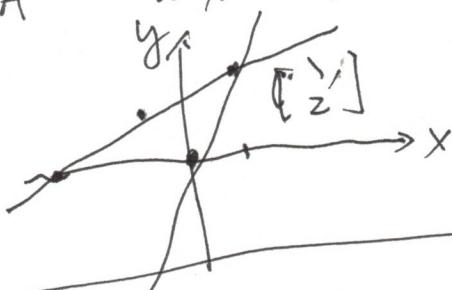


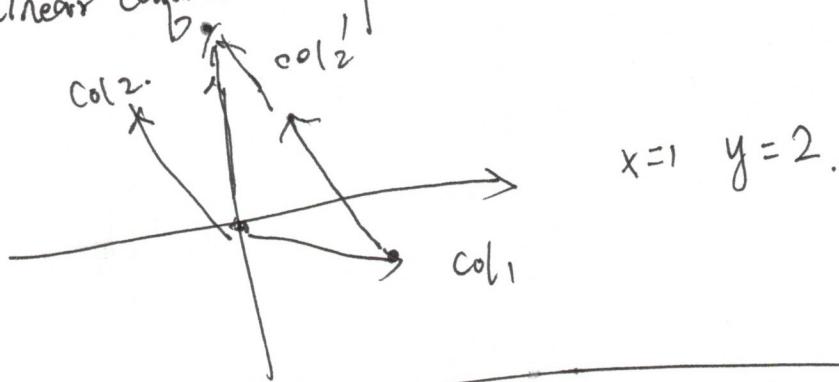
ecture 1

$$A = \begin{bmatrix} 2 & -1 \\ 1 & 2 \end{bmatrix}, \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 0 \\ 3 \end{bmatrix} \rightarrow \boxed{\text{Row picture}}$$



$$x \begin{bmatrix} 2 \\ -1 \end{bmatrix} + y \begin{bmatrix} -1 \\ 2 \end{bmatrix} = \begin{bmatrix} 0 \\ 3 \end{bmatrix} \rightarrow \boxed{\text{Column picture}}$$

linear combination of columns



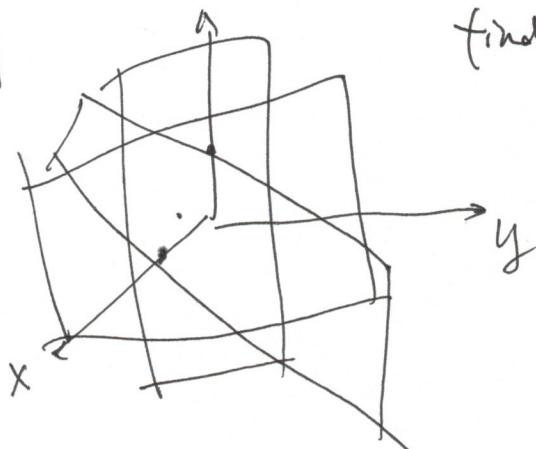
3x3 example:

$$\begin{aligned} 2x - y &= 0 \\ x + 2y - z &= 1 \\ -3y + 4z &= 4 \end{aligned}$$

$$A = \begin{bmatrix} 2 & -1 & 0 \\ 1 & 2 & -1 \\ 0 & -3 & 4 \end{bmatrix}$$

$$b = \begin{bmatrix} 0 \\ 1 \\ 4 \end{bmatrix}$$

Row Picture:



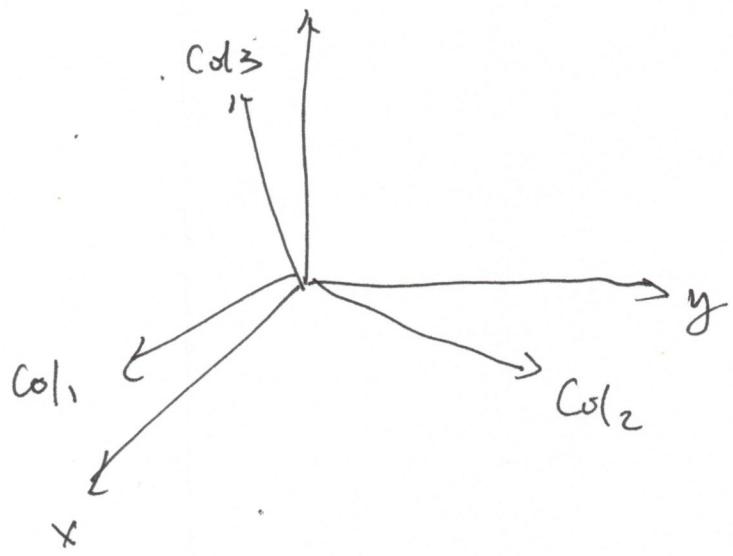
find the point where 3 plane meet.

Column Picture

$$x \begin{bmatrix} 2 \\ -1 \\ 0 \end{bmatrix} + y \begin{bmatrix} -1 \\ 2 \\ -3 \end{bmatrix} + z \begin{bmatrix} 0 \\ -1 \\ 4 \end{bmatrix} = \begin{bmatrix} 0 \\ -1 \\ 4 \end{bmatrix}$$

$$b = \begin{bmatrix} 1 \\ 1 \\ -3 \end{bmatrix}$$

$$x=1 \quad y=1 \quad z=0$$



$$x = 0$$

$$y = 0$$

$$z = 1$$

Can I solve  $Ax = b$  ? for every  $b$ .

or do the linear combination of all columns fill the  $\mathbb{R}^3$  space.

Yes,

Intuitively, if  $Col_1/Col_2/Col_3$  are on the same plane. It is singular / non-invertible.  
(not independent)

---

$Ax = b$  is a combination of columns of A

## Lecture 2: elimination expressed as matrix

elimination

$$\begin{bmatrix} 1 & 2 & | & 1 \\ 3 & 8 & | & 1 \\ 0 & 4 & | & 1 \end{bmatrix} \quad \begin{bmatrix} 2 \\ 12 \\ 2 \end{bmatrix}$$

$$Ax = b$$

pivot

$$\rightarrow \begin{array}{ccc|cc} 1 & 2 & | & 2 & 2 \\ 3 & 8 & | & 12 & 6 \\ 0 & 4 & | & 2 & -10 \end{array} \rightarrow \begin{array}{ccc|cc} 1 & 2 & | & 2 & 2 \\ 0 & 2 & | & 6 & 6 \\ 0 & 4 & | & 2 & -10 \end{array} \rightarrow \begin{array}{ccc|cc} 1 & 2 & | & 2 & 2 \\ 0 & 2 & | & 6 & 6 \\ 0 & 0 & | & 5 & -10 \end{array}$$

elimination from A to U (upper triangle)

determinant = multiple by diagles.

bosck  
substitution

$$\begin{aligned} x + 2y + z &= 2 \\ 2y - 2z &= 6 \\ 5z &= -10 \end{aligned} \quad \begin{aligned} x &= 2 \\ y &= 1 \\ z &= -2 \end{aligned}$$

Matrices.

