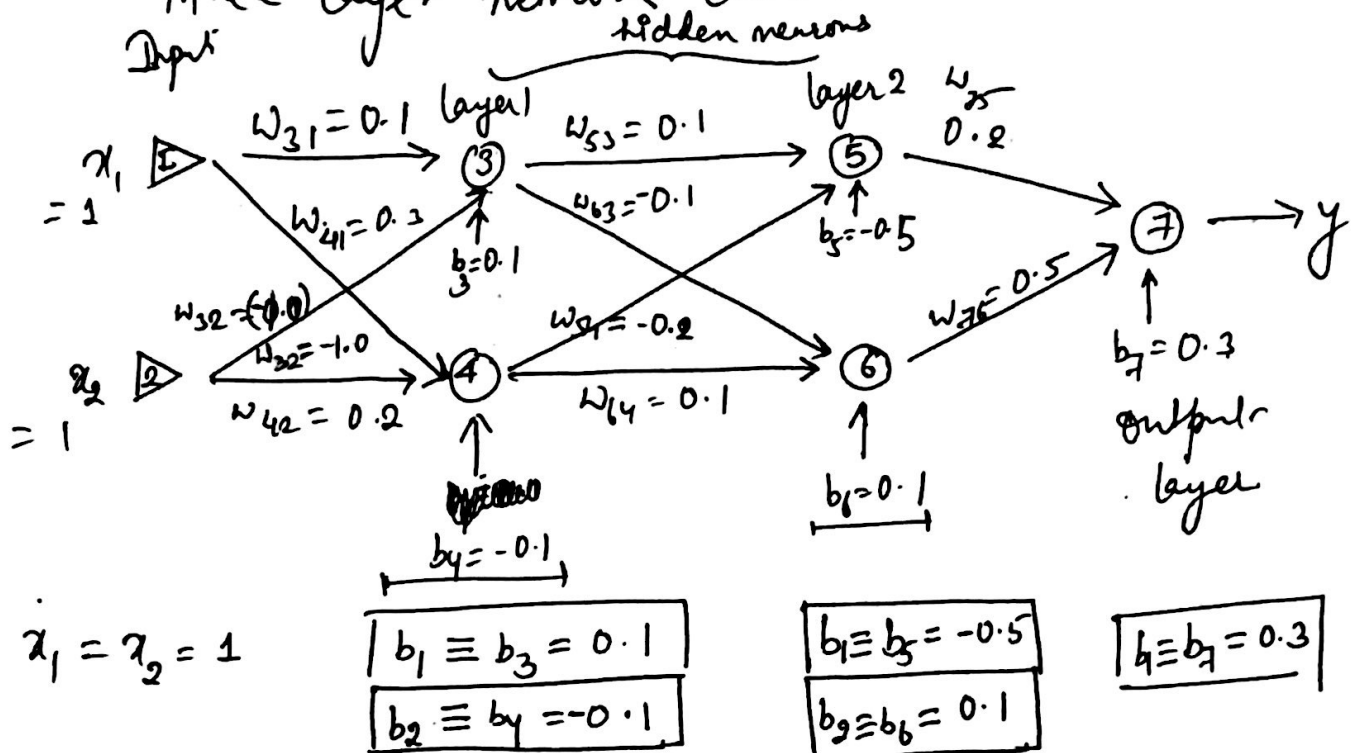


SOFT COMPUTING - 2

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Ans 1

Three layer network can be shown as :-



b_i represent bias value of each neuron.

w_{ji} synaptic weight of neuron j that is feed from neuron i

y_d = output of the network = 1

Induced local field of neuron j in layer l

$$V_j^{(l)}(n) = \sum_i w_{ji}^{(l)}(n) \cdot y_i^{(l-1)}(n)$$

n = number of iteration = 1 (given in Question)

layer 1

$$V_3' = b_3 + w_{31}x_1 + w_{32}x_2 = 0.1 + 0.1 \cdot 1 + (-1.0) \cdot 1 = -0.8 \approx 0.8$$

$$V_4' = b_4 + w_{41}x_1 + w_{42}x_2 = -0.1 + (0.3) \cdot 1 + (0.2) \cdot 1 = 0.4$$

