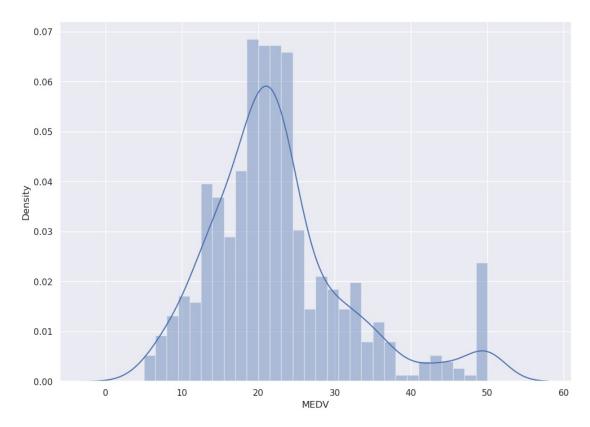
```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
dataset = pd.read_csv("HousingData.csv")
dataset
        CRIM
                ZN
                     INDUS
                            CHAS
                                    NOX
                                             RM
                                                  AGE
                                                           DIS
                                                                RAD
                                                                     TAX
\
0
     0.00632
              18.0
                      2.31
                             0.0 0.538 6.575
                                                 65.2
                                                       4.0900
                                                                  1
                                                                     296
1
                                  0.469 6.421
     0.02731
               0.0
                      7.07
                             0.0
                                                 78.9
                                                       4.9671
                                                                  2
                                                                     242
2
     0.02729
               0.0
                      7.07
                             0.0 0.469
                                         7.185
                                                 61.1
                                                       4.9671
                                                                  2
                                                                     242
3
     0.03237
               0.0
                      2.18
                             0.0
                                  0.458
                                         6.998
                                                 45.8
                                                       6.0622
                                                                  3
                                                                     222
     0.06905
4
               0.0
                      2.18
                             0.0
                                  0.458
                                         7.147
                                                 54.2
                                                       6.0622
                                                                  3
                                                                     222
. .
         . . .
                . . .
                                     . . .
                                            . . .
     0.06263
                    11.93
                                  0.573
                                         6.593
                                                 69.1
                                                                     273
501
               0.0
                             0.0
                                                       2.4786
                                                                  1
502
     0.04527
               0.0
                    11.93
                                  0.573 6.120
                                                 76.7
                                                       2.2875
                                                                     273
                             0.0
                                                                  1
     0.06076
                    11.93
                                  0.573 6.976
503
               0.0
                             0.0
                                                 91.0
                                                       2.1675
                                                                  1
                                                                     273
504
     0.10959
               0.0
                    11.93
                             0.0 0.573 6.794
                                                 89.3
                                                       2.3889
                                                                     273
                                                                  1
     0.04741
               0.0 11.93
                             0.0 0.573 6.030
                                                                  1 273
505
                                                  NaN
                                                       2.5050
     PTRATIO
                       LSTAT
                              MEDV
                    В
0
        15.3
              396.90
                        4.98
                              24.0
              396.90
                              21.6
1
        17.8
                        9.14
2
        17.8
              392.83
                        4.03
                              34.7
3
        18.7
              394.63
                        2.94
                              33.4
4
        18.7
              396.90
                         NaN
                              36.2
                         . . .
                               . . .
. .
         . . .
501
        21.0
              391.99
                              22.4
                         NaN
502
        21.0
                              20.6
              396.90
                        9.08
503
        21.0
                              23.9
              396.90
                        5.64
504
        21.0
              393.45
                        6.48
                              22.0
505
        21.0
              396.90
                        7.88
                              11.9
```

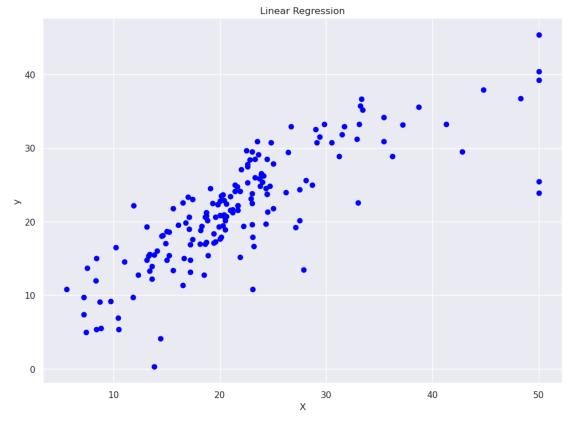
[506 rows x 14 columns]

dataset.isnull().sum()

