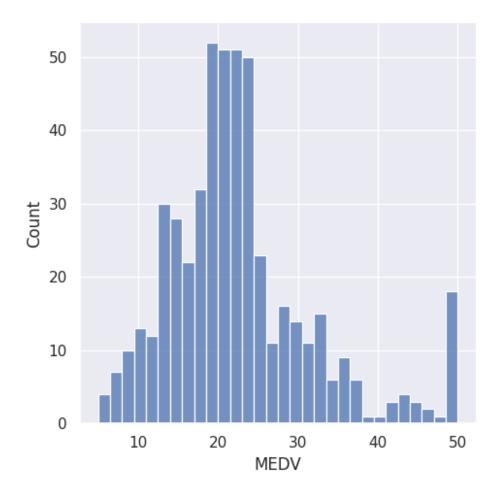
```
import pandas as pd
import matplotlib.pyplot as plt
import numpy as np
import seaborn as sns
df = pd.read csv("/content/HousingData.csv")
df
        CRIM
                ZN
                     INDUS
                            CHAS
                                    NOX
                                             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
                      7.07
                                  0.469 6.421
     0.02731
               0.0
                             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
              396.90
                              20.6
                        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 \times 14 columns]
```

df.isnull().sum()

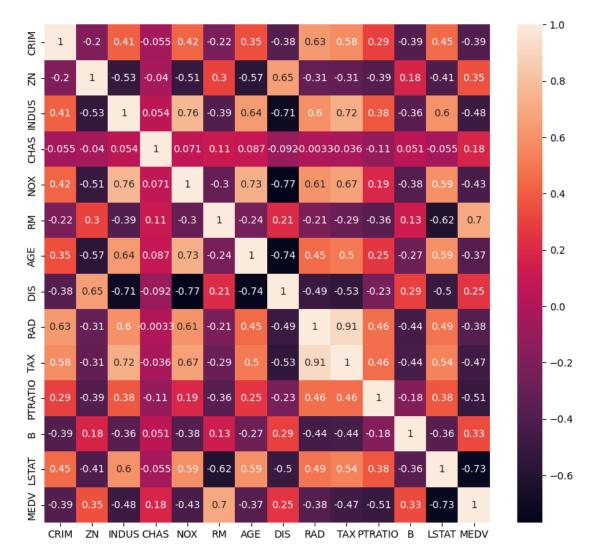
```
CRIM
            20
\mathsf{ZN}
            20
INDUS
            20
CHAS
            20
NOX
             0
RM
             0
AGE
            20
DIS
             0
RAD
             0
TAX
             0
PTRATIO
             0
             0
LSTAT
            20
MEDV
             0
dtype: int64
df.fillna(method ='ffill',inplace=True)
df.isnull().sum()
CRIM
            0
            0
\mathsf{ZN}
INDUS
            0
CHAS
            0
            0
NOX
RM
            0
AGE
            0
            0
DIS
RAD
            0
            0
TAX
PTRATIO
            0
            0
В
LSTAT
            0
MEDV
            0
dtype: int64
sns.set(rc={'figure.figsize':(11.7,8.27)})
sns.displot(df['MEDV'], bins=30)
plt.show()
```



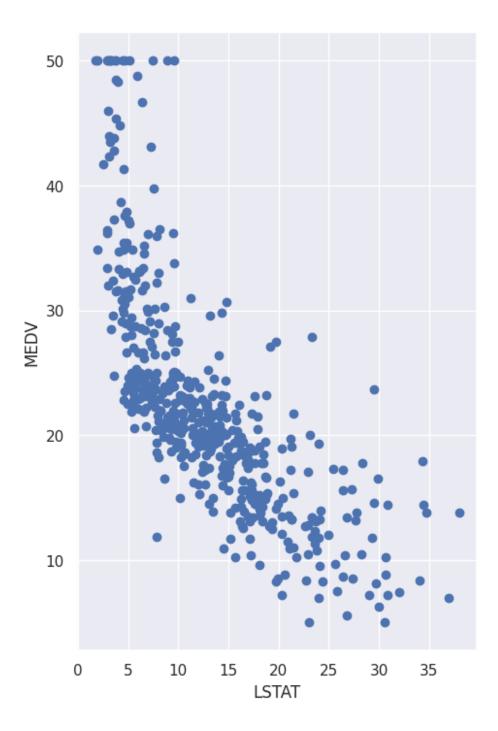
corr = df.corr()
corr

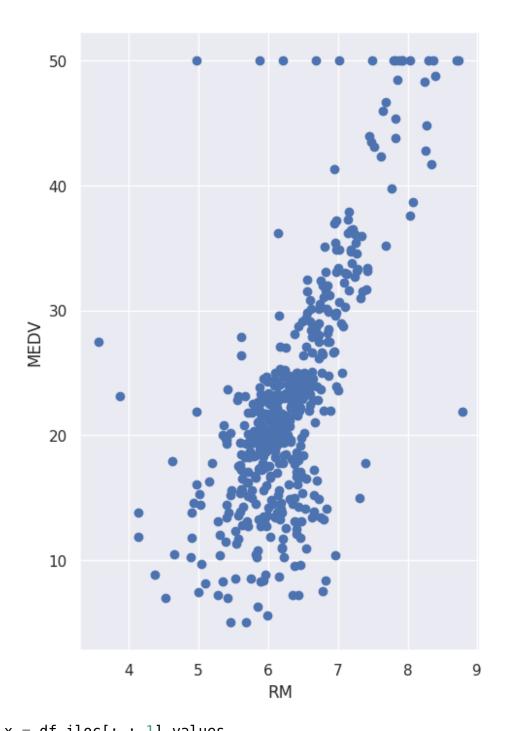
	CRIM	ZN	INDUS	CHAS	NOX	RM	
AGE \							
CRIM 1	1.000000	-0.200017	0.407614	-0.055486	0.422366	-0.221545	
0.349144							
ZN - 6	9.200017	1.000000	-0.526815	-0.040399	-0.511056	0.302417	-
0.565795							
INDUS 0	9.407614	-0.526815	1.000000	0.054064	0.763651	-0.391676	
0.639985							
	0.055486	-0.040399	0.054064	1.000000	0.070867	0.106797	
0.087057							
-	9.422366	-0.511056	0.763651	0.070867	1.000000	-0.302188	
0.729830							
	9.221545	0.302417	-0.391676	0.106797	-0.302188	1.000000	-
0.235960							
	9.349144	-0.565795	0.639985	0.087057	0.729830	-0.235960	
1.000000							
DIS - G	9.379364	0.653589	-0.708027	-0.092318	-0.769230	0.205246	-
0.744977							
RAD 6	9.627112	-0.310159	0.595129	-0.003339	0.611441	-0.209847	

```
0.452431
TAX
         0.584277 -0.313370 0.720760 -0.035822 0.668023 -0.292048
0.503086
PTRATIO 0.290771 -0.385452 0.383248 -0.109451 0.188933 -0.355501
0.254110
        -0.386169 0.175763 -0.356977 0.050608 -0.380051 0.128069 -
0.272501
         0.452563 - 0.408091 \quad 0.601579 - 0.054583 \quad 0.589646 - 0.615892
LSTAT
0.592692
MEDV
        -0.391776   0.353895   -0.483725   0.183844   -0.427321   0.695360   -
0.371413
              DIS
                        RAD
                                  TAX
                                        PTRATIO
                                                        В
                                                              LSTAT
MEDV
CRIM
        -0.379364
                   0.627112
                             0.584277  0.290771  -0.386169  0.452563  -
0.391776
ZN
         0.653589 -0.310159 -0.313370 -0.385452 0.175763 -0.408091
0.353895
INDUS
        -0.708027 0.595129 0.720760 0.383248 -0.356977
                                                           0.601579 -
0.483725
CHAS
        -0.092318 -0.003339 -0.035822 -0.109451 0.050608 -0.054583
0.183844
NOX
        -0.769230
                   0.611441   0.668023   0.188933   -0.380051   0.589646   -
0.427321
RM
         0.205246 -0.209847 -0.292048 -0.355501 0.128069 -0.615892
0.695360
AGE
        -0.744977 0.452431 0.503086 0.254110 -0.272501 0.592692 -
0.371413
         1.000000 - 0.494588 - 0.534432 - 0.232471 0.291512 - 0.501481
DIS
0.249929
                            RAD
        -0.494588
                   1.000000
0.381626
                   0.910228
                             1.000000
                                      0.460853 -0.441808
TAX
        -0.534432
                                                           0.542354 -
0.468536
                                      1.000000 -0.177383 0.377126 -
PTRATIO -0.232471
                   0.464741
                             0.460853
0.507787
         0.291512 - 0.444413 - 0.441808 - 0.177383 1.000000 - 0.363509
В
0.333461
LSTAT
        -0.501481  0.487754  0.542354  0.377126 -0.363509
                                                           1.000000 -
0.732489
         0.249929 \ -0.381626 \ -0.468536 \ -0.507787 \ \ 0.333461 \ -0.732489
MEDV
1.000000
plt.subplots(figsize=(10,9))
sns.heatmap(corr,annot = True)
<Axes: >
```



```
features = ['LSTAT','RM']
target = df['MEDV']
for i,col in enumerate(features):
  plt.subplot(1 ,len(features), i+1)
  x = df[col]
  y = target
  plt.scatter(x,y)
  plt.xlabel(col)
  plt.ylabel('MEDV')
  plt.show()
```





x = df.iloc[:,:-1].values
y = df.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=0.2,random_state=0)

from sklearn.linear_model import LinearRegression
reg = LinearRegression()

```
req.fit(x train,y_train)
print(reg.coef ,reg.intercept )
[-1.22341809e-01 4.41363869e-02 -1.34419395e-03 2.55872583e+00
 -1.56313766e+01 3.69035255e+00 -6.13607953e-03 -1.41415596e+00
  2.38632648e-01 -1.09194727e-02 -1.03687110e+00 8.11270396e-03
 -4.90318222e-01] 38.188474058519745
from sklearn.metrics import r2 score
y pred train = reg.predict(x train)
r2 score(y train,y pred train)
0.7731269304116333
y_pred_test = reg.predict(x_test)
r2_score(y_test,y_pred_test)
0.5724529611713153
reg.predict(df.iloc[3:4,0:13])
/usr/local/lib/python3.10/dist-packages/sklearn/base.py:432:
UserWarning: X has feature names, but LinearRegression was fitted
without feature names
 warnings.warn(
array([28.65583765])
df.iloc[3:4,:]
      CRIM
            ZN INDUS CHAS
                               NOX
                                            AGE
                                                     DIS
                                                         RAD
                                                              TAX
                                        RM
PTRATIO \
                        0.0 0.458 6.998 45.8 6.0622
3 0.03237
           0.0
                 2.18
                                                            3
                                                              222
18.7
        B LSTAT MEDV
3 394.63
           2.94 33.4
from sklearn.metrics import mean squared error
rmse = np.sqrt(mean squared error(y train,y pred train))
rmse
4.395090120040799
```