```
import pandas as pd
import numpy as np
import sklearn
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
from sklearn.metrics import mean squared error, r2 score
from sklearn.preprocessing import StandardScaler
df=pd.read_csv('BostonHousing.csv')
df.head(10)
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                                                   65.2 4.0900
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        0.02731
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                                  0 0.469
                                           6.421
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                                                         4.9671
                                                                   2
                                                                      242
                                                                               17.8
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print(df.isnull().sum())
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                а
     dtype: int64
print(df.describe())
₹
                                           indus
                                                         chas
     count
            506.000000
                         506.000000
                                      506.000000
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                                                               506.000000
                                                                            501.000000
              3.613524
                                       11.136779
                                                    0.069170
                                                                 0.554695
                                                                              6.284341
                          11.363636
     mean
               8.601545
                          23.322453
                                        6.860353
                                                    0.253994
                                                                 0.115878
                                                                              0.705587
     std
              0.006320
                           0.000000
                                        0.460000
                                                    0.000000
                                                                 0.385000
                                                                              3.561000
     min
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                                                    0.000000
                                                                 0.449000
                                                                              5.884000
     25%
              0.082045
     50%
              0.256510
                           0.000000
                                        9.690000
                                                    0.000000
                                                                 0.538000
                                                                              6.208000
     75%
              3,677083
                          12.500000
                                       18.100000
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                                                                 0.624000
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     count
             68.574901
                           3.795043
                                        9.549407
                                                  408.237154
                                                                18.455534
                                                                            356.674032
     mean
              28.148861
                           2.105710
                                        8.707259
                                                  168.537116
                                                                 2.164946
                                                                             91,294864
     std
     min
              2.900000
                           1.129600
                                        1.000000
                                                  187.000000
                                                                12.600000
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             45.025000
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                                        4.000000
                                                  279.000000
                                                                17,400000
                                                                            375.377500
     50%
             77,500000
                           3,207450
                                        5,000000
                                                  330,000000
                                                                19,050000
                                                                            391,440000
     75%
             94,075000
                           5.188425
                                       24.000000
                                                  666.000000
                                                                20,200000
                                                                            396.225000
     max
            100,000000
                          12.126500
                                       24.000000
                                                  711.000000
                                                                22,000000
                                                                            396.900000
                  lstat
            506.000000
                         506.000000
     count
```

12.653063

7.141062

1.730000

6.950000

11,360000

16.955000

37,970000

mean std

min

25%

50%

75%

max

22.532806

9.197104

5.000000

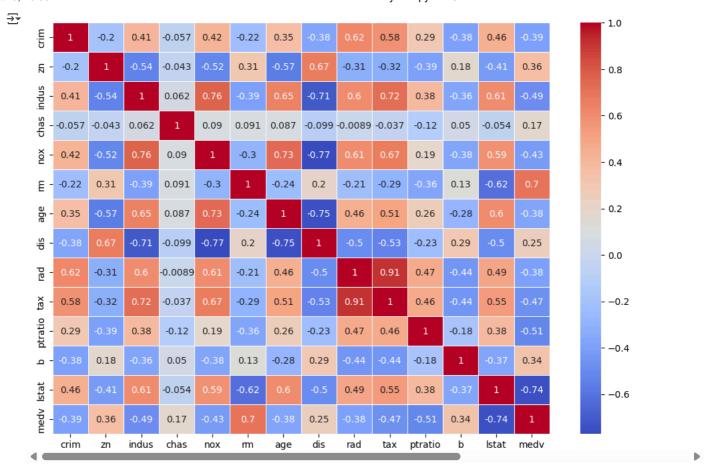
17.025000

21,200000

25.000000

50.000000

```
df = df.dropna()
print(df.isnull().sum())
→ crim
     zn
               0
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     ptratio
     lstat
               0
     medv
               0
     dtype: int64
print(df.isnull().sum())
→ crim
     indus
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               0
               0
     rad
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     tax
     ptratio
               0
               0
     lstat
               0
     medv
               0
     dtype: int64
X = df.drop('medv', axis=1)
y = df['medv']
corr = df.corr()
corr.shape
→ (14, 14)
corr_matrix = df.corr()
plt.figure(figsize=(12, 8))
sns.heatmap(corr_matrix, annot=True, cmap='coolwarm', linewidths=0.5)
plt.show()
```



```
x = df[['rm','ptratio','lstat']]
```

