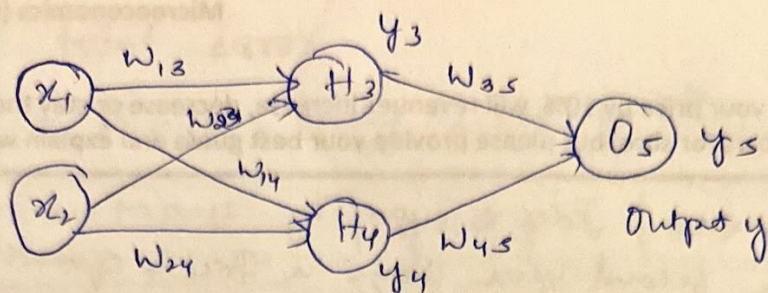


## Step 8:



Given inputs:

$$x_1 = 0.35$$

$$x_2 = 0.9$$

$$\eta = 1$$

output / Target values:

$$\eta = 0.5$$

Weights:

$$w_{13} = 0.1$$

$$w_{14} = 0.4$$

$$w_{23} = 0.8$$

$$w_{24} = 0.6$$

$$w_{35} = 0.3$$

$$w_{45} = 0.9$$

Activation function is logistic function

Given by sigmoid curve =  $\frac{1}{1+e^{-x}}$

Forward pass: compute output  $y_3$ ,  $y_4$  and  $y_5$

$$a_j = \sum_i (w_{ij} * x_i) \quad y_j = F(a_j) = \frac{1}{1+e^{-a_j}}$$

$$\begin{aligned} a_1 &= (w_{13} * x_1) + (w_{23} * x_2) \\ &= (0.1 * 0.35) + (0.8 * 0.9) = 0.755 \end{aligned}$$

$$y_3 = f(a_1) = \frac{1}{1+e^{-0.755}} = \underline{\underline{0.68}}$$

$$\begin{aligned} a_2 &= (w_{14} * x_1) + (w_{24} * x_2) \\ &= (0.4 * 0.35) + (0.6 * 0.9) = 0.68 \end{aligned}$$

$$y_4 = f(a_2) = \frac{1}{1+e^{-0.68}} = \underline{\underline{0.6637}}$$

$$\begin{aligned} a_3 &= (w_{35} * y_3) + (w_{45} * y_4) \\ &= (0.3 * 0.68) + (0.9 * 0.6637) = 0.801 \end{aligned}$$

$$y_5 = f(a_3) = \frac{1}{1+e^{-0.801}} = \underline{\underline{0.69}} \rightarrow \text{output of network}$$



Calculating total error

$$\text{Error} = \text{MSE} = \frac{1}{N} \sum (y - \hat{y})^2$$

$$\text{a. Error} = \text{MAE} = |y - \hat{y}| = |0.5 - 0.69| = \underline{\underline{0.19}}$$

$$\boxed{\text{Error} = 0.19}$$

Backward pass :

$$\Delta w_{ji} = \eta \delta_j o_i \rightarrow \text{output at that node}$$

$\downarrow$                        $\downarrow$   
weight update      learning rate

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

If  $j$  is an output unit

$$\delta_j = o_j (1 - o_j) \sum_k \delta_k w_{kj}$$

If  $j$  is a hidden unit

Let's calculate partial derivatives for  $t_3$ ,  $t_4$  &  $o_5$   
i.e.  $\delta_3$ ,  $\delta_4$ ,  $\delta_5$

For output unit  $o_5$ :

$$\begin{aligned} \delta_5 &= y(1-y) (y_{\text{target}} - y) \\ &= 0.69 * (1 - 0.69) * (0.5 - 0.69) \end{aligned}$$

$$\boxed{\delta_5 = -0.0406}$$

For hidden unit:

$$\begin{aligned} \delta_3 &= y_3(1-y_3) w_{35} * \delta_5 \\ &= 0.68 * (1 - 0.68) * (0.3 * -0.0406) \end{aligned}$$

$$\boxed{\delta_3 = 0.00265}$$



$$\delta_4 = y_4(1 - y_4)w_{45} + \delta_5$$

$$= 0.6637 * (1 - 0.6637) * (0.9 + -0.0406)$$

$$\boxed{\delta_4 = -0.0082}$$

Since we have partial derivatives

let us compute new weights

$$\Delta w_{ji} = \eta \delta_j o_i$$

$$\Delta w_{45} = \eta \delta_5 y_4$$

$$= 1 * -0.0406 * 0.6637$$

$$\Delta w_{45} = -0.0269$$

$$w_{45}(\text{new}) = \Delta w_{45} + w_{45}(\text{old})$$

$$= -0.0269 + 0.9$$

$$\boxed{w_{45}(\text{new}) = 0.8731}$$

$$\Delta w_{14} = \eta \delta_4 x_1 = 1 * -0.0082 * 0.35$$

$$\Delta w_{14} = -0.00287$$

$$w_{14}(\text{new}) = \Delta w_{14} + w_{14}(\text{old})$$

$$= -0.00287 + 0.4$$

$$w_{14}(\text{new}) = -0.00287 + 0.4 = 0.3971$$

$$\boxed{w_{14}(\text{new}) = 0.3971}$$

$$\Delta w_{13} = \eta \delta_3 x_1 = 1 * (-0.00265) * 0.35$$

$$w_{13}(\text{new}) = \Delta w_{13} + w_{13}(\text{old}) = [1 * (-0.00265) * 0.35] + 0.1 = 0.0971$$

$$w_{23}(\text{new}) = \Delta w_{23} + w_{23}(\text{old}) = [1 * (-0.00265) * 0.9] + 0.8 = 0.7974$$



$$W_{24}(\text{new}) = \Delta W_{24} + W_{24}(\text{old}) = (1 + (-0.0082) * 0.4) + 0.6$$

$$\boxed{W_{24}(\text{new}) = 0.5926}$$

$$W_{35}(\text{new}) = \Delta W_{35} + W_{35}(\text{old}) = (1 + (-0.0406) * 0.68) + 0.3$$

$$\boxed{W_{35}(\text{new}) = 0.2724}$$

thus, weights are updated after backward pass  
and this loop goes on until a minima is reached