

# Lecture “Advanced Data Analytics”

## Problem Set 4

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### Exercise 1

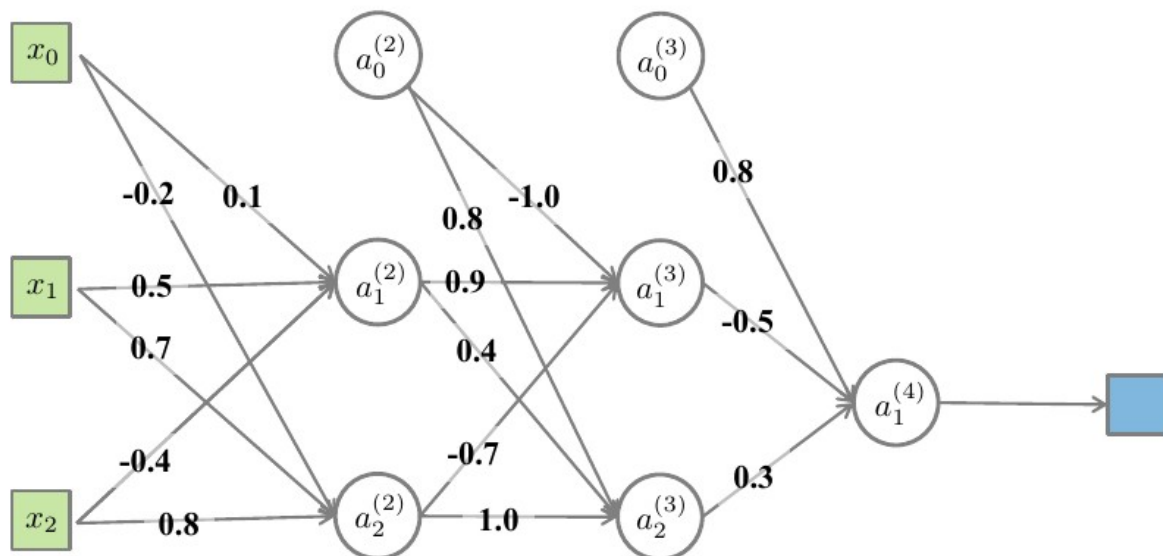
#### Artificial Neural Nets

You are given the feed-forward network (multi-layer perceptron) below, consisting of sigmoid-perceptrons, and the corresponding weights. Per default,  $x_0 = 1$  for the bias.

Write a piece of code that implements the network below with the weights specified.

Determine the output for the following inputs:

- a)  $x_1 = 0.7$ ,  $x_2 = 1.0$
- b)  $x_1 = -0.5$ ,  $x_2 = 2.5$
- c)  $x_1 = -3.0$ ,  $x_2 = 3.0$



## **Exercise 2**

### **Artificial Neural Nets**

Consider the sigmoid activation function for a node in a neural network:

$$g(a) = \frac{1}{1 + \exp(-a)}$$

Show that the derivative of this activation function takes the simple form:

$$\frac{d}{da}g(a) = g(a) (1 - g(a))$$

## **Exercise 3**

### **Artificial Neural Nets**

Apply the simple multi-layer perceptron class distributed in the class (located **data/mlp.py**) to classify the MNIST data set (located **data/mnist.pkl.gz**)

To do so, proceed as follows.

- a) Describe in few words the MNIST data set.
  - b) Describe every function in **mlp.py** with few words. What are they doing?
  - c) Implement the routine below to run a test classification on the data set.
- Play with different 3 sizes of test / training sets. What are the respective performances?

## **Exercise 4**

Apply the simple multi-layer perceptron class distributed in the class (located **data/mlp.py**) to the following dataset (taken from Anscombe's quartet):

(x1 , y1 ) = (10.0, 9.14), (x2 , y2 ) = (8.0, 8.14), (x3 , y3 ) = (13.0, 8.74),  
(x4 , y4 ) = (9.0, 8.77), (x5 , y5 ) = (11.0, 9.26), (x6 , y6 ) = (14.0, 8.10),  
(x7 , y7 ) = (6.0, 6.13), (x8 , y8 ) = (4.0, 3.10), (x9 , y9 ) = (12.0, 9.13),  
(x10 , y10 ) = (7.0, 7.26), (x11 , y11 ) = (5.0, 4.74).

- a) Use the neural network library **mlp.py** to show that a feedforward network with one hidden layer consisting of one unit and a feedforward network with no hidden layers, each using only linear activation functions, do not outperform linear regression based on ordinary least squares (OLS; for OLS, you may use the scikit library).
- b) Also demonstrate that a neural network with a hidden layer of three neurons using the tanh activation function and an output layer using the linear activation function captures the non-linearity and outperforms the linear regression.