

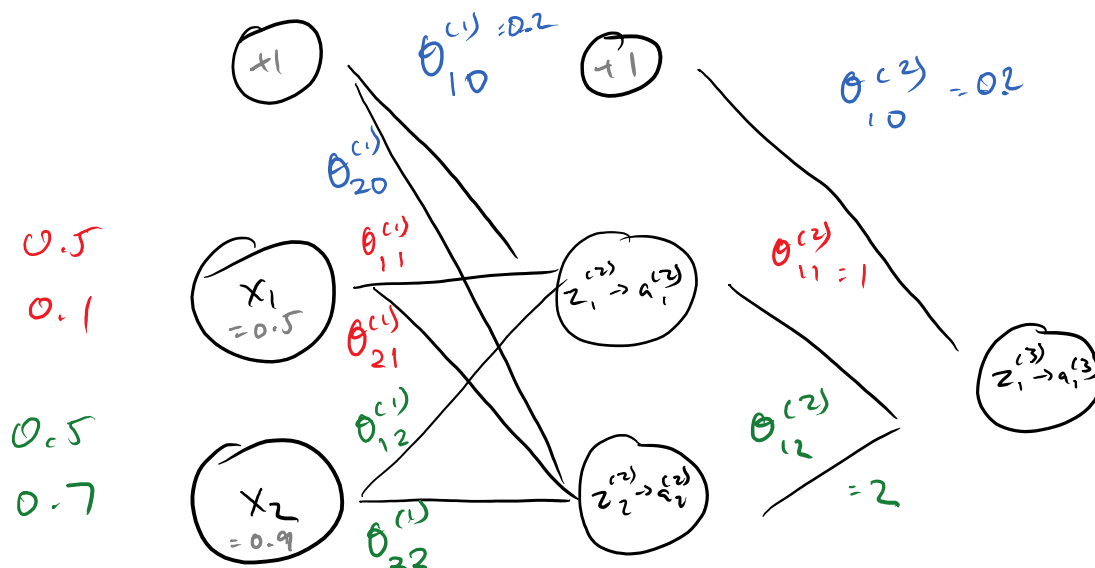
Machine Learning Assignment 3 - Julian Tsang

Friday, November 4, 2016 11:00 PM

3. (GRADED 3 points) We have a Neural Network with 1 hidden layer. Input and hidden layer both have 2 nodes. There is 1 output node. The values of theta for bias nodes are 0.2. The vector $\theta^{(1)}$ for layer 1 is: [0.5, 0.1, 0.5, 0.7] and $\theta^{(2)}$ for layer 2 is [1, 2].

1. Calculate by hand the activations of all nodes for $x_1 = 0.5$ and $x_2 = 0.9$.
2. Suppose the correct output is 1. Calculate the errors for all nodes and the updates of the weights (for 1 iteration).

1)



$$z_1^{(2)} = (\theta_{10}^{(1)} * 1) + (\theta_{11}^{(1)} * x_1) + (\theta_{12}^{(1)} * x_2) = (0.2 * 1) + (0.5 * 0.5) + (0.5 * 0.9) = 0.9$$

$$a_1^{(2)} = g(z_1^{(2)}) = \frac{1}{1 + e^{-z_1^{(2)}}} = \frac{1}{1 + e^{-0.9}} \approx 0.7109$$

$$z_2^{(2)} = (\theta_{20}^{(1)} * 1) + (\theta_{21}^{(1)} * x_1) + (\theta_{22}^{(1)} * x_2) = (0.2 * 1) + (0.1 * 0.5) + (0.7 * 0.9) = 0.88$$

$$a_2^{(2)} = g(z_2^{(2)}) = \frac{1}{1 + e^{-z_2^{(2)}}} = \frac{1}{1 + e^{-0.88}} \approx 0.70682$$

$$z_1^{(3)} = (\theta_{10}^{(2)} * 1) + (\theta_{11}^{(2)} * a_1^{(2)}) + (\theta_{12}^{(2)} * a_2^{(2)}) = (0.2 * 1) + (1 * 0.7109) + (2 * 0.70682) \approx 2.32454$$

$$a_1^{(3)} = g(z_1^{(3)}) = \frac{1}{1 + e^{-z_1^{(3)}}} = \frac{1}{1 + e^{-2.32454}} \approx 0.910889$$

- 2) Correct output = 1. Calculate all errors for nodes and update weights (theta values)

Calculate error between correct output and predicted value:

$$\delta_1^{(3)} = a_1^{(3)} - y^i = 0.910889 - 1 = -0.08911$$

Using this error, calculate errors for nodes in hidden layer:

$$\delta_1^{(2)} = \theta_{11}^{(2)} \delta_1^{(3)} * a_1^{(2)} (1 - a_1^{(2)}) = (1 * -0.08911) * 0.7109(1 - 0.7109) = -0.018314$$

$$\delta_2^{(2)} = \theta_{12}^{(2)} \delta_1^{(3)} * a_2^{(2)} (1 - a_2^{(2)}) = (2 * -0.08911) * 0.70682(1 - 0.70682) = -0.03693$$

