

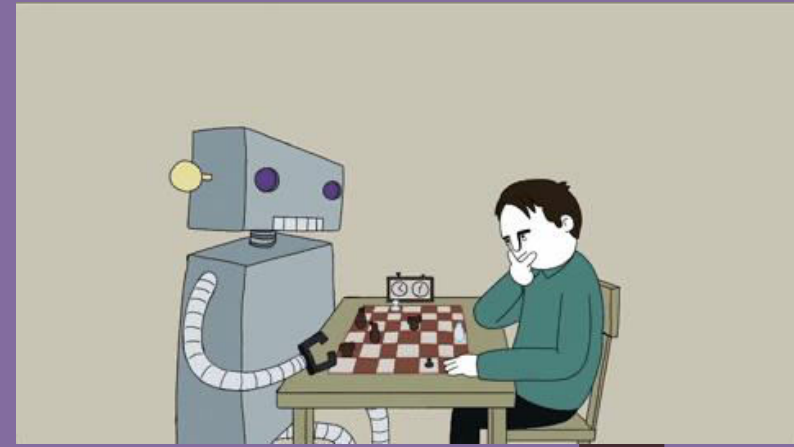
# CSE4006

# DEEP LEARNING

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# Module No. 2

## Practical Deep Networks

### 8 Hours

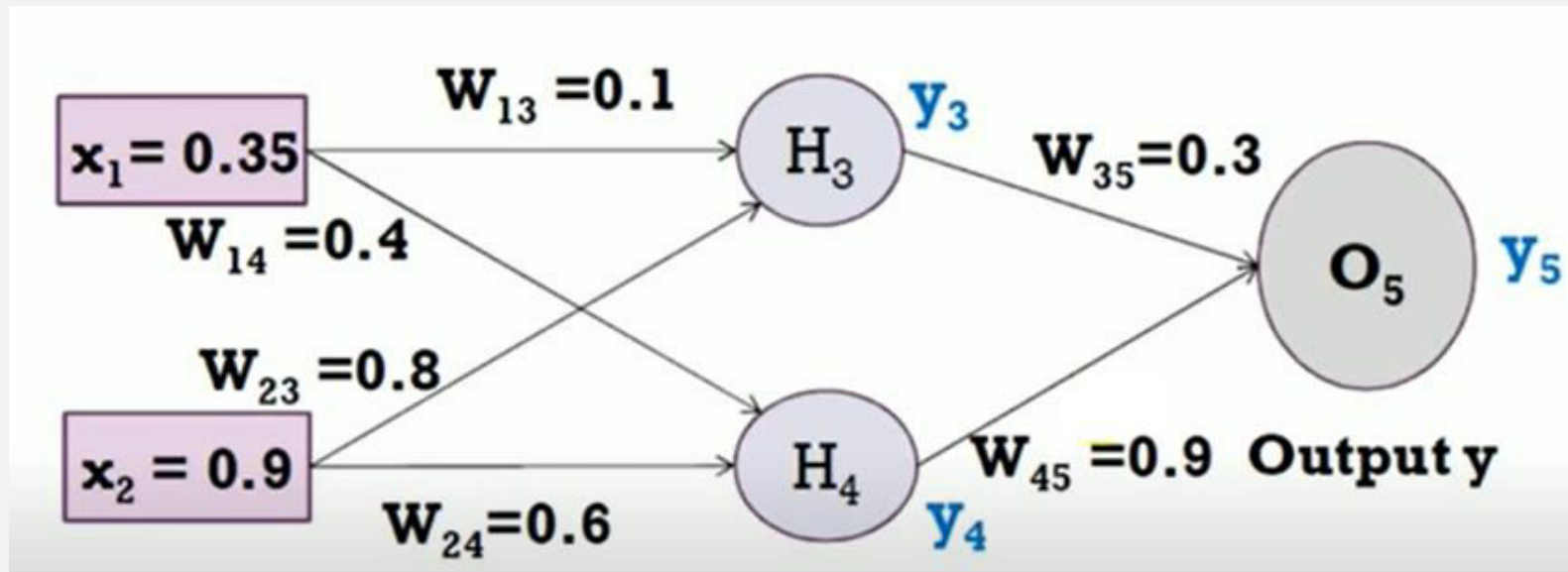
- Multilayer Perceptron
- Gradient based Learning
- Backpropagation Algorithm
- Regularization for Deep Learning
- Optimization for training deep models

A decorative L-shaped frame composed of thick, dark purple lines. One part of the frame is on the left, extending from the top-left towards the center, and the other part is on the right, extending from the bottom-right towards the center. They meet at the bottom-right corner, forming a large, open square frame.

