3/20/25, 1:22 PM ASSIGNMENT4

ASSIGNMENT 4

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.

```
In [1]:
          import numpy as np
           import pandas as pd
           import matplotlib.pyplot as plt
           import seaborn as sns
          df = pd.read_csv("BostonHousing.csv")
In [49]:
In [50]:
          df.head()
Out[50]:
                 crim
                             indus
                                    chas
                                            nox
                                                         age
                                                                  dis
                                                                       rad
                                                                                  ptratio
                                                                                                   Istat i
                                                    rm
                                                                             tax
           0.00632
                               2.31
                                                              4.0900
                                                                            296
                                                                                          396.90
                                                                                                   4.98
                       18.0
                                           0.538
                                                  6.575
                                                         65.2
                                                                                     15.3
                                                                         1
                                                  6.421
              0.02731
                        0.0
                               7.07
                                           0.469
                                                         78.9
                                                               4.9671
                                                                            242
                                                                                          396.90
                                                                                                   9.14
                                                                         2
                                                                                     17.8
           2 0.02729
                        0.0
                               7.07
                                                                                          392.83
                                                                                                   4.03
                                           0.469
                                                 7.185
                                                         61.1
                                                              4.9671
                                                                            242
                                                                                     17.8
             0.03237
                         0.0
                                                  6.998
                                                         45.8
                                                               6.0622
                                                                                           394.63
                                                                                                   2.94
                               2.18
                                           0.458
                                                                         3
                                                                            222
                                                                                     18.7
             0.06905
                        0.0
                               2.18
                                           0.458 7.147
                                                         54.2 6.0622
                                                                         3
                                                                            222
                                                                                          396.90
                                                                                                   5.33
                                                                                     18.7
          df.tail()
In [51]:
Out[51]:
                   crim
                              indus chas
                                              nox
                                                           age
                                                                        rad
                                                                              tax ptratio
                                                                                                    Istat
                                                     rm
                0.06263
                          0.0
                               11.93
                                                                              273
                                                                                            391.99
           501
                                            0.573
                                                   6.593
                                                          69.1
                                                                2.4786
                                                                                      21.0
                                                                                                    9.67
                0.04527
                          0.0
                               11.93
                                                   6.120
                                                          76.7
                                                                2.2875
                                                                              273
                                                                                            396.90
                                                                                                    9.08
           502
                                            0.573
                                                                                      21.0
           503
                0.06076
                         0.0
                               11.93
                                            0.573
                                                   6.976
                                                          91.0
                                                                2.1675
                                                                              273
                                                                                      21.0
                                                                                            396.90
                                                                                                     5.64
           504
                0.10959
                          0.0
                               11.93
                                            0.573
                                                   6.794
                                                          89.3
                                                                2.3889
                                                                              273
                                                                                      21.0
                                                                                            393.45
                                                                                                    6.48
                0.04741
                         0.0
                              11.93
                                            0.573 6.030 80.8
                                                               2.5050
                                                                             273
                                                                                      21.0
                                                                                            396.90
                                                                                                    7.88
In [52]:
          print("The shape of the data is: ")
           df.shape
```

3/20/25, 1:22 PM ASSIGNMENT4

The shape of the data is: Out[52]: (506, 14) In [53]: df.isnull().sum() Out[53]: crim 0 zn 0 indus chas 0 0 nox 0 rm age 0 0 dis rad 0 tax 0 ptratio 0 0 lstat 0 medv 0 dtype: int64 In [66]: X = df[['crim', 'zn', 'indus', 'chas', 'nox', 'rm', 'age', 'dis', 'rad', 'tax', 'ptratio', 'b', 'lstat']] Y = df['medv']In [68]: from sklearn.model_selection import train_test_split X_train, X_test, y_train, y_test = train_test_split(X,Y, test_size = 0.3, random_st In [69]: **from** sklearn.preprocessing **import** StandardScaler sc = StandardScaler() X_train = sc.fit_transform(X_train) X_test = sc.fit_transform(X_test) In [70]: from sklearn.linear_model import LinearRegression lr = LinearRegression() lr.fit(X_train, y_train) y_pred = lr.predict(X_test) In [71]: print(y_pred)

3/20/25, 1:22 PM ASSIGNMENT4

```
[30.77834822 38.66581541 16.96705276 26.79035818 20.36068008 24.89386356
         18.98948959 15.44554414 24.50205701 22.12455513 27.16709892 20.60218969
         -4.91312377 23.45520373 20.44300739 27.97454642 21.82481134 6.80793603
         43.13876668 19.08804672 28.93737925 31.91943808 12.05148619 25.41493356
         19.60614164 17.23543021 24.61263361 16.907528 24.25623469 20.73183868
         23.98650325 26.66317526 27.45468242 19.57475267 18.15660164 19.94422395
         33.04896681 21.26475982 25.83570427 26.36436267 15.22545112 33.61610719
         45.24385042 18.83620774 28.91584593 18.49470905 14.91054896 27.65199578
         21.75498681 32.36964545 23.36168054 36.5955736 16.82098601 27.59874838
         42.39261672 24.39945725 20.42785358 34.46608842 26.52068304 14.0304141
         24.03170468 32.40960974 33.59067192 17.33583061 22.46180876 18.28308242
         22.01801285 27.57486429 32.54366102 13.95067422 21.96577449 29.21079363
         12.65101384 17.15704754 25.56951945 6.99478235 22.73334313 43.93983367
         20.03571324 10.919632 22.50309934 14.46771561 23.07376266 10.52688537
         24.71136556 34.38262811 21.02382869 27.04400189 30.94450114 21.54068028
         27.48732329 7.31708375 21.56254771 16.57285469 14.43338806 22.26113509
         26.42907376 1.28686278 15.98179144 18.13930581 23.62097091 26.3208465
         12.12867204 20.66733158 25.32719152 14.23008516 19.48076211 26.9289352
         22.02461937 26.11181168 9.83654844 21.30376854 23.13155127 28.99045653
         34.46072055 17.31422502 36.81803955 14.0639201 22.38719669 30.47720506
         16.96724187 26.23079837 6.30719991 25.6891189 27.44337177 24.41723783
         26.72754089 35.52652858 23.8342233 41.30725508 15.51580837 27.19720138
         18.99437335 22.50080685 11.95557604 23.13463621 23.4954008 34.27666272
         33.53948706 16.55984176 17.98511069 31.00757978 26.51337445 18.4765867
         7.99167823 28.32517448 26.1665569 19.032183 14.47502635 42.88762567
         18.69947012 19.82530166]
In [72]: from sklearn.metrics import mean squared error
         rmse = np.sqrt(mean_squared_error(y_test, y_pred))
         print("Root Mean squared Error is:")
         print(rmse)
        Root Mean squared Error is:
        4.912717301969202
In [73]: print("Training accuracy is:")
         lr.score(X_train, y_train)
        Training accuracy is:
Out[73]: 0.7434997532004697
In [74]: print("Testing accuracy is:")
         lr.score(X_test, y_test)
        Testing accuracy is:
Out[74]: 0.6761000049033605
```