from sklearn.model_selection import train_test_split
xtrain, xtest, ytrain, ytest =train_test_split(x, y, test_size =0.2,random_state = 2)

from sklearn.linear_model import LinearRegression
lm = LinearRegression()
model=lm.fit(xtrain,ytrain)

xtrain

		rm	ptratio	lstat
	496	5.390	19.2	21.14
	55	7.249	17.9	4.81
	341	7.241	15.5	5.49
	159	6.510	14.7	7.39
	398	5.453	20.2	30.59
	23	5.813	21.0	19.88
	75	6.286	18.7	8.94
	498	6.019	19.2	12.92
	16	5.935	21.0	6.58
	173	6.416	16.6	9.04
400 rows × 3 columns				

ytrain_pred=lm.predict(xtrain)
ytest_pred=lm.predict(xtest)

df

y = df['medv']

```
₹
             crim
                    zn indus chas
                                                          dis rad tax ptratio
                                                                                      b 1stat medv
                                      nox
                                              rm
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          0.00632
                   18.0
                          2.31
                                  0 0.538 6.575 65.2 4.0900
                                                                 1 296
                                                                            15.3 396.90
                                                                                          4.98
                                                                                                24.0
          0.02731
                    0.0
                          7.07
                                  0 0.469 6.421 78.9 4.9671
                                                                 2 242
                                                                            17.8 396.90
       1
                                                                                          9.14
                                                                                                21.6
                                                                 2 242
       2
          0.02729
                    0.0
                          7.07
                                  0 0.469
                                           7.185 61.1 4.9671
                                                                            17.8
                                                                                392.83
                                                                                          4.03
                                                                                                34.7
                                  0 0.458 6.998 45.8 6.0622
                                                                 3 222
       3
          0.03237
                    0.0
                          2 18
                                                                            18 7 394 63
                                                                                          2 94
                                                                                                33 4
                                  0 0.458 7.147 54.2 6.0622
                                                                 3 222
                                                                            18.7 396.90
          0.06905
                    0.0
                          2.18
                                                                                          5.33
                                                                                                36.2
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                                                                 1 273
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                                                                                          5.64
                                                                                                23.9
      504 0.10959
                    0.0
                         11.93
                                  0 0.573 6.794 89.3 2.3889
                                                                 1 273
                                                                            21.0 393.45
                                                                                          6.48
                                                                                                22.0
      505 0.04741
                    0.0
                         11.93
                                  0 0.573 6.030 80.8 2.5050
                                                                 1 273
                                                                            21.0 396.90
                                                                                          7.88
                                                                                                11.9
     501 rows × 14 columns
testdata=[[6.575,15.3,4.98]]
test pred = lm.predict(testdata)
test_pred
/usr/local/lib/python3.11/dist-packages/sklearn/utils/validation.py:2739: UserWarning: X does not have valid feature names, but Line
       warnings.warn(
     array([31.17744056])
df1=pd.DataFrame(ytrain_pred,ytrain)
df2=pd.DataFrame(ytest_pred,ytest)
df1
₹
                   0
      medv
      19.7
           12.843707
      35.4 32.020235
           33.626917
      32.7
      23.3
           29.945748
      5.0
            6.543438
       ...
          13.926973
      14.5
      21.4 24.599352
      21.2 20.587343
      23.1 22.516369
      23.6 26.907501
     400 rows × 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:',mse1)
    MSE on test data: 22.733525859178872
     MSE on training data: 28.51317586815709
r2 = lm.score(xtest, ytest)
rmse = (np.sqrt(mean_squared_error(ytest, ytest_pred)))
print('r\text{-squared: }\{\}'\text{ .format(r2)})
print('----')
print('root mean squared error: {}'.format(rmse))
    r-squared: 0.7034416121361311
```

root mean squared error: 4.767968735130179

```
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

