```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline
from sklearn.datasets import load_boston
boston =load boston()
data = pd.DataFrame(boston.data)
data.columns = boston.feature names
data.head()
     CRIM
             ZN INDUS CHAS NOX
                                      RM
                                           AGE
                                                  DIS
                                                       RAD
                                                              TAX
  0.00632 18.0
                 2.31
                        0.0
                             0.538 6.575
                                          65.2
                                                4.0900
                                                       1.0
                                                            296.0
1 0.02731
            0.0
                 7.07
                        0.0
                             0.469 6.421
                                          78.9
                                                4.9671 2.0
                                                            242.0
2 0.02729
            0.0
                 7.07
                        0.0
                             0.469 7.185 61.1 4.9671 2.0
                                                            242.0
3 0.03237
            0.0
                 2.18
                        0.0
                             0.458 6.998 45.8 6.0622 3.0 222.0
4 0.06905
            0.0
                 2.18
                        0.0 0.458 7.147
                                          54.2 6.0622 3.0
                                                            222.0
  PTRATIO
               В
                  LSTAT
0
     15.3
           396.90
                   4.98
1
     17.8
           396.90
                   9.14
2
     17.8
           392.83
                   4.03
3
     18.7
           394.63
                   2.94
4
     18.7 396.90
                   5.33
data['PRICE'] = boston.target
data
       CRIM
               ZN
                  INDUS CHAS
                                 NOX
                                        RM
                                             AGE
                                                    DIS
                                                         RAD
TAX \
    0.00632
             18.0
                   2.31
                          0.0 0.538 6.575
                                            65.2
                                                  4.0900
                                                         1.0
296.0
    0.02731
              0.0
                   7.07
                          0.0 0.469 6.421 78.9
                                                  4.9671
                                                         2.0
1
242.0
    0.02729
              0.0
                   7.07
                          0.0
                               0.469 7.185
                                            61.1
                                                  4.9671
                                                         2.0
242.0
    0.03237
                          0.0 0.458 6.998 45.8
              0.0
                   2.18
                                                  6.0622
                                                         3.0
3
222.0
    0.06905
              0.0
                   2.18
                          0.0 0.458 7.147 54.2
                                                  6.0622
                                                         3.0
222.0
```

```
501 0.06263
              0.0 11.93 0.0 0.573 6.593 69.1 2.4786
                                                           1.0
273.0
502 0.04527
              0.0 11.93
                           0.0 0.573 6.120 76.7
                                                    2.2875
                                                            1.0
273.0
503 0.06076
              0.0 11.93
                           0.0 0.573 6.976 91.0
                                                    2.1675
                                                            1.0
273.0
504 0.10959
              0.0 11.93
                           0.0 0.573 6.794 89.3
                                                    2.3889
                                                            1.0
273.0
505 0.04741
             0.0 11.93 0.0 0.573 6.030 80.8 2.5050 1.0
273.0
    PTRATIO
                  В
                     LSTAT
                            PRICE
0
       15.3
             396.90
                      4.98
                             24.0
1
       17.8
             396.90
                      9.14
                             21.6
2
       17.8
             392.83
                      4.03
                             34.7
3
       18.7
             394.63
                      2.94
                             33.4
4
       18.7
             396.90
                      5.33
                             36.2
501
       21.0
             391.99
                      9.67
                             22.4
       21.0
502
             396.90
                      9.08
                             20.6
503
       21.0
             396.90
                      5.64
                             23.9
       21.0
504
             393.45
                      6.48
                             22.0
       21.0 396.90
505
                      7.88
                             11.9
[506 rows x 14 columns]
data.isnull().sum()
CRIM
          0
ZN
          0
INDUS
          0
CHAS
          0
NOX
          0
RM
           0
           0
AGE
DIS
          0
          0
RAD
          0
TAX
PTRATIO
          0
В
           0
LSTAT
          0
          0
PRICE
dtype: int64
corr = data.corr()
corr.shape
(14, 14)
```

```
plt.figure(figsize=(20,20))
sns.heatmap(corr, cbar=True, square= True, fm='.1f', annot=True,
annot_kws={'size':15}, cmap='Blues')
plt.show()
```

