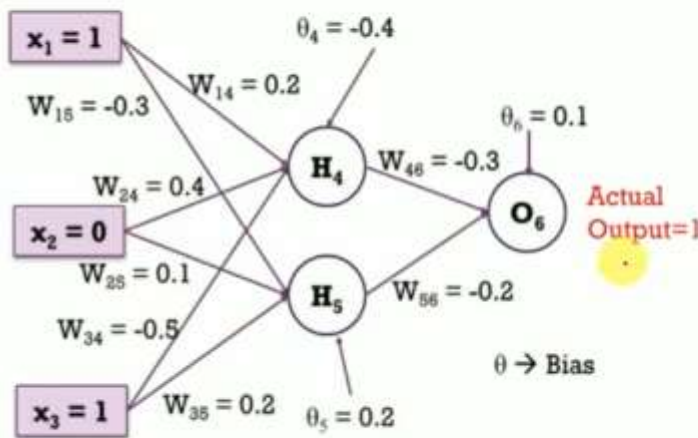
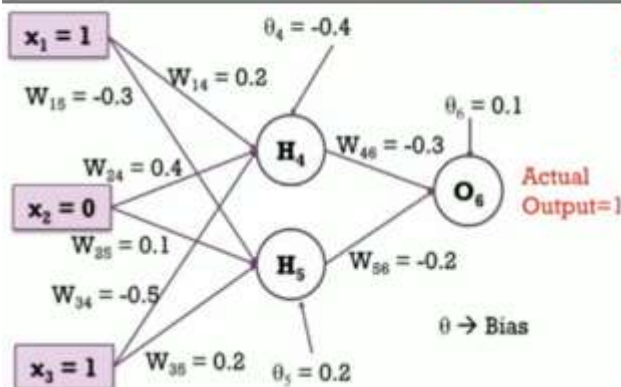


Back Propagation Solved Example - 2



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.

Back Propagation Solved Example - 2



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

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

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

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

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

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

Back Propagation Solved Example - 2

- Each weight changed by:

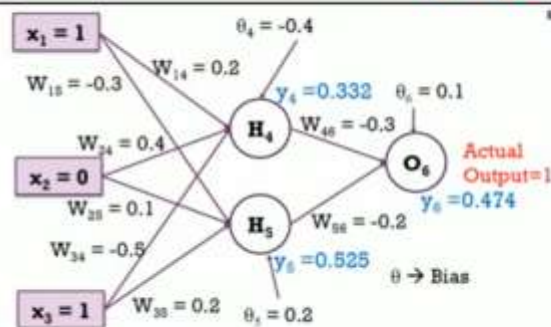
$$\Delta w_{ji} = \eta \delta_j o_i$$

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

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

- where η is a constant called the learning rate
- t_j is the correct teacher output for unit j
- δ_j is the error measure for unit j

Back Propagation Solved Example - 2



Compute new weights

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

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

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

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

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

- Backward Pass: Compute δ_4, δ_5 and δ_6 .

For output unit:

$$\delta_6 = y_6(1 - y_6)(y_{\text{target}} - y_6)$$

$$= 0.474 * (1 - 0.474) * (1 - 0.474) = 0.1311$$

For hidden unit:

$$\delta_5 = y_5(1 - y_5) w_{56} * \delta_6$$

$$= 0.525 * (1 - 0.525) * (-0.2 * 0.1311) = -0.0065$$

$$= y_4(1 - y_4) w_{46} * \delta_6$$

$$= 0.332 * (1 - 0.332) * (-0.3 * 0.1331) = -0.0087$$

Back Propagation Solved Example - 2

- Similarly, update all other weights

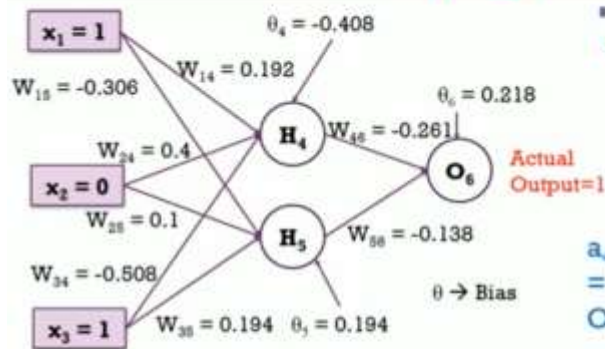
i	j	w_{ij}	δ_j	x_i	η	Updated w_{ij}
4	6	-0.3	0.1311	0.332	0.9	-0.261
5	6	-0.2	0.1311	0.525	0.9	-0.138
1	4	0.2	-0.0087	1	0.9	0.192
1	5	-0.3	-0.0065	1	0.9	-0.306
2	4	0.4	-0.0087	0	0.9	0.4
2	5	0.1	-0.0065	0	0.9	0.1
3	4	-0.5	-0.0087	1	0.9	-0.508
3	5	0.2	-0.0065	1	0.9	0.194

Back Propagation Solved Example - 2

- Similarly, update bias weights

θ_j	Previous θ_j	δ_j	η	Updated θ_j
Θ_6	0.1	0.1311	0.9	0.218
Θ_5	0.2	-0.0065	0.9	0.194
Θ_4	-0.4	-0.0087	0.9	-0.408

Back Propagation Solved Example - 2



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

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

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

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

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

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