```
import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
```

data = pd.read_csv("housing.csv")

data

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

506 rows × 14 columns

X=data.iloc[:,:-1].values
Y=data.iloc[:,-1].values

data.head()

	CRIM	ZN	INDUS	CHAS	NOX	RM	AGE	DIS	RAD	TAX	PTRATIO	В	LSTAT	MEDV
0	0.00632	18.0	2.31	0.0	0.538	6.575	65.2	4.0900	1	296	15.3	396.90	4.98	24.0
1	0.02731	0.0	7.07	0.0	0.469	6.421	78.9	4.9671	2	242	17.8	396.90	9.14	21.6
2	0.02729	0.0	7.07	0.0	0.469	7.185	61.1	4.9671	2	242	17.8	392.83	4.03	34.7
3	0.03237	0.0	2.18	0.0	0.458	6.998	45.8	6.0622	3	222	18.7	394.63	2.94	33.4
4	0.06905	0.0	2.18	0.0	0.458	7.147	54.2	6.0622	3	222	18.7	396.90	NaN	36.2

data.info()

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

Data	COTUIIII	(cocar 14 corunn	٥).
#	Column	Non-Null Count	Dtype
0	CRIM	486 non-null	float64
1	ZN	486 non-null	float64
2	INDUS	486 non-null	float64
3	CHAS	486 non-null	float64
4	NOX	506 non-null	float64
5	RM	506 non-null	float64
6	AGE	486 non-null	float64
7	DIS	506 non-null	float64

```
8
     RAD
              506 non-null
                              int64
 9
     TAX
              506 non-null
                              int64
 10
    PTRATIO
              506 non-null
                              float64
 11
              506 non-null
                              float64
 12
    LSTAT
              486 non-null
                              float64
13 MEDV
              506 non-null
                              float64
dtypes: float64(12), int64(2)
memory usage: 55.5 KB
```

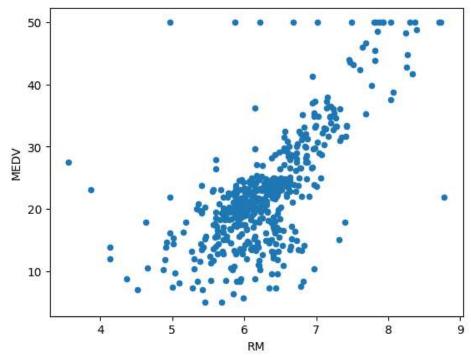
data['CRIM'].fillna(data['CRIM'].mean() , inplace = True)
data['ZN'].fillna(data['ZN'].mean() , inplace = True)
data['INDUS'].fillna(data['INDUS'].mean() , inplace = True)
data['CHAS'].fillna(data['CHAS'].mean() , inplace = True)
data['AGE'].fillna(data['AGE'].mean() , inplace = True)
data['LSTAT'].fillna(data['LSTAT'].mean() , inplace = True)

data.isnull().sum()

CRIM 0 ΖN 0 **INDUS** 0 CHAS 0 NOX 0 RM0 AGE 0 DIS 0 RAD 0 TAX 0 PTRATIO 0 **LSTAT** MEDV dtype: int64

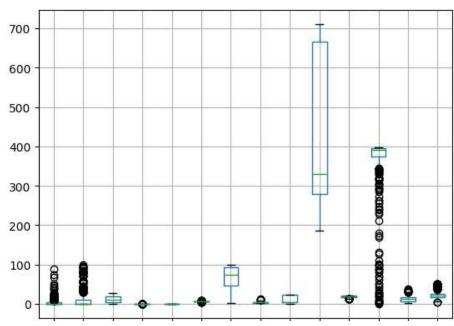
data.plot.scatter('RM', 'MEDV')

<Axes: xlabel='RM', ylabel='MEDV'>



data.boxplot(column_names, rot=15)

<Axes: >



x=data.iloc[:,:-1].values
y=data.iloc[:,-1].values

plt.figure(figsize=(15,10))
sns.heatmap(data.select_dtypes(include=['int','float']).corr(),annot=True,center = 2)
plt.show()

