

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

(Reverse ~~order~~ the row operations to get inverses)

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

So $\begin{bmatrix} 2 & -1 & -1 \\ -1 & 2 & -1 \\ -1 & -1 & 2 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ -1/2 & 1 & 0 \\ -1/2 & -1 & 1 \end{bmatrix} \begin{bmatrix} 2 & -1 & -1 \\ 0 & 3/2 & -3/2 \\ 0 & 0 & 0 \end{bmatrix}$ (check.)

A L U

Last time: Started LU decomposition

$n \times n$ matrix $\rightarrow A = L U$

lower triangular upper triangular

How? A $\xrightarrow{\text{Eliminate lower left variables}}$ U

upper triangular

Same as: multiply A by lower triangular elimination matrices

$A \rightarrow L_1 A \rightarrow L_2 L_1 A \rightarrow \dots \rightarrow L_m \dots L_2 L_1 A = U$

not L yet

Then: $A = L_1^{-1} L_2^{-1} \dots L_m^{-1} U$

This is L!

4x4 Example: $A = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 2 & 3 & 4 \\ 1 & 3 & 6 & 10 \\ 1 & 4 & 10 & 20 \end{bmatrix}$

Eliminate lower left:

Row 2 - Row 1
Row 3 - Row 1
Row 4 - Row 1

$\rightarrow \begin{bmatrix} 1 & 1 & 1 & 1 \\ 0 & 1 & 2 & 3 \\ 0 & 2 & 5 & 9 \\ 0 & 3 & 9 & 19 \end{bmatrix}$

Row 3 - 2Row 2
Row 4 - 3Row 2

$\rightarrow \begin{bmatrix} 1 & 1 & 1 & 1 \\ 0 & 1 & 2 & 3 \\ 0 & 0 & 1 & 3 \\ 0 & 0 & 3 & 10 \end{bmatrix}$

or: multiply by $\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & -2 & 1 & 0 \\ 0 & -3 & 0 & 1 \end{bmatrix}$

Row - 3Row 3

$\rightarrow \begin{bmatrix} 1 & 1 & 1 & 1 \\ 0 & 1 & 2 & 3 \\ 0 & 0 & 1 & 3 \\ 0 & 0 & 0 & 1 \end{bmatrix}$

or: multiply by $\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & -3 & 1 \end{bmatrix}$

or: multiply by $\begin{bmatrix} 1 & 0 & 0 & 0 \\ -1 & 1 & 0 & 0 \\ -1 & 0 & 1 & 0 \\ -1 & 0 & 0 & 1 \end{bmatrix}$

Upper triangular; this is U

$$A \rightarrow \begin{bmatrix} 1 & 0 & 0 & 0 \\ -1 & 1 & 0 & 0 \\ -1 & 0 & 1 & 0 \\ -1 & 0 & 0 & 1 \end{bmatrix} A \rightarrow \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & -2 & 1 & 0 \\ 0 & -3 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ -1 & 1 & 0 & 0 \\ -1 & 0 & 1 & 0 \\ -1 & 0 & 0 & 1 \end{bmatrix} A \rightarrow$$

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

Not L yet; this is L^{-1}

Solve for A : $A = \left(\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & -3 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & -2 & 1 & 0 \\ 0 & -3 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ -1 & 1 & 0 & 0 \\ -1 & 0 & 1 & 0 \\ -1 & 0 & 0 & 1 \end{bmatrix} \right)^{-1} U$

Switch order when you invert a product \rightarrow $\begin{bmatrix} 1 & 0 & 0 & 0 \\ -1 & 1 & 0 & 0 \\ -1 & 0 & 1 & 0 \\ -1 & 0 & 0 & 1 \end{bmatrix}^{-1} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & -2 & 1 & 0 \\ 0 & -3 & 0 & 1 \end{bmatrix}^{-1} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & -3 & 1 \end{bmatrix}^{-1}$

Reverse the row operations \rightarrow $\begin{bmatrix} 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 \\ 1 & 0 & 1 & 0 \\ 1 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 2 & 1 & 0 \\ 0 & 3 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 3 & 1 \end{bmatrix}$

$$\begin{bmatrix} 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 \\ 1 & 2 & 1 & 0 \\ 1 & 3 & 0 & 1 \end{bmatrix} \Leftarrow \begin{bmatrix} 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 \\ 1 & 2 & 1 & 0 \\ 1 & 3 & 3 & 1 \end{bmatrix} \leftarrow \text{This is } L!$$

Conclusion: $\begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 2 & 3 & 4 \\ 1 & 3 & 6 & 10 \\ 1 & 4 & 10 & 20 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 \\ 1 & 2 & 1 & 0 \\ 1 & 3 & 3 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 0 & 1 & 2 & 3 \\ 0 & 0 & 1 & 3 \\ 0 & 0 & 0 & 1 \end{bmatrix}$

$A \quad \quad \quad L \quad \quad \quad U$

LU decomposition works whenever you can do elimination

$A \rightarrow U$ without switching rows. How to find the entries of L more efficiently?

