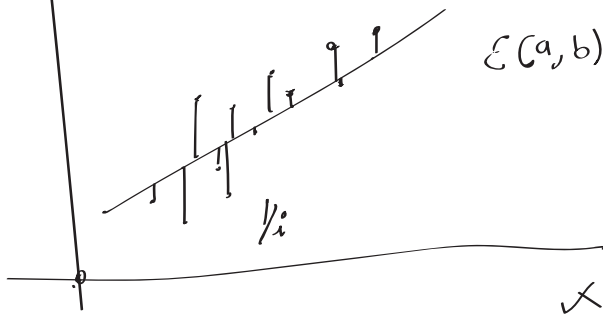


Regressión

$f(x)$



$f(x, \theta) \theta(a, b)$

diferencia o el error es menor
menor

$$\mathcal{E}(a, b) = \sum_{i=1}^N |ax_i + b - y_i|$$

$l_2 \rightarrow$ distancia

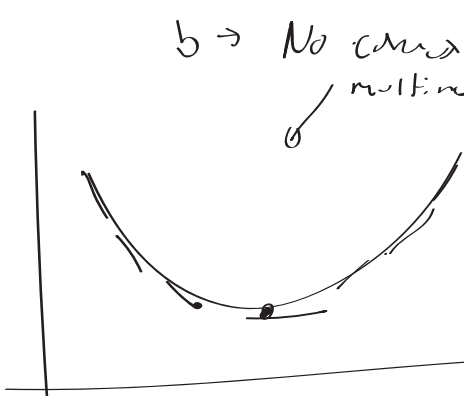
$$\mathcal{E}(a, b) = \sum_{i=1}^N (ax_i + b - y_i)^2$$

Convexa

Concava

$\theta = \arg \min_{\theta} MSE(\theta)$

$\mathcal{E}(a)$

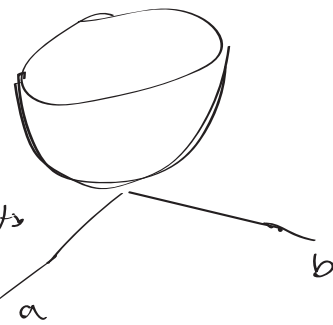


$b \rightarrow$ No cambia
multimodal

~~MSE~~

Optimizer

Encontrar la mejor
solución entre
todas las tests



¿Cuál a y b minimizan el error?

$$\frac{\partial \mathcal{E}(a, b)}{\partial a} = 0$$

$$\frac{\partial \mathcal{E}(\theta)}{\partial \theta} = 0$$

$b = 0$ restricción

$$\mathcal{E}(a, b) = \sum_{i=1}^N (ax_i + b - y_i)^2$$

$$\mathcal{E}(a) = \sum_{i=1}^N (ax_i - y_i)^2$$

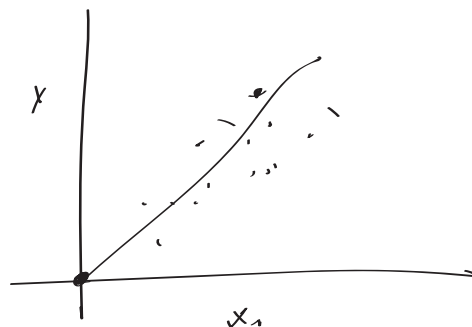
$\arg \min_a \mathcal{E}(a)$

$$\frac{\partial \mathcal{E}(a)}{\partial a} = 2 \sum_{i=1}^N (ax_i - y_i) \cdot x_i$$

$$a \sum x_i^2 - \sum y_i x_i = 0$$

$$a \sum x_i^2 = \sum y_i x_i$$

$$a = \frac{\sum y_i x_i}{\sum x_i^2}$$



$$f(a,b) = \sum_{i=1}^N (ax_i + b - y_i)^2$$

$$2 \sum (ax_i + b - y_i) \cdot 1$$

$$\sum ax_i + \sum b - \sum y_i = 0$$

$$\frac{\partial f(a,b)}{\partial a} = 0 \quad \frac{\partial f(a,b)}{\partial b} = 0$$

$$a \sum x_i^2 + b \sum x_i = \sum y_i x_i \quad \dots 1$$

$$\sum ax_i + \sum_{i=1}^N b = \sum y_i \quad \dots 2$$

$$\sum_{i=1}^N b = Nb$$

$$\begin{bmatrix} a \sum x_i^2 + Nb \sum x_i & : & \sum y_i x_i \\ a \sum x_i + Nb & : & \sum y_i \end{bmatrix}$$

$$\text{row } 1 \times N - 2 \sum x$$

$$a \sum x + Nb = \sum y$$

$$b = \frac{\sum y - a \sum x}{N}$$

$$aN \sum x^2 - a(\sum x)^2 = N \sum yx - \sum y \sum x$$

$$a = \frac{N \sum yx - \sum y \sum x}{N \sum x^2 - (\sum x)^2}$$

$$b = \frac{\sum y - \left(\frac{N \sum yx - \sum y \sum x}{N \sum x^2 - (\sum x)^2} \right) \sum x}{N}$$

$$b = \frac{N \sum y \sum x^2 - \sum y (\sum x)^2 - N \sum yx \sum x + \sum y \sum x^2}{N (N \sum x^2 - (\sum x)^2)}$$

$$b = \frac{\sum y \sum x^2 - \sum yx \sum x}{N \sum x^2 - (\sum x)^2}$$