$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 2 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 3 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 2 & | & 1 \\ 3 & 8 & | & 1 \\ 0 & 4 & | & 1 \end{bmatrix}$$

$E_{32}$

$E_{21}$

A

$$[a, b, c] \begin{bmatrix} A \end{bmatrix} =$$

$$E_{32}(E_{21}A) = U \quad \text{it is associative}$$

Permutation matrix

Inverse:

$$\begin{bmatrix} 1 & 0 & 0 \\ 3 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ -3 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} = I$$

$E^{-1}$

### Lecture 3

Matrix multiplication

$$1. \text{ element wise : } C_{ij} = \sum_{k=0}^n a_{ik} b_{kj}$$

2. column wise :

$$\begin{bmatrix} \begin{array}{|c|c|c|} \hline & & c_1 \\ \hline & & \end{array} \end{bmatrix} = \begin{bmatrix} \begin{array}{|c|} \hline Ac_1 \\ \hline \end{array} \end{bmatrix}$$

$A \quad B \quad C$   
 $m \times n \quad n \times p$

each column of  $C$  is a linear combination of ~~the~~ columns of  $A$

3. row wise :

$$\begin{bmatrix} \begin{array}{|c|} \hline r \\ \hline \end{array} \end{bmatrix} \begin{bmatrix} \begin{array}{|c|} \hline \vdots \\ \hline B \\ \hline \end{array} \end{bmatrix} = \begin{bmatrix} \begin{array}{|c|} \hline rB \\ \hline \end{array} \end{bmatrix}$$

$A \quad B$

each row of  $C$  is a linear combination of rows of  $B$

4. Column of  $A$  \* Row of  $B$ . ? Why

$$A B = \text{Sum of } (\text{Column of } A) (\text{Row of } B)$$

5. by Block

$$A_1 B_1 + A_2 B_3$$

$$\begin{bmatrix} \begin{array}{|c|c|} \hline A_1 & A_2 \\ \hline A_3 & A_4 \\ \hline \end{array} \end{bmatrix} \begin{bmatrix} \begin{array}{|c|c|} \hline B_1 & B_2 \\ \hline B_3 & B_4 \\ \hline \end{array} \end{bmatrix} = \begin{bmatrix} \downarrow & \\ \text{+} & \end{bmatrix}$$

$A \quad B$

Inverse matrix:

$$A^T A = I = AA^T \text{ for square matrix.}$$

Inverted or non-singular.

(hard to prove)

(left inverse equal  
to right inverse)

Singular matrix:

$$A = \begin{bmatrix} 1 & 3 \\ 2 & 7 \end{bmatrix} = \text{any } AB \rightarrow \text{regular column space on the same line.}$$

We can find  $Ax = 0$ , so it is singular.  
proof:

because if  $A^T$  exist  $\Rightarrow A^T A x = A^T 0 = 0$

Showing  $A^{-1}$

$$\begin{bmatrix} 1 & 3 \\ 2 & 7 \end{bmatrix} \begin{bmatrix} a & c \\ b & d \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \Rightarrow x = 0$$

$A \quad A^T \quad ]$

$A \times$  column of  $A^{-1}$  = Column of  $I$ .

Gauss-Jordan (solve 2 at once)

$$\begin{bmatrix} 1 & 3 \\ 2 & 7 \end{bmatrix} \begin{bmatrix} a \\ b \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 3 \\ 2 & 7 \end{bmatrix} \begin{bmatrix} c \\ d \end{bmatrix} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

Augmented Matrix

$$\begin{bmatrix} 1 & 3 & | & 1 & 0 \\ 2 & 7 & | & 0 & 1 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 3 & | & 1 & 0 \\ 0 & 1 & | & -2 & 1 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 0 & | & 7 & -3 \\ 0 & 1 & | & -7 & 1 \end{bmatrix}$$

$$1. \quad ABB^T A^{-1} = I$$

$$2. \quad AA^T = I$$

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

$$E \quad A \quad U \quad A = L_{\text{lower}} U_{\text{upper}}$$

$$\begin{bmatrix} 1 & 0 \\ -4 & 1 \end{bmatrix} \begin{bmatrix} 2 & 1 \\ 8 & 7 \end{bmatrix} = \begin{bmatrix} 2 & 1 \\ 0 & 3 \end{bmatrix} \quad \begin{bmatrix} 2 & 1 \\ 8 & 7 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 4 & 1 \end{bmatrix} \begin{bmatrix} 2 & 1 \\ 0 & 3 \end{bmatrix}$$

---

$$\begin{bmatrix} 1 & 0 \\ 4 & 1 \end{bmatrix} \begin{bmatrix} 2 & 0 \\ 0 & 3 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$E_{32} E_{31} E_{21} A = U$$

$$A = E_2^{-1} E_3^{-1} E_{32}^{-1} U \sim \text{this is better because less compute dependency.}$$

$$A = LU \quad \text{if no row exchanges, the multipliers go directly into } L. \quad (\text{ie, } E_{31}^{-1} \text{ have no dependency of } E_{21}^{-1})$$

---

How many operation : it takes to eliminate?  $O(N^3)$

Permutations:  $\begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \quad \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \quad \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}$  permutation group

$$\begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \quad \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \quad \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}$$

$$P^{-1} = P^T$$

Permutation in  $\mathbb{B} A = L U \rightarrow$  allow row exchange

$$PA = LU$$

$P$  = identity matrix with reordered rows...

$n!$  counts the possible reordering.

$P$  is invertible, and  $P^T = P^{-1}$ ?  $P^T P = I$

Transpose =  $(A^T)_{ij} = A_{ji}$

Symmetric:  $A^T = A$

matrices

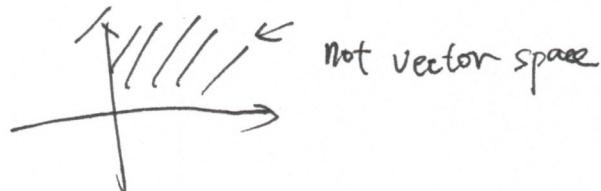
$R^T R$  is always symmetric?

$$\boxed{R_{ij} - \cancel{\textcircled{E}}(R^T R)^T = \cancel{R^T R} R^T R^{TT} = R^T R}$$

## Lecture 5: Vector spaces (chapter 3)

1.  $\mathbb{R}^2$  = all 2 dim real number vectors  
= the plane.

$\mathbb{R}^n$  = all vectors with  $n$  Real components.



Example of  $\mathbb{R}^2$  subspace:  
1. lines through orig.  
2. origin / zero vector

3.  $\mathbb{R}^2$

---

$A = \begin{bmatrix} 1 & 3 \\ 2 & 3 \\ 4 & 1 \end{bmatrix}$  column in  $\mathbb{R}^3$   
all their linear combination form a subspace.  
called column space.  $C(A)$ .

---

Lecture 6.: 2 subspaces = P and L

PUL  $\neq$  is not a subspace. (usually)

$P \cap L$  is a subspace for any two subspaces P and L

because any  $v, w$ ,  $v+w$  in both subspaces,

$\neq v+w$  and  $c v$  are in the subspace;

all linear combination form a subspace.

column space of  $A$  is a subspace of  $\mathbb{R}^4$   $C(A)$

$$A = \begin{bmatrix} 1 & 2 \\ 2 & 3 \\ 3 & 4 \\ 4 & 5 \end{bmatrix}$$

= all linear combination of columns.

