```
In [1]: from sklearn.datasets import fetch_california_housing
In [2]: from sklearn.model_selection import train_test_split
In [3]: from sklearn.linear_model import LinearRegression
In [4]: from sklearn.metrics import mean_absolute_error,mean_squared_error,r2_score
In [5]: data = fetch_california_housing() # data = pd.read_csv('houseprice.csv')
```

```
In [11]: print(data)
```

```
6.98412698, ...,
                                                                      2.5555
{'data': array([[
                    8.3252
                                  41.
5556,
          37.88
                     , -122.23
                                    ],
          8.3014
                                         6.23813708, ...,
                                                             2.10984183,
       Γ
                         21.
          37.86
                       -122.22
                                    ],
          7.2574
                                         8.28813559, ...,
                                                             2.80225989,
                         52.
                     , -122.24
          37.85
                                    ],
           1.7
                                         5.20554273, ...,
                         17.
                                                             2.3256351 ,
          39.43
                       -121.22
                                    ],
          1.8672
                         18.
                                         5.32951289, ...,
                                                            2.12320917,
          39.43
                      -121.32
                                    ],
           2.3886
                                         5.25471698, ...,
                                                             2.61698113,
                         16.
                                    ]]), 'target': array([4.526, 3.585, 3.52
          39.37
                      -121.24
1, ..., 0.923, 0.847, 0.894]), 'frame': None, 'target_names': ['MedHouseVa
l'], 'feature_names': ['MedInc', 'HouseAge', 'AveRooms', 'AveBedrms', 'Popul
ation', 'AveOccup', 'Latitude', 'Longitude'], 'DESCR': '.. _california_housi
ng_dataset:\n\nCalifornia Housing dataset\n-----\n\n**D
ata Set Characteristics:**\n\n
                                 :Number of Instances: 20640\n\n
of Attributes: 8 numeric, predictive attributes and the target\n\n
bute Information:\n

    MedInc

                                          median income in block group\n
- HouseAge
                median house age in block group\n
                                                         - AveRooms
                                                                         ave
rage number of rooms per household\n
                                            - AveBedrms
                                                            average number o
f bedrooms per household\n
                                  - Population
                                                  block group population\n
- AveOccup
                average number of household members\n
                                                             - Latitude
block group latitude\n

    Longitude

                                              block group longitude\n\n
Missing Attribute Values: None\n\nThis dataset was obtained from the StatLib
repository.\nhttps://www.dcc.fc.up.pt/~ltorgo/Regression/cal housing.html\n
\nThe target variable is the median house value for California districts,\ne
xpressed in hundreds of thousands of dollars ($100,000).\n\nThis dataset was
derived from the 1990 U.S. census, using one row per census\nblock group. A
block group is the smallest geographical unit for which the U.S.\nCensus Bur
eau publishes sample data (a block group typically has a population\nof 600
to 3,000 people).\n\nA household is a group of people residing within a hom
e. Since the average\nnumber of rooms and bedrooms in this dataset are provi
ded per household, these\ncolumns may take surprisingly large values for blo
ck groups with few households\nand many empty houses, such as vacation resor
ts.\n\nIt can be downloaded/loaded using the\n:func:`sklearn.datasets.fetch
california_housing` function.\n\n.. topic:: References\n\n
                                                            - Pace, R. Kel
ley and Ronald Barry, Sparse Spatial Autoregressions,\n
                                                             Statistics and
Probability Letters, 33 (1997) 291-297\n'}
```

```
In [12]: x = data.data
In [14]: y = data.target
In [15]: x_train, x_test, y_train, y_test = train_test_split(x,y,test_size=0.2)
```

```
In [16]: model = LinearRegression()
In [17]: model.fit(x train,y train)
Out[17]: LinearRegression()
         In a Jupyter environment, please rerun this cell to show the HTML representation or
         trust the notebook.
         On GitHub, the HTML representation is unable to render, please try loading this page
         with nbviewer.org.
In [18]: y pred = model.predict(x test)
In [19]: y_pred
Out[19]: array([1.02221485, 1.48808491, 1.75052818, ..., 3.20591363, 1.07120023,
                 2.945452391)
In [20]: | mae = mean_absolute_error(y_test,y_pred)
In [21]: |mse = mean_squared_error(y_test,y_pred)
In [22]: r2_score = r2_score(y_test,y_pred)
In [23]: print('Mean Absolute Error', mae)
         Mean Absolute Error 0.5345160416296426
In [24]: print('Mean Square Error',mse)
         Mean Square Error 0.5206071326972168
In [25]: print('R2 Score',r2_score)
         R2 Score 0.6074818254825243
 In [ ]:
```