

Systems of Eqs: $A \in \mathbb{R}^{m \times n}$

$m > n$
A is tall
(lin ind cols)

$$y = Ax$$

Applications

data analysis
if redundant parameters
→ lin dep cols
prob no soln
(unless $y \in R(A)$)

Find x to minimize

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

$$\Rightarrow \frac{\partial}{\partial x} |y - Ax|^2 = \frac{\partial}{\partial x} (y - Ax)^T (y - Ax)$$

$$= \frac{\partial}{\partial v} v^T v \Big|_{y-Ax} \frac{\partial (y-Ax)}{\partial x}$$

$$= -2(y - Ax)^T (+A) = 0$$

$$= y^T A - x^T A^T A = 0$$

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

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

A square invertible
 $m = n$

$$y = Ax$$

$$x = \bar{A}^T y$$

unique soln

$m < n$
A is fat.
→ (lin ind rows)

$$y = Ax$$

Applications

constraints
if redundant constraints
→ lin dep rows

many solns
subspace of sols

If \bar{x} satisfies
 $y = A\bar{x} \nmid z \in N(A)$

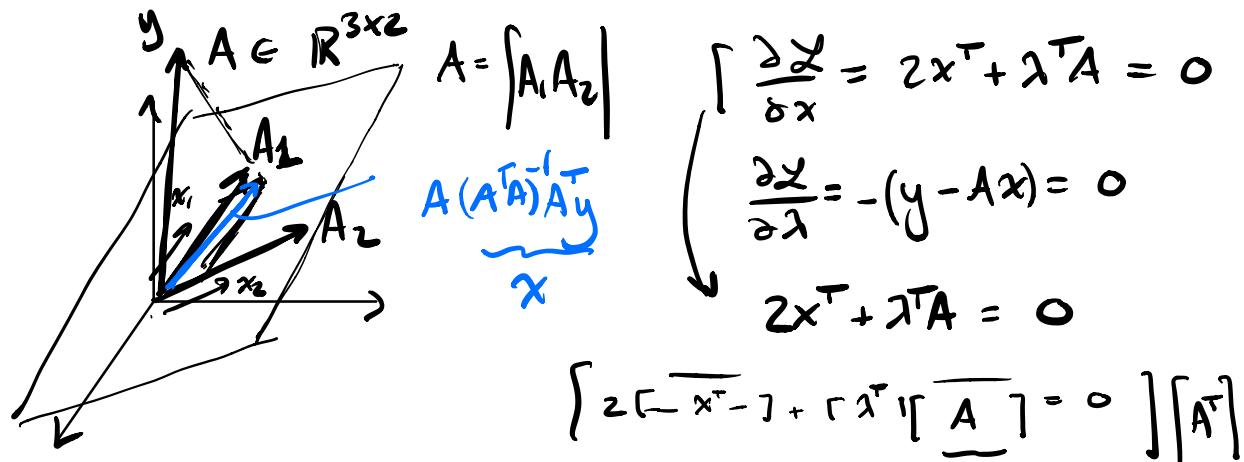
$y = A(\bar{x} + z) = A\bar{x}$
set of solns

$$\bar{x} + N(A)$$

Find x
 $\min_x |x|^2 \Leftarrow$
s.t. $y = Ax \}$

Lagrangian: $y \geq Ax$

$$\mathcal{L} = x^T x - \lambda^T (y - Ax)$$



Summary:

$$\text{LS: } \hat{x} = \underline{(A^T A)^{-1}} A^T y$$

(tall A) $\underline{[A^T]^{-1}}$

$$2x^T + \lambda^T A A^T = 0$$

\uparrow invertible

$$y = Ax$$

$$2y^T + \lambda^T A A^T = 0$$

$$\lambda^T = -2y^T (A A^T)^{-1}$$

$$\begin{aligned} \text{MIN: } \hat{x} &= \underline{A^T} \underline{(A A^T)^{-1}} y \\ \text{NORM: } (fat A) &\quad \rightarrow \underline{[A^T]^{-1}} \underline{A^T} \\ &\quad z \left[\begin{array}{c|c} 3 & 2 \\ \hline 4 & 2 \\ 2 & 2 \end{array} \right] \underline{A^T} \end{aligned}$$

