

# ARTIFICIAL NEURAL NETWORKS

**Scholastic Video Book Series**

Artificial Neural Networks

Part 3

(Back Propagation)

(with English Narrations)

**Scholastic**

home.video.tutor

@gmail.com

<http://scholastictutors.webs.com>

(<http://scholastictutors.webs.com/Scholastic-Book-NeuralNetworks-Part03-2013-10-22.pdf>)

**Scholastic**  
home.video.tutor  
@gmail.com

**©Scholastic Tutors (Oct, 2013)**

ISVT 911-0-20-131022-1

**Scholastic**  
home.video.tutor  
@gmail.com

# Artificial Neural Networks - #3

## Back-Propagation

Click here to see the video

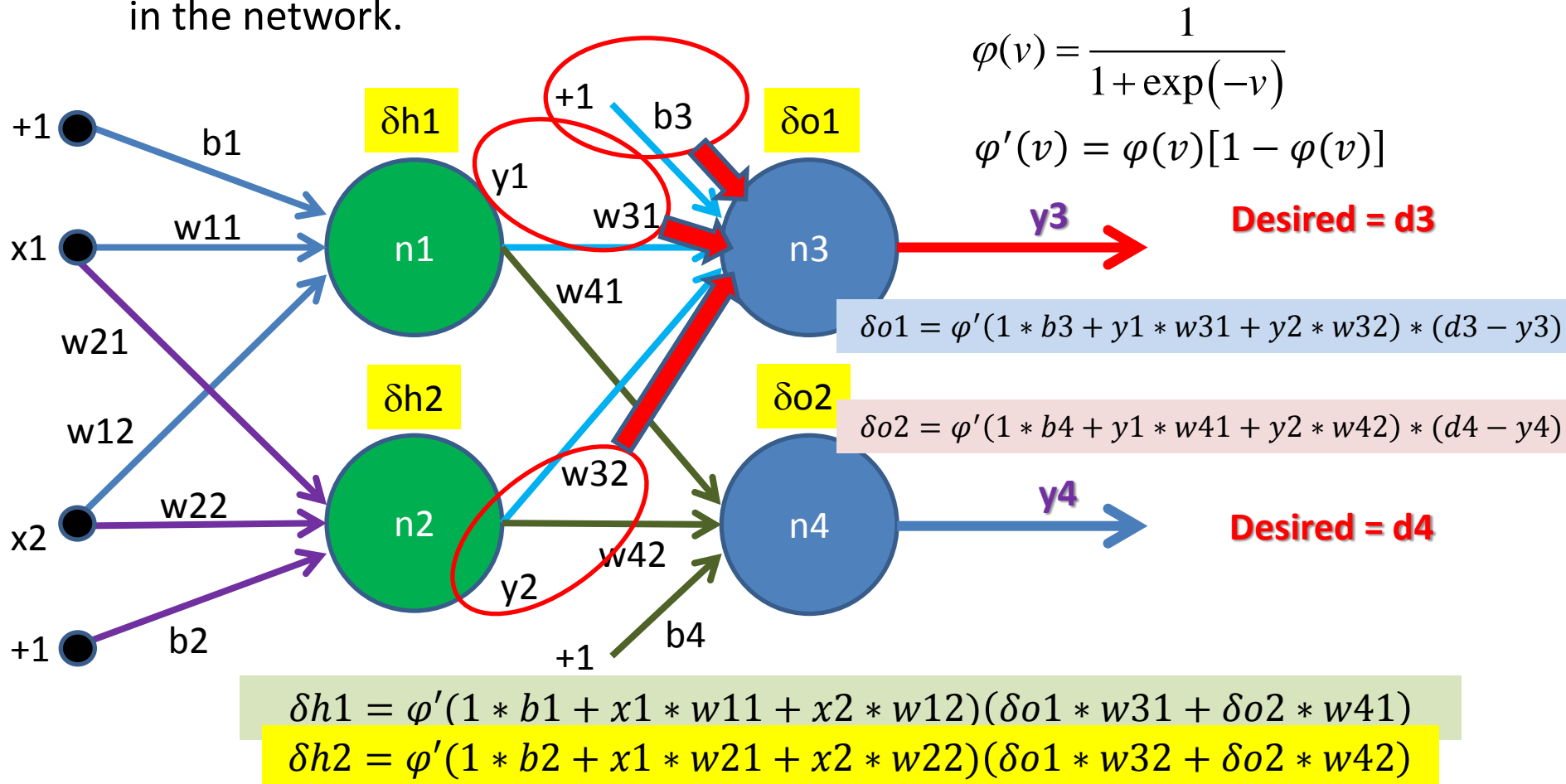
<http://youtu.be/I2I5ztVfUSE>

<http://scholastictutors.webs.com>

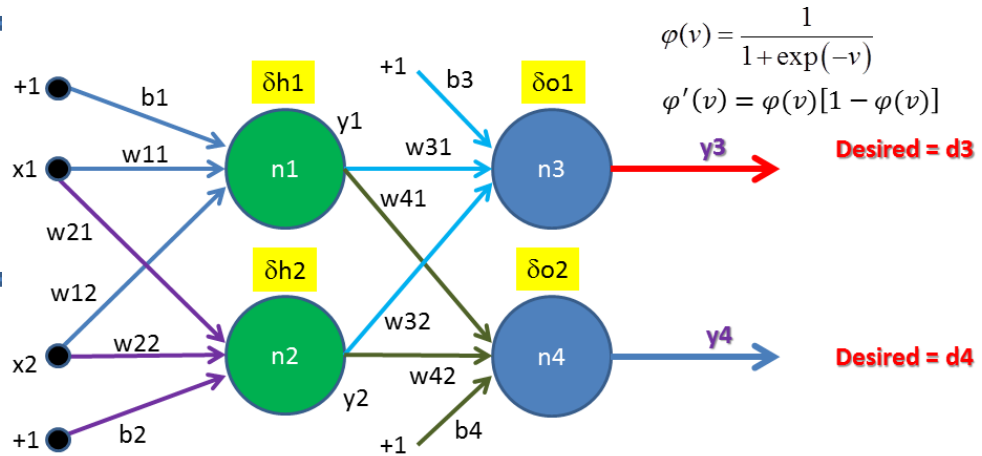
(ANN-003)

## Back-Propagation Algorithm

- Step 1:** Calculate the local gradients ( $\delta o1$ ,  $\delta o2$ ,  $\delta h1$  and  $\delta h2$ ) for the nodes in the network.



