```
In [1]: import pandas as pd
import numpy as np
```

In [2]: #reading a csv file
 df = pd.read\_csv("https://raw.githubusercontent.com/YBIFoundation/Dataset/ma

In [3]: df

## Out[3]:

	CRIM	ZN	INDUS	CHAS	NX	RM	AGE	DIS	RAD	TAX	PTRATIO	В
0	0.00632	18.0	2.31	0	0.538	6.575	65.2	4.0900	1	296.0	15.3	396.90
1	0.02731	0.0	7.07	0	0.469	6.421	78.9	4.9671	2	242.0	17.8	396.90
2	0.02729	0.0	7.07	0	0.469	7.185	61.1	4.9671	2	242.0	17.8	392.83
3	0.03237	0.0	2.18	0	0.458	6.998	45.8	6.0622	3	222.0	18.7	394.63
4	0.06905	0.0	2.18	0	0.458	7.147	54.2	6.0622	3	222.0	18.7	396.90
501	0.06263	0.0	11.93	0	0.573	6.593	69.1	2.4786	1	273.0	21.0	391.99
502	0.04527	0.0	11.93	0	0.573	6.120	76.7	2.2875	1	273.0	21.0	396.90
503	0.06076	0.0	11.93	0	0.573	6.976	91.0	2.1675	1	273.0	21.0	396.90
504	0.10959	0.0	11.93	0	0.573	6.794	89.3	2.3889	1	273.0	21.0	393.45
505	0.04741	0.0	11.93	0	0.573	6.030	80.8	2.5050	1	273.0	21.0	396.90

506 rows × 14 columns

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## In [4]: df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 506 entries, 0 to 505
Data columns (total 14 columns):

<b>D</b> G C G	COLUMNIS	(	٠,٠
#	Column	Non-Null Count	Dtype
0	CRIM	506 non-null	float64
1	ZN	506 non-null	float64
2	INDUS	506 non-null	float64
3	CHAS	506 non-null	int64
4	NX	506 non-null	float64
5	RM	506 non-null	float64
6	AGE	506 non-null	float64
7	DIS	506 non-null	float64
8	RAD	506 non-null	int64
9	TAX	506 non-null	float64
10	PTRATIO	506 non-null	float64
11	В	506 non-null	float64
12	LSTAT	506 non-null	float64
13	MEDV	506 non-null	float64

dtypes: float64(12), int64(2)

memory usage: 55.4 KB

```
df.isnull().sum()
In [7]:
Out[7]: CRIM
                       0
          ZN
                       0
          INDUS
                       0
          CHAS
                       0
          NX
                       0
          RM
                       0
          AGE
                       0
          DIS
                       0
          RAD
                       0
          TAX
                       0
          PTRATIO
                       0
                       0
                       0
          LSTAT
          MEDV
                       0
          dtype: int64
In [8]:
          df.describe()
Out[8]:
                                     ΖN
                                              INDUS
                                                                                               AGE
                       CRIM
                                                          CHAS
                                                                         NX
                                                                                    RM
                  506.000000
                              506.000000
                                         506.000000
                                                     506.000000
                                                                 506.000000
                                                                             506.000000
                                                                                         count
                    3.613524
                                           11.136779
                                                       0.069170
                                                                   0.554695
                                                                               6.284634
                                                                                          68.574901
           mean
                               11.363636
                    8.601545
                               23.322453
                                           6.860353
                                                       0.253994
                                                                   0.115878
                                                                               0.702617
                                                                                          28.148861
             std
                    0.006320
                                0.000000
                                           0.460000
                                                       0.000000
                                                                   0.385000
                                                                               3.561000
                                                                                           2.900000
            min
            25%
                    0.082045
                                0.000000
                                           5.190000
                                                       0.000000
                                                                   0.449000
                                                                               5.885500
                                                                                          45.025000
            50%
                    0.256510
                                0.000000
                                           9.690000
                                                       0.000000
                                                                   0.538000
                                                                               6.208500
                                                                                          77.500000
            75%
                    3.677083
                               12.500000
                                           18.100000
                                                       0.000000
                                                                   0.624000
                                                                               6.623500
                                                                                          94.075000
                                                                   0.871000
                                                                                         100.000000
                   88.976200
                              100.000000
                                           27.740000
                                                        1.000000
                                                                               8.780000
            max
```

In [13]: y = df[['MEDV']]

```
In [14]: y
```

Out	[14]	:
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	MEDV
0	24.0
1	21.6
2	34.7
3	33.4
4	36.2
501	22.4
502	20.6
503	23.9
504	22.0
505	11.9

506 rows × 1 columns

In [16]: x = df.drop("MEDV", axis = 1)

In [17]: x

Out[17]:

	CRIM	ZN	INDUS	CHAS	NX	RM	AGE	DIS	RAD	TAX	PTRATIO	В
0	0.00632	18.0	2.31	0	0.538	6.575	65.2	4.0900	1	296.0	15.3	396.90
1	0.02731	0.0	7.07	0	0.469	6.421	78.9	4.9671	2	242.0	17.8	396.90
2	0.02729	0.0	7.07	0	0.469	7.185	61.1	4.9671	2	242.0	17.8	392.83
3	0.03237	0.0	2.18	0	0.458	6.998	45.8	6.0622	3	222.0	18.7	394.63
4	0.06905	0.0	2.18	0	0.458	7.147	54.2	6.0622	3	222.0	18.7	396.90
501	0.06263	0.0	11.93	0	0.573	6.593	69.1	2.4786	1	273.0	21.0	391.99
502	0.04527	0.0	11.93	0	0.573	6.120	76.7	2.2875	1	273.0	21.0	396.90
503	0.06076	0.0	11.93	0	0.573	6.976	91.0	2.1675	1	273.0	21.0	396.90
504	0.10959	0.0	11.93	0	0.573	6.794	89.3	2.3889	1	273.0	21.0	393.45
505	0.04741	0.0	11.93	0	0.573	6.030	80.8	2.5050	1	273.0	21.0	396.90

