# Gaussian Elimination & LU Decompositions

# Square System of Equations

Consider

$$Ax = b$$

where

$$A = [a_{ij}] \longleftarrow n \times n \text{ matrix}$$

$$b \leftarrow n \times 1$$
 vector.

In expanded form,

$$a_{11}x_1 + \cdots + a_{1n}x_n = b_1$$
  
 $a_{21}x_1 + \cdots + a_{2n}x_n = b_2$   
 $\vdots$   
 $a_{n1}x_1 + \cdots + a_{nn}x_n = b_n$ 

#### Gaussian Elimination



Carl Friedrich Gauss (1777-1855)

A Summary of the Evolution of Gaussian Elimination

# Gaussian Elimination With No Pivoting (GENP)

$$A \longrightarrow A^{(1)} \longrightarrow \cdots \longrightarrow A^{(n-1)} =: U$$
 (upper triangular).  $b \longrightarrow b^{(1)} \longrightarrow \cdots \longrightarrow b^{(n-1)}$ .

# Gaussian Elimination With No Pivoting (GENP)

$$A \longrightarrow A^{(1)} \longrightarrow \cdots \longrightarrow A^{(n-1)} =: U$$
 (upper triangular).  
 $b \longrightarrow b^{(1)} \longrightarrow \cdots \longrightarrow b^{(n-1)}$ .

#### **Step 1:** Create zeros in the first column of A:

$$A \longrightarrow \underbrace{\begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ 0 & a_{22}^{(1)} & \cdots & a_{2n}^{(1)} \\ \vdots & \vdots & \ddots & \vdots \\ 0 & a_{n2}^{(1)} & \cdots & a_{nn}^{(1)} \end{bmatrix}}_{=:A^{(1)}}; b \longrightarrow \underbrace{\begin{bmatrix} b_1 \\ b_2^{(1)} \\ \vdots \\ b_n^{(1)} \end{bmatrix}}_{=:b^{(1)}}$$

where

$$a_{ij}^{(1)} = a_{ij} - \underbrace{\frac{a_{i1}}{a_{11}}}_{=:m:} a_{1j}; \quad b_i^{(1)} = b_i - \frac{a_{i1}}{a_{11}} b_1; \quad i = 2:n, j = 2:n.$$

Here  $a_{11} \leftarrow$  pivot (assumed non zero);  $m_{i1} \leftarrow$  multipliers;



**Step k:** Create zeros in column k of  $A^{(k-1)}$ :

The same operations are performed on  $b^{(k-1)}$ :

$$b^{(k-1)} \longrightarrow \left[egin{array}{c} b_1 \ b_2^{(1)} \ dots \ b_k^{(k-1)} \ b_{k+1}^{(k)} \ dots \ b_n^{(k)} \end{array}
ight] =: b^{(k)}$$

where for i = k + 1 : n, j = k + 1 : n,

$$a_{ij}^{(k)} = a_{ij}^{(k-1)} - \underbrace{\frac{a_{ik}^{(k-1)}}{a_{kk}^{(k-1)}}}_{=:m_{ik}} a_{kj}^{(k-1)}; \quad b_{i}^{(k)} = b_{i}^{(k-1)} - \frac{a_{ik}^{(k-1)}}{a_{kk}^{(k-1)}} b_{k}^{(k-1)};$$

Here  $a_{kk}^{(k-1)} \leftarrow$  pivot (assumed non zero);  $m_{ik} \leftarrow$  multipliers;



