CS 4210 – Assignment #4

### Maximum Points: 100 pts.

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**Note 1:** Your submission header must have the format as shown in the above-enclosed rounded rectangle.

**Note 2:** Homework is to be done individually. You may discuss the homework problems with your fellow students, but you are NOT allowed to copy – either in part or in whole – anyone else’s answers.

**Note 3:** Your deliverable should be a .pdf file submitted through Gradescope until the deadline. Do not forget to assign a page to each of your answers when making a submission. In addition, source code (.py files) should be added to an online repository (e.g., github) to be downloaded and executed later.

**Note 4:** All submitted materials must be legible. Figures/diagrams must have good quality.

**Note 5:** Please use and check the Canvas discussion for further instructions, questions, answers, and hints. The bold words/sentences provide information for a complete or accurate answer.

1. [30 points] Train the perceptron networks below to solve the following logical problems:
   1. [15 points] logical AND problem correctly classifying the dataset instances. Use the parameters: learning rate ɳ=0.4, initial weights = 1, and activation function = heaviside.

|  |  |  |
| --- | --- | --- |
|  |  |  |
| 0 | 0 | 0 |
| 0 | 1 | 0 |
| 1 | 0 | 0 |
| 1 | 1 | 1 |

 Datset





Table, calendar

Description automatically generated

* 1. [15 points] logical NOT problem correctly classifying the dataset instances. Use the parameters: learning rate ɳ=0.1, initial weights = 0, and activation function = heaviside.

Dataset

|  |  |
| --- | --- |
|  |  |
| 0 | 1 |
| 1 | 0 |





Graphical user interface, application

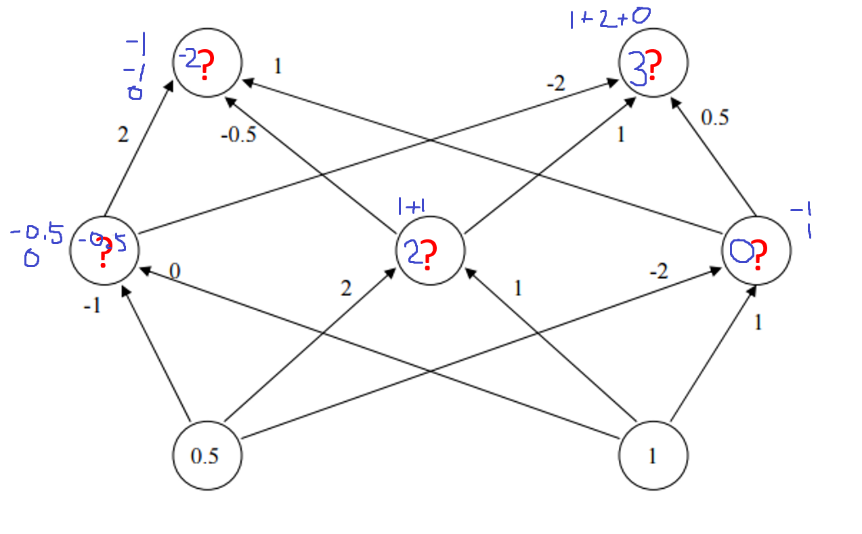
Description automatically generated

Solution format: Include the solution table (as illustrated in the lecture) with all variables and values calculated for each iteration.

1. [15 points] Complete the Python program (perceptron.py) that will read the file optdigits.tra to build a Single Layer Perceptron and a Multi-Layer Perceptron classifiers. You will simulate a grid search, trying to find which combination of two hyperparameters (learning rate and shuffle) leads you to the best prediction performance for each classifier. To test the accuracy of those distinct models, you will use the file optdigits.tes. You should update and print the accuracy of each classifier, together with the hyperparameters when it is getting higher.

