & \dots & m_{n,n-1} & 1 \end{bmatrix}$$

Permutation matrices.

We assumed that  $A\vec{x} = \vec{b}$  can be solved by Gaussian elimination without row interchanges.

What if the row interchanges are required? Introduce a class of matrices allowing for rearranging, or permuting rows of a given matrix.

An  $n \times n$  permutation matrix  $P = \{P_{ij}\}$  is a matrix obtained from  $I_n$ , the identity matrix, by rearranging its rows.  $P$  has exactly one nonzero entry in each row and in each column, and

each nonzero entry is 1.

$$P = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix} \quad A = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix}$$

$$PA = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix} =$$

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{31} & a_{32} & a_{33} \\ a_{21} & a_{22} & a_{23} \end{bmatrix}.$$

Multiplying by  $P$  has the effect of interchanging the second and third rows (columns if multiplying on the right) of  $A$ .

Assume  $k_1, \dots, k_n$  is a permutation of the integers  $1, \dots, n$  and  $P = [P_{ij}]$  where

$$P_{ij} = \begin{cases} 1, & j = k_i \\ 0, & \text{otherwise} \end{cases}$$

Then

(1)  $PA$  permutes the rows of  $A$ :

$$PA = \begin{bmatrix} a_{k_1,1} & a_{k_1,2} & \dots & a_{k_1,n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{k_n,1} & a_{k_n,2} & \dots & a_{k_n,n} \end{bmatrix}$$

(2)  $P^{-1}$  exists and  $P^{-1} = P^t$ .

There is a rearrangement of the equations in the system  $A\vec{x} = \vec{b}$  such that no row interchanges are needed, meaning that there exists a permutation matrix  $P$  for which

$$PA\vec{x} = P\vec{b}$$

can be solved without row interchanges. Then

$$PA = LU,$$

as above.

$$A = P^{-1}LU = (P^t L)U.$$

$P^t L$  is not lower triangular unless  $P = I$ .

# Chapter 6.5: Matrix Factorization



## Theorem (6.19)

*If Gaussian elimination can be performed on the linear system  $A\mathbf{x} = \mathbf{b}$  without row interchanges, then the matrix  $A$  can be factored into the product of a lower-triangular matrix  $L$  and an upper-triangular matrix  $U$ , that is,  $A = LU$ , where  $m_{ji} = a_{ji}^{(i)} / a_{ii}^{(i)}$ ,*

$$U = \begin{bmatrix} a_{11}^{(1)} & a_{12}^{(1)} & a_{1n}^{(1)} \\ 0 & a_{22}^{(2)} & \\ & & a_{n-1,n}^{(n-1)} \\ 0 & & 0 & a_{nn}^{(n)} \end{bmatrix}, \text{ and } L = \begin{bmatrix} 1 & 0 & 0 \\ m_{21} & 1 & \\ & & 1 \\ m_{n1} & & m_{n,n-1} & 1 \end{bmatrix}.$$



# Chapter 6.5: Matrix Factorization



## Permutation matrix

An  $n \times n$  **permutation matrix**  $P = [p_{ij}]$  is a matrix obtained by rearranging the rows of  $I_n$ , the identity matrix. This gives a matrix with precisely one nonzero entry in each row and in each column, and each nonzero entry is a 1.

NOTE: Any nonsingular matrix  $A$  can be factored into  $A = P^t L U$ .

Example.

# 1(a). Solve

$$\begin{bmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ -1 & 0 & 1 \end{bmatrix} \begin{bmatrix} 2 & 3 & -1 \\ 0 & -2 & 1 \\ 0 & 0 & 3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 2 \\ -1 \\ 1 \end{bmatrix}$$

$\parallel$                        $\parallel$                        $\parallel$   
 $L$                        $U$                        $\vec{x}$

Let  $U\vec{x} = \vec{y}$

①  $L\vec{y} = \begin{bmatrix} 2 \\ -1 \\ 1 \end{bmatrix}$

$$\begin{bmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ -1 & 0 & 1 \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix} = \begin{bmatrix} 2 \\ -1 \\ 1 \end{bmatrix}$$

