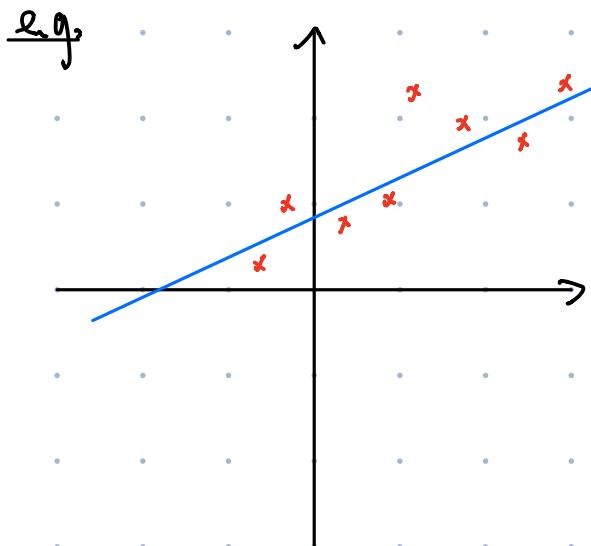


Office hour: Thur. ~~12-1:30~~ → Fri. 12-1:30 PDT.

## Least square problem:



Given data

$$\{(x_1, y_1), \dots, (x_n, y_n)\}.$$

Want to find

$$y = ax + b$$

$$y_1 \sim ax_1 + b \\ y_2 \sim ax_2 + b \\ \vdots$$

approximating the data.

i.e. find  $a, b$  s.t.

$$\begin{bmatrix} x_1 & 1 \\ x_2 & 1 \\ \vdots & \vdots \\ x_n & 1 \end{bmatrix} \begin{bmatrix} a \\ b \end{bmatrix}$$

"closest" to

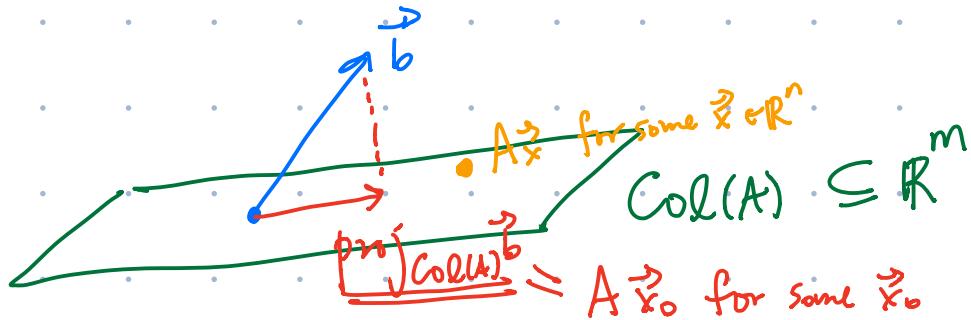
$$\begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix}$$

Def:  $A$ :  $m \times n$  matrix,  $\vec{b} \in \mathbb{R}^m$

A Least square soln of " $A\vec{x} = \vec{b}$ " is an  $\vec{x}_0 \in \mathbb{R}^n$

s.t.

$$\|\vec{b} - A\vec{x}\| \leq \|\vec{b} - A\vec{x}_0\| \quad \forall \vec{x} \in \mathbb{R}^n.$$



For any  $\vec{x}_0 \in \mathbb{R}^n$  s.t.  $A\vec{x}_0 = \text{proj}_{\text{Col}(A)} \vec{b}$ ,

We have  $\|\vec{b} - A\vec{x}_0\| \leq \|\vec{b} - A\vec{x}\| \quad \forall \vec{x} \in \mathbb{R}^n$ .

$$\Rightarrow b - A\vec{x}_0 = b - \text{proj}_{\text{Col}(A)} \vec{b} \in \text{Col}(A)^\perp$$

$$\Leftrightarrow \underline{A^T(b - A\vec{x}_0)} = 0 \quad (\because \text{Col}(A)^\perp = \text{Null}(A^T))$$

$$\Rightarrow \underline{A^T A \vec{x}_0 = A^T \vec{b}}$$

"normal eq'" for  $A\vec{x} = \vec{b}$ .

