

The input pattern is ( 0.52, 0.75, 0.97 ), and the desired output pattern is ( 0.24, 0.17, 0.65). The initial weight matrices are as follows:

**M-1** Matrix of weights from input layer to hidden layer

$$\begin{array}{cc} 0.6 & - 0.4 \\ 0.2 & 0.8 \\ - 0.5 & 0.3 \end{array}$$

**M-2** Matrix of weights from hidden layer to output layer

$$\begin{array}{ccc} -0.90 & 0.43 & 0.25 \\ 0.11 & - 0.67 & - 0.75 \end{array}$$

The threshold values (or bias) for neurons in the hidden layer are 0.2 and 0.3, while those for the output neurons are 0.15, 0.25, and 0.05, respectively.

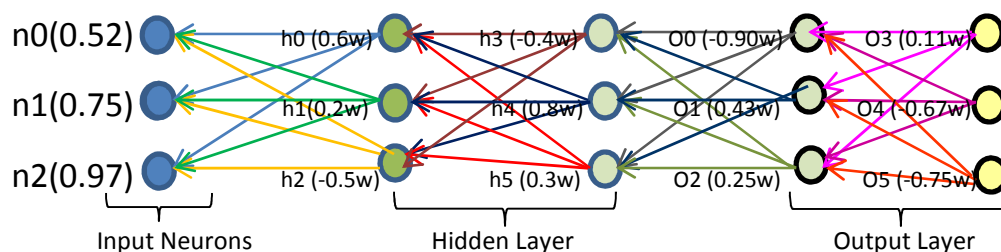
We calculate the output of each neuron in the hidden and output layers as follows. We add a bias or threshold value to the activation of a neuron (call this result  $x$ ) and use the **sigmoid** function below to get the output.

$$f(x) = 1 / (1 + e^{-x})$$

Learning parameters used are 0.2 for the connections between the hidden layer neurons and output neurons and 0.15 for the connections between the input neurons and the neurons in the hidden layer. These values as you recall are the same as in the previous illustration, to make it easy for you to follow the calculations by comparing them with similar calculations in the preceding sections.

Answer:

Schematic conceptual graph for neural net of the above mentioned case



Vector data for the Input Neurons  $\mathbf{N}^{\rightarrow} = (0.52, 0.75, 0.97)$

Vector data for the Desired Output pattern  $\mathbf{DO}^{\rightarrow} = (0.24, 0.17, 0.65)$

Weights Set for the Hidden Layer 00 Neurons  $\mathbf{WH00} = \{0.6, 0.2, -0.5\}$

Weights Set for the Hidden Layer 01 Neurons  $\mathbf{WH01} = \{-0.4, 0.8, 0.3\}$

Weights Set for the Output Layer 00 Neurons  $\mathbf{WO00} = \{-0.90, 0.43, 0.25\}$

Weights Set for the Output Layer 01 Neurons  $\mathbf{WO01} = \{0.11, -0.67, -0.75\}$

Bias Value for Neurons in the Hidden Layers is 0.2 and 0.3 respectively, while Threshold limit for Neurons in the Output Layer are 0.15, 0.25 and 0.05 for each neuron in the row respectively.

Learning rates ( $\mu$ ) used for connection between the input and hidden layer neurons is set to be 0.15, while connection for the layer between hidden and output neurons is set to be 0.2.

### Activating Neurons in the Hidden Layer 00 (NH00)

Neurons in the Hidden Layer 00 are  $h_0$ ,  $h_1$  and  $h_2$  respectively.

$$\begin{aligned} F(h_0) = F(n_0, n_1, n_2) &= \sum(W_i I_i) = \{ (h_0 * n_0) + (h_0 * n_1) + (h_0 * n_2) \} \\ &= \{ (0.6 * 0.52) + (0.6 * 0.75) + (0.6 * 0.97) \} \\ &= 0.312 + 0.45 + 0.582 \\ &= 1.344 \end{aligned}$$

$$h_0 \text{ Bias} : 0.2 + 1.344 = 1.544$$

$$FSigmoid(x) = \frac{1}{1+e^{-x}} = \frac{1}{1+e^{-1.544}} = 0.824$$

$$h_0 \text{ Activation } (h_0\_acti) = \mathbf{0.824}$$

$$\begin{aligned} F(h_1) = F(n_0, n_1, n_2) &= \sum(W_i I_i) = \{ (h_1 * n_0) + (h_1 * n_1) + (h_1 * n_2) \} \\ &= \{ (0.2 * 0.52) + (0.2 * 0.75) + (0.2 * 0.97) \} \\ &= 0.104 + 0.15 + 0.194 \\ &= 0.448 \end{aligned}$$

