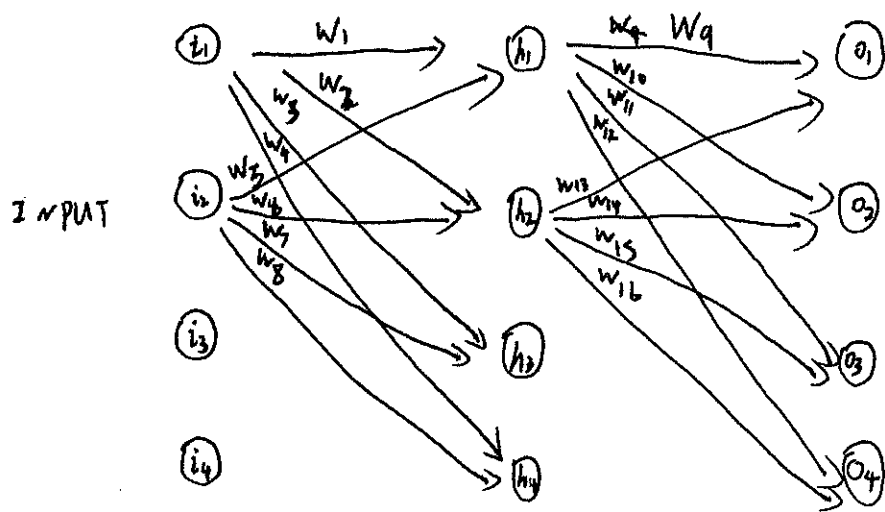


BP In Traditional Networks



INPUT: $i_1=0.05, i_2=0.1, i_3, i_4$

LABEL OUTPUT: $0, 0.01, 0.99, 0.3, 0.4$

Initial weight:

$w_1=0.01, w_2=0.4, w_3=0.5, w_4=0.6,$

$w_5=0.08, w_6=0.04, w_7=0.6, w_8=0.8, \dots$

Activation Function = sigmoid = $\frac{1}{1 + e^{-x}}$
 η : learning rate

STEP 1: INPUT \rightarrow Hidden \rightarrow OUTPUT (Forward Propagation)

$$h_1: w_1 \times i_1 + w_5 \times i_2 + w_{i_3 \rightarrow h_1} \times i_3 + w_{i_4 \rightarrow h_1} \times i_4 = net_{h_1}$$

$$out_{h_1} = \frac{1}{1 + e^{-net_{h_1}}}$$

$$h_2: w_2 \times i_1 + w_6 \times i_2 + w_{i_3 \rightarrow h_2} \times i_3 + w_{i_4 \rightarrow h_2} \times i_4 = net_{h_2}$$

$$out_{h_2} = \frac{1}{1 + e^{-net_{h_2}}}$$

$$o_1: out_{h_1} = i_{h_1}, out_{h_2} = i_{h_2} \dots$$

$$i_{h_1} \times w_9 + i_{h_2} \times w_{13} + i_{h_3} \times w_{h_3 \rightarrow o_1} + i_{h_4} \times w_{h_4 \rightarrow o_1} = net_{o_1}$$

$$out_{o_1} = \frac{1}{1 + e^{-net_{o_1}}}$$

$$o_2: i_{h_1} \times w_{10} + i_{h_2} \times w_{14} + i_{h_3} \times w_{h_3 \rightarrow o_2} + i_{h_4} \times w_{h_4 \rightarrow o_2} = net_{o_2}$$

$$out_{o_2} = \frac{1}{1 + e^{-net_{o_2}}}$$

STEP 1 Forward Propagation Done. LABEL ($o_1, o_2, o_3, o_4 \dots$)
 OUTPUT ($out_{o_1}, out_{o_2}, out_{o_3}, out_{o_4} \dots$)

STEP 2 Back Propagation

① Square Error. $E_{\text{total}} = \sum (E_{o1} + E_{o2} \dots) = \sum \frac{1}{2} (\text{target} - \text{output})^2$

$$E_{o1} = \frac{1}{2} (O_1 - \text{out}_{o1})^2 = \frac{1}{2} (0.01 - \text{out}_{o1})^2 = e_1$$

$$E_{o2} = \frac{1}{2} (O_2 - \text{out}_{o2})^2 = \frac{1}{2} (0.99 - \text{out}_{o2})^2 = e_2$$

$$E_{\text{total}} = e_1 + e_2 + \dots = e_{\text{total}}$$

② UPDATE Hid \rightarrow OUTPUT weight

~~$\frac{\partial E_{\text{total}}}{\partial w_q}$~~ $\frac{\partial E_{\text{total}}}{\partial w_q} = \frac{\partial E_{\text{total}}}{\partial \text{out}_{o1}} \times \frac{\partial \text{out}_{o1}}{\partial \text{net}_{o1}} \times \frac{\partial \text{net}_{o1}}{\partial w_q}$

$$= \left(-2 \times \frac{1}{2} (\text{target}_{o1} - \text{out}_{o1}) \right) \times (\text{out}_{o1} (1 - \text{out}_{o1})) \times (\text{out}_{h1} \times w_q^0) = E'_{w_q}$$

$$E'_{w_q} = -(\text{target}_{o1} - \text{out}_{o1}) \times \text{out}_{o1} (1 - \text{out}_{o1}) = \delta_{o1} \times \text{out}_{h1}$$

UPDATE: $w_{q\text{new}} = w_q - \eta \times \frac{\partial E_{\text{total}}}{\partial w_q} = w_q - \eta \delta_{o1} \times \text{out}_{h1}$

③ UPDATE Hid \rightarrow INPUT weight

~~$\frac{\partial E_{\text{total}}}{\partial w_1}$~~
$$\frac{\partial E_{\text{total}}}{\partial w_1} = \frac{\partial E_{\text{total}}}{\partial \text{out}_{h1}} \times \frac{\partial \text{out}_{h1}}{\partial \text{net}_{h1}} \times \frac{\partial \text{net}_{h1}}{\partial w_1}$$

$$\frac{\partial E_{o1}}{\partial \text{out}_{h1}} = \frac{\partial E_{o1}}{\partial \text{net}_{o1}} \times \frac{\partial \text{net}_{o1}}{\partial \text{out}_{h1}}$$

$$= E_{o1} E'_{h1}$$

$$= \left(\frac{\partial E_{o1}}{\partial \text{out}_{h1}} + \frac{\partial E_{o2}}{\partial \text{out}_{h1}} \right) \times \frac{\partial \text{out}_{h1}}{\partial \text{net}_{h1}} \times \frac{\partial \text{net}_{h1}}{\partial w_1}$$

$$= (E'_{h1} + E'_{h2}) \times \text{out}_{h1} (1 - \text{out}_{h1}) \times i_1 = E'_1$$

UPDATE: $w_{1\text{new}} = w_1 - \eta \frac{\partial E_{\text{total}}}{\partial w_1} = w_1 - \eta E'_1$

Back Propagation Done

KEEP Iteration