```
In this assignment students will build the random forest model after normalizing the variable to house pricing from boston data set.
In [1]: 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.preprocessing import StandardScaler
         from sklearn import datasets
         boston = datasets.load_boston()
          features = pd.DataFrame(boston.data,columns=boston.feature_names)
         targets = boston.target
         C:\Users\Rahul kashyap\Anaconda3\lib\site-packages\statsmodels\tools\_testing.py:19: FutureWarning: pandas.util.testing is depr
         ecated. Use the functions in the public API at pandas.testing instead.
           import pandas.util.testing as tm
In [2]: features.head()
Out[2]:
                     ZN INDUS CHAS NOX RM AGE
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                                                                                    B LSTAT
                                                                           15.3 396.90
          0 0.00632 18.0
                           2.31
                                  0.0 0.538 6.575 65.2 4.0900 1.0 296.0
                                                                                        4.98
                                                                           17.8 396.90
          1 0.02731
                    0.0
                                  0.0 0.469 6.421 78.9 4.9671 2.0 242.0
                                                                                        9.14
                           7.07
                                                                           17.8 392.83
                     0.0
                                  0.0 0.469 7.185 61.1 4.9671 2.0 242.0
                                                                                        4.03
          2 0.02729
                           7.07
                                                                           18.7 394.63
          3 0.03237
                     0.0
                           2.18
                                  0.0 0.458 6.998 45.8 6.0622
                                                             3.0 222.0
                                                                                        2.94
                                  0.0 0.458 7.147 54.2 6.0622 3.0 222.0
          4 0.06905 0.0 2.18
                                                                           18.7 396.90
                                                                                        5.33
In [3]: targets
Out[3]: array([24., 21.6, 34.7, 33.4, 36.2, 28.7, 22.9, 27.1, 16.5, 18.9, 15.,
                18.9, 21.7, 20.4, 18.2, 19.9, 23.1, 17.5, 20.2, 18.2, 13.6, 19.6,
                15.2, 14.5, 15.6, 13.9, 16.6, 14.8, 18.4, 21. , 12.7, 14.5, 13.2,
                13.1, 13.5, 18.9, 20., 21., 24.7, 30.8, 34.9, 26.6, 25.3, 24.7,
                21.2, 19.3, 20., 16.6, 14.4, 19.4, 19.7, 20.5, 25., 23.4, 18.9,
                 35.4, 24.7, 31.6, 23.3, 19.6, 18.7, 16., 22.2, 25., 33., 23.5,
                 19.4, 22. , 17.4, 20.9, 24.2, 21.7, 22.8, 23.4, 24.1, 21.4, 20. ,
                 20.8, 21.2, 20.3, 28., 23.9, 24.8, 22.9, 23.9, 26.6, 22.5, 22.2,
                 23.6, 28.7, 22.6, 22. , 22.9, 25. , 20.6, 28.4, 21.4, 38.7, 43.8,
                 33.2, 27.5, 26.5, 18.6, 19.3, 20.1, 19.5, 19.5, 20.4, 19.8, 19.4,
                21.7, 22.8, 18.8, 18.7, 18.5, 18.3, 21.2, 19.2, 20.4, 19.3, 22.
                20.3, 20.5, 17.3, 18.8, 21.4, 15.7, 16.2, 18., 14.3, 19.2, 19.6,
                23. , 18.4, 15.6, 18.1, 17.4, 17.1, 13.3, 17.8, 14. , 14.4, 13.4,
                15.6, 11.8, 13.8, 15.6, 14.6, 17.8, 15.4, 21.5, 19.6, 15.3, 19.4,
                17. , 15.6, 13.1, 41.3, 24.3, 23.3, 27. , 50. , 50. , 50. , 22.7,
                25., 50., 23.8, 23.8, 22.3, 17.4, 19.1, 23.1, 23.6, 22.6, 29.4,
                 23.2, 24.6, 29.9, 37.2, 39.8, 36.2, 37.9, 32.5, 26.4, 29.6, 50.