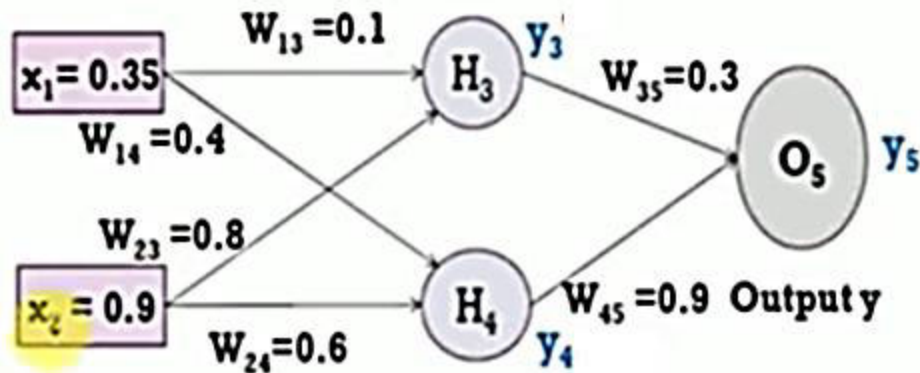
# BACKPROPAGATION SOLVED EXERCISE - 1

# Backpropagation Algorithm Exercise - 1

- Assume that the neurons have a Sigmoid Activation function, perform a forward pass and a backward pass on the network. Assume that the actual output of  $y$  is 0.5 and learning rate is 1. Perform another forward pass.



# Backpropagation Algorithm Exercise - 1



- Forward Pass: Compute output for  $y_3$ ,  $y_4$  and  $y_5$ .

$$a_j = \sum_l (w_{lj} * x_l) \quad y_j = F(a_j) = \frac{1}{1 + e^{-a_j}}$$

$$\begin{aligned} a_1 &= (w_{13} * x_1) + (w_{23} * x_2) \\ &= (0.1 * 0.35) + (0.8 * 0.9) = 0.755 \\ y_3 &= f(a_1) = 1 / (1 + e^{-0.755}) = 0.68 \end{aligned}$$

$$\begin{aligned} a_2 &= (w_{14} * x_1) + (w_{24} * x_2) \\ &= (0.4 * 0.35) + (0.6 * 0.9) = 0.68 \\ y_4 &= f(a_2) = 1 / (1 + e^{-0.68}) = 0.6637 \end{aligned}$$

$$\text{Error} = y_{\text{target}} - y_5 = -0.19$$

$$0.5 - 0.69$$

$$\begin{aligned} a_3 &= (w_{35} * y_3) + (w_{45} * y_4) \\ &= (0.3 * 0.68) + (0.9 * 0.6637) = 0.801 \\ y_5 &= f(a_3) = 1 / (1 + e^{-0.801}) = \mathbf{0.69 \text{ (Network Output)}} \end{aligned}$$

# Backpropagation Algorithm Exercise - 1

- Each weight changed by:

