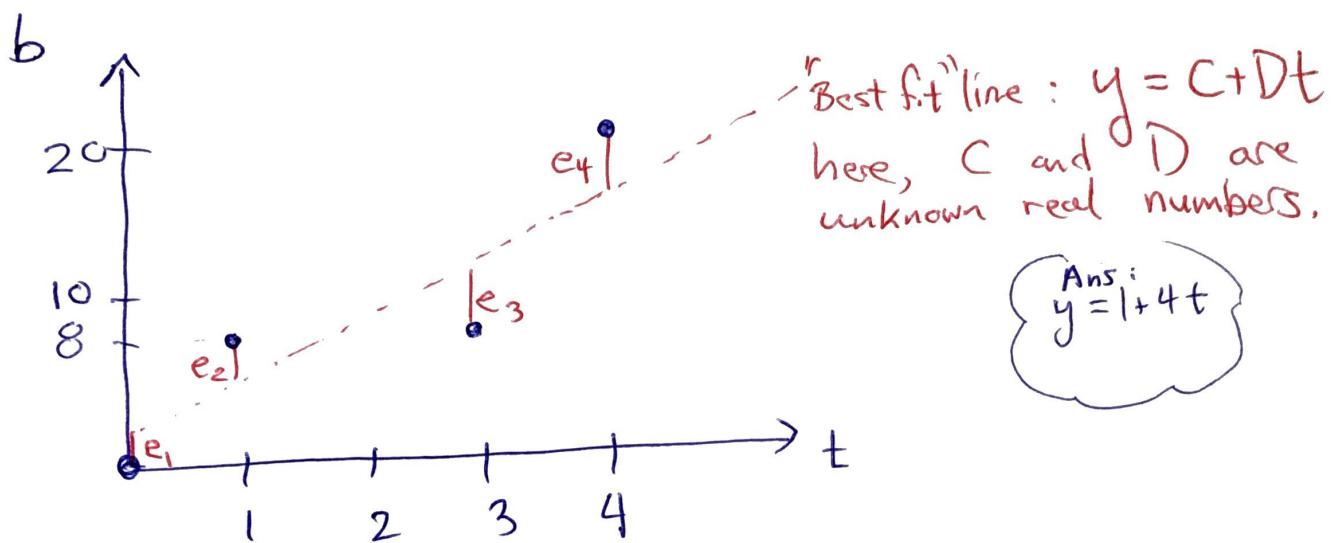


Linear Regression

Data: $(\text{time}, \text{obsesvation})$

$$t \quad b$$
$$(0,0), (1,8), (3,8), (4,20)$$



We want to fit a line. How best to do it? The line equation is $y = C + Dt$.

Ideally, our best-fit line passes through all the points:

$$\begin{cases} C + 0 \cdot D = 0 \\ C + 1 \cdot D = 8 \\ C + 3D = 8 \\ C + 4D = 20 \end{cases}$$

This is a system of 4 linear eqns
in 2 unknowns^(C and D). We can write it
in matrix form as

$$\begin{bmatrix} 1 & 0 \\ 1 & 1 \\ 1 & 3 \\ 1 & 4 \end{bmatrix} \begin{bmatrix} C \\ D \end{bmatrix} = \begin{bmatrix} 0 \\ 8 \\ 8 \\ 20 \end{bmatrix}$$

$\underbrace{\hspace{1cm}}_{\vec{x}}$

$$A = \begin{bmatrix} 1 & t_1 \\ 1 & t_2 \\ 1 & t_3 \\ 1 & t_4 \end{bmatrix} \quad \vec{b} = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \\ b_4 \end{bmatrix}$$

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

Bad news: this system is "overdetermined".
There is no choice of C and D
which yields equality $A\vec{x} = \vec{b}$. After all,
our points do not lie on a line.

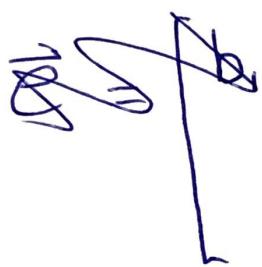
Instead, what we aim for is

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

or equivalently that

$$\vec{b} - A\vec{x} \approx 0$$

This is a vector. We call it \vec{e} .
~~called the vector of residuals.~~



$$\vec{e} = \begin{bmatrix} e_1 \\ e_2 \\ e_3 \\ e_4 \end{bmatrix}$$

~~$b_1 - (C + Dt_1)$~~
 ~~$b_2 - (C + Dt_2)$~~
 ~~$b_3 - (C + Dt_3)$~~
 ~~$b_4 - (C + Dt_4)$~~

where $e_i = b_i - (C + Dt_i)$

In our case,

$$\vec{e} = \begin{bmatrix} 0 - C \\ 8 - (C + D) \\ 8 - (C + 3D) \\ 20 - (C + 4D) \end{bmatrix}$$

This is ~~the~~ the vector of vertical distances, (known as residuals).

We want ~~to minimize~~ this vector to be close to 0.

One approach ~~is~~ is to find values of C and D which minimize the length of the residual vector \vec{e} .