**Step** n-1: Create a zero in the (n, n-1) of  $A^{(n-2)}$ :

$$A^{(n-2)} \longrightarrow egin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ & a_{22}^{(1)} & \cdots & a_{2n}^{(1)} \\ & & \ddots & \vdots \\ & & a_{nn}^{(n-1)} \end{bmatrix}; \ b^{(n-2)} \longrightarrow b^{(n-1)};$$
 $=:A^{(n-1)} \ (also \ called \ U)$ 

where assuming pivot  $a_{n-1,n-1}^{(n-2)} \neq 0$  and using multiplier

$$m_{n,n-1} := \frac{a_{n,n-1}^{(n-2)}}{a_{n-1,n-1}^{(n-2)}},$$

$$a_{nn}^{(n-1)} = a_{nn}^{(n-2)} - m_{n,n-1}a_{n-1,n}^{(n-2)}; \quad b_n^{(n-1)} = b_n^{(n-2)} - m_{n,n-1}b_{n-1}^{(n-2)}.$$

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where assuming pivot  $a_{n-1,n-1}^{(n-2)} \neq 0$  and using multiplier

$$\begin{split} m_{n,n-1} &:= \frac{a_{n,n-1}^{(n-2)}}{a_{n-1,n-1}^{(n-2)}}, \\ a_{nn}^{(n-1)} &= a_{nn}^{(n-2)} - m_{n,n-1} a_{n-1,n}^{(n-2)}; \quad b_n^{(n-1)} = b_n^{(n-2)} - m_{n,n-1} b_{n-1}^{(n-2)}. \end{split}$$

- ▶ The system is transformed to  $Ux = b^{(n-1)}$ .
- ▶ The pivots at each step are on the diagonal of *U*!
- ► All steps have to be repeated to solve any new system Ax = c if the multipliers used in the GENP are not saved.



Let

```
L = \begin{bmatrix} 1 & & & & & & & & & & & & \\ m_{21} & 1 & & & & & & & & & \\ m_{31} & m_{32} & \ddots & & & & & & & \\ \vdots & \vdots & & \ddots & & & & & & & \\ m_{k1} & m_{k2} & \cdots & \cdots & 1 & & & & \\ m_{k+1,1} & m_{k+1,2} & \cdots & \cdots & m_{k+1,k} & \ddots & & \\ \vdots & \vdots & & & \vdots & & \vdots & & 1 & \\ m_{n1} & m_{n2} & \cdots & \cdots & m_{nk} & \cdots & m_{n,n-1} & 1 \end{bmatrix}.
```

Let

Then A = LU!

*LU* **decomposition:** A square matrix A is said to have an LU decomposition if there exists a unit lower triangular matrix L and an upper triangular matrix U such that A = LU.

Let

$$L = \begin{bmatrix} 1 & & & & & & & & & & & & \\ m_{21} & 1 & & & & & & & & & \\ m_{31} & m_{32} & \ddots & & & & & & \\ \vdots & \vdots & & \ddots & & & & & & \\ m_{k1} & m_{k2} & \cdots & \cdots & 1 & & & & \\ m_{k+1,1} & m_{k+1,2} & \cdots & \cdots & m_{k+1,k} & \ddots & & \\ \vdots & \vdots & & & \vdots & & \vdots & & 1 \\ m_{n1} & m_{n2} & \cdots & \cdots & m_{nk} & \cdots & m_{n,n-1} & 1 \end{bmatrix}.$$

Then A = LU! This needs a proof!

LU **decomposition:** A square matrix A is said to have an LU decomposition if there exists a unit lower triangular matrix L and an upper triangular matrix U such that A = LU.

**Proof:** In step *k* of GENP

$$A^{(k)} = \underbrace{\begin{bmatrix} 1 & & & & & \\ 0 & \ddots & & & & \\ \vdots & \cdots & 1 & & & \\ 0 & & -m_{k+1,k} & \ddots & & \\ \vdots & & \vdots & & \ddots & \\ 0 & \cdots & -m_{nk} & \cdots & 1 \end{bmatrix}}_{=:M_k} A^{(k-1)}$$

Then

$$U = A^{(n-1)} = M_{n-1}M_{n-2}\cdots M_{k}\cdots M_{2}M_{1}A$$

where  $M_k$ , k = 1, ..., n-1 are the *multiplier* matrices or *Gauss transforms* of Gaussian Elimination.