                 32., 29.8, 34.9, 37., 30.5, 36.4, 31.1, 29.1, 50., 33.3, 30.3,
                 34.6, 34.9, 32.9, 24.1, 42.3, 48.5, 50., 22.6, 24.4, 22.5, 24.4,
                20. , 21.7, 19.3, 22.4, 28.1, 23.7, 25. , 23.3, 28.7, 21.5, 23. ,
                26.7, 21.7, 27.5, 30.1, 44.8, 50., 37.6, 31.6, 46.7, 31.5, 24.3,
                 31.7, 41.7, 48.3, 29., 24., 25.1, 31.5, 23.7, 23.3, 22., 20.1,
                22.2, 23.7, 17.6, 18.5, 24.3, 20.5, 24.5, 26.2, 24.4, 24.8, 29.6,
                 42.8, 21.9, 20.9, 44., 50., 36., 30.1, 33.8, 43.1, 48.8, 31.,
                 36.5, 22.8, 30.7, 50., 43.5, 20.7, 21.1, 25.2, 24.4, 35.2, 32.4,
                 32. , 33.2, 33.1, 29.1, 35.1, 45.4, 35.4, 46. , 50. , 32.2, 22. ,
                20.1, 23.2, 22.3, 24.8, 28.5, 37.3, 27.9, 23.9, 21.7, 28.6, 27.1,
                20.3, 22.5, 29., 24.8, 22., 26.4, 33.1, 36.1, 28.4, 33.4, 28.2,
                22.8, 20.3, 16.1, 22.1, 19.4, 21.6, 23.8, 16.2, 17.8, 19.8, 23.1,
                21. , 23.8, 23.1, 20.4, 18.5, 25. , 24.6, 23. , 22.2, 19.3, 22.6,
                19.8, 17.1, 19.4, 22.2, 20.7, 21.1, 19.5, 18.5, 20.6, 19., 18.7,
                 32.7, 16.5, 23.9, 31.2, 17.5, 17.2, 23.1, 24.5, 26.6, 22.9, 24.1,
                18.6, 30.1, 18.2, 20.6, 17.8, 21.7, 22.7, 22.6, 25., 19.9, 20.8,
                16.8, 21.9, 27.5, 21.9, 23.1, 50., 50., 50., 50., 50., 13.8,
                13.8, 15., 13.9, 13.3, 13.1, 10.2, 10.4, 10.9, 11.3, 12.3, 8.8,
                 7.2, 10.5, 7.4, 10.2, 11.5, 15.1, 23.2, 9.7, 13.8, 12.7, 13.1,
                12.5, 8.5, 5., 6.3, 5.6, 7.2, 12.1, 8.3, 8.5, 5., 11.9,
                27.9, 17.2, 27.5, 15., 17.2, 17.9, 16.3, 7., 7.2, 7.5, 10.4,
                 8.8, 8.4, 16.7, 14.2, 20.8, 13.4, 11.7, 8.3, 10.2, 10.9, 11.
                 9.5, 14.5, 14.1, 16.1, 14.3, 11.7, 13.4, 9.6, 8.7, 8.4, 12.8,
                10.5, 17.1, 18.4, 15.4, 10.8, 11.8, 14.9, 12.6, 14.1, 13. , 13.4,
                15.2, 16.1, 17.8, 14.9, 14.1, 12.7, 13.5, 14.9, 20. , 16.4, 17.7,
                19.5, 20.2, 21.4, 19.9, 19. , 19.1, 19.1, 20.1, 19.9, 19.6, 23.2,
                29.8, 13.8, 13.3, 16.7, 12. , 14.6, 21.4, 23. , 23.7, 25. , 21.8,
                20.6, 21.2, 19.1, 20.6, 15.2, 7., 8.1, 13.6, 20.1, 21.8, 24.5,
                23.1, 19.7, 18.3, 21.2, 17.5, 16.8, 22.4, 20.6, 23.9, 22. , 11.9])
In [4]: features.describe()
 Out[4]:
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                  3.613524
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                  3.677083
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                                                                                        12.126500
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           max
         features.dtypes
 Out[5]:
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         INDUS
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          CHAS
                     float64
          NOX
                     float64
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                     float64
         AGE
         DIS
                     float64
         RAD
                     float64
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          TAX
                     float64
         PTRATIO
                     float64
         LSTAT
                     float64
         dtype: object
In [7]: from sklearn.ensemble import RandomForestClassifier
 In [8]: rand_clf = RandomForestClassifier(random_state=6)
 In [9]: from sklearn.ensemble import RandomForestRegressor
In [10]: features.describe()
Out[10]:
                                                                                                                       PTRATIO
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                                     11.136779
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                                                                     6.284634
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          mean
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           75%
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                 88.976200 100.000000
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In [11]: import seaborn as sns
