

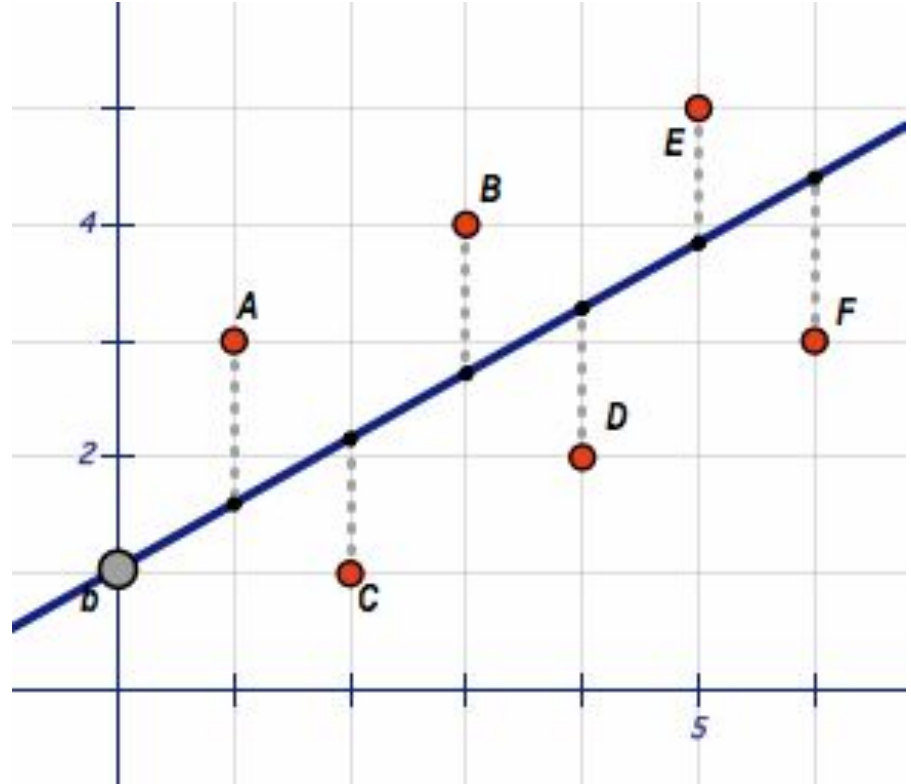
# Algorithms Club

6.3.24

# Least Squares

**Input** - Set of points in 2D space

**Output** - A line that minimizes the sum of squares of residuals



Cool Gif

<https://www.r-bloggers.com/2020/12/least-squares-as-springs-the-shiny-app/>

Good YT Video

<https://www.youtube.com/watch?v=EDPCsD6BzWE>

# Pseudocode Overview

1. Compute  $A^T A$  and  $A^T b$
2. Form augmented matrix
3. Row Reduce

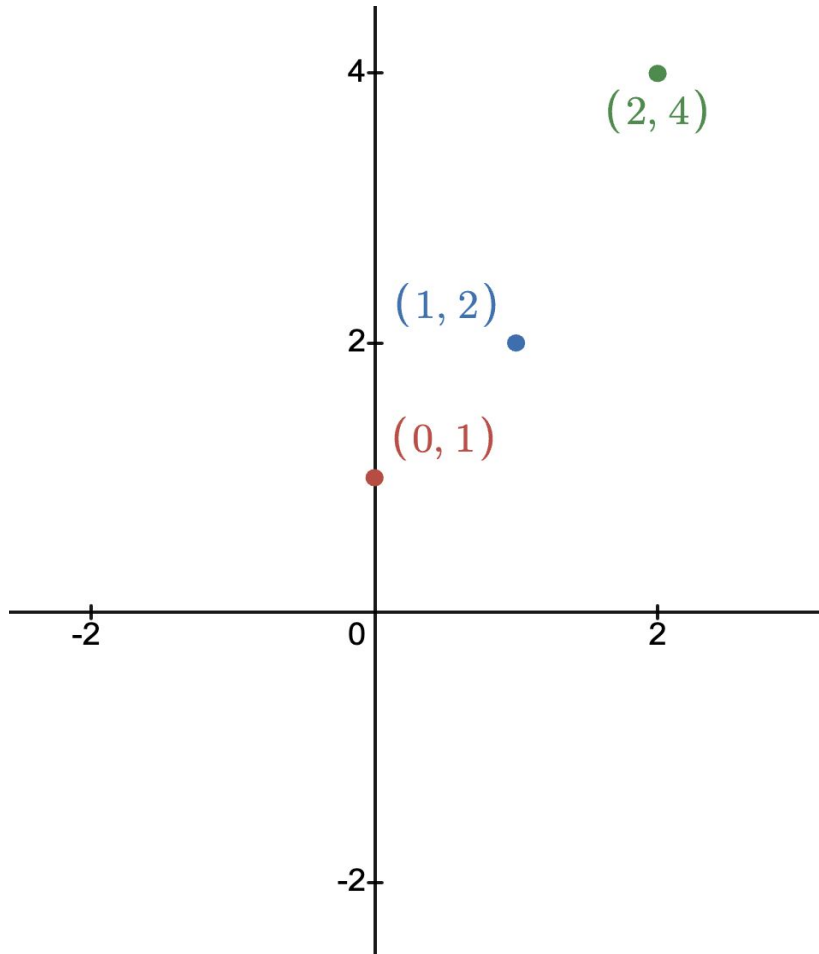
$$\mathbf{Ax} = \mathbf{b}$$



