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$$\begin{aligned}
 \text{a) } \tanh(x) &= \frac{e^x - e^{-x}}{e^x + e^{-x}} \\
 \frac{d \tanh(x)}{dx} &= \frac{(e^x + e^{-x})(e^x + e^{-x}) - (e^x - e^{-x})(e^x - e^{-x})}{(e^x + e^{-x})^2} \\
 i_{\text{net}} &= \frac{(e^x + e^{-x})^2 - (e^x - e^{-x})^2}{(e^x + e^{-x})^2} \\
 i_{\text{net}} &= \left| \frac{\frac{e^x - e^{-x}}{e^x + e^{-x}}^2 - (\tanh(x))^2}{1} \right|
 \end{aligned}$$

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

$$w_{ji}^{n+1} = w_{ji}^{old} + (\Delta w_{ji})$$

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

$$\begin{aligned}
 \frac{\partial E_d}{\partial w_{ji}} &= \frac{\partial E_d}{\partial \text{net}_i} \frac{\partial \text{net}_i}{\partial w_{ji}} & \frac{\partial \text{net}_i}{\partial w_{ji}} &= (\sum x_{ji} w_{ji})' = x_{ji}
 \end{aligned}$$

Case 1: j is an output unit

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

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

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

$$\begin{aligned}
 \frac{\partial E_d}{\partial o_j} &= \frac{\partial}{\partial o_j} \left[\frac{1}{2} \sum_{k \in \text{outputs}} (t_k - o_k)^2 \right] \\
 &= \frac{\partial}{\partial o_j} \left[\frac{1}{2} (t_j - o_j)^2 \right] \\
 &= -(t_j - o_j)
 \end{aligned}$$

$$-\delta_j x_{ji} (1 - o_j^2)$$

Case 2: j is a hidden unit.

$$\begin{aligned}\frac{\partial E_d}{\partial \text{net}_j} &= \sum_{\text{Kedownstream}(j)} \frac{\partial E_d}{\partial \text{net}_k} \frac{\partial \text{net}_k}{\partial \text{net}_j} \\&= \sum_{\text{Kedownstream}(j)} -\delta_k \frac{\partial \text{net}_k}{\partial \text{net}_j} \\&= \sum_{\text{Kedownstream}(j)} -\delta_k \frac{\partial \text{net}_k}{\partial o_j} \frac{\partial o_j}{\partial \text{net}_j} \\&= \sum_{\text{Kedownstream}(j)} -\delta_k w_{kj} (1 - o_j^2) \\&= \delta_j = (1 - o_j^2) \sum_{\text{Kedownstream}(j)} -\delta_k w_{kj}\end{aligned}$$

$$\Delta w_{ji} = -n \frac{\partial E_d}{\partial w_{ji}} = n \delta_j k_{ji}$$

Summary:

Case 1: j is output layer node:

$$\delta_j = (t_j - o_j) (1 - o_j^2)$$

$$w_{ji}^{\text{new}} = w_{ji}^{\text{old}} + \Delta w_{ji} \quad \text{where } \Delta w_{ji} = n \delta_j k_{ji}$$

Case 2: j is hidden layer node:

$$\begin{aligned}\delta_j &= (1 - o_j^2) \sum_{\text{Kedownstream}(j)} -\delta_k w_{kj} \\w_{ji}^{\text{new}} &= w_{ji}^{\text{old}} + \Delta w_{ji} \quad \text{where } \Delta w_{ji} = n \delta_j k_{ji}\end{aligned}$$

$$\text{ReLU}(x) = \begin{cases} x & ; x \in (0, \infty) \\ 0 & ; x \in (-\infty, 0] \end{cases}$$

$$\frac{d \text{ReLU}(x)}{dx} = \begin{cases} 1 & ; x \in (0, \infty) \\ 0 & ; x \in (-\infty, 0] \end{cases}$$

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

$$w_{ji}^{\text{new}} = w_{ji}^{\text{old}} + \Delta w_{ji} \quad \Delta w_{ji} = -n \frac{\partial E_d}{\partial w_{ji}}$$

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

$$\frac{\partial \text{net}_j}{\partial w_{ji}} = (\sum_{k \in \text{outputs}} w_{kj}) = x_{ji}$$

$$\frac{\partial E_d}{\partial \text{net}_j}$$

Case 1: j is an output unit

$$o_j = \text{ReLU}(\text{net}_j)$$

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

$$\frac{\partial o_j}{\partial \text{net}_j} = \begin{cases} 1 & ; \text{net}_j \in (0, \infty) \\ 0 & ; \text{net}_j \in (-\infty, 0] \end{cases}$$