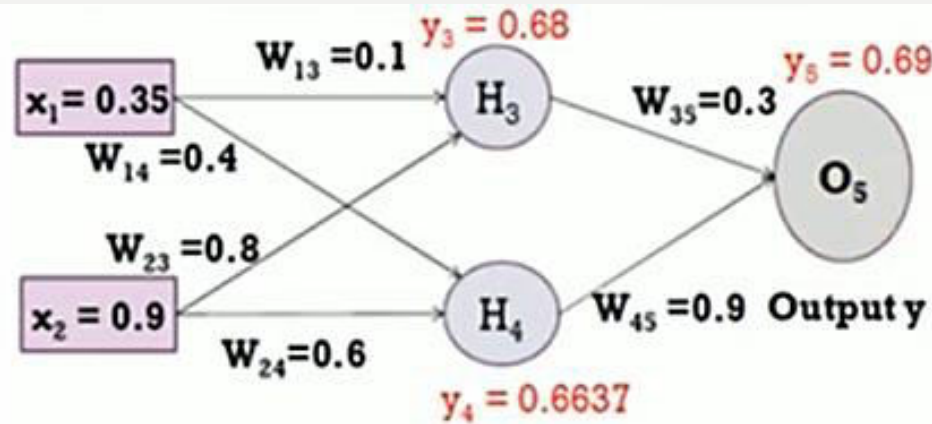
$$\Delta w_{ji} = \eta \delta_j o_i$$

$$\delta_j = o_j(1 - o_j)(t_j - o_j) \quad \text{if } j \text{ is an output unit}$$

$$\delta_j = o_j(1 - o_j) \sum_k \delta_k w_{kj} \quad \text{if } j \text{ is a hidden unit}$$

- where  $\eta$  is a constant called the learning rate
- $t_j$  is the correct teacher output for unit  $j$
- $\delta_j$  is the error measure for unit  $j$

# Backpropagation Algorithm Exercise - 1



• Backward Pass: Compute  $\delta_3$ ,  $\delta_4$  and  $\delta_5$ .

For output unit:

$$\delta_5 = y(1-y)(y_{\text{target}} - y) = 0.69 * (1 - 0.69) * (0.5 - 0.69) = -0.0406$$

For hidden unit:

$$\delta_3 = y_3(1-y_3) w_{35} * \delta_5 = 0.68 * (1 - 0.68) * (0.3 * -0.0406) = -0.00265$$

$$\delta_4 = y_4(1-y_4) w_{45} * \delta_5 = 0.6637 * (1 - 0.6637) * (0.9 * -0.0406) = -0.0082$$

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

$$\delta_j = o_j(1-o_j)(t_j - o_j) \quad \text{if } j \text{ is an output unit}$$

$$\delta_j = o_j(1-o_j) \sum_k \delta_k w_{kj} \quad \text{if } j \text{ is a hidden unit}$$



# Backpropagation Algorithm Exercise - 1

Compute new weights

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

$$\Delta w_{45} = \eta \delta_5 y_4 = 1 * -0.0406 * 0.6637 = -0.0269$$

$$w_{45}(\text{new}) = \Delta w_{45} + w_{45}(\text{old}) = -0.0269 + (0.9) = 0.8731$$