In [12]: import seaborn as sns
         plt.figure(figsize=(20,25), facecolor='white')
         plotnumber = 1
         for column in features:
             if plotnumber<=15 :</pre>
                  ax = plt.subplot(4,4,plotnumber)
                  sns.distplot(features[column])
                  plt.xlabel(column, fontsize=20)
                  #plt.ylabel('Salary',fontsize=20)
             plotnumber+=1
         plt.tight_layout()
          0.40
                                                                                                             20.0
                                                                            0.08
          0.35
                                                                                                             17.5
                                           0.20
          0.30
                                                                                                             15.0
                                                                            0.06
                                           0.15
          0.25
                                                                                                             12.5
          0.20
                                                                                                             10.0
                                                                            0.04
                                           0.10
                                                                                                              7.5
          0.05
                                                                                          INDUS
                                                           ZN
                        CRIM
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                                            0.5
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                                            0.3
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                                            0.1
                                0.8 0.9
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                                                          RM
                         NOX
                                                                                           AGE
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          0.10
                                          0.0025
                                          0.0020
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                                          0.0010
          0.02 -
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                        LSTAT
In [14]: scalar = StandardScaler()
         x_transfrom = scalar.fit_transform(features)
In [15]: from sklearn.model_selection import train_test_split,GridSearchCV
In [16]: targets
Out[16]: array([24., 21.6, 34.7, 33.4, 36.2, 28.7, 22.9, 27.1, 16.5, 18.9, 15.,
                18.9, 21.7, 20.4, 18.2, 19.9, 23.1, 17.5, 20.2, 18.2, 13.6, 19.6,
                15.2, 14.5, 15.6, 13.9, 16.6, 14.8, 18.4, 21. , 12.7, 14.5, 13.2,
                13.1, 13.5, 18.9, 20., 21., 24.7, 30.8, 34.9, 26.6, 25.3, 24.7,
                21.2, 19.3, 20., 16.6, 14.4, 19.4, 19.7, 20.5, 25., 23.4, 18.9,
                35.4, 24.7, 31.6, 23.3, 19.6, 18.7, 16., 22.2, 25., 33., 23.5,
                19.4, 22., 17.4, 20.9, 24.2, 21.7, 22.8, 23.4, 24.1, 21.4, 20.,
                20.8, 21.2, 20.3, 28., 23.9, 24.8, 22.9, 23.9, 26.6, 22.5, 22.2,
                23.6, 28.7, 22.6, 22. , 22.9, 25. , 20.6, 28.4, 21.4, 38.7, 43.8,
                33.2, 27.5, 26.5, 18.6, 19.3, 20.1, 19.5, 19.5, 20.4, 19.8, 19.4,
                21.7, 22.8, 18.8, 18.7, 18.5, 18.3, 21.2, 19.2, 20.4, 19.3, 22.
                20.3, 20.5, 17.3, 18.8, 21.4, 15.7, 16.2, 18., 14.3, 19.2, 19.6,
                23. , 18.4, 15.6, 18.1, 17.4, 17.1, 13.3, 17.8, 14. , 14.4, 13.4,
                15.6, 11.8, 13.8, 15.6, 14.6, 17.8, 15.4, 21.5, 19.6, 15.3, 19.4,
                17. , 15.6, 13.1, 41.3, 24.3, 23.3, 27. , 50. , 50. , 50. , 22.7,
                25. , 50. , 23.8, 23.8, 22.3, 17.4, 19.1, 23.1, 23.6, 22.6, 29.4,
                23.2, 24.6, 29.9, 37.2, 39.8, 36.2, 37.9, 32.5, 26.4, 29.6, 50.
