

6) Linear Regression

$$\mathcal{L} = (w @ x - y)^2$$

$$= (w_1 x_1 + w_2 x_2 + b - y)^2$$

$$\frac{\partial \mathcal{L}}{\partial w_2} = 2(w_1 x_1 + w_2 x_2 + b - y) \frac{\partial (w_1 x_1 + w_2 x_2 + b - y)}{\partial w_2}$$

$$= 2(w_1 x_1 + w_2 x_2 + b - y) x_2$$

$$= 2 \epsilon x$$

$$\text{where } \epsilon = w @ x - y$$

7) Linear Classifier

$$f_k = \frac{\exp(\xi_k)}{\sum_j \exp(\xi_j)}$$

$$\frac{\partial f_s}{\partial \xi_3} = \frac{1}{\text{DEN}} \cdot \frac{\partial \text{NUM}}{\partial \xi_3} - \frac{\text{NUM}}{\text{DEN}^2} \cdot \frac{\partial \text{DEN}}{\partial \xi_3}$$

$$= \frac{1}{\sum_j \exp(\xi_j)} \cdot \frac{\partial \exp(\xi_s)}{\partial \xi_3} - \frac{\exp(\xi_s)}{(\sum_j \exp(\xi_j))^2} \frac{\partial \text{DEN}}{\partial \xi_3}$$

$$= 0 - \frac{\exp(\xi_s)}{(\sum_j \exp(\xi_j))^2} \frac{\partial (\sum_j \exp(\xi_j))}{\partial \xi_3}$$

$$\frac{\partial e^{\xi_3}}{\partial \xi_3} = e^{\xi_3}$$

$$\frac{\partial f_s}{\partial \xi_3} = - \frac{\exp(\xi_s) \exp(\xi_3)}{(\sum_j \exp(\xi_j))^2} = -f_s f_3$$

10) PRIVACY

