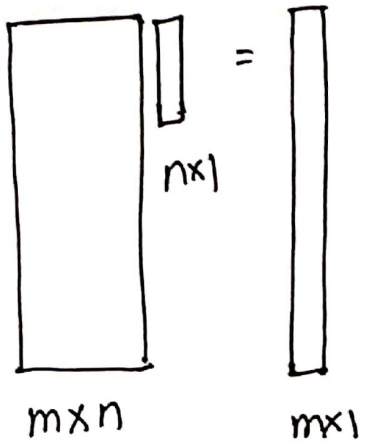


Module 3 - Lecture 4

Nov 21, 2019.

Least-Squares.

$$A \vec{x} = \vec{b}$$



$A$  is not a square mat.

$m > n$

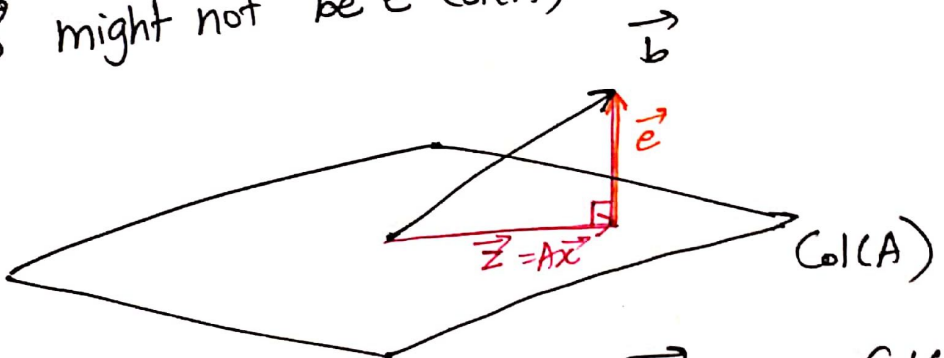
More equations than unknowns

"Overdetermined system"

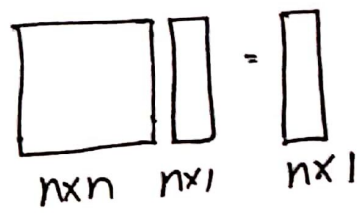
Issue:  $A$  is not square  
 $\rightarrow A$  has no inverse.

$A\vec{x} \in \text{Col}(A)$ .

$\vec{b}$  might not be  $\in \text{Col}(A)$



Square system.



$A^T: n \times m$

$A^T A: n \times n$

$\hat{\vec{x}}$ : best of estimate of  $\vec{x}$

The orthogonal projection of  $\vec{b}$  onto  $\text{Col}(A)$  minimizes

Solution: "Project"  $\vec{b}$  onto  $\text{col}(A)$ .

$$\hat{\vec{x}} = (A^T A)^{-1} A^T \cdot \vec{b}$$

$$A \hat{\vec{x}} = A (A^T A)^{-1} A^T \cdot \vec{b}$$

$$\|A\vec{x} - \vec{b}\|^2 = \|\vec{e}\|^2$$

Example:

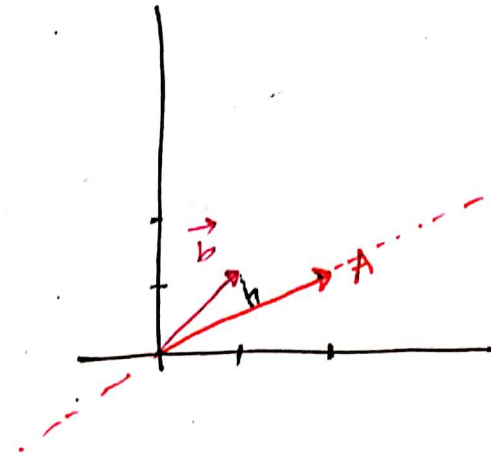
$$A = \begin{bmatrix} 2 \\ 1 \end{bmatrix}$$

2x1

$$\vec{b} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

$$A \cdot x = \vec{b}$$

$$\begin{bmatrix} 2 \\ 1 \end{bmatrix} [x] = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$



① Does GE work?

$$\left[ \begin{array}{c|c} 2 & 1 \\ 1 & 1 \end{array} \right] \rightarrow \left[ \begin{array}{c|c} 1 & 1/2 \\ 1 & 1 \end{array} \right] \rightarrow \left[ \begin{array}{c|c} 1 & 1/2 \\ 0 & 1/2 \end{array} \right]$$

No solution.

②  $\hat{x} = (A^T A)^{-1} A^T \vec{b}$

$$A^T = [2 \quad 1]$$

$$A^T A = [2 \quad 1] \begin{bmatrix} 2 \\ 1 \end{bmatrix} = 5$$

$$(A^T A)^{-1} = 1/5$$

$$\hat{x} = \frac{1}{5} [2 \quad 1] \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \frac{3}{5}$$

1801 Jan Piazzi

22 observations.