$$\Delta w_{14} = \eta \delta_4 x_1 = 1 * -0.0082 * 0.35 = -0.00287$$

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

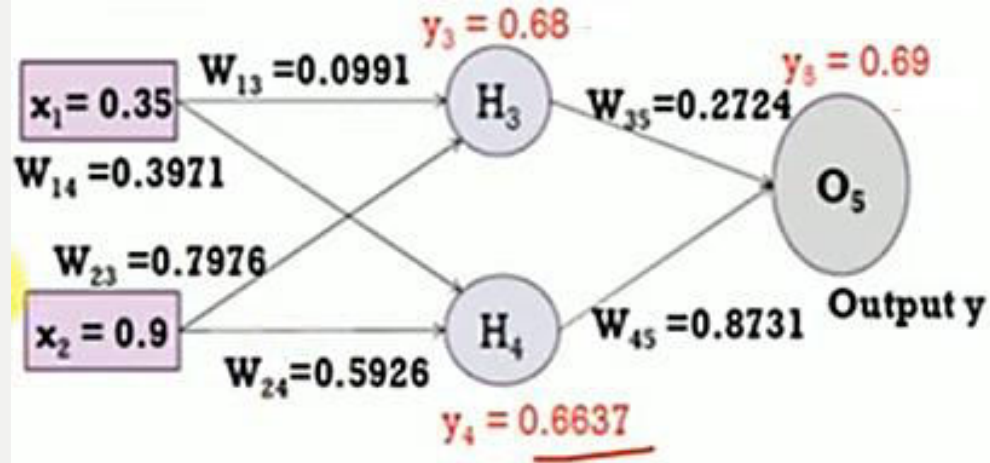


# Backpropagation Algorithm Exercise - 1

- Similarly, update all other weights

<b>i</b>	<b>j</b>	<b><math>w_{ij}</math></b>	<b><math>\delta_i</math></b>	<b><math>x_i</math></b>	<b><math>\eta</math></b>	<b>Updated <math>w_{ij}</math></b>
1	3	0.1	-0.00265	0.35	1	0.0991
2	3	0.8	-0.00265	0.9	1	0.7976
1	4	0.4	-0.0082	0.35	1	0.3971
2	4	0.6	-0.0082	0.9	1	0.5926
3	5	0.3	-0.0406	0.68	1	0.2724
4	5	0.9	-0.0406	0.6637	1	0.8731

# Backpropagation Algorithm Exercise - 1



$$\text{Error} = y_{\text{target}} - y_5 = -0.182$$

- Forward Pass: Compute output for  $y_3$ ,  $y_4$  and  $y_5$ .

$$a_j = \sum_l (w_{l,j} * x_l) \quad y_j = F(a_j) = \frac{1}{1 + e^{-a_j}}$$

$$a_1 = (w_{13} * x_1) + (w_{23} * x_2) = (0.0991 * 0.35) + (0.7976 * 0.9) = 0.7525$$

$$y_3 = f(a_1) = 1 / (1 + e^{-0.7525}) = 0.6797$$

$$a_2 = (w_{14} * x_1) + (w_{24} * x_2) = (0.3971 * 0.35) + (0.5926 * 0.9) = 0.6723$$

$$y_4 = f(a_2) = 1 / (1 + e^{-0.6723}) = 0.6620$$

$$a_3 = (w_{35} * y_3) + (w_{45} * y_4) = (0.2724 * 0.6797) + (0.8731 * 0.6620) = 0.7631$$

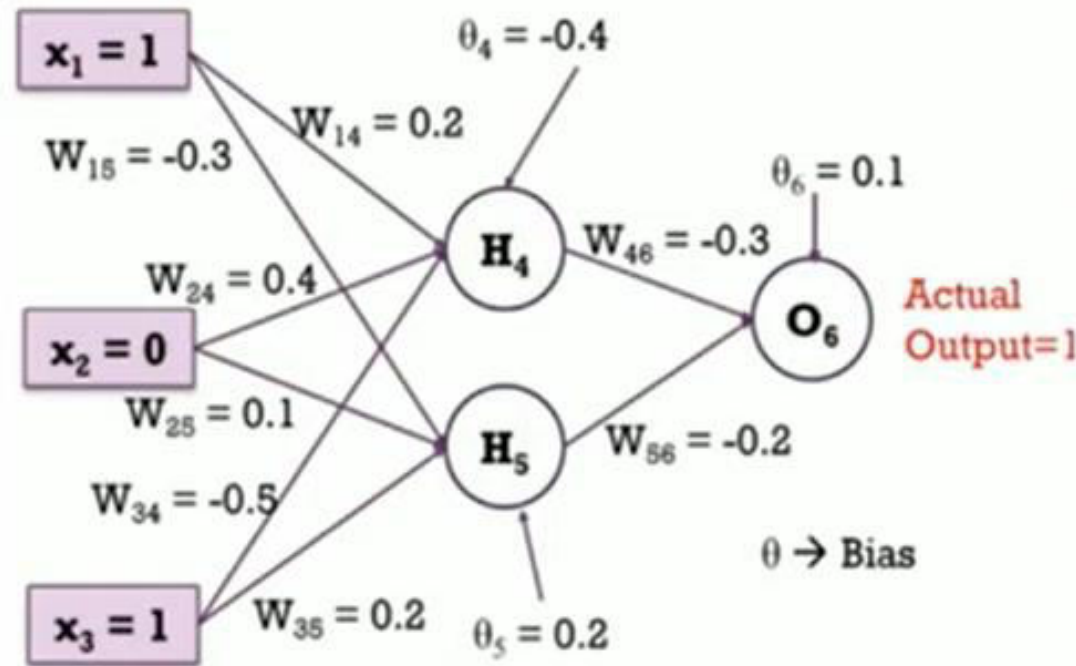
$$y_5 = f(a_3) = 1 / (1 + e^{-0.7631}) = \mathbf{0.6820 \text{ (Network Output)}}$$



