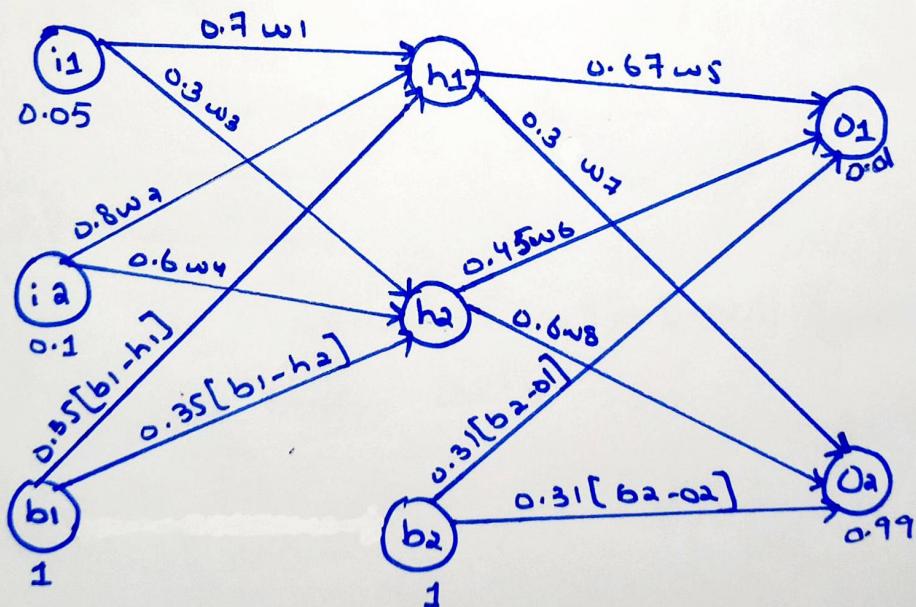


## Question NO # 1

Please use the following network to update all the weights except the bias weights using backward propagation for MLP. Input is given as ( $i_1 = 0.05$ ) and ( $i_2 = 0.1$ ) and target values for ( $o_1 = 0.01$  and ( $o_2 = 0.99$ ) and learning rate = 1. Execute for only one iteration. All neurons have sigmoid activation function.



## Forward propagation:-

- Weighted sum for the hidden layer.  
Find  $H_1, H_2$ .

$$\begin{aligned}H_1 &= (w_1 \times i_1) + (w_2 \times i_2) + (b_1 \times 0.35) \\&= (0.7 \times 0.05) + (0.8 \times 0.1) + (1 \times 0.35) \\&= 0.035 + 0.08 + 0.35.\end{aligned}$$

$$H_1 = 0.465$$

$$\begin{aligned}H_2 &= (w_3 \times i_1) + (w_4 \times i_2) + (b_2 \times 0.35) \\&= (0.3 \times 0.05) + (0.6 \times 0.1) + (1 \times 0.35) \\&= 0.015 + 0.06 + 0.35\end{aligned}$$

$$H_2 = 0.425$$

- Apply the sigmoid activation function to the hidden layer units.

$$h_1 = \frac{1}{1 + e^{-0.465}}$$

$$h_1 = \frac{1}{1 + 0.628}$$

$$h_2 = \frac{1}{1 + e^{-0.425}}$$

$$h_2 = \frac{1}{1 + 0.653}$$

$$h_2 = 0.6046$$

→ Weighted sum for the output layer units:  
Find  $o_1, o_2$ .

$$\begin{aligned} o_1 &= (w_5 \times h_1) + (w_6 \times h_2) + (b_2 \times 0.31) \\ &= (0.67 \times 0.614) + (0.45 \times 0.604) + (1 \times 0.31) \\ &= 0.411 + 0.2718 + 0.31 \end{aligned}$$

$$o_1 = 0.9928$$

$$\begin{aligned} o_2 &= (w_7 \times h_1) + (w_8 \times h_2) + (b_2 \times 0.31) \\ &= (0.3 \times 0.614) + (0.6 \times 0.604) + (1 \times 0.31) \\ &= 0.1842 + 0.3624 + 0.31 \end{aligned}$$

$$o_2 = 0.8584$$

→ Apply the sigmoid activation function to the output layer units.

