

Last time:

$$\det(A) = \begin{cases} 0 & , A \text{ is not invertible} \\ (-1)^k \cdot d_1 \cdot \dots \cdot d_n & , \text{otherwise} \end{cases}$$

where  $PA = L \cdot \text{diag}(d_1, \dots, d_n) \cdot U$

$\nwarrow$   $k$ : how many row swaps  
P requires,

$$\det(AB) = \det(A) \cdot \det(B)$$

$$\det(A^T) = \det(A).$$

How to compute.

$$\textcircled{1} \det A = \overset{(*)}{a_{11}} \cdot C_{11} + a_{12} C_{12} + \dots + a_{1n} C_{1n}$$

where  $C_{1k} = (-1)^{1+k} \cdot \begin{vmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & \dots & \dots & a_{nn} \end{vmatrix}$

delete 1st row & kth column

$(n-1) \times (n-1)$

called  
the  $(1, k)$   
Cofactor

We say  $(*)$  computes  $\det$  using  
the first row.

$\det A = a_{m1} C_{m1} + a_{m2} C_{m2} + \dots + a_{mn} C_{mn}$   
 computes  $\det$  using the  $m^{\text{th}}$  row

$\det A = a_{1k} C_{1k} + a_{2k} C_{2k} + \dots + a_{nk} C_{nk}$   
 computes using the  $k^{\text{th}}$  column

Ex:  $A = \begin{pmatrix} 2 & 3 & 4 \\ 1 & 0 & 1 \\ 7 & 0 & 5 \end{pmatrix}$

$\det A = 2 \cdot C_{11} + 3 \cdot C_{12} + 4 \cdot C_{13}$

$1^{\text{st}} \text{ row} = 2 \cdot (-1)^{1+1} \begin{vmatrix} 0 & 1 \\ 0 & 5 \end{vmatrix} + 3 \cdot (-1)^{1+2} \begin{vmatrix} 1 & 1 \\ 7 & 5 \end{vmatrix} + 4 \cdot (-1)^{1+3} \begin{vmatrix} 1 & 0 \\ 7 & 0 \end{vmatrix}$

$= 2 \cdot 0 - 3 \cdot (5 - 7) + 4 \cdot 0 = 6$

$\det A = -3 \cdot \begin{vmatrix} 1 & 1 \\ 7 & 5 \end{vmatrix} + 0 \cdot * + 0 \cdot *$   
 $2^{\text{nd}} \text{ column} = 6$

ex:  $B = \begin{bmatrix} 1 & 0 & 1 & 0 \\ 2 & 1 & 1 & 0 \\ 3 & 2 & 0 & 1 \\ 4 & 3 & -1 & 1 \end{bmatrix}$

$\det B = 0 \cdot C_{14} + 0 \cdot C_{24} + 1 \cdot C_{34} + 1 \cdot C_{44}$   
 wrt 4th column  $= (-1)^7 \cdot \begin{vmatrix} 1 & 0 & 1 \\ 2 & 1 & 1 \\ 4 & 3 & -1 \end{vmatrix} + (-1)^8 \cdot \begin{vmatrix} 1 & 0 & 1 \\ 2 & 1 & 1 \\ 3 & 2 & 0 \end{vmatrix}$

$\begin{vmatrix} 1 & 0 & 1 \\ 1 & 1 & 0 \\ 5 & 3 & 0 \end{vmatrix} = 3 - 5 = -2$

②  $A \cdot A^{-1} = I.$

$1 = \det I = \det(AA^{-1}) = \det A \cdot \det A^{-1}$   
 $\Rightarrow \det(A^{-1}) = 1/\det A.$

③  $Q$  is orthogonal, i.e.  $Q^T Q = I.$

Then  $1 = \det Q \cdot \det Q^T = (\det Q)^2$

$\Rightarrow \det Q = \pm 1.$

$\begin{vmatrix} 1 & 0 \\ 0 & 1 \end{vmatrix} = 1, \quad \begin{vmatrix} 0 & 1 \\ 1 & 0 \end{vmatrix} = -1.$

④ Let  $A_{n \times n} = Q \cdot R$  ↗ upper  
↘ orthog.  
 $(a_1 \vdots \dots \vdots a_n)$   $Q = (p_1 \vdots \dots \vdots p_n)$

$$\det A = (\pm 1) \cdot \det R = \pm (p_1^T a_1) \cdot \dots \cdot (p_n^T a_n)$$

Applications.

① Computing  $A^{-1}$ .

$$\det A = a_{11} \cdot C_{11} + a_{12} C_{12} + \dots + a_{1n} C_{1n}$$

$$a_{11} \cdot C_{21} + a_{12} C_{22} + \dots + a_{1n} C_{2n}$$

$$\sum_{j=1}^n \begin{vmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ a_{31} & a_{32} & \dots & a_{3n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{vmatrix} = 0$$

↖  
wrt  
1st row

Therefore

$$\begin{matrix} A & C^T \\ \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{pmatrix} & \begin{pmatrix} C_{11} & C_{21} & \dots & C_{n1} \\ C_{12} & C_{22} & \dots & C_{n2} \\ \vdots & \vdots & \ddots & \vdots \\ C_{1n} & C_{2n} & \dots & C_{nn} \end{pmatrix} \end{matrix}$$

$$= \begin{pmatrix} \det A & 0 & \dots & 0 \\ 0 & \det A & \dots & 0 \\ \vdots & 0 & \ddots & 0 \\ 0 & 0 & \dots & \det A \end{pmatrix}$$

$$= \det A \cdot I$$

Hence if  $\det A \neq 0$

$$A \cdot \left( \frac{1}{\det A} \cdot C^T \right) = I \quad \text{so that}$$

$$A^{-1} = C^T / \det A$$

Here  $C$  is the cofactors matrix,  
with  $ij$  entry  $= C_{ij}$ .

