## **Assignment 2**

## Instructions:

- 1) Use **Python** programming. **You may use libraries**.
- 2) Handle missing data as and when required using any approach.
- 3) There are two questions, each of 50 marks. You will be submitting two python code files named as "q1.py" and "q2.py".
- 4) You will prepare a **README** file to explain how to execute your code.
- 5) You will print the outputs in a ".txt" file and also provide the plots.
- 6) All source code files, results files and documents should be kept in a folder named "roll1\_and\_roll2\_a2". Zip the folder and upload it on Moodle.
- 7) Kindly download the dataset from: " <a href="https://archive.ics.uci.edu/ml/datasets/Abalone">https://archive.ics.uci.edu/ml/datasets/Abalone</a>"

## **Question 1: Unsupervised Learning (30)**

- 1) Apply PCA (select number of components by preserving 95% of total variance). (in-built function allowed for PCA).
- 2) Plot the graph for PCA.
- 3) Using the features extracted from PCA, apply K-Means Clustering. Vary the value of K from 2 to 8. Plot the graph of K vs normalised mutual information (NMI). Report the value of K for which the NMI is maximum. (in-built function not allowed for K-Means).
- 4) Prepare a **report** including all your results.

[10+5+10+5]

## **Question 2: Supervised Learning (70)**

- 1) Normalise the data using **Standard Scalar Normalisation**. Randomly divide the Dataset into 80% for training and 20% for testing. Encode categorical variables using appropriate encoding method (in-built function not allowed for normalization, sampling and encoding).
- 2) Implement the **binary SVM classifier** using the following kernels: **Linear, Quadratic, Radial Basis function**. Report the **accuracy** for each. (in-built function allowed).
- 3) Build an MLP classifier (in-built function allowed). for the given dataset. Use stochastic gradient descent optimiser. Keep learning rate as 0.001 and batch size of 32. Vary the number of hidden layers and number of nodes in each hidden layer as follows and report the accuracy of each:
  - a. 1 hidden layer with 16 nodes
  - b. 2 hidden layers with 256 and 16 nodes respectively.
- 4) Using the best accuracy model from part 3, vary the learning rate as 0.1, 0.01, 0.001, 0.0001 and 0.00001. Plot the **learning rate vs accuracy** graph.
- 5) Use **backward elimination method** on the best model found in part 3 to select the best set of features. **Print the features**.
- 6) Apply **ensemble** learning (**max voting** technique) using SVM with quadratic, SVM with radial basis function and the best accuracy model from part 3. Report the **accuracy**.
- 7) Prepare a **report** including all your results.