

Assignment 3: Neural Network By Hand!

PART 1: Derivation of Backpropagation.

recall: $\forall l = 1, \dots, N_{\text{layer}},$

$$z^{(i,l)} = (W^{(l)})^T a^{(i,l-1)} + b^{(l)} \quad (*)$$

$$a^{(i,l)} = f(z^{(i,l)}) \quad (**)$$

with,

$$\left\{ \begin{array}{l} W^{(l)} = (w_{jk})_{\substack{1 \leq j \leq m_{l-1} \\ 1 \leq k \leq m_l}}; \quad W^{(l)} \in \mathbb{R}^{m_{l-1} \times m_l} \\ a^{(i,l-1)} \in \mathbb{R}^{m_{l-1}} \\ b^{(l)} \in \mathbb{R}^{m_l}, \quad b^{(l)} = (b_k^{(l)})_{1 \leq k \leq m_l} \end{array} \right.$$

therefore,

$$\left\{ \begin{array}{l} Z^{(i,l)} = (z_k^{(i,l)})_{1 \leq k \leq m_l} \in \mathbb{R}^{m_l} \quad \text{with } z_k^{(i,l)} = \sum_{j=1}^{m_{l-1}} w_{kj} a_j^{(i,l-1)} + b_k^{(l)} \quad \forall l=1, \dots, N_{\text{layer}} \\ \text{with } z_k^{(i,l)} = \sum_{j=1}^{m_{l-1}} w_{kj} a_j^{(i,l-1)} + b_k^{(l)} \quad \forall 1 \leq k \leq m_l \end{array} \right. \quad (1)$$

and

$$a^{(i,l)} = (a_k^{(i,l)})_{1 \leq k \leq m_l} \in \mathbb{R}^{m_l} \quad \text{with } a_k^{(i,l)} = f(z_k^{(i,l)}) \quad (2)$$

Now let's derive the equations:

→ Equation (108): Error at the Last layer:

$$\delta^{(i,L)} = f'(z^{(i,L)}) \odot L'(a^{(i,L)}, y^{(i)})$$

$$\delta^{(i,L)} = \left(\delta_k^{(i,L)} \right)_{1 \leq k \leq m_L} \in \mathbb{R}^{m_L}, \quad \text{with}$$

$$\delta_k^{(i,L)} = f'(z_k^{(i,L)}) L'(a_k^{(i,L)}, y^{(i)}) \quad (3)$$

where, $z_k^{(i,L)}$ and $a_k^{(i,L)}$ are given by (1) and (2)

→ Equation (100): Derivative of the Cost at the Last layer:

$$\frac{\partial \text{Cost}}{\partial W^{(L)}} = \frac{1}{m} \sum_{i=1}^m a^{(i,L-1)} \otimes \delta^{(i,L)}$$

$$= \frac{1}{m} \sum_{i=1}^m a^{(i,L-1)} (\delta^{(i,L)})^T$$

$$\frac{\partial \text{Cost}}{\partial W^{(L)}} = \left(\frac{1}{m} \sum_{i=1}^m C_{jk}^{(i,L)} \right)_{\substack{1 \leq j \leq m_{L-1} \\ 1 \leq k \leq m_L}} \in \mathbb{R}^{m_{L-1} \times m_L}$$

with $C_{jk}^{(i,L)} = a_j^{(i,L-1)} \times \delta_k^{(i,L)}$ where $\delta_k^{(i,L)}$ is given by (3)

→ Equation (110) :

$$g^{(i, L-1)} = f(Z^{(i, L-1)}) \odot [W^{(L)} g^{(i, L)}]$$

$$g^{(i, L-1)} = \left(g_j^{(i, L-1)} \right)_{1 \leq j \leq m_{L-1}} \in \mathbb{R}^{m_{L-1}}$$

$$\text{with } g_j^{(i, L-1)} = f(Z_j^{(i, L-1)}) \times \sum_{k=1}^{m_L} w_{jk} g_k^{(i, L)} \quad (4)$$

→ Equation (111) :

$$\begin{aligned} \frac{\partial \text{Cost}}{\partial W^{(L-1)}} &= \frac{1}{m} \sum_{i=1}^m a^{(i, L-2)} \otimes [g^{(i, L-1)}] \\ &= \frac{1}{m} \sum_{i=1}^m a^{(i, L-2)} (g^{(i, L-1)})^T \end{aligned}$$

$$\frac{\partial \text{Cost}}{\partial W^{(L-1)}} = \left(\frac{1}{m} \sum_{i=1}^m C_{jk}^{(i, L-2)} \right)_{\substack{1 \leq j \leq m_{L-2} \\ 1 \leq k \leq m_{L-1}}} \in \mathbb{R}^{m_{L-2} \times m_{L-1}}$$

$$\text{with, } C_{jk}^{(i, L-2)} = a_j^{(i, L-2)} \times g_k^{(i, L-1)}$$

where $g_k^{(i, L-1)}$ is given by (4) .

→ Equation (4.2):

$$g^{(i,l)} = f'(z^{(i,l)}) \otimes [W^{(l+1)} g^{(i,l+1)}]$$

$$g^{(i,l)} = (g_j^{(i,l)})_{1 \leq j \leq n_l} \in \mathbb{R}^{n_l}$$

$$\text{with } g_j^{(i,l)} = f'(z_j^{(i,l)}) \sum_{k=1}^{n_{l+1}} w_{jk} g_k^{(i,l+1)}, \quad \forall 1 \leq j \leq n_l \quad (5)$$

→ Equation (4.3):

$$\frac{\partial \text{Cost}}{\partial W^{(l)}} = \frac{1}{m} \sum_{i=1}^m a^{(i,l-1)} \otimes [g^{(i,l)}]$$

$$= \frac{1}{m} \sum_{i=1}^m a^{(i,l-1)} (g^{(i,l)})^T$$

$$= \left(\frac{1}{m} \sum_{i=1}^m c_{jk}^{(i,l)} \right)_{\substack{1 \leq j \leq n_{l-1} \\ 1 \leq k \leq n_l}} \in \mathbb{R}^{n_{l-1} \times n_l}$$

$$\text{with, } c_{jk}^{(i,l)} = a_j^{(i,l-1)} \times g_k^{(i,l)}$$