Does  $Ax = b$  have a solution for every  $b$ . No

$$Ax = b$$

$$\begin{bmatrix} 1 & 2 \\ 2 & 3 \\ 3 & 4 \\ 4 & 5 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \\ b_4 \end{bmatrix}$$

for which  $b$  have answers?

We can solve  $Ax = b$  when  $b$  is in  $A$ 's column space ( $C(A)$ )

Nullspace of  $A$

of  $Ax = 0$  in  $\mathbb{R}^3$

all solution  $x$

$$\begin{bmatrix} 1 & 2 \\ 2 & 3 \\ 3 & 4 \\ 4 & 5 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

$N(A)$  contains

$$P \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}, c \begin{pmatrix} 1 \\ -1 \\ 1 \end{pmatrix}$$

if  $v, w$  is in Nullspace, then  
any linear combination is  
also in Nullspace  
i.e.  $Av = 0$   $Aw = 0$   
 $A(cv + dw) = 0$

the

$$Ax = \begin{bmatrix} 1 \\ 2 \\ 3 \\ 4 \end{bmatrix}$$

is not a subspace  $\rightarrow$  as zero is not in the space  
is a plane not through origin.

2 ways of construct subspaces

1. column combinations

2. solve equations

## Lecture 7.

$$A = \begin{bmatrix} 1 & 2 & 2 & 2 \\ 2 & 4 & 6 & 8 \\ 3 & 6 & 8 & 10 \end{bmatrix}$$

$$\begin{bmatrix} ① & 2 & 2 & 2 \\ 0 & 0 & ② & 4 \\ 0 & 0 & 2 & 4 \end{bmatrix}$$

Rank of A  
= # of pivots  
= 2.

**echelon**  $\rightarrow$   $\begin{bmatrix} ① & 2 & 2 & 2 \\ 0 & 0 & ② & 4 \\ 0 & 0 & 0 & 0 \end{bmatrix} = U$

↑  
2 pivot columns  
↓  
free columns

elimination  $\Rightarrow$  Won't change null space

1. elimination find free
2. give 1, 0 to free variables
3. find special solutions,  
combine them

$r = 2$ .  
 $n - r = 2 \Rightarrow$  free variables. ~~for mxn matrix A~~.

rref(A)

$$\begin{bmatrix} 1 & 2 & 2 & 2 \\ 0 & 0 & 2 & 4 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

R reduced row echelon form

$$\begin{bmatrix} 1 & 2 & 0 & -2 \\ 0 & 0 & 2 & 4 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

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

$$R = \text{rref}(A) = \begin{bmatrix} I & F \\ 0 & 0 \end{bmatrix} \quad \begin{matrix} \leftarrow r \text{ pivot rows} \\ \downarrow \quad \quad \quad n - r \text{ free columns} \\ \quad \quad \quad r \text{ pivot column} \end{matrix}$$

Magic!!!

$$RN = 0$$

$$N = \begin{bmatrix} -F \\ I \end{bmatrix} \rightarrow N \text{ is nullspace}$$

## Lecture 8. solve $Ax = b$

$$\left[ \begin{array}{cccc|c} 1 & 2 & 2 & 2 & b_1 \\ 2 & 4 & 6 & 8 & b_2 \\ 3 & 6 & 8 & 10 & b_3 \end{array} \right] \rightarrow \left[ \begin{array}{cccc|c} 1 & 2 & 2 & 2 & b_1 \\ 0 & 0 & 2 & 4 & b_2 - 2b_1 \\ 0 & 0 & 2 & 4 & b_3 - 3b_1 \end{array} \right] \left[ \begin{array}{cccc|c} 1 & 2 & 2 & 2 & b_1 \\ 0 & 0 & 2 & 4 & b_2 - 2b_1 \\ 0 & 0 & 0 & 0 & b_3 - b_2 - b_1 \end{array} \right]$$

Solvability condition on the right hand size.

$Ax = b$  solvable when  $b$  is in  $C(A)$

if combination of rows of  $A$  give zero row

then the same combination of entries of  $b$  must give 0.

To find complete sol'n to  $Ax = b$ .

- ①  $x_{\text{particular}} :$  1. Set free variable to zero  
2. Solve  $Ax = b$  for pivot

+

②  $x_{\text{nullspace}}$

$$A x_p = b$$

$$x = x_p + x_n$$

$$A x_n = 0$$

$$A(x_p + x_n) = b$$

$$x = \begin{bmatrix} -2 \\ 9/2 \\ 0 \end{bmatrix} + c_1 \begin{bmatrix} -2 \\ 1 \\ 0 \end{bmatrix} + c_2 \begin{bmatrix} 2 \\ 0 \\ 1 \end{bmatrix}$$

a plane through  $x_{\text{particular}}$ .

nullspace

$m$  by  $n$  matrix  $A$  of rank  $r$  ( $r \leq m, r \leq n$ )

1. Full rank column means  $\boxed{r=n}$ , no free variable  
 $N(A) = \{\text{zero vector}\}$

Solution to  $Ax=b$  :  $x=x_p$  unique solution if it exists

$$A = \begin{bmatrix} 1 & 3 \\ 2 & 1 \\ 0 & 1 \\ 5 & 1 \end{bmatrix} R = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \\ 0 & 0 \end{bmatrix}$$

(or 1 solution)

2. Full Row Rank means  $r=m$

Can solve  $Ax=b$  for every  $b$

Left with  $n-r$  free variables

$$A = \begin{bmatrix} 1 & 2 & 6 & 5 \\ 3 & 1 & 1 & 1 \end{bmatrix}$$

$$R = \begin{bmatrix} 1 & 0 & \boxed{-F} \\ 0 & 1 & \boxed{-E} \end{bmatrix}$$

3.  $r=n=M$

$A = \begin{bmatrix} 1 & 2 \\ 3 & 1 \end{bmatrix}$   $N(A) = \{\text{zero}\}$ ,  $Ax=b$  exists for any  $b$ .

$$R = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = I$$

$r = m = n$	$r = \frac{n}{m}$ $R = [I]$ 1 solution	$r < m$ $R = [I \ F]$ (0 or 1 solution)	$r > m$ $R = [I \ F]$ ( <del>for</del> infinite many solutions)
$r < m, r < n$ $R = [I \ F]$ (0 or $\infty$ solutions)			

# Lecture 9: Independent Spanning $\Leftrightarrow$ basis

1. suppose  $A$  is  $m \times n$  with  $n < m$ ,

Then there are infinite solutions to  $Ax = 0$   
(because there are free variables.)

2. Independence

Vectors  $x_1, x_2, \dots, x_n$  are independent  
if no combinations give zero vector (except zero)

Repeat  $\therefore$  when  $v_1, \dots, v_n$  are columns of  $A$

They are independent if  $N(A)$  is {zero vector}

They are dependent if  $Ac = 0$  for some non zero  $c$ .  
 $\boxed{\text{rank } = n}$

$\boxed{\text{rank } < n}$

Vectors  $v_1, \dots, v_n$

Span

The space consists of all spans of those vectors.