$$h_1 \text{ Bias} : 0.2 + 0.448 = 0.648$$

$$h_1 \text{ Activation } (h_1\_acti) = \mathbf{0.656}$$

Sources of Reference: C++ Fuzzy Network by V.Rao

Neural Networks and Back Propagation Algorithm by M.CilimKovic

$$\begin{aligned}
 F(h_2) = F(n_0, n_1, n_2) &= \sum(W_i l_i) = \{ (h_2 * n_0) + (h_2 * n_1) + (h_2 * n_2) \} \\
 &= \{ (-0.5 * 0.52) + (-0.5 * 0.75) + (-0.5 * 0.97) \} \\
 &= -0.26 \quad - \quad 0.375 \quad - \quad 0.485 \\
 &= -1.12
 \end{aligned}$$

$$h_2 \text{ Bias} : 0.2 + (-1.12) = -0.92$$

$$h_2 \text{ Activation } (h_2\_Acti) = \mathbf{0.284}$$

## Activating Neurons in the Hidden Layer 01 (NH01)

Neurons in the Hidden Layer 01 are h3, h4 and h5 respectively.

$$\begin{aligned}
 F(h_3) = F(h_0, h_1, h_2) &= \sum(W_i l_i) = \{ (-0.4 * h_0) + (-0.4 * h_1) + (-0.4 * h_2) \} \\
 &= (-0.4 * 0.824) + (-0.4 * 0.656) + (-0.4 * 0.284) \\
 &= -0.329 \quad - \quad 0.262 \quad - \quad 0.113 \\
 &= -0.704
 \end{aligned}$$

$$h_3 \text{ Bias} : 0.3 - 0.704 = -0.404$$

$$h_3 \text{ Activation } (h_3\_acti) = \mathbf{0.400}$$

$$\begin{aligned}
 F(h_4) = F(h_0, h_1, h_2) &= \sum(W_i l_i) = \{ (0.8 * h_0) + (0.8 * h_1) + (0.8 * h_2) \} \\
 &= (0.8 * 0.824) + (0.8 * 0.656) + (0.8 * 0.284) \\
 &= 0.659 \quad + \quad 0.524 \quad + \quad 0.227 \\
 &= 1.41
 \end{aligned}$$

$$h_4 \text{ Bias} : 0.3 + 1.41 = 1.71$$

$$h_4 \text{ Activation } (h_4\_acti) = \mathbf{0.846}$$

$$\begin{aligned}
 F(h_5) = F(h_0, h_1, h_2) &= \sum(W_i l_i) = \{ (0.3 * h_0) + (0.3 * h_1) + (0.3 * h_2) \} \\
 &= (0.3 * 0.824) + (0.3 * 0.656) + (0.3 * 0.284) \\
 &= 0.247 \quad + \quad 0.196 \quad + \quad 0.085 \\
 &= 0.528
 \end{aligned}$$

$$h_5 \text{ Bias} : 0.3 + 0.528 = 0.828$$

$$h_5 \text{ Activation } (h_5\_acti) = \mathbf{0.695}$$

Sources of Reference: C++ Fuzzy Network by V.Rao

Neural Networks and Back Propagation Algorithm by M.CilimKovic

All neurons' stimuli calculated from the above would then be transmitted to the neurons in the end layer which responsible for finalizing and satisfying the overall neurons' stimulus.

## Activating Neurons in the Output Layer O0 (NO00)

Neurons in the output layer NO00 : O0 , O1 and O2 respectively.

$$\begin{aligned} F(O0) = F(h3, h4, h5) &= \sum(W_j I_j) = \{ (O0 * h3) + (O0 * h4) + (O5 * h5) \} \\ &= (-0.90 * 0.400) + (-0.90 * 0.846) + (-0.90 * 0.695) \\ &= -0.36 \quad \quad \quad -0.761 \quad \quad \quad -0.625 \\ &= -1.746 \end{aligned}$$

$$O0 \text{ Bias} : 0.15 - 1.746 = -1.596$$

$$FSigmoid(x) = \frac{1}{1 + e^{1.596}} = 0.168$$

$$O0 \text{ Activation } (O0\_acti) = \mathbf{0.168}$$

$$\begin{aligned} F(O1) = F(h3, h4, h5) &= \sum(W_j I_j) = \{ (O1 * h3) + (O1 * h4) + (O1 * h5) \} \\ &= (0.43 * 0.400) + (0.43 * 0.846) + (0.43 * 0.695) \\ &= 0.172 \quad \quad \quad +0.363 \quad \quad \quad +0.298 \\ &= 0.833 \end{aligned}$$