- 0.6

CRIM	1.0	-0.2	0.4	-0.1	0.4	-0.2	0.4	-0.4	0.6	0.6	0.3	-0.4	0.5	-0.4
N -	-0.2	1.0	-0.5	-0.0	-0.5	0.3	-0.6	0.7	-0.3	-0.3	-0.4	0.2	-0.4	0.4
SUDNI	0.4	-0.5	1.0	0.1	0.8	-0.4	0.6	-0.7	0.6	0.7	0.4	-0.4	0.6	-0.5
CHAS	-0.1	-0.0	0.1	1.0	0.1		0.1	-0.1	-0.0	-0.0	-0.1	0.0	-0.1	0.2
XON -	0.4	-0.5	0.8	0.1	1.0	-0.3	0.7	-0.8	0.6	0.7	0.2	-0.4	0.6	-0.4
M -	-0.2	0.3	-0.4		-0.3	1.0	-0.2	0.2	-0.2	-0.3	-0.4	0.1	-0.6	0.7
AGE	0.4	-0.6	0.6		0.7	-0.2	1.0	-0.7	0.5	0.5	0.3	-0.3	0.6	-0.4
DIS -	-0.4	0.7	-0.7	-0.1	-0.8	0.2	-0.7	1.0	-0.5	-0.5	-0.2	0.3	-0.5	0.2
RAD -	0.6	-0.3	0.6	-0.0	0.6	-0.2	0.5	-0.5	1.0	0.9	0.5	-0.4	0.5	-0.4
TAX	0.6	-0.3	0.7	-0.0	0.7	-0.3	0.5	-0.5	0.9	1.0	0.5	-0.4	0.5	-0.5
PTRATIO	0.3	-0.4	0.4	-0.1	0.2	-0.4	0.3	-0.2	0.5	0.5	1.0	-0.2	0.4	-0.5
a -	-0.4	0.2	-0.4	0.0	-0.4	0.1	-0.3	0.3	-0.4	-0.4	-0.2	1.0	-0.4	0.3
LSTAT	0.5	-0.4	0.6	-0.1	0.6	-0.6	0.6	-0.5	0.5	0.5	0.4	-0.4	1.0	-0.7
PRICE -	-0.4	0.4	-0.5	0.2	-0.4	0.7	-0.4	0.2	-0.4	-0.5	-0.5	0.3	-0.7	1.0
	CRIM	ΖŃ	INDUS	CHAS	NOX	RM	AGE	Dis	RAD	TAX	PTRATIO	В	LSTAT	PRICE