Thm:  $A^T A \vec{x}_0 = A^T \vec{b}$  always has a sol<sup>n</sup>, and the sol<sup>n</sup>(s) are the least square sol<sup>n</sup> of  $A\vec{x} = \vec{b}$ .

Thm  $A: \overset{\text{real}}{m \times n}$ . The following are equivalent:

- 1)  $A\vec{x} = \vec{b}$  has a unique least square sol<sup>n</sup>
- 2)  $A^T A$  is invertible
- 3) columns of  $A$  are l.i.  $\Leftrightarrow \text{Null}(A) = \{\vec{0}\}$

pf. 1)  $\Leftrightarrow$  2) clear.

2)  $\Leftrightarrow$  3):  $\text{Null}(A^T A) \supseteq \text{Null}(A) \Rightarrow "2) \Rightarrow 3)"$

$$\text{Null}(A^T A) \subseteq \text{Null}(A)$$

$$A^T A \vec{v} = \vec{0} \quad \langle A\vec{v}, A\vec{v} \rangle$$

$$0 = \begin{bmatrix} \vec{v}^T & A^T & A \vec{v} \end{bmatrix} = \begin{bmatrix} \vec{v} \\ A\vec{v} \\ A\vec{v} \end{bmatrix}^T \begin{bmatrix} \vec{v} \\ A\vec{v} \\ A\vec{v} \end{bmatrix} \stackrel{\parallel}{=} \vec{v}^T \vec{v} \Rightarrow \vec{v} = \vec{0} \quad \square$$

Rmk When  $A$  has full columns,  $A = QR$

the unique least square sol<sup>b</sup> of  $\vec{A}\vec{x} = \vec{b}$  is given by:

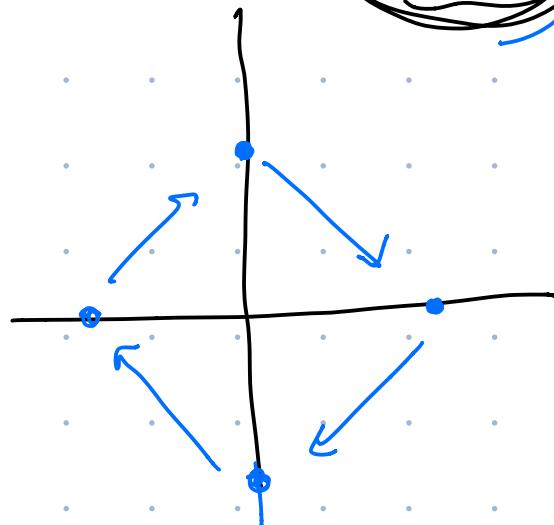
$$\boxed{\vec{x}_0 = R^{-1} Q^T \vec{b}.}$$

Check:  $A^T A \vec{x}_0 \stackrel{??}{=} A^T \vec{b}$

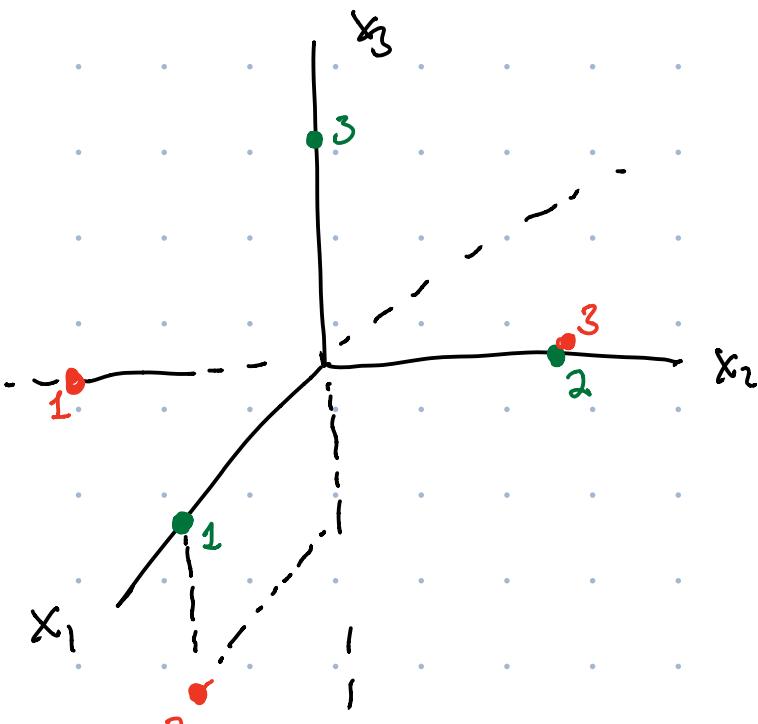
$\cancel{R^T Q^T Q R R^T Q^T \vec{b}}$        $\cancel{R^T Q^T \vec{b}}$

