In []: #5. Data Analytics I:

#Create a Linear Regression Model using Python/R to predict home prices using #(https://www.kaggle.com/c/boston-housing). The Boston Housing dataset contain #Boston through different parameters. There are 506 samples and 14 feature var #The objective is to predict the value of prices of the house using the given

#if 1st not wprk then 2nd try..boston.csv nnnnn #mam ke pdf wala

In [8]: import numpy as np

import pandas as pd

import matplotlib.pyplot as plt

```
from sklearn.datasets import load_boston
In [9]:
        boston=load boston()
        C:\ProgramData\Anaconda3\lib\site-packages\sklearn\utils\deprecation.py:87:
        FutureWarning: Function load_boston is deprecated; `load_boston` is deprecat
        ed in 1.0 and will be removed in 1.2.
            The Boston housing prices dataset has an ethical problem. You can refer
        to
            the documentation of this function for further details.
            The scikit-learn maintainers therefore strongly discourage the use of th
        is
            dataset unless the purpose of the code is to study and educate about
            ethical issues in data science and machine learning.
            In this special case, you can fetch the dataset from the original
            source::
                import pandas as pd
                import numpy as np
                data url = "http://lib.stat.cmu.edu/datasets/boston"
                raw_df = pd.read_csv(data_url, sep="\s+", skiprows=22, header=None)
                data = np.hstack([raw_df.values[::2, :], raw_df.values[1::2, :2]])
                target = raw df.values[1::2, 2]
            Alternative datasets include the California housing dataset (i.e.
            :func:`~sklearn.datasets.fetch california housing`) and the Ames housing
            dataset. You can load the datasets as follows::
                from sklearn.datasets import fetch california housing
                housing = fetch california housing()
            for the California housing dataset and::
                from sklearn.datasets import fetch openml
                housing = fetch_openml(name="house_prices", as_frame=True)
            for the Ames housing dataset.
          warnings.warn(msg, category=FutureWarning)
```

```
In [10]: data = pd.DataFrame(boston.data)
```

```
In [11]:
          data.columns = boston.feature names
          data.head()
Out[11]:
               CRIM
                      ZN INDUS CHAS
                                        NOX
                                               RM AGE
                                                           DIS RAD
                                                                      TAX PTRATIO
                                                                                        B LST.
           0 0.00632 18.0
                            2.31
                                    0.0 0.538 6.575
                                                    65.2 4.0900
                                                                 1.0 296.0
                                                                               15.3 396.90
                                                                                             4.
           1 0.02731
                            7.07
                                    0.0 0.469 6.421 78.9 4.9671
                                                                 2.0 242.0
                                                                               17.8 396.90
                      0.0
                                                                                             9.
           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.
           3 0.03237
                                    0.0 0.458 6.998 45.8 6.0622
                                                                 3.0 222.0
                                                                               18.7 394.63
                      0.0
                            2.18
                                                                                             2.
             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.
In [12]: data['PRICE'] = boston.target
In [13]: data.isnull().sum()
Out[13]: CRIM
                      0
          ΖN
                      0
          INDUS
                      0
          CHAS
                      0
          NOX
                      0
                      0
          RM
                      0
          AGE
          DIS
                      0
          RAD
                      0
          TAX
                      0
                      0
          PTRATIO
                      0
          LSTAT
                      0
          PRICE
                      0
          dtype: int64
In [14]: | x = data.drop(['PRICE'], axis = 1)
          y = data['PRICE']
In [15]:
          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 [16]:
          import sklearn
          from sklearn.linear_model import LinearRegression
          lm = LinearRegression()
          model=lm.fit(xtrain, ytrain)
```

```
In [17]: ytrain_pred = lm.predict(xtrain)
   ytest_pred = lm.predict(xtest)

In [18]: df=pd.DataFrame(ytrain_pred,ytrain)
   df=pd.DataFrame(ytest_pred,ytest)

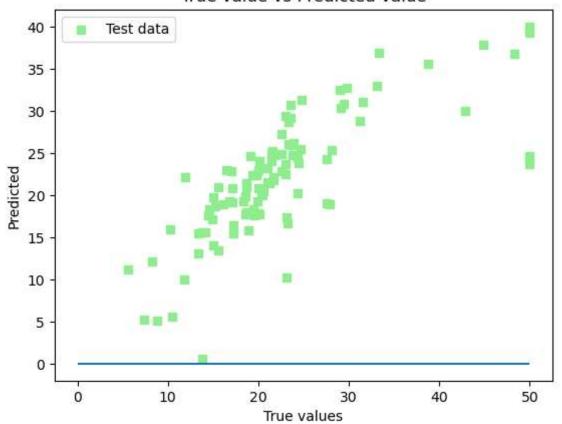
In [19]: from sklearn.metrics import mean_squared_error, r2_score
   mse = mean_squared_error(ytest, ytest_pred)
   print(mse)
   mse = mean_squared_error(ytrain_pred,ytrain)
   print(mse)

33.44897999767653
   19.326470203585725

In [20]: mse = mean_squared_error(ytest, ytest_pred)
   print(mse)
```

33.44897999767653

True value vs Predicted value



In []: