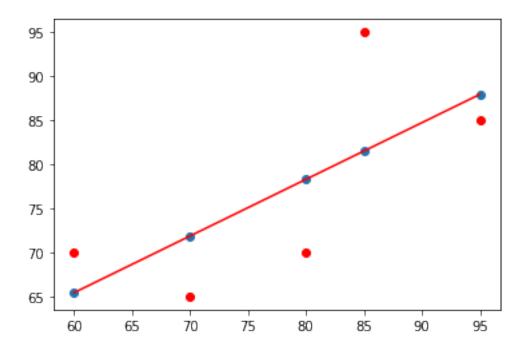
pr4-1

April 22, 2024

```
[1]: import pandas as pd
     import numpy as np
     import matplotlib.pyplot as plt
[3]: x=np.array([95,85,80,70,60])
     y=np.array([85,95,70,65,70])
[5]: model= np.polyfit(x, y, 1)
     model
[5]: array([ 0.64383562, 26.78082192])
[6]: predict = np.poly1d(model)
     predict(65)
[6]: 68.63013698630137
[7]: y_pred = predict(x)
     y_pred
[7]: array([87.94520548, 81.50684932, 78.28767123, 71.84931507, 65.4109589])
[8]: from sklearn.metrics import r2_score
     r2_score(y,y_pred)
[8]: 0.4803218090889322
[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 0x2655a795790>



```
[13]: from sklearn.datasets import fetch_california_housing
      housing = fetch_california_housing()
[14]: data = pd.DataFrame(housing.data)
      data
[15]:
[15]:
                  0
                        1
                                  2
                                             3
                                                     4
                                                               5
                                                                      6
      0
             8.3252
                     41.0
                           6.984127
                                     1.023810
                                                 322.0
                                                        2.555556
                                                                  37.88 -122.23
                           6.238137
                                     0.971880
                                                2401.0
                                                        2.109842
                                                                  37.86 -122.22
      1
             8.3014
                     21.0
      2
             7.2574
                     52.0
                           8.288136
                                     1.073446
                                                 496.0
                                                        2.802260
                                                                  37.85 -122.24
      3
             5.6431
                                                 558.0
                                                                  37.85 -122.25
                     52.0
                           5.817352
                                     1.073059
                                                        2.547945
      4
             3.8462
                     52.0
                           6.281853
                                     1.081081
                                                 565.0
                                                        2.181467
                                                                  37.85 -122.25
                                                 845.0 2.560606
      20635
             1.5603
                     25.0
                           5.045455
                                     1.133333
                                                                  39.48 -121.09
      20636
             2.5568
                     18.0
                           6.114035
                                     1.315789
                                                 356.0 3.122807
                                                                  39.49 -121.21
      20637
             1.7000
                           5.205543
                                     1.120092
                                                1007.0 2.325635
                                                                  39.43 -121.22
                     17.0
      20638
             1.8672
                     18.0
                           5.329513
                                     1.171920
                                                 741.0 2.123209
                                                                  39.43 -121.32
      20639
             2.3886
                     16.0
                           5.254717
                                     1.162264
                                                1387.0 2.616981
                                                                  39.37 -121.24
      [20640 rows x 8 columns]
[16]: data.columns = housing.feature_names
      data.head()
```

```
[16]:
        MedInc HouseAge AveRooms AveBedrms Population AveOccup Latitude \
      0 8.3252
                    41.0 6.984127
                                      1.023810
                                                     322.0 2.555556
                                                                         37.88
      1 8.3014
                     21.0 6.238137
                                      0.971880
                                                    2401.0 2.109842
                                                                         37.86
      2 7.2574
                    52.0 8.288136
                                      1.073446
                                                     496.0 2.802260
                                                                         37.85
      3 5.6431
                    52.0 5.817352
                                      1.073059
                                                     558.0 2.547945
                                                                         37.85
      4 3.8462
                    52.0 6.281853
                                      1.081081
                                                     565.0 2.181467
                                                                         37.85
        Longitude
      0
           -122.23
          -122.22
      1
          -122.24
      2
      3
          -122.25
           -122.25
      4
[17]: data['MedInc'] = housing.target
[18]: data.isnull().sum()
                    0
[18]: MedInc
     HouseAge
                    0
      AveRooms
                    0
      AveBedrms
                   0
     Population
     AveOccup
     Latitude
                    0
     Longitude
      dtype: int64
[19]: x = data.drop(['MedInc'],axis = 1)
      y = data['MedInc']
[20]: from sklearn.model_selection import train_test_split
      xtrain, xtest, ytrain, ytest=train_test_split(x,y, test_size=0.2,random_state=0)
[27]: from sklearn.linear_model import LinearRegression
      lm = LinearRegression()
      lm.fit(xtrain, ytrain)
      ytrain_pred = lm.predict(xtrain)
      ytest_pred = lm.predict(xtest)
[28]: df=pd.DataFrame(ytrain_pred,ytrain)
      df=pd.DataFrame(ytest_pred,ytest)
```

```
[32]: from sklearn.metrics import mean_squared_error, r2_score
mse = mean_squared_error(ytest,ytest_pred)
print(mse)
```

0.8195128774610305

```
[35]: mse = mean_squared_error(ytrain_pred,ytrain)
print(mse)
```

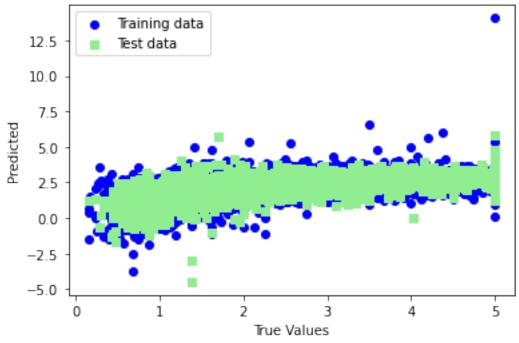
0.7957123632209536

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

0.8195128774610305

```
[44]: 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



[]:	
[]:	