**Exercise:**  $b^{(n-1)} = M_{n-1}M_{n-2}\cdots M_1b$ .



Note that,

$$U = A^{(n-1)} = M_{n-1} \underbrace{\left(M_{n-2} \cdots \underbrace{\left(M_{k} \cdots \underbrace{\left(M_{2} \underbrace{\left(M_{1} A\right)}_{=A^{(1)}}\right)}\right)}_{=A^{(n-2)}}\right)}_{=A^{(n-2)}}$$

and

$$M_k = I_n - \begin{vmatrix} 0 \\ \vdots \\ 0 \\ m_{k+1,k} \\ \vdots \end{vmatrix} e_k^T, \quad k = 1: n-1,$$

#### Observe that

$$M_k^{-1} = I_n + \begin{bmatrix} 0 \\ \vdots \\ 0 \\ m_{k+1,k} \\ \vdots \\ m_{kn} \end{bmatrix} e_k^T, \quad k = 1: n-1, \text{ (Prove this!)}$$

 $\blacktriangleright \text{ For } i_1 < \dots < i_p,$ 

$$M_{i_1}^{-1}\cdots M_{i_p}^{-1}=I_n+\sum_{i=i_1}^{i_p}\left|egin{array}{c} \vdots \\ 0 \\ m_{i+1,i} \\ \vdots \\ m_{r} \end{array}
ight|e_i^T, ext{ (Prove this!)}$$

So  $U = M_{n-1}M_{n-2}\cdots M_2M_1A$  implies,

$$A = M_1^{-1} M_2^{-1} \cdots M_{n-1}^{-1} U = \begin{pmatrix} I_n + \sum_{k=1}^{n-1} \begin{pmatrix} 0 \\ \vdots \\ 0 \\ m_{k+1,k} \\ \vdots \\ m_{nk} \end{pmatrix} e_k^T \end{pmatrix} U$$

In Step k,

$$A^{(k)} = M_{k}A^{(k-1)}$$

$$= \begin{pmatrix} I_{n} - \begin{bmatrix} 0 \\ \vdots \\ 0 \\ m_{k+1,k} \\ \vdots \\ m_{nk} \end{bmatrix} e_{k}^{T} A^{(k-1)} = A^{(k-1)} - \begin{bmatrix} 0 \\ \vdots \\ 0 \\ m_{k+1,k} \\ \vdots \\ m_{nk} \end{bmatrix} e_{k}^{T}A^{(k-1)}$$

$$= A^{(k-1)} - \begin{bmatrix} 0 \\ \vdots \\ 0 \\ m_{k+1,k} \\ \vdots \\ m_{nk} \end{bmatrix} \begin{bmatrix} 0 & \cdots & 0 & a_{k}^{(k-1)} & \cdots & a_{kn}^{(k-1)} \end{bmatrix}$$

rank one update of  $A^{(k-1)}$ 

Now

$$\begin{bmatrix} 0 \\ \vdots \\ 0 \\ m_{k+1,k} \\ \vdots \\ m_{nk} \end{bmatrix} \begin{bmatrix} 0 & \cdots & 0 & a_{kk}^{(k-1)} & \cdots & a_{kn}^{(k-1)} \end{bmatrix} = \begin{bmatrix} \mathbf{0} & \mathbf{0} \\ \mathbf{0} & \widehat{M}_k \end{bmatrix},$$

where 
$$\widehat{M}_k = \begin{bmatrix} m_{k+1,k} \\ \vdots \\ m_{nk} \end{bmatrix} \begin{bmatrix} a_{kk}^{(k-1)} & \cdots & a_{kn}^{(k-1)} \end{bmatrix}$$
.