Diagonal entries of L : all 1's

Below-diagonal entries: Put l_{ij} in row i , column j if you did the operation $\text{Row } i \rightarrow \text{Row } i - l_{ij} \text{ Row } j$

Example: $A = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 2 & 3 \\ 1 & 4 & 9 \end{bmatrix}$ $\xrightarrow[\text{Row 3} - \boxed{1} \text{ Row 1}]{\text{Row 2} - \boxed{1} \text{ Row 1}} \begin{bmatrix} 1 & 1 & 1 \\ 0 & 1 & 2 \\ 0 & 3 & 8 \end{bmatrix} \xrightarrow{\text{Row 3} - \boxed{3} \text{ Row 2}} \begin{bmatrix} 1 & 1 & 1 \\ 0 & 1 & 2 \\ 0 & 0 & 2 \end{bmatrix}$

l_{21} l_{31} l_{32}

$$\begin{bmatrix} 1 & 1 & 1 \\ 0 & 1 & 2 \\ 0 & 0 & 2 \end{bmatrix} = U \quad L = \begin{bmatrix} 1 & 0 & 0 \\ 1 & 1 & 0 \\ 1 & 3 & 1 \end{bmatrix}$$

check that this is correct: $LU = \begin{bmatrix} 1 & 0 & 0 \\ 1 & 1 & 0 \\ 1 & 3 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 \\ 0 & 1 & 2 \\ 0 & 0 & 2 \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 2 & 3 \\ 1 & 4 & 9 \end{bmatrix} = A$ ✓

Summarize LU decomposition algorithm:

$A \xrightarrow{\text{lower left elimination}} U \rightsquigarrow A = L U$ ← upper triangular

↑
lower triangular

$$L = \begin{bmatrix} 1 & 0 & \dots & 0 \\ & 1 & \dots & 0 \\ & & \ddots & \vdots \\ & & & 1 \end{bmatrix}$$

l_{ij}

Comes from operation
 $\text{Row } i \rightarrow \text{Row } i - l_{ij} \text{ Row } j$
 note the minus sign

Warning 1: Formula for L only works if you do the elimination operations in the right order:

Eliminate Column 1 variables using Row 1 first,
 Eliminate Column 2 variables using Row 2 next,
 ⋮

Warning 2: LU decomposition only works if you can go from A to U without switching rows.

Now: What is the use of LU decomposition?

It turns one hard system of equations $A\vec{x} = \vec{b}$ into two easy triangular systems:

$$A\vec{x} = \vec{b} \rightsquigarrow \underbrace{LU\vec{x} = \vec{b}}_{\text{call this vector } \vec{y}}$$

First solve $L\vec{y} = \vec{b}$ for \vec{y} , then solve $U\vec{x} = \vec{y}$ for \vec{x}
same \vec{y}

check that this works: Does $A\vec{x}$ really equal \vec{b} ?

$$A\vec{x} = (LU)\vec{x} = L(\underbrace{U\vec{x}}_{\text{This is } \vec{y}}) = \underbrace{L\vec{y}}_{\text{This is } \vec{b}} = \vec{b} \quad \checkmark$$

Example solve

$$\begin{cases} x_1 + x_2 + x_3 + x_4 = 1 \\ x_1 + 2x_2 + 3x_3 + 4x_4 = 0 \\ x_1 + 3x_2 + 6x_3 + 10x_4 = -1 \\ x_1 + 4x_2 + 10x_3 + 20x_4 = 0 \end{cases}$$

Already know: $A = \underbrace{\begin{bmatrix} 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 \\ 1 & 2 & 1 & 0 \\ 1 & 3 & 3 & 1 \end{bmatrix}}_L \underbrace{\begin{bmatrix} 1 & 1 & 1 & 1 \\ 0 & 1 & 2 & 3 \\ 0 & 0 & 1 & 3 \\ 0 & 0 & 0 & 1 \end{bmatrix}}_U$

First solve $L\vec{y} = \vec{b}$:

$$\begin{aligned} y_1 &= 1 \\ y_1 + y_2 &= 0 \rightarrow y_2 = -1 \\ y_1 + 2y_2 + y_3 &= -1 \rightarrow y_3 = -1 - 2(-1) = 0 \\ y_1 + 3y_2 + 3y_3 + y_4 &= 0 \rightarrow y_4 = -1 - 3(-1) = 2 \end{aligned}$$

Now solve $U\vec{x} = \vec{y}$:

$$\begin{aligned} x_1 + x_2 + x_3 + x_4 &= 1 \rightarrow 1 - 5 - (-6) - 2 = 0 \\ x_2 + 2x_3 + 3x_4 &= -1 \rightarrow x_2 = -1 - 2(-6) - 3(2) = 5 \\ x_3 + 3x_4 &= 0 \rightarrow x_3 = -3(2) = -6 \\ x_4 &= 2 \end{aligned}$$

so $\vec{x} = \begin{bmatrix} 0 \\ 5 \\ -6 \\ 2 \end{bmatrix}$

Note: The two triangular systems are easy to solve, but we (52) still have to work hard to find L and U.