Output of neuron j in layer 1 is given by

$$y_j^{(1)} = \phi_j(v_j^{(1)})$$

where $\phi_j(v)$ is the activation function.

for layer 1 $\phi(v) = \frac{1}{1 + e^{-v}}$

Output of neuron 3 is

$$y_3 = \phi_3(v_3) = \frac{1}{1 + e^{-v_3}} = \frac{1}{1 + e^{-0.8}} = 0.310949$$

$$\approx 0.31095$$

$$\approx 0.31$$

Output of neuron 4 is

$$y_4 = \phi_4(v_4) = \frac{1}{1 + e^{-v_4}} = \frac{1}{1 + e^{-0.4}} = 0.59869$$

$$\approx 0.59$$

Layer 2 (Hidden layer)

y_3 and y_4 are inputs to neuron 5 and 6 respectively

Induced local field of neuron 5

$$v_5 = b_5 + w_{53} \cdot y_3 + w_{54} \cdot y_4$$

$$= -0.5 + (0.1 * 0.31095) + (-0.2 * 0.59869)$$

$$= -0.5 + 0.031095 + (-0.119738)$$

$$v_5 = -0.58864 \approx -0.58864 \approx -0.59$$

Induced local field of neuron 6

$$v_6 = b_6 + w_{63} \cdot y_3 + w_{64} \cdot y_4$$

$$= 0.1 + (-0.1 * 0.31095) + (0.1 * 0.59869)$$

$$v_6 = 0.1 + -0.031095 + 0.059869 = 0.128869$$

$$= 0.1 - 0.031 + 0.059869 = 0.128869 \approx 0.13$$

Backward Computation

Output layer

Activation ~~func~~ of output layer is $\phi(v) = v$
 partial derivative

$$\phi'(v) = \frac{\partial}{\partial v} (v) = 1$$

Error

$$e_j(v) = \underbrace{d_j(v)}_{j^{\text{th}} \text{ Neuron}} - \underbrace{y_j(v)}_{\text{desired output}}$$

$$e_7(v_7) = 1 - y_7(v_7) = 1 - 0.26148$$

$$e_7(v_7) = 0.73852$$

Local gradient of neuron 7 is δ_7

$$\delta_7 = e_j \phi_j'(v_j)$$

$$\delta_7 = e_7 * \phi_7'(v_7) = 0.73852 * 1 = 0.73852$$

change in synaptic weight $\Delta w_{ji} = \eta \delta_j y_i$

$$\Delta w_{76} = 0.01 * 0.73852 * 0.1305 = 0.000963$$

$$\Delta w_{75} = 0.01 * 0.73852 * (-0.5189) = -0.00383$$

updated weights $w_{ji}(n+1) = w_{ji}(n) + \Delta w_{ji}$

$$w_{76}(n+1) = w_{76}(n) + \Delta w_{76} = 0.5 + 0.000963 = 0.500963$$

$$w_{75}(n+1) = w_{75}(n) + \Delta w_{75} = 0.2 - 0.00383 = 0.19617$$

Output of layer 2

$$y_5 = \phi_5(v_5) = \frac{1 - e^{-2v_5}}{1 + e^{-2v_5}} = \frac{1 - e^{-2 \cdot (-0.59)}}{1 + e^{-2 \cdot (-0.59)}}$$

$$y_5 = \frac{1.99600}{2.99600} = 0.5189$$

$$y_6 = \phi_6(v_6) = \frac{1 - e^{-2 \cdot (0.8067)}}{1 + e^{-2 \cdot (0.8067)}} = \frac{1 - 0.8067}{1 + 0.8067} = \frac{0.1933}{1.8067}$$

$$y_6 = 0.1305$$

layer 3 (Output-layer)

Induced local field of neuron 7 :

$$\begin{aligned} v_7 &= b_7 + w_{75} y_5 + w_{76} y_6 \\ &= 0.3 + (0.2 \cdot (-0.5189)) + 0.5 \cdot (0.1305) \\ &= 0.35598 - 0.0999 \\ v_7 &= 0.25598 \approx 0.26148 \end{aligned}$$

