# **D7**

In [1]: import pandas as pd
 import numpy as np
 import matplotlib.pyplot as plt
 import seaborn as sns

In [2]: df=pd.read\_csv(r"C:\Users\user\Downloads\11\_winequality-red.csv")
df

## Out[2]:

	fixed acidity	volatile acidity	citric acid	residual sugar	chlorides	free sulfur dioxide	total sulfur dioxide	density	рН	sulphates	alco
0	7.4	0.700	0.00	1.9	0.076	11.0	34.0	0.99780	3.51	0.56	
1	7.8	0.880	0.00	2.6	0.098	25.0	67.0	0.99680	3.20	0.68	
2	7.8	0.760	0.04	2.3	0.092	15.0	54.0	0.99700	3.26	0.65	
3	11.2	0.280	0.56	1.9	0.075	17.0	60.0	0.99800	3.16	0.58	
4	7.4	0.700	0.00	1.9	0.076	11.0	34.0	0.99780	3.51	0.56	
				•••							
1594	6.2	0.600	80.0	2.0	0.090	32.0	44.0	0.99490	3.45	0.58	1
1595	5.9	0.550	0.10	2.2	0.062	39.0	51.0	0.99512	3.52	0.76	1
1596	6.3	0.510	0.13	2.3	0.076	29.0	40.0	0.99574	3.42	0.75	1
1597	5.9	0.645	0.12	2.0	0.075	32.0	44.0	0.99547	3.57	0.71	1
1598	6.0	0.310	0.47	3.6	0.067	18.0	42.0	0.99549	3.39	0.66	1

1599 rows × 12 columns

# In [3]: df.head(10)

### Out[3]:

	fixed acidity	volatile acidity	citric acid	residual sugar	chlorides	free sulfur dioxide	total sulfur dioxide	density	рН	sulphates	alcohol
0	7.4	0.70	0.00	1.9	0.076	11.0	34.0	0.9978	3.51	0.56	9.4
1	7.8	0.88	0.00	2.6	0.098	25.0	67.0	0.9968	3.20	0.68	9.8
2	7.8	0.76	0.04	2.3	0.092	15.0	54.0	0.9970	3.26	0.65	9.8
3	11.2	0.28	0.56	1.9	0.075	17.0	60.0	0.9980	3.16	0.58	9.8
4	7.4	0.70	0.00	1.9	0.076	11.0	34.0	0.9978	3.51	0.56	9.4
5	7.4	0.66	0.00	1.8	0.075	13.0	40.0	0.9978	3.51	0.56	9.4
6	7.9	0.60	0.06	1.6	0.069	15.0	59.0	0.9964	3.30	0.46	9.4
7	7.3	0.65	0.00	1.2	0.065	15.0	21.0	0.9946	3.39	0.47	10.0
8	7.8	0.58	0.02	2.0	0.073	9.0	18.0	0.9968	3.36	0.57	9.5
9	7.5	0.50	0.36	6.1	0.071	17.0	102.0	0.9978	3.35	0.80	10.5
4											•

# In [4]: df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1599 entries, 0 to 1598
Data columns (total 12 columns):

Ducu	coramiis (cocar iz cor	a	
#	Column	Non-Null Count	Dtype
0	fixed acidity	1599 non-null	float64
1	volatile acidity	1599 non-null	float64
2	citric acid	1599 non-null	float64
3	residual sugar	1599 non-null	float64
4	chlorides	1599 non-null	float64
5	free sulfur dioxide	1599 non-null	float64
6	total sulfur dioxide	1599 non-null	float64
7	density	1599 non-null	float64
8	рН	1599 non-null	float64
9	sulphates	1599 non-null	float64
10	alcohol	1599 non-null	float64
11	quality	1599 non-null	int64

dtypes: float64(11), int64(1)

memory usage: 150.0 KB

```
In [5]: df.describe()
```

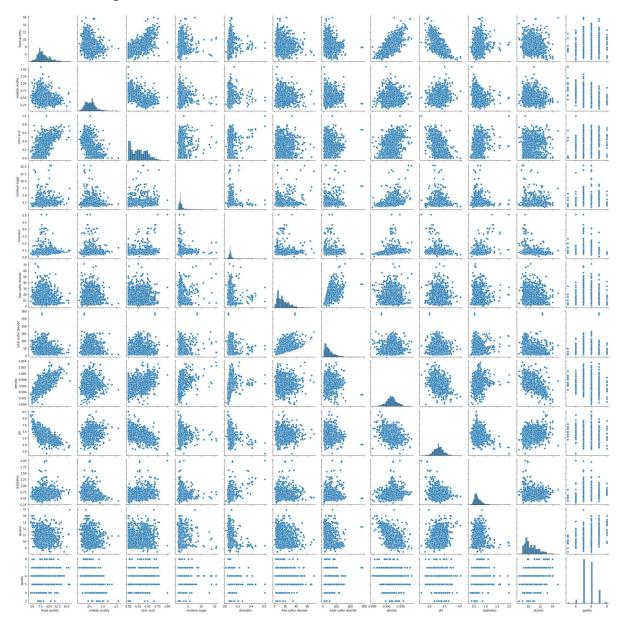
Out[5]:

	fixed acidity	volatile acidity	citric acid	residual sugar	chlorides	free sulfur dioxide	total sulf dioxic
count	1599.000000	1599.000000	1599.000000	1599.000000	1599.000000	1599.000000	1599.00000
mean	8.319637	0.527821	0.270976	2.538806	0.087467	15.874922	46.46779
std	1.741096	0.179060	0.194801	1.409928	0.047065	10.460157	32.89532
min	4.600000	0.120000	0.000000	0.900000	0.012000	1.000000	6.00000
25%	7.100000	0.390000	0.090000	1.900000	0.070000	7.000000	22.00000
50%	7.900000	0.520000	0.260000	2.200000	0.079000	14.000000	38.00000
75%	9.200000	0.640000	0.420000	2.600000	0.090000	21.000000	62.00000
max	15.900000	1.580000	1.000000	15.500000	0.611000	72.000000	289.00000

```
In [6]: df.columns
```

In [7]: sns.pairplot(df)

Out[7]: <seaborn.axisgrid.PairGrid at 0x26e27b9b9d0>

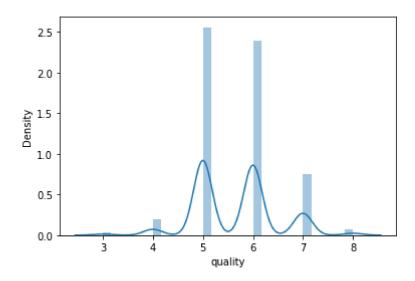


### In [8]: | sns.distplot(df['quality'])

C:\ProgramData\Anaconda3\lib\site-packages\seaborn\distributions.py:2557: Fut ureWarning: `distplot` is a deprecated function and will be removed in a futu re version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

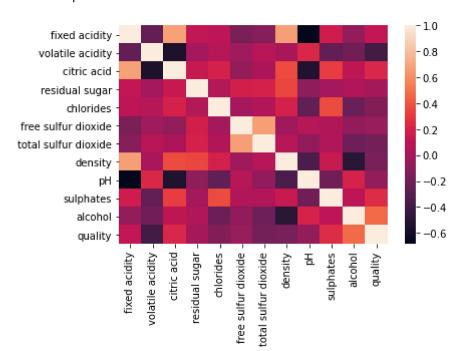
warnings.warn(msg, FutureWarning)

Out[8]: <AxesSubplot:xlabel='quality', ylabel='Density'>



#### In [10]: sns.heatmap(df1.corr())

#### Out[10]: <AxesSubplot:>



```
In [11]: x=df1[['fixed acidity', 'volatile acidity', 'citric acid', 'residual sugar',
                  'chlorides', 'free sulfur dioxide', 'total sulfur dioxide', 'density',
                  'pH', 'sulphates', 'alcohol']]
          y=df1['quality']
In [12]: | from sklearn.model_selection import train_test_split
          x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.3)
In [13]: from sklearn.linear_model import LinearRegression
          lr=LinearRegression()
          lr.fit(x_train,y_train)
Out[13]: LinearRegression()
In [14]: |print(lr.intercept_)
          5.845811866151241
In [15]: coeff = pd.DataFrame(lr.coef_,x.columns,columns=['Co-efficient'])
          coeff
Out[15]:
                            Co-efficient
                fixed acidity
                             -0.007488
              volatile acidity
                             -1.026444
                  citric acid
                             -0.054716
              residual sugar
                              0.000125
                  chlorides
                             -1.842561
           free sulfur dioxide
                              0.005322
           total sulfur dioxide
                             -0.003336
                    density
                             -0.883610
```

pН

sulphates

alcohol

-0.648912

0.905897

0.295879

```
In [16]:
         prediction=lr.predict(x_test)
          plt.scatter(y_test,prediction)
Out[16]: <matplotlib.collections.PathCollection at 0x26e3046c400>
           7.0
           6.5
           6.0
           5.5
           5.0
           4.5
                                         5.5
                              4.5
               3.0
                    3.5
                         4.0
                                    5.0
                                              6.0
                                                    6.5
                                                         7.0
In [17]: print(lr.score(x_test,y_test))
          0.3504856066186194
In [18]: from sklearn.linear model import Ridge, Lasso
In [19]: | rr=Ridge(alpha=10)
          rr.fit(x_train,y_train)
Out[19]: Ridge(alpha=10)
In [20]: |rr.score(x_test,y_test)
Out[20]: 0.33712542505350773
In [21]: la=Lasso(alpha=10)
          la.fit(x_train,y_train)
Out[21]: Lasso(alpha=10)
In [22]: la.score(x_test,y_test)
Out[22]: -0.004986667842787673
In [23]: | from sklearn.linear_model import ElasticNet
          en=ElasticNet()
          en.fit(x_train,y_train)
Out[23]: ElasticNet()
In [24]: |print(en.coef_)
                       -0.
                                     0.
                                                  0.
                                                              -0.
                                                                           0.00152978
                                    -0.
                                                  0.
                                                               0.
           -0.00433797 -0.
                                                                         ]
```