

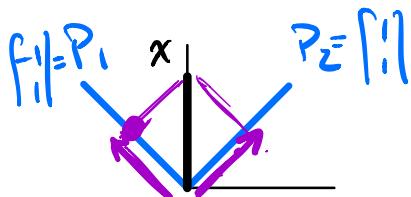
Basis : set of vectors
 • span a space
 • lin.ind.

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

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

$$P = \{P_1, \dots, P_n\}$$

Ex: $P = [P_1, P_2]$



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

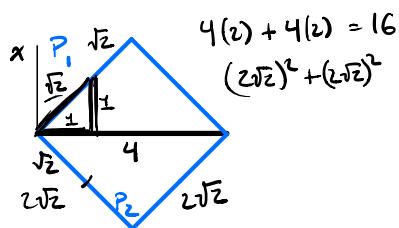
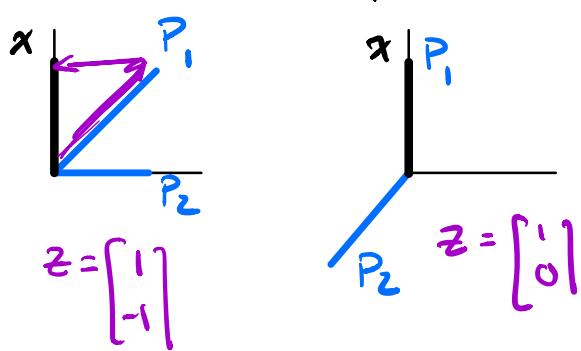
$$z = P^{-1}x$$

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

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

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

↓
basis vectors ↓
coords of x
w.r.t. the basis



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

changing basis = coordinate transformation

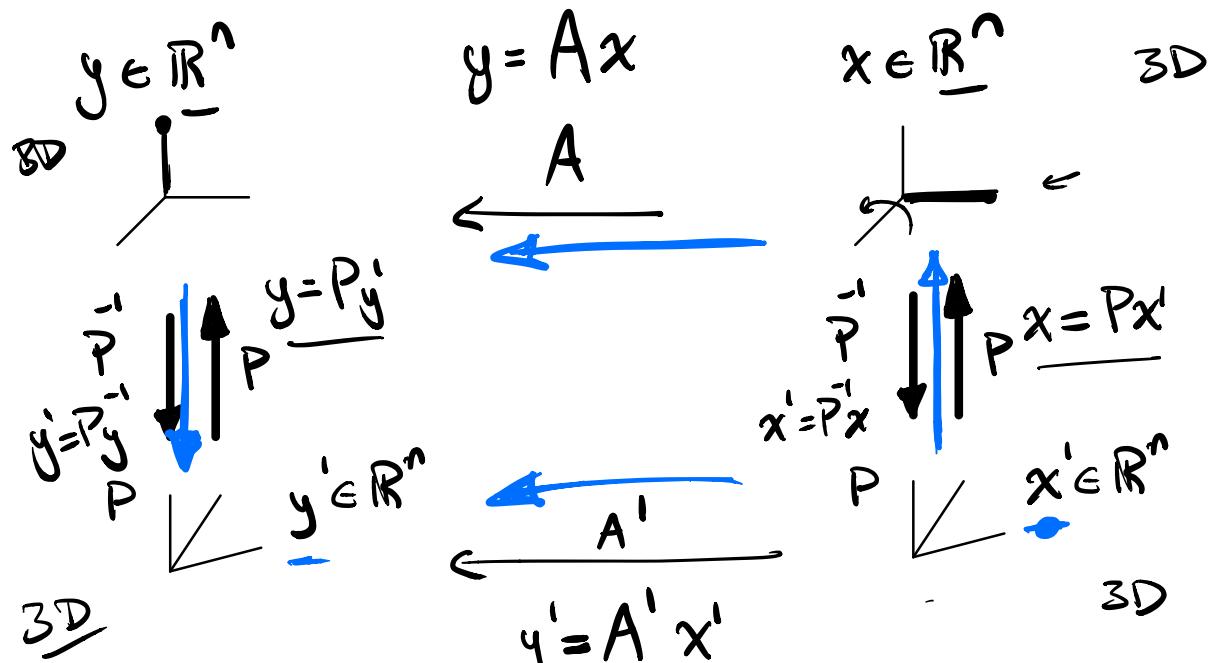
P : new basis, coordinate transform from the standard basis