$$O1 \text{ Bias} : 0.25 + 0.833 = 1.083$$

$$O1 \text{ Activation } (O1\_acti) = \mathbf{0.747}$$

$$\begin{aligned} F(O2) = F(h3, h4, h5) &= \sum(W_j I_j) = \{ (O2 * h3) + (O2 * h4) + (O2 * h5) \} \\ &= (0.25 * 0.400) + (0.25 * 0.846) + (0.25 * 0.695) \\ &= 0.1 \quad \quad \quad +0.211 \quad \quad \quad +0.173 \\ &= 0.484 \end{aligned}$$

$$O2 \text{ Bias} : 0.05 + 0.484 = 0.534$$

$$O2 \text{ Activation } (O2\_acti) = \mathbf{0.630}$$

## Activating Neurons in the Output Layer 01 (NO01)

Neurons in the output layer 01 are O3 , O4 and O5 respectively.

$$\begin{aligned} F(O3) &= F(O0, O1, O2) = \sum(W_j I_j) = \{ (O3 * O0) + (O3 * O1) + (O3 * O2) \} \\ &= (0.11 * 0.168) + (0.11 * 0.747) + (0.11 * 0.630) \\ &= 0.018 \quad + \quad 0.082 \quad + \quad 0.069 \\ &= 0.169 \end{aligned}$$

$$\text{O3 Bias} : 0.15 + 0.169 = 0.319$$

$$\text{O3 Activation (O3_acti)} = \mathbf{0.579}$$

$$\begin{aligned} F(O4) &= F(O0, O1, O2) = \sum(W_j I_j) = \{ (O4 * O0) + (O4 * O1) + (O4 * O2) \} \\ &= (-0.67 * 0.168) + (-0.67 * 0.747) + (-0.67 * 0.630) \\ &= -0.112 \quad - \quad 0.500 \quad - \quad 0.422 \\ &= -1.034 \end{aligned}$$

$$\text{O4 Bias} : 0.25 - 1.034 = -0.784$$

$$\text{O4 Activation (O4_acti)} = \mathbf{0.313}$$

$$\begin{aligned} F(O5) &= F(O0, O1, O2) = \sum(W_j I_j) = \{ (O5 * O0) + (O5 * O1) + (O5 * O2) \} \\ &= (-0.75 * 0.168) + (-0.75 * 0.747) + (-0.75 * 0.630) \\ &= -0.126 \quad - \quad 0.560 \quad - \quad 0.472 \\ &= -1.158 \end{aligned}$$

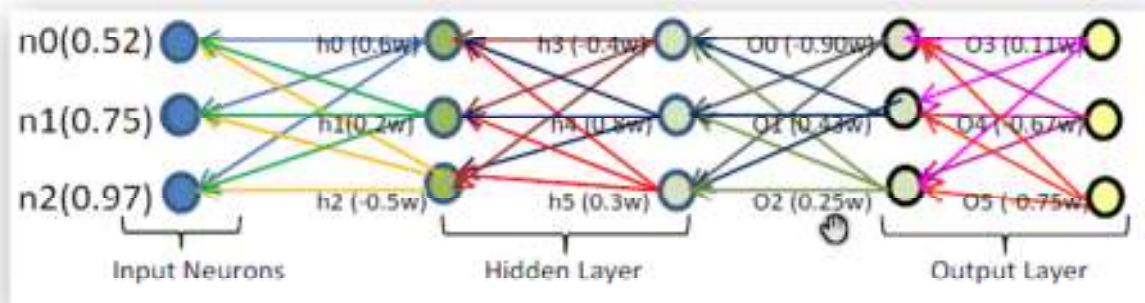
$$\text{O5 Bias} : 0.05 - 1.158 = -1.108$$

$$\text{O5 Activation (O5_acti)} = \mathbf{0.248}$$

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## Calculating Errors in the Last End Output Layer

Neurons in the last end layer are O3, O4 and O5 respectively. Look at the diagram charts below depicting position of artificial neurons settled in the layers copulatively arranged for them.



