

CS 4375 Assignment 2: NN

1)

$$1.1) \text{ Error } (E_d) = \frac{1}{2} \sum_k (t_k - o_k)^2$$

$$\Delta w_{ji} = -\eta \frac{\partial E_d}{\partial w_{ji}}$$

$$\text{Net}_j = \sum_{i=1}^n x_{ji} w_{ji}$$

$$\frac{\partial E_d}{\partial w_{ji}} = \frac{\partial E_d}{\partial \text{net}_j} \times \frac{\partial \text{net}_j}{\partial w_{ji}}$$

$$\frac{\partial \text{net}_j}{\partial w_{ji}} = x_{ji}$$

$$\frac{\partial E_d}{\partial \text{net}_j} = \frac{\partial E_d}{\partial o_j} \times \frac{\partial o_j}{\partial \text{net}_j}$$

$$\frac{\partial E_d}{\partial o_j} = -(t_j - o_j)$$

$$a) \tanh(x) = \frac{e^x - e^{-x}}{e^x + e^{-x}} = o_j \quad x = \text{net}_j$$

$$\frac{\partial o_j}{\partial \text{net}_j} = \frac{(e^x + e^{-x})(e^x + e^{-x}) - (e^x - e^{-x})(e^x - e^{-x})}{(e^x + e^{-x})^2}$$

$$= \frac{4}{(e^x + e^{-x})^2} = 1 - (\tanh(x))^2 = 1 - (o_j)^2$$

Case 1: output unit:

$$\frac{\partial E_d}{\partial \text{net}_j} = -(t_j - o_j)(1 - o_j^2) = -\delta_j$$

$$\therefore \Delta w_{ji} = -\eta \frac{\partial E_d}{\partial w_{ji}} = \boxed{\eta (t_j - o_j) (1 - o_j^2) x_{ji}}$$

Case 2: j is hidden layer.

$$\frac{\partial E_d}{\partial net_j} = -\sum_k \delta_k w_{kj} (1 - o_j^2)$$

$$\therefore \delta_j = (1 - o_j^2) \sum_k \delta_k w_{kj}$$

$$\therefore \Delta w_{ji} = \boxed{\eta (1 - o_j^2) \sum_k \delta_k w_{kj} x_{ji}}$$

$$b) \text{Relu}(x) = \max(0, x) = \begin{cases} 0 & x \leq 0 \\ x & x > 0 \end{cases}$$

$$= o_j$$

$$\frac{\partial o_j}{\partial net_j} = \begin{cases} 0 & x \leq 0 \\ 1 & x > 0 \end{cases}$$

Case 1: Output unit:

$$\frac{\partial E_d}{\partial net_j} = \begin{cases} 0 & x \leq 0 \\ -(t_j - o_j) & x > 0 \end{cases}$$

$$\Delta w_{ji} = \begin{cases} 0 & x \leq 0 \\ -\eta (t_j - o_j) x_{ji} & x > 0 \end{cases}$$

Case 2: j is hidden layer.

$$\frac{\partial E_d}{\partial \text{net}_j} = \begin{cases} 0 & x \leq 0 \\ \sum_k -w_{kj} \delta_k & x > 0 \end{cases}$$

$$\therefore \Delta w_{ji} = \begin{cases} 0 & x \leq 0 \\ \eta x_j \sum_k -w_{kj} \delta_k & x > 0 \end{cases}$$

$$1.2 \quad w_\theta = w_0 + w_1(x_1 + x_1^2) + \dots + w_n(x_n + x_n^2)$$

$$w_{j,\text{new}} = w_{j,\text{old}} - \eta \frac{\partial J(\theta)}{\partial w_j}$$

$$\frac{\partial J(\theta)}{\partial w_j} = \frac{\partial}{\partial w_j} \frac{1}{2m} \sum_{i=1}^m (w_\theta(x^{(i)}) - y^{(i)})^2$$

$$= \frac{\partial}{\partial w_j} \frac{1}{2m} \sum_{i=1}^m ((w_0 + w_1(x_1 + x_1^2) + \dots + w_n(x_n + x_n^2)) - y^{(i)})^2$$

$$= \frac{1}{m} \sum_{i=1}^m (w_\theta(x^{(i)}) - y^{(i)}) (x_j + x_j^2)^{(i)}$$

$$\therefore w_{j,\text{new}} = w_{j,\text{old}} - \eta \frac{1}{m} \sum_{i=1}^m (w_\theta x^i - y^i) (x_j + x_j^2)^i$$

$$1.3 \quad o_3 = h(x_1 w_{31} + x_2 w_{32})$$

$$o_4 = h(x_1 w_{41} + x_2 w_{42})$$

$$O_5 = h(w_{53} h(x_1 w_{31} + x_2 w_{32}) + w_{54} h(x_1 w_{41} + x_2 w_{42}))$$

b)

$$O_{\text{hidden layer}} = h(w^{(1)} \cdot x)$$

$$O_{\text{output}} = h(w^{(2)} \cdot h(w^{(1)} \cdot x))$$

$$c) h_s(z) = \frac{1}{1+e^{-z}} \quad h_f(z) = \frac{e^z - e^{-z}}{e^z + e^{-z}}$$

$$h_f(z) = \frac{e^z(1 - e^{-2z})}{e^z(1 + e^{-2z})}$$

$$\therefore h_f(z) = \frac{1 - e^{-2z}}{1 + e^{-2z}} = h_s(2z) - \frac{e^{-2z}}{1 + e^{-2z}}$$

$$= h_s(2z) - \frac{(1 + e^{-2z})}{1 + e^{-2z}} + \frac{1}{1 + e^{-2z}}$$

$$h_f(z) = \boxed{2 h_s(2z) - 1}$$

\therefore Since parameters differ only by linear transformation and constants, output functions are similar: $\therefore O_5 = h_f(w^{(2)} h_f(w^{(1)}(x)))$

$$= h_f(w^{(2)} (2 \cdot h_s(2 \cdot w^{(1)}(x)) + \begin{bmatrix} 1 \\ 1 \end{bmatrix}))$$

$$\therefore O_5 = 2 h_s(2 w^{(2)} (2 \cdot h_s(2 \cdot w^{(1)}(x)) + \begin{bmatrix} 1 \\ 1 \end{bmatrix})) + 1$$

1.4)

Q: 4.16

$$E(\vec{w}) = \frac{1}{2} \sum_{d \in D} \sum_{k \in \text{outputs}} (t_{kd} - o_{kd})^2 + \gamma \sum_{i,j} w_{ji}^2$$

$$\Delta w_{ji} = -\eta \frac{\partial E}{\partial w_{ji}} + \frac{\partial}{\partial w_{ji}} \left(\gamma \sum_{i,j} w_{ji}^2 \right)$$

$$\therefore \Delta w_{ji} = -\eta \left(\frac{\partial E_d}{\partial w_{ji}} + 2\gamma w_{ji} \right)$$

Output layer $\therefore \Delta w_{ji} = -\eta \left((t_j - o_j) o_j (1 - o_j) x_{ji} + 2\gamma w_{ji} \right)$

Hidden layer $\Delta w_{ji} = -\eta \left(\delta_j x_{ji} - 2\gamma w_{ji} \right)$

where $\delta_j = o_j (1 - o_j) \sum_{k \in \text{outputs}} \delta_k w_{kj}$