

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

(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:

$$A \xrightarrow{\text{lower left elimination}} U \quad \rightsquigarrow \quad A = L U$$

$\uparrow$  lower triangular       $\leftarrow$  upper triangular

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

$\boxed{l_{ij}}$

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:

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

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?

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

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

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  $A$

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

$\text{jth column now}$

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

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