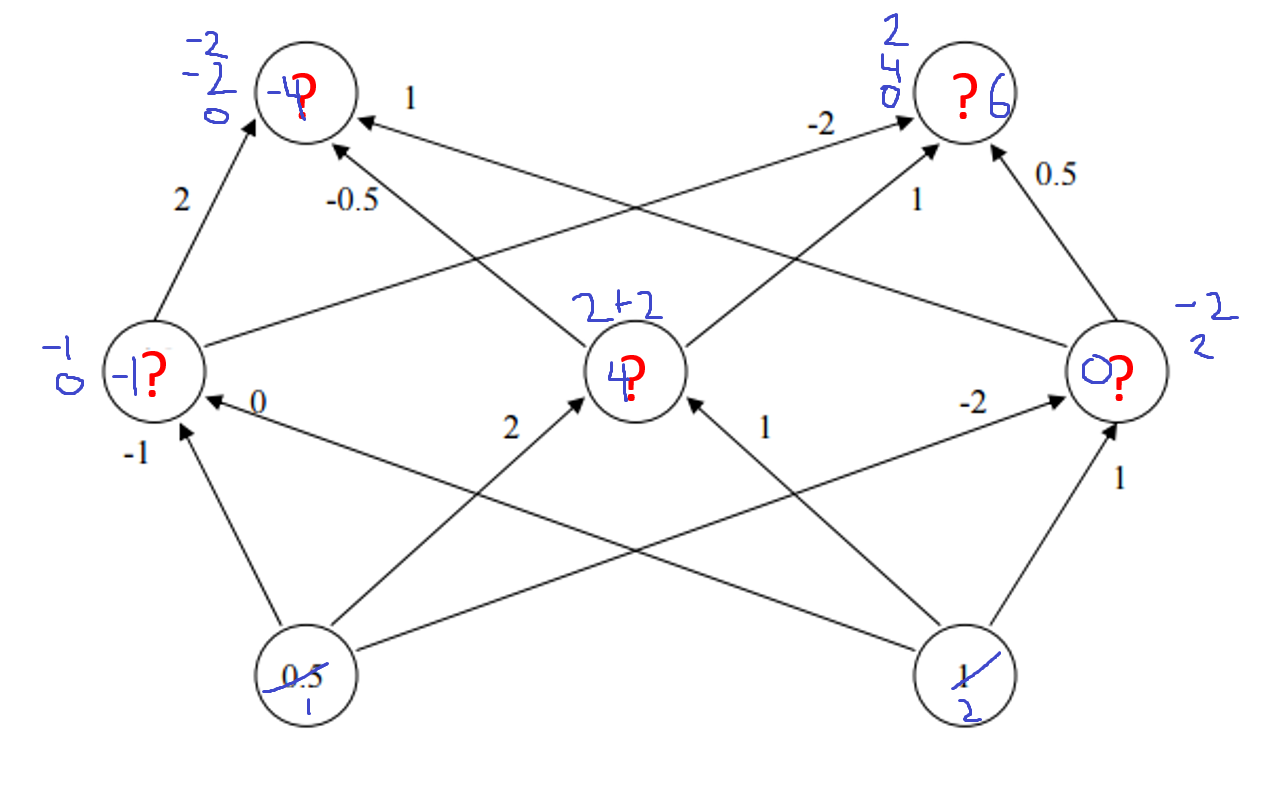
https://github.com/leolanggeng/GA\_perceptron\_DL

1. [20 points] The figure below is a network of linear neurons, that is the output of each neuron is identical to its input (linear activation function). The numbers at the connections indicate the weights of the links.
   1. [10 points] Find the value at the output nodes of the network for the given input (0.5,1).

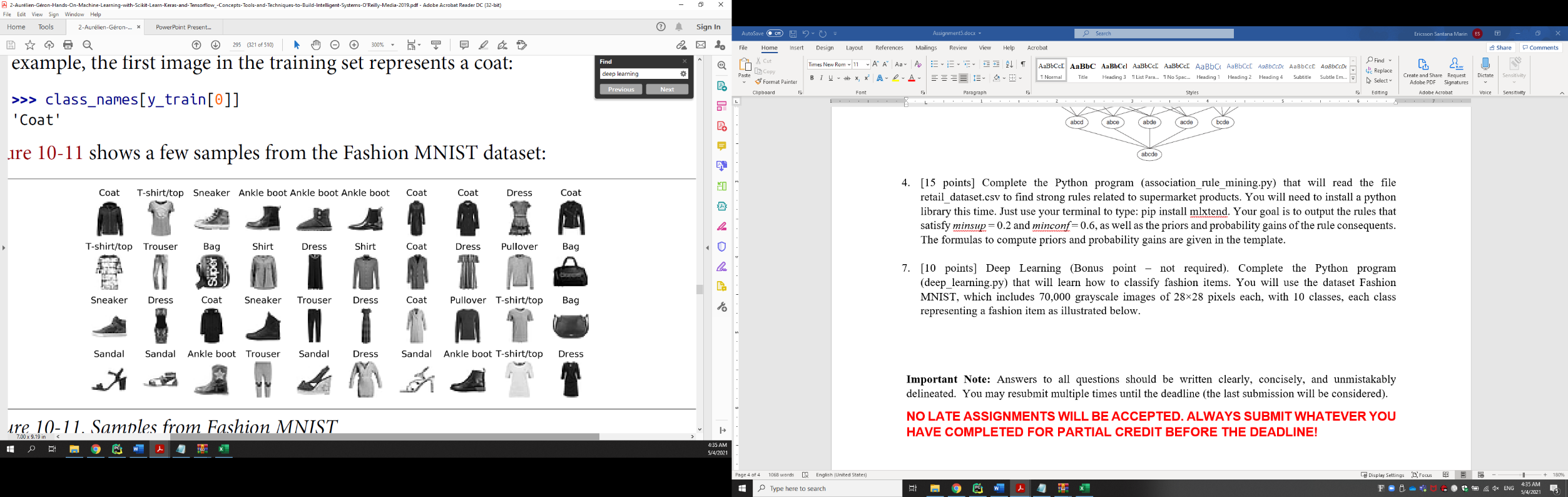


* 1. [5 points] Find the value at the output nodes of the network for the input (1,2). [5 points] Do you need to repeat the computations all over again? Why?