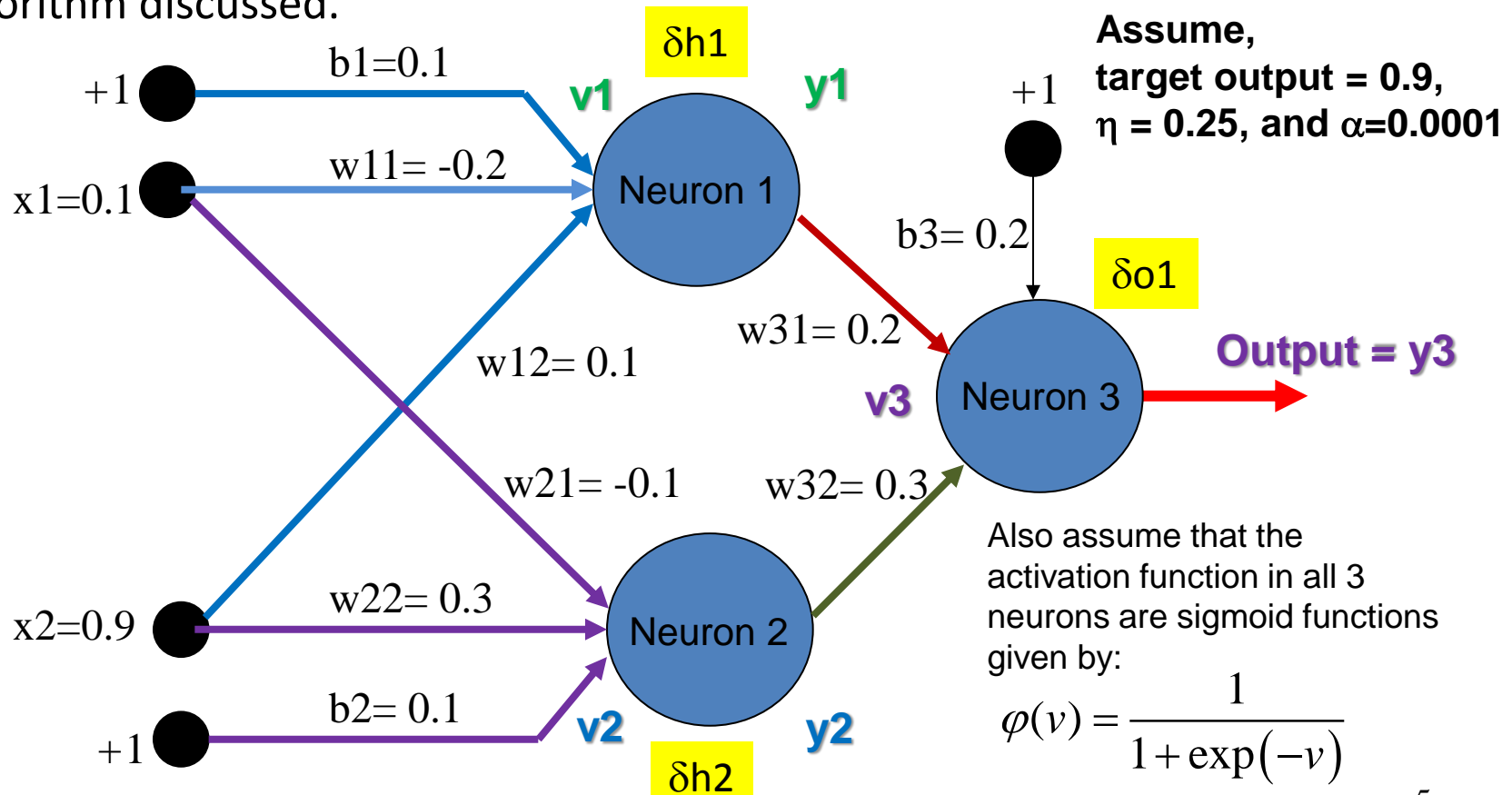
## Back-Propagation Algorithm

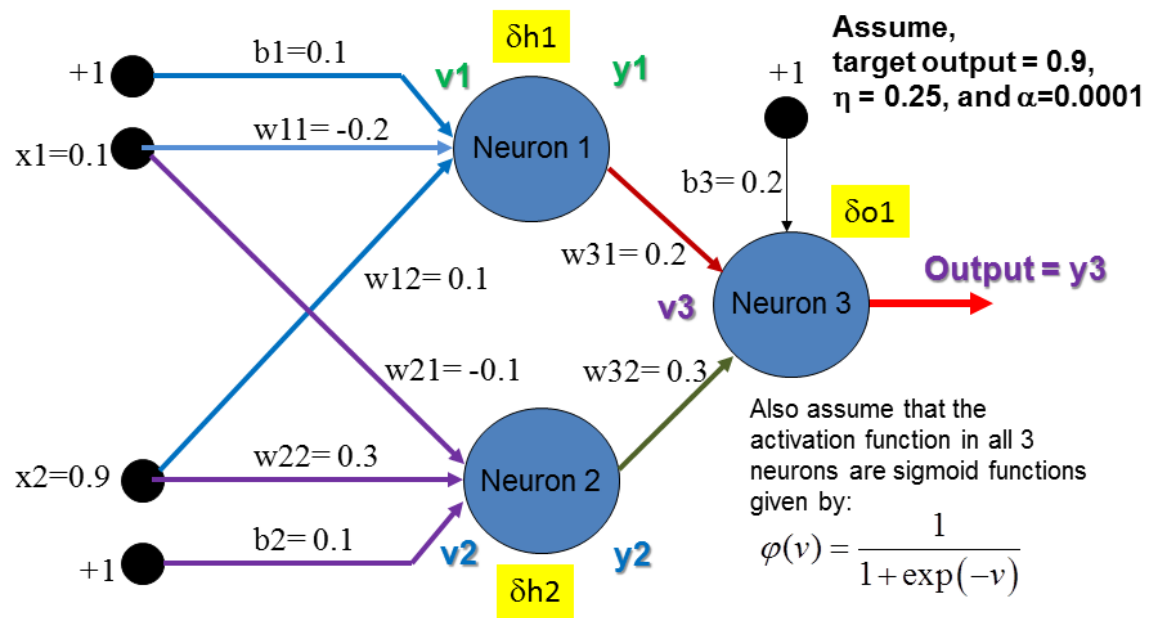


- **Step 2:** Adjust the weights of the network using the learning rule:
- $w(n + 1) = w(n) + \alpha * w(n - 1) + \eta * \delta(n) * y$
- $w_{31}(n + 1) = w_{31}(n) + \alpha * w_{31}(n - 1) + \eta * \delta o1(n) * y1$
- $w_{41}(n + 1) = w_{41}(n) + \alpha * w_{41}(n - 1) + \eta * \delta o2(n) * y1$
- $w_{32}(n + 1) = w_{32}(n) + \alpha * w_{32}(n - 1) + \eta * \delta o1(n) * y2$
- $w_{42}(n + 1) = w_{42}(n) + \alpha * w_{42}(n - 1) + \eta * \delta o2(n) * y2$
- $w_{11}(n + 1) = w_{11}(n) + \alpha * w_{11}(n - 1) + \eta * \delta h1(n) * x1$
- $w_{21}(n + 1) = w_{21}(n) + \alpha * w_{21}(n - 1) + \eta * \delta h2(n) * x1$
- $w_{12}(n + 1) = w_{12}(n) + \alpha * w_{12}(n - 1) + \eta * \delta h1(n) * x2$
- $w_{22}(n + 1) = w_{22}(n) + \alpha * w_{22}(n - 1) + \eta * \delta h2(n) * x2$
- $b3(n + 1) = b3(n) + \alpha * b3(n - 1) + \eta * \delta o1(n) * 1$
- $b4(n + 1) = b4(n) + \alpha * b4(n - 1) + \eta * \delta o2(n) * 1$
- $b1(n + 1) = b1(n) + \alpha * b1(n - 1) + \eta * \delta h1(n) * 1$
- $b2(n + 1) = b2(n) + \alpha * b2(n - 1) + \eta * \delta h2(n) * 1$