$$\mathbf{A}^T \mathbf{Ax} = \mathbf{A}^T \mathbf{b}$$

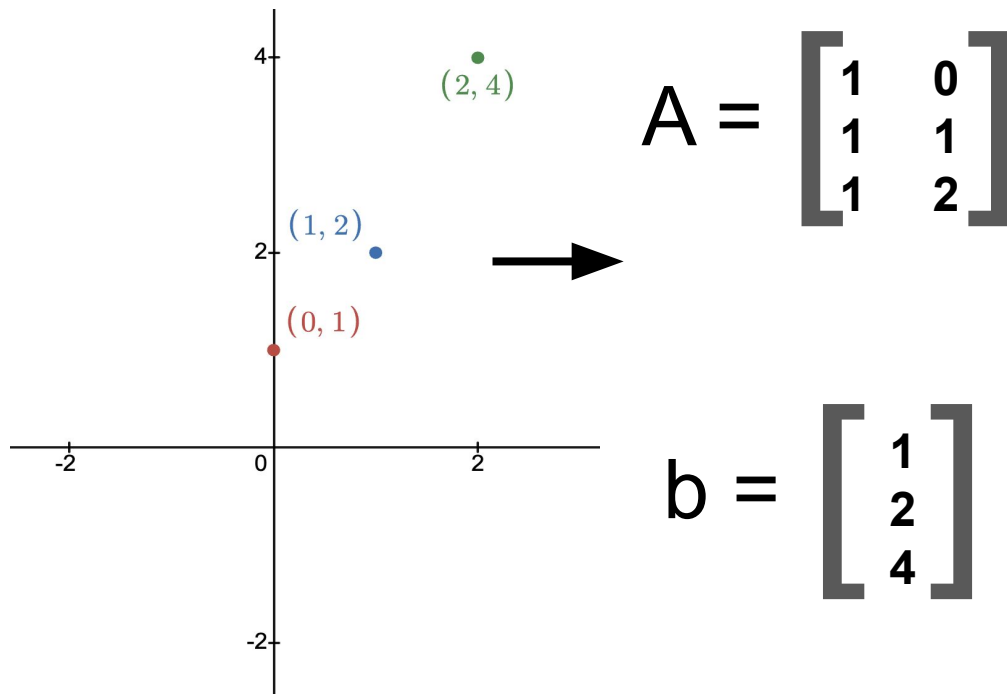
# In Depth Pseudocode

1. **Input-** Set of points in 2D space



# In Depth Pseudocode

1. **Input-** Set of points in 2D space
2. Make A matrix and b vector



# In Depth Pseudocode

3. Compute  $A^T A$  and  $A^T b$

$$\begin{bmatrix} 1 & 1 & 1 \\ 0 & 1 & 2 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 1 & 1 \\ 1 & 2 \end{bmatrix} = \begin{bmatrix} 3 & 3 \\ 3 & 5 \end{bmatrix} \quad A^T A$$

