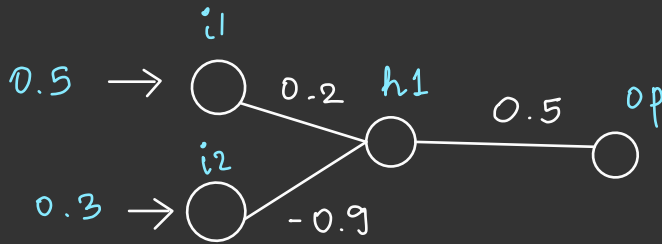


1. Neural Networks

i) An observation with two variables (0.5, 0.3) is input into the given neural network what is the predicted output using sigmoid activation function?



$$i_1 = 0.5 \quad w_1 = 0.2$$

$$i_2 = 0.3 \quad w_2 = -0.9$$

$$w_3 = 0.5$$

$$h_1 = (i_1 \times w_1 + i_2 \times w_2)$$

$$= 0.5 \times 0.2 + 0.3 \times -0.9$$

$$= -0.17$$

$$h_{1f} = \frac{1}{1 + \frac{1}{e^h}} \quad (\text{applying sigmoid fn})$$

$$= \frac{1}{1 + \frac{1}{e^{-0.17}}}$$

$$= 0.457$$

$$op = h_{1f} \times \omega_3$$

$$= 0.457 \times 0.5$$

$$= 0.2285$$

$$op_f = \frac{1}{1 + \frac{1}{e^{op}}}$$

$$= \frac{1}{1 + \frac{1}{e^{0.2285}}}$$

$$= \underline{\underline{0.557}} \quad \text{is the output}$$

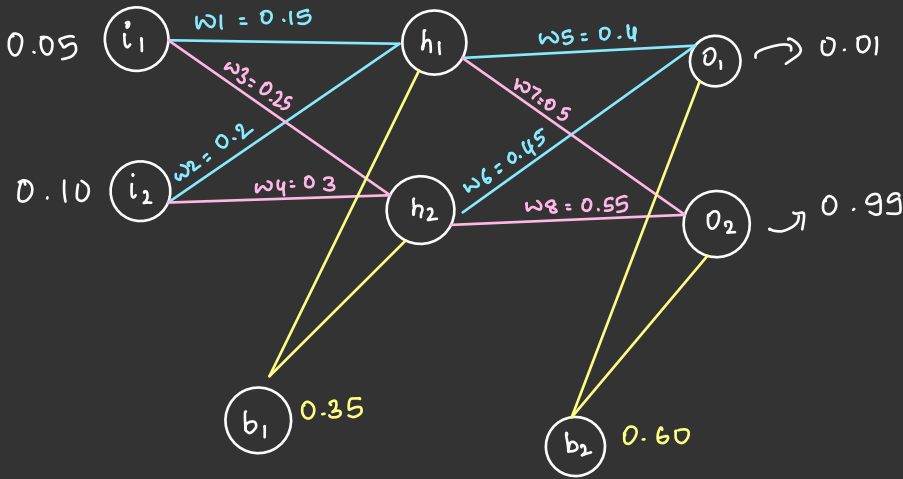
2) Given inputs for neural network are

$$i_1 = 0.05 \quad i_2 = 0.10$$

Required output : 0.01 and 0.99

\downarrow
 t_1

\downarrow
 t_2



Step 1: Forward pass

Calculate net input at each hidden layer node

$$h_1 = (w_1 \times i_1) + (w_2 \times i_2) + b_1$$

$$= (0.15 \times 0.05) + (0.2 \times 0.1) + 0.35$$

$$= \underline{\underline{0.3775}}$$

Apply sigmoid activation function

$$h_{1f} = \frac{1}{1 + \frac{1}{e^h}} = \frac{1}{1 + \frac{1}{e^{0.3775}}}$$

$$= \underline{\underline{0.593}}$$

III^{ly} for h_2 :

$$h_2 = (w_3 \times i_1) + (w_4 \times i_2) + b_1$$

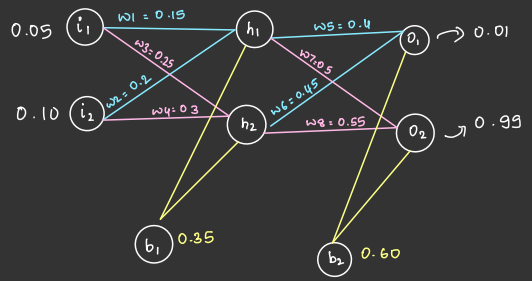
$$= (0.25 \times 0.05) + (0.3 \times 0.1) + 0.35$$

