Assignment No. 4

Aim: 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). 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.

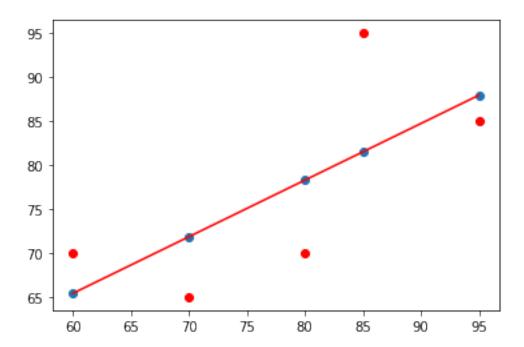
Code:

[8]: 0.4803218090889326

```
[1]: import pandas as pd
     import numpy as np
     import matplotlib.pyplot as plt
[2]: x=np.array([95,85,80,70,60])
     y=np.array([85,95,70,65,70])
[3]: model= np.polyfit(x, y, 1)
[4]: model
[4]: array([ 0.64383562, 26.78082192])
[5]: predict = np.poly1d(model)
     predict(65)
[5]: 68.63013698630137
[6]: y_pred= predict(x)
     y_pred
[6]: array([87.94520548, 81.50684932, 78.28767123, 71.84931507, 65.4109589])
[8]: from sklearn.metrics import r2_score
     r2_score(y, y_pred)
```

```
[9]: y_line = model[1] + model[0]* x
plt.plot(x, y_line, c = 'r')
plt.scatter(x, y_pred)
plt.scatter(x,y,c='r')
```

[9]: <matplotlib.collections.PathCollection at 0x2a8395862b0>



```
[10]: import numpy as np
     import pandas as pd
     import matplotlib.pyplot as plt
[11]: from sklearn.datasets import load_boston
     boston = load_boston()
[12]: data = pd.DataFrame(boston.data)
[13]: data.columns = boston.feature_names
     data.head()
[13]:
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                                              RM
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        0.00632 18.0
                        2.31
                               0.0 0.538
                                                  65.2 4.0900
                                                                1.0
                                                                     296.0
                                           6.575
     1 0.02731
                  0.0
                        7.07
                                           6.421
                                                  78.9 4.9671
                                                                2.0
                                                                     242.0
                               0.0 0.469
     2 0.02729
                  0.0
                        7.07
                               0.0 0.469
                                           7.185
                                                  61.1 4.9671
                                                                2.0
                                                                     242.0
                                                  45.8 6.0622
     3 0.03237
                  0.0
                        2.18
                               0.0 0.458
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     4 0.06905
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```

```
B LSTAT
         PTRATIO
      0
            15.3 396.90
                          4.98
            17.8 396.90
                           9.14
      1
      2
                           4.03
            17.8 392.83
      3
            18.7 394.63
                           2.94
            18.7 396.90
                           5.33
[14]: data['PRICE'] = boston.target
[15]: data.isnull().sum()
[15]: CRIM
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      иох
                 0
      RM
                 0
      AGE
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      R.AD
      TAX
      PTRATIO
                 0
     LSTAT
                 0
      PRICE
                 0
      dtype: int64
[16]: x = data.drop(['PRICE'], axis = 1)
      y = data['PRICE']
[18]: from sklearn.model_selection import train_test_split
      xtrain, xtest, ytrain, ytest = train_test_split(x, y, test_size =0.
       \hookrightarrow 2, random_state = 0)
[19]: import sklearn
      from sklearn.linear_model import LinearRegression
      lm = LinearRegression()
      model=lm.fit(xtrain, ytrain)
[24]: ytrain_pred = lm.predict(xtrain)
      ytest_pred = lm.predict(xtest)
[25]: df=pd.DataFrame(ytrain_pred,ytrain)
      df=pd.DataFrame(ytest_pred,ytest)
[26]: from sklearn.metrics import mean_squared_error, r2_score
      mse = mean_squared_error(ytest, ytest_pred)
```

```
print(mse)
mse = mean_squared_error(ytrain_pred,ytrain)
```

33.448979997676524

```
[27]: print(mse)
```

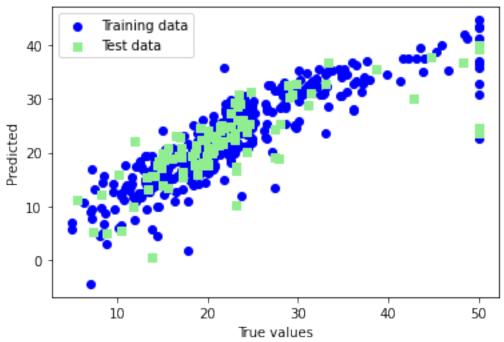
19.326470203585725

```
[]: mse = mean_squared_error(ytest, ytest_pred)
print(mse)
```

33.448979997676524

```
[]: 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.plot()
   plt.show()
```

True value vs Predicted value



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