

```

import numpy as np
import pandas as pd
import sklearn

from sklearn.datasets import load_boston
df = load_boston()
df.keys()
print(df.DESCR)
print(df.feature_names)
print(df.filename)
print(df.target)
boston = pd.DataFrame(df.data, columns=df.feature_names)
boston.head()
boston['MEDV']=df.target
boston.isnull()
boston.isnull().sum()
from sklearn.model_selection import train_test_split
X=boston.drop('MEDV', axis=1)
Y=boston ['MEDV']
X_train, X_test, Y_train,Y_test = train_test_split(X, Y, test_size = 0.15)
print(X_train.shape)
print(X_test.shape)
print(Y_train.shape)
print(Y_test.shape)
from sklearn.linear_model import LinearRegression
from sklearn.metrics import mean_squared_error
lin_model = LinearRegression()
lin_model.fit(X_train, Y_train)
Y_train_predict = lin_model.predict(X_train)
rmse = (np.sqrt(mean_squared_error(Y_train, Y_train_predict)))
print("The models performance for training sets")
print('rmse is {}'.format(rmse))
print("\n")
#on testing set
Y_test_predict = lin_model.predict(X_test)
rmse = (np.sqrt(mean_squared_error(Y_test, Y_test_predict)))
print("the model performance for testing set")
print('rmse is{}'.format(rmse))

```

.. _boston_dataset:

Boston house prices dataset

****Data Set Characteristics:****

:Number of Instances: 506

:Number of Attributes: 13 numeric/categorical predictive. Median Value (att

:Attribute Information (in order):

- CRIM per capita crime rate by town
- ZN proportion of residential land zoned for lots over 25,000 sq
- INDUS proportion of non-retail business acres per town
- CHAS Charles River dummy variable (= 1 if tract bounds river; 0 o
- NOX nitric oxides concentration (parts per 10 million)
- RM average number of rooms per dwelling

- AGE proportion of owner-occupied units built prior to 1940
- DIS weighted distances to five Boston employment centres
- RAD index of accessibility to radial highways
- TAX full-value property-tax rate per \$10,000
- PTRATIO pupil-teacher ratio by town
- B $1000(B_k - 0.63)^2$ where B_k is the proportion of black people
- LSTAT % lower status of the population
- MEDV Median value of owner-occupied homes in \$1000's

:Missing Attribute Values: None

:Creator: Harrison, D. and Rubinfeld, D.L.

This is a copy of UCI ML housing dataset.

<https://archive.ics.uci.edu/ml/machine-learning-databases/housing/>

This dataset was taken from the StatLib library which is maintained at Carnegie

The Boston house-price data of Harrison, D. and Rubinfeld, D.L. 'Hedonic prices and the demand for clean air', J. Environ. Economics & Management, vol.5, 81-102, 1978. Used in Belsley, Kuh & Welsch, 'Regression diagnostics ...', Wiley, 1980. N.B. Various transformations are used in the table on pages 244-261 of the latter.

The Boston house-price data has been used in many machine learning papers that problems.

.. topic:: References

- Belsley, Kuh & Welsch, 'Regression diagnostics: Identifying Influential Da
- Quinlan, R. (1993). Combining Instance-Based and Model-Based Learning. In P

['CRIM' 'ZN' 'INDUS' 'CHAS' 'NOX' 'RM' 'AGE' 'DIS' 'RAD' 'TAX' 'PTRATIO'
'B' 'LSTAT']

boston_house_prices.csv

```
[24.  21.6 34.7 33.4 36.2 28.7 22.9 27.1 16.5 18.9 15.  18.9 21.7 20.4
 18.2 19.9 23.1 17.5 20.2 18.2 13.6 19.6 15.2 14.5 15.6 13.9 16.6 14.8
 18.4 21.  12.7 14.5 13.2 13.1 13.5 18.9 20.  21.  24.7 30.8 34.9 26.6
 25.3 24.7 21.2 19.3 20.  16.6 14.4 19.4 19.7 20.5 25.  23.4 18.9 35.4]
```

[Colab paid products](#) - [Cancel contracts here](#)