

1. for the output layer neuron uses the identity function $f(x) = x$

$$E_d(W) = \frac{1}{2} \sum_{k \in \text{outputs}} (t_k - o_k)^2 \quad \Delta W_{ji} = -\eta \frac{\partial E_d}{\partial W_{ji}}$$

$$\frac{\partial E_d}{\partial W_{ji}} = \frac{\partial E_d}{\partial \text{net}_j} \cdot \frac{\partial \text{net}_j}{\partial W_{ji}} \quad \frac{\partial \text{net}_j}{\partial W_{ji}} = x_{ji}$$

$$\frac{\partial E_d}{\partial \text{net}_j} = \frac{\partial E_d}{\partial o_j} \cdot \frac{\partial o_j}{\partial \text{net}_j} \quad \frac{\partial E_d}{\partial o_j} = \frac{\partial}{\partial o_j} \left[\frac{1}{2} \sum_{k \in \text{outputs}} (t_k - o_k)^2 \right] = \frac{\partial}{\partial o_j} \left[\frac{1}{2} (t_j - o_j)^2 \right]$$

$$= -(t_j - o_j) \quad \frac{\partial o_j}{\partial \text{net}_j} = 1 \quad \frac{\partial E_d}{\partial o_j} \cdot \frac{\partial o_j}{\partial \text{net}_j} = -(t_j - o_j) \quad \Delta W_{ji} = \eta (t_j - o_j) x_{ji}$$

for hidden layer uses tanh function $f(x) = \tanh(x) = \frac{e^x - e^{-x}}{e^x + e^{-x}}$

$$\frac{\partial E_d}{\partial W_{ji}} = \frac{\partial E_d}{\partial \text{net}_j} \cdot \frac{\partial \text{net}_j}{\partial W_{ji}} \quad \frac{\partial \text{net}_j}{\partial W_{ji}} = x_{ji}$$

$$\frac{\partial E_d}{\partial \text{net}_j} = \sum_{k \in \text{Downstream}(j)} \frac{\partial E_d}{\partial \text{net}_k} \cdot \frac{\partial \text{net}_k}{\partial \text{net}_j} = \sum_{k \in \text{Downstream}(j)} -\delta_k \cdot \frac{\partial \text{net}_k}{\partial \text{net}_j} = \sum_{k \in \text{Downstream}(j)} \frac{\partial \text{net}_k}{\partial o_j} \cdot \frac{\partial o_j}{\partial \text{net}_j}$$

$$= \sum_{k \in \text{Downstream}(j)} -\delta_k W_{kj} \cdot \frac{\partial o_j}{\partial \text{net}_j} = \sum_{k \in \text{Downstream}(j)} -\delta_k W_{kj} (1 - o_j^2) \quad \delta_j = (1 - o_j^2) \sum_{k \in \text{Downstream}(j)} \delta_k W_{kj}$$

$$\Delta W_{ji} = \eta \delta_j x_{ji}$$

2. $o = W_0 + W_1(x_1 + x_1^2) + \dots + W_n(x_n + x_n^2)$

$$E(W) = \frac{1}{2} \sum_{d \in D} (t_d - o_d)^2 \quad \frac{\partial E}{\partial W_i} = \frac{\partial}{\partial W_i} \frac{1}{2} \sum_{d \in D} (t_d - o_d)^2 = \frac{1}{2} \sum_{d \in D} \frac{\partial}{\partial W_i} (t_d - o_d)^2$$

$$= \frac{1}{2} \sum_{d \in D} (t_d - o_d) \frac{\partial}{\partial W_i} (t_d - o_d) = \sum_{d \in D} (t_d - o_d) (-x_{id} + x_{id}^2) = \sum_{d \in D} (t_d - o_d) (-x_{id} - x_{id}^2)$$

$$\Delta W_i = -\eta \sum_{d \in D} (t_d - o_d) (-x_{id} - x_{id}^2)$$

3. a. $\text{net}_3 = x_1 \cdot W_{31} + x_2 \cdot W_{32} \quad \text{net}_4 = x_1 \cdot W_{41} + x_2 \cdot W_{42}$

$$\text{net}(y_5) = h(\text{net}_3) \cdot W_{53} + h(\text{net}_4) \cdot W_{54} \quad o = h(\text{net}(y_5))$$

$$\text{b. output} = h(h(x \cdot W^{(1)}) \cdot W^{(2)})$$

$$\text{c. } h_1(x) = \frac{1}{1+e^{-x}} = \frac{e^x}{1+e^x} \quad h_2(x) = \tanh(x) = \frac{e^x - e^{-x}}{e^x + e^{-x}}$$

$$h_2(x) = 2h_1(2x) - 1$$

$$\text{output} = h_1(h_1(\text{net}_3) \cdot W_{53} + h_1(\text{net}_4) \cdot W_{54})$$

$$\text{output} = h_2(h_2(\text{net}_3) \cdot W_{53} + h_2(\text{net}_4) \cdot W_{54}) = 2h_1(2h_2(\text{net}_3)W_{53} + 2h_2(\text{net}_4)W_{54})$$

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4. $W_{ji} = W_{ji} + \Delta W_{ji} \quad \Delta W_{ji} = -\eta \frac{\partial E_d}{\partial W_{ji}}$

for output layer $\frac{\partial E_d}{\partial W_{ji}} = \frac{\partial}{\partial W_{ji}} \frac{1}{2} \sum_{d \in D} \sum_{k \in \text{outputs}} (t_{kd} - O_{kd})^2 + \frac{\partial}{\partial W_{ji}} r W_{ji}^2$

$$= -(t_j - O_j) O_j (1 - O_j) x_{ji} + 2r W_{ji} \quad \Delta W_{ji} = \eta (t_j - O_j) O_j (1 - O_j) x_{ji} - 2r \eta W_{ji}$$

$$W_{ji} = W_{ji} + \Delta W_{ji} = (1 - 2r\eta) W_{ji} + \eta (t_j - O_j) O_j (1 - O_j) x_{ji}$$

for hidden layer $\delta_j = O_j(1 - O_j) \sum_{k \in \text{hidden layer}} \delta_k W_{kj} \quad \Delta W_{ji} = \eta \delta_j x_{ji} - 2r \eta W_{ji}$

$$W_{ji} = W_{ji} + \Delta W_{ji} = (1 - 2r\eta) W_{ji} + \eta \delta_j x_{ji}$$