

Basis: set of vectors

- span a space
- lin. ind.

$$\mathbb{R}^n \quad x \in \mathbb{R}^n$$

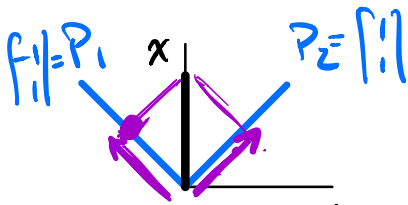
basis: cols of  $P \in \mathbb{R}^{n \times n}$

$$P = [P_1 \dots P_n]$$

$$x = Pz = P_1 z_1 + \dots + P_n z_n$$

$\downarrow$  basis vectors  $\rightarrow$  coords of  $x$  w.r.t. the basis  $P$

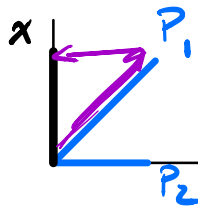
Ex:  $P = [P_1 \ P_2]$



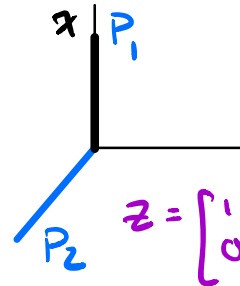
$$x = Pz = P_1 z_1 + P_2 z_2$$

$$z = P^{-1}x \quad z = \begin{bmatrix} 1/2 \\ 1/2 \end{bmatrix}$$

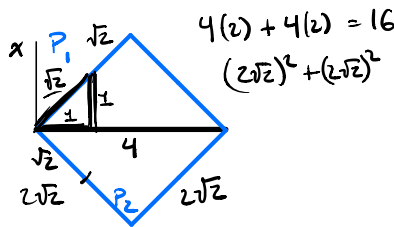
$$P^{-1} = \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix}^{-1} = \frac{1}{2} \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}$$



$$z = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$



$$z = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$



$P$  is a basis for  $\mathbb{R}^n$

changing basis = coordinate transformation

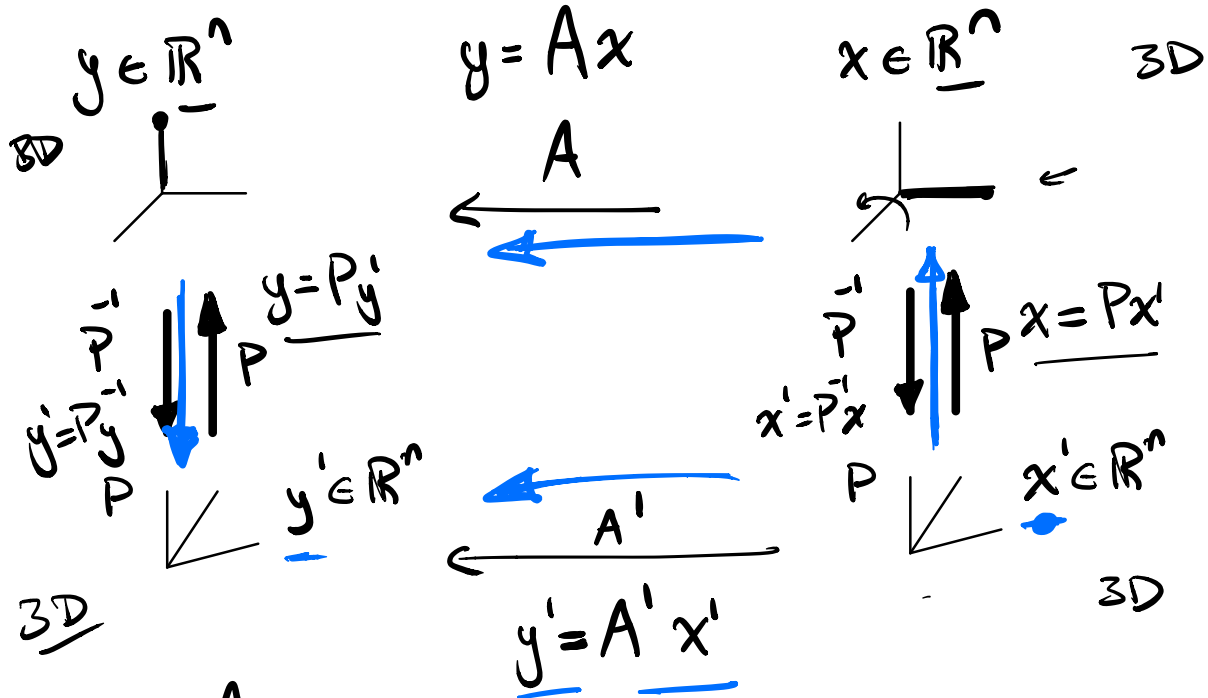
$P$ : new basis, coordinate transform from the standard basis