$| x = Px' \rightarrow \text{coords w.r.t. } P.$
 ↑
 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



$$y = Ax$$

$$Py' = APx'$$

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

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

Similarity transformation on A
 $A \in A'$ are similar

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

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

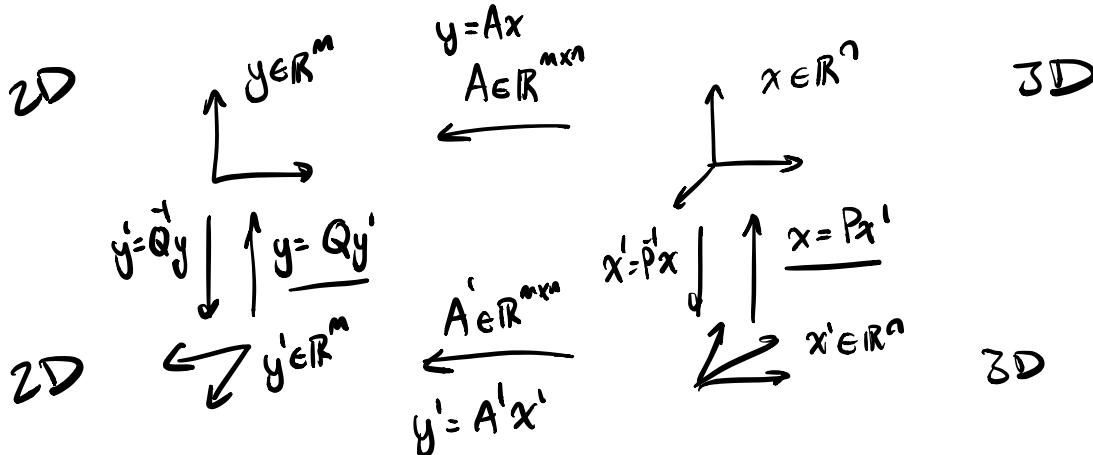
Note: in general mattices don't commute

$$AB \neq BA$$

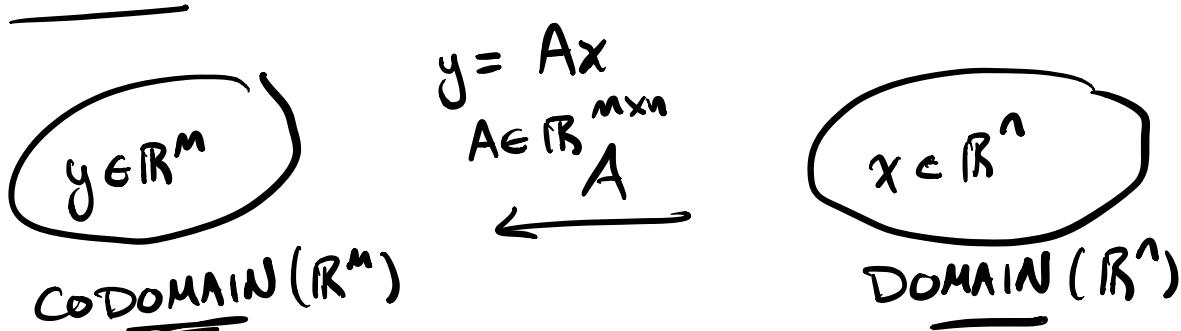
2 rotations... R_1, R_2

\rightarrow [distributivity
associativity
commutativity]

More general



$$\begin{aligned} y &= Ax \\ Qy' &= APx' \Rightarrow y' = Q^{-1}APx' \quad A' = Q^{-1}AP \\ (\text{Similarity}) \quad \text{when } Q = P. \end{aligned}$$



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

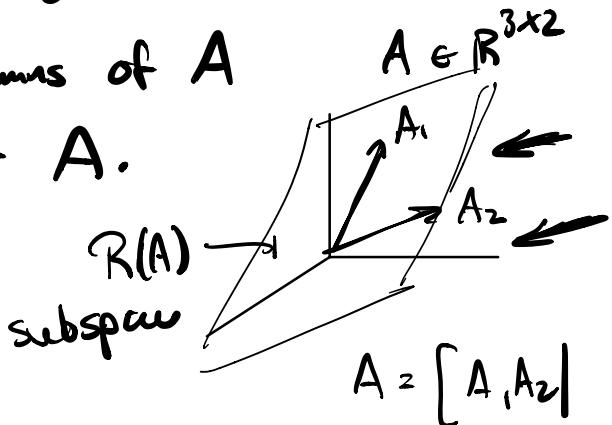
$$\text{Range}(A^T) = R(A^T) \subseteq \text{DOMAIN}$$