Basis for a space is a sequence of vectors

$v_1, v_2, \dots, v_d$ . With two properties

1. They are independent.

2. They span the space

Example  $\mathbb{R}^3$

Space is  $\mathbb{R}^3$

one basis is

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

$$\begin{bmatrix} 1 \\ 1 \\ 2 \end{bmatrix}, \begin{bmatrix} 2 \\ 1 \\ 5 \end{bmatrix}$$

n vectors give basis if  $n \times n$  matrix has n  
as column vectors  
if  $n \times n$  matrix is  
invertible

given a space

every basis for the space has the same number of vectors

Dimension of the space.

$\text{rank}(A) = \# \text{pivot columns} = \text{dimension of } C(A)$

matrix

space

$$\dim(A) = r$$

$\dim N(A) = \text{number of free variable} = n - r$

Null space vector really tells the vectors make columns  
dependent.

Lecture 10. 4 subspaces of  $A$  in  $\mathbb{R}^{m \times n}$

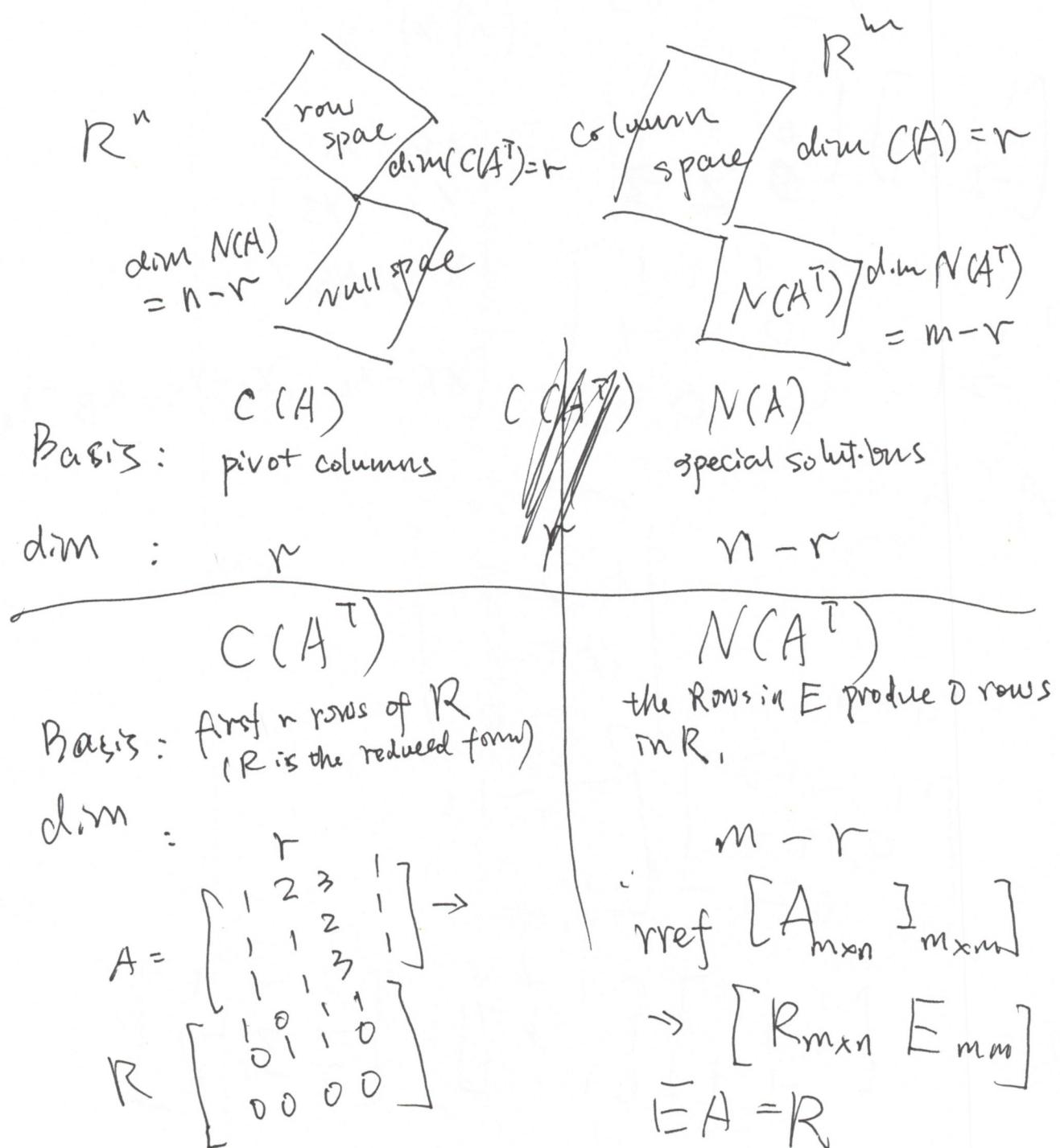
column space  $C(A)$  in  $\mathbb{R}^m$

Null space  $N(A)$  in  $\mathbb{R}^n$

Row space  $C(A^T)$  in  $\mathbb{R}^n$

left Null space  $N(A^T)$  in  $\mathbb{R}^m$

---



new vector space,  $M$   
all  $3 \times 3$  matrices !!  $A + B, CA$

Subspace of  $M$ : all upper triangles

symmetric matrices

diagonal matrices  $D \rightarrow \text{dim} = 3$

## Lecture 14. Orthogonal vectors and subspaces

Orthogonal vectors:  $x^T y = 0$

$$\text{Pythagoras: } \|x\|^2 + \|y\|^2 = \|x+y\|^2$$

$$x^T x + y^T y = (x+y)^T (x+y)$$

$$= x^T x + y^T y + x^T y + y^T x$$

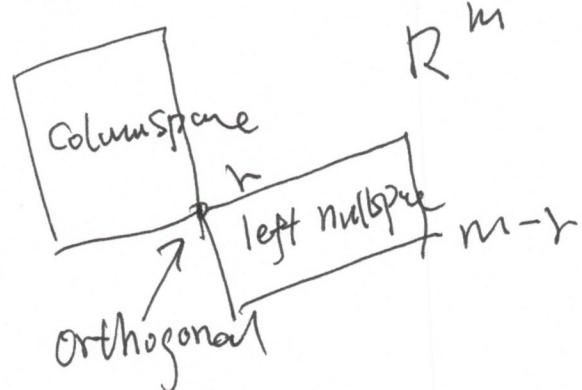
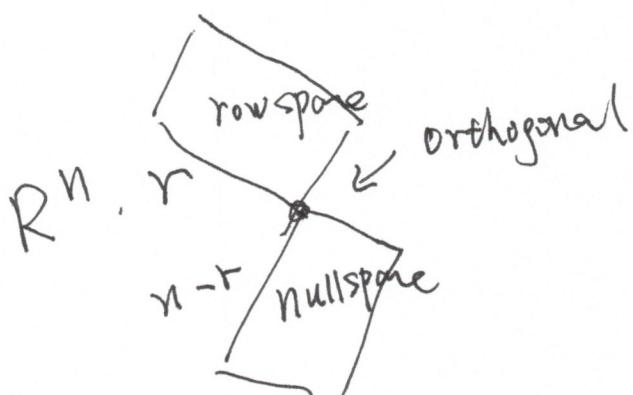
$$2x^T y = 0$$

Subspace  $S$  is orthogonal to subspace  $T$ :

means: every vector in  $S$  is orthogonal to every vector in  $T$

row space is orthogonal to nullspace.

why?  $\cancel{x} A x = 0$   $\begin{bmatrix} r_1 \\ r_2 \\ \vdots \\ r_n \end{bmatrix} \begin{bmatrix} x \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$



Nullspace and Row space are orthogonal, their dim add up to  $R^n$ :  $\rightarrow$  Orthogonal Complements

Nullspace contains ALL vectors  $\perp$  row space.