$$x = Pz \rightarrow \text{coords w.r.t. } P.$$

$\uparrow$   
 coords w.r.t. standard basis

Inverse coord transform:  $x' = P^{-1}x$   
 How do coord transforms affect matrices?

$A \in \mathbb{R}^{n \times n}$  square



$$y = Ax$$

$$Py' = APx'$$

$$y' = \underline{P^{-1}AP}x'$$

$$A' = P^{-1}AP$$

← Similarity transform on  $A$   
 $A$  &  $A'$  are similar

Two similar matrices perform the same transformation but w.r.t different coord. systems.

$$P(A' = P^{-1}AP)P^{-1} \Rightarrow \boxed{A = PA'P^{-1}}$$

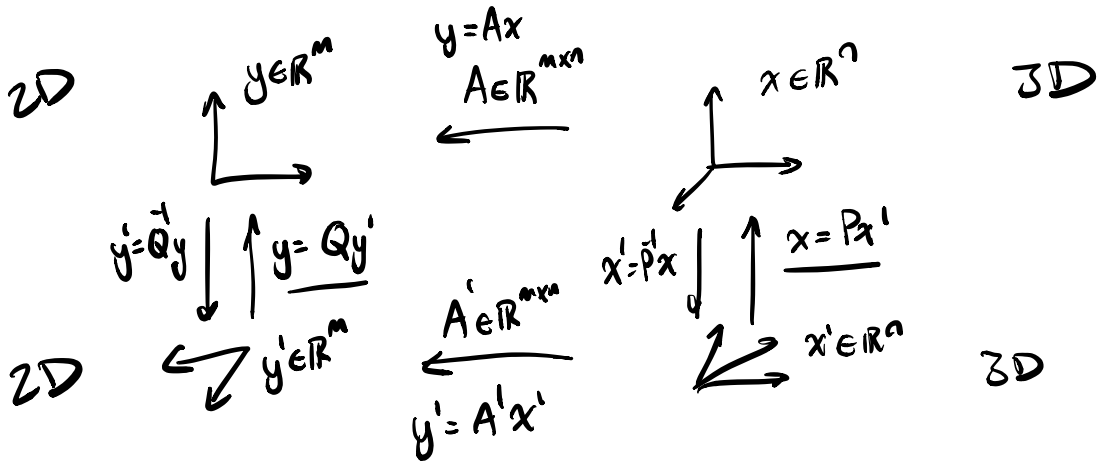
Note: in general matrices don't commute

$$AB \neq BA$$

2 rotations...  $R_1, R_2$

→ [distributivity  
associativity  
commutativity]

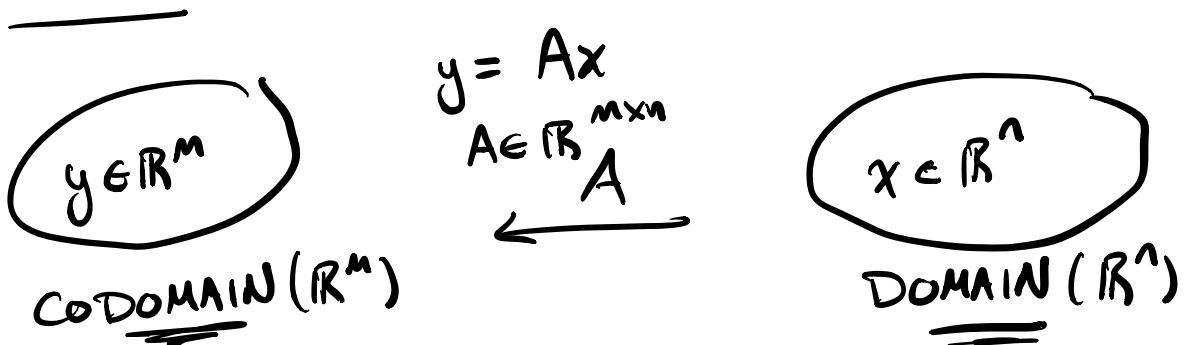
More general



$$y = Ax$$

$$Qy' = APx' \Rightarrow y' = Q^{-1}APx' \quad A' = Q^{-1}AP$$

(similarity) when  $Q = P$ .