$$= 0.3925$$

$$h_{2f} = \frac{1}{1 + \frac{1}{e^{h_2}}} = \frac{1}{1 + \frac{1}{e^{0.3925}}}$$

$$\underline{\underline{h_{2f} = 0.597}}$$

III^{ly} for O_1



$$O_1 = (w_5 \times h_1) + (w_6 \times h_2) + b_2$$

$$= (0.4 \times 0.593) + (0.45 \times 0.597) + 0.6$$

$$= \underline{\underline{1.106}}$$

$$O_{1f} = \frac{1}{1 + \frac{1}{e^{O_1}}} = \frac{1}{1 + \frac{1}{e^{1.106}}}$$

$$\underline{\underline{O_{1f} = 0.751}}$$

$$O_2 = (w_7 \times h_1) + (w_8 \times h_2) + b_2$$

$$= (0.5 \times 0.593) + (0.55 \times 0.597) + 0.6$$

$$= 1.225$$

$$O_{2f} = \frac{1}{1 + \frac{1}{e^{O_2}}} = \underline{\underline{0.773}}$$

Now calculate total error

$$E_{\text{total}} = \sum \frac{1}{2} (\text{target} - \text{output})^2$$

Given target for $O_1 = 0.01 = t_1$

- " - $O_2 = 0.99 = t_2$

obtained $O_{1f} = 0.751$

$O_{2f} = 0.773$

$$E_{\text{total}} = \frac{1}{2} (0.01 - 0.751)^2 + \frac{1}{2} (0.99 - 0.773)^2$$

$$= 0.298$$

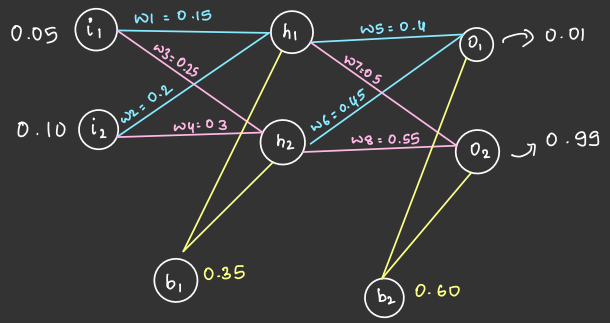
Step 2 Backward pass

- to update weights in order to
reduce error and obtain required output

$$i) \quad \frac{\partial E_{total}}{\partial w_5}$$



use chain rule



$$\frac{\partial E_{total}}{\partial w_5} = \frac{\partial E_{total}}{\partial o_{1f}} \times \frac{\partial o_{1f}}{\partial o_1} \times \frac{\partial o_1}{\partial w_5}$$

$$\text{WRT } E_{total} = \frac{1}{2} \left[(t_1 - o_{1f})^2 + (t_2 - o_{2f})^2 \right]$$

$$\therefore \frac{\partial E_{total}}{\partial o_{1f}} = \frac{1}{2} \cdot 2 \left((t_1 - o_{1f}) \cdot (-1) \right)$$

$$= - (t_1 - o_{1f})$$

$$= - (0.01 - 0.751)$$

$$= 0.741$$

————— (1)

$$\frac{\partial \sigma_{1f}}{\partial \sigma_1} \rightarrow ?$$

$$\text{wkt } \sigma_{1f} = \frac{1}{1 + e^{-\sigma_1}} = (1 + e^{-\sigma_1})^{-1}$$

Chain rule:

$$\begin{aligned} \frac{\partial \sigma_{1f}}{\partial \sigma_1} &= (-1)(1 + e^{-\sigma_1})^{-2} \cdot e^{-\sigma_1} \cdot (-1) \\ &= \frac{e^{-\sigma_1}}{(1 + e^{-\sigma_1})^2} \end{aligned}$$

Subs

$$\frac{\partial \sigma_{1f}}{\partial \sigma_1} = \frac{e^{-1.106}}{(1 + e^{-1.106})^2} = \underline{\underline{0.186}} \quad - (2)$$

$$\frac{\partial O_1}{\partial w_5} \rightarrow ?$$

$$O_1 = (w_5 \times h_{1f}) + (w_6 \times h_{2f}) + (b_2)$$

$$\frac{\partial O_1}{\partial w_5} = h_{1f} = \underline{\underline{0.593}} \quad \text{--- (3)}$$

$$\therefore \frac{\partial E_{\text{total}}}{\partial w_5} = \textcircled{1} \times \textcircled{2} \times \textcircled{3}$$

$$= 0.741 \times 0.186 \times 0.593$$

$$= \underline{\underline{0.082}} \quad \text{--- (4)}$$

Now to decrease error, subtract (4)

from w_5 . Repeat same for all weights