506 rows × 13 columns

In [20]: from sklearn.model\_selection import train\_test\_split X\_train, X\_test, y\_train, y\_test = train\_test\_split(x, y, test\_size=0.22, range)

In [21]: X\_train

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	CRIM	ZN	INDUS	CHAS	NX	RM	AGE	DIS	RAD	TAX	PTRATIO	В
334	0.03738	0.0	5.19	0	0.515	6.310	38.5	6.4584	5	224.0	20.2	389.40
181	0.06888	0.0	2.46	0	0.488	6.144	62.2	2.5979	3	193.0	17.8	396.90
227	0.41238	0.0	6.20	0	0.504	7.163	79.9	3.2157	8	307.0	17.4	372.08
434	13.91340	0.0	18.10	0	0.713	6.208	95.0	2.2222	24	666.0	20.2	100.63
180	0.06588	0.0	2.46	0	0.488	7.765	83.3	2.7410	3	193.0	17.8	395.56
106	0.17120	0.0	8.56	0	0.520	5.836	91.9	2.2110	5	384.0	20.9	395.67
270	0.29916	20.0	6.96	0	0.464	5.856	42.1	4.4290	3	223.0	18.6	388.65
348	0.01501	80.0	2.01	0	0.435	6.635	29.7	8.3440	4	280.0	17.0	390.94
435	11.16040	0.0	18.10	0	0.740	6.629	94.6	2.1247	24	666.0	20.2	109.85
102	0.22876	0.0	8.56	0	0.520	6.405	85.4	2.7147	5	384.0	20.9	70.80

394 rows × 13 columns

In [22]:

X\_test

## Out[22]:

	CRIM	ZN	INDUS	CHAS	NX	RM	AGE	DIS	RAD	TAX	PTRATIO	E
173	0.09178	0.0	4.05	0	0.510	6.416	84.1	2.6463	5	296.0	16.6	395.50
274	0.05644	40.0	6.41	1	0.447	6.758	32.9	4.0776	4	254.0	17.6	396.90
491	0.10574	0.0	27.74	0	0.609	5.983	98.8	1.8681	4	711.0	20.1	390.1 <sup>-</sup>
72	0.09164	0.0	10.81	0	0.413	6.065	7.8	5.2873	4	305.0	19.2	390.9 <sup>-</sup>
452	5.09017	0.0	18.10	0	0.713	6.297	91.8	2.3682	24	666.0	20.2	385.09
325	0.19186	0.0	7.38	0	0.493	6.431	14.7	5.4159	5	287.0	19.6	393.68
335	0.03961	0.0	5.19	0	0.515	6.037	34.5	5.9853	5	224.0	20.2	396.90
56	0.02055	85.0	0.74	0	0.410	6.383	35.7	9.1876	2	313.0	17.3	396.90
437	15.17720	0.0	18.10	0	0.740	6.152	100.0	1.9142	24	666.0	20.2	9.32
409	14.43830	0.0	18.10	0	0.597	6.852	100.0	1.4655	24	666.0	20.2	179.36

112 rows × 13 columns

```
In [23]:
          y_train
Out[23]:
                MEDV
                 20.7
           334
           181
                 36.2
           227
                 31.6
           434
                 11.7
           180
                 39.8
           106
                 19.5
           270
                 21.1
           348
                 24.5
           435
                 13.4
           102
                 18.6
          394 rows × 1 columns
In [24]:
          y_test
Out[24]:
                MEDV
           173
                 23.6
           274
                 32.4
           491
                 13.6
            72
                 22.8
           452
                 16.1
           325
                 24.6
           335
                 21.1
                 24.7
            56
           437
                  8.7
           409
                 27.5
          112 rows × 1 columns
          from sklearn.linear_model import LinearRegression
In [27]:
          model = LinearRegression()
          model.fit(X_train, y_train)
Out[27]: LinearRegression()
In [31]: |model.intercept_
Out[31]: array([30.19463764])
```

```
In [29]: model.coef_
Out[29]: array([[-1.11553969e-01, 2.95152462e-02, 3.86075523e-02,
                   2.79971538e+00, -1.64618910e+01, 4.37750378e+00,
                  -7.77843388e-03, -1.43434767e+00, 2.55926441e-01,
                  -1.06030604e-02, -9.14793654e-01, 1.28877921e-02,
                  -5.15350135e-01]])
In [32]: | # y = 30.194 + (-1.11553969e - 01) * CRIM + 2.95152462e - 02 * ZN . . . . . . . . . . . .
         y_predict = model.predict(X_test)
In [33]: y_predict
Out[33]: array([[28.97528729],
                 [36.06237867],
                 [14.78397171],
                 [25.12703189],
                 [18.70716381],
                 [23.17927344],
                 [17.649153
                 [14.26888579],
                 [22.86584011],
                 [20.64941593],
                 [24.87153295],
                 [18.62998967],
                 [-6.31043237],
                 [21.73416181],
                 [19.2665216],
                 [26.08512913],
                 [20.59744865],
                 [ 5.70346772],
                 [40.44438231],
                 F47 FF334004 T
In [34]: #using rmse
          from sklearn.metrics import mean_squared_error
In [35]:
          mean_squared_error(y_test,y_predict)
Out[35]: 23.055926174103806
 In [ ]:
 In [ ]:
 In [ ]:
```