$$o_1 = \frac{1}{1 + e^{-0.9928}}$$

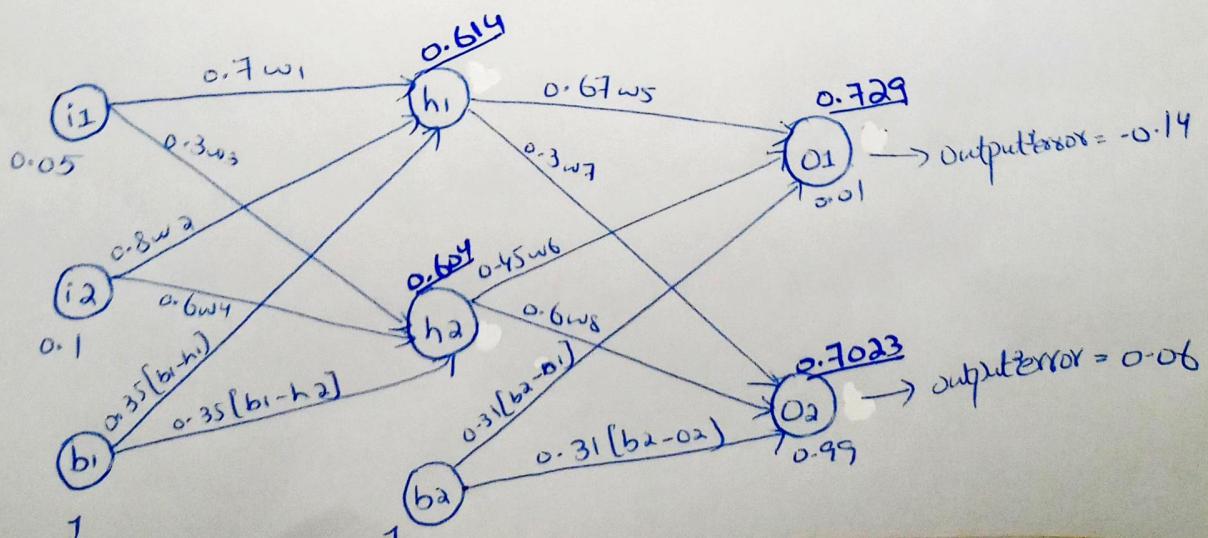
$$o_1 = \frac{1}{1 + 0.3705}$$

$$o_1 = 0.729$$

$$o_2 = \frac{1}{1 + e^{-0.8584}}$$

$$o_2 = \frac{1}{1 + 0.4238}$$

$$o_2 = 0.7023$$



## Back propagation

→ Calculate error:-

• For output units:-

$$\text{Target } (o_1 = 0.01)$$

$$\text{output}_1 = o_1(1-o_1)(\text{Target} - o_1)$$

$$= 0.729(1-0.729)(0.01-0.729)$$

$$= 0.729(0.271)(-0.719)$$

$$\boxed{\text{output}_1 = -0.14}$$

$$\text{Target } (o_2 = 0.99)$$

$$\text{output}_2 = o_2(1-o_2)(\text{Target} - o_2)$$

$$= 0.7023(1-0.7023)(0.99-0.7023)$$

$$= 0.7023(0.2977)(0.2877)$$

$$\boxed{\text{output}_2 = 0.06}$$

• Total error:-

$$\text{Total error} = 0.5 \times ((\text{Target } o_1 - \text{sigmoid}(o_1))^2 + (\text{Target } o_2 - \text{sigmoid}(o_2))^2)$$

$$= 0.5 \times ((0.01 - 0.729)^2 + (0.99 - 0.7023)^2)$$

$$= 0.5 \left( (-0.719)^2 + (0.287)^2 \right)$$

$$= 0.5 (0.5169 + 0.0823)$$

$$= 0.5 (0.5992)$$

$$\boxed{\text{Total error} = 0.2996}$$

## Backward Propagation:-

→ Updating the weights between the hidden layer units.