$$\begin{bmatrix} 1 & 1 & 1 \\ 0 & 1 & 2 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 4 \end{bmatrix} = \begin{bmatrix} 7 \\ 10 \end{bmatrix} \quad A^T b$$

# In Depth

## Pseudocode

3. Compute  $A^T A$  and  $A^T b$
4. Make augmented matrix

$$\begin{bmatrix} 3 & 3 & 7 \\ 3 & 5 & 10 \end{bmatrix}$$



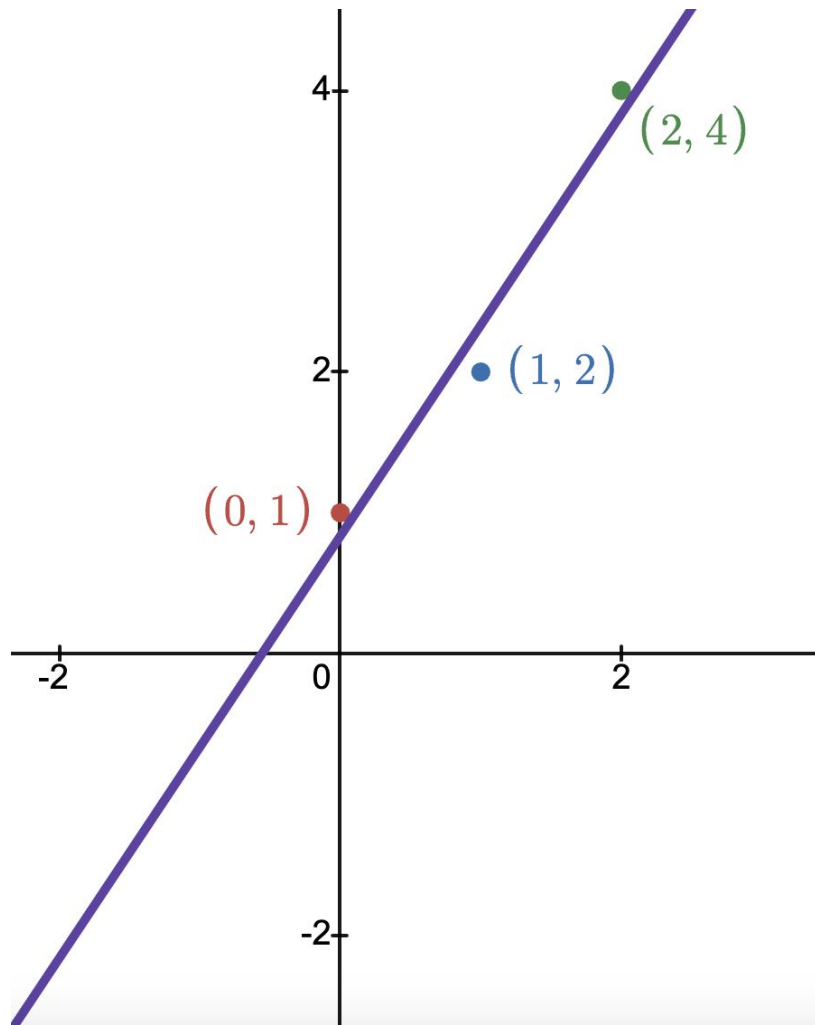
# In Depth Pseudocode

3. Compute  $A^T A$  and  $A^T b$
4. Make augmented matrix
5. Rref of matrix

$$\begin{bmatrix} 3 & 3 & 7 \\ 3 & 5 & 10 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 0 & 5/6 \\ 0 & 1 & 3/2 \end{bmatrix}$$