$$x = \begin{vmatrix} x_1 \\ x_2 \\ x_3 \end{vmatrix} = 3 \int \begin{vmatrix} y_1 \\ y_2 \end{vmatrix} = 3 \int \begin{vmatrix} 2 \\ 2 \end{vmatrix} \frac{1}{2} \underline{\underline{X}}^{-1} \begin{vmatrix} y_1 \\ y_2 \end{vmatrix}$$

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

$$(A z)^T = [0 \dots 0]$$

$$\begin{aligned} |x^* + z|^2 &= x^{*T} x^* + z^T z \\ |x^*|^2 &\leq |x^*|^2 + \underline{|z|^2} \end{aligned}$$

$$\begin{aligned} 2x^T &= -\lambda^T A \\ 2x^T &= 2y^T (A A^T)^{-1} A \\ \Rightarrow \boxed{x^* = A^T (A A^T)^{-1} y} &\quad \leftarrow \end{aligned}$$

can show

$$|x^*| \leq |x^* + z| \text{ if } z \in N(A)$$

$$\begin{aligned} |x^*|^2 &\leq |x^* + z|^2 \\ &\downarrow \\ &(x^* + z)^T (x^* + z) \\ \nearrow & \searrow \\ x^{*T} x^* + z^T x^* + x^{*T} z + z^T z \end{aligned}$$

Rank & Fund of Lin Alg.

Review: row rank (A) = # of lin ind rows = $\dim R(A)$
col rank (A) = # of lin ind cols = $\dim (R(A))$

$A = [A_1 \dots A_n]$ $R(A)$ = lin comb of cols.

$A = \begin{bmatrix} -\bar{a}_1^T \\ -\bar{a}_m^T \end{bmatrix}$ $A^T = \begin{bmatrix} 1 & \dots & 1 \\ \bar{a}_1 & \dots & \bar{a}_m \end{bmatrix}$ $R(A^T)$ = lin combs of rows

row rank = col rank = rank $\quad \text{rank}(A) = \text{rank}(A^T)$
 $\text{rank}(A) = rk(A)$ $\quad rk(A) = rk(A^T)$

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

$$\underbrace{rk(A)}_k + \underbrace{\dim N(A)}_{n-k} = n$$

Proof: by construction $A \in \mathbb{R}^{m \times n}$

$A \rightarrow$ take a set of lin ind cols of A that span $R(A)$

$$A = \begin{bmatrix} \underline{B_1} & \underline{B_2} \end{bmatrix} \Rightarrow \underline{B_2} = \underline{B_1} V \quad \text{for some matrix of coeffs } V$$

$\begin{array}{l} \text{cols are lin ind.} \\ \vdots \text{span } R(A) \end{array}$ $\downarrow \text{cols are lin dep on cols of } B_1$ $= B_1 \begin{bmatrix} V_1 & \dots & V_{n-k} \end{bmatrix}$

$$B_1 \in \mathbb{R}^{m \times k}$$

$$B_2 \in \mathbb{R}^{m \times (n-k)}$$

$$V \in \mathbb{R}^{k \times (n-k)}$$

Many different choices for B . $A = \begin{bmatrix} 1 & 1 & 1 & 1 & 1 & 1 \end{bmatrix}$

$$A = [B_1 B_2 V] = B_1 [I V]$$

$$N = \begin{bmatrix} -V \\ I \end{bmatrix} \quad AN = B_1 [I V] \begin{bmatrix} -V \\ I \end{bmatrix} = B_1 [-V + V] = 0$$

explicit construction of a basis for nullspace
of A . \rightarrow cols of $N = \begin{bmatrix} -V \\ I \end{bmatrix}$