Activation function for output-layer is $\phi(v) = v$

$$\therefore \phi_7(v_7) = v_7 = 0.26148$$

Hidden layer

$$\phi(v) = \frac{1 - e^{-2v}}{1 + e^{-2v}}$$

$$\begin{aligned} \phi'(v) &= \frac{(1 + e^{-2v})(2e^{-2v}) - (1 - e^{-2v})(-2e^{-2v})}{(1 + e^{-2v})^2} \\ &= \frac{2e^{-2v} + 2e^{-4v} + 2e^{-2v} - 2e^{-4v}}{(1 + e^{-2v})^2} \end{aligned}$$

$$\phi'(v) = \frac{4e^{-2v}}{(1 + e^{-2v})^2}$$

Neuron 5 :- $\phi'_5(v_5) = \frac{4e^{-2v_5}}{(1 + e^{-2v_5})^2} = \frac{4 * e^{-2(-0.59)}}{[1 + e^{-2(-0.59)}]^2}$

$$= \frac{4 * (2.99600)}{(1 + 2.99600)^2} = \frac{12.43822}{15.98896} = \frac{13.017}{18.0997}$$

$$\phi'_5(v_5) = 0.7192$$

Neuron 6 : $\phi'_6(v_6) = \frac{4e^{-2v_6}}{(1 + e^{-2v_6})^2} = \frac{4e^{-2(0.128869)}}{(1 + e^{-2(0.128869)})^2}$

$$\phi'_6(v_6) = \frac{4 * 0.80675}{(1 + 0.80675)^2} = \frac{3.0912}{3.143} = \frac{3.0912}{3.143} = 0.98358$$

Local gradient of Neuron 5 :

$$\begin{aligned} \delta_5 &= \phi'_5(v_5) \sum_k \delta_k^{(3)} w_{k5} = \phi'_5(v_5) \delta_7 w_{75} \\ &= 0.7192 * 0.73852 * 0.2 \\ \delta_5 &= 0.10623 \end{aligned}$$

Neuron 6:
$$\delta_6 = \phi'_6(v_6) \sum_k \delta_k^{(3)} w_{k6}^{(3)} = \phi'_6(v_6) \delta_7 w_{76}$$

$$= 0.98358 * 0.73852 * 0.5$$

$$\delta_6 = \text{[redacted]} = 0.3632$$

We calculated

$$\delta_j^{(l)}(n) = \phi'_j(v_j^{(l)}(n)) \sum_k \delta_k^{(l+1)}(n) w_{kj}^{(l+1)}(n)$$

For neuron j , hidden layer l , synaptic weight is

$$w_{ji}^{(l)}(n+1) = w_{ji}^{(l)}(n) + \eta \delta_j^{(l)}(n) y_i^{(l-1)}(n) + \alpha \Delta w_{ji}^{(l)}(n-1)$$

α : momentum constant

When $\alpha = 0$,

$$w_{ji}^{(l)}(n+1) = w_{ji}^{(l)}(n) + \eta \delta_j^{(l)}(n) y_i^{(l-1)}(n)$$

Neuron 5: Synaptic weights from neurons 3 and 4 are:

$$w_{53}(n+1) = w_{53}(n) + \eta \delta_5(n) y_3(n)$$

$$= 0.1 + (0.01 * 0.10623 * 0.31)$$

$$w_{53}(n+1) = 0.10033 \approx 0.10033$$

$$w_{54}(n+1) = w_{54}(n) + \eta \delta_5(n) y_4(n)$$

$$= -0.2 + 0.01 * 0.10623 * 0.59869$$

$$w_{54}(n+1) = -0.19936$$

Neuron 6: Synaptic weights from neuron 3 and 4 are:

$$\begin{aligned} w_{63}(n+1) &= w_{63}(n) + \eta \delta_6(n) y_3(n) \\ &= -0.1 + 0.01 * 0.3632 * 0.31 \\ &= -0.09887 \approx -0.09887 \end{aligned}$$

$$\begin{aligned} w_{64}(n+1) &= w_{64}(n) + \eta \delta_6(n) y_4(n) \\ &= 0.1 + (0.01 * 0.3632 * 0.59869) \\ &= 0.10217 \end{aligned}$$