1. Range  $(A) = \mathcal{R}(A) \subseteq \text{CODOMAIN}$ .
2. Nullspace  $(A) = \mathcal{N}(A) \subseteq \text{DOMAIN}$

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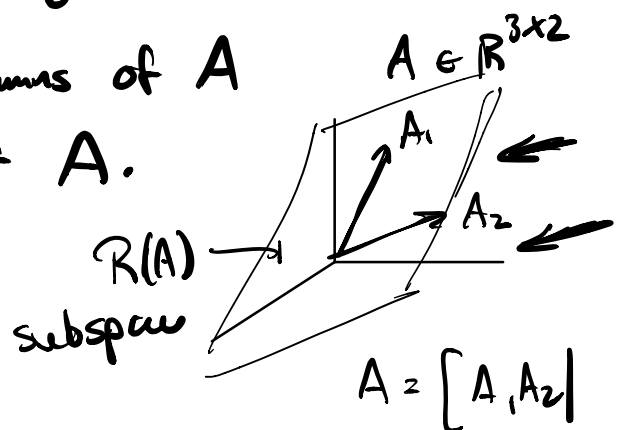

$$\text{Range}(A^T) = \mathcal{R}(A^T) \subseteq \text{DOMAIN}$$

$$\text{Nullspace}(A^T) = \mathcal{N}(A^T) \subseteq \text{CODOMAIN}$$

## Range & Nullspace

1. Range  $R(A) = \{y \in \mathbb{R}^m \mid y = Ax, x \in \mathbb{R}^n\}$

$R(A)$  is span of columns of  $A$   
"columns space of  $A$ ."



2. Nullspace  $N(A) = \{x \in \mathbb{R}^n \mid Ax = \underset{\substack{\downarrow \\ \text{0 vector}}}{0}, x \in \mathbb{R}^n\}$

if  $R(A) = \mathbb{R}^m$ :  $A$  is onto (surjective)

$$\exists x \text{ s.t. } y = Ax \text{ for any } y \in \mathbb{R}^m$$

if  $N(A) = \{0\} \rightarrow A$  has a trivial nullspace

$\times$   $N(A) = 0$   $A$  is one-to-one (injective)

if  $y = Ax$  then  $x$  is unique  
only  $x$  that maps to  $y$ .

if not  $y = Ax, y = Ax'$  for  $x \neq x'$   $Ax = y = Ax'$

$$\Rightarrow A(\underline{x} - \underline{x}') = \underline{0}$$

$\underline{x} \neq \underline{x}'$

more than 1 solution to a system of lin. eqns. means a non trivial nullspace.

Some properties of  $N(A)$ :

- if  $\underline{x} \in N(A)$ , then  $\underline{x} \perp$  rows of  $A$ .

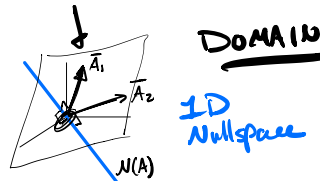
$$A = \begin{bmatrix} -\bar{A}_1^T \\ \vdots \\ -\bar{A}_m^T \end{bmatrix} \quad Ax = \begin{bmatrix} -\bar{A}_1^T \\ \vdots \\ -\bar{A}_m^T \end{bmatrix} x = \begin{bmatrix} \bar{A}_1^T x \\ \vdots \\ \bar{A}_m^T x \end{bmatrix} = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$$