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

$$\begin{aligned} \frac{\partial E_d}{\partial \text{net}_j} &= \begin{cases} -(t_j - o_j) & ; \text{net}_j \in (0, \infty) \\ 0 & ; \text{net}_j \in (-\infty, 0] \end{cases} &= \frac{\partial}{\partial o_j} \left[\frac{1}{2} (t_j - o_j)^2 \right] \\ &= -8_j &= -(t_j - o_j) \end{aligned}$$

$$\Delta w_{ji} = -n \frac{\partial E_d}{\partial w_{ji}} = n \delta_j x_{ji}$$

Case 2: j is a hidden unit:

$$\frac{\partial E_d}{\partial \text{net}_j} = \sum_{k \in \text{downstream}(j)} \frac{\partial E_d}{\partial \text{net}_k} \frac{\partial \text{net}_k}{\partial \text{net}_j}$$

$$= \sum_{k \in \text{downstream}(j)} -\delta_k \frac{\partial \text{net}_k}{\partial \text{net}_j}$$

$$= \sum_{k \in \text{downstream}(j)} -\delta_k \frac{\partial \text{net}_k}{\partial \sigma_j} \frac{\partial \sigma_j}{\partial \text{net}_j}$$

$$\delta_j = \begin{cases} \sum_{k \in \text{downstream}(j)} -\delta_k w_{kj} & ; \text{net}_j \in (0, \infty) \\ 0 & ; \text{net}_j \in (-\infty, 0] \end{cases}$$

$$\Delta w_{ji} = -n \frac{\partial E_d}{\partial w_{ji}} = n \delta_j x_{ji}$$

Summary: 8 where $\sigma_j = \text{ReLU}(\text{net}_j)$

Case 1: j is output layer node:

$$\delta_j = \begin{cases} (t_s - \sigma_j) & , \text{net}_j \in (0, \infty) \\ 0 & , \text{net}_j \in (-\infty, 0] \end{cases}$$

$$w_{ji}^{\text{new}} = w_{ji}^{\text{old}} + \Delta w_{ji} \quad \text{where } \Delta w_{ji} = n \delta_j x_{ji}$$

Case 2: j is hidden layer node:

$$\delta_j = \begin{cases} \sum_{k \in \text{downstream}(j)} -\delta_k w_{kj} & ; \text{net}_j \in (0, \infty) \\ 0 & ; \text{net}_j \in (-\infty, 0] \end{cases}$$

$$w_{ji}^{\text{new}} = w_{ji}^{\text{old}} + \Delta w_{ji} \quad \text{where } \Delta w_{ji} = n \delta_j x_{ji}$$

$$2. \quad O = w_0 + w_1(x_1 + x_1^2) + \dots + w_n(x_n + x_n^2)$$

$$E = \frac{1}{2} \sum (t - O)^2$$

$$w_i^{\text{new}} = w_i^{\text{old}} + \Delta w_i \quad \Delta w_j = -n \left(\frac{\partial E}{\partial w_i} \right)$$

$$\frac{\partial E}{\partial w_i} = \frac{\partial E}{\partial (t - O)} \frac{\partial (t - O)}{\partial w_i}$$

$$\frac{\partial E}{\partial (t - O)} = (t - O)$$

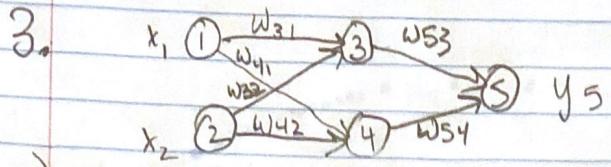
$$\frac{\partial (t - O)}{\partial w_i} = -\frac{\partial O}{\partial w_i} = -\frac{\partial (w_i(x_i + x_i^2))}{\partial w_i} = -(x_i + x_i^2)$$

$$\frac{\partial E}{\partial w_i} = -(t - O)(x_i + x_i^2)$$

$$w_i^{\text{new}} = w_i^{\text{old}} + \Delta w_j$$

$$\Delta w_j = -n \frac{\partial E}{\partial w_i}$$

$$\Delta w_j = n(t - O)(x_i + x_i^2)$$



a)

$$x_3 = w_{31}x_1 + w_{32}x_2$$

$$x_4 = w_{41}x_1 + w_{42}x_2$$

$$y_3 = h(x_3)$$

$$y_4 = h(x_4)$$

$$x_5 = w_{53}y_3 + w_{54}y_4$$

$$y_5 = h(x_5)$$

b)

