

Statistical Machine Learning
Winter 2022
Assignment - 4 [Rubric](#)
Deadline : *14th May, 11 : 59PM*

April 2022

1 Instructions

- You are free to use either python or MATLAB for this assignment.
- You can use inbuilt libraries for Math, plotting, and handling the data (eg. NumPy, Pandas, Matplotlib).
- Usage instructions for other libraries can be found in the question.
- Only (*.py) and (*.m) files should be submitted for code.
- Create a (*.pdf) report explaining your assumptions, approach, results, and any further detail asked in the question.
- You should be able to replicate your results if required.

2 Question:1 [\[4 Marks\]](#)

Use [MNIST](#) dataset and follow below instructions to solve this question.
Create Gradient Boosting Classifier from [scratch](#) with following instructions.

- As a base model, use DecisionTreeRegression(max_depth = 1) from [sklearn](#).
- Use **Number of Iterations M = 5**[\[0.6 marks for each iterations= 3 marks\]](#) and **learning rate = 0.1**
- Plot iteration-wise training and testing accuracy [1 marks](#)
- Report all the assumptions that you made, and report final testing accuracy.

3 Question:2 [4 Marks]

Use FMNIST(Fashion MNIST) dataset and follow below instructions to solve this question.

Create a feed forward Neural Network from with following instructions.

NOTE : You are allowed use **PyTorch** or **TensorFlow** and take advantage of all the functionality of these DL frameworks to implement this architecture.

Do not need to implement anything from scratch.

- Input layer must have **784** nodes and output layer have **appropriate** number of nodes. 0.5 marks
- use any number of hidden layers with any number of neurons depend on your computation power availability.0.5 marks
- Use **Multiclass Cross Entropy loss** function, **Stochastic Gradient descent(SGD)** optimizer and **random weight initialization** techniques to train the model.0.5 marks
- Report all the **hyperparameters** that you have assumed like batch size , learning rate etc.0.5 marks
- Plot epoch wise training loss[1 Marks].Report testing accuracy[0.5 marks] and classwise testing accuracy[0.5 Marks].

4 Question:3 [4 Marks]

Use MNIST dataset and follow below instructions to solve this question.

NOTE : To solve this question you are allowed to use any python packages or DL framework.

- Create a **AutoEncoder** with following instructions:[2 marks]
 - Create feed forward Neural Network as follows:[1 Marks]
 - * **Input layer** : (input = 784, output = 512, activation = ReLU)
 - * **Hidden layer** : (input = 512, output = 128, activation = ReLU)
 - * **Latent Space** : (input = 128, output = 64, activation = ReLU)
 - * **Hidden layer** : (input = 64, output = 128, activation = ReLU)
 - * **Hidden layer** : (input = 128, output = 512, activation = ReLU)
 - * **Output layer** : (input = 512, output = ?, activation = ReLU)
 - Use **appropriate** loss function, and state why you used this loss function.
 - Use **Adam** optimizer to optimize the loss function with proper learning rate.

- Use training data to train the autoencoder and plot epoch-wise loss.[1 Marks]
- After Training autoEncoder remove the decoder from the autoEncoder architecture.
- Create Classification Model called **MNIST Classification Model** with following configuration.[2 Marks]
- Use the encoder and then argument it with the following.[1 Marks]
 - **Input Layer** :(input = 64 , output = 32 , activation = ReLU)
 - **output Layer** : (input = 32 , output = ? , activation = Softmax)
- Use **MultiClass Cross Entropy loss** function and **Adam** optimizer to optimize the loss function with **appropriate** learning rate.
- Train **MNIST Classification Model** using training dataset.Plot epoch-wise training loss.
- Test using testing dataset and report accuracy and classwise accuracy.[1 Marks]

5 Question : 4[4 Marks]

Use MNIST dataset and below instructions to solve this question.

NOTE: Use of any bagging libraries are not allowed.

- Create a Bagging or Bootstrap model where the **bag size = 3** and the **base model = DecisionTreeClassifier**. [1 marks]
- Use **Majority voting** techniques for final prediction. [1 marks]
- Report accuracy [1 marks and classwise accuracy [1 marks on the testing dataset.