                 32., 29.8, 34.9, 37., 30.5, 36.4, 31.1, 29.1, 50., 33.3, 30.3,
                34.6, 34.9, 32.9, 24.1, 42.3, 48.5, 50., 22.6, 24.4, 22.5, 24.4,
                20. , 21.7, 19.3, 22.4, 28.1, 23.7, 25. , 23.3, 28.7, 21.5, 23. ,
                26.7, 21.7, 27.5, 30.1, 44.8, 50., 37.6, 31.6, 46.7, 31.5, 24.3,
                31.7, 41.7, 48.3, 29., 24., 25.1, 31.5, 23.7, 23.3, 22., 20.1,
                22.2, 23.7, 17.6, 18.5, 24.3, 20.5, 24.5, 26.2, 24.4, 24.8, 29.6,
                42.8, 21.9, 20.9, 44., 50., 36., 30.1, 33.8, 43.1, 48.8, 31.,
                36.5, 22.8, 30.7, 50., 43.5, 20.7, 21.1, 25.2, 24.4, 35.2, 32.4,
                32. , 33.2, 33.1, 29.1, 35.1, 45.4, 35.4, 46. , 50. , 32.2, 22. ,
                20.1, 23.2, 22.3, 24.8, 28.5, 37.3, 27.9, 23.9, 21.7, 28.6, 27.1,
                20.3, 22.5, 29., 24.8, 22., 26.4, 33.1, 36.1, 28.4, 33.4, 28.2,
                22.8, 20.3, 16.1, 22.1, 19.4, 21.6, 23.8, 16.2, 17.8, 19.8, 23.1,
                21. , 23.8, 23.1, 20.4, 18.5, 25. , 24.6, 23. , 22.2, 19.3, 22.6,
                19.8, 17.1, 19.4, 22.2, 20.7, 21.1, 19.5, 18.5, 20.6, 19., 18.7,
                32.7, 16.5, 23.9, 31.2, 17.5, 17.2, 23.1, 24.5, 26.6, 22.9, 24.1,
                18.6, 30.1, 18.2, 20.6, 17.8, 21.7, 22.7, 22.6, 25., 19.9, 20.8,
                16.8, 21.9, 27.5, 21.9, 23.1, 50., 50., 50., 50., 50., 13.8,
                13.8, 15., 13.9, 13.3, 13.1, 10.2, 10.4, 10.9, 11.3, 12.3, 8.8,
                 7.2, 10.5, 7.4, 10.2, 11.5, 15.1, 23.2, 9.7, 13.8, 12.7, 13.1,
                12.5, 8.5, 5., 6.3, 5.6, 7.2, 12.1, 8.3, 8.5, 5., 11.9,
                27.9, 17.2, 27.5, 15., 17.2, 17.9, 16.3, 7., 7.2, 7.5, 10.4,
                 8.8, 8.4, 16.7, 14.2, 20.8, 13.4, 11.7, 8.3, 10.2, 10.9, 11.
                 9.5, 14.5, 14.1, 16.1, 14.3, 11.7, 13.4, 9.6, 8.7, 8.4, 12.8,
                10.5, 17.1, 18.4, 15.4, 10.8, 11.8, 14.9, 12.6, 14.1, 13., 13.4,
                15.2, 16.1, 17.8, 14.9, 14.1, 12.7, 13.5, 14.9, 20. , 16.4, 17.7,
                19.5, 20.2, 21.4, 19.9, 19. , 19.1, 19.1, 20.1, 19.9, 19.6, 23.2,
                29.8, 13.8, 13.3, 16.7, 12. , 14.6, 21.4, 23. , 23.7, 25. , 21.8,
                20.6, 21.2, 19.1, 20.6, 15.2, 7., 8.1, 13.6, 20.1, 21.8, 24.5,
                23.1, 19.7, 18.3, 21.2, 17.5, 16.8, 22.4, 20.6, 23.9, 22. , 11.9])
In [17]: x_train,x_test,y_train,y_test = train_test_split(x_transfrom,targets,test_size = 0.25,random_state=355)
In [18]: rgr = RandomForestRegressor()
In [19]: params={ 'max_depth':[None,5,10,15,20,30,50,70],
                      'min_samples_leaf':[1,2,5,10,15,20],
                      'min_samples_split':[2,5,10,15,20],
                      'max_features':[4,7,9,10,13]
In [20]: grid_search = GridSearchCV(estimator=rgr,
                               param_grid=params,
                               cv=5,
                              n_jobs =-1)
In [21]: grid search.fit(x train,y train)
         C:\Users\Rahul kashyap\Anaconda3\lib\site-packages\sklearn\model_selection\_search.py:814: DeprecationWarning: The default of t
         he `iid` parameter will change from True to False in version 0.22 and will be removed in 0.24. This will change numeric results
         when test-set sizes are unequal.