## Back-Propagation - Example

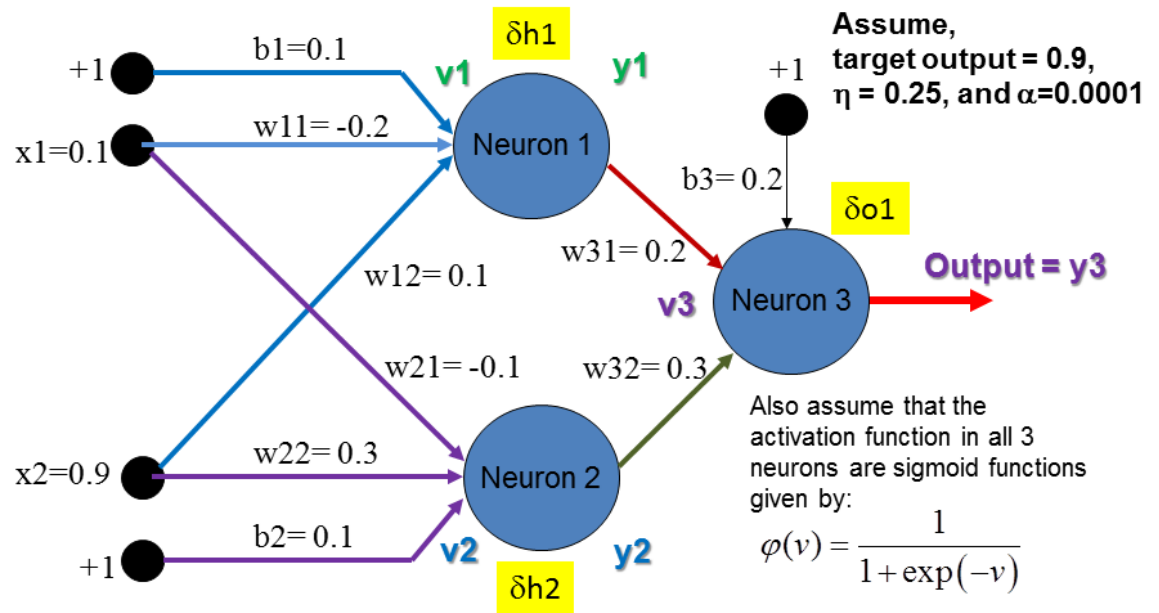
- Let's perform a complete forward and backward sweep of the feedforward network (2-2-1 architecture) shown below using the back propagation algorithm discussed.





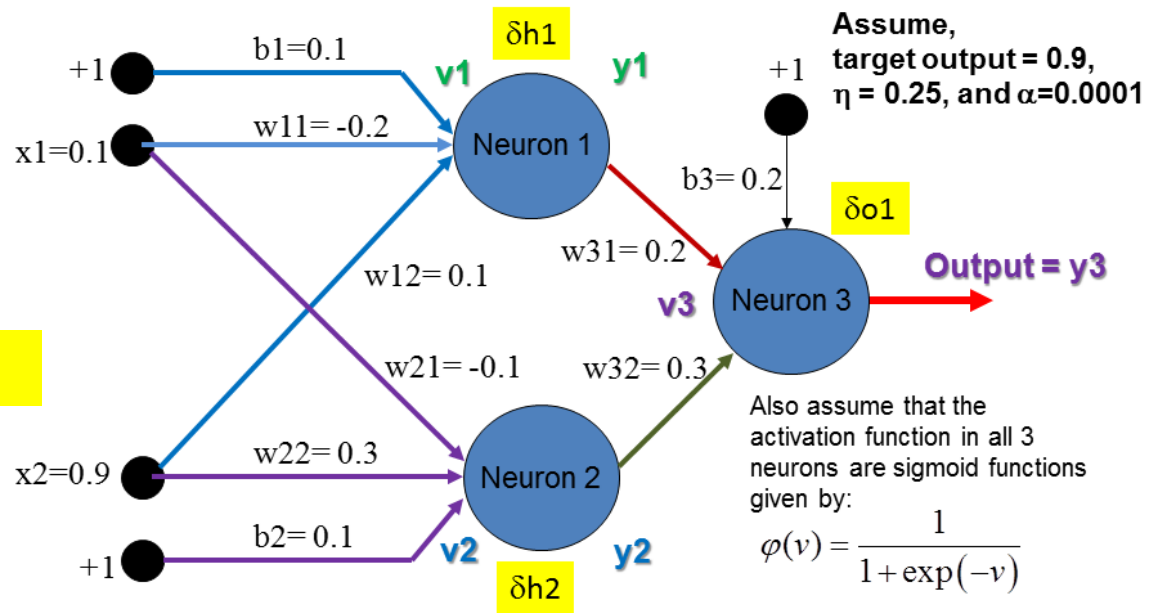
- $v_1 = 1 * b_1 + x_1 * w_{11} + x_2 * w_{12} = 1 * 0.1 + 0.1 * (-0.2) + 0.9 * 0.1 = 0.17$
- $y_1 = \phi(v_1) = \phi(0.17) = \frac{1}{1 + \exp(-0.17)} = 0.542$
- $v_2 = 1 * b_2 + x_1 * w_{21} + x_2 * w_{22} = 1 * 0.1 + 0.1 * (-0.1) + 0.9 * 0.3 = 0.36$
- $y_2 = \phi(v_2) = \phi(0.36) = \frac{1}{1 + \exp(-0.36)} = 0.589$
- $v_3 = 1 * b_3 + y_1 * w_{31} + y_2 * w_{32} = 1 * 0.2 + 0.542 * 0.2 + 0.589 * 0.3 = 0.485$
- $y_3 = \phi(v_3) = \phi(0.485) = \frac{1}{1 + \exp(-0.485)} = 0.619$
- Therefore:  $e = d_3 - y_3 = 0.9 - 0.619 = 0.281$