$$\mathbf{DO}^{\rightarrow} = (0.24, 0.17, 0.65)$$

$$\begin{aligned} \text{O3 Error} &= \text{O3} * (1 - \text{O3}) * (\mathbf{DO}^{\rightarrow} - \text{O3}) \\ &= 0.579 * (1 - 0.579) * (0.24 - 0.579) \\ &= -\mathbf{0.082} \end{aligned}$$

$$\begin{aligned} \text{Rate of Changes } \Delta W_{03} &= \eta * \text{O3 Error} * \text{O0} \\ &= 0.2 * (-0.082) * 0.168 \\ &= -\mathbf{0.002} \end{aligned}$$

Momentum Rate of Changes in this case is assumed to be always zero, thus  $(\alpha * \Delta(t - 1)) = 0$ .

$$\begin{aligned} W_{03} \text{ New} &= W_{03} \text{ Old} + \Delta W_{03} + (\alpha * \Delta(t - 1)) \\ &= 0.11 - 0.002 + 0 \end{aligned}$$

$$\text{Updated Weight for O3 : } = \mathbf{0.10}$$


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$$\begin{aligned} \text{O4 Error} &= \text{O4} * (1 - \text{O4}) * (\mathbf{DO}^{\rightarrow} - \text{O4}) \\ &= 0.313 * (1 - 0.313) * (0.17 - 0.313) \\ &= -\mathbf{0.030} \end{aligned}$$

$$\begin{aligned} \Delta W_{04} &= \eta * \text{O4 Error} * \text{O1} \\ &= 0.2 * (-0.030) * 0.747 \\ &= -0.004 \end{aligned}$$

Sources of Reference: C++ Fuzzy Network by V.Rao  
Neural Networks and Back Propagation Algorithm by M.CilimKovic

$$\begin{aligned} W04 \text{ New} &= W04 \text{ Old} + \Delta W04 + (\alpha * \Delta(t - 1)) \\ &= -0.67 - 0.004 + 0 \end{aligned}$$

$$O4's \text{ Updated Weight} = -0.67$$


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$$\begin{aligned} O5 \text{ Error} &= O5 * (1 - O5) * (DO^{\rightarrow} - O5) \\ &= 0.248 * (1 - 0.248) * (0.65 - 0.248) \\ &= \mathbf{0.074} \end{aligned}$$

$$\begin{aligned} \Delta W05 &= \eta * O5 \text{ Error} * O2 \\ &= 0.2 * 0.074 * 0.630 \\ &= 0.009 \end{aligned}$$

$$\begin{aligned} W05 \text{ New} &= W05 \text{ Old} + \Delta W05 + (\alpha * \Delta(t - 1)) \\ &= -0.75 + 0.009 + 0 \end{aligned}$$

$$O5's \text{ Updated Weight} = -0.74$$

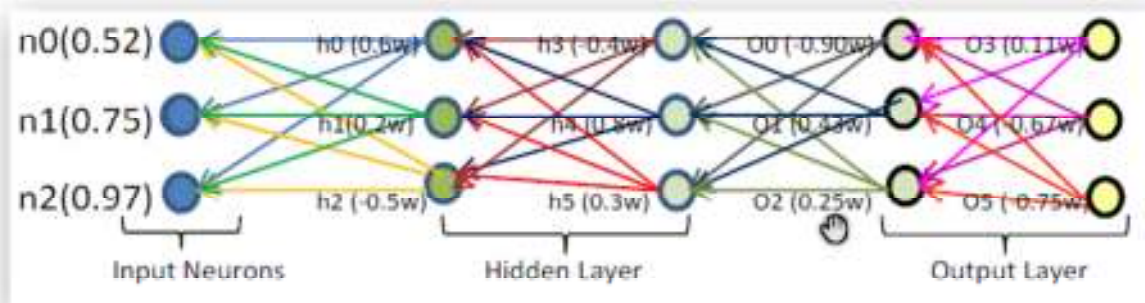
So, the New Updated Weights Set for the Last End Output Layer is { 0.10 , -0.67 , -0.74 }.

$$WO01 = \{ 0.10 , -0.67 , -0.74 \}$$

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## Calculating Errors in the 1<sup>st</sup> (First) Output Layer (NO00)

Neurons in the First (1<sup>st</sup>) Output Layer are O0, O1 and O2 respectively.



$$\begin{aligned} \text{O0 Error} &= \text{O3 Error} * \text{WO3 New} \\ &= -0.082 * 0.10 \end{aligned}$$

$$\text{O0 Error} = \mathbf{-0.0082}$$

$$\begin{aligned} \Delta \text{WO0} &= \eta * \text{O0 Error} * h3 \\ &= 0.2 * (-0.0082) * 0.400 \\ &= -0.0006 \end{aligned}$$

$$\begin{aligned} \text{WO0 New} &= \text{WO0 Old} + \Delta \text{WO0} + (\alpha * \Delta(t-1)) \\ &= -0.90 - 0.0006 + 0 \\ \text{WO0 New} &= \mathbf{-0.90} \end{aligned}$$