Coming: "solve"  $Ax = b$  when there is no solution. (i.e. noise)

( $m > n$ )

~~ATA~~ symmetric:

square:

$n \times m \quad m \times n$

$n \times n$



$$ATA\hat{x} = A^T b$$

---

$$N(A^T A) = N(A).$$

$$\text{rank of } A^T A = \text{rank of } A$$

$A^T A$  is invertible if  $A$

has independent columns

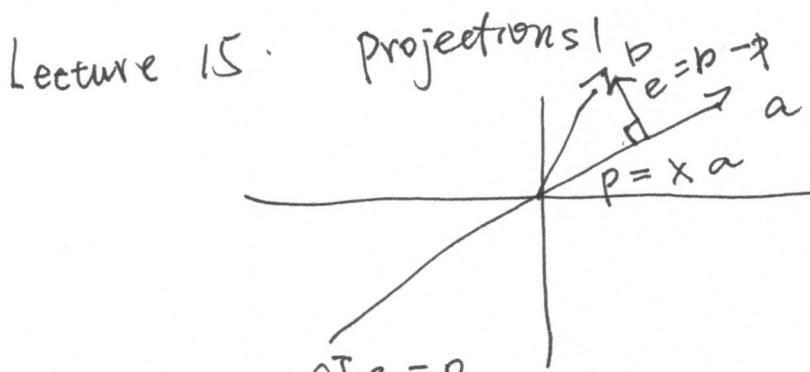
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Any solution of  $Ax = b$  can have be presented

as  $x_r + x_n$  where  $x_r$  is in Row space

and  $x_n$  is in Nullspace.

why? because  $C(A^T)$  and  $N(A)$  are ortho components  
of  $\mathbb{R}^n \rightarrow$  so any  $x$  can be presented  
as a  $x_r + x_n$  ~~where~~.



$$a^T e = 0$$

$$a^T(b - xa) = 0$$

$$xa^T a = a^T b$$

$$x = \frac{a^T b}{a^T a}$$

$$p = ax$$

$$P = a \frac{a^T b}{a^T a}$$

$$\text{Projection} = Pb \quad P = \frac{aa^T}{a^T a}$$

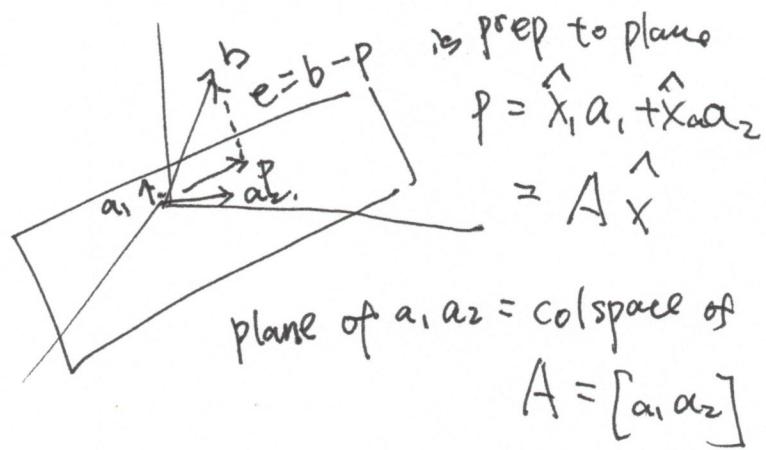
↑  
projection matrix

$C(P)$  = line through  $a$ .

$$\begin{array}{l} \text{rank}(P)=1 \\ * P^T = P. \quad P^2 = P \end{array} \quad (\text{projection twice})$$

Why project? because  $Ax=b$  may have no solution.

Solve  $\overset{\curvearrowleft}{Ax} = P$  instead.  
 → project  $b$  onto column space.



$$P = A \hat{X} \quad \text{Find } \hat{X}$$

key:  $b - A \hat{X}$  is prep to plane

$$a_1^T (b - A \hat{X}) = 0 \quad a_2^T (b - A \hat{X}) = 0$$

$$\begin{bmatrix} a_1^T \\ a_2^T \end{bmatrix} (b - A \hat{X}) = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \quad \begin{cases} e^{(b - A \hat{X})} \text{ in } N(A^T) \\ e \perp C(A) \end{cases}$$

$$A^T (b - A \hat{X}) = 0.$$

$$\boxed{A^T A \hat{X} = A^T b}$$

$$\hat{X} = (A^T A)^{-1} A^T b$$

$$P = \underbrace{A (A^T A)^{-1}}_{\text{id } a^T a} A^T b$$

$$\text{matrix } P = A (A^T A)^{-1} A^T \xrightarrow{\text{A is not invertible.}}$$