- $v1 = 0.17, y1 = 0.542$
- $v2 = 0.36, y2 = 0.589$
- $v3 = 0.485, y3 = 0.619$
- $e = 0.281$



- $\delta o1 = \phi'(v3) * (d3 - y3) = \phi'(0.4851) * 0.281 = \phi(0.4851)[1 - \phi(0.4851)] * 0.281 = 0.619[1 - 0.619] * 0.281 = 0.0663$
- $\delta h1 = \phi'(v1) * (\delta o1 * w31) = \phi'(0.17) * (0.0663 * 0.2) = \phi(0.17)[1 - \phi(0.17)] * 0.01362 = 0.542[1 - 0.542] * 0.01362 = 0.0033$
- $\delta h2 = \phi'(v2) * (\delta o1 * w32) = \phi'(0.36) * (0.0663 * 0.3) = \phi(0.36)[1 - \phi(0.36)] * 0.01989 = 0.589[1 - 0.589] * 0.01989 = 0.0049$

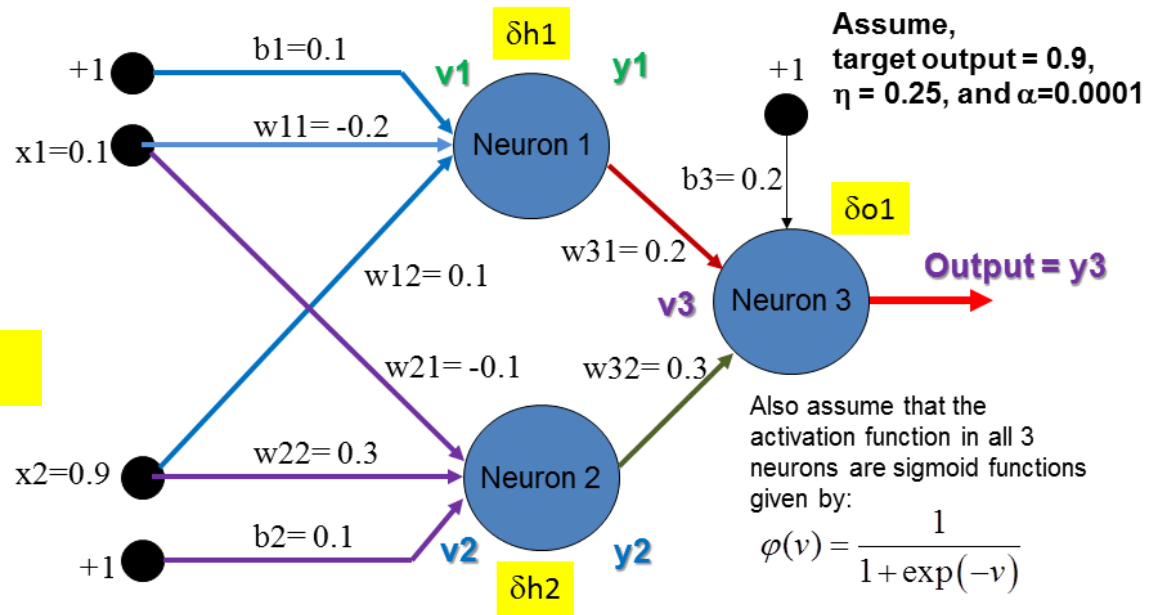
- $y1 = 0.542, y2 = 0.589$
- $\delta o1 = 0.0663,$
- $\delta h1 = 0.0033$
- $\delta h2 = 0.0049$



- **Step 2:** Adjust the weights of the network using the learning rule:
- $w(n+1) = w(n) + \alpha * w(n-1) + \eta * \delta(n) * y$
- $w31(n+1) = w31(n) + \alpha * w31(n-1) + \eta * \delta o1(n) * y1$
- $w31(n+1) = 0.2 + 0.0001 * 0.2 + 0.25 * 0.0663 * 0.542 = 0.2090$
- $w32(n+1) = w32(n) + \alpha * w32(n-1) + \eta * \delta o1(n) * y2$
- $w32(n+1) = 0.3 + 0.0001 * 0.3 + 0.25 * 0.0663 * 0.589 = 0.3098$

