

$$\begin{aligned}
 A &= LU \\
 &= LDU'
 \end{aligned}$$

square \swarrow A
 L : lower triangular 1's on diag
 U : upper triangular
 D : diagonal
 U' : upper triangular 1's on diag
 Gaussian elimination

$$U = DU'$$

• This only holds for A 's whose Gaussian elimination does not involve row swaps

• If A does require row swaps, then we get the factorization

$$PA = LU$$

permutation matrix \swarrow P
 as before \underbrace{LU}

Permutation matrices : a $n \times n$ matrix with lots of 0's and a single 1 on each row & on each column

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

$\sigma(1) \quad \sigma(3)$
 $\{1, 2, 3\}$
 $\sigma(2)$

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

every entry in this 2×2 is 0

$\{2, 1, 4, 3\}$
 $\sigma(1) \quad \sigma(2) \quad \sigma(3) \quad \sigma(4)$

let $\sigma(i)$ be the index of the column on which the 1 on row i lies, for all $i \in \{1, \dots, n\}$
 $\{\sigma(1), \dots, \sigma(n)\}$ is a permutation of $\{1, \dots, n\}$

there are $n! = 1 \cdot 2 \cdot \dots \cdot n$ such permutations

there $n!$ permutation matrices of size $n \times n$

P_{ij} from our second class correspond to permutation

$$\{1, 2, \dots, i-1, j, i+1, \dots, j-1, i, j+1, \dots, n\}$$

$$A = \begin{bmatrix} 0 & 0 & 3 \\ 1 & 1 & 2 \\ 1 & 2 & -1 \end{bmatrix} \xrightarrow{\text{swap } \pi_1 \text{ and } \pi_3} \begin{bmatrix} 1 & 2 & -1 \\ 1 & 1 & 2 \\ 0 & 0 & 3 \end{bmatrix}$$

$$\xrightarrow{\pi_2 - \pi_1} \begin{bmatrix} 1 & 2 & -1 \\ 0 & -1 & 3 \\ 0 & 0 & 3 \end{bmatrix} = U$$

in general

$$PA = LU$$

for a suitably chosen permutation matrix P

$$E_{21}^{(-1)} P_{13} A = U \quad \bigg| \quad (E_{21}^{(-1)})^{-1}$$

$$P_{13} A = \underbrace{E_{21}^{(1)}}_L \cdot U = \begin{bmatrix} 1 & & \\ 1 & 1 & \\ 0 & 0 & 1 \end{bmatrix} \cdot \begin{bmatrix} 1 & 2 & -1 \\ 0 & -1 & 3 \\ 0 & 0 & 3 \end{bmatrix}$$

To get $PA = LU$

you have to do all row swaps before all the row eliminations

(if you do all your row swaps after row eliminations, then you get a
 $A = LPU$ factorization)

how do you know which row swaps to do

- stare at the matrix and guess
- do Gaussian elimination in the order from class

and record all row swaps that you do along the way

then come back to the original matrix, do the same row swaps first (before any elimination) and then do elimination on resulting matrix

(add multiples of row 1 to lower rows
then swap
add multiples of row 2 to lower rows
then swap)

Take A rectangular ($m \times n$)

Def: the **transpose** of A , denoted by A^T , is the $n \times m$ matrix obtained by switching the roles of rows and columns

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

Properties:

$$\bullet (A^T)^T = A$$

$$\bullet (A + B)^T = A^T + B^T$$

$$\bullet (AB)^T = B^T A^T$$

$m \times n$ $n \times p$ $p \times n$ $n \times m$

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

if A is square and invertible

$$\underline{V} = \begin{bmatrix} v_1 \\ \vdots \\ v_n \end{bmatrix} \cdot \begin{bmatrix} w_1 \\ \vdots \\ w_n \end{bmatrix} = v_1 w_1 + \dots + v_n w_n$$

$$\underline{V}^T = \begin{bmatrix} v_1 & \dots & v_n \end{bmatrix} \begin{bmatrix} w_1 \\ \vdots \\ w_n \end{bmatrix} = \begin{bmatrix} v_1 w_1 + \dots + v_n w_n \end{bmatrix}$$

$\underbrace{\hspace{1cm}}_{1 \times n} \quad \underbrace{\hspace{1cm}}_{n \times 1}$

Moral:

$$V \cdot W = \underbrace{V^T W}_{1 \times 1 \text{ matrix}}$$

number

compose

$$\begin{matrix} & \swarrow & \searrow \\ V & W^T & \\ n \times 1 & 1 \times n & \end{matrix} = n \times n \text{ matrix}$$

important for SVD

Definition: a square matrix S is called symmetric if $S = S^T$

$$\begin{bmatrix} 4 & 3 & 6 \\ 3 & 7 & -1 \\ 6 & -1 & 5 \end{bmatrix} = \begin{bmatrix} 4 & 3 & 6 \\ 3 & 7 & -1 \\ 6 & -1 & 5 \end{bmatrix}$$

$$\begin{bmatrix} 4 & 3 & 6 \\ 3 & 7 & -1 \\ 6 & -1 & 5 \end{bmatrix}$$

Theorem: for any matrix A (even $m \times n$),
the matrix $S = A^T A$ is symmetric

Proof: $S^T = (\underline{A^T} \underline{A})^T = \underline{A}^T (\underline{A^T})^T = A^T A = S$

Def: an antisymmetric matrix A is one such that $A = -A^T$

Theorem: The LDU factorization of a symmetric matrix S takes the form

$$L = \begin{bmatrix} 1 & 0 & 0 \\ * & 1 & 0 \\ * & * & 1 \end{bmatrix} \Rightarrow L^T = \begin{bmatrix} 1 & * & * \\ 0 & 1 & * \\ 0 & 0 & 1 \end{bmatrix}$$

$$S = L D L^T$$

lower triang
is on diag

diag

upper triang
is on diag

Proof (for S is invertible \rightsquigarrow unique LDU factorization)

take $S = LDU \Rightarrow S^T = (LDU)^T = U^T D^T L^T \Rightarrow \begin{matrix} L = U^T \\ U = L^T \end{matrix}$

lower triang $\parallel D$ upper triang

example: $O = L \cdot O \cdot U$ for any L & U

zero matrix

\Downarrow
 LDU is not unique for singular matrices