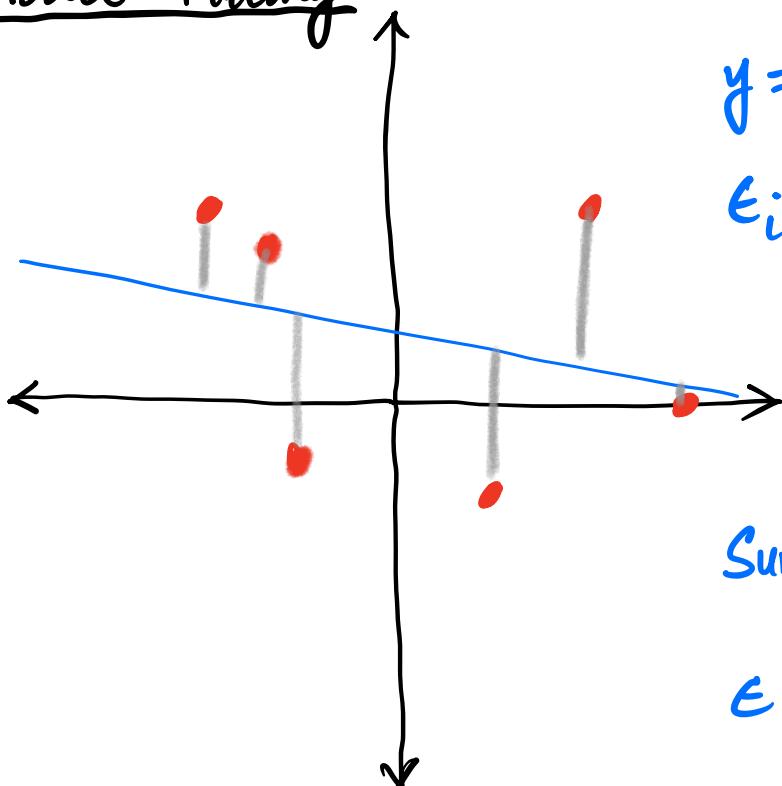


Model Fitting



$$y = mx + c$$

$$\epsilon_i = y_i - mx_i - c$$

Sum of squared errors

$$E = \sum_{i=1}^n (y_i - mx_i - c)^2$$

? ?

In order to find the best-fitting line we find m and c by solving the following optimization problem:

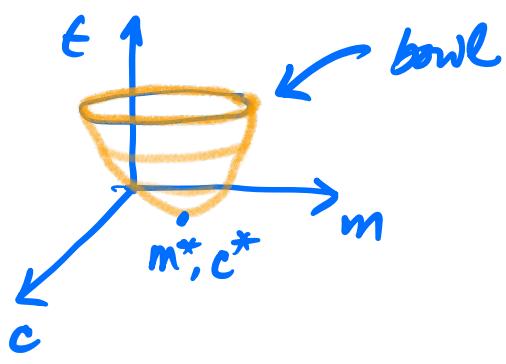
$$m^*, c^* = \underset{m, c}{\operatorname{argmin}} \sum_{i=1}^n (y_i - mx_i - c)^2$$

\because the error ϵ is quadratic

We can solve the optimization

by setting $\frac{\partial E}{\partial c} = 0$ and

$$\frac{\partial E}{\partial m} = 0.$$



$$\epsilon = \sum_{i=1}^n (\gamma_i - mx_i - c)^2$$

$$= \| \begin{bmatrix} \epsilon_1 \\ \epsilon_2 \\ \vdots \\ \epsilon_n \end{bmatrix} \|^2$$

$$= \| \vec{y} - \vec{A}\vec{p} \|^2$$

$$= \vec{y}^T \vec{y} - 2(\vec{A}\vec{p})^T \vec{y} + (\vec{A}\vec{p})^T \vec{A}\vec{p}$$

Compute $\frac{\partial \epsilon}{\partial \vec{p}} = ?$

$$\frac{\partial \epsilon}{\partial \vec{p}} = -2\vec{A}^T \vec{y} + 2\vec{A}^T \vec{A}\vec{p}$$

↑
Set $\frac{\partial \epsilon}{\partial \vec{p}} = 0$ and solve for \vec{p} .