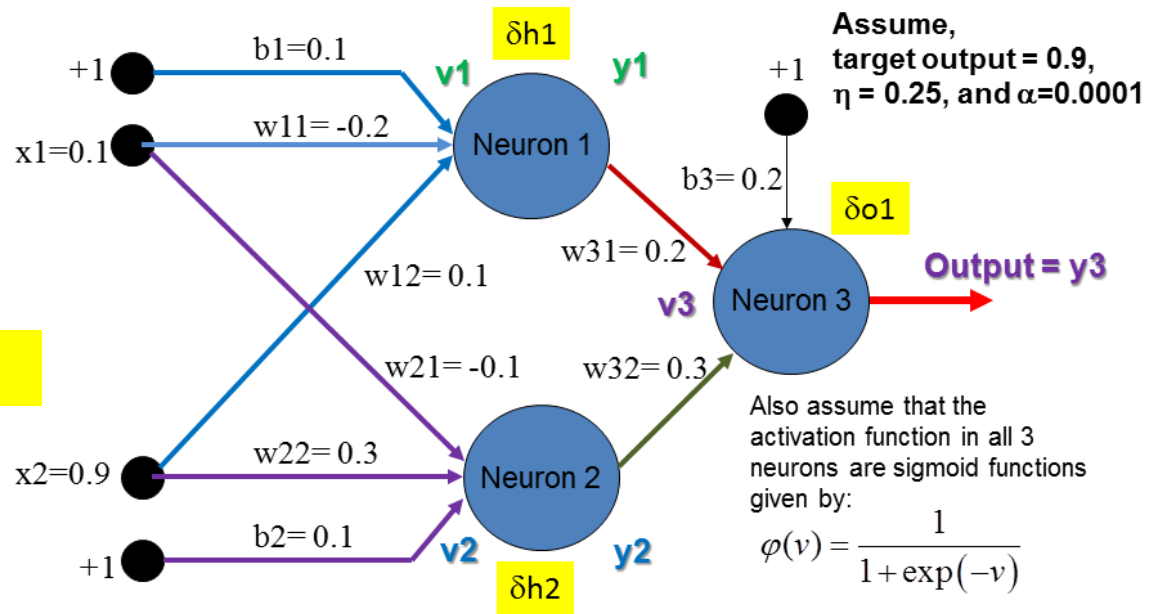


- $y1 = 0.542, y2 = 0.589$
- $\delta o1 = 0.0663,$
- $\delta h1 = 0.0033$
- $\delta h2 = 0.0049$



- **Step 2:** Adjust the weights of the network using the learning rule:
- $w(n+1) = w(n) + \alpha * w(n-1) + \eta * \delta(n) * y$
- $w11(n+1) = w11(n) + \alpha * w11(n-1) + \eta * \delta h1(n) * x1$
- $w11(n+1) = (-0.2) + 0.0001 * (-0.2) + 0.25 * 0.0033 * 0.1 = -0.1999$
- $w21(n+1) = w21(n) + \alpha * w21(n-1) + \eta * \delta h2(n) * x1$
- $w21(n+1) = (-0.1) + 0.0001 * (-0.1) + 0.25 * 0.0049 * 0.1 = -0.0999$

- $y1 = 0.542, y2 = 0.589$
- $\delta o1 = 0.0663,$
- $\delta h1 = 0.0033$
- $\delta h2 = 0.0049$