Picture

$$\Rightarrow \bar{A}_i^T x = 0 \quad i=1, \dots, m$$

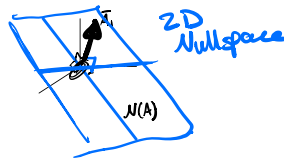
$$A \in \mathbb{R}^{2 \times 3}$$

$$A = \begin{bmatrix} -\bar{A}_1^T \\ -\bar{A}_2^T \end{bmatrix}$$



$$A \in \mathbb{R}^{1 \times 3}$$

$$A = [-\bar{A}_1^T]$$

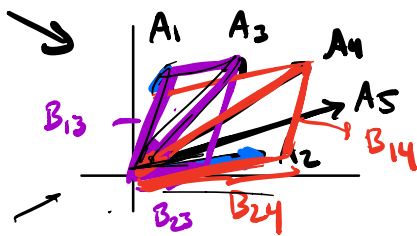


Construct a basis for  $N(A)$ :

$$A \in \mathbb{R}^{2 \times 5}$$

$$A = [A_1 \ A_2 \ A_3 \ A_4 \ A_5]$$

CODOMAIN



$y = Ax$  non unique solutions  
cols of  $A$  are lin dep.

$$y = A(\underline{x} + \underline{z}) = Ax + A\underline{z} \quad \underline{z} \in N(A)$$

$$A = \left[ \underbrace{A_1 \ A_2}_{\text{lin ind cols}} \mid \underbrace{A_3 \ A_4 \ A_5}_{\text{lin dep cols}} \right]$$

$$y = [A_1 \ \dots \ A_n] \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} +$$

$$A_3 = [A_1, A_2 | B_3] \rightarrow \begin{matrix} \text{coords of} \\ A_3 \text{ wrt. } A_1, A_2 \end{matrix} \quad B_3 = \begin{bmatrix} B_{13} \\ B_{23} \end{bmatrix}$$

$$A_4 = [A_1, A_2 | B_4] \leftarrow \begin{matrix} \text{coords of } A_4, A_5 \\ \text{wrt. } A_1, A_2 \end{matrix} \quad B_4 = \begin{bmatrix} B_{14} \\ B_{24} \end{bmatrix}$$

Consider:

$$[A_1, A_2 | A_3 \ A_4 \ A_5] \begin{bmatrix} B_3 \\ -1 \\ 0 \\ 0 \end{bmatrix} = [A_1, A_2] B_3 - A_3 = 0$$

$$[A_1, A_2 | A_3 \ A_4 \ A_5] \begin{bmatrix} B_4 \\ 0 \\ -1 \\ 0 \end{bmatrix} = [A_1, A_2] B_4 - A_4 = 0$$

$$\begin{matrix} 2 \\ 3 \end{matrix} \begin{bmatrix} B_3 & B_4 & B_5 \\ -1 & 0 & 0 \\ 0 & -1 & 0 \\ 0 & 0 & -1 \end{bmatrix} \rightarrow \text{cols are a basis for } N(A).$$

General:  $A \in \mathbb{R}^{m \times n}$  w/  $k$  lin ind cols

$$A = \underbrace{[A_1 \dots A_k]}_{\text{lin ind}} \underbrace{[A_{k+1} \dots A_n]}_{\text{lin dep on}}$$

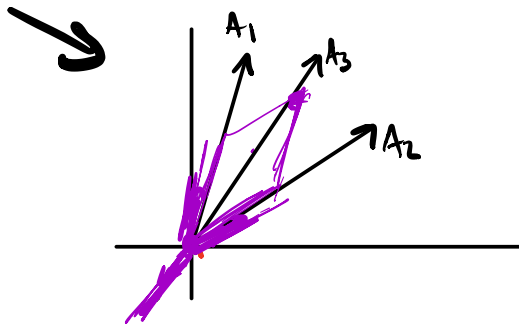
• find coords of  $A_{k+1} \dots A_n$  wrt.  $[A_1 \dots A_k]$

$$A_j = [A_1 \dots A_k | B_j] \quad j = k+1, \dots, n$$

$$B = [B_{k+1} \dots B_n]$$

$$N = \begin{matrix} k \\ n-k \end{matrix} \begin{bmatrix} B \\ -I \end{bmatrix} \quad \begin{matrix} N \in \mathbb{R}^{(k+(n-k)) \times (n-k)} \\ N \in \mathbb{R}^{(n \times (n-k))} \end{matrix}$$

•  $AN = 0 \leftarrow$  every column of  $N$  is in nullspace of  $A$



$$A \in \mathbb{R}^{2 \times 3}$$

$A \in \mathbb{R}^{m \times n}$  w  $k$  lin ind cols...  
 $\rightarrow$  find  $n-k$  vectors in nullspace

cols of  $N = \begin{bmatrix} B \\ -I \end{bmatrix}$

- cols are lin ind.
- $\exists z$  s.t.  $x = Nz$  for any  $x \in N(A)$

lin ind.  $\leftarrow$

$$\left. \begin{array}{l} \text{lindep } \exists z \neq 0 \\ Az = 0 \end{array} \right\}$$

Assume  $Nz = 0$   $\checkmark$   
 show  $z = 0$

$$- \underbrace{A_1 z_1 + \dots + A_n z_n}_{z_j} = 0$$

$$Nz = \begin{bmatrix} Bz \\ -z \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \Rightarrow z = 0$$

$$N = \begin{bmatrix} B \\ -I \end{bmatrix} = \begin{bmatrix} B_{k+1} & B_n \\ \hline -1 & 0 & 0 \\ \hline 0 & 0 & -1 \end{bmatrix}$$

if  $x \in \mathcal{N}(A)$  then  $\exists z$  s.t.  $x = Nz$  Span

$$A = [A_1 \dots A_k \mid A_{k+1} \dots A_n]$$

$$\left[ \begin{array}{c|c} A_1 \dots A_k & [A_{k+1} \dots A_n] \end{array} \right] \begin{array}{l} \downarrow B \\ \downarrow B \end{array} \quad x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

$$Ax = [A_1 \dots A_k \mid [I \mid B]] \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = [A_1 \dots A_k \mid \underline{x_1 + Bx_2}] = 0$$

$$x_1 + Bx_2 = 0 \Rightarrow -Bx_2 = +x_1$$

$$x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} -Bx_2 \\ x_2 \end{bmatrix} = \begin{bmatrix} B \\ -I \end{bmatrix} (-x_2) \quad z = -x_2$$

$A \in \mathbb{R}^{m \times n}$  w  $k$  lin ind cols...  
 $\rightarrow$  find  $n-k$  vectors in nullspace

$$\left. \begin{array}{l} \# \text{ of lin ind cols} + \dim \mathcal{N}(A) = n \\ \underline{k} \quad + \quad n-k \quad = \quad n \end{array} \right\}$$

Rank Nullity Thm:  $A \in \mathbb{R}^{m \times n}$  nullity

$$rk(A) + \dim \mathcal{N}(A) = n$$



Matrix rank:

col rank: # of lin ind cols.

row rank: # of lin ind rows.

col rank = row rank = rank

PROOF CLEVER. (END OF CLASS)

$$\boxed{\text{rk}(A) = \text{rk}(A^T) = \text{rk}(A^T A) = \text{rk}(A A^T)}$$

$A^T A, A A^T$ : Grammians      pseudo inverses

$$A \in \mathbb{R}^{m \times n} \quad [ \quad ]$$

$$A^T A \in \mathbb{R}^{n \times n} \quad A A^T \in \mathbb{R}^{m \times m}$$

$$\boxed{\text{rk}(A) = \text{rk}(A^T A):}$$

$\uparrow$   $\mathbb{R}^{m \times n}$        $\mathbb{R}^{n \times n}$

proof:  $N(A) = N(A^T A)$  ← rank-nullity

Assume  $Ax = 0 \Rightarrow \underbrace{A^T A x}_{0} = 0$  ✓

$$A^T A x = 0 \Rightarrow x^T A^T A x = 0 \quad |Ax| = 0$$

$$|Ax|^2 = 0 \quad Ax = 0 \quad \checkmark$$

A tall  $\begin{bmatrix} n \\ m \end{bmatrix}$   $m > n$

A fat  $m \begin{bmatrix} n \\ \end{bmatrix}$  ✓

$$A^T A = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} \quad A A^T = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$A^T A = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \quad A A^T = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$$

if  $A$  not square one of these Gramians is big & other is small.