- lin ind $Nz = \begin{bmatrix} -V \\ I \end{bmatrix} z = \begin{bmatrix} -Vz \\ z \end{bmatrix} = 0 \rightarrow z = 0$

- if $Ax = 0$
 $\Rightarrow x \in \text{span}(N)$

$$x = \begin{bmatrix} -V \\ I \end{bmatrix} x_2$$

$$Ax = [B_1 B_2 V] \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = 0$$

$$= B_1 x_1 + B_2 V x_2 = 0$$

$$\Rightarrow B_1 x_1 = -B_2 V x_2$$

$$B_1^T (B_1 x_1 = -B_2 V x_2)$$

$$B_1^T B_1 x_1 = -B_1^T B_2 V x_2$$

since cols of B_1 are lin ind.

cols of N are
a basis

for nullspace
of A

$B_1^T B_1 \rightarrow$ invertible

$$x_1 = -V x_2$$

$$x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} -V x_2 \\ I x_2 \end{bmatrix} = \begin{bmatrix} -V \\ I \end{bmatrix} x_2$$

Row perspective on Nullspace ...

$$A = \begin{bmatrix} -\bar{a}_1^T \\ -\bar{a}_n^T \end{bmatrix} \quad \text{if } x \in N(A)$$

$$\bar{a}_i^T x = 0 \quad \forall i$$

$$Ax = \begin{bmatrix} \bar{a}_1^T x \\ \bar{a}_n^T x \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \quad x \in N(A) \text{ then}$$

$x \perp$ rows of A .

$$\Rightarrow \underline{R(A^T)} \perp \underline{N(A)}$$

Properties of Rank $f(x) = Ax \quad A \in \mathbb{R}^{m \times n}$

- $\text{rk}(A) \leq \min(m, n)$

$$\text{rk}(A) = \min(m, n) \Rightarrow \text{full rank matrix}$$

full rank fat matrix has m lin ind rows

full rank tall matrix has n lin ind cols

not full rank \rightarrow rank deficient

- $\text{rk}(A) = 0 \Rightarrow A = 0$

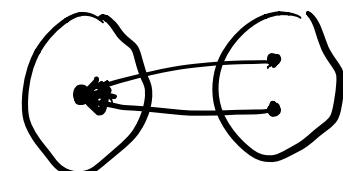
- A has rank n : "full col rank" has to be square or tall.

f is one-to-one / injective $\rightarrow \underline{N(A) = \{0\}}$

every x has a unique $y = f(x)$ nullspace is trivial

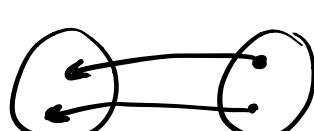
Not injective injective $= Ax$ if $z \in N(A)$

$$z \neq 0$$



CODOM

DOM



CODOM

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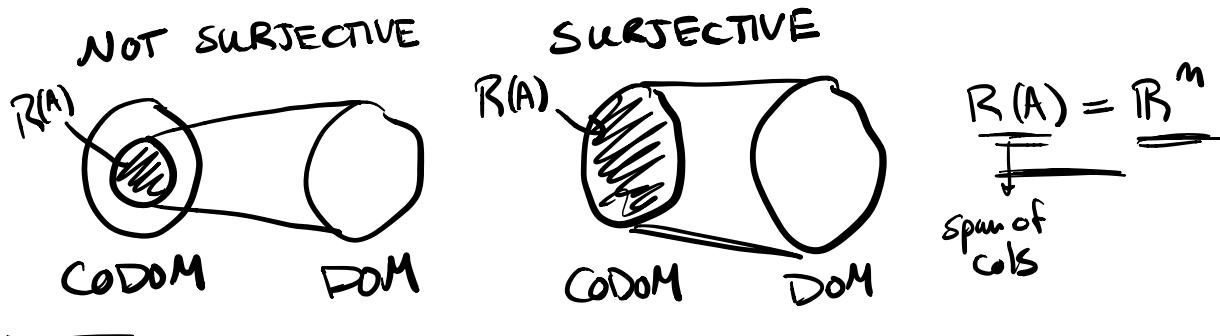
$$Ax = y = Ax + z$$

