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In [1]: #Name: Atharv Santosh Danave
         #Roll no: 11
         #Practical no: 04
         #Academic year: 2024-25
 In [4]: import pandas as pd
 In [5]: import numpy as np
 In [6]: import matplotlib.pyplot as plt
 In [7]: x=np.array([95,85,80,70,60])
         y=np.array([85,95,70,65,70])
 In [8]: model=np.polyfit(x,y,1)
 In [9]: model
 Out[9]: array([ 0.64383562, 26.78082192])
In [10]: predict=np.poly1d(model)
         predict(65)
Out[10]: 68.63013698630135
In [11]: y pred=predict(x)
         y_pred
Out[11]: array([87.94520548, 81.50684932, 78.28767123, 71.84931507, 65.4109589])
In [12]: from sklearn.metrics import r2 score
         r2_score(y,y_pred)
Out[12]: 0.4803218090889323
In [13]: 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')
Out[13]: <matplotlib.collections.PathCollection at 0x7f0649321940>
        95
        90
        85
        80
        75
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        65
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```

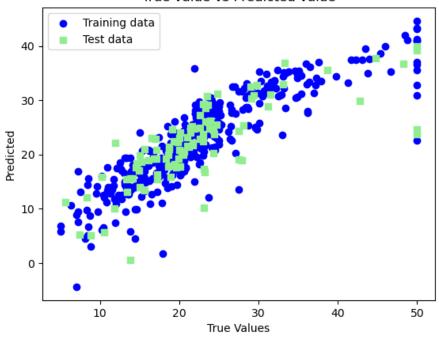
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```
In [1]: import numpy as np
In [3]: import pandas as pd
In [5]: import matplotlib.pyplot as plt
In [61]: data = pd.read csv("https://raw.githubusercontent.com/selva86/datasets/master/BostonHousing.csv")
In [13]: data.head()
Out[13]:
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In [15]: data.tail()
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In [29]: data.isnull().sum()
Out[29]: crim
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          dtype: int64
In [71]: x = data.iloc[:,0:13]
         y = data.iloc[:,-1]
In [73]: 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
In [77]: from sklearn.linear model import LinearRegression
In [79]: lm=LinearRegression()
In [86]: model=lm.fit(xtrain, ytrain)
In [88]: ytrain pred=lm.predict(xtrain)
         ytest pred=lm.predict(xtest)
```

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In [90]: df=pd.DataFrame(ytrain_pred,ytrain)
In [92]: df=pd.DataFrame(ytest_pred,ytest)
In [94]: from sklearn.metrics import mean_squared_error,r2_score
In [96]: mse=mean_squared_error(ytest,ytest_pred)
In [98]: print(mse)
        33.44897999767632
In [100... | mse=mean_squared_error(ytrain_pred,ytrain)
In [102... print(mse)
        19.326470203585725
In [106... 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



In []:

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