if  $A$  has rank  $n$ :  $A^T A$  invertible

if  $A$  has rank  $m$ :  $A A^T$  invertible

$A$  is full col rank if its cols are lin ind  $rk(A)=n$

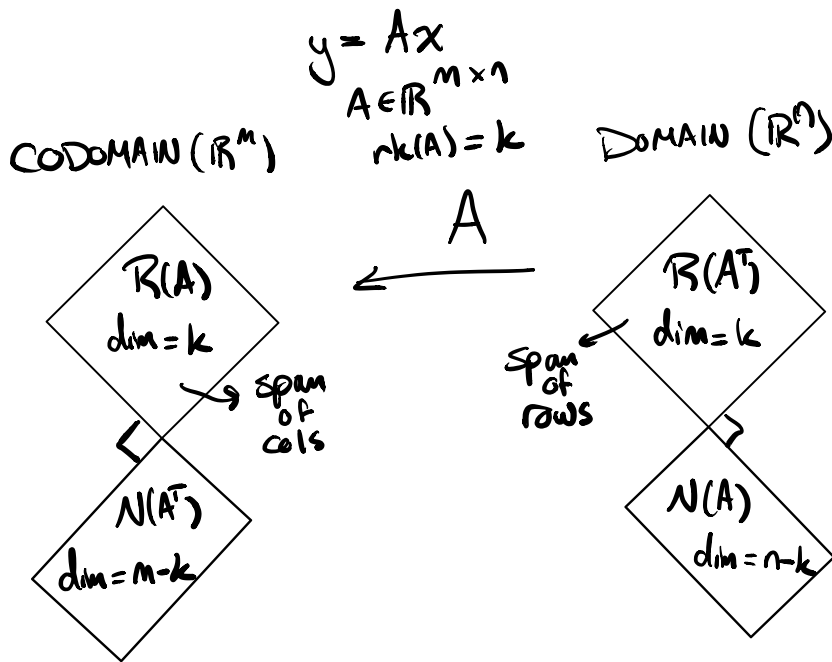
$A$  is full row rank if its rows are lin ind  $rk(A)=m$

Square matrix  $A$

- $A$  is invertible / non singular

- $A$  is full row & col rank ←

# Fundamental Theorem of Linear Algebra



$V \perp W$  if  $v \in V, w \in W \Rightarrow v^T w = 0$

$R(A) \perp N(A^T)$

$R(A^T) \perp N(A)$

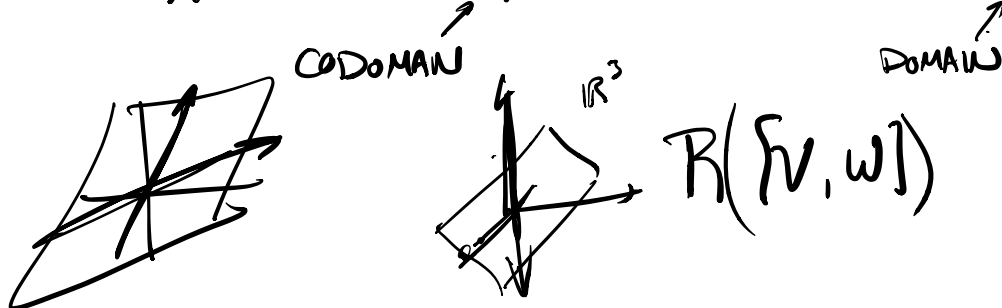
(vectors in  $N(A)$  orthogonal to rows.)

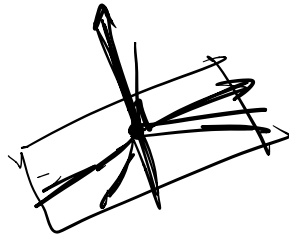
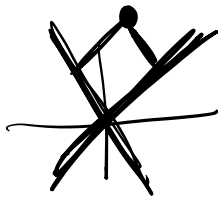
vector space Direct Sum of 2 subspaces

$V \oplus W = \{v+w \mid \forall v \in V, \forall w \in W\}$

$R(A) \oplus N(A^T) = \mathbb{R}^m$

$R(A^T) \oplus N(A) = \mathbb{R}^n$





# Systems of Linear Equations

A tall

→ A full col rank

→  $A^T A$  invertible

$$|y| = |A| |x|$$

y might not be in range of A.

**no solution**  
(overconstrained)

↓  
closest x.

$$\min_x |y - Ax|^2$$

⇒ LEAST SQUARES Regression

$$y = Ax$$

A square  
(invertible)

**a unique solution**

$$x = A^{-1} y$$

A fat.

