

Today

→ ① Finish up LL : round-off
slides # 83/84

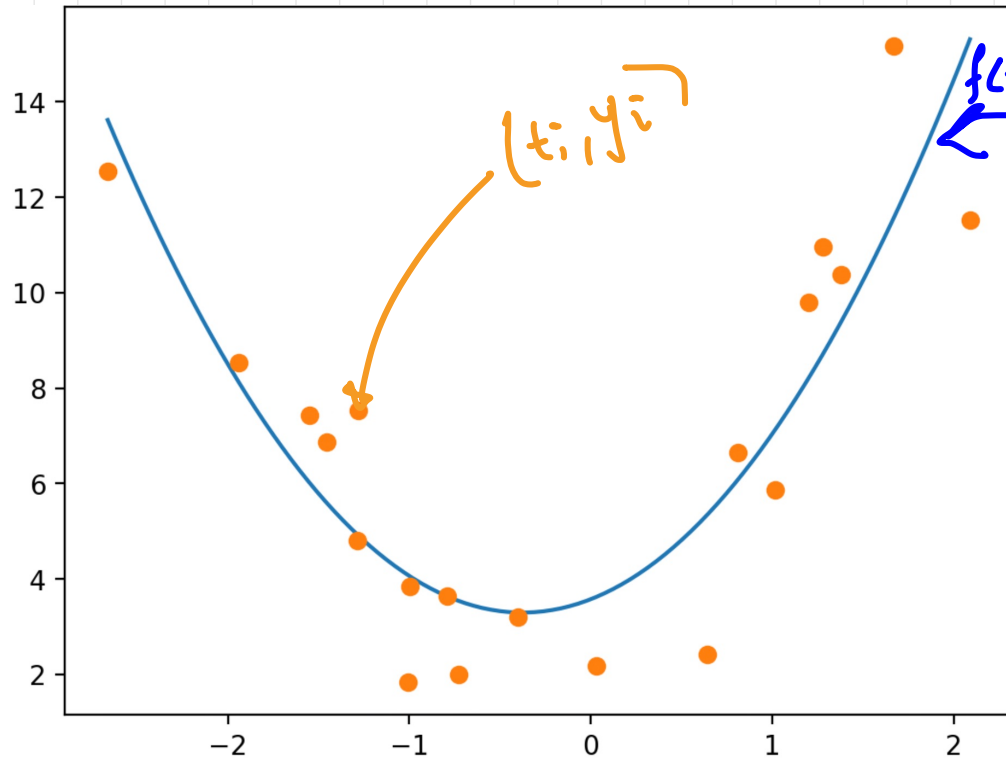
② Linear Least Squares

① Think about small ε

$$A = \begin{bmatrix} \varepsilon & 1 \\ 1 & 1 \end{bmatrix}$$

② linear least squares:

We have data (t_i, y_i)



We want a function

$$f(t) = x_1 + x_2 t + x_3 t^2$$

So that

$$y_i = f(t_i) = x_1 + x_2 t_i + x_3 t_i^2$$

$$y_i = f(t_i) = x_1 + x_2 t_i + x_3 t_i^2$$

↓

$$y_1 = x_1 + x_2 t_1 + x_3 t_1^2$$

$$y_2 = x_1 + x_2 t_2 + x_3 t_2^2$$

⋮

$$y_m = x_1 + x_2 t_m + x_3 t_m^2$$

→

$$\begin{bmatrix} 1 & t_1 & t_1^2 \\ 1 & t_2 & t_2^2 \\ \vdots & \vdots & \vdots \\ 1 & t_m & t_m^2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_m \end{bmatrix}$$

$m \times 3$ 3×1 $m \times 1$

OR so we have: Find x such that

$$A x = b$$

$m \times n$ $n \times 1$ $m \times 1$

Q: Does this have a solution?

$m = n$ ✓ yes

$m > n$ ✗ no, only if
 $b \in \text{colspan}(A)$

Different question: Find x such that

$$\|Ax - b\|_2 \rightarrow \text{minimized}$$

Q: solution exist?

✓ yes

Q: unique?