$\cancel{R^T Q^T Q} Q^T \vec{b}$

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



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



$\lambda_1, \dots, \lambda_k$  are distinct eigenvalues of  $A$ .

generalized eigenspace  $A: n \times n$   $\lambda$  eigenvalues of  $A$

$$V_\lambda^g := \{ \vec{v} \in \mathbb{C}^n \mid (A - \lambda I)^k \vec{v} = \vec{0} \text{ for some } k \geq 1 \}.$$

Thm:  $\mathbb{C}^n = V_{\lambda_1}^g \oplus \dots \oplus V_{\lambda_k}^g$

Sketch of proof:

K=1:  $A$  has only one eigenvalue  $\lambda_0$

$$\xrightarrow{\text{char. poly}} p(\lambda) = (\lambda_0 - \lambda)^n$$

Need to show:  $\mathbb{C}^n = V_{\lambda_0}^g$

i.e.  $\forall \vec{v} \in \mathbb{C}^n, \exists k \geq 1$  s.t.  $(A - \lambda_0 I)^k \vec{v} = \vec{0}$

Cayley-Hamilton thm: If  $p$  is the char. poly. of  $A$ ,

then  $p(A) = 0$

$$\underbrace{\lambda^n + b_{n-1}\lambda^{n-1} + \dots + b_0}_{\text{zero matrix}}$$

$$A^n + b_{n-1} A^{n-1} + \dots + b_0 I = 0$$

(In HW5 #6, you proved C-H for diag. A)

char. poly. of  $A \Leftrightarrow (\lambda_0 - \lambda)^n$

By C-H  $\Rightarrow$   $(\lambda_0 \mathbb{I} - A)^n = \mathbb{0}$

$$\Rightarrow (A - \lambda_0 \mathbb{I})^n \vec{v} = \vec{0}$$

$$\Rightarrow \mathbb{C}^n = V_{\lambda_0}^g \quad \square$$

Rank  $A$  is nilpotent ( $\underline{A^k = 0}$  for some  $k \geq 1$ )

In HW1,  $\Rightarrow 0$  is the only eigenvalue of  $A$ .

$\xrightarrow{k}$   
C-H

Suppose the statement is true for any linear transf. w/

$\leq k-1$  eigenvalues

Suppose  $A$  has  $k$  eigenvalues

WTS:  $A$  also has such descmp.

HWS extra credit:  $T: V \rightarrow V$  finite dim V.s.  $V$ .  $\ker^s(T)$

$$\mathbb{0} \subseteq \ker(T) \subseteq \ker(T^2) \subseteq \dots \subseteq \underbrace{\ker(T^k) = \ker(T^{k+1}) = \dots}_{\text{ker}^s(T)}$$

$$V \supseteq \text{Im}(T) \supseteq \text{Im}(T^2) \supseteq \dots \supseteq \underbrace{\text{Im}(T^k) = \text{Im}(T^{k+1}) = \dots}_{\text{Im}^s(T)}$$

- $V = \ker^s(T) \oplus \text{Im}^s(T)$
- $T$  preserves this decomp.:  $T(\ker^s(T)) \subseteq \ker^s(T)$   
 $T(\text{Im}^s(T)) \subseteq \text{Im}^s(T)$
- $T|_{\ker^s(T)}$  nilpotent,  $T|_{\text{Im}^s(T)}$  invertible

Apply \* to:  $T = T_{(A - \lambda_k \mathbb{I})}$

$$\mathbb{C}^n = \boxed{\ker^s(A - \lambda_k \mathbb{I})} \oplus \boxed{\text{Im}^s(A - \lambda_k \mathbb{I})}$$

$$V_{\lambda_k}^s = \{ \vec{v} \mid (A - \lambda_k \mathbb{I})^l \vec{v} = \vec{0} \text{ for some } l \geq 1 \}$$

Claim:  $T_{A - \lambda_k \mathbb{I}}|_{\text{Im}^s(A - \lambda_k \mathbb{I})}$  only has eigenvalues  $\lambda_1, \dots, \lambda_{k-1}$ .

PF  $T_{A - \lambda_k \mathbb{I}}|_{\text{Im}^s(A - \lambda_k \mathbb{I})}: \text{Im}^s(A - \lambda_k \mathbb{I}) \rightarrow \text{Im}^s(A - \lambda_k \mathbb{I})$  invertible.

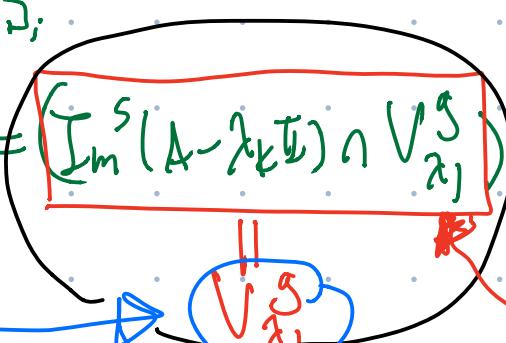
$$\forall \vec{v} \in \text{Im}^s(A - \lambda_k \mathbb{I}) \setminus \{ \vec{0} \},$$

$$(A - \lambda_k \mathbb{I})\vec{v} \neq \vec{0}$$

$\Rightarrow \lambda_k$  is not an eigenvalue of  $T_{A - \lambda_k \mathbb{I}}|_{\text{Im}^s(A - \lambda_k \mathbb{I})}$

By inductive hypothesis;

$$\text{Im}^s(A - \lambda_k \mathbb{I}) = \boxed{\text{Im}^s(A - \lambda_k \mathbb{I}) \cap V_{\lambda_1}^s} \oplus \dots \oplus \boxed{\text{Im}^s(A - \lambda_k \mathbb{I}) \cap V_{\lambda_k}^s}$$



$$V_{\lambda_k}^s$$

Obs:  $(A - \lambda_1 \mathbb{I})$  preserves  $\ker^S(A - \lambda_k \mathbb{I}) \oplus \text{Im}^S(A - \lambda_k \mathbb{I})$

$$\boxed{\bigcup_{\lambda_1}^S = \ker^S(A - \lambda_1 \mathbb{I}) = (\ker^S(A - \lambda_1 \mathbb{I}) \cap \ker^S(A - \lambda_k \mathbb{I})) \oplus (\ker^S(A - \lambda_1 \mathbb{I}) \cap \text{Im}^S(A - \lambda_k \mathbb{I}))}$$

Prop  $S: V \rightarrow V$ ,  $V = U \oplus W$ ,  $S(U) \subseteq U$ ,  $S(W) \subseteq W$ ,  
then  $\ker^S(S) = (\ker^S(S) \cap U) \oplus (\ker^S(S) \cap W)$

The only thing left:  $\boxed{\ker^S(A - \lambda_1 \mathbb{I}) \cap \ker^S(A - \lambda_k \mathbb{I}) = \{0\}}$

$$A \circledast \vec{v} \in W \subseteq \mathbb{C}^n$$

$\xrightarrow[S_0]{\quad}$

$\forall \vec{v}, (A - \lambda_1 \mathbb{I})^k \vec{v} = \vec{0}$  for some  $k \geq 1$ .

Claim: The only eigenvalue of  $A$  on  $W$  is  $\lambda_1$

$$Aw = \lambda' w \quad w \neq \vec{0}$$

$$(A - \lambda_1 \mathbb{I})^k w = \vec{0}$$

$$\Rightarrow (\lambda' - \lambda_1)^k w = \vec{0}$$

$$\Rightarrow \lambda' = \lambda_1$$

Cayley-Hamilton. (sketch of proof)

$$f: M_{n \times n}(\mathbb{R}) \longrightarrow M_{n \times n}(\mathbb{C})$$

$$A \longmapsto \underline{P_A(A)}$$

where  $P_A$  is the char. poly. of  $A$ :

- $A$  is diagonalizable,  $f(A) = P_A(A) = 0$
- $\{ \text{diagonalizable matrices} \} \subseteq M_{n \times n}(\mathbb{C})$  is "dense"
- $f$  is continuous.

$\Rightarrow f(A) = 0$  for any  $A$ .