$$(a+b)^2 = a^2 + 2ab + b^2$$

$$-2\vec{A}^T \vec{y} + 2\vec{A}^T \vec{A} \vec{p} = 0$$

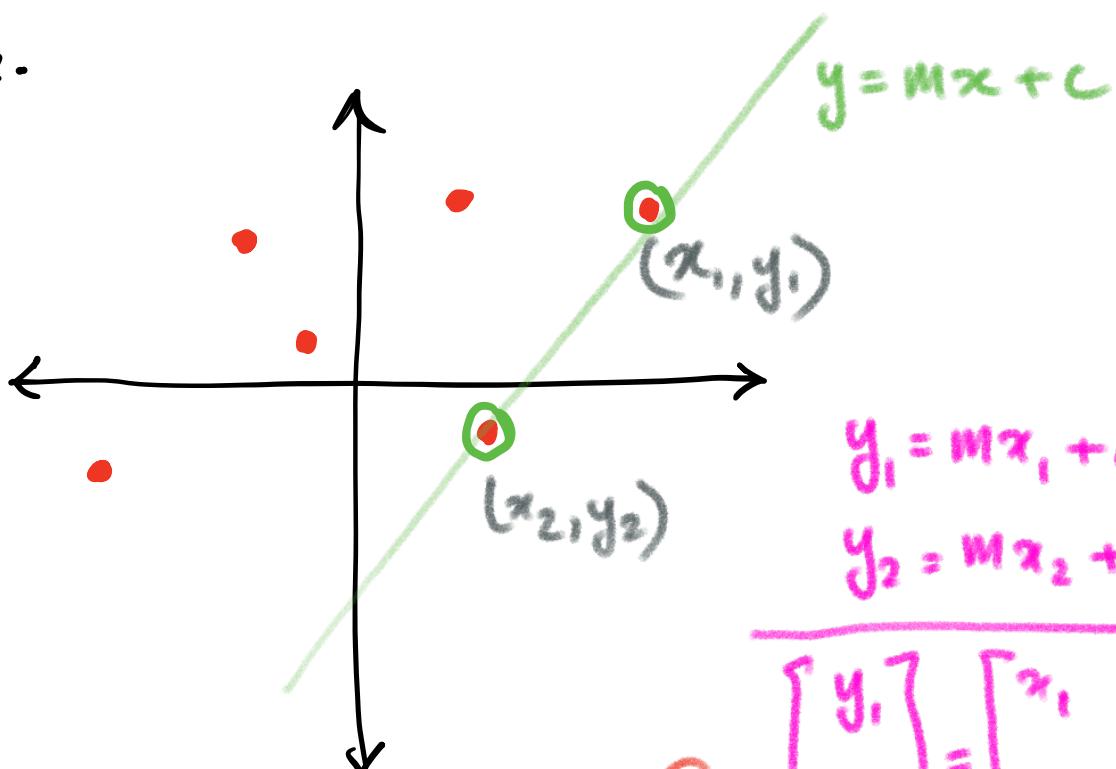
$$\Rightarrow \vec{A}^T \vec{A} \vec{p} = \vec{A}^T \vec{y}$$

$$\Rightarrow \vec{p} = (\vec{A}^T \vec{A})^{-1} \vec{A}^T \vec{y}$$

\downarrow \downarrow \downarrow
 $[m]$ $n \times n$ $n \times 1$
 \downarrow
 $n \times 2$
 $\underbrace{\quad}_{2 \times n}$
 $\underbrace{\quad}_{2 \times 2}$
 $\underbrace{\quad}_{2 \times 1}$

least squares fit
of a linear model
in our case
the lowly 2D
line.

Detour.



$$\textcircled{1} \quad \begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = \begin{bmatrix} x_1 & 1 \\ x_2 & 1 \end{bmatrix} \begin{bmatrix} m \\ c \end{bmatrix}$$

+ CHANGE OF NOTATION.

$$\textcircled{1} \quad A \vec{x} = \vec{b}$$

BACK TO OUR NOTATION.

$$A \vec{p} = \vec{y}$$

How to solve this equation?

$$\vec{p} = \vec{A}^{-1} \vec{y}$$

We want to fit the line to multiple points.

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

is not a square matrix.

$$\Rightarrow A \vec{p} = \vec{y}$$

$$\Rightarrow A^T A \vec{p} = A^T \vec{y}$$

linear regression.

$$\Rightarrow \vec{p} = \underline{(A^T A)^{-1}} A^T \vec{y}.$$

pseudo-inverse.

