

#### deeplearning.ai

# Basics of Neural Network Programming

### Logistic Regression Gradient descent

#### Logistic regression recap

$$\Rightarrow z = w^{T}x + b$$

$$\Rightarrow \hat{y} = a = \sigma(z)$$

$$\Rightarrow \mathcal{L}(a, y) = -(y \log(a) + (1 - y) \log(1 - a))$$

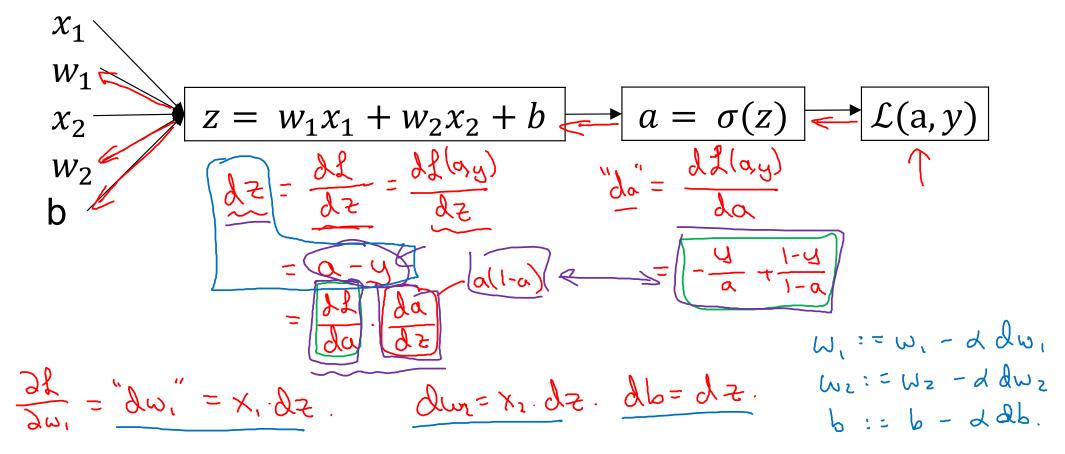
$$x_{1}$$

$$y_{2} = \omega_{1}x_{1} + \omega_{2}x_{2} + b$$

$$y_{3} = \omega_{1}x_{2} + \omega_{2}x_{3} + b$$

$$y_{4} = \omega_{1}x_{1} + \omega_{2}x_{2} + b$$

#### Logistic regression derivatives



Andrew Ng



deeplearning.ai

# Basics of Neural Network Programming

Gradient descent on m examples

#### Logistic regression on *m* examples

$$\frac{J(\omega,b)}{J(\omega,b)} = \frac{1}{m} \sum_{i=1}^{m} \chi(\alpha^{(i)}, y^{(i)})$$

$$\Rightarrow \alpha^{(i)} = \gamma^{(i)} = G(z^{(i)}) = G(\omega^{T} \chi^{(i)} + b)$$

$$\frac{\partial}{\partial \omega_{i}} \mathcal{I}(\omega, b) = \frac{1}{m} \sum_{i=1}^{m} \frac{\partial}{\partial \omega_{i}} \mathcal{I}(a^{(i)}, y^{(i)})$$

$$\frac{\partial}{\partial \omega_{i}} \mathcal{I}(\omega, b) = \frac{1}{m} \sum_{i=1}^{m} \frac{\partial}{\partial \omega_{i}} \mathcal{I}(a^{(i)}, y^{(i)})$$

### Logistic regression on *m* examples

$$J=0; d\omega_{1}=0; d\omega_{2}=0; db=0$$

$$For i=1 to m$$

$$Z^{(i)}=\omega^{T}\chi^{(i)}+b$$

$$Q^{(i)}=6(Z^{(i)})$$

$$J+=-[y^{(i)}(\log_{1}Q^{(i)}+(1-y^{(i)})\log_{1}(1-q^{(i)})]$$

$$dZ^{(i)}=Q^{(i)}-y^{(i)}$$

$$dZ^{(i)}=Q^{(i)}-y^{(i)}$$

$$dZ^{(i)}=Q^{(i)}-y^{(i)}$$

$$dZ^{(i)}=Q^{(i)}$$

$$dZ^{(i)}$$

$$d\omega_1 = \frac{\partial J}{\partial \omega_1}$$

$$W_1 := W_1 - d d w_1$$
 $W_2 := W_2 - d d w_2$ 
 $b := b - d d b$