Now

$$\begin{bmatrix} 0 \\ \vdots \\ 0 \\ m_{k+1,k} \\ \vdots \\ m_{nk} \end{bmatrix} \begin{bmatrix} 0 & \cdots & 0 & a_{kk}^{(k-1)} & \cdots & a_{kn}^{(k-1)} \end{bmatrix} = \begin{bmatrix} \mathbf{0} & \mathbf{0} \\ \mathbf{0} & \widehat{M}_k \end{bmatrix},$$

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.

As

$$m_{ik}a_{kk}^{(k-1)} = \left(a_{ik}^{(k-1)}/a_{kk}^{(k-1)}\right)a_{kk}^{(k-1)} = a_{ik}^{(k-1)}, \ i = k+1:n,$$

the first column of  $\widehat{M}_k$  is

$$\begin{bmatrix} a_{k+1,k}^{(k-1)} \\ \vdots \\ a_{k-1}^{(k-1)} \end{bmatrix}.$$

Therefore,

$$A^{(k)} = A^{(k-1)} - \begin{bmatrix} \mathbf{0} & \mathbf{0} \\ \mathbf{0} & \widehat{M}_k \end{bmatrix}$$

$$= \left[ \begin{array}{c|c|c} A_{11}^{(k-1)} & A_{12}^{(k-1)} \\ \hline A_{21}^{(k-1)} & A_{22}^{(k-1)} \end{array} \right]$$

$$-\begin{bmatrix} 0 & 0 & 0 & 0 & \cdots & 0 \\ \vdots & \cdots & \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & 0 & \cdots & 0 \\ \hline 0 & 0 & a_{k+1,k}^{(k-1)} & \begin{bmatrix} m_{k+1,k} \\ \vdots \\ m_{nk} \end{bmatrix} \begin{bmatrix} a_{k,k+1}^{(k-1)} & \cdots & a_{kn}^{(k-1)} \end{bmatrix} \end{bmatrix}$$

where

$$A_{11}^{(k-1)} \to k \times k; \quad A_{21}^{(k-1)} \to k \times (n-k);$$
  
 $A_{21}^{(k-1)} \to (n-k) \times k; \quad A_{22}^{(k-1)} \to (n-k) \times (n-k).$ 

As 
$$A_{21}^{(k-1)} = \begin{bmatrix} 0 & 0 & a_{k+1,k}^{(k-1)} \\ \vdots & \cdots & \vdots & \vdots \\ 0 & 0 & a_{nk}^{(k-1)} \end{bmatrix}$$
, therefore,
$$A^{(k)} = \begin{bmatrix} A_{11}^{(k-1)} & A_{12}^{(k-1)} \\ A_{12}^{(k)} & A_{12}^{(k)} \end{bmatrix}$$

where

$$A_{22}^{(k)} = A_{22}^{(k-1)} - \begin{bmatrix} & & & & \\ & \vdots & & \\ & & & \\ & & & & \end{bmatrix} \begin{bmatrix} a_{k,k+1}^{(k-1)} & \cdots & a_{kn}^{(k-1)} \end{bmatrix}$$

$$= \begin{bmatrix} a_{k+1,k+1}^{(k-1)} & \cdots & a_{k+1,n}^{(k-1)} \\ \vdots & \ddots & \vdots \\ a_{n,k+1}^{(k-1)} & \cdots & a_{nn}^{(k-1)} \end{bmatrix} - \begin{bmatrix} a_{k+1,k}^{(k-1)} \\ \vdots \\ a_{n,k}^{(k-1)} \\ \vdots \\ a_{n,k+1}^{(k-1)} & \cdots & a_{nn}^{(k-1)} \end{bmatrix} \begin{bmatrix} a_{k,k+1}^{(k-1)} & \cdots & a_{kn}^{(k-1)} \end{bmatrix}$$