$I - P$  is also projection  
it is projecting to  $e$

$$(I - P)b \perp Pb$$

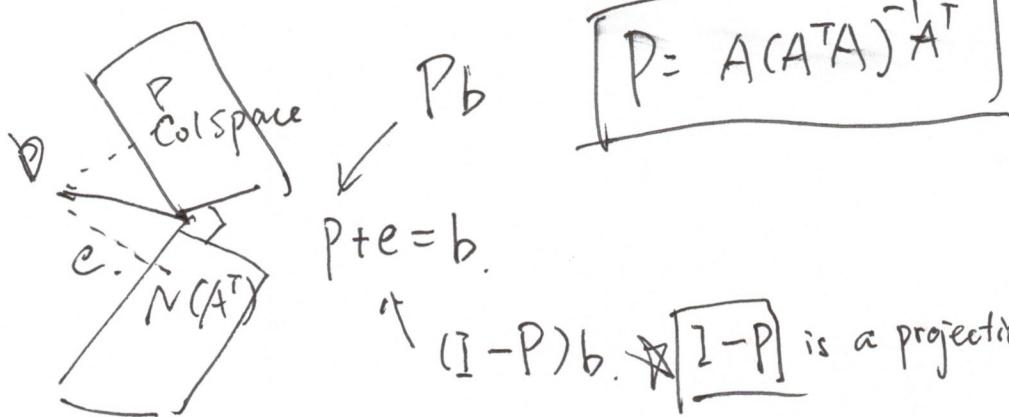
$$(I - P)\text{ space other}$$

$$P^T = P$$

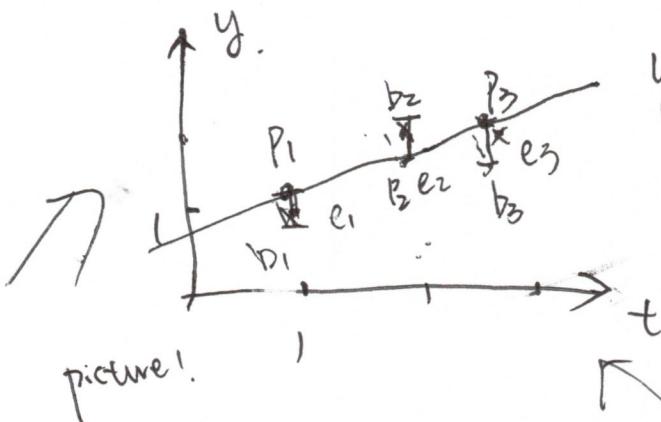
$$P^2 = P = P^2 = A (A^T A)^{-1} A^T A (A^T A)^{-1} A^T$$

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

## Lecture 16. Projection matrix and least square error.



$(I - P)b$ .  $\boxed{I - P}$  is a projection onto  $\perp$  space,  $N(A^T)$



$$y = C + Dt \quad Ax = b$$

$$\begin{aligned} C + D &= 1 \\ C + 2D &= 2 \\ C + 3D &= 2 \end{aligned}$$

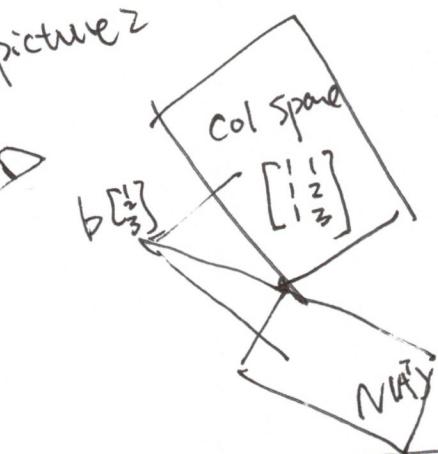
$$\begin{bmatrix} 1 & 1 \\ 1 & 2 \\ 1 & 3 \end{bmatrix} \begin{bmatrix} C \\ D \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \\ 2 \end{bmatrix}$$

$$\text{Minimize } \|Ax - b\|^2 = \|e\|^2.$$

$$= e_1^2 + e_2^2 + e_3^2$$

$$= (C+D-1)^2 + (C+2D-2)^2 + (C+3D-2)^2$$

equal to find a projector onto the column space



$$\text{Find } \hat{x} = \begin{bmatrix} \hat{x}_1 \\ \hat{x}_2 \\ \hat{x}_3 \end{bmatrix}, P \quad \begin{bmatrix} 1 & 1 & 1 \\ 1 & 2 & 3 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 1 & 2 \\ 1 & 3 \end{bmatrix} = \begin{bmatrix} 3 & 6 \\ 6 & 14 \end{bmatrix} \quad \begin{aligned} 3\hat{x}_1 + 6\hat{x}_2 &= 5 \\ 6\hat{x}_1 + 14\hat{x}_2 &= 11 \end{aligned}$$

$$\hat{x}_3 = \frac{1}{2}$$

$$\hat{x}_1 = \frac{2}{3}, \hat{x}_2 = \frac{1}{3}$$

$$A^T A \hat{x} = A^T b$$

If  $A$  has independent columns, then  $A^T A$  is invertible.

Proof. suppose  $A^T A x = 0$ ,  $x$  must be 0.

$$\Rightarrow x^T A^T A x = 0$$

$$\Rightarrow \|Ax\| = 0.$$

$$\Rightarrow Ax = 0.$$

because  $A$  has independent columns, so  $x$  is zero.

Columns are definitely independent, if they are perpendicular unit vector. like  $\begin{bmatrix} 1 \\ 0 \end{bmatrix} \begin{bmatrix} 0 \\ 1 \end{bmatrix} \begin{bmatrix} 0 \\ 0 \end{bmatrix}$

Orthogonal Vectors.

## Lecture 17.

Orthogonal basis.

Orthogonal matrix  $Q$

Gram-Schmidt  $A \rightarrow Q$

Orthonormal vectors.

$$q_i^T q_j = \begin{cases} 0 & \text{if } i \neq j \\ 1 & \text{if } i = j \end{cases}$$

$$Q = [q_1 \dots q_n] \quad Q^T Q = I = \begin{bmatrix} a_1 \\ \vdots \\ a_m \end{bmatrix} \begin{bmatrix} q_1 & \dots & q_n \end{bmatrix} = I$$

∅ ( $Q$  doesn't have to be square!!!).

Orthogonal matrix  $\rightarrow$  matrix is square and orthonormal

If  $Q$  is square.  $Q^T Q = I$ . tell us.  $Q^T = Q^{-1}$

i.e. permutation matrix.

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

$$Q = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}$$

$$Q = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}$$

$$Q = \frac{1}{2} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -1 & 1 & -1 \\ 1 & 1 & -1 & -1 \\ 1 & -1 & -1 & 1 \end{bmatrix}$$

$$Q = \frac{1}{3} \begin{bmatrix} 1 & -2 \\ 2 & -1 \\ 2 & 2 \end{bmatrix}$$

Project onto  $Q'$  column space.

$$P = Q(Q^T Q)^{-1} Q^T = \underline{\underline{Q Q^T}} \left( \begin{smallmatrix} \text{symmetric} \\ \text{square=itself} \end{smallmatrix} \right). \quad \left\{ \begin{array}{l} I \text{ is } Q \text{ is square} \end{array} \right\}$$

$$A^T A x = A^T b. \quad (\text{for least square error})$$

$$Q^T Q \hat{x} = Q^T b.$$

$$\boxed{\hat{x} = Q^T b} \rightarrow$$

$$\boxed{\hat{x}_i = q_i^T b}$$

Gram-Schmidt.

independent  
vectors  $a, b \rightarrow$  orthogonal  $\rightarrow$  orthonormal

$$a = A, B = b - P_b = b - \frac{A a a^T}{a^T a} b$$

$$C = c - P_c = c - \frac{A^T c}{A^T A} A - \frac{B^T c}{B^T B} B.$$

$\xrightarrow{\text{ratio of } A}$   
 $c$ 's projection on  $A$

$c$ 's projection on  $B$

$$a = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}, b = \begin{bmatrix} 0 \\ 0 \\ 2 \end{bmatrix}, B = \begin{bmatrix} 1 \\ 0 \\ 2 \end{bmatrix} - \frac{1}{3} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$$

$$Q = Q = \begin{bmatrix} q_1 & q_2 \end{bmatrix} = \begin{bmatrix} \frac{1}{\sqrt{3}} & 0 \\ \frac{1}{\sqrt{3}} & -\frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{2}} \end{bmatrix} \quad A \perp B$$

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

same column space!!!