# In Depth Pseudocode

3. Compute  $A^T A$  and  $A^T b$
4. Make augmented matrix
5. Rref of matrix
6. Plot



$Ax = b$  - overdetermined

$$b = \underbrace{b}_{\text{projection Col}(A)} + \underbrace{b - b_{\text{projection Col}(A)}}_{\text{projection Col}(A)}$$

orthogonal to  $\text{Col}(A)$   
= orthogonal to  $\text{Row}(A^T)$   
=  $\text{Null}(A^T)$

Therefore

$$A^T A x = A^T b \quad \leftarrow \text{Normal equation}$$

$$Ax = \underbrace{A(A^T A)^{-1} A^T}_{\text{projection matrix}} b = b_{\text{projection Col}(A)}$$

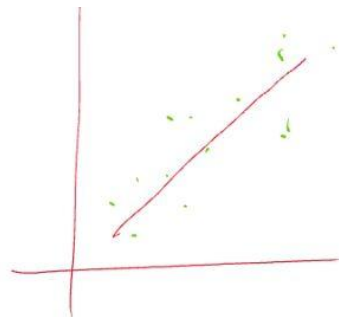
$$\begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix} = \begin{bmatrix} 1 & x_1 \\ 1 & x_2 \\ \vdots & \vdots \\ 1 & x_n \end{bmatrix} \begin{bmatrix} \beta_0 \\ \beta_1 \end{bmatrix}$$

"b" "A" "x"

This is overdetermined  
so cannot solve!!

Instead solve for

$$Ax = b_{\text{projection Col}(A)}$$



# row echelon form (ref) of a matrix

**Step 1.** Begin with an  $m \times n$  matrix  $A$ . If  $A = 0$ , go to Step 7.

**Step 2.** Determine the leftmost non-zero column.

**Step 3.** Use elementary row operations to put a 1 in the topmost position (we call this position pivot position) of this column.

**Step 4.** Use elementary row operations to put zeros (strictly) below the pivot position.

**Step 5.** If there are no more non-zero rows (strictly) below the pivot position, then go to Step 7.

**Step 6.** Apply Step 2-5 to the submatrix consisting of the rows that lie (strictly below) the pivot position.

**Step 7.** The resulting matrix is in row-echelon form

1)  $[A]_{m \times n}$

2)  $[A]$  : find leftmost column

3)  $[A]$  : Use elementary row operation

4)  $[A]$  : Make them zeros with row operation  
for example:  
 $R_2 \rightarrow R_2 - (\text{Multi factor}) \times R_1$

Move along the column and perform same operation

- if any pivot element is zero, switch row. immediately below, if not keep looking at the rows below when found switch rows

5) If no-more non-zeros (strictly) below the pivot position, you found a row echelon form

# Reduced row echelon form (rref) of a matrix

we can reduce a Row Echelon Form to the Reduced Row Echelon Form:

**Step 8.** Determine all the leading ones in the row-echelon form obtained in Step 7.

**Step 9.** Determine the right most column containing a leading one (we call this column pivot column).

**Step 10.** Use elementary row operations to erase all the non-zero entries above the leading one in the pivot column.

**Step 11.** If there are no more columns containing leading ones to the left of the pivot column, then go to Step 13.

**Step 12.** Apply Step 9-11 to the submatrix consisting of the columns that lie to the left of the pivot column.

**Step 13.** The resulting matrix is in reduced row-echelon form.

Solving Normal equation ??

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