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

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 $N(A) = \{x \in \mathbb{R}^n \mid Ax = \underline{0}, x \in \mathbb{R}^n\}$
 $\underline{0}$ vector

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

$\exists x$ s.t. $y = Ax$ 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 $x \neq x'$ $Ax = y = Ax'$

$$\Rightarrow A(\underbrace{x - x'}_{\neq 0}) = 0$$

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

Some properties of $N(A)$:

- if $x \in N(A)$, then $x \perp$ 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

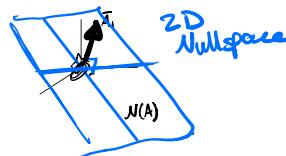
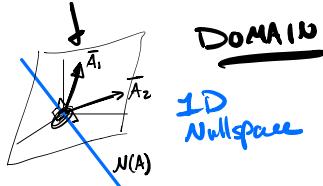
$$A \in \mathbb{R}^{2 \times 3}$$

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

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

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

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

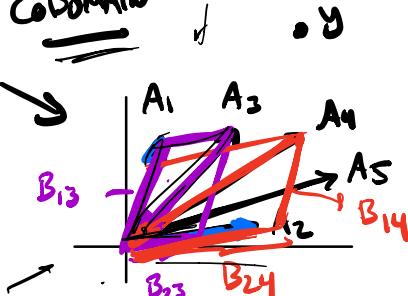


Construct a basis for $N(A)$:

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

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

CODOMAIN



$y = Ax$ non unique solutions

cols of A are lin dep.

$$\bullet A = \left[\begin{array}{c|cc} A_1 \ A_2 & A_3 \ A_4 \ A_5 \end{array} \right] \quad \begin{array}{l} \text{lin ind} \\ \text{cols} \end{array} \quad \begin{array}{l} \text{lin dep} \\ \text{cols} \end{array}$$

$$y = A(x + z) = Ax + Az \quad z \in N(A)$$

$$y = \underbrace{\begin{bmatrix} A_1 & \cdots & A_n \end{bmatrix}}_{\star} \left(\begin{bmatrix} x_1 \\ \vdots \\ x_m \end{bmatrix} \right) +$$

$$A_3 = [A_1 A_2 | B_3] \xrightarrow{\substack{\text{coords of} \\ A_3 \text{ w.r.t. } A_1, A_2}} B_3 = \begin{pmatrix} B_{13} \\ B_{23} \end{pmatrix}$$

$$A_4 = [A_1 A_2 | B_4] \xrightarrow{\substack{\text{coords of} \\ A_4 \text{ w.r.t. } A_1, A_2}} B_4 = \begin{pmatrix} B_{14} \\ B_{24} \end{pmatrix}$$

$$A_5 = [A_1 A_2 | B_5] \xrightarrow{\substack{\text{w.r.t.} \\ A_1, A_2}} B_5$$

Consider:

$$\left[\begin{array}{c|cc} A_1 & A_2 & A_3 & A_4 & A_5 \\ \hline A_1 & A_2 & B_3 & -1 & 0 \\ & & \begin{pmatrix} 0 & 0 & 0 \end{pmatrix} & & \end{array} \right] = \left[\begin{array}{c|c} A_1 & A_2 \\ \hline A_1 & B_3 - A_3 = 0 \end{array} \right]$$

$$\left[\begin{array}{c|cc} A_1 & A_2 & A_3 & A_4 & A_5 \\ \hline A_1 & A_2 & B_4 & 0 & -1 \\ & & \begin{pmatrix} 0 & 0 & 0 \end{pmatrix} & & \end{array} \right] = \left[\begin{array}{c|c} A_1 & A_2 \\ \hline A_1 & B_4 - A_4 = 0 \end{array} \right]$$

$\xrightarrow{-3}$

$$\left[\begin{array}{ccc} B_3 & B_4 & B_5 \\ -1 & 0 & 0 \\ 0 & -1 & 0 \\ 0 & 0 & -1 \end{array} \right] \rightarrow \text{cols are a basis for } N(A).$$

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

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

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

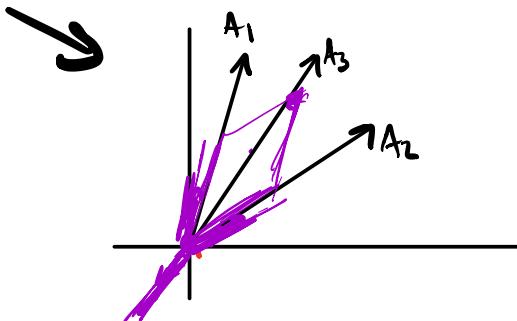
$$A_j = [A_1 \dots A_k | B_j] \quad j = k+1, \dots, n$$

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

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

- $AN = O \quad \rightarrow$ 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...