Hidden layer 1 (Neurons 3 and 4)

$$\phi(v) = \frac{1}{1+e^{-v}}$$

$$\begin{aligned} \phi'(v) &= \frac{(1+e^{-v})(0) - 1(1-e^{-v})}{(1+e^{-v})^2} \\ &= \frac{e^{-v}}{(1+e^{-v})^2} \end{aligned}$$

$$\begin{aligned} \phi'_3(v_3) &= \frac{e^{-v_3}}{(1+e^{-v_3})^2} = \frac{e^{+0.8}}{(1+e^{+0.8})^2} = \frac{2.2255}{(1+2.2255)^2} \\ \phi'_3(v_3) &= 0.2139 \end{aligned}$$

$$\phi'_4(v_4) = \frac{e^{-v_4}}{(1+e^{-v_4})^2} = \frac{e^{-0.4}}{(1+e^{-0.4})^2} = 0.24026$$

Local Gradient:-

$$\begin{aligned}\text{Neuron 3: } \delta_3 &= \phi'_3(v_3) \sum_k \delta_k w_{k3} \\ &= \phi'_3(v_3) [\delta_5 w_{53} + \delta_6 w_{63}] \\ &= 0.2139 (0.10623 * 0.1 + 0.3632 * 0.1) \\ &= -0.00549\end{aligned}$$

$$\begin{aligned}\text{Neuron 4: } \delta_4 &= \phi'_4(v_4) \sum_k \delta_k w_{k4} \\ &= \phi'_4(v_4) (\delta_5 w_{54} + \delta_6 w_{64}) \\ &= 0.24026 (0.10623 * 0.2 + 0.3632 * 0.1) \\ &= 0.24026 (-0.02207 + 0.037034) \\ \delta_4 &= 0.00362\end{aligned}$$

Synaptic weights update from Neuron 1 and 2:

Neuron 3 & Neuron 4

$$\begin{aligned}w_{31}(n+1) &= w_{31}(n) + \eta \delta_3(n) y_1(n) \\ &= 0.1 + 0.01 * (-0.00549) * 1 = 0.0999451\end{aligned}$$

$$\begin{aligned}w_{32}(n+1) &= w_{32}(n) + \eta \delta_3(n) y_2(n) \\ &= -0.1 + 0.01 * (-0.00549) * 1 = -0.1000549\end{aligned}$$

$$\begin{aligned}w_{41}(n+1) &= w_{41}(n) + \eta \delta_4(n) y_1(n) \\ &= 0.3 + 0.01 * 0.00362 * 1 = 0.3000362\end{aligned}$$

$$\begin{aligned}w_{42}(n+1) &= w_{42}(n) + \eta \delta_4(n) y_2(n) \\ &= 0.2 + 0.01 * 0.00362 * 1 = 0.2000362\end{aligned}$$

Bias Update

Output-layer or Neuron 7

$$b_7(n+1) = b_7(n) + \eta \delta_7 = 0.3 + 0.01 * 0.73852 \\ = 0.3073852$$

Hidden layer 2

$$\text{Neuron 5: } b_5(n+1) = b_5(n) + \eta \delta_5 = -0.5 + 0.01 * 0.10623 \\ = -0.4989377$$

$$\text{Neuron 6: } b_6(n+1) = b_6(n) + \eta \delta_6 = 0.1 + 0.01 * 0.3632 \\ = 0.103632$$

Hidden layer 1

$$\text{Neuron 3: } b_3(n+1) = b_3(n) + \eta \delta_3 = 0.1 + 0.01 * 0.00549 \\ = 0.0999451$$

$$\text{Neuron 4: } b_4(n+1) = b_4(n) + \eta \delta_4 \\ = -0.1 + 0.01 * 0.00362 = -0.099964$$