$$A = QR.$$

$$[q_1 q_2] =$$

$$[q_1 q_2] \begin{pmatrix} a_1^T q_1 & * \\ 0 & a_2^T q_2 \end{pmatrix}$$

$q_2 \perp a_1$   
because Gram Schmidt does

$R$  is upper triangular

# Lecture 18..

Determinants  $\det A = |A|$

$$\textcircled{1} \quad \det I = 1.$$

$$\textcircled{2} \quad \begin{matrix} \text{Exchange rows} \\ \text{the sign of det} \end{matrix}$$

property of  $\det$

$$\leftarrow \det P = \begin{cases} +1 & \text{even} \\ -1 & \text{odd} \end{cases} \text{ number of exchanges.}$$

$$\textcircled{3a} \quad \begin{vmatrix} 1 & 0 \\ 0 & 1 \end{vmatrix} = 1 \quad \begin{vmatrix} a & b \\ c & d \end{vmatrix} \stackrel{3a}{=} t \begin{vmatrix} ab \\ cd \end{vmatrix} \quad \begin{vmatrix} ab \\ cd \end{vmatrix} = ad - bc$$

$$\textcircled{3b} \quad \begin{vmatrix} 0 & 1 \\ 1 & 0 \end{vmatrix} = -1 \quad \begin{vmatrix} a+a' & b+b' \\ c & d \end{vmatrix} = \begin{vmatrix} a & b \\ c & d \end{vmatrix} + \begin{vmatrix} a' & b' \\ c & d \end{vmatrix}.$$

linear combinations

FOR EACH ROW

$$\det(A+B) \cancel{\equiv \det A + \det B}$$

$$\det A + \cancel{\det B}$$

$$\textcircled{4} \quad 2 \text{ equal rows} \rightarrow \det = 0.$$

Exchange those rows  $\rightarrow$  same matrix.  $\Rightarrow \det 0.$

$$\textcircled{5} \quad \text{Subtract } l \times \text{row } i \text{ from row } k.$$

DET doesn't change

$$\begin{aligned} &\rightarrow \begin{vmatrix} a & b \\ -la & d-lb \end{vmatrix} \\ &= \begin{vmatrix} a & b \\ c & d \end{vmatrix} - l \begin{vmatrix} a & b \\ a & b \end{vmatrix} \\ &= \begin{vmatrix} a & b \\ c & d \end{vmatrix} - 0 \quad \textcircled{4} \end{aligned}$$

(3.a/3.b)

⑥ Row of zeros  $\rightarrow \det A = 0$   
 $\left| \begin{array}{cc} a & b \\ c & d \end{array} \right| \equiv \left| \begin{array}{cc} 0 \cdot a & 0 \cdot b \\ c & d \end{array} \right| = 0 \left| \begin{array}{cc} ab \\ cd \end{array} \right| = 0$

⑦  $\det U = \begin{vmatrix} d_1 & & * \\ 0 & \dots & d_n \end{vmatrix} = d_1 \cdot d_2 \cdots d_n$

↓ prove by eliminations  $\text{L.R}$  then.

(rule 5)

we get  $\det U' = \begin{vmatrix} d_1 & 0 & & \\ 0 & d_2 & \dots & \\ & & \ddots & \\ & & & d_n \end{vmatrix}$

(rule 3a)

↓  $d_n \cdot d_2 d_1 \begin{vmatrix} 1 & 0 & & \\ 0 & 1 & \dots & \\ & & \ddots & \\ & & & 1 \end{vmatrix}$

(rule 1)

↓  $d_n \cdot d_2 d_1 \cdot *$

⑧

$\det A = 0$

row of zeros when column L.R elimination

When A is singular.

$\det A \neq 0$

$\rightarrow$  L.R elimination

When A is invertible.

$\downarrow D \rightarrow d_1 d_2 \cdots d_n$

$$\left[ \begin{array}{cc} a & b \\ c & d \end{array} \right] \rightarrow \left[ \begin{array}{cc} a & b \\ 0 & d - \frac{c}{a}b \end{array} \right] \rightarrow \det A = ad - bc$$

$$\textcircled{9} \star \det AB = (\det A)(\det B) // \cancel{\det A^{-1}}$$

$$\det A^{-1} = \frac{1}{\det A} \quad \text{use above property.}$$

(10)

for diagonal matrix - trivial to prove.

for other matrix, need go through elimination.

$$\det A^2 = (\det A)^2$$

$$\det 2A = 2^n \det A \quad \dots \text{(like volume)}$$

$$\textcircled{10} \quad \det A^T = \det A. \rightarrow \text{all previous rows rule apply to columns!}$$

$$\text{prove } |A^T| = |A|$$

$$|U^T L^T| = |L U|$$

$$|U^T| |L^T| = |L| |U|$$

$\uparrow$  all triangular, so the det is only on diagonals!

$\nexists$  rule (2) implies permutations are either odd or even!

Lecture 19. Determinant formulae and cofactors.

①  $\det I = 1$

② SIGN REVERSE. Exchange two rows

③ LINEAR ON EACH Row separately.

$$\begin{vmatrix} a & b \\ c & d \end{vmatrix} = \begin{vmatrix} a & 0 \\ c & d \end{vmatrix} + \begin{vmatrix} 0 & b \\ c & d \end{vmatrix} = \cancel{\begin{vmatrix} a & 0 \\ 0 & d \end{vmatrix}} + \cancel{\begin{vmatrix} a & 0 \\ c & d \end{vmatrix}} + \begin{vmatrix} 0 & b \\ c & 0 \end{vmatrix} + \cancel{\begin{vmatrix} 0 & b \\ 0 & d \end{vmatrix}}$$

(3) (2)

$= ad - bc.$

2x2

$$\begin{vmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix} = \begin{vmatrix} a_{11} & & \\ & a_{22} & \\ & & a_{33} \end{vmatrix} + \begin{vmatrix} a_{11} & & \\ & a_{23} & \\ & & a_{32} \end{vmatrix} + \begin{vmatrix} a_{12} & & \\ & a_{21} & \\ & & a_{33} \end{vmatrix} + \begin{vmatrix} a_{12} & & \\ & a_{23} & \\ & & a_{31} \end{vmatrix} + \begin{vmatrix} a_{13} & & \\ & a_{21} & \\ & & a_{32} \end{vmatrix} + \begin{vmatrix} a_{13} & & \\ & a_{22} & \\ & & a_{31} \end{vmatrix}$$
$$= a_{11}a_{22}a_{33} - a_{11}a_{23}a_{32} - a_{12}a_{21}a_{33} + a_{12}a_{23}a_{31} + a_{13}a_{21}a_{32} - a_{13}a_{22}a_{31}$$

3x3

Big formula:  $n \times n$  det

$$\det A = \sum_{n! \text{ terms}} \pm a_{i_1} a_{i_2} a_{i_3} \dots a_{i_n}$$

$(\alpha \beta \gamma \dots \omega) \not\in$  Permutation of  $(1, 2 \dots n)$

Example

$$\begin{vmatrix} 0 & 0 & 1 & 1 \\ 0 & 1 & 0 & 0 \\ 1 & 0 & 0 & 0 \end{vmatrix} = 0 \rightarrow \text{singular}$$

$$(4, 3, 2, 1) \rightarrow + (-3, 2, 1, 4) \rightarrow -1$$

Cofactors!  $3 \times 3$  in parens.

