# Pattern Recognition and Machine Learning 2022 Winter Semester

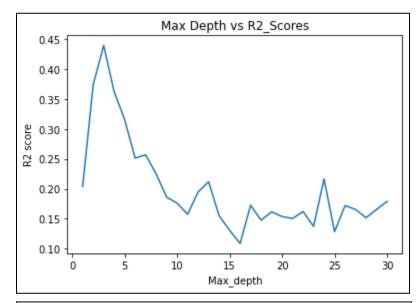
## Report - Lab Assignment - 3

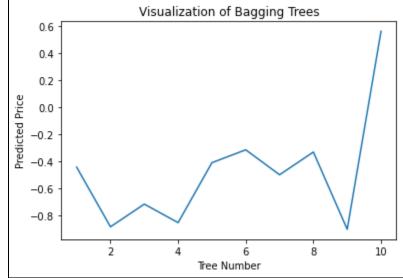
#### **Question - 1**

- Firstly, I Imported the necessary data manipulation libraries pandas and numpy.
- Then I created a DataFrame from the given Housing.csv file.
- Then, I created a visual plot of the data using the seaborn library.
- Then, I checked the data for null values and dropped an entire row if found any.
- After that, I encoded the features with object dtypes and updated the current DataFrame df.
- Afterwards, I splitted the data into Training and Testing set in a 70:30 ratio.
- Then, I used a DecisionTreeRegressor Class of sklearn.tree module to predict the labels and found out the accuracy as 47.37%.
- Then, after performing 5 fold cross-validation, I calculated the r2\_scores at different max\_depths and created a plot of max\_depth vs the scores obtained corresponding to them.
- I found out the best max\_depth = 3 for which the r2\_score was maximum.
- Then, I created a bagging function with n\_estimators = 10, that means it would create 10 different trees.
- In the bagging function, all the 10 different trees made different predictions and then I printed out the R2 Scores of all the different trees separately.
- Also, I plotted the graph of the tree number and its corresponding R2\_score.
- After that, the average R2 score of all the different trees came out to be -0.47.
- Then, I created a function which combined all the trees into one using their weighted averages and I finally got the final predictions of the combined tree.
- The final R2\_score of the combined trees came out to be 0.039.
- When I increased and decreased the max\_depth by 2 units of earlier, I found out the R2 Scores as -0.002 and 0.0566 respectively.
- Then, I used the RandomForestRegressor Class of sklearn.ensemble module to make the label predictions and found out the following:
  - Mean Squared Error 1276332056747.2378
  - Mean Absolute Error 829701.5955
  - ❖ R2 Score 0.63205

- Then, I used the AdaBoostRegressor Class of sklearn.ensemble module to make the label predictions and found out the following:
  - Mean Squared Error 1629098192592.5369
  - ❖ Mean Absolute Error 974926.992
  - ❖ R2 Score 0.5303

### **Some Plots:**





#### **Question - 2**

- Firstly, I Imported the necessary data manipulation libraries pandas and numpy.
- Also, I installed the xgboost and lightGBM libraries using
  - !pip install xgboost
  - !pip install lightgbm
- Then, I created a DataFrame df2 from the data "Breast\_cancer\_data.csv" and visualized the data using seaborn library.
- Then, I checked for the null values in the Dataframe and dropped the corresponding row if I found any null.
- Afterwards, I splitted the data into Training and Testing set in a 70:30 ratio.
- Then, I used a DecisionTreeClassifier Class of sklearn.tree module to predict the labels and found out the accuracy as 92.3976 %.
- Then, after performing 5 fold cross-validation, I calculated the r2\_scores at different max\_depths and created a plot of max\_depth vs the scores obtained corresponding to them.
- I found out the best max\_depth = 4 for which the r2\_score was maximum.
- Then, I implemented xgboost classifier with subsample = 0.7 and max\_depth = 4 and predicted the labels keeping n\_estimators = 10 and found out the accuracy score on the training data as 0.8919 and 0.9356 on the testing set.
- Similarly, I implemented the LightGBM classifier with max\_depth = 3 and predicted the models for different values of num leaves from 2 to 14.
- I found out that at fixed max\_depth = 3 and num\_leaves = 8, the model started showing overfitting and accuracy score didn't change further.
- Then, I fixed num\_leaves = 8 and checked the accuracy for different values of max\_depth from 2 to 14 and found out that at a maximum depth of 6 and for num leaves = 8, the model started showing overfitting.
- Also, the parameters such as min\_data\_in\_leaf, max\_depth and num\_leaves
  with proper tuning can give us a model with better accuracy and can avoid
  overfitting to a great extent.