✗ yes, if A is full rank

How do we find \underline{x} ?

s.t.

$$\underline{x} \leftarrow \underset{\underline{x}}{\text{min}} \|Ax - b\|_2$$

Two main approaches:

- 1) direct minimization
- 2) transform A

$$\begin{aligned}
 \textcircled{1} \text{ let } f(x) &= \|Ax - b\|_2^2 \\
 &= \|b - Ax\|_2^2 \\
 &= (b - Ax)^T (b - Ax) \\
 &= \underbrace{b^T}_{(b^T - x^T A^T)} b - \underbrace{x^T A^T b}_{b^T A x} - \underbrace{b^T A x}_{b^T A x} + x^T A^T A x \\
 &= b^T b - 2 x^T A^T b + x^T A^T A x
 \end{aligned}$$

$\|v\|_2^2 = v^T v$

→ find $\nabla_x f$

→ set $\nabla_x f = 0$

→ find x

Two ways to find $\nabla f(x)$

1) write all components out with

$\sum_{i,j}$

2) $\lim_{\tau \rightarrow 0} \frac{f(x + \tau \cdot z) - f(x)}{\tau}$

try it!

$$f(\underline{y}) = \underline{b}^T \underline{b} - 2 \underline{x}^T A^T \underline{b} + \underline{x}^T A^T A \underline{x}$$

$$\rightarrow \nabla f(\underline{x}) = 0 - 2 A^T \underline{b} + 2 A^T A \underline{x} \equiv 0$$

set $\equiv 0$
 \rightarrow

$$2 A^T A \underline{x} = 2 A^T \underline{b}$$

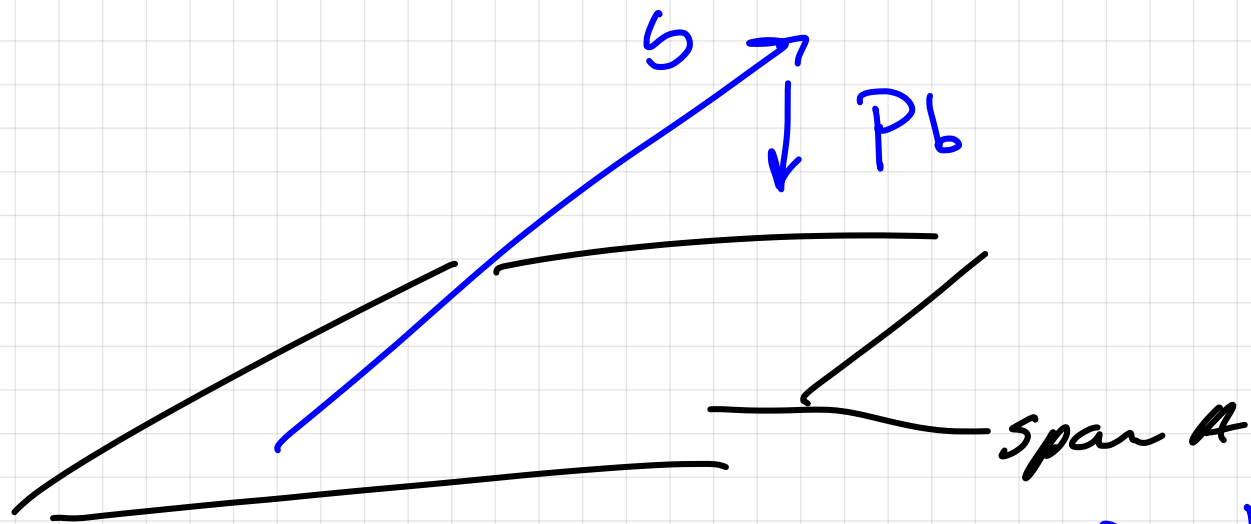
$$\rightarrow A^T A \underline{x} = A^T \underline{b}$$

The Normal Equations

$$\underbrace{n \times m \quad m \times n}_{n \times n} \cdot n \times 1 = n \times m \quad n \times 1$$

$$n \times n \cdot n \times 1 = n \times 1$$

$$\hat{A} \hat{\underline{x}} = \hat{\underline{b}}$$



We want a projection of b

$$P^2 = P$$

$$P = P^T$$

know $A^T A x = A^T b$
 $n \times n \quad m \times m$

$$\rightarrow x = \underbrace{(A^T A)^{-1}}_{A^+} A^T b$$

A^+ pseudoinverse
 (page 113)

$$\rightarrow Ax = \underbrace{A(A^T A)^{-1} A^T}_{P} b$$

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

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

$$= P$$

$$\begin{bmatrix} 1 \\ 3 \end{bmatrix} \underline{x} = \begin{bmatrix} -1 \\ 4 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 3 \end{bmatrix} \begin{bmatrix} 1 \\ 3 \end{bmatrix} \underline{x} = \begin{bmatrix} 1 & 3 \end{bmatrix} \begin{bmatrix} -1 \\ 4 \end{bmatrix}$$

$$10 \underline{x} = 11$$

$$\Rightarrow \underline{x} = \frac{11}{10}$$

$$A^T A x = A^T b$$

↓

x

→ min $\|Ax - b\|_2$

~~↓~~
 $Ax = b$

$$y = 1 \cdot x + b \cdot x + a^2 \cdot x^2$$

$$y = 1 \cdot x + b \cdot x + d \cdot x^2$$

$$a = \sqrt{d}$$

107