           DeprecationWarning)
         C:\Users\Rahul kashyap\Anaconda3\lib\site-packages\sklearn\ensemble\forest.py:245: FutureWarning: The default value of n_estima
         tors will change from 10 in version 0.20 to 100 in 0.22.
           "10 in version 0.20 to 100 in 0.22.", FutureWarning)
Out[21]: GridSearchCV(cv=5, error_score='raise-deprecating',
                       estimator=RandomForestRegressor(bootstrap=True, criterion='mse',
                                                       max_depth=None,
                                                       max_features='auto',
                                                       max_leaf_nodes=None,
                                                       min_impurity_decrease=0.0,
                                                       min_impurity_split=None,
                                                       min_samples_leaf=1,
                                                       min_samples_split=2,
                                                       min_weight_fraction_leaf=0.0,
                                                       n_estimators='warn', n_jobs=None,
                                                       oob_score=False, random_state=None,
                                                       verbose=0, warm_start=False),
                       iid='warn', n_jobs=-1,
                       param_grid={'max_depth': [None, 5, 10, 15, 20, 30, 50, 70],
                                   'max_features': [4, 7, 9, 10, 13],
                                   'min_samples_leaf': [1, 2, 5, 10, 15, 20],
                                   'min_samples_split': [2, 5, 10, 15, 20]},
                       pre_dispatch='2*n_jobs', refit=True, return_train_score=False,
                       scoring=None, verbose=0)
In [22]: RandomForestRegressor(bootstrap=True, criterion='mse', max_depth=70,
                                max_features=7, max_leaf_nodes=None,
                                min_impurity_decrease=0.0, min_impurity_split=None,
                                min_samples_leaf=1, min_samples_split=5,
                                min_weight_fraction_leaf=0.0, n_estimators=10,
                                n_jobs=None, oob_score=False, random_state=None,
                                verbose=0, warm start=False)
Out[22]: RandomForestRegressor(bootstrap=True, criterion='mse', max_depth=70,
                                max_features=7, max_leaf_nodes=None,
                                min_impurity_decrease=0.0, min_impurity_split=None,
                                min_samples_leaf=1, min_samples_split=5,
                                min_weight_fraction_leaf=0.0, n_estimators=10,
                                n_jobs=None, oob_score=False, random_state=None,
                                verbose=0, warm_start=False)
In [23]: grid_search.best_score_
Out[23]: 0.8593319527830265
In [24]: pred = grid_search.predict(x_test)
In [25]: from sklearn.metrics import accuracy_score, confusion_matrix, roc_curve, roc_auc_score
In [26]: pred
Out[26]: array([12.65083333, 17.90361111, 30.7122381 , 20.78016667, 8.03240476,
                           , 22.29516667, 19.17447619, 46.03892857, 14.52275 ,
                 22.147
                 22.38625 , 21.5262381 , 21.22092857, 19.63636111, 10.44829762,
                 31.79666667, 21.3185 , 19.49933333, 21.8724127 , 15.39655952,
                18.84897619, 23.89588889, 33.13828571, 23.38633333, 30.25133333,
                 21.99702381, 31.13416667, 35.27133333, 27.93061905, 20.69342857,
                 21.88033333, 23.823
