

I.

we will use $\tanh(x) = \frac{e^x - e^{-x}}{e^x + e^{-x}}$ activation

case I: $\frac{\partial E_d}{\partial net_k}$ for output unit

$$= \frac{\partial E_d}{\partial o_k} \cdot \frac{\partial o_k}{\partial net_k}$$

$$\frac{\partial E_d}{\partial o_k} = \frac{1}{2} \sum_{k \text{ - outputs}} \frac{\partial}{\partial o_j} (t_k - o_k)^2$$

$$= \frac{1}{2} \sum_{k \text{ - outputs}} 2(t_k - o_k) \frac{\partial}{\partial o_k} (t_k - o_k)$$

$$= \sum_{k \text{ - outputs}} -(t_k - o_k)$$

$$\frac{\partial O_K}{\partial \text{net}_K} = 1$$

$$\frac{\partial E_d}{\partial \text{net}_K} = -(t_K - O_K)$$

Let's say

$$-(t_K - O_K) = -\delta_K$$

we will call the partial of Error w.r.t net for any unit K as follows.

$$\frac{\partial E_d}{\partial \text{net}_j} = -\delta_K$$

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

$$\Delta w_{ji} = \eta \delta_K x_{ki} \quad \text{where } \delta_K = (t_K - O_K)$$

$$\boxed{\delta_0 \leftarrow (t_K - O_K)}$$

output

Case II: K is a hidden unit

$$\begin{aligned}\frac{\partial E_d}{\partial \text{net}_h} &= \sum_{K \in \text{DownStream}(h)} \frac{\partial E_d}{\partial \text{net}_K} \frac{\partial \text{net}_K}{\partial \text{net}_h} \\&= \sum_{K \in \text{DownStream}(h)} -\delta_K \frac{\partial \text{net}_K}{\partial \text{net}_h} \frac{\partial O_h}{\partial \text{net}_h} \\&= \sum_{K \in \text{DownStream}(h)} -\delta_K \omega_{Kh} \frac{\partial O_h}{\partial \text{net}_h}\end{aligned}$$

Now, $\frac{\partial O_h}{\partial \text{net}_h} = (1 - O_h)^2 \rightarrow$ tanh instead of sigmoid

$$\therefore \frac{\partial E_d}{\partial \text{net}_h} = -(1 - O_h^2) \sum_{K \in \text{DownStream}(h)} \omega_{Kh} \delta_K$$

$$\boxed{\delta_h \leftarrow (1 - O_h^2) \sum_{K \in \text{DownStream}(h)} \omega_{Kh} \delta_K}$$

2. $0 = \omega_0 + \omega_1 (x_1 + x_2)^2 +$
 Ans $0 = \omega_0 + \omega_1 x_1 + \omega_1 x_1^2 +$

\Rightarrow First, Error function is defined as

$$E(\vec{w}) = \frac{1}{2} \sum_{d \in D} (t_d - o_d)^2$$

\Rightarrow update rule is the same $\omega_i = \omega_i + \Delta \omega_i$
 where $\Delta \omega_i = -\eta \frac{\partial E}{\partial \omega_i}$

$$\begin{aligned} \Rightarrow \text{For } \omega_0, \frac{\partial E}{\partial \omega_0} &= \frac{\partial}{\partial \omega_0} \frac{1}{2} \sum_{d \in D} (t_d - o_d)^2 \\ &= \frac{1}{2} \sum_{d \in D} \frac{\partial}{\partial \omega_0} (t_d - o_d)^2 \\ &= \frac{1}{2} \sum_{d \in D} 2(t_d - o_d) \frac{\partial}{\partial \omega_0} (t_d - o_d) \\ &= \sum_{d \in D} (t_d - o_d) (-1) \\ &= - \sum_{d \in D} (t_d - o_d) \end{aligned}$$

$$\boxed{\text{So, } \Delta \omega_0 = \eta \sum_{d \in D} (t_d - o_d)}$$

\Rightarrow For $\omega_1, \omega_2, \dots, \omega_n$

$$\begin{aligned} \frac{\partial E}{\partial \omega_i} &= \frac{\partial}{\partial \omega_i} \frac{1}{2} \sum_{d \in D} (t_d - o_d)^2 \\ &= \frac{1}{2} \sum_{d \in D} \frac{\partial}{\partial \omega_i} (t_d - o_d)^2 \\ &= \frac{1}{2} \sum_{d \in D} 2(t_d - o_d) \frac{\partial}{\partial \omega_i} (t_d - o_d) \\ &= \sum_{d \in D} (t_d - o_d) (-(x_{id} + x_{id}^2)) \end{aligned}$$

$$\boxed{\text{So, } \Delta \omega_i = \eta \sum_{d \in D} (t_d - o_d) (x_{id} + x_{id}^2)}$$