Gradient to update theta values

Since there is only 1 iteration, $\Delta_{ij}^{(L)} = 0$

$$\Delta^{(1)} := \begin{bmatrix} a_1^{(1)} \delta_1^{(2)} & a_2^{(1)} \delta_1^{(2)} \\ a_1^{(1)} \delta_2^{(2)} & a_2^{(1)} \delta_2^{(2)} \end{bmatrix} = \begin{bmatrix} (0.5)(-0.018314) & (0.9)(-0.018314) \\ (0.5)(-0.03693) & (0.9)(-0.03693) \end{bmatrix} = \begin{bmatrix} -0.009157 & -0.0164826 \\ -0.018465 & -0.033237 \end{bmatrix}$$

$$\Delta_{ij}^{(l)} := \Delta_{ij}^{(l)} + a_j \delta_i^{(l+1)}$$

Update Weights:

$$\theta_{11}^{(1)} := \theta_{11}^{(1)} + a_1^{(1)} \delta_1^{(2)} = 0.5 - 0.009157 = 0.490843$$

$$\theta_{21}^{(1)} := \theta_{21}^{(1)} + a_1^{(1)} \delta_2^{(2)} = 0.1 - 0.018465 = 0.081535$$

$$\theta_{12}^{(1)} := \theta_{12}^{(1)} + a_2^{(1)} \delta_1^{(2)} = 0.5 - 0.0164826 = 0.4835174$$

$$\theta_{22}^{(1)} := \theta_{22}^{(1)} + a_2^{(1)} \delta_2^{(2)} = 0.7 - 0.033237 = 0.666763$$

$$\theta_{11}^{(2)} := \theta_{11}^{(2)} + a_1^{(2)} \delta_1^{(3)} = 1 - (0.7109)(-0.08911) = 1.0633483$$

$$\theta_{12}^{(2)} := \theta_{12}^{(2)} + a_2^{(2)} \delta_1^{(3)} = 2 - (0.70682)(-0.08911) = 2.06298473$$

4. (GRADED 2 point) Mitchell chapter 4 (p. 124):

In this exercise you will use a perceptron (instead of sigmoid function), that is explained in the slides of Mitchell. You can find those slides [here](#). Please refer to pages 78 and 79.

Exercise 4.1

What are the values of weights w_0 , w_1 and w_2 for the perceptron whose decision surface is illustrated in figure b on page 79?

Assume the surface crosses the x_1 axis at -1 and the x_2 axis at 2.

Exercise 4.2

(a) Design a two-input perceptron that implements the Boolean function A AND (NOT B).

(b) Design a two-layer network of perceptrons that implements A XOR B.

Exercise 4.1

A line that crosses the x_1 axis at -1 and the x_2 at 2 has the equation $x_2 = 2x_1 + 2$, which can be rearranged to: $0 = -x_2 + 2x_1 + 2$.

Since the notation is $w_0 + w_1x_1 + w_2x_2 = 0$

Then:

$$w_0 = -1$$

$$w_1 = 2$$

$$w_2 = 2$$

Exercise 4.2

a) Truth Table for A and ($NOT B$)

x_1	x_2	$h_\theta(x)$	Representation
0	0	-1	$x_0 < 0$
0	1	-1	$x_0 + x_2 < 0$
1	0	1	$x_0 + x_1 > 0$
1	1	-1	$x_0 + x_1 + x_2 < 0$

Weights:

$$x_0 = -10$$

$$x_1 = 15$$

$$x_2 = -10$$

b) XOR, exclusive OR

x_1	x_2	$a_1^{(2)}$ <i>OR</i>	$a_2^{(2)}$ <i>NOT x_1 OR NOT x_2</i>	$h_\theta(x)$ <i>AND</i>
0	0	0	1	0
0	1	1	1	1
1	0	1	1	1
1	1	1	0	0

Representation for *OR*

$x_0 < 0$
$x_0 + x_2 > 0$
$x_0 + x_1 > 0$
$x_0 + x_1 + x_2 > 0$

Weights:

$$x_0 = -5$$

$$x_1 = 10$$

$$x_2 = 15$$

Representation for *NOT x_1 OR NOT x_2*

$x_0 > 0$
$x_0 + x_2 > 0$
$x_0 + x_1 > 0$
$x_0 + x_1 + x_2 < 0$

Weights:

$$x_0 = 10$$

$$x_1 = -5$$

$$x_2 = -6$$

Representation for *AND*

$x_0 + x_2 < 0$

$x_0 + x_1 + x_2 > 0$
$x_0 + x_1 + x_2 > 0$
$x_0 + x_1 < 0$

Weights:

$$x_0 = -10$$

$$x_1 = 5$$

$$x_2 = 6$$

