

## + Algorithm with Backpropagation

**Algorithm: Backpropagation.** Neural network learning for classification or prediction, using the backpropagation algorithm.

**Input:**

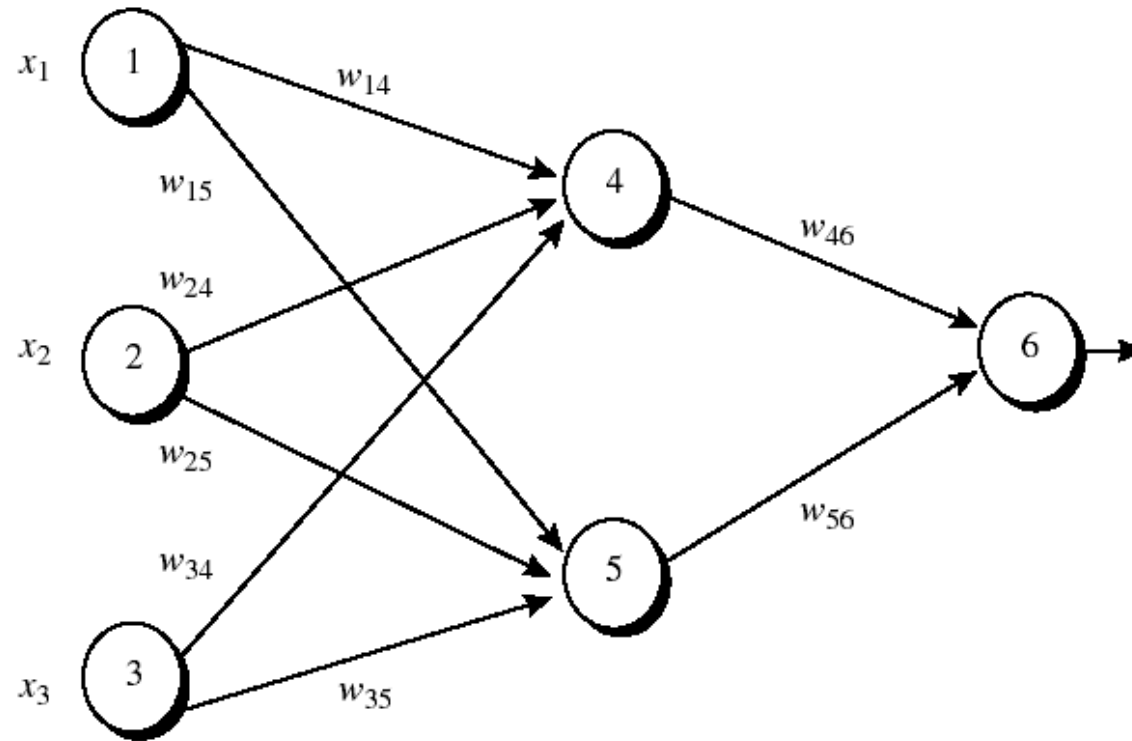
- $D$ , a data set consisting of the training tuples and their associated target values;
- $l$ , the learning rate;
- *network*, a multilayer feed-forward network.

**Output:** A trained neural network.

## + Algorithm with Backpropagation

- (1) Initialize all weights and biases in *network*;
- (2) while terminating condition is not satisfied {
- (3)     for each training tuple  $\mathbf{X}$  in  $D$  {
- (4)         // Propagate the inputs forward:
- (5)         for each input layer unit  $j$  {
- (6)              $O_j = I_j$ ; // output of an input unit is its actual input value
- (7)         for each hidden or output layer unit  $j$  {
- (8)              $I_j = \sum_i w_{ij} O_i + \theta_j$ ; // compute the net input of unit  $j$  with respect to the previous layer,  $i$
- (9)              $O_j = \frac{1}{1+e^{-I_j}}$ ; } // compute the output of each unit  $j$
- (10)         // Backpropagate the errors:
- (11)         for each unit  $j$  in the output layer
- (12)              $Err_j = O_j(1 - O_j)(T_j - O_j)$ ; // compute the error
- (13)         for each unit  $j$  in the hidden layers, from the last to the first hidden layer
- (14)              $Err_j = O_j(1 - O_j) \sum_k Err_k w_{jk}$ ; // compute the error with respect to the next higher layer,  $k$
- (15)         for each weight  $w_{ij}$  in *network* {
- (16)              $\Delta w_{ij} = (l) Err_j O_i$ ; // weight increment
- (17)              $w_{ij} = w_{ij} + \Delta w_{ij}$ ; } // weight update
- (18)         for each bias  $\theta_j$  in *network* {
- (19)              $\Delta \theta_j = (l) Err_j$ ; // bias increment
- (20)              $\theta_j = \theta_j + \Delta \theta_j$ ; } // bias update
- (21)         } }

+ Example



Initial input, weight, and bias values.

$x_1$	$x_2$	$x_3$	$w_{14}$	$w_{15}$	$w_{24}$	$w_{25}$	$w_{34}$	$w_{35}$	$w_{46}$	$w_{56}$	$\theta_4$	$\theta_5$	$\theta_6$
1	0	1	0.2	-0.3	0.4	0.1	-0.5	0.2	-0.3	-0.2	-0.4	0.2	0.1

The net input and output calculations.

<i>Unit <math>j</math></i>	<i>Net input, <math>I_j</math></i>	<i>Output, <math>O_j</math></i>
4	$0.2 + 0 - 0.5 - 0.4 = -0.7$	$1/(1 + e^{0.7}) = 0.332$
5	$-0.3 + 0 + 0.2 + 0.2 = 0.1$	$1/(1 + e^{-0.1}) = 0.525$
6	$(-0.3)(0.332) - (0.2)(0.525) + 0.1 = -0.105$	$1/(1 + e^{0.105}) = 0.474$

Calculation of the error at each node.

<i>Unit <math>j</math></i>	<i>Err<sub><math>j</math></sub></i>
6	$(0.474)(1 - 0.474)(1 - 0.474) = 0.1311$
5	$(0.525)(1 - 0.525)(0.1311)(-0.2) = -0.0065$
4	$(0.332)(1 - 0.332)(0.1311)(-0.3) = -0.0087$

+

Calculations for weight and bias updating.

<i>Weight or bias</i>	<i>New value</i>
$w_{46}$	$-0.3 + (0.9)(0.1311)(0.332) = -0.261$
$w_{56}$	$-0.2 + (0.9)(0.1311)(0.525) = -0.138$
$w_{14}$	$0.2 + (0.9)(-0.0087)(1) = 0.192$
$w_{15}$	$-0.3 + (0.9)(-0.0065)(1) = -0.306$
$w_{24}$	$0.4 + (0.9)(-0.0087)(0) = 0.4$
$w_{25}$	$0.1 + (0.9)(-0.0065)(0) = 0.1$
$w_{34}$	$-0.5 + (0.9)(-0.0087)(1) = -0.508$
$w_{35}$	$0.2 + (0.9)(-0.0065)(1) = 0.194$
$\theta_6$	$0.1 + (0.9)(0.1311) = 0.218$
$\theta_5$	$0.2 + (0.9)(-0.0065) = 0.194$
$\theta_4$	$-0.4 + (0.9)(-0.0087) = -0.408$