CRIM	1	-0.18	0.39	-0.052	0.41	-0.22	0.34	-0.37	0.61	0.57	0.27	-0.37	0.43	-0.3
NZ -	-0.18	1.	-0.51	-0.036	-0.5	0.32	-0.54	0.64	-0.31	-0.31	-0.4	0.17	-0.41	0.3
SNDNS	0.39	-0.51	1	0.058	0.74	-0.38	0.61	-0.7	0.59	0.72	0.38	-0.35	0.57	-0.4
CHAS	-0.052	-0.036	0.058	1	0.073	0.1	0.075	-0.092	0.0014	-0.031	-0.11	0.05	-0.046	0.1
XON -	0.41	-0.5	0.74	0.073	1	-0.3	0.71	-0.77	0.61	0.67	0.19	-0.38	0.57	-0.4
RM -	-0.22	0.32	-0.38	0.1	-0.3	1	-0.24	0.21	-0.21	-0.29	-0.36	0.13	-0.6	0.
AGE	0.34	-0.54	0.61	0.075	0.71	-0.24	1	-0.72	0.45	0.5	0.26	-0.27	0.57	-0.3
DIS	-0.37	0.64	-0.7	-0.092	-0.77	0.21	-0.72	1	-0.49	-0.53	-0.23	0.29	-0.48	0.2
RAD	0.61	-0.31	0.59	0.0014	0.61	-0.21	0.45	-0.49	1	0.91	0.46	-0.44	0.47	-0.3
TAX -	0.57	-0.31	0.72	-0.031	0.67	-0.29	0.5	-0.53	0.91	1	0.46	-0.44	0.52	-0.4
PTRATIO	0.27	-0.4	0.38	-0.11	0.19	-0.36	0.26	-0.23	0.46	0.46	1	-0.18	0.37	-0.5
В -	-0.37	0.17	-0.35	0.05	-0.38	0.13	-0.27	0.29	-0.44	-0.44	-0.18	1	-0.37	0.3
LSTAT	0.43	-0.41	0.57	-0.046	0.57	-0.6	0.57	-0.48	0.47	0.52	0.37	-0.37	1	-0.
2	0.20	0.27	0.40	0.10	0.42	0.7	0.20	0.25	0.20	0.47	0.51	0.32	0.72	,

```
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=1)
print(X_train.shape)
print(X_test.shape)
print(y_train.shape)
print(y_test.shape)
     (404, 13)
     (102, 13)
     (404,)
     (102,)
X = data[['LSTAT', 'RM', 'PTRATIO', 'INDUS', 'TAX', 'NOX' , 'RAD' ,
'AGE' , 'CRIM' , 'ZN']]
Y = data['MEDV']
seed= 1
X_train , X_test, Y_train , Y_test = train_test_split(X, Y,
test size=0.20, random state=seed)
X.shape
     (506, 10)
Y.shape
     (506,)
from sklearn.linear_model import LinearRegression
LR=LinearRegression()
LR.fit(X_train , Y_train)
LinearRegression()
     ▼ LinearRegression
     LinearRegression()
y_pred= LR.predict(X_test)
from sklearn import metrics
import numpy as np
print("Mean Absolute Error:",metrics.mean_absolute_error(y_test,y_pred))
print("Mean Squred Error:",metrics.mean_squared_error(y_test,y_pred))
print("Root Mean Squred Error:",metrics.mean_squared_error(y_test,y_pred))
     Mean Absolute Error: 4.311333848096257
     Mean Squred Error: 29.58597268132346
     Root Mean Squred Error: 29.58597268132346
print("Mean Absolute Error:",metrics.mean_absolute_error(y_test,y_pred))
     Mean Absolute Error: 4.311333848096257
plt.scatter(y_test, y_pred, c = 'Blue')
plt.xlabel("Price: in $1000's")
```

nlt.vlahel("Predicted value")