Detour.

Line: $y = mx + c$

Quadratic model: $y = ax^2 + bx + c$

linear in

a, b, c

$$y = f(x; a, b, c)$$

①

$$\epsilon_i = y_i - ax_i^2 - bx_i - c \quad (x_1, y_1) \dots (x_n, y_n)$$

x

② or ③

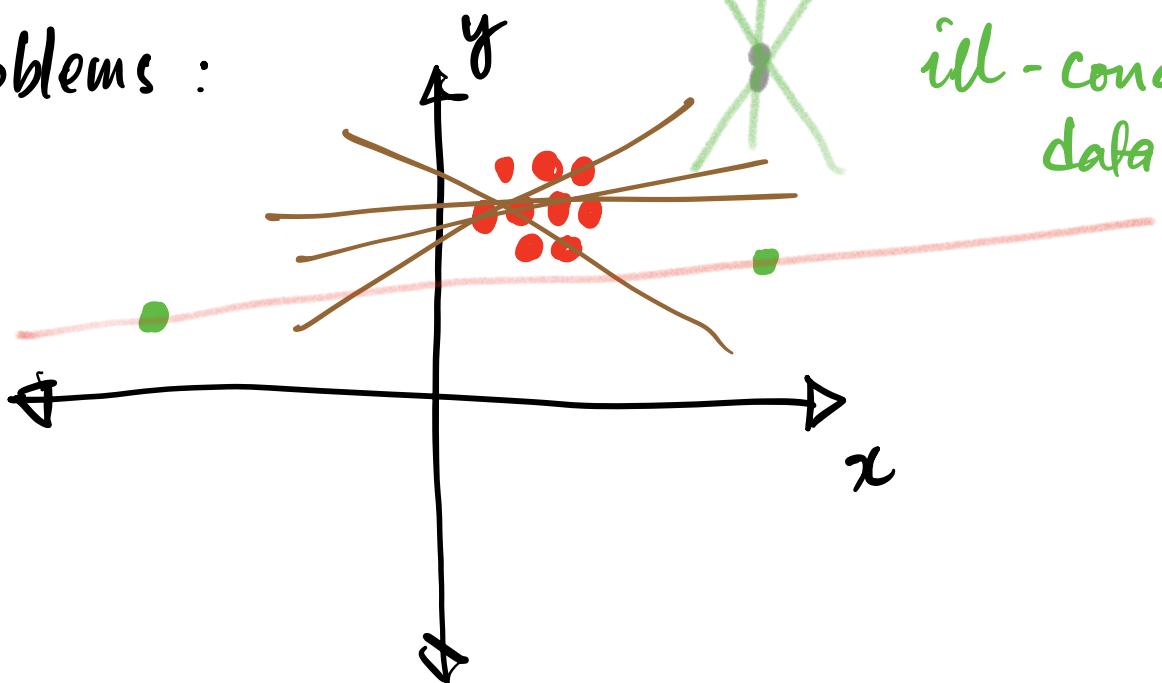
$$\vec{y} = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix}$$

$$A = \begin{bmatrix} x_1^2 & x_1 & 1 \\ x_2^2 & x_2 & 1 \\ \vdots & \vdots & \vdots \\ x_n^2 & x_n & 1 \end{bmatrix}$$