A full row rank  
( $AA^T$ ) invertible

$$|y| = [A] |x|$$

**a continuum or subspace of solns**

if  $y = Ax$   
∃  $\text{Span}(N) = N(A)$

$x + Nz$  is also a soln for any z

$$Ax + \underbrace{AN}_{0} z$$

Smallest x.

$$\min_x |x|^2$$

s.t.  $y = Ax$

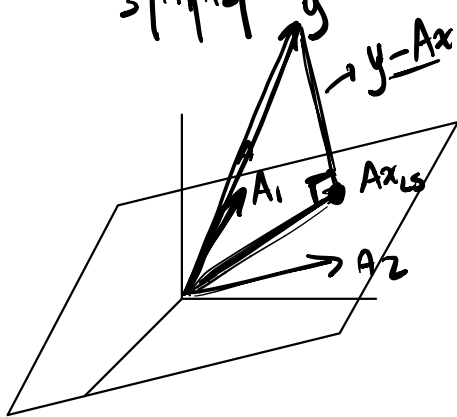
MIN. NORM SOLN.

# LEAST SQUARES

$$\rightarrow \min_x \|y - Ax\|^2 = (y - Ax)^T (y - Ax)$$

ex.

$$A = \frac{1}{3} \begin{bmatrix} A_1 \\ A_2 \end{bmatrix}$$



$$\min_x \frac{1}{2} y^T y - z^T A x + x^T A^T A x = J(x)$$

$$\frac{\partial J}{\partial x} = 0: -z^T A + 2x^T A^T A = 0$$

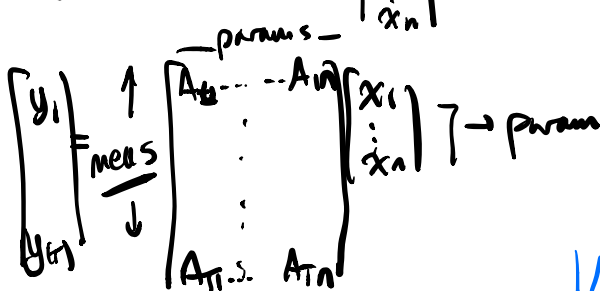
$$x^T (A^T A) = y^T A$$

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

$$x_{LS} = (A^T A)^{-1} A^T y$$

$$\text{Proj}_A y = A x_{LS} = A (A^T A)^{-1} A^T y$$

$$y(t) = [A_{t1} \dots A_{tn}] \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} \quad \text{model}$$



# MIN NORM:

$$y = [-A \quad -1] x$$

$$x = x_0 + N z$$

$$\min_x \|x\|^2$$

$$\text{s.t. } y = Ax \leftarrow$$

$$x_{mn} = A^T (A A^T)^{-1} y$$

$$y = A x_{mn} = A A^T (A A^T)^{-1} y = y$$

arbitrary lin comb.

$$\rightarrow x = x_{mn} + N z \quad \text{RN} = U(A)$$

$$\|x\|^2 = x^T x = x_{mn}^T x_{mn} + 2 x_{mn}^T N z + z^T N^T N z$$

$$= y^T \overbrace{(A A^T)^{-1}}^I A A^T (A A^T)^{-1} y + 2 y^T (A A^T)^{-1} A N z + z^T N^T N z$$

$$\|N z\|^2$$

$$(z^T N^T) (Nz) = y^T (AA^T)^{-1} y + \|Nz\|^2$$

$$\Rightarrow z=0$$

$$(AA^T)^{-1} (AA^T)$$

$$\|x\|^2 = y^T (AA^T)^{-1} y$$

$$\boxed{x_{LS} = (A^T A)^{-1} A^T y}$$

$$\boxed{x_{mn} = A^T (AA^T)^{-1} y}$$

$(A^T A)^{-1} A^T$ : left inverse

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

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

$(A^T (AA^T)^{-1})$ : right inverse

$$AA^T (AA^T)^{-1} = I$$

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

$$\underline{\underline{(A^T A)^{-1} A^T}} \quad \underline{\underline{A^T (AA^T)^{-1}}}$$

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

Moore Penrose Pseudo inverse:

$$x = A^+ y \leftarrow \text{min norm least squares solution}$$

Singular Value Decomposition