#### Algorithm for GENP/LU

```
for k = 1 : n - 1
      if a_{kk} \neq 0 (multiplier computation begins)
            for i = k + 1 : n
                  a_{ik} = a_{ik}/a_{kk};
            end
      else
                   exit {'zero pivot encountered'}
      end
                  (multiplier computation ends)
      for i = k + 1: n (matrix update begins)
            for j = k + 1 : n
                  a_{ii} = a_{ii} - a_{ik}a_{ki}
            end
      end
                  (matrix update ends)
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```

**Exercise:** Show that the flop count of LU decomposition of an  $n \times n$  matrix is  $\frac{2}{3}n^3 + O(n^2)$  flops.



## Algorithm for GENP/LU with higher level BLAS

```
for k = 1:n-1  \text{if } A(k,k) \neq 0 \qquad \text{(multiplier computation begins)} \\ A(k+1:n,k) = A(k+1:n,k)/A(k,k); \\ \text{else} \\ \text{exit \{'zero pivot encountered'\}} \\ \text{end} \qquad \text{(multiplier computation ends)} \\ \text{(matrix update)} \\ A(k+1:n,k+1:n) = A(k+1:n,k+1:n) - A(k+1:n,k)*A(k,k+1:n); \\ \text{end}
```

Pseudocode for solving  $n \times n$  system Ax = b:

- 1. Find *LU* decomposition of *A*.  $(\frac{2}{3}n^3 + O(n^2) \text{ flops})$
- 2. Solve Ly = b for y.  $(n^2 \text{ flops})$
- 3. Solve Ux = y for x.  $(n^2 \text{ flops})$

**Total flops:**  $\frac{2}{3}n^3 + O(n^2)$  flops.

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First step need NOT be repeated for solving other systems with same *A*.

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But the algorithm does not always work!

**Theorem:** A nonsingular square matrix has an *LU* decomposition if and only if all its leading principal submatrices are nonsingular.

Additionally for such matrices, the *LU* decomposition is unique.



- 1. Checking A for existence of LU decomposition is not possible in practice.
  - (i) Numerically it is only possible to ascertain how close A and its leading principal submatrices are to being singular.
  - (ii) Ascertaining the proximity of *A* and its leading principal submatrices to a singular matrix will cost more flops than finding the *LU* factors.

- 1. Checking *A* for existence of *LU* decomposition is not possible in practice.
- 2. Even if A has an LU decomposition, computing it is a numerically unstable process.
  - Small pivots can lead to large multipliers and result in instability in finite precision arithmetic.

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What is this?

For each k = 1 : n - 1

- 1. Find  $a_{pk}^{(k-1)}$  such that  $|a_{pk}^{(k-1)}| = \max_{k \le j \le n} |a_{jk}^{(k-1)}|$ .
- 2. If  $p \neq k$  interchange rows k and p.
- 3. Perform the usual GE steps to create zeros in column k.



#### **GEPP**

$$A^{(k-1)} = \begin{bmatrix} a_{11} & \cdots & \cdots & a_{1k} & a_{1,k+1} & \cdots & a_{1k} \\ a_{22}^{(1)} & \cdots & a_{2k}^{(1)} & a_{2,k+1}^{(1)} & \cdots & a_{2n}^{(1)} \\ & \ddots & \vdots & \vdots & \cdots & \vdots \\ & & a_{kk}^{(k-1)} & a_{k,k+1}^{(k-1)} & \cdots & a_{kn}^{(k-1)} \\ & & & a_{k+1,k}^{(k-1)} & a_{k+1,k+1}^{(k-1)} & \cdots & a_{k+1,n}^{(k-1)} \\ & & \vdots & & \vdots & & \vdots \\ & & & a_{n,k}^{(k-1)} & a_{n,k+1}^{(k-1)} & \cdots & a_{nn}^{(k-1)} \end{bmatrix}$$

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$$\begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}, \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ 1 & 0 & 0 & 0 \end{bmatrix}.$$

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**Transposition:** A transposition is a permutation matrix obtained by interchanging only two rows or two columns of an identity matrix.

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**Theorem** Given any  $n \times n$  matrix A, there exists a permutation P such that PA has an LU decomposition.