both $x, x+z$ give you y

- A has rank m : "full row rank" has to be square or fat

f is onto / surjective

every y has at least one x s.t. $y = f(x)$



full row rank

$$\underline{R(A)} = \underline{R}^m$$

span of
cols

full col
rank

$$\underline{N(A)} = 0$$

subspace \perp
to rows of A

- A is square $M=n$

A is invertible \iff rank n

Note: rank 1 matrices / dyads $\frac{1}{n}$ col repeated w different scalings

$$A = \underbrace{ab^T}_{\text{outer prod}} \quad a = \begin{bmatrix} a_1 \\ \vdots \\ a_m \end{bmatrix} \quad b = \begin{bmatrix} b_1 \\ \vdots \\ b_n \end{bmatrix}$$

or equiv
 $\frac{1}{n}$ row repeated w diff. scalings

$$A = \begin{bmatrix} ab_1 & \dots & ab_n \end{bmatrix} = \begin{bmatrix} a_1 & b^T \\ \vdots & \vdots \\ a_m & b^T \end{bmatrix} \leftarrow$$

- $B \in \mathbb{R}^{n \times k}$
 $\text{rk}(AB) \leq \min(\text{rk}(A), \text{rk}(B))$
- $B \in \mathbb{R}^{n \times k}, \text{rk}(B)=n$ unproven
- $C \in \mathbb{R}^{l \times m}, \text{rk}(C)=m$
 $\text{rk}(CA) = \text{rk}(A)$
- Sylvester's Rank inequality
 $\text{rk}(A) + \text{rk}(B) - n \leq \text{rk}(AB)$
- Frobenius
 $\text{rk}(AB) + \text{rk}(BC) \leq \text{rk}(B) + \text{rk}(ABC)$
- Subadditivity:
 $\text{rk}(A+B) \leq \text{rk}(A) + \text{rk}(B)$
 \Rightarrow a rank- k matrix can be written
as a sum of k rank-1 matrices
and no fewer

$$A = \begin{bmatrix} a_1 & a_2 & a_3 = a_1 + a_2 \end{bmatrix} = \begin{bmatrix} a_1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} + \begin{bmatrix} a_2 \end{bmatrix} \begin{bmatrix} 0 & 1 & 1 \\ 1 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

$$\text{rk}(A) = 2.$$

- $\text{rk}(A^T A) = \text{rk}(A A^T) = \underline{\text{rk}(A)} = \underline{\text{rk}(A^T)}$
- $A^T A, A A^T$: "Gramians"
- $x \in N(A^T A)$ iff $x \in N(A)$
- if $x \in N(A)$: $Ax = 0$ $A^T A x = 0 = A^T 0 = 0$
- if $x \in N(A^T A)$ $A^T A x = 0$ $x^T A^T A x = 0$
- $$|\underline{Ax}|^2 = 0 \quad Ax = 0$$
- $x \in N(A)$

Defs: vector space \mathcal{W}

Orthogonal Complement

subspace $\mathcal{V}_1 \subseteq \mathcal{W}$

$$\mathcal{V}_1^\perp = \{w \in \mathcal{W} \mid w^T v = 0, v \in \mathcal{V}_1\}$$

set of all vectors \perp to all vectors in \mathcal{V}_1

Direct Sum:

$$\mathcal{V}_1 \subseteq \mathcal{W} \quad \mathcal{V}_2 \subseteq \mathcal{W}$$

$$\overline{\mathcal{V}_1 \oplus \mathcal{V}_2} = \{w \in \mathcal{W} \mid w = v_1 + v_2, v_1 \in \mathcal{V}_1, v_2 \in \mathcal{V}_2\}$$