The length of a vector v is $\|v\| = \sqrt{v_1^2 + v_2^2 + v_3^2}$

Minimizing a length is ~~is~~ equivalent to minimizing its square, which allows us to ~~fend off~~ fend off the pesky square root. So our new goal is:

~~find~~ find C, D which minimize ~~$\|e\|^2$~~

$$\|e\|^2 = e_1^2 + e_2^2 + e_3^2 + e_4^2.$$

In our problem,

$$\|\vec{e}\|^2 = \cancel{\|(c-10)^2\|} c^2 + (8-c-D)^2 + (8-c-3D)^2 \\ + (20-c-4D)^2$$

use calculus: (call RHS $f(c, D)$)

~~Step 1 Set derivative~~
Step 1 Set $\nabla f = \vec{0}$ (here $\nabla = \begin{bmatrix} \frac{\partial}{\partial c} \\ \frac{\partial}{\partial D} \end{bmatrix}$)

$$f_c = 0 = 2c + 2(8-c-D)(-1) + 2(8-c-3D)(-1)$$

$$+ 2(20-c-4D)(-1)$$

$$= 2c - 16 + 2c + 2D - 16 + 2c + 6D$$

~~=~~
$$-16 + 2c + 8D$$

$$= -16 + 8c + 16D$$

$$f_D = \frac{\partial f}{\partial D} = -2(8-c-D) - 6(8-c-3D) - 8(20-c-4D)$$

$$= -16 + 2c + 2D - 48 + 6c + 18D - 160 + 8c + 32D$$

$$= -224 + 16c + 52D$$

We have a system of equations

$$\begin{cases} 8C + 16D = 72 \\ 16C + 52D = 224 \end{cases}$$

Dividing both equations by 2 ^(to simplify) gives

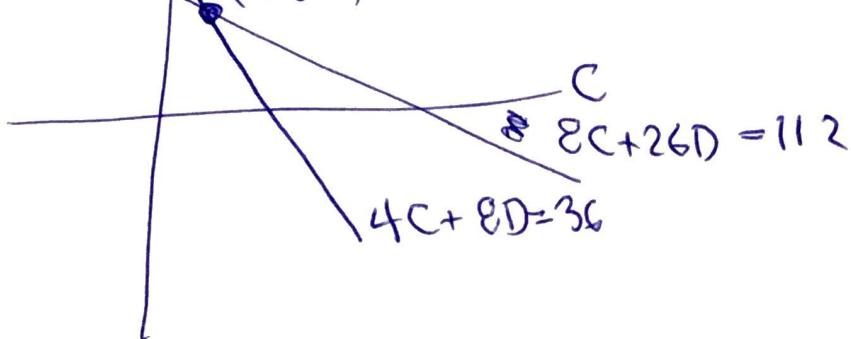
$$\begin{cases} 4C + 8D = 36 \\ 8C + 26D = 112 \end{cases}$$

Normal Eqs

Solving this System gives $(C,D) = (1,4)$

Q: How do we know this critical point

is a minimum?



Recall $Ax \approx b$, with $A = \begin{bmatrix} 1 & 0 \\ 1 & 1 \\ 1 & 3 \\ 1 & 4 \end{bmatrix}$, $\vec{b} = \begin{bmatrix} 0 \\ 8 \\ 8 \\ 20 \end{bmatrix}$,

~~unknown $x = [C]$~~
and unknown $\vec{x} = [C]$.

Transpose: If $A\vec{x} \approx \vec{b}$ then $A^T A \vec{x} \approx A^T \vec{b}$

$$A^T = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 0 & 1 & 3 & 4 \end{bmatrix} \quad (2 \times 4) \quad (4 \times 2)$$

$$A^T A = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 0 & 1 & 3 & 4 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 1 & 1 \\ 1 & 3 \\ 1 & 4 \end{bmatrix} = \begin{bmatrix} 4 & 8 \\ 8 & 26 \end{bmatrix} \quad (2 \times 2)$$

$$A^T \vec{b} = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 0 & 1 & 3 & 4 \end{bmatrix} \begin{bmatrix} 0 \\ 8 \\ 8 \\ 20 \end{bmatrix} = \begin{bmatrix} 36 \\ 112 \end{bmatrix} \quad (2 \times 1)$$

So $A^T A \vec{x} = A^T b$ is the equation

$$\begin{bmatrix} 4 & 8 \\ 8 & 26 \end{bmatrix} \begin{bmatrix} C \\ D \end{bmatrix} = \begin{bmatrix} 36 \\ 112 \end{bmatrix} \quad \text{or} \quad \begin{cases} 4C + 8D = 36 \\ 8C + 26D = 112 \end{cases}$$

The normal equations!

If A has L.I. columns

Theorem: The vector \vec{x} that minimizes $\|A\vec{x} - \vec{b}\|^2$ is the solution to the normal eqns

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

The vector $\vec{x} = (A^T A)^{-1} A^T \vec{b}$ is the least squares solution to $A\vec{x} = \vec{b}$.

Remember, $\vec{x} = \begin{bmatrix} c \\ d \end{bmatrix}$ so finding \vec{x} gives us the best-fit line.

Easy to summarize:

If $A\vec{x} = \vec{b}$ has no solution, multiply by A^T and solve
 $A^T A \vec{x} = A^T \vec{b}$