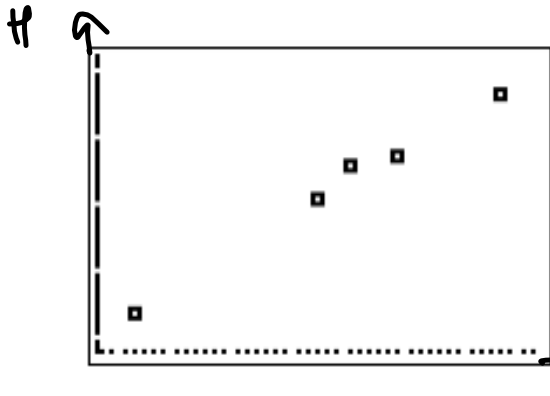


# Least Squares : Orthogonal Projections

Femur: 38 56 59 64 74

Humerus: 41 63 70 72 84



There is no single line  $y = mx + b$  that is satisfied by all the data points.

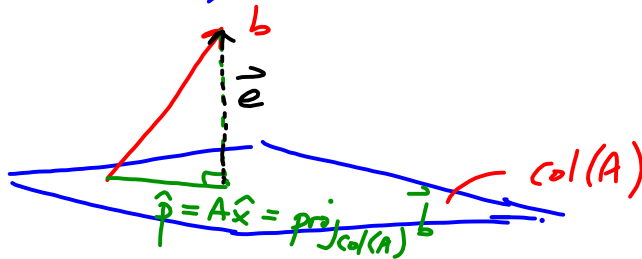
Make it fit: Assume  $y = mx + b$  is the line that models this relationship. Then it must be true that:

$$\begin{cases} 38m + b = 41 \\ 56m + b = 63 \\ 59m + b = 70 \\ 64m + b = 72 \\ 74m + b = 84 \end{cases} \rightarrow \begin{bmatrix} 38 & 1 \\ 56 & 1 \\ 59 & 1 \\ 64 & 1 \\ 74 & 1 \end{bmatrix} \begin{bmatrix} m \\ b \end{bmatrix} = \begin{bmatrix} 41 \\ 63 \\ 70 \\ 72 \\ 84 \end{bmatrix}$$

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

We know from the graph that this is an inconsistent system. Recall: If  $A\vec{x} = \vec{b}$  has no solution  $\vec{x}$ ,  $\vec{b}$  is not in  $\text{col}(A)$ .

$$A\vec{x} = \vec{b}, \text{ and } \vec{b} \notin \text{col}(A) \dots$$



If  $\vec{b}$  is not in  $\text{col}(A)$ , what's the closest we could get to  $\vec{b}$  while in  $\text{col}(A)$ ? The vector  $\vec{p} = \text{proj}_{\text{col}(A)} \vec{b}$ .

Also, since  $\vec{p} \in \text{col}(A)$ , we know

$\vec{p} = A\hat{x}$  for some heretofore undetermined vector  $\hat{x}$ .

$$\vec{b} = \vec{e} + \vec{p} \implies \vec{e} = \vec{b} - \vec{p} \\ = \vec{b} - A\hat{x}$$

$\vec{e} \perp \text{col}(A)$ , so  $\vec{e}$  is in  $\text{null}(A^T)$ .

$$\text{So } A^T \vec{e} = \vec{0}$$

$$A^T (\vec{b} - A\hat{x}) = \vec{0}$$

$$A^T \vec{b} - A^T A \hat{x} = \vec{0}$$

$$A^T A \hat{x} = A^T \vec{b}$$

"normal equations"

Generally speaking,  $A^T$  will be invertible only if  $A$  is square and has linearly independent columns. In a least-squares situation,  $A^T$  is usually not square, hence not invertible, i.e., we usually cannot split  $A^T A$  apart. However,  $A^T A$  is ALWAYS square and will always be invertible if  $A$  has linearly independent columns.