# BACKPROPAGATION SOLVED EXERCISE- 2



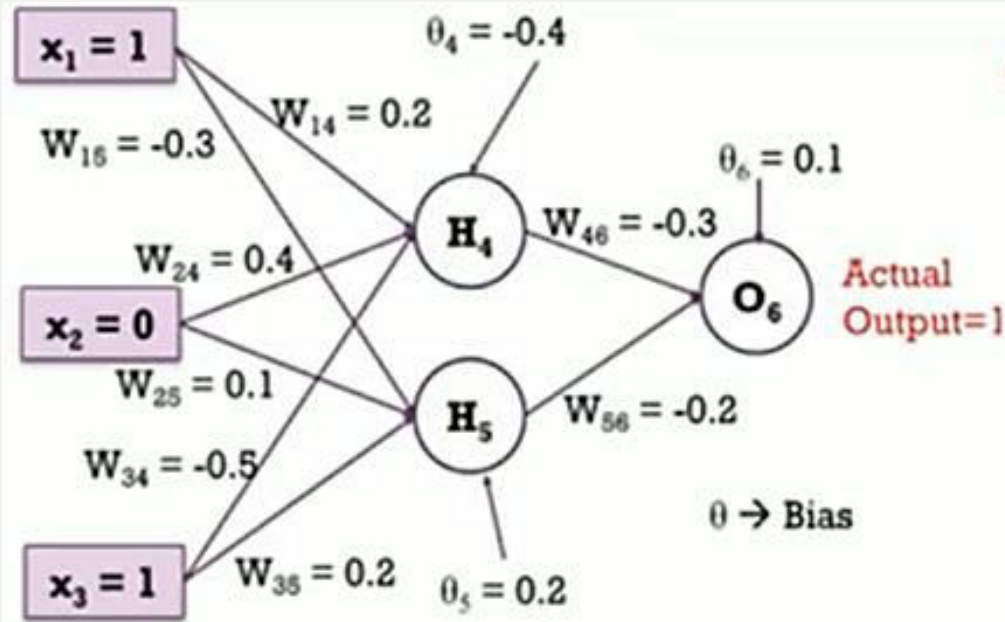
# Backpropagation Algorithm Exercise - 2



Assume that the neurons have a sigmoid activation function, perform a forward pass and a backward pass on the network. Assume that the actual output of  $y$  is 1 and learning rate is 0.9. Perform another forward pass.



# Backpropagation Algorithm Exercise - 2



$$\text{Error} = y_{\text{target}} - y_6 = 0.526$$

- Forward Pass: Compute output for  $y_4$ ,  $y_5$  and  $y_6$ .

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

$$\begin{aligned} a_4 &= (w_{14} * x_1) + (w_{24} * x_2) + (w_{34} * x_3) + \theta_4 \\ &= (0.2 * 1) + (0.4 * 0) + (-0.5 * 1) + (-0.4) = -0.7 \\ O(H_4) = y_4 &= f(a_4) = 1 / (1 + e^{0.7}) = 0.332 \end{aligned}$$

$$\begin{aligned} a_5 &= (w_{15} * x_1) + (w_{25} * x_2) + (w_{35} * x_3) + \theta_5 \\ &= (-0.3 * 1) + (0.1 * 0) + (0.2 * 1) + (0.2) = 0.1 \\ O(H_5) = y_5 &= f(a_5) = 1 / (1 + e^{-0.1}) = 0.525 \end{aligned}$$

$$\begin{aligned} a_6 &= (w_{46} * H_4) + (w_{56} * H_5) + \theta_6 \\ &= (-0.3 * 0.332) + (-0.2 * 0.525) + 0.1 = -0.105 \\ O(O_6) = y_6 &= f(a_6) = 1 / (1 + e^{0.105}) = \mathbf{0.474} \end{aligned}$$