$$\Delta w_{ij} = \eta s_j o_i$$

$\eta$  is a learning rate which is equal to 1

$$\Delta w_5 = \eta \overset{\text{error}}{o_1 h_2}$$

$$= 1(-0.14)(0.614)$$

$$\Delta w_5 = -0.085$$

$$w_5(\text{new}) = \Delta w_5 + w_5(\text{old})$$

$$w_5(\text{new}) = -0.085 + 0.67$$

$$\boxed{w_5(\text{new}) = 0.585}$$

$$\Delta w_7 = \eta_{02}^{\text{new}} h_1$$

$$= 1 \times 0.06 \times 0.614$$

$$\Delta w_7 = 0.036$$

$$w_7(\text{new}) = \Delta w_7 + w_7(\text{old})$$

$$= 0.036 + 0.3$$

$$w_7(\text{new}) = 0.336$$

$$\Delta w_6 = \eta_{01}^{\text{new}} h_2$$

$$= 1 \times -0.14 \times 0.604$$

$$\Delta w_6 = -0.084$$

$$w_6(\text{new}) = \Delta w_6 + w_6(\text{old})$$

$$= -0.084 + 0.45$$

$$w_6(\text{new}) = 0.366$$

$$\Delta w_8 = \eta_{02}^{\text{new}} h_2$$

$$= 1 \times 0.06 \times 0.604$$

$$\Delta w_8 = 0.036$$

$$w_8(\text{new}) = \Delta w_8 + w_8(\text{old})$$

$$= 0.036 + 0.6$$

$$w_8(\text{new}) = 0.636$$

$$\Delta\omega_1 = \eta h_1 i_1$$

$$= 1 \times 0.614 \times 0.05$$

$$\Delta\omega_1 = 0.0307$$

$$\omega_1(\text{new}) = \Delta\omega_1 + \omega_1(\text{old})$$

$$= 0.0307 + 0.7$$

$$\boxed{\omega_1(\text{new}) = 0.73}$$

$$\Delta\omega_3 = \eta h_3 i_1$$

$$= 1 \times 0.604 \times 0.05$$

$$\Delta\omega_3 = 0.0302$$

$$\omega_3(\text{new}) = \Delta\omega_3 + \omega_3(\text{old})$$

$$= 0.0302 + 0.3$$

$$\boxed{\omega_3(\text{new}) = 0.3302}$$

$$\Delta\omega_2 = \eta h_1 i_2$$

$$= 1 \times 0.614 \times 0.1$$

$$\Delta\omega_2 = 0.0614$$

$$\omega_2(\text{new}) = \Delta\omega_2 + \omega_2(\text{old})$$

$$= 0.0614 + 0.8$$