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$$\begin{aligned} \text{O1 Error} &= \text{O4 Error} * \text{W04 New} \\ &= -0.030 * (-0.67) \\ &= \mathbf{0.020} \end{aligned}$$

$$\begin{aligned} \Delta \text{W01} &= \eta * \text{O1 Error} * h4 \\ &= 0.2 * 0.020 * 0.846 \\ &= 0.0033 \end{aligned}$$

$$\begin{aligned} \text{W01 New} &= \text{W01 Old} + \Delta \text{W01} + (\alpha * \Delta(t-1)) \\ &= 0.43 + 0.0033 + 0 \\ \text{W01 New} &= \mathbf{0.43} \end{aligned}$$


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$$\begin{aligned} \text{O2 Error} &= \text{O5 Error} * \text{WO5 New} \\ &= 0.074 * (-0.74) \\ &= \mathbf{-0.054} \end{aligned}$$

Sources of Reference: C++ Fuzzy Network by V.Rao  
Neural Networks and Back Propagation Algorithm by M.CilimKovic



$$\begin{aligned}
 \Delta WO2 &= u_1 * O2 \text{ Error} * h5 \\
 &= 0.2 * (-0.054) * 0.695 \\
 &= -0.007
 \end{aligned}$$

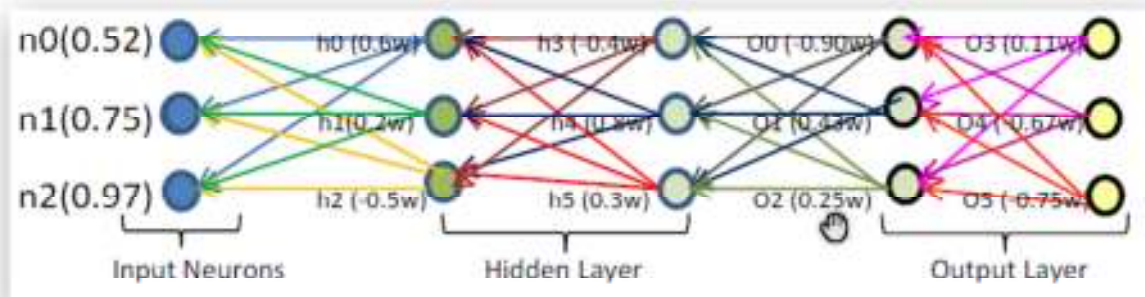
$$\begin{aligned}
 WO2 \text{ New} &= WO2 \text{ Old} + \Delta WO2 + (\alpha * \Delta(t-1)) \\
 &= 0.25 - 0.007 + 0 \\
 WO2 \text{ New} &= \mathbf{0.24}
 \end{aligned}$$

New Updated Weights Set for the 1<sup>st</sup> Layer Output is  $WO00 = \{-0.90, 0.43, 0.24\}$

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## Calculating Errors in the Last End Hidden Layer (NH01)

Neurons laid in this layer are h3, h4 and h5 respectively.



$$\begin{aligned}
 h3 \text{ Error} &= O0 \text{ Error} * WO0 \text{ New} \\
 &= -0.0082 * -0.90 \\
 &= 0.0073
 \end{aligned}$$

$$\begin{aligned}
 \Delta WH3 &= u_1 * h3 \text{ Error} * h0 \\
 &= 0.2 * 0.0073 * 0.824 \\
 &= 0.0012
 \end{aligned}$$

$$\begin{aligned}
 WH3 \text{ New} &= WH3 \text{ Old} + \Delta WH3 + (\alpha * \Delta(t-1)) \\
 &= -0.4 + 0.0012 + 0 \\
 &= \mathbf{-0.39}
 \end{aligned}$$

$$\begin{aligned}
 \text{h4 Error} &= \text{O1 Error} * \text{WO1 New} \\
 &= 0.020 * 0.43 \\
 &= 0.0086 \\
 \\
 \Delta \text{WH4} &= \eta * \text{h4 Error} * \text{h1} \\
 &= 0.2 * 0.0086 * 0.656 \\
 &= 0.0011 \\
 \\
 \text{WH4 New} &= \text{WH4 Old} + \Delta \text{WH4} + ( \alpha * \Delta ( t - 1 ) ) \\
 &= 0.8 + 0.0011 + 0 \\
 &= \mathbf{0.8}
 \end{aligned}$$


