ML_5_Assignment

March 18, 2019

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In [1]: """In this assignment students will build the random forest model after normalizing th
       variable to house pricing from boston data set."""
        import numpy as np
        import pandas as pd
        import matplotlib.pyplot as plt
        import seaborn as sns
       from sklearn.model_selection import train_test_split
       from sklearn import datasets
       boston = datasets.load_boston()
       features = pd.DataFrame(boston.data,columns = boston.feature_names)
       targets = boston.target
       features.head(5)
Out[1]:
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       0 0.00632 18.0
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          PTRATIO
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             15.3 396.90
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             17.8 396.90
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       2
             17.8 392.83
                            4.03
       3
             18.7
                   394.63
                            2.94
       4
             18.7 396.90
                            5.33
In [2]: # The target column contains the prices which we are going to predict.
        # Lets add the target data to the dataframe as a new column = 'Price'
       features['Price'] = boston.target
       features.head(5)
Out[2]:
                                                    AGE
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       0 0.00632 18.0
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       2 0.02729
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       3 0.03237 0.0
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0.0 0.458 7.147 54.2 6.0622 3.0 222.0
       4 0.06905 0.0
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                        B LSTAT Price
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                                   24.0
       1
             17.8 396.90 9.14
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             17.8 392.83
                          4.03
                                   34.7
       3
             18.7 394.63 2.94
                                   33.4
             18.7 396.90
                          5.33
       4
                                   36.2
In [3]: # Splitting into train and test parts
       X = features.drop(['Price'],axis = 1).values
       y = features['Price'].values
       X_train,X_test,y_train,y_test = train_test_split(X,y,test_size = 0.30,random_state = 4
In [4]: # Apply standard Scaling
       from sklearn.preprocessing import StandardScaler
       sc_X = StandardScaler()
       X_train = sc_X.fit_transform(X_train)
       X_test = sc_X.fit_transform(X_test)
In [5]: # Applying the model
       from sklearn.ensemble import RandomForestRegressor
       regressor = RandomForestRegressor()
       regressor.fit(X_train,y_train)
       y_pred = regressor.predict(X_test)
C:\ProgramData\Anaconda3\lib\site-packages\sklearn\ensemble\forest.py:246: FutureWarning: The
  "10 in version 0.20 to 100 in 0.22.", FutureWarning)
In [6]: # Getting the r2 score
       from sklearn.metrics import r2_score
        score = r2_score(y_test,y_pred)
        score
Out[6]: 0.8182031508239296
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