



deeplearning.ai

Basics of Neural Network

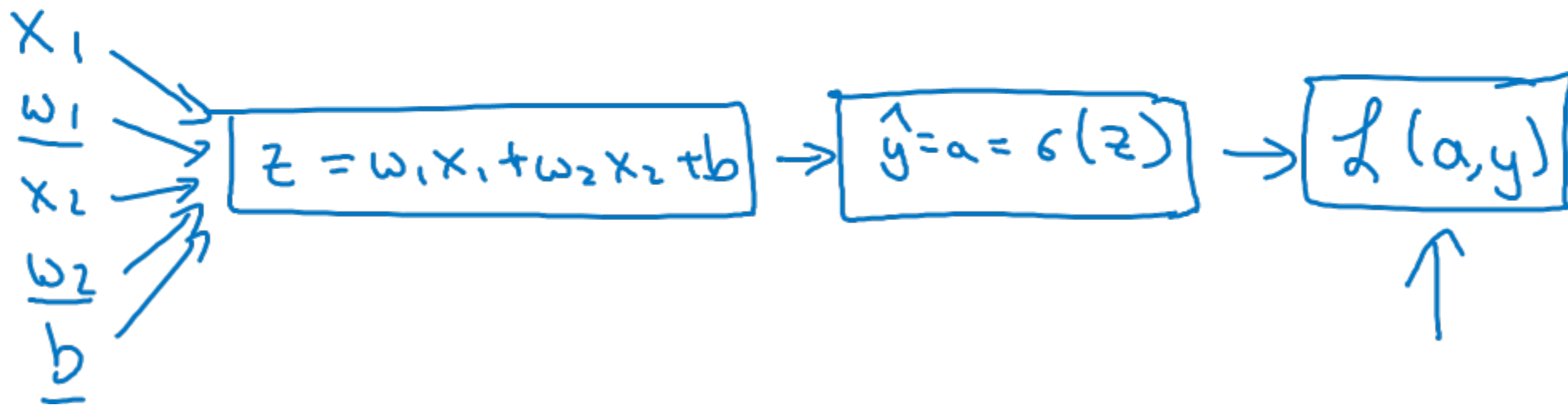
Programming
Logistic Regression
Gradient descent

Logistic regression recap

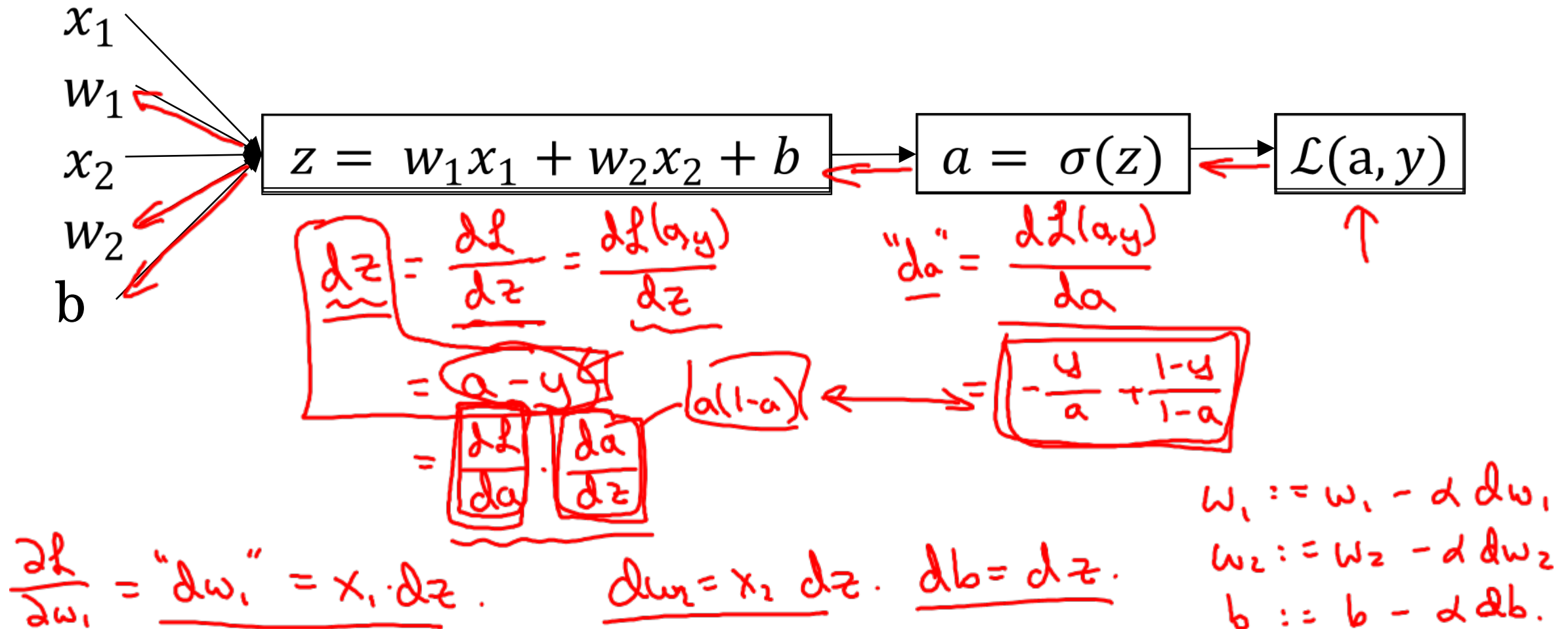
→ $z = w^T x + b$

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

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



Logistic regression derivatives





deeplearning.ai

Basics of Neural Network — Programming — Gradient descent on *m* examples

Logistic regression on m examples

$$\underline{J(w, b)} = \underline{\frac{1}{m} \sum_{i=1}^m \ell(a^{(i)}, y^{(i)})}$$

$$\rightarrow a^{(i)} = \hat{y}^{(i)} = \sigma(z^{(i)}) = \sigma(w^T x^{(i)} + b)$$

$$(x^{(i)}, y^{(i)})$$

$$\underline{dw_1^{(i)}}, \underline{dw_2^{(i)}}, \underline{db^{(i)}}$$

$$\underline{\frac{\partial}{\partial w_1} J(w, b)} = \frac{1}{m} \sum_{i=1}^m \underbrace{\frac{\partial}{\partial w_1} \ell(a^{(i)}, y^{(i)})}_{\underline{dw_1^{(i)}} - (x^{(i)}, y^{(i)})}$$

Logistic regression on m examples

$$J=0; \underline{dw}_1=0; \underline{dw}_2=0; \underline{db}=0$$

→ For $i=1$ to m

$$z^{(i)} = \mathbf{w}^T \mathbf{x}^{(i)} + b$$

$$a^{(i)} = \sigma(z^{(i)})$$

$$J += -[y^{(i)} \log a^{(i)} + (1-y^{(i)}) \log(1-a^{(i)})]$$

$$\underline{dz}^{(i)} = a^{(i)} - y^{(i)}$$

$$dw_1 += x_1^{(i)} dz^{(i)}$$

$$dw_2 += x_2^{(i)} dz^{(i)}$$

$$db += dz^{(i)}$$

$n=2$

dw_3
 \vdots
 dw_n

$J /= m \leftarrow$

$$\underline{dw}_1 /= m; \quad \underline{dw}_2 /= m; \quad \underline{db} /= m. \leftarrow$$

$$\underline{dw}_1 = \frac{\partial J}{\partial w_1}$$

$$w_1 := w_1 - \alpha \underline{dw}_1$$

$$w_2 := w_2 - \alpha \underline{dw}_2$$

$$b := b - \alpha \underline{db}$$

Vectorization