ex:  $A = \begin{pmatrix} 1 & 0 & 1 \\ 2 & 1 & 1 \\ 3 & 2 & 0 \end{pmatrix}$   $C_{11} = -2, C_{12} = +3,$   
 $C_{13} = +1, C_{21} = +2$

$$C_{22} = -3, C_{23} = -2, C_{31} = -1,$$

$$C_{32} = +1, C_{33} = +1.$$

cofactors matrix  $C = \begin{pmatrix} -2 & 3 & 1 \\ 2 & -3 & -2 \\ -1 & 1 & 1 \end{pmatrix}$

$$\det A = \begin{vmatrix} 1 & 0 & 1 \\ 1 & 1 & 0 \\ 3 & 2 & 0 \end{vmatrix} = -1 \neq 0$$

Hence

$$A^{-1} = (-1) \cdot \begin{pmatrix} -2 & 2 & -1 \\ 3 & -3 & 1 \\ 1 & -2 & 1 \end{pmatrix} = \frac{C^T}{\det A}$$

## [2] CRAMER'S RULE.

We want to solve  $Ax=b$  with  $A$  invertible:

$$x = A^{-1}b = \frac{1}{\det A} \begin{pmatrix} C_{11} & C_{21} & C_{n1} \\ C_{12} & C_{22} & C_{n2} \\ & \ddots & \\ & & C_{nn} \end{pmatrix} \begin{pmatrix} b_1 \\ \vdots \\ b_n \end{pmatrix}$$

Thus

$$x_1 = \frac{1}{\det A} (b_1 C_{11} + b_2 C_{21} + \dots + b_n C_{n1})$$

$$= \frac{1}{\det A} \cdot \begin{vmatrix} b_1 & a_{12} & \dots & a_{1n} \\ b_2 & a_{22} & \dots & a_{2n} \\ \vdots & \ddots & & \\ b_n & a_{n2} & \dots & a_{nn} \end{vmatrix}$$

Similarly

CRAMER'S RULE

$$x_j = \frac{1}{\det A} \cdot \det \left( \begin{array}{c} \text{matrix obtained} \\ \text{from } A \text{ by deleting} \\ \text{its } j\text{th column \& instead} \\ \text{writing the vector } b \end{array} \right)$$

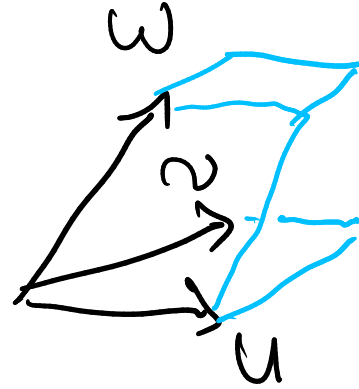
[3] area of a parallelogram

$$= \|u \times v\| = |\det(u : v)|$$



volume of a parallelepiped

$$= u \cdot (v \times w) = |\det[u : v : w]|$$



volume of a solid in  $\mathbb{R}^n$  spanned by  $u_1, \dots, u_n$  is defined to be

$$|u_1 : \dots : u_n|$$

Observe that if  $u_1, \dots, u_n$  are lin. dependent then the volume = 0.



$$\boxed{4} \left| \begin{array}{cc|c} a_{11} & a_{12} & \\ a_{21} & a_{22} & \\ & & \ddots \\ & & a_{nn} \end{array} \right| = \left| \begin{array}{cc|c} a_{11} & a_{12} & \\ 0 & * & \\ & & \ddots \\ & & a_{nn} \end{array} \right|$$

(Assume no permutation needed to reach the echelon form.)

The determinants of the 2 top left  $2 \times 2$  matrices are equal too.

This shows that  $A$  is invertible

$\Leftrightarrow$  every  $k \times k$  top left submatrix is invertible

I.e.

$\det A \neq 0 \Leftrightarrow$  every  $k \times k$  top left submatrix has  $\det \neq 0$ .

# EIGENVALUES & EIGENVECTORS

defn: For  $A_{n \times n}$  square, a nonzero vector  $v \in \mathbb{R}^n$  & a  $\lambda \in \mathbb{R}$  satisfying

$$Av = \lambda v$$

↓  
to be  
modified  
later.

are called an eigenvector  $v$  corresponding to eigenvalue  $\lambda$ .

ex: •  $A = I_{n \times n}$ . Every  $v \in \mathbb{R}^n$  is an evector & there is just one eval: 1.  $I \cdot v = 1 \cdot v$

• For any rotation of  $\mathbb{R}^3$  with a rotation axis  $\text{span}(u)$ .

Any  $cu$  is an evector with the corresponding eval 1.

• If  $Av = \lambda v$ ,  $v \neq 0$  then  
for any  $c \in \mathbb{R} - \{0\}$ ,  $A(cv) = \lambda(cv)$   
i.e. if  $v$  is evector corresp to  $\lambda$   
then  $\forall cv$  is " "  $\lambda$ .

• If  $u$  &  $v$  are evector corresp  
to  $\lambda$  then

$$\begin{aligned} A(cu + v) &= cAu + Av \\ &= c\lambda u + \lambda v \\ &= \lambda(cu + v) \end{aligned}$$

i.e.  $cu + v$  is also an evector  
corresp to  $\lambda$ .

thm: Let  $\lambda$  be an eval for  $A$

Then the set of all eectors corr.  
to  $\lambda$  is a subspace of  $\mathbb{R}^n$ .

We call that the eigenspace  
corresponding to  $\lambda$ , denoted by  $E_\lambda$ .