Yes, because the value of the nodes is coming from the previous level nodes, which also comes from the previous level, and etc. Thus, as we change the input (which is at the base), we also change all of the node value and thus need recalculate.



1. [15 points] Deep Learning. Complete the Python program (deep\_learning.py) that will learn how to classify fashion items. You will use the dataset Fashion MNIST, which includes 70,000 grayscale images of 28×28 pixels each, with 10 classes, each class representing a fashion item as illustrated below. You will use Keras to load the dataset which includes 60,000 images for training and 10,000 for test. Your goal is to train and test multiple deep neural networks and check their corresponding performances, always updating the highest accuracy found. This time you will use a separate function named build\_model() to define the architectures of your neural networks. Finally, the weights of the best model will be printed, together with the architecture and the learning curves. To install TensorFlow use: python -m pip install --upgrade tensorflow.

https://github.com/leolanggeng/GA\_perceptron\_DL

1. [20 points] Consider the dataset below.

|  |  |  |
| --- | --- | --- |
| Outlook | Temperature | PlayTennis |
| Sunny | Hot | No |
| Overcast | Cool | Yes |
| Overcast | Hot | Yes |
| Rain | Cool | No |
| Overcast | Mild | Yes |

We will apply *steady state* Genetic Algorithms ( = 0.5) to solve this classification problem, by using a similar approach of the Find-S algorithm. If the instance matches the rule pre-condition of the chromosome, you predict according to its post-condition, otherwise you predict the opposite class defined by the chromosome (there must be a prediction for each instance then). Follow the planning below to build your solution.

* Representation: single if-then rule by using bit strings (binary encoding).
  + Outlook <*Sunny, Overcast, Rain>*
  + Temperature <*Hot, Mild, Cool >*
  + Examples:
  + = *Overcast* 010
  + = *Overcast*  Rain 011
  + = *Sunny* *Overcast*  Rain 111
  + Outlook = (*Overcast* *Rain*) (Temperature = *Hot*) 011100
  + Outlook = *Sunny* Temperature = *Hot* then PlayTennis = *yes* 1001001
* Initial population (Chromosomes) : (C1=100.100.1, C2=010.010.1, C3=101.100.0, C4=110.110.0). Population size should remain the same (4 individuals) over time.
* Fitness function: accuracy
* Penalty criterion: no penalty.
* Selection method: The best two chromosomes are carried over to the next generation. The other two are selected for crossover by using the roulette wheel (simulation).
* Crossover strategy: single-point crossover with mask 1110000 in the 1st generation, two-point crossover with mask 0001100 in the 2nd generation. Use the following chromosomes to perform crossover (simulating the process of spinning the roulette wheel) according to the relative fitness (sectors of a roulette wheel), generating two offspring for each crossover:
  + (1st and 3rd) in the 1st generation
  + (1st and 2nd) in the 2nd generation
  + Mutation: on the 6th bit of the chromosome(s) 1011000 generated during the 2nd generation.
  + Termination criteria: accuracy = 1.0. **Return the corresponding chromosome(s) – your model.**

Solution format:

Fitness(C1) = 0.2 1001001

Fitness(C2) = 0.6 0100101

Fitness(C3) = 0.8 1011000

Fitness(C4) = 0.4 1101100

1st generation (C1, C2, C3, C4):

Pr(C1) = 0.1 (4th) 1001001

Pr(C4) = 0.2 (3rd) 110|1100

**Pr(C2) = 0.3 (****2nd)** 0100101

**Pr(C3) = 0.4 (1st)** 101|1000

C5 = 110|1000 → C5 = 1101000

C6 = 101|1100 → C6 = 1011100

Fitness(C2) = 0.6 0100101

Fitness(C3) = 0.8 1011000

Fitness(C5) = 0.6 1101000

Fitness(C6) = 0.8 1011100

2nd generation (C2, C3, C5, C6)

Pr(C2) = 0.2143 (4th) 0100101

Pr(C5) = 0.2143 (3rd) 1101000

**Pr(C3) = 0.2857 (2nd)** 101|10|00

**Pr(C6) = 0.2857 (1st)** 101|11|00

C7 = 101|11|00 → C7 = 1011100

C8 = 101|10|00 → C8 = 10110***1***0

Applying mutation on 1011000

Fitness(C3) = 0.8

Fitness(C6) = 0.8

Fitness(C7) = 0.8

Fitness(C8) = 1.0

Final answer: 1011010

**Important Note:** Answers to all questions should be written clearly, concisely, and unmistakably delineated. You may resubmit multiple times until the deadline (the last submission will be considered).

**NO LATE ASSIGNMENTS WILL BE ACCEPTED. ALWAYS SUBMIT WHATEVER YOU HAVE COMPLETED FOR PARTIAL CREDIT BEFORE THE DEADLINE!**