Solving systems using LU works best when you have a lot of systems with the same A:

$$A\vec{x}_1 = \vec{b}_1, A\vec{x}_2 = \vec{b}_2, A\vec{x}_3 = \vec{b}_3, \dots, A\vec{x}_{1000} = \vec{b}_{1000}$$

Even though you have 1000 systems, you only need to find LU once.

Some statistics from the textbook: (if A is $n \times n$)

To find LU, it usually takes about $\frac{2}{3}n^3$ arithmetic operations
 \nearrow multiplication and addition

To solve a triangular system usually takes about $2n^2$ arithmetic operations

So what if you solve 1000 systems individually?

$$\underbrace{A \xrightarrow{\frac{2}{3}n^3} U \xrightarrow{2n^2} \text{solution}}_{1000 \text{ times}} \longrightarrow 2000 \left(\frac{n^3}{3} + n^2 \right) \text{ operations}$$

$$\begin{aligned} \text{Using LU: } A &\xrightarrow{\text{once}} U \xrightarrow{\frac{2}{3}n^3} \text{solution} \\ \underbrace{\text{Two triangular systems}}_{1000 \text{ times}} &\xrightarrow{4n^2} \frac{2}{3}n^3 + 4000n^2 \text{ operations} \end{aligned}$$

If n is big, then n^2 is much less than n^3 , so solving by LU is almost 1000 times faster:

$$\frac{2000 \left(\frac{n^3}{3} + n^2 \right)}{\frac{2}{3}n^3 + 4000n^2} = \frac{\frac{2000n^3}{3} \left(1 + \frac{3}{n} \right)}{\frac{2}{3}n^3 \left(1 + \frac{6000}{n} \right)} = 1000 \cdot \underbrace{\frac{1 + 3/n}{1 + 6000/n}}_{\text{Goes to 1 as } n \rightarrow \infty}$$

Transposes and Permutation

One final matrix operation: transpose

↑
switch rows and columns of a matrix

$m \times n$ matrix A $\xrightarrow[\text{into columns}]{\text{turn rows}}$ $n \times m$ matrix A^T

$$A = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{bmatrix} \longrightarrow A^T = \begin{bmatrix} 1 & 4 \\ 2 & 5 \\ 3 & 6 \end{bmatrix}$$

(i, j) -entry of $A^T = (j, i)$ -entry of A

$$\begin{array}{c} \text{jth row} \rightarrow \left[\begin{array}{c|c} & A_{ji} \\ \hline & \end{array} \right] \xrightarrow{\quad} A^T = \left[\begin{array}{c|c} & (A^T)_{ij} \\ \hline & \end{array} \right] \leftarrow \begin{array}{l} \text{ith row} \\ \text{now} \end{array} \\ \uparrow \\ \text{ith column} \end{array}$$

jth column now

Another viewpoint: "Flip A over the diagonal" to get $A^T =$

$$A = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{bmatrix} \longrightarrow A^T = \begin{bmatrix} 1 & 4 & 7 \\ 2 & 5 & 8 \\ 3 & 6 & 9 \end{bmatrix}$$

Rules $(A+B)^T = A^T + B^T$

$$(AB)^T = ??$$

$$(A^{-1})^T = ??$$

$$(A+B)^T_{ij} = (A+B)_{ji}$$

$$= A_{ji} + B_{ji}$$

$$= A^T_{ij} + B^T_{ij} = (A^T + B^T)_{ij}$$

(i, j) -entry of $(AB)^T = (j, i)$ -entry of AB

$$= (\text{jth row of } A) \cdot (\text{ith column of } B)$$

$$= (\text{jth column of } A^T) \cdot (\text{ith row of } B^T)$$

$$= (\text{ith row of } B^T) \cdot (\text{jth column of } A^T)$$

$$= (i, j)\text{-entry of } B^T A^T$$

so $(AB)^T = B^T A^T$ (switch order, just like for inverse)

$$A^{-1}A = I \longrightarrow (A^{-1}A)^T = I^T \longrightarrow A^T(A^{-1})^T = I$$

$(A^{-1})^T$ looks like inverse of A^T .

So: $(A^{-1})^T = (A^T)^{-1}$ This means that A^T is invertible if A is, and A^T isn't invertible if A isn't.

Example $A = \begin{bmatrix} 1 & 3 \\ 1 & 4 \end{bmatrix} \xrightarrow{\text{Inverse}} A^{-1} = \frac{1}{\underbrace{(1)(4) - (1)(3)}} \begin{bmatrix} 4 & -3 \\ -1 & 1 \end{bmatrix}$

\downarrow Transpose

\downarrow Transpose

$$A^T = \begin{bmatrix} 1 & 1 \\ 3 & 4 \end{bmatrix} \xrightarrow{\text{Inverse}} (A^{-1})^T = \begin{bmatrix} 4 & -1 \\ -3 & 1 \end{bmatrix} = (A^{-1})^T$$