

Enunciados c y d - punto 7 Mínimos cuadrados

c

$$J(\vec{\theta}) = \sum_{i=1}^N \left(\frac{y_i - M(x_i, \vec{\theta})}{\sigma_i} \right)^2$$

Derivada parcial:

$$\frac{\partial J(\vec{\theta})}{\partial \theta_i} = \frac{\partial \left(\sum_{i=1}^N \left(\frac{y_i - M(x_i, \vec{\theta})}{\sigma_i} \right)^2 \right)}{\partial \theta_i}$$

$$= \sum_{i=1}^N 2 \left(\frac{y_i - M(x_i, \vec{\theta})}{\sigma_i} \right) \left(\frac{-\partial M(x_i, \vec{\theta})}{\partial \theta_i} \right)$$

$$= -2 \sum_{i=1}^N \left(\frac{y_i - M(x_i, \vec{\theta})}{\sigma_i} \right) \left(\frac{\partial M(x_i, \vec{\theta})}{\partial \theta_i} \right)$$

d

$$\vec{\theta}_{j+1} = \vec{\theta}_j - \gamma \left(-2 \sum_{i=1}^N (y_i - M(x_i, \vec{\theta}_j)) \nabla_{\theta} M(x_i, \vec{\theta}_j) \right)$$

Descenso del gradiente \rightarrow dirección opuesta al gradiente

$$x_n = x_{n-1} - \underbrace{\beta}_{\text{distancia}} \nabla_x f(x_n)$$

$\vec{\theta}_j \rightarrow x_{n-1}$ (vector inicial) $\frac{1}{\gamma} \rightarrow \beta$ (distancia)

$$\nabla_{\theta} x^2(\vec{\theta}) =$$

$$\left[-2 \sum_{i=1}^N (y_i - M(x_i, \vec{\theta}_0)) \left(\frac{\partial M(x_i, \vec{\theta}')}{\partial \theta_0} \right), -2 \sum_{i=1}^N (y_i - M(x_i, \vec{\theta}_1)) \right]$$

$$\left(\frac{\partial M(x_i, \vec{\theta}')}{\partial \theta_1} \right), -2 \sum_{i=1}^N (y_i - M(x_i, \vec{\theta}_2)) \left(\frac{\partial M(x_i, \vec{\theta}')}{\partial \theta_2} \right) \right]$$

$$\nabla_{\theta} M(x_i, \vec{\theta}_j) = \left[\frac{\partial M(x_i, \vec{\theta}')}{\partial \theta_0}, \frac{\partial M(x_i, \vec{\theta}')}{\partial \theta_1}, \frac{\partial M(x_i, \vec{\theta}')}{\partial \theta_2} \right]$$