The provided dataset contains 569 data instances. Each data instance has 30 features that are computed from a digitized image of a fine needle aspirate (FNA) of a breast mass. They describe the characteristics of the cell nuclei present in the image. (UCI Dataset). Each instance is classified into malignant or Benign.

Name your Jupyter notebook as **DATA240_Assignment2_Name_SJSU_ID**.

Split the dataset into 70% training and 30% test and provide the following experiments. Report the accuracy for the test set as a performance measurement for all the following tasks

- 1. Use "from sklearn.tree import DecisionTreeClassifier".
 - a) Train a DT classifier with Entropy (C1) and GINI (C2) and compare the performance.
 - b) Visualize the C1 and C2 by using the "graphviz" library
 - c) Prune C1 and C2 by limiting the depth and compare their performance with the unpruned versions.
 - d) Use depth 1,...,20 and plot the performance for C1 and C2 separately.
 - e) Choose the best value for depth and visualize C1 and C2.
- 2. Use "from sklearn.ensemble import RandomForestClassifier".
 - a) Train an RF classifier with 10 estimators and compare the performance for the test set with C1.
 - b) Change the number of estimators from 10,50,100,500, 1000, and plot the performance.
 - c) Perform 5 fold cross-validation and report the performance for RF classifier with 50 estimators
 - d) Plot the feature importance for RF with 200 estimators using the mean decrease in impurity and also feature permutation and explain the plots.
- 3. Use "from sklearn.ensemble import AdaBoostClassifier"
 - a) Train a classifier with 10 estimators and compare the performance with C1 and RF in 2a.
 - b) Change the number of estimators from 10,50,100,500, 1000, and plot the performance.
 - c) Perform 5 fold cross-validation and report the performance for classifier with 50 estimators
- 4. Use "from sklearn.naive_bayes import GaussianNB"
 - a) Train a classifier and compare the performance for the test set with C1 and 2a and 3a.
- 5. Use PCA and print the Cumulative proportion. Using Cumulative proportion, only keep the features that account for more than 95% (ratio of variance to keep) of the total variation associated with all the original variables.