$$\begin{pmatrix} x_3 \\ x_4 \end{pmatrix} = w^{(1)} X$$

$$\begin{pmatrix} y_3 \\ y_4 \end{pmatrix} = h(w^{(1)} X)$$

$$x_5 = w^{(2)} h(w^{(1)} X)$$

$$y_5 = h(w^{(2)} h(w^{(1)} X))$$

c) Sigmoid $h = \sigma$
 tanh $h = t$

$$y_{50} = A y_{5t} + B$$

$$\frac{1}{1 + e^{-x_5}} = A \frac{e^{x_5} - \bar{e}^{x_5}}{e^{x_5} + \bar{e}^{x_5}} + B$$

$$\frac{1}{1 + e^{-x_5}} = \frac{1}{1 + e^{-x_5}} \cdot \frac{e^{x_5} - \bar{e}^{x_5}}{e^{x_5} + \bar{e}^{x_5}} + \frac{2e^{-x_5}}{(1 + e^{-x_5})(e^{x_5} + \bar{e}^{-x_5})}$$

$$\frac{1}{1 + e^{-x_5}} = \frac{(e^{x_5} + \bar{e}^{-x_5})}{(1 + e^{-x_5})(e^{x_5} + \bar{e}^{-x_5})}$$

$$\frac{1}{1 + e^{-x_5}} = \frac{1}{1 + e^{-x_5}}$$

$$A = \frac{1}{1 + e^{-x_5}} \quad B = \frac{2e^{-x_5}}{(1 + e^{-x_5})(e^{x_5} + \bar{e}^{-x_5})}$$

$$\frac{\partial o_j}{\partial w_{ji}} = \frac{\partial O(\text{net}_j)}{\partial w_{ji}}$$

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

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

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

$$\frac{\partial E_d}{\partial w_{ji}} = \frac{\partial}{\partial w_{ji}} \left[\frac{1}{2} \sum_{d \in D} \sum_{k \in \text{outputs}} (t_{kd} - o_{kd})^2 + \gamma \sum_{i,j} w_{ji}^2 \right]$$

$$\begin{aligned} \frac{\partial E_d}{\partial w_{ji}} &= \frac{\partial}{\partial w_{ji}} \left[\frac{1}{2} \sum_{d \in D} \sum_{k \in \text{outputs}} (t_{kd} - o_{kd})^2 \right] + \frac{\partial}{\partial w_{ji}} \left[\gamma \sum_{i,j} w_{ji}^2 \right] \\ &= \frac{\partial}{\partial w_{ji}} \left[\frac{1}{2} \sum_{d \in D} \sum_{k \in \text{outputs}} (t_{kd} - o_{kd})^2 \right] \cdot \frac{\partial \text{net}_j}{\partial w_{ji}} + \gamma \sum_{i,j} \frac{\partial}{\partial w_{ji}} (w_{ji}^2) \end{aligned}$$

$$\begin{aligned} \frac{\partial E_d}{\partial w_{ji}} &= \frac{\partial}{\partial w_{ji}} \left[\frac{1}{2} \sum_{d \in D} \sum_{k \in \text{outputs}} (t_{kd} - o_{kd})^2 \right] \frac{\partial o_j}{\partial \text{net}_j} X_{ji} + 2\gamma \sum_{i,j} w_{ji} \\ \frac{\partial E_d}{\partial w_{ji}} &= \frac{\partial}{\partial w_{ji}} \left[\frac{1}{2} \sum_{d \in D} \sum_{k \in \text{outputs}} (t_{kd} - o_{kd})^2 \right] \frac{\partial o_j}{\partial \text{net}_j} X_{ji} + \dots \\ &= \underbrace{\left[\sum_{d \in D} (t_{kd} - o_{kd})^2 \right]}_{\text{dotted}} \frac{\partial o_j}{\partial \text{net}_j} X_{ji} + \dots \\ &= \underbrace{\left[\sum_{d \in D} (t_{kd} - o_{kd}) \right]}_{\text{dotted}} (-1) o_j (1 - o_j) X_{ji} + 2\gamma \sum_{i,j} w_{ji} \end{aligned}$$

$$= \underbrace{(-X_{ji} o_j + X_{ji} o_j^2)}_{\text{dotted}} \sum_{d \in D} (t_{kd} - o_{kd}) + 2\gamma \sum_{i,j} w_{ji}$$