$$\Delta w_4 = \eta h_2 i_2$$

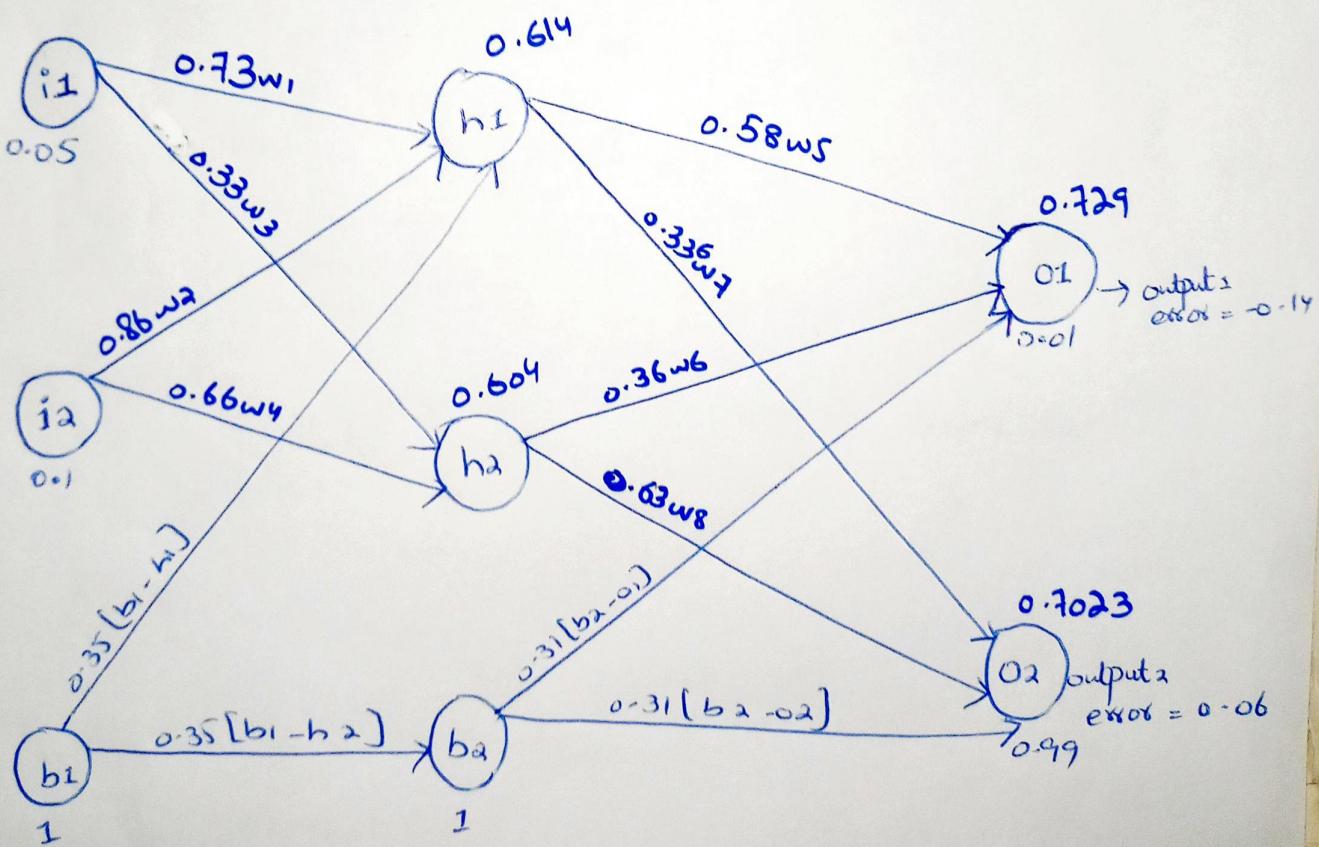
$$= 1 \times 0.604 \times 0.1$$

$$\Delta w_4 = 0.0604$$

$$w_4(\text{new}) = \Delta w_4 + w_4(\text{old})$$

$$= 0.0604 + 0.6$$

$$w_4(\text{new}) = 0.6604$$



## Forward Propagation With Updated Weights:-

→ Calculate the weighted sum for the hidden layer units:-

$$\begin{aligned} h_1 &= (\omega_1 \times i_1) + (\omega_2 \times i_2) + (b_1 \times 0.35) \\ &= (0.73 \times 0.05) + (0.86 \times 0.1) + (1 \times 0.35) \\ &= 0.0365 + 0.086 + 0.35 \end{aligned}$$

$$h_1 = 0.4725$$

$$\begin{aligned} h_2 &= (\omega_3 \times i_1) + (\omega_4 \times i_2) + (b_1 \times 0.35) \\ &= (0.33 \times 0.05) + (0.66 \times 0.1) + (1 \times 0.35) \\ &= 0.0165 + 0.066 + 0.35 \end{aligned}$$

$$h_2 = 0.4325$$

→ Apply the sigmoid activation function to the hidden layer units:-

$$h_1 = \frac{1}{1 + e^{-0.4725}}$$

$$h_1 = \frac{1}{1 + 0.623}$$

$$h_1 = 0.6159$$

$$h_2 = \frac{1}{1 + e^{-0.4325}}$$

$$h_2 = \frac{1}{1 + 0.64}$$

$$h_2 = 0.606$$

→ Weighted sum for the output layer units

$$o_1 = (w_5 * h_1^{\text{output}}) + (w_6 * h_2^{\text{output}}) + (b_2 * 0.31)$$

$$o_1 = (0.58 * 0.614) + (0.38 * 0.604) + (1 * 0.31)$$

$$o_1 = (0.356 + 0.229 + 0.31)$$

$$o_1 = 0.895$$

$$o_2 = (w_7 * h_1) + (w_8 * h_2) + (b_2 * 0.31)$$

$$= (0.336 * 0.614) + (0.63 * 0.604) + (1 * 0.31)$$

$$= 0.206 + 0.3805 + 0.31$$



$$o_2 = 0.896$$

→ Apply the sigmoid activation function to the output layer units:-

$$o_1 = \frac{1}{1 + e^{-0.895}}$$

$$o_1 = 0.709$$

$$o_2 = \frac{1}{1 + e^{-0.896}}$$

$$o_2 = 0.71$$