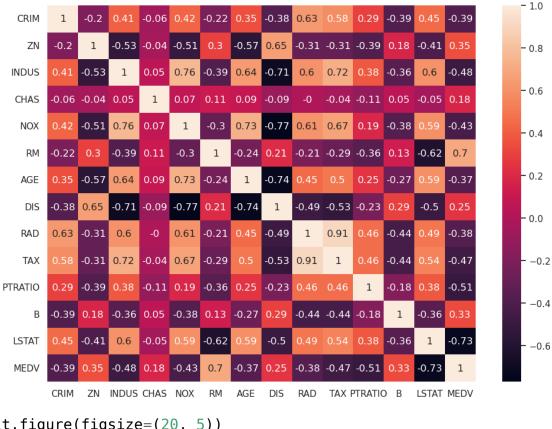
```
CRIM
           20
           20
\mathsf{ZN}
INDUS
           20
CHAS
           20
NOX
            0
RM
            0
           20
AGE
DIS
            0
RAD
            0
TAX
            0
PTRATIO
            0
            0
LSTAT
           20
MEDV
            0
dtype: int64
dataset.fillna(method = 'ffill', inplace = True)
dataset.isnull().sum()
CRIM
           0
ZN
           0
INDUS
           0
CHAS
           0
           0
NOX
RM
           0
AGE
           0
DIS
           0
RAD
           0
           0
TAX
PTRATIO
           0
           0
В
LSTAT
           0
MEDV
           0
dtype: int64
sns.set(rc={'figure.figsize':(11.7,8.27)})
sns.distplot(dataset['MEDV'], bins=30)
plt.show()
/home/mmcoe/anaconda3/lib/python3.9/site-packages/seaborn/
distributions.py:2619: FutureWarning: `distplot` is a deprecated
function and will be removed in a future version. Please adapt your
code to use either `displot` (a figure-level function with similar
flexibility) or `histplot` (an axes-level function for histograms).
  warnings.warn(msg, FutureWarning)
```



```
X = dataset.iloc[:, :-1].values
y = dataset.iloc[:, -1].values
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y,
test_size=1/3, random_state = 0)
from sklearn.linear model import LinearRegression
from sklearn.metrics import mean squared error
regressor = LinearRegression()
regressor.fit(X train, y train)
LinearRegression()
y pred = regressor.predict(X test)
plt.scatter(y_test, y_pred, color = "blue")
#plt.plot(y_test, regressor.predict(y_pred), color = "green")
plt.title("Linear Regression")
plt.xlabel("X")
plt.ylabel("y")
plt.show()
```

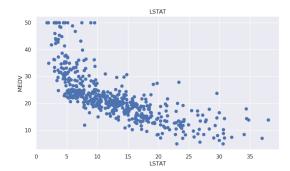


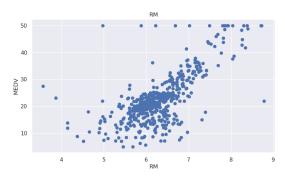
```
print(X_train.shape)
(337, 13)
print(y_train.shape)
(337,)
print(X_test.shape)
(169, 13)
print(y_test.shape)
(169,)
correlation_matrix = dataset.corr().round(2)
sns.heatmap(data=correlation_matrix, annot=True)
<AxesSubplot:>
```



```
plt.figure(figsize=(20, 5))
```