# Backpropagation Algorithm Exercise - 2

- Each weight changed by:

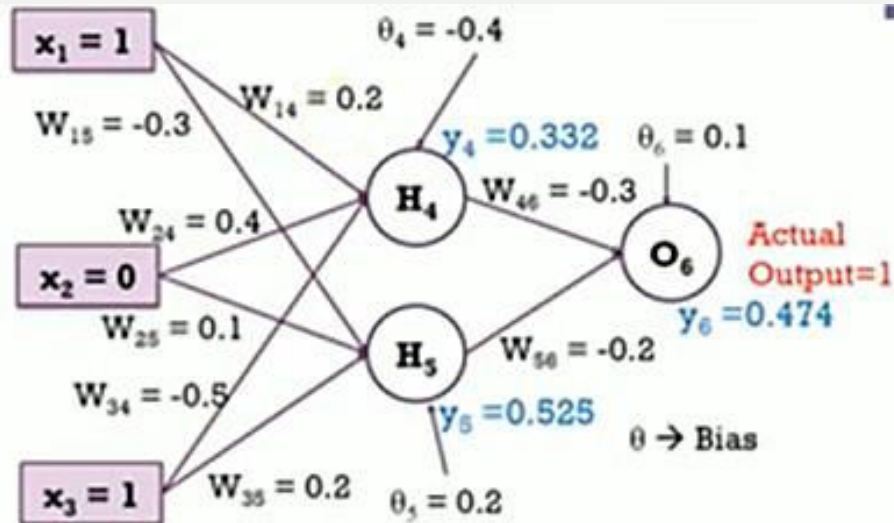
$$\Delta w_{ji} = \eta \delta_j o_i$$

$$\delta_j = o_j(1 - o_j)(t_j - o_j) \quad \text{if } j \text{ is an output unit}$$

$$\delta_j = o_j(1 - o_j) \sum_k \delta_k w_{kj} \quad \text{if } j \text{ is a hidden unit}$$

- where  $\eta$  is a constant called the learning rate
- $t_j$  is the correct teacher output for unit  $j$
- $\delta_j$  is the error measure for unit  $j$

# Backpropagation Algorithm Exercise - 2



Compute new weights

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

$$\Delta w_{46} = \eta \delta_6 y_4 = 0.9 * 0.1311 * 0.332 = 0.03917$$

$$w_{46}(\text{new}) = \Delta w_{46} + w_{46}(\text{old}) = 0.03917 + (-0.3) = -0.261$$

$$\Delta w_{14} = \eta \delta_4 x_1 = 0.9 * -0.0087 * 1 = -0.0078$$

$$w_{14}(\text{new}) = \Delta w_{14} + w_{14}(\text{old}) = -0.0078 + 0.2 = \underline{0.192}$$

- Backward Pass: Compute  $\delta_4$ ,  $\delta_5$  and  $\delta_6$ .

For output unit:

$$\begin{aligned} \delta_6 &= y_6(1-y_6)(y_{\text{target}} - y_6) \\ &= 0.474 * (1-0.474) * (1-0.474) = 0.1311 \end{aligned}$$

For hidden unit:

$$\begin{aligned} \delta_5 &= y_5(1-y_5) w_{56} * \delta_6 \\ &= 0.525 * (1 - 0.525) * (-0.2 * 0.1311) = -0.0065 \end{aligned}$$

$$\begin{aligned} &= y_4(1-y_4) w_{46} * \delta_6 \\ &= 0.332 * (1 - 0.332) * (-0.3 * 0.1331) = -0.0087 \end{aligned}$$