→ 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:

Assume $Nz = 0$

Show $z = 0$

l indep $\exists z \neq 0$

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

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

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

If $x \in N(A)$ then $\exists z$ s.t. $x = Nz$ Span

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

$$\left[\underbrace{A_1 \dots A_k}_{\text{lin. ind.}} | \underbrace{[A_1 \dots A_k] [B_{k+1} \dots B_n]}_B \right] \quad x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

$$Ax = [A_1 \dots A_k] \left[\begin{array}{c|c} I & B \end{array} \right] \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = [A_1 \dots A_k] \begin{bmatrix} x_1 + Bx_2 \\ x_2 \end{bmatrix} = 0$$

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

$$x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} -Bx_2 \\ x_2 \end{bmatrix} = \begin{bmatrix} B \\ -I \end{bmatrix} \begin{bmatrix} -x_2 \\ 1 \end{bmatrix} \quad z = -x_2$$

$A \in \mathbb{R}^{m \times n}$ w k lin ind cols...
 → Sind $n-k$ Vektoren in Nullspace

$$\begin{aligned} \# \text{ of lin ind cols} + \dim N(A) &= n \\ \underbrace{k}^{\text{rank}(A)} + n-k &= n \end{aligned}$$

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

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

Matrix rank:

col rank: # of lin ind cols.

row rank: # of lin ind rows.

col rank = rowrank = 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

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

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

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

$\underset{\mathbb{R}^{M \times n}}{A}$ $\underset{\mathbb{R}^{n \times n}}{A^T A}$

proof: $N(A) = N(A^T A) \leftarrow \text{rank-nullity}$

Assume $Ax = 0 \Rightarrow \underset{0}{A^T A x} = 0$

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

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

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

A fat $\begin{bmatrix} n \\ m \end{bmatrix} \quad n > m$

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

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

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

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

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 $\text{rk}(A)=n$

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

Square matrix A

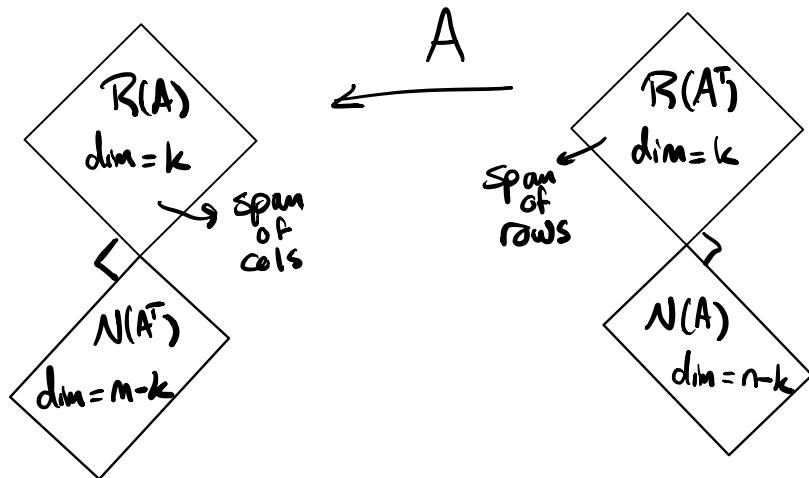
- A is invertible / non singular
- A is full row & col rank \leftarrow

Fundamental Theorem of Linear Algebra

$$y = Ax$$

$A \in \mathbb{R}^{m \times n}$
 $\text{rk}(A) = k$

CODOMAIN (\mathbb{R}^m) DOMAIN (\mathbb{R}^n)



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

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

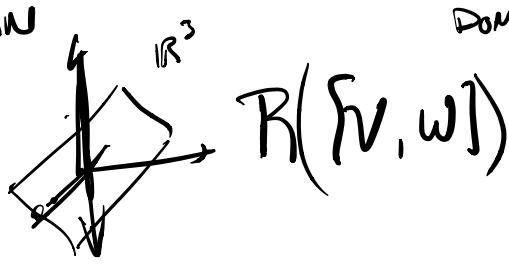
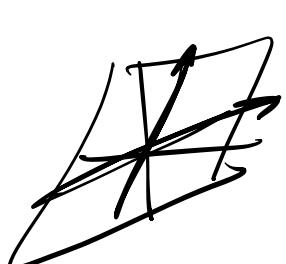
$R(A^T) \perp N(A)$
 (vectors in $N(A)$ orthogonal to rows.)

