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

localhost:8888/notebooks/D9-7.ipynb

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 [6]: df.info()

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

#	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 [7]: df.describe()
```

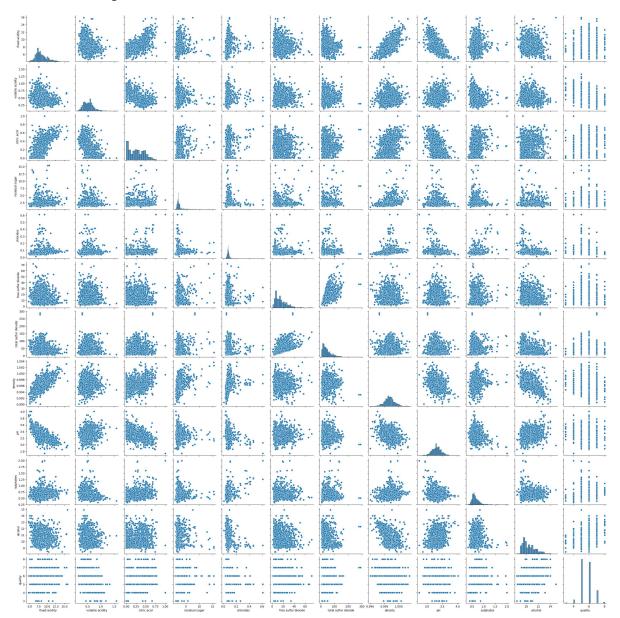
Out[7]:

	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
4							

```
In [4]: df.columns
```

In [5]: sns.pairplot(df)

Out[5]: <seaborn.axisgrid.PairGrid at 0x2489f53b6d0>



```
In [6]: | sns.distplot(df['quality'])
```

C:\Users\user\AppData\Local\Temp\ipykernel 4092\3043670886.py:1: UserWarning:

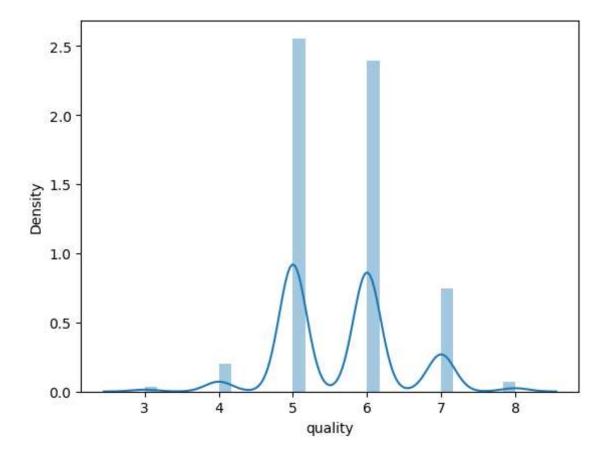
`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751 (https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751)

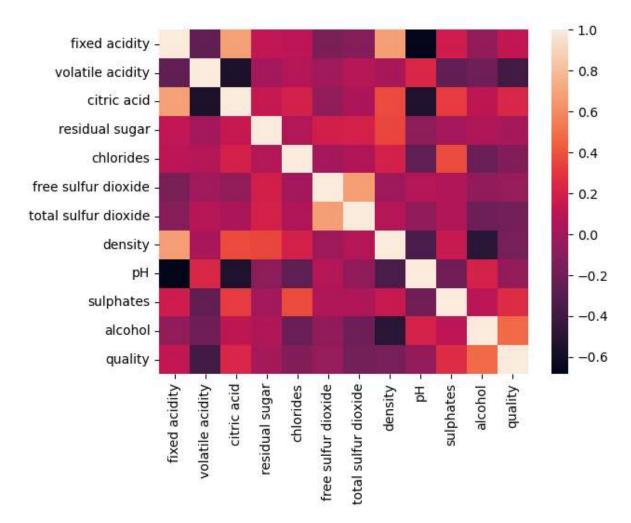
sns.distplot(df['quality'])

Out[6]: <Axes: xlabel='quality', ylabel='Density'>



```
In [8]: sns.heatmap(df1.corr())
```

Out[8]: <Axes: >



- In [11]: 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)
- Out[12]: v LinearRegression LinearRegression()
- In [13]: print(lr.intercept_)

1.8371596870274778

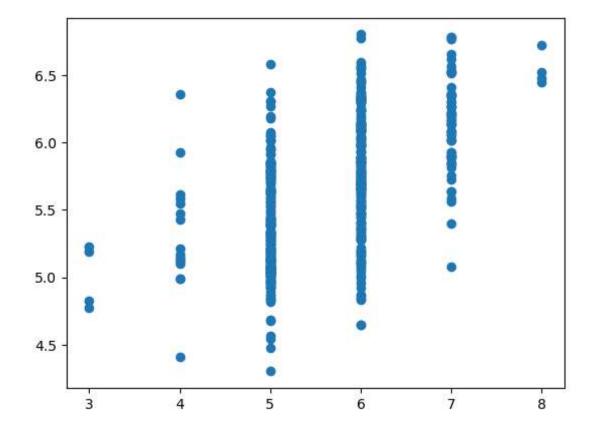
```
In [14]: coeff = pd.DataFrame(lr.coef_,x.columns,columns=['Co-efficient'])
coeff
```

Out[14]:

	Co-efficient
fixed acidity	-0.003015
volatile acidity	-1.193810
citric acid	-0.303974
residual sugar	0.015383
chlorides	-1.856641
free sulfur dioxide	0.004369
total sulfur dioxide	-0.004300
density	3.309743
рН	-0.651472
sulphates	0.939148
alcohol	0.290871

```
In [15]: prediction=lr.predict(x_test)
plt.scatter(y_test,prediction)
```

Out[15]: <matplotlib.collections.PathCollection at 0x248b09268d0>



```
In [16]: print(lr.score(x_test,y_test))
         0.30762818465530084
In [17]: from sklearn.linear_model import Ridge,Lasso
In [18]: | rr=Ridge(alpha=10)
         rr.fit(x_train,y_train)
Out[18]:
               Ridge
          Ridge(alpha=10)
In [19]: rr.score(x_test,y_test)
Out[19]: 0.3098447423349697
In [20]: la=Lasso(alpha=10)
         la.fit(x_train,y_train)
Out[20]:
                Lasso
          Lasso(alpha=10)
In [21]: la.score(x_test,y_test)
Out[21]: -2.275654888835937e-05
In [ ]:
```