# Backpropagation Algorithm Exercise - 2

Similarly, update all other weights

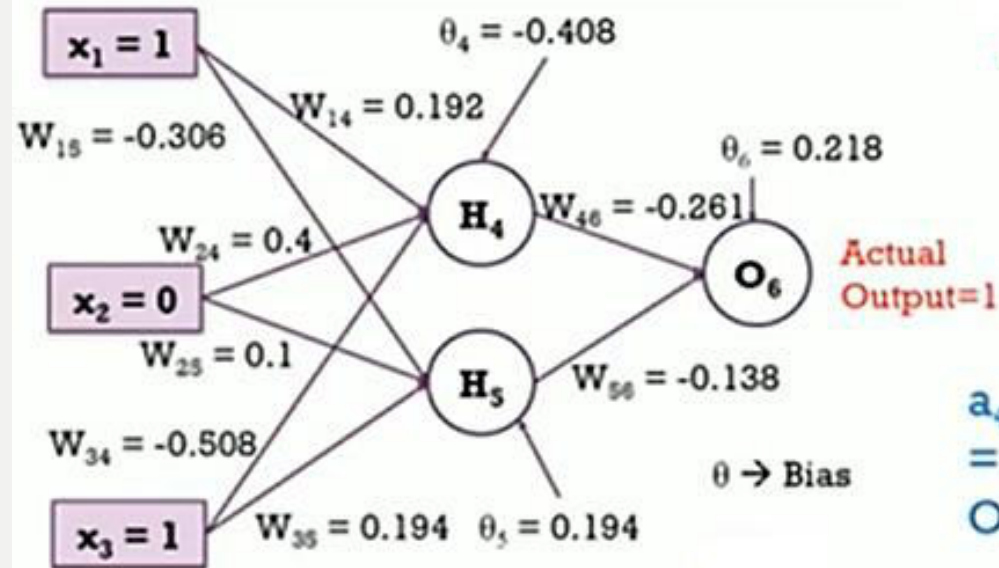
<b>i</b>	<b>j</b>	<b><math>w_{ij}</math></b>	<b><math>\delta_j</math></b>	<b><math>x_i</math></b>	<b><math>\eta</math></b>	<b>Updated <math>w_{ij}</math></b>
4	6	-0.3	0.1311	0.332	0.9	-0.261
5	6	-0.2	0.1311	0.525	0.9	-0.138
1	4	0.2	-0.0087	1	0.9	0.192
1	5	-0.3	-0.0065	1	0.9	-0.306
2	4	0.4	-0.0087	0	0.9	0.4
2	5	0.1	-0.0065	0	0.9	0.1
3	4	-0.5	-0.0087	1	0.9	-0.508
3	5	0.2	-0.0065	1	0.9	0.194

# Backpropagation Algorithm Exercise - 2

- Similarly, update bias weights

$\theta_j$	Previous $\theta_j$	$\delta_j$	$\eta$	Updated $\theta_j$
$\Theta_6$	0.1	0.1311	0.9	0.218
$\Theta_5$	0.2	-0.0065	0.9	0.194
$\Theta_4$	-0.4	-0.0087	0.9	-0.408

# Backpropagation Algorithm Exercise - 2



$$\text{Error} = y_{\text{target}} - y_6 = 0.485$$

- Forward Pass: Compute output for  $y_4$ ,  $y_5$  and  $y_6$ .

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

$$\begin{aligned} a_4 &= (w_{14} * x_1) + (w_{24} * x_2) + (w_{34} * x_3) + \theta_4 \\ &= (0.192 * 1) + (0.4 * 0) + (-0.508 * 1) + (-0.408) = -0.724 \\ O(H_4) = y_4 &= f(a_4) = 1 / (1 + e^{0.724}) = 0.327 \end{aligned}$$

$$\begin{aligned} a_5 &= (w_{15} * x_1) + (w_{25} * x_2) + (w_{35} * x_3) + \theta_5 \\ &= (-0.306 * 1) + (0.1 * 0) + (0.194 * 1) + (0.194) = 0.082 \\ O(H_5) = y_5 &= f(a_5) = 1 / (1 + e^{-0.082}) = 0.520 \end{aligned}$$

$$\begin{aligned} a_6 &= (w_{46} * H_4) + (w_{56} * H_5) + \theta_6 \\ &= (-0.261 * 0.327) + (-0.138 * 0.520) + 0.218 = 0.061 \\ O(O_6) = y_6 &= f(a_6) = 1 / (1 + e^{-0.061}) = \mathbf{0.515 \text{ (Network Output)}} \end{aligned}$$