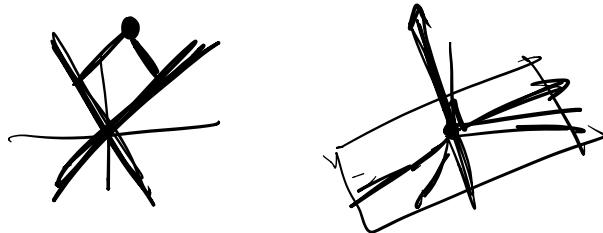
$\xrightarrow{\text{vector space}}$ Direct Sum of 2 subspaces

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

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

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





Systems of Linear Equations

A tall

→ A full col rank

→ $A^T A$ invertible

$$\begin{bmatrix} y \\ \vdots \\ y \end{bmatrix} = \begin{bmatrix} A \\ \vdots \\ A \end{bmatrix} \begin{bmatrix} x \\ \vdots \\ x \end{bmatrix}$$

y might not be in range of A.

no solution
(overconstrained)

↓
closest x.

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

⇒ LEAST REGRESSION
SQUARES

$$y = Ax$$

A square
(invertible)

a unique
solution

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

A fat.

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

$$\begin{bmatrix} y \\ \vdots \\ y \end{bmatrix} = \begin{bmatrix} A \\ \vdots \\ A \end{bmatrix} \begin{bmatrix} x \\ \vdots \\ x \end{bmatrix}$$

a continuum
or subspace
of solns

if $y = Ax$
 $\in \text{span}(N) = N(A)$

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

$$Ax + \underbrace{ANz}_0$$

smallest x.

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

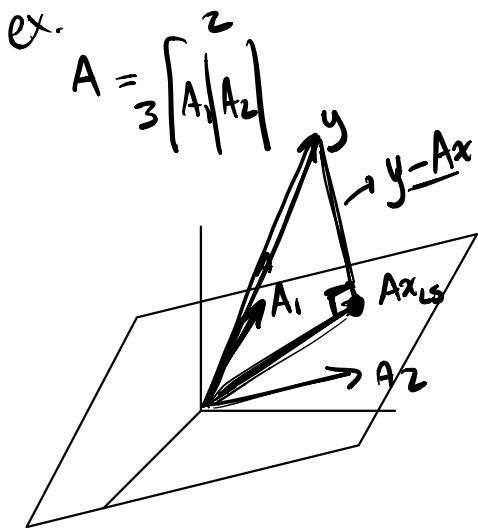
$$x \text{ s.t. } y = Ax$$

MIN. NORM SOLN.

LEAST SQUARES

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

ex.



$$\min_x \|y - Ax\|^2 = J(x)$$

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

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

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

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

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

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

$$\begin{bmatrix} y_1 \\ \vdots \\ y_n \end{bmatrix} \stackrel{\text{meas}}{\uparrow} \begin{bmatrix} A_{11} & \cdots & A_{1n} \\ \vdots & \ddots & \vdots \\ A_{n1} & \cdots & A_{nn} \end{bmatrix} \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} \stackrel{\text{param}}{\rightarrow}$$

MIN NORM:

$$y = [-A - I] \begin{bmatrix} x \\ z \end{bmatrix}$$

$$x = x_0 + Nz$$

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

x

st.

$$y = Ax$$

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

$$y = Ax_{mn}$$

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

$$= y$$

$$\rightarrow x = x_{mn} + Nz \quad R(N) = N(A)$$

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

I

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

~~$$+ 2 y^T (A A^T)^{-1} A N z$$~~

~~$$+ z^T N^T N z$$~~

$$|Nz|^2$$

$$(z^T N^T) \underbrace{(Nz)}_{\text{---}} = y^T (A A^T)^{-1} y + |Nz|^2$$

$$(A A^T)^{-1} (A A^T) \Rightarrow z=0$$

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

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

$$\boxed{x_{ML} = \bar{A}^T (\underline{A A^T})^{-1} y}$$

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

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

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

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

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

Moore Penrose
Pseudo inverse:

$$A^T (A^T A)^{-1} \quad x = A^T y \leftarrow \begin{matrix} \min \text{ norm} \\ \text{least squares} \\ \text{solution} \end{matrix}$$

Singular Value Decomposition