Forward-substitution:

$$y_1 = 2$$

$$2y_1 + y_2 = -1$$

$$4 + y_2 = -1$$

$$y_2 = -5$$

$$-y_1 + y_3 = 1$$

$$-2 + y_3 = 1$$

$$y_3 = 3$$

$$\vec{y} = \begin{bmatrix} 2 \\ -5 \\ 3 \end{bmatrix}$$

$$\textcircled{2} \begin{bmatrix} 2 & 3 & -1 \\ 0 & -2 & 1 \\ 0 & 0 & 3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 2 \\ -5 \\ 3 \end{bmatrix}$$

$$3x_3 = 3$$

$$x_3 = 1$$

$$-2x_2 + 1 = -5$$

$$x_2 = \frac{-6}{-2} = 3$$

$$2x_1 + 9 - 1 = 2$$

$$2x_1 = 2 - 8 = -6$$

$$x_1 = -3$$

$$\vec{x} = \begin{bmatrix} -3 \\ 3 \\ 1 \end{bmatrix}.$$

#3(a)

$$A = \begin{bmatrix} 1 & 2 & -1 \\ 2 & 4 & 0 \\ 0 & 1 & -1 \end{bmatrix} \xrightarrow{E_2 - 2E_1 \rightarrow E_2} \begin{bmatrix} 1 & 2 & -1 \\ 0 & 0 & 2 \\ 0 & 1 & -1 \end{bmatrix} \xrightarrow{E_2 \leftrightarrow E_3}$$

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

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

$$PA = LU$$

$$PA = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} 1 & 2 & -1 \\ 2 & 4 & 0 \\ 0 & 1 & -1 \end{bmatrix} = \begin{bmatrix} 1 & 2 & -1 \\ 0 & 1 & -1 \\ 2 & 4 & 0 \end{bmatrix}$$

$$m_3 = 2$$

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

$$A = (P^t L) U = \begin{bmatrix} 1 & 0 & 0 \\ 2 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} 1 & 2 & -1 \\ 0 & 1 & -1 \\ 0 & 0 & 2 \end{bmatrix} = \begin{bmatrix} 1 & 2 & -1 \\ 2 & 4 & 0 \\ 0 & 1 & -1 \end{bmatrix}^{-9-}$$

$$P^t L = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 2 & 0 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 2 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix}$$

# Chapter 6.5: Matrix Factorization



## Algorithm 6.4: LU FACTORIZATION

To factor the  $n \times n$  matrix  $A = [a_{ij}]$  into the product of the lower-triangular matrix  $L = [l_{ij}]$  and the upper-triangular matrix  $U = [u_{ij}]$ ; that is,  $A = LU$ , where the main diagonal of either  $L$  or  $U$  consists of all ones:

INPUT dimension  $n$ ; the entries  $a_{ij}$ ,  $1 \leq i, j \leq n$  of  $A$ ; the diagonal  $l_{11} = \dots = l_{nn} = 1$  of  $L$  or the diagonal  $u_{11} = \dots = u_{nn} = 1$  of  $U$ .

OUTPUT the entries  $l_{ij}$ ,  $1 \leq j \leq i$ ,  $1 \leq i \leq n$  of  $L$  and the entries,  $u_{ij}$ ,  $i \leq j \leq n$ ,  $1 \leq i \leq n$  of  $U$ .

Step 1 Select  $l_{11}$  and  $u_{11}$  satisfying  $l_{11}u_{11} = a_{11}$ .

If  $l_{11}u_{11} = 0$  then OUTPUT ('Factorization impossible');  
STOP.

Step 2 For  $j = 2, \dots, n$  set  $u_{1j} = a_{1j}/l_{11}$ ; (*First row of  $U$ .*)  
 $l_{j1} = a_{j1}/u_{11}$ . (*First column of  $L$ .*)

# Chapter 6.5: TMatrix Factorization



## Algorithm 6.4: LU FACTORIZATION

Step 3 For  $i = 2, \dots, n - 1$  do Steps 4 and 5.

Step 4 Select  $l_{ij}$  and  $u_{ij}$  satisfying  $l_{ij}u_{ij} = a_{ij} - \sum_{k=1}^{i-1} l_{ik}u_{kj}$ .

If  $l_{ij}u_{ij} = 0$  then OUTPUT ('Factorization impossible');  
STOP.

Step 5 For  $j = i + 1, \dots, n$

set  $u_{ij} = \frac{1}{l_{ii}} \left[ a_{ij} - \sum_{k=1}^{i-1} l_{ik}u_{kj} \right];$  (*ith row of U.*)

$l_{ji} = \frac{1}{u_{ii}} \left[ a_{ji} - \sum_{k=1}^{i-1} l_{jk}u_{ki} \right].$  (*ith column of L.*)

Step 6 Select  $l_{nn}$  and  $u_{nn}$  satisfying  $l_{nn}u_{nn} = a_{nn} - \sum_{k=1}^{n-1} l_{nk}u_{kn}$ .

(Note: If  $l_{nn}u_{nn} = 0$ , then  $A = LU$  but  $A$  is singular.)

Step 7 OUTPUT ( $l_{ij}$  for  $j = 1, \dots, i$  and  $i = 1, \dots, n$ );

OUTPUT ( $u_{ij}$  for  $j = i, \dots, n$  and  $i = 1, \dots, n$ );

STOP.