3.	Node	Net	output
	1	X_1	X_1
	2	X_2	X_2
	3	$net_3 = w_{31}X_1 + w_{32}X_2$	$X_3 = F(net_3)$
	4	$net_4 = w_{41}X_1 + w_{42}X_2$	$X_4 = F(net_4)$
	5	$net_5 = w_{53}X_3 + w_{54}X_4$	$X_5 = F(net_5)$

$$\begin{aligned}
 \textcircled{a} \quad \text{O/p } Y_5 &= F(net_5) \\
 &= F(w_{53}X_3 + w_{54}X_4) \\
 &= F(w_{53}[F(w_{31}X_1 + w_{32}X_2)] \\
 &\quad + w_{54}[F(w_{41}X_1 + w_{42}X_2)])
 \end{aligned}$$

$$\textcircled{b} \quad H[w^{(2)}, H(w^{(1)} \cdot X)]$$

$$\begin{aligned}
 \textcircled{c} \quad h_1(x) &= \frac{1}{1+e^{-x}} & h_2(x) &= \frac{e^x - e^{-x}}{e^x + e^{-x}} \\
 &= \frac{e^x}{1+e^x} & &= \frac{e^{2x} - 1}{e^{2x} + 1}
 \end{aligned}$$

$$h_1(2x) = \frac{e^{2x}}{1+e^{2x}}$$

$$2h_1(2x) = \frac{2e^{2x}}{e^{2x}+1}$$

$$\begin{aligned}
 2h_1(2x) - 1 &= \frac{2e^{2x}}{e^{2x}+1} - 1 \\
 &= \frac{e^{2x}-1}{e^{2x}+1} = h_2(x)
 \end{aligned}$$

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

$\therefore h_2(x)$ is a rescaled $h_1(x)$ function

→ here $h_1(x)$ & $h_2(x)$ is differing by linear transformations & constants.

→ $h_1(2x)$ multiply by 2 is linear transformation of $h_2(x)$, where subtracting 1 from $2h_1(2x)$ is constant scaling of $h_2(x)$.

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

$$w_{ji} \leftarrow w_{ji} + \Delta w_{ji}$$

$$\Delta w_{ji} = -\eta \frac{\partial E(\vec{w})}{\partial w_{ji}}$$

$$\frac{\partial E(\vec{w})}{\partial w_{ji}} = \underbrace{\frac{\partial}{\partial w_{ji}} \frac{1}{2} \sum_{d \in D} \sum_{k \in O} (t_{kd} - o_{kd})^2}_{(1)} + \underbrace{\gamma \frac{\partial}{\partial w_{ji}} w_{ji}^2}_{(2)}$$

the first term already derived in Back Propagation Algo.

$$\therefore \frac{\partial E(\vec{w})}{\partial w_{ji}} = -(t_j - o_j) o_j (1 - o_j) x_{ji} + 2\gamma w_{ji}$$

⇒ For output nodes,
 $w_{ji} \leftarrow w_{ji} + \eta (t_j - o_j) o_j (1 - o_j) x_{ji} - 2\gamma w_{ji}$

$$w_{ji} \leftarrow (1 - 2\eta\gamma) w_{ji} + \eta (t_j - o_j) o_j (1 - o_j) x_{ji}$$

$$w_{ji} \leftarrow \beta w_{ji} + \eta \delta_j x_{ji}$$

$$\text{where } \beta = 1 - 2\eta\gamma \text{ \& } \delta_j = (t_j - o_j) o_j (1 - o_j)$$

⇒ For hidden layers,
 $w_{ji} \leftarrow \beta w_{ji} + \eta \delta_j x_{ji}$

$$\text{where } \beta = 1 - 2\eta\gamma \text{ \& } \delta_j = o_j (1 - o_j) \sum_{k \in \text{Downstream}(j)} \delta_k w_{kj}$$

⇒ So we have multiplier β in both derivatives.