1801 June

### Kepler's Law

We know: Planet will have an elliptical orbit.  
(x,y) pairs.

$$ax^2 + by^2 + cxy + dx + ey = 1$$

Eq<sup>n</sup> for an ellipse.

Vector of unknowns:  $\begin{bmatrix} a \\ b \\ c \\ d \\ e \end{bmatrix}$

Known:  $(x_1, y_1) (x_2, y_2) \dots$

Unknowns:  $a, b, c, d, e.$

$$\begin{bmatrix} x_1^2 & y_1^2 & x_1 y_1 & x_1 & y_1 \\ x_2^2 & y_2^2 & x_2 y_2 & x_2 & y_2 \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ x_n^2 & y_n^2 & \dots & x_n & y_n \end{bmatrix}$$

A

$$\begin{bmatrix} a \\ b \\ c \\ d \\ e \end{bmatrix}$$

$$= \begin{bmatrix} 1 \\ \vdots \\ \vdots \\ \vdots \\ 1 \end{bmatrix}$$

$$ax_1^2 + by_1^2 + cx_1 y_1 + dx_1 + ey_1 = 1$$

# Linear Regression

$$y = m \cdot x + c$$

unknowns.

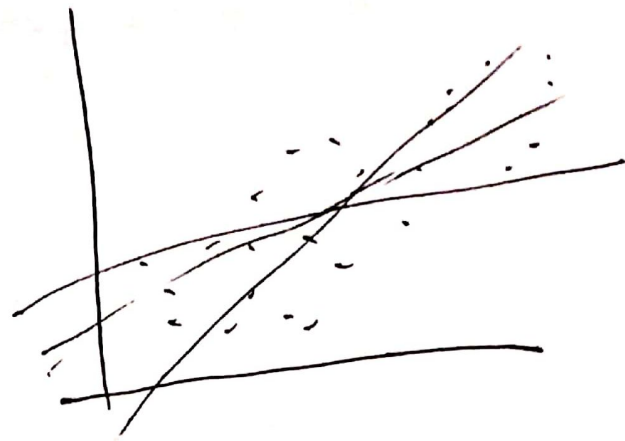
Line-of-best-fit.

$$y_1 = m \cdot x_1 + c$$

$$y_2 = m \cdot x_2 + c$$

⋮

$(x_1, y_1)$   
 $(x_2, y_2)$   
⋮  
 $(x_n, y_n)$  } Data



$$\underbrace{\begin{bmatrix} x_1 & 1 \\ x_2 & 1 \\ \vdots & \vdots \\ x_n & 1 \end{bmatrix}}_A \cdot \underbrace{\begin{bmatrix} m \\ c \end{bmatrix}}_w = \underbrace{\begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix}}_b$$

- $\hat{\vec{x}} = (A^T A)^{-1} A^T \vec{b}$

- LS can fail if  $A^T A$  is ~~not~~ non-invertible.

$(A^T A)^{-1}$  does not exist  $\iff (A^T A)$  has a non-trivial nullspace.

Thm:  $\text{Null}(A^T A) = \text{Null}(A)$

---

$\text{Null}(A)$  is non trivial.

---

$A \cdot \vec{z} = \vec{0} \rightarrow$  Col of  $A$  are linearly dependant.

Proof: (1) If  $\vec{u} \in N(A)$ , then  $\vec{u} \in N(A^T A)$ .

(2) If  $\vec{w} \in N(A^T A)$ , then  $\vec{w} \in N(A)$

(1)  $A \cdot \vec{u} = \vec{0}$  Want!  $A^T A \cdot \vec{u} = \vec{0}$

Mult by  $A^T \Rightarrow A^T(A \cdot \vec{u}) = A^T \cdot \vec{0} = \vec{0}$ .

$\Rightarrow \vec{u} \in N(A^T A)$ .

$$(2) \quad \vec{w} \in N(A^T A)$$

$$A^T A \cdot \vec{w} = 0$$

Want!  $A\vec{w} = 0$

Consider:  $A\vec{w} = \vec{q}$

$$\langle \vec{q}, \vec{q} \rangle = (A\vec{w})^T (A\vec{w})$$

~~$$\vec{w}^T A^T A \vec{w}$$~~

$$= \vec{w}^T \underbrace{A^T A}_{\text{red wavy}} \vec{w}$$

$$= \vec{w}^T \cdot 0$$

$$= 0 \cdot 0$$

$$\|\vec{q}\| = 0$$

Then  $\vec{q} = 0$ .

$$\vec{w} \in N(A)$$