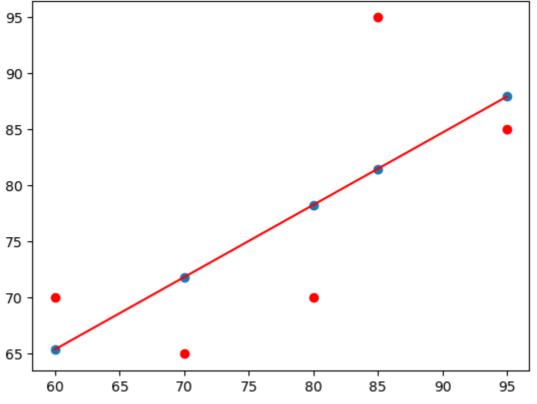
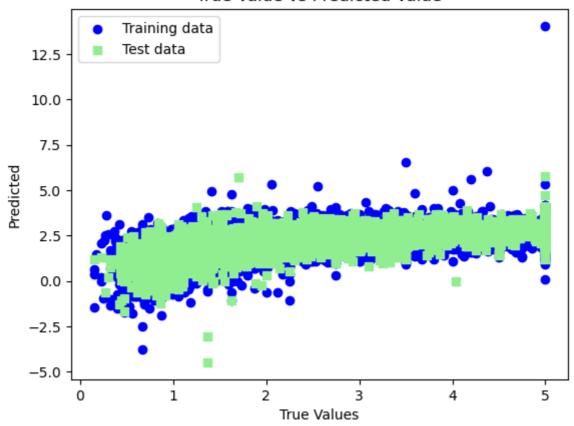
```
import pandas as pd
In [1]:
         import numpy as np
         import matplotlib.pyplot as plt
        x=np.array([95,85,80,70,60])
In [2]:
        y=np.array([85,95,70,65,70])
In [3]:
        model = np.polyfit(x,y,1)
        model
        array([ 0.64383562, 26.78082192])
Out[3]:
        predict = np.poly1d(model)
In [4]:
        predict(65)
        68.63013698630137
Out[4]:
In [5]:
        y_pred = predict(x)
        y_pred
        array([87.94520548, 81.50684932, 78.28767123, 71.84931507, 65.4109589])
Out[5]:
        from sklearn.metrics import r2_score
In [6]:
         r2_score(y,y_pred)
        0.4803218090889326
Out[6]:
        y_{line} = model[1] + model[0]* x
In [7]:
        plt.plot(x,y_line,c = 'r')
        plt.scatter(x,y_pred)
        plt.scatter(x,y,c='r')
        <matplotlib.collections.PathCollection at 0x19c62e03cd0>
Out[7]:
```



```
In [12]:
           from sklearn.datasets import fetch california housing
           housing = fetch_california_housing()
           data = pd.DataFrame(housing.data)
In [14]:
           data
In [15]:
Out[15]:
                      0
                                     2
                                               3
                                                      4
                                                               5
                                                                      6
                                                                              7
               0 8.3252 41.0
                              6.984127 1.023810
                                                   322.0 2.555556 37.88
                                                                        -122.23
               1 8.3014 21.0 6.238137 0.971880 2401.0 2.109842 37.86 -122.22
               2 7.2574
                         52.0 8.288136
                                       1.073446
                                                  496.0
                                                         2.802260 37.85
                                                                        -122.24
               3 5.6431 52.0
                              5.817352
                                       1.073059
                                                   558.0
                                                        2.547945 37.85
                                                                        -122.25
                  3.8462
                         52.0
                              6.281853
                                        1.081081
                                                   565.0
                                                         2.181467
                                                                  37.85
                                                                        -122.25
           20635 1.5603 25.0
                             5.045455
                                       1.133333
                                                  845.0
                                                         2.560606
                                                                  39.48 -121.09
           20636 2.5568
                        18.0
                              6.114035
                                       1.315789
                                                   356.0
                                                        3.122807 39.49 -121.21
                  1.7000
                         17.0
                              5.205543
                                        1.120092
                                                 1007.0
                                                         2.325635
                                                                  39.43
                                                                        -121.22
                 1.8672 18.0
                             5.329513
                                                  741.0
                                                         2.123209 39.43 -121.32
           20638
                                       1.171920
           20639 2.3886 16.0 5.254717 1.162264 1387.0 2.616981 39.37 -121.24
          20640 rows × 8 columns
In [16]:
           data.columns = housing.feature_names
           data.head()
Out[16]:
              MedInc HouseAge
                                 AveRooms AveBedrms
                                                        Population
                                                                    AveOccup
                                                                              Latitude
                                                                                        Longitude
                                   6.984127
           0
               8.3252
                            41.0
                                               1.023810
                                                              322.0
                                                                     2.55556
                                                                                  37.88
                                                                                           -122.23
           1
               8.3014
                            21.0
                                   6.238137
                                               0.971880
                                                             2401.0
                                                                     2.109842
                                                                                  37.86
                                                                                           -122.22
           2
               7.2574
                            52.0
                                   8.288136
                                                              496.0
                                                                     2.802260
                                                                                  37.85
                                                                                           -122.24
                                               1.073446
           3
                            52.0
               5.6431
                                   5.817352
                                               1.073059
                                                              558.0
                                                                     2.547945
                                                                                  37.85
                                                                                           -122.25
               3.8462
                            52.0
                                   6.281853
                                               1.081081
                                                              565.0
                                                                     2.181467
                                                                                  37.85
                                                                                           -122.25
           data['MedInc'] = housing.target
In [17]:
           data.isnull().sum()
In [18]:
          MedInc
                          0
Out[18]:
                          0
          HouseAge
          AveRooms
                          0
          AveBedrms
                          0
          Population
                          0
          Ave0ccup
                          0
          Latitude
                          0
          Longitude
                          0
           dtype: int64
```

```
x = data.drop(['MedInc'],axis = 1)
In [19]:
         y = data['MedInc']
         from sklearn.model_selection import train_test_split
In [20]:
         xtrain, xtest, ytrain, ytest=train_test_split(x,y, test_size=0.2,random_state=0)
         import sklearn
In [21]:
         from sklearn.linear_model import LinearRegression
         lm = LinearRegression()
         model=lm.fit(xtrain,ytrain)
In [22]: ytrain_pred = lm.predict(xtrain)
         ytest_pred = lm.predict(xtest)
         df=pd.DataFrame(ytrain_pred,ytrain)
In [24]:
         df=pd.DataFrame(ytest_pred,ytest)
         from sklearn.metrics import mean squared error, r2 score
In [25]:
         mse = mean_squared_error(ytest,ytest_pred)
         print(mse)
         0.8195128774610314
         mse = mean_squared_error(ytrain_pred,ytrain)
In [26]:
         print(mse)
         0.7957123632209536
         mse = mean_squared_error(ytest,ytest_pred)
In [27]:
         print(mse)
         0.8195128774610314
In [29]: 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.plot()
         plt.show()
```

True Value vs Predicted Value



In []: Name= Anurag Jadhav Roll Ne13171