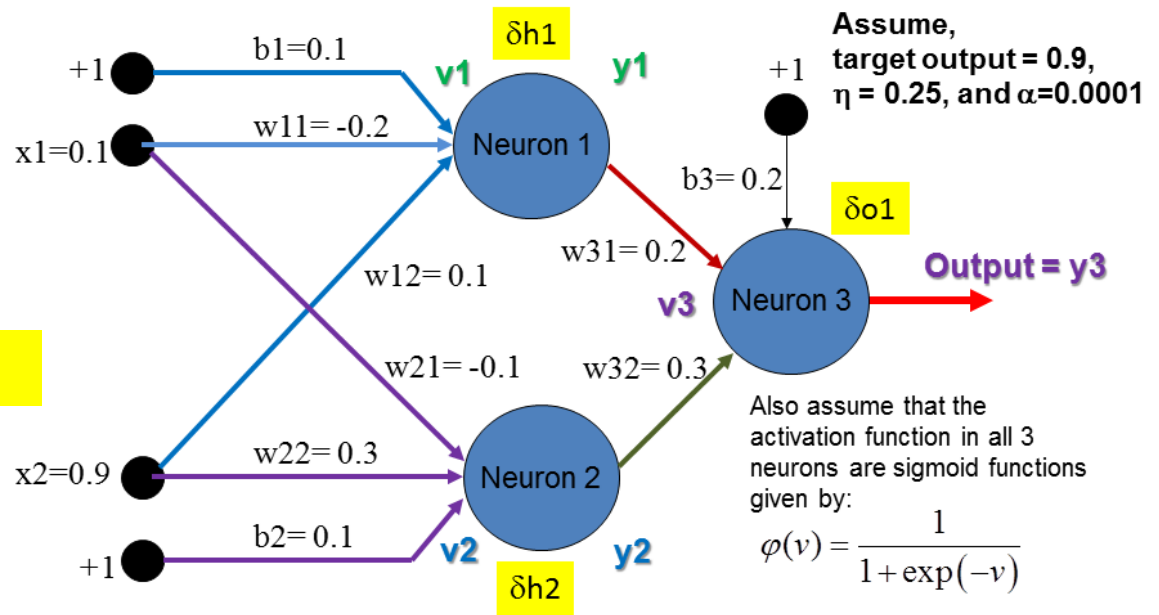


- **Step 2:** Adjust the weights of the network using the learning rule:
- $w(n+1) = w(n) + \alpha * w(n-1) + \eta * \delta(n) * y$
- $w12(n+1) = w12(n) + \alpha * w12(n-1) + \eta * \delta h1(n) * x2$
- $w12(n+1) = 0.1 + 0.0001 * 0.1 + 0.25 * 0.0033 * 0.9 = 0.1008$
- $w22(n+1) = w22(n) + \alpha * w22(n-1) + \eta * \delta h2(n) * x2$
- $w22(n+1) = 0.3 + 0.0001 * 0.3 + 0.25 * 0.0049 * 0.9 = 0.3011$

Click here to see the video

<http://youtu.be/l2I5ztVfUSE>

- $y1 = 0.542, y2 = 0.589$
- $\delta o1 = 0.0663,$
- $\delta h1 = 0.0033$
- $\delta h2 = 0.0049$



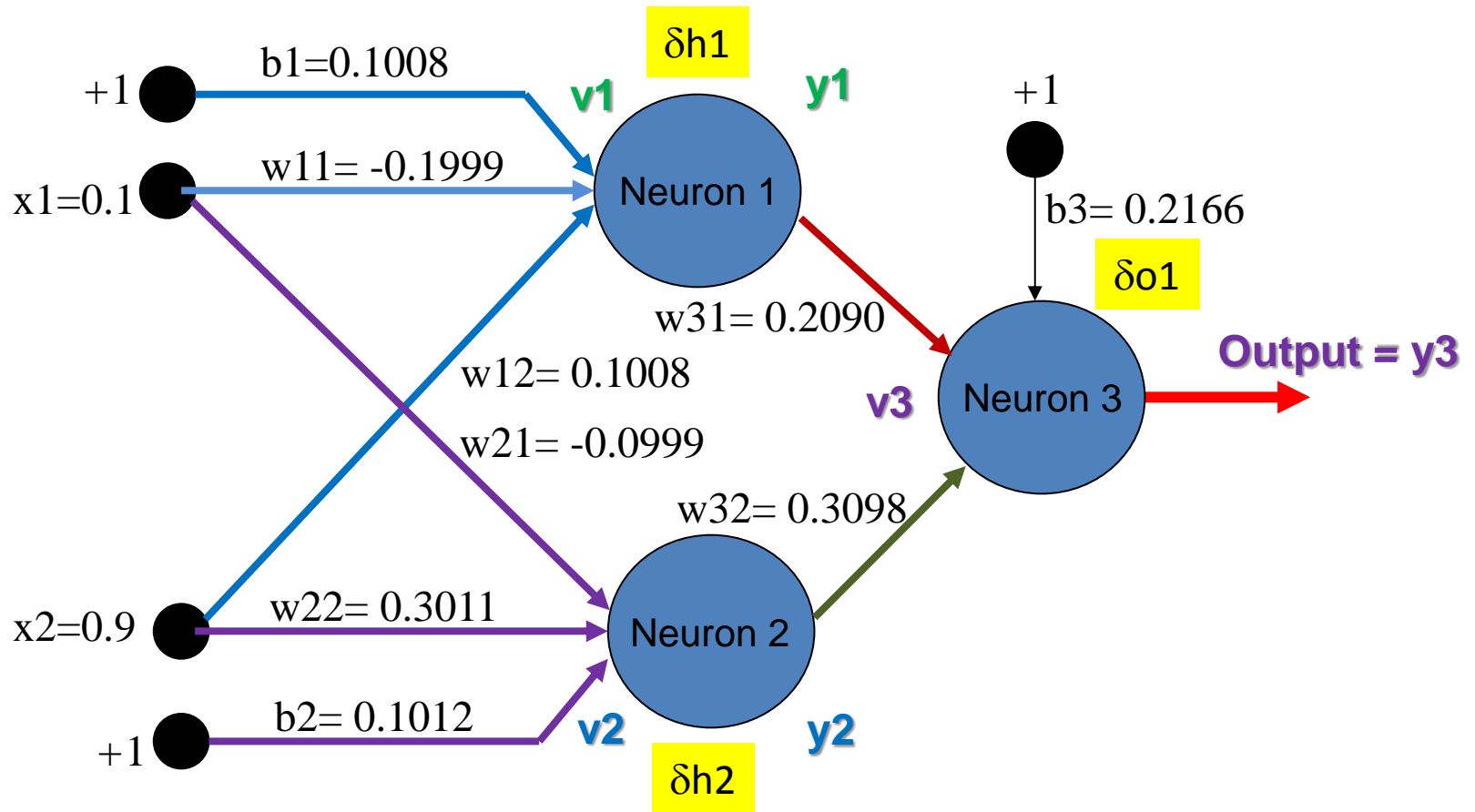
- **Step 2:** Adjust the weights of the network using the learning rule:

Click here to see the video

<http://youtu.be/I2I5ztVfUSE>

- $b(n+1) = b(n) + \alpha * b(n-1) + \eta * \delta(n) * 1$
- $b3(n+1) = b3(n) + \alpha * b3(n-1) + \eta * \delta o1(n) * 1$
- $b3(n+1) = 0.2 + 0.0001 * 0.2 + 0.25 * 0.0663 * 1 = 0.2166$
- $b1(n+1) = b1(n) + \alpha * b1(n-1) + \eta * \delta h1(n) * 1$
- $b1(n+1) = 0.1 + 0.0001 * 0.2 + 0.25 * 0.0033 * 1 = 0.1008$
- $b3(n+1) = b3(n) + \alpha * b3(n-1) + \eta * \delta 2(n) * 1$
- $b3(n+1) = 0.1 + 0.0001 * 0.1 + 0.25 * 0.0049 * 1 = 0.1012$

## After one Complete Forward & Backward Pass

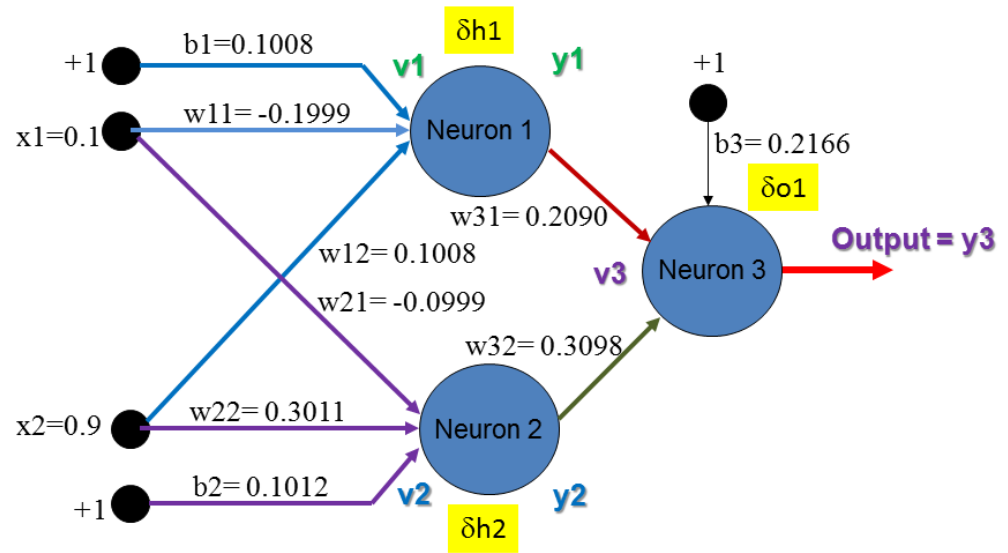


Click here to see the video

<http://youtu.be/l2I5ztVfUSE>

## After one Complete Forward & Backward Pass

- $v1 = 0.17 \rightarrow v1 = 0.1715$
- $y1 = 0.542 \rightarrow y1 = 0.5428$
- $v2 = 0.36 \rightarrow v2 = 0.3622$
- $y2 = 0.589 \rightarrow y2 = 0.5896$
- $v3 = 0.4851 \rightarrow v3 = 0.5127$
- $y3 = 0.619 \rightarrow y3 = 0.6254$
- $e = d3 - y3 = 0.9 - 0.619 = 0.281 \rightarrow e = d3 - y3 = 0.9 - 0.6254 = 0.2746$



Click here to see the video

<http://youtu.be/I2I5ztVfUSE>

## After a few more Complete Forward & Backward Passes

- After second pass  $e = 0.2683$
- After third pass  $e = 0.2623$
- After forth pass  $e = 0.2565$
- After 100 passes  $e = 0.0693$
- After 200 passes  $e = 0.0319$
- After 500 passes  $e = 0.0038$
- Error is getting reduced after each pass.

**Scholastic**  
home.video.tutor  
@gmail.com

# Artificial Neural Networks - #3

## Back Propagation

Click here to see the video

<http://youtu.be/l2I5ztVfUSE>

**END of the Book**

**If you like to see similar solutions to any Mathematics problems please**  
contact us at: [home.video.tutor@gmail.com](mailto:home.video.tutor@gmail.com) with your request.

<http://scholastictutors.webs.com>

(ANN-003)



**Scholastic**  
home.video.tutor  
@gmail.com

**Scholastic Video Book Series**

Artificial Neural Networks

Part 3

(Back Propagation)

(with English Narrations)

(END)

**Videos at:** <http://www.youtube.com/user/homevideotutor>

(<http://scholastictutors.webs.com/Scholastic-Book-NeuralNetworks-Part03-2013-10-22.pdf>)

**Scholastic**  
home.video.tutor  
@gmail.com

**©Scholastic Tutors (Sep, 2013)**

ISVT 911-0-20-131022-1