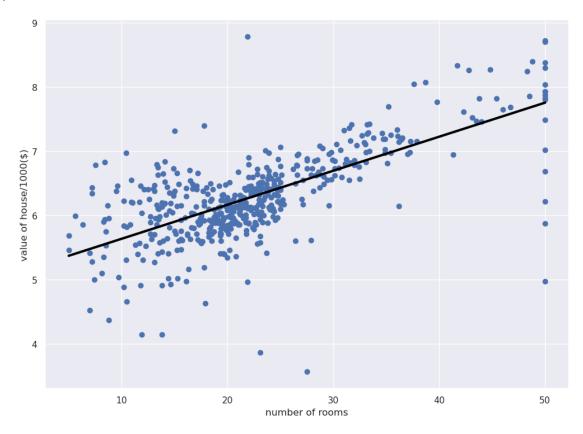
```
features = ['LSTAT', 'RM']
target = dataset['MEDV']
for i, col in enumerate(features):
    plt.subplot(1, len(features) , i+1)
    x = dataset[col]
    y = target
    plt.scatter(x, y, marker='o')
    #plt.plot(X, y, color = "green")
    plt.title(col)
    plt.xlabel(col)
    plt.ylabel('MEDV')
```





```
X rooms = dataset.MEDV
y price = dataset.RM
X rooms = np.array(X rooms).reshape(-1,1)
y_price = np.array(y_price).reshape(-1,1)
print(X rooms.shape)
print(y_price.shape)
(506, 1)
(506, 1)
X_train_1, X_test_1, Y_train_1, Y_test_1 = train_test_split(X_rooms,
y_price, test_size = 0.2, random state=0)
print(X train 1.shape)
print(X_test_1.shape)
print(Y train 1.shape)
print(Y test 1.shape)
(404, 1)
(102, 1)
(404, 1)
(102, 1)
reg 1 = LinearRegression()
reg 1.fit(X train 1, Y train 1)
y train predict 1 = reg 1.predict(X train 1)
rmse = (np.sqrt(mean_squared_error(Y_train_1, y_train_predict_1)))
r2 = round(reg 1.score(X train 1, Y train 1),2)
print("The model performance for training set")
print("-----
print('RMSE is {}'.format(rmse))
print('R2 score is {}'.format(r2))
print("\n")
The model performance for training set
RMSE is 0.4920435120933795
R2 score is 0.5
prediction space = np.linspace(min(X rooms), max(X rooms)).reshape(-
plt.scatter(X_rooms,y_price)
plt.plot(prediction space, reg 1.predict(prediction space), color =
'black', linewidth = 3)
```

```
plt.ylabel('value of house/1000($)')
plt.xlabel('number of rooms')
plt.show()
```



## dataset

`	CRIM	ZN	INDUS	CHAS	NOX	RM	AGE	DIS	RAD	TAX
0	0.00632	18.0	2.31	0.0	0.538	6.575	65.2	4.0900	1	296
1	0.02731	0.0	7.07	0.0	0.469	6.421	78.9	4.9671	2	242
2	0.02729	0.0	7.07	0.0	0.469	7.185	61.1	4.9671	2	242
3	0.03237	0.0	2.18	0.0	0.458	6.998	45.8	6.0622	3	222
4	0.06905	0.0	2.18	0.0	0.458	7.147	54.2	6.0622	3	222
501	0.06263	0.0	11.93	0.0	0.573	6.593	69.1	2.4786	1	273
502	0.04527	0.0	11.93	0.0	0.573	6.120	76.7	2.2875	1	273
503	0.06076	0.0	11.93	0.0	0.573	6.976	91.0	2.1675	1	273

```
0.0 11.93 0.0 0.573 6.794 89.3 2.3889
504 0.10959
                                                            1 273
505 0.04741 0.0 11.93 0.0 0.573 6.030 89.3 2.5050 1 273
    PTRATIO
              B LSTAT
                           MEDV
             396.90
0
       15.3
                     4.98
                           24.0
       17.8
                     9.14
                           21.6
1
             396.90
2
       17.8
             392.83
                     4.03
                           34.7
                           33.4
             394.63
3
       18.7
                     2.94
4
       18.7
             396.90
                           36.2
                     2.94
                     ...
14.33
                           . . .
. .
       . . .
                . . .
             391.99
                           22.4
501
       21.0
       21.0
502
             396.90
                     9.08
                           20.6
       21.0
                     5.64
                           23.9
503
             396.90
504
       21.0
             393.45
                      6.48
                           22.0
505
       21.0
            396.90
                      7.88
                           11.9
```

[506 rows x 14 columns]