span of both spaces

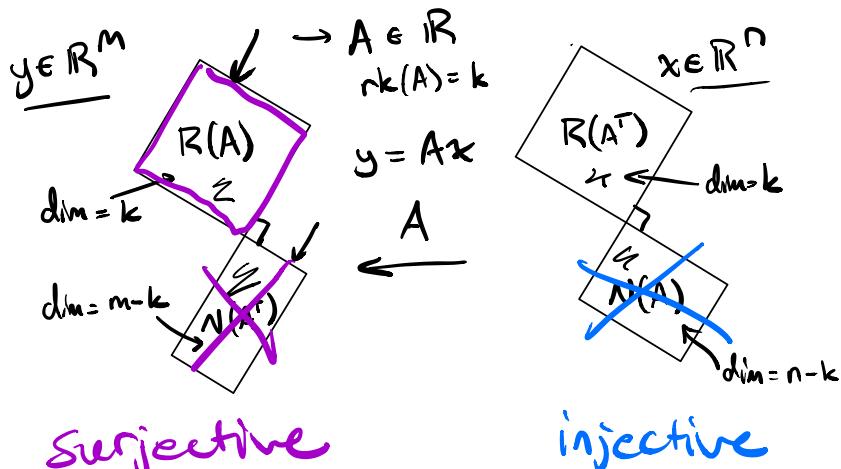
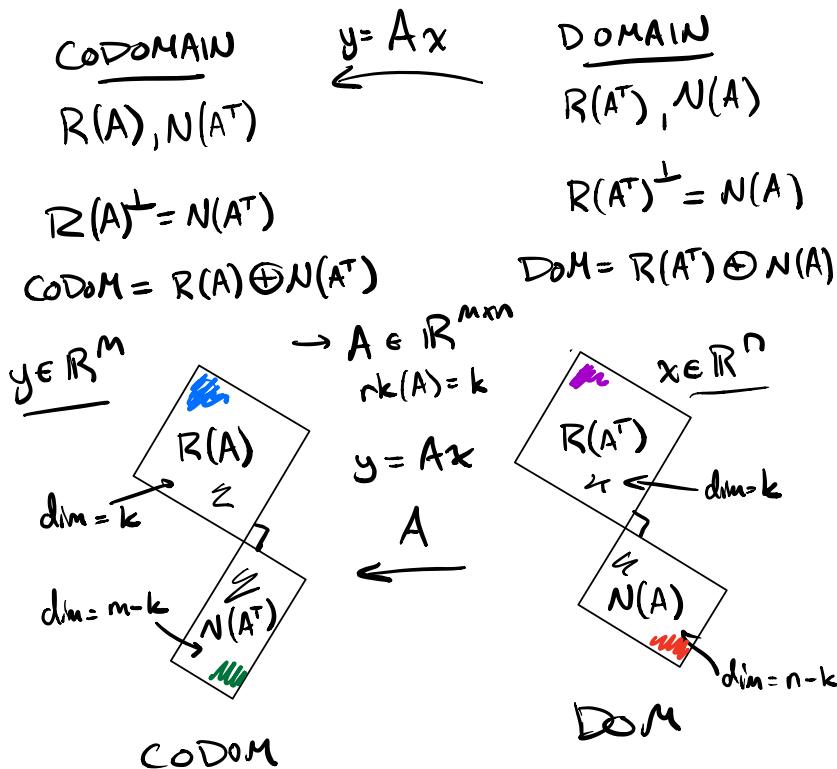
$$\mathcal{V} \subseteq \mathcal{W} \quad \mathcal{W} = \mathcal{V} \oplus \mathcal{V}^\perp$$

FUND. THM OF LIN ALG: $A \in \mathbb{R}^{m \times n}$

4 subspaces: $R(A)$, $R(A^T)$, $N(A)$, $N(A^T)$

right
nullspace

left
nullspace



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

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

$$A = [A_1 \ A_2 \ A_3] = \begin{bmatrix} \bar{a}_1^T \\ \bar{a}_2^T \end{bmatrix}$$