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$$\begin{aligned}
 \text{h5 Error} &= \text{O2 Error} * \text{W02 New} \\
 &= -0.054 * 0.24 \\
 &= -0.012 \\
 \\
 \Delta \text{WH5} &= \eta * \text{h5 Error} * \text{h2} \\
 &= 0.2 * -0.012 * 0.284 \\
 &= -0.0006 \\
 \\
 \text{WH5 New} &= \text{WH5 Old} + \Delta \text{WH5} \\
 &= 0.3 - 0.0006 \\
 &= \mathbf{0.29}
 \end{aligned}$$

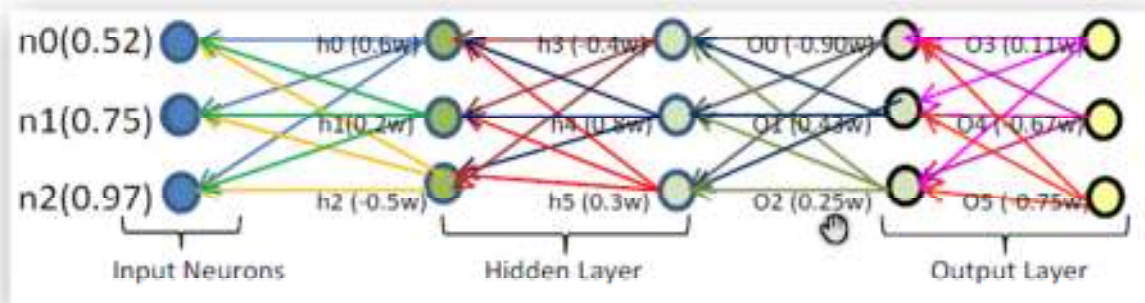
**So, the new adjusted weights set for this end hidden layer (NH01) is WH01 = { -0.39 , 0.8 , 0.29 }**

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## Calculating Errors in the First ( 1<sup>st</sup> ) Hidden Layer (NH00)

Neurons in this layer are h0 , h1 and h2 respectively. The determined Learning Rate ( $\eta$ ) for this layer to the input layer neurons is set to be 0.15.



$$\begin{aligned}
 h0 \text{ Error} &= h3 \text{ Error} * WH3 \text{ New} \\
 &= 0.0073 * -0.39 \\
 &= -0.0028
 \end{aligned}$$

$$\begin{aligned}
 WH0 \text{ New} &= WH0 \text{ Old} + \Delta WH0 + 0 \\
 &= 0.6 - 0.00021 + 0 \\
 &= \mathbf{0.59}
 \end{aligned}$$

$$\begin{aligned}
 \Delta WH0 &= \eta * h0 \text{ Error} * n0 \\
 &= 0.15 * -0.0028 * 0.52 \\
 &= -0.00021
 \end{aligned}$$


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$$\begin{aligned}
 h1 \text{ Error} &= h4 \text{ Error} * WH4 \text{ New} \\
 &= 0.0086 * 0.8 \\
 &= 0.0068
 \end{aligned}$$

$$\begin{aligned}
 \Delta WH1 &= \eta * h1 \text{ Error} * n1 \\
 &= 0.15 * 0.0068 * 0.75 \\
 &= 0.00076
 \end{aligned}$$

$$\begin{aligned}
 WH1 \text{ New} &= WH1 \text{ Old} + \Delta WH1 + 0 \\
 &= 0.2 + 0.00076 + 0 \\
 &= \mathbf{0.2}
 \end{aligned}$$


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$$\begin{aligned}
 h2 \text{ Error} &= h5 \text{ Error} * WH5 \text{ New} \\
 &= -0.012 * 0.29 \\
 &= -0.0034
 \end{aligned}$$

$$\begin{aligned}
 \Delta WH2 &= \eta * h2 \text{ Error} * n2 \\
 &= 0.15 * -0.0034 * 0.97 \\
 &= -0.00049
 \end{aligned}$$

$$\begin{aligned}
 \text{WH2 New} &= \text{WH2 Old} + \Delta\text{WH2} + 0 \\
 &= -0.5 - 0.00049 + 0 \\
 &= \mathbf{-0.5}
 \end{aligned}$$

So, the new adjusted weights set for this hidden layer NH00 is:  
 WH00 = { 0.59 , 0.2 , -0.5 }