$$\det = a_{11}(a_{22}a_{33} - a_{23}a_{32}) \leftarrow \det \text{ of } 2 \times 2$$

$$+ a_{12}(-a_{21}a_{33} + a_{23}a_{31})$$

$$+ a_{13}(a_{21}a_{32} - a_{22}a_{31})$$

$$\begin{vmatrix} a_{11} & 0 & 0 \\ 0 & a_{22} & a_{23} \\ 0 & a_{32} & a_{33} \end{vmatrix} + \begin{vmatrix} 0 & a_{12} & 0 \\ a_{21} & 0 & a_{23} \\ a_{31} & 0 & a_{33} \end{vmatrix}$$
  
$$\begin{vmatrix} a_{21} & a_{22} & a_{13} \\ a_{31} & a_{32} & \end{vmatrix}$$

Cofactor of  $a_{ij} = C_{ij}$

$\pm \det$  ( $n-1$  matrix  
with row  $i$ ; col  $j$  erased)

$\rightarrow i+j$  |  $i+j$   
+ even | - odd

Cofactor formula: (along row 1)

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

$$\begin{vmatrix} a & b \\ c & d \end{vmatrix} = ad + b(-c)$$

Ex.

$$A_4 = \begin{vmatrix} 1 & 1 & 0 & 0 \\ 1 & 1 & 1 & 0 \\ 0 & 1 & 1 & 1 \\ 0 & 0 & 1 & 1 \end{vmatrix}$$
$$|A_1| = 1 \quad |A_2| = 0 \quad |A_3| = -1 \quad |A_4| =$$
$$|A_4| = 1 \cdot |A_3| + 1 \cdot |A_{32}| = 1$$
$$|A_n| = |A_{n-1}| - |A_{n-2}|.$$

## Lecture 20.

1. Formulae for  $A^{-1}$

$$\begin{bmatrix} a & b \\ c & d \end{bmatrix}^{-1} = \frac{1}{ad-bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$$

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

$C \rightarrow$  cofactor matrix

Check  $A C^T = (\det A) I$

$$\begin{bmatrix} a_{11} & \dots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{n1} & \dots & a_{nn} \end{bmatrix} \begin{bmatrix} C_{11} & \dots & C_{1n} \\ C_{21} & \ddots & \vdots \\ \vdots & & C_{nn} \end{bmatrix} = \begin{bmatrix} \det A & & & \\ & \det A & & \\ & & \ddots & \\ & & & \det A \end{bmatrix}$$

the non-diagonal value is zero, is because it is equivalent to take det of a matrix with two equal rows.

$$Ax = b$$

$$x = A^{-1}b = \frac{1}{\det A} C^T b.$$

### CRAMER'S RULE

$$x_1 = \frac{\det B_1}{\det A}, \dots, (C_{11}b_1 + b_{21}b_2 + \dots)$$

$$x_2 = \frac{\det B_2}{\det A}$$

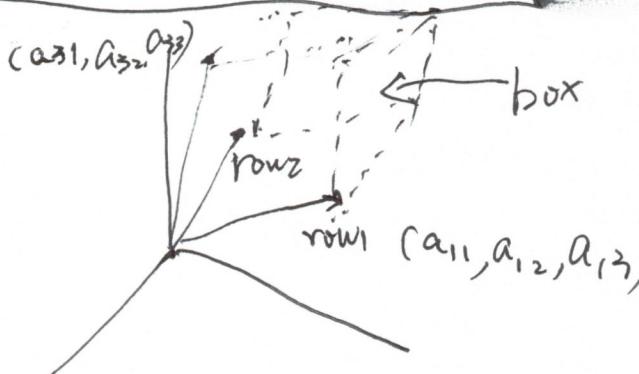
$$x_j = \frac{\det B_j}{\det A}$$

$A$  with column 1 replaced by  $b$ .

$$B_1 = \begin{bmatrix} b & n-1 \\ \text{columns} & \text{of } A \end{bmatrix}$$

$$B_j = \begin{bmatrix} A \text{ with column } j \\ \text{replaced by } b \end{bmatrix}$$

$|\det A| = \text{volume of box}$   $3 \times 3$



$$A = I \rightarrow 1$$

$$A = Q \rightarrow 1$$

to prove it: just  
check 3 properties !!

$$\begin{aligned} Q^T Q &= I \rightarrow \det(Q^T) \det(Q) = \det(I) \\ \det(Q)^2 &= 1 \end{aligned}$$

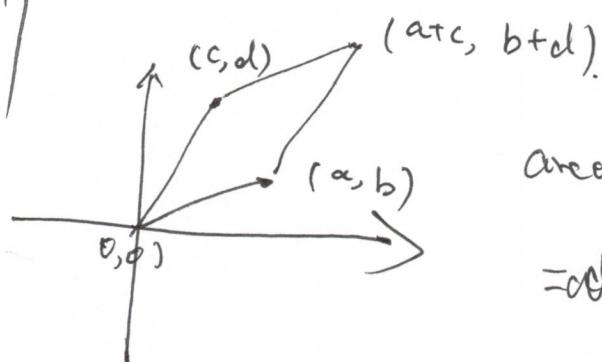
$|\det A| = \text{volume box} \rightarrow$  property 1 ✓  $(\det I = 1)$

property 2 ✓  $(\text{sign. doesn't mind.})$

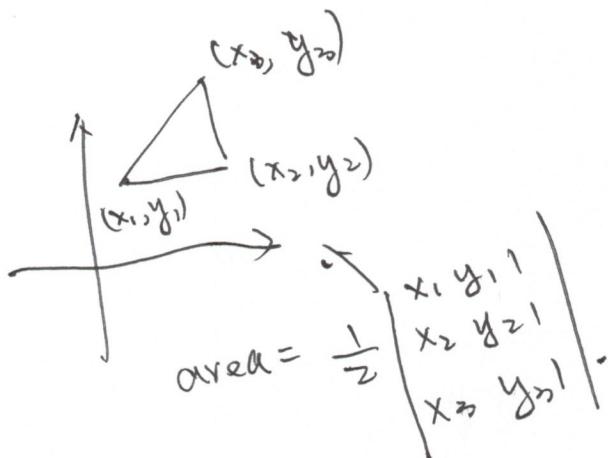
property 3.a ✓ factor a row,  $\Leftrightarrow$  increase the  
row length by  $n$

property 3.b.

$$\begin{vmatrix} a+cd, b+d \\ c, d \end{vmatrix} = \begin{vmatrix} a, b \\ c, d \end{vmatrix} + \begin{vmatrix} c, d \\ c, d \end{vmatrix}$$



$$\begin{aligned} \text{area} &= \det \begin{bmatrix} ab \\ cd \end{bmatrix} \\ &= ad - bc \end{aligned}$$



$$\frac{1}{2} (ad - bc) \text{ for triangle}$$