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

"least squares solution"

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

$$\begin{bmatrix} 38 & 1 \\ 56 & 1 \\ 59 & 1 \\ 64 & 1 \\ 74 & 1 \end{bmatrix} \begin{bmatrix} m \\ b \end{bmatrix} = \begin{bmatrix} 41 \\ 63 \\ 70 \\ 72 \\ 84 \end{bmatrix}$$

On Calc:

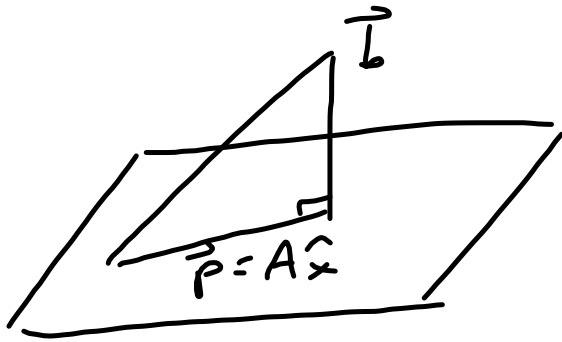
$$([A]^T [A])^{-1} [A]^T [B]$$

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

$$\hat{x} = \left( \begin{bmatrix} 38 & 56 & 59 & 64 & 74 \\ 1 & 1 & 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} 38 & 1 \\ 56 & 1 \\ 59 & 1 \\ 64 & 1 \\ 74 & 1 \end{bmatrix} \right)^{-1} \begin{bmatrix} 38 & 56 & 59 & 64 & 74 \\ 1 & 1 & 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} 41 \\ 63 \\ 70 \\ 72 \\ 84 \end{bmatrix}$$

$$= \begin{bmatrix} 17633 & 291 \\ 291 & 5 \end{bmatrix}^{-1} A^T \vec{b} = \begin{bmatrix} 1.196900\dots \\ -3.659586\dots \end{bmatrix}$$

$$\text{So } y = 1.1969x - 3.6596$$



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

$$\text{Proj}_{\text{col}(A)} \vec{b} = A \hat{x} = \underbrace{A(A^T A)^{-1} A^T}_{m \times m \text{ projection matrix}} \vec{b}.$$

$m \times m$  projection matrix for projecting any vector  $\vec{b}$  onto  $\text{col}(A)$ .

$(3, -1, 2)$  onto the subspace  $V$  with basis  $B = \{(1, 1, 2), (0, -1, 3)\}$

$$A = \begin{bmatrix} 1 & 0 \\ 1 & -1 \\ 2 & 3 \end{bmatrix} \quad \text{So } V = \text{col}(A).$$

Proj.  $(3, -1, 2)$  onto  $\text{col}(A)$ .

$$P = \underbrace{A(A^T A)^{-1} A^T}_{m \times m \text{ projection matrix}} \text{ for projection } \vec{b} \text{ onto cols of } A.$$

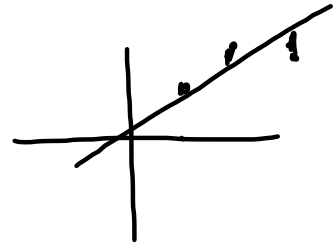
$$P = \frac{a a^T}{a^T a} \text{ for projection } \vec{b} \text{ onto } \vec{a}.$$

Calculus :

Suppose  $a_1 m = b_1$

$a_2 m = b_2$

$a_3 m = b_3$



What should  $m$  be?

Least-Squares:

Minimize sum of squared errors:

$$\sum (E^2) = \sum (a_i m - b_i)^2$$

$$\sum (E^2) = (a_1 m - b_1)^2 + (a_2 m - b_2)^2 + (a_3 m - b_3)^2$$

$$\frac{d(\sum(E^2))}{dm} = 2(a_1 m - b_1)a_1 + 2(a_2 m - b_2)a_2 + 2(a_3 m - b_3)a_3$$

$$= 2a_1^2 m - 2a_1 b_1 + 2a_2^2 m - 2a_2 b_2 + 2a_3^2 m - 2a_3 b_3$$

Min. could only occur when  $\frac{d(\sum(E^2))}{dm} = 0 \dots$

$$2a_1^2 m + 2a_2^2 m + 2a_3^2 m = 2a_1 b_1 + 2a_2 b_2 + 2a_3 b_3$$

$$m = \frac{a_1 b_1 + a_2 b_2 + a_3 b_3}{a_1^2 + a_2^2 + a_3^2}$$

$$m = \frac{a \cdot b}{a \cdot a} = \frac{a^T b}{a^T a}$$

## Least-Squares : QR

$$\text{If } A^T A \hat{x} = A^T \vec{b}$$

Since  $A = QR$ ,

$$A^T = (QR)^T = R^T Q^T$$

$$\underbrace{(R^T Q^T) QR}_{I} \hat{x} = R^T Q^T \vec{b}$$

$$R^T R \hat{x} = R^T Q^T \vec{b}$$

$R^T$  is invertible, so

$$R \hat{x} = Q^T \vec{b} \quad (R \text{ is upper } \Delta).$$

Solvable via  
back substitution