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Data Analytics I Create a Linear Regression Model using Python/R to predict home prices using Boston Housing Dataset (https://www.kaggle.com/c/boston-housing (https://www.kaggle.com/c/boston-housing)). The Boston Housing dataset contains information about various houses in Boston through different parameters. There are 506 samples and 14 feature variables in this dataset. The objective is to predict the value of prices of the house using the given features.

```
import numpy as np
 In [ ]:
          import pandas as pd
          import matplotlib.pyplot as plt
          #Step 2: Import the Boston Housing dataset
          from sklearn.datasets import load_boston
          boston = load_boston()
In [10]: data = pd.DataFrame(boston.data)
In [28]: data.shape
Out[28]: (506, 14)
In [11]:
          data.columns = boston.feature_names
          data.head()
Out[11]:
               CRIM
                     ZN INDUS CHAS NOX
                                              RM AGE
                                                          DIS RAD
                                                                     TAX PTRATIO
                                                                                      B LSTAT
          0 0.00632
                     18.0
                                                                                           4.98
                            2.31
                                       0.538
                                             6.575
                                                   65.2
                                                        4.0900
                                                                1.0
                                                                   296.0
                                                                              15.3
                                                                                  396.90
                                   0.0
             0.02731
                            7.07
                                   0.0 0.469
                                                        4.9671
                                                                2.0
                                                                   242.0
                                                                                  396.90
                     0.0
                                             6.421
                                                                              17.8
                                                                                           9.14
          2 0.02729
                     0.0
                           7.07
                                   0.0 0.469 7.185
                                                   61.1 4.9671
                                                               2.0 242.0
                                                                             17.8 392.83
                                                                                           4.03
          3 0.03237
                     0.0
                           2.18
                                   0.0 0.458 6.998
                                                   45.8 6.0622
                                                                   222.0
                                                                              18.7
                                                                                  394.63
                                                                                           2.94
          4 0.06905
                     0.0
                           2.18
                                   0.0 0.458 7.147
                                                   54.2 6.0622
                                                               3.0 222.0
                                                                             18.7 396.90
                                                                                           5.33
In [12]: data['PRICE'] = boston.target
In [13]: data.isnull().sum()
Out[13]: CRIM
          TNDUS
                     0
          CHAS
                     0
         NOX
                     0
          RM
                     a
          ΔGF
                     0
          DIS
                     0
          RAD
                     0
          TAX
                     0
          PTRATIO
                     0
                     0
          В
          LSTAT
                     0
          PRICE
                     0
          dtype: int64
In [14]: x = data.drop(['PRICE'], axis = 1)
          y = data['PRICE']
In [16]:
         from sklearn.model_selection import train_test_split
          xtrain, xtest, ytrain, ytest =train_test_split(x, y, test_size =0.2,random_state = 0)
In [17]:
          import sklearn
          from sklearn.linear_model import LinearRegression
          lm = LinearRegression()
          model=lm.fit(xtrain, ytrain)
In [25]: lm.intercept_
Out[25]: 38.138692713393205
In [26]: lm.coef_
Out[26]: array([-1.18410318e-01, 4.47550643e-02, 5.85674689e-03, 2.34230117e+00,
                  -1.61634024e+01, 3.70135143e+00, -3.04553661e-03, -1.38664542e+00,
                  2.43784171e-01, -1.09856157e-02, -1.04699133e+00, 8.22014729e-03,
                 -4.93642452e-01)
In [18]:
          ytrain_pred = lm.predict(xtrain)
          ytest_pred = lm.predict(xtest)
In [19]: df=pd.DataFrame(ytrain_pred,ytrain)
          df=pd.DataFrame(ytest_pred,ytest)
```

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```
In [20]: from sklearn.metrics import mean_squared_error, r2_score
                mse = mean_squared_error(ytest, ytest_pred)
                mse = mean_squared_error(ytrain_pred,ytrain)
                print(mse)
                33.450708967691185
                19.330019357349375
In [21]: mse = mean_squared_error(ytest, ytest_pred)
In [23]: plt.scatter(ytrain ,ytrain_pred,c='blue',marker='o',label='Training data')
    plt.scatter(ytest,ytest_pred ,c='lightgreen',marker='s',label='Test data')
    plt.xlabel('True values')
    plt.ylabel('Predicted')
    plt.title("True value vs Predicted value")
    plt.legend(loc= 'upper left')
    #plt.hlines(y=0,xmin=0,xmax=50)
    plt.nlot()
                plt.plot()
                plt.show()
                                          True value vs Predicted value
                                 Training data
                     40
                                  Test data
                     30
                     20
                     10
                      0
                                                                                              50
                                                                               40
                                                                30
                                                        True values
```

In []: .