                                        , 13.96366667, 22.50574603, 15.46866667,
                 23.82338889, 40.04303571, 19.18384127, 25.52978571, 30.33347619,
                 21.09528571, 14.10986111, 18.15900794, 15.71613889, 21.91083333,
                 20.15211905, 25.75408333, 14.06716667, 26.06371429, 18.46778571,
                 20.11814286, 19.08634127, 18.33580952, 33.11809524, 46.34985714,
                 7.47597619, 43.89107143, 26.88925 , 16.57675 , 20.81911905,
                17.14788889, 20.52295238, 8.63253571, 20.80957143, 26.36133333,
                 9.82417857, 23.6115 , 31.04642857, 20.73116667, 19.25902381,
                 16.41666667, 20.53016667, 28.9554127, 19.50514286, 20.49404762,
                 45.81486508, 14.90513889, 15.70472222, 20.38209524, 14.85891667,
                 8.18347619, 19.31644444, 31.0565 , 19.93007143, 20.1482619 ,
                 35.68476984, 11.88404762, 18.36727778, 16.12147222, 28.68304762,
                 23.09033333, 20.94358333, 32.73142857, 21.76815476, 14.8125
                 22.55783333, 19.3725 , 19.15944444, 33.17178571, 24.29940476,
                15.59383333, 31.22328571, 15.58897222, 21.81658333, 19.89916667,
                 15.69994048, 34.24752381, 15.96947619, 20.33840476, 19.46411111,
                 27.9355556, 47.81152381, 22.21766667, 26.33764286, 44.189
                 48.99935714, 11.72232143, 31.41082051, 20.33692857, 20.03466667,
                19.91284127, 10.52996429, 38.48783333, 14.74815476, 31.65032051,
                11.10752381, 36.15835714])
In [27]: y_test
Out[27]: array([15.2, 18., 30.5, 23.4, 10.2, 22.2, 21., 19.1, 50., 16.7, 20.6,
                25. , 22.6, 17.8, 8.3, 30.3, 20.5, 18.5, 29.6, 23.2, 18.4, 22.2,
                33.1, 23.8, 36., 21.4, 26.6, 34.9, 24.1, 20.7, 24.5, 25., 14.9,
                20.1, 16.4, 24.4, 35.2, 18.9, 23.5, 27., 20.3, 15.4, 15., 15.2,
                24.3, 17.1, 29.1, 14.9, 25., 20., 18.6, 27.1, 16.6, 23.6, 50.,
                 7.4, 46., 27.1, 15.2, 20.4, 14.4, 24.7, 5.6, 19.6, 24.6, 9.7,
                28.7, 23.9, 16. , 23.8, 19.4, 19.6, 24. , 19.9, 17.4, 38.7, 13.1,
                13.5, 19.9, 13.3, 7.2, 20.5, 31.1, 22.9, 20.6, 33.4, 13.9, 17.3,
                13.6, 26.7, 26.4, 22., 28.7, 15., 16.7, 22.9, 18.7, 18.8, 35.4,
                24.7, 13.1, 33.1, 14.1, 21.4, 14.1, 13.4, 34.7, 15.6, 21.7, 19.8,
                27.5, 44.8, 36.2, 22.5, 50., 50., 9.6, 36.1, 21.8, 19.3, 16.8,
                 8.8, 50., 13.8, 34.6, 5., 37.2])
In [28]: from sklearn.metrics import r2_score
In [29]: r2_score(y_test,pred)
```

Acuraccy with Hyparameter: 87.23 Acuraccy with without Hyparameter: 85.18

In [33]: print("Acuraccy with Hyparameter:" ,format(r2_score(y_test,pred)*100,'.2f')) print("-"*50) print("Acuraccy with without Hyparameter:",format(r2_score(y_test,whyp)*100,".2f"))

C:\Users\Rahul kashyap\Anaconda3\lib\site-packages\sklearn\ensemble\forest.py:245: FutureWarning: The default value of n_estima

"10 in version 0.20 to 100 in 0.22.", FutureWarning) Out[30]: RandomForestRegressor(bootstrap=True, criterion='mse', max_depth=None, max_features='auto', max_leaf_nodes=None, min_impurity_decrease=0.0, min_impurity_split=None, min_samples_leaf=1, min_samples_split=2, min_weight_fraction_leaf=0.0, n_estimators=10, n_jobs=None, oob_score=False, random_state=None, verbose=0, warm start=False) In [31]: whyp = rgr.predict(x_test) In [32]: r2_score(y_test,whyp) Out[32]: 0.8518371064711073

tors will change from 10 in version 0.20 to 100 in 0.22.

Out[29]: 0.8723187534127839

In [30]: rgr.fit(x_train,y_train)