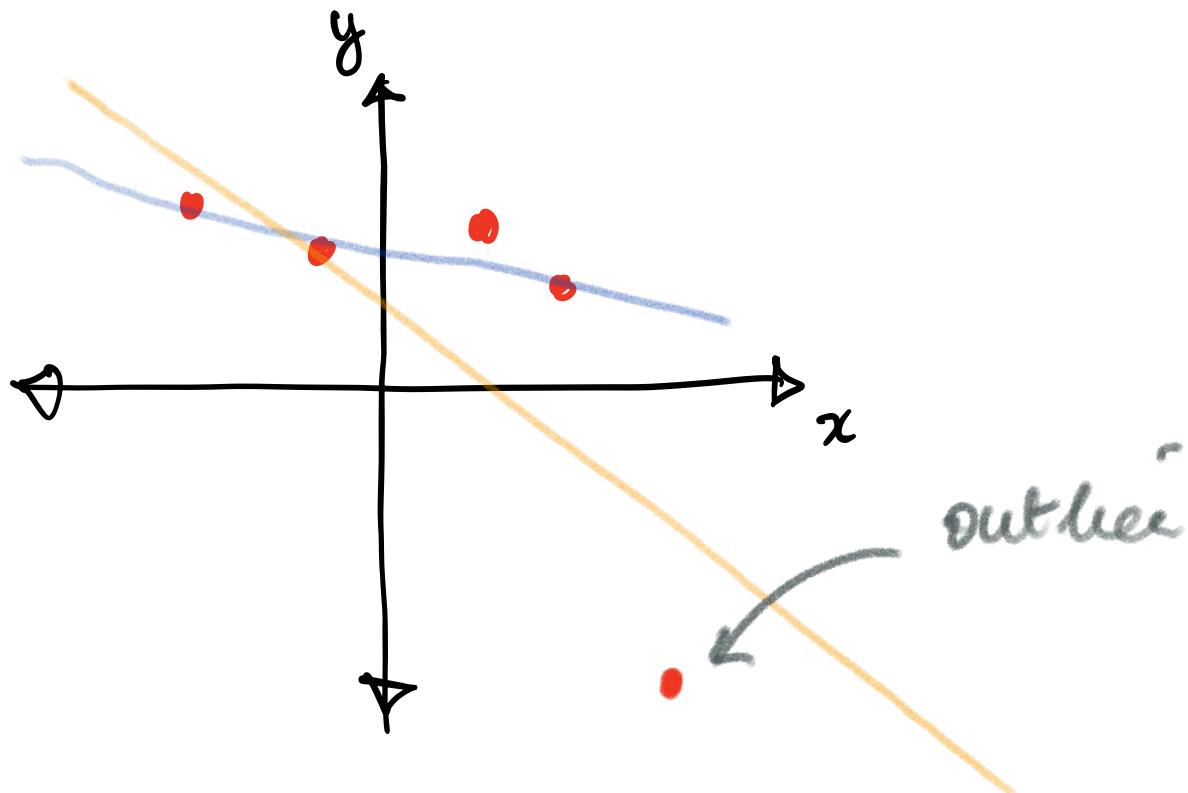
$$\vec{p} = \begin{bmatrix} a \\ b \\ c \end{bmatrix}$$

$$\Rightarrow \vec{p} = (A^T A)^{-1} A^T \vec{y}$$

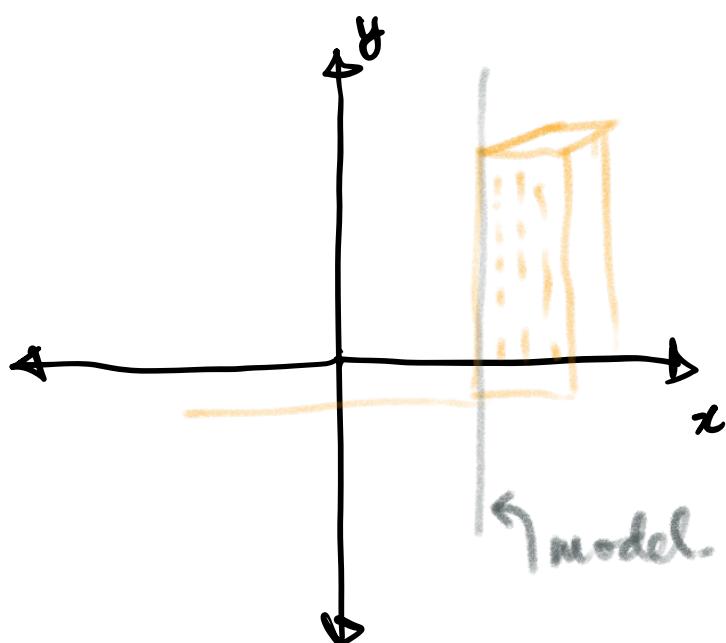
Problems:



ill-conditioned
data.



CHALLENGE



Dops. Can't fit.
line is vertical.
Slope $m \rightarrow \infty$.

$y = mx + c$
 \downarrow
 Fitting using
edge pixel locations.
 $(x_1, y_1) \dots (x_n, y_n)$

Use Implicit Equation of a line:

$$ax + by + c = 0$$