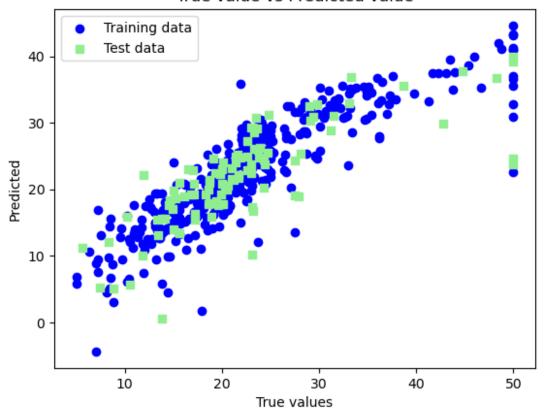
```
x = data.drop(['PRICE'], axis = 1)
y = data['PRICE']
```

```
from sklearn.model selection import train test split
xtrain, xtest, ytrain, ytest =train test split(x, y, test size
=0.2, random state =0)
import sklearn
from sklearn.linear model import LinearRegression
lm = LinearRegression()
# Train the model using the training sets
model=lm.fit(xtrain,ytrain)
xtrain
        CRIM ZN INDUS CHAS NOX
                                         RM
                                              AGE
                                                     DIS
                                                           RAD
TAX \
     0.35809
               0.0 6.20 1.0 0.507 6.951
220
                                             88.5 2.8617
                                                           8.0
307.0
     0.15876 0.0 10.81
                           0.0 0.413
                                      5.961
                                             17.5 5.2873
71
                                                           4.0
305.0
     0.11329 30.0 4.93
240
                           0.0 0.428
                                      6.897
                                             54.3 6.3361
                                                           6.0
300.0
                           0.0 0.524
                                      6.012
6
     0.08829 12.5 7.87
                                             66.6 5.5605
                                                           5.0
311.0
417 25.94060 0.0 18.10
                           0.0 0.679
                                      5.304
                                             89.1 1.6475 24.0
666.0
. .
                           . . .
. . .
323
     0.28392
               0.0 7.38
                           0.0 0.493
                                      5.708 74.3 4.7211
                                                           5.0
287.0
192
     0.08664 45.0 3.44
                           0.0 0.437 7.178
                                             26.3 6.4798
                                                           5.0
398.0
117
              0.0 10.01
                           0.0 0.547
                                      6.021
                                             82.6 2.7474
                                                           6.0
     0.15098
432.0
47
              0.0
                    6.91
                           0.0 0.448
     0.22927
                                      6.030
                                             85.5 5.6894
                                                           3.0
233.0
172
     0.13914
               0.0
                    4.05
                           0.0 0.510 5.572
                                             88.5 2.5961
                                                           5.0
296.0
    PTRATIO
                В
                    LSTAT
220
       17.4
             391.70
                     9.71
       19.2
             376.94
71
                     9.88
       16.6
             391.25
                    11.38
240
6
       15.2
             395.60
                    12.43
417
       20.2
             127.36
                    26.64
        . . .
323
       19.6
             391.13
                    11.74
192
       15.2
             390.49
                    2.87
117
       17.8
             394.51
                    10.30
47
       17.9
             392.74
                    18.80
172
       16.6
            396.90 14.69
```

```
[404 rows x 13 columns]
vtrain pred=lm.predict(xtrain)
ytest pred=lm.predict(xtest)
testdata=[[0.00632,18.0,2.31,0.0,0.538,6.575,65.2,4.0900,1.0,296.0,15.
3,396.90,4.98]]
test pred = lm.predict(testdata)
test pred
C:\ProgramData\Anaconda3\lib\site-packages\sklearn\base.py:450:
UserWarning: X does not have valid feature names, but LinearRegression
was fitted with feature names
 warnings.warn(
array([30.49949836])
df1=pd.DataFrame(ytrain pred,ytrain)
df2=pd.DataFrame(ytest pred,ytest)
df1
PRICE
26.7
       32.556927
21.7
       21.927095
22.0
      27.543826
22.9
      23,603188
10.4
      6.571910
. . .
18.5
       19.494951
36.4
      33.326364
19.2
       23.796208
16.6
      18.458353
23.1 23.249181
[404 rows x 1 columns]
from sklearn.metrics import mean squared error, r2 score
mse = mean_squared_error(ytest, ytest_pred)
print('MSE on test data:',mse)
mse1 = mean_squared_error(ytrain_pred, ytrain)
print('MSE on training data:',msel)
MSE on test data: 33.44897999767653
MSE on training data: 19.326470203585725
#from sklearn.metrics import mean squared error
#def linear metrics():
r2 = lm.score(xtest, ytest)
rmse = (np.sqrt(mean squared error(ytest, ytest pred)))
```

```
print('r-squared: {}' .format(r2))
print('----
print('root mean squared error: {}'.format(rmse))
r-squared: 0.5892223849182507
root mean squared error: 5.783509315085135
#plotting the linear regression model
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



testdata=[[0.00632,18.0,2.31,0.0,0.538,6.575,65.2,4.0900,1.0,296.0,15.3,396.90,4.98]]

```
test_pred = lm.predict(testdata)
test_pred

C:\ProgramData\Anaconda3\lib\site-packages\sklearn\base.py:450:
UserWarning: X does not have valid feature names, but LinearRegression
was fitted with feature names
   warnings.warn(
array([30.49949836])
```