import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model\_selection import train\_test\_split
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import accuracy\_score

In [2]: wine\_dataset=pd.read\_csv('winequality-red.csv')

In [3]: wine\_dataset.shape

Out[3]: (1599, 12)

In [4]: wine\_dataset.head()

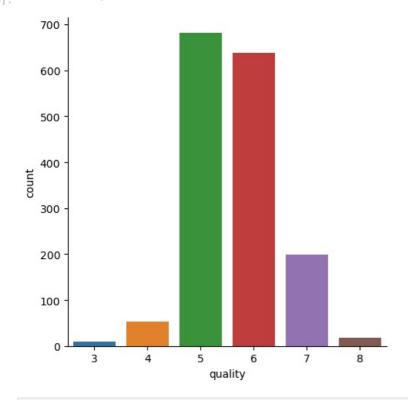
Out[4]:		fixed acidity	volatile acidity	citric acid	residual sugar	chlorides	free sulfur dioxide	total sulfur dioxide	density	рН	sulphates	alcohol	quality
	0	7.4	0.70	0.00	1.9	0.076	11.0	34.0	0.9978	3.51	0.56	9.4	5
	1	7.8	0.88	0.00	2.6	0.098	25.0	67.0	0.9968	3.20	0.68	9.8	5
	2	7.8	0.76	0.04	2.3	0.092	15.0	54.0	0.9970	3.26	0.65	9.8	5
	3	11.2	0.28	0.56	1.9	0.075	17.0	60.0	0.9980	3.16	0.58	9.8	6
	4	7.4	0.70	0.00	1.9	0.076	11.0	34.0	0.9978	3.51	0.56	9.4	5

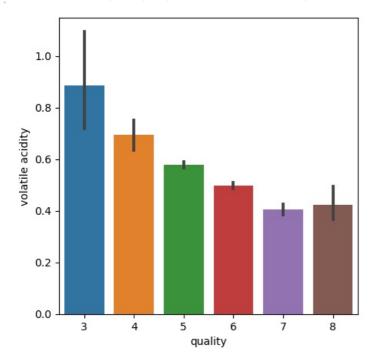
In [5]: wine\_dataset.describe()

Out[5]:		fixed acidity	volatile acidity	citric acid	residual sugar	chlorides	free sulfur dioxide	total sulfur dioxide	density	рН	sulphates	
	count	1599.000000	1599.000000	1599.000000	1599.000000	1599.000000	1599.000000	1599.000000	1599.000000	1599.000000	1599.000000	1
	mean	8.319637	0.527821	0.270976	2.538806	0.087467	15.874922	46.467792	0.996747	3.311113	0.658149	
	std	1.741096	0.179060	0.194801	1.409928	0.047065	10.460157	32.895324	0.001887	0.154386	0.169507	
	min	4.600000	0.120000	0.000000	0.900000	0.012000	1.000000	6.000000	0.990070	2.740000	0.330000	
	25%	7.100000	0.390000	0.090000	1.900000	0.070000	7.000000	22.000000	0.995600	3.210000	0.550000	
	50%	7.900000	0.520000	0.260000	2.200000	0.079000	14.000000	38.000000	0.996750	3.310000	0.620000	
	75%	9.200000	0.640000	0.420000	2.600000	0.090000	21.000000	62.000000	0.997835	3.400000	0.730000	
	max	15.900000	1.580000	1.000000	15.500000	0.611000	72.000000	289.000000	1.003690	4.010000	2.000000	

In [6]: sns.catplot(x='quality',data=wine\_dataset,kind='count')

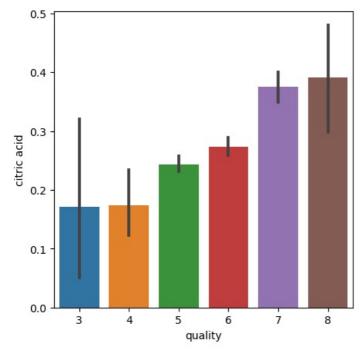
Out[6]: <seaborn.axisgrid.FacetGrid at 0x21a129ffa00>



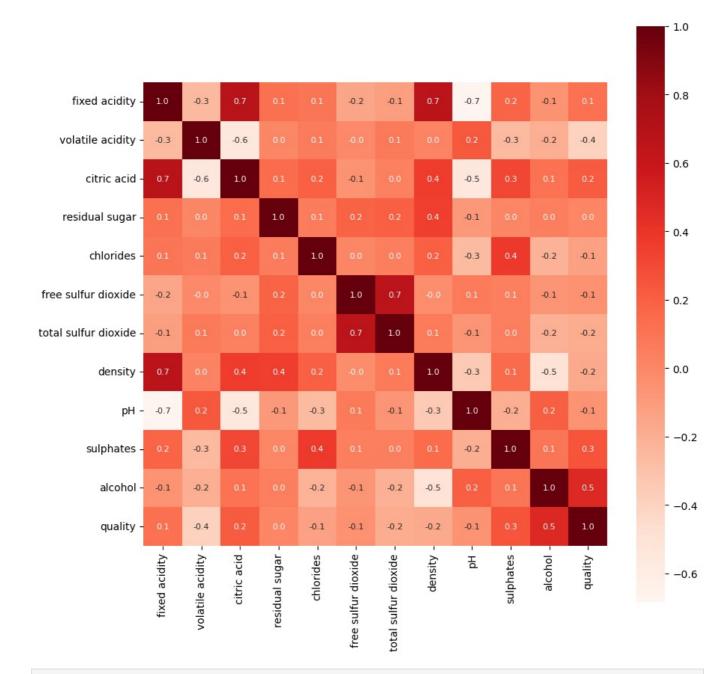


```
In [8]: plot=plt.figure(figsize=(5,5))
sns.barplot(x='quality',y='citric acid',data=wine_dataset)
```

Out[8]: <Axes: xlabel='quality', ylabel='citric acid'>



```
In [9]: correlation=wine_dataset.corr()
In [13]: plt.figure(figsize=(10,10))
    sns.heatmap(correlation,cbar=True, square=True, fmt='.1f',annot=True, annot_kws={'size':8},cmap='Reds')
Out[13]: <Axes: >
```



In [14]: x=wine\_dataset.drop('quality',axis=1)
In [15]: print(x)

```
0
                         7.4
                                          0.700
                                                        0.00
                                                                          1.9
                                                                                   0.076
         1
                          7.8
                                          0.880
                                                        0.00
                                                                          2.6
                                                                                   0.098
                                          0.760
         2
                         7.8
                                                        0.04
                                                                          2.3
                                                                                   0.092
         3
                                                                                   0.075
                         11.2
                                          0.280
                                                        0.56
                                                                          1.9
         4
                         7.4
                                          0.700
                                                        0.00
                                                                          1.9
                                                                                   0.076
                                                                                   0.090
         1594
                         6.2
                                          0.600
                                                        0.08
                                                                          2.0
         1595
                         5.9
                                          0.550
                                                        0.10
                                                                          2.2
                                                                                   0.062
         1596
                          6.3
                                          0.510
                                                        0.13
                                                                          2.3
                                                                                   0.076
         1597
                          5.9
                                          0.645
                                                        0.12
                                                                          2.0
                                                                                   0.075
         1598
                                          0.310
                                                        0.47
                                                                                   0.067
                         6.0
                                                                          3.6
               free sulfur dioxide total sulfur dioxide density
                                                                       рΗ
                                                                          sulphates
         0
                                                     34.0 0.99780
                                                                     3.51
                                                                                0.56
                               11.0
                                                     67.0 0.99680
                                                                                0.68
         1
                               25.0
                                                                     3.20
         2
                               15.0
                                                     54.0
                                                          0.99700
                                                                     3.26
                                                                                0.65
         3
                                                     60.0 0.99800
                               17.0
                                                                     3.16
                                                                                0.58
         4
                                                     34.0 0.99780
                               11.0
                                                                     3.51
                                                                                0.56
         1594
                               32.0
                                                     44.0 0.99490
                                                                     3.45
                                                                                0.58
         1595
                               39.0
                                                     51.0
                                                           0.99512
                                                                     3.52
                                                                                0.76
                               29.0
         1596
                                                     40.0 0.99574
                                                                                0.75
                                                                     3 42
         1597
                               32.0
                                                     44.0 0.99547
                                                                     3.57
                                                                                0.71
         1598
                               18.0
                                                     42.0 0.99549
                                                                     3.39
                                                                                0.66
               alcohol
         0
                   9.4
         1
                   9.8
         2
                   9.8
         3
                   9.8
         4
                   9.4
         1594
                  10.5
         1595
                  11.2
         1596
                  11.0
         1597
                  10.2
         1598
                  11.0
         [1599 rows x 11 columns]
In [17]: y=wine dataset['quality'].apply(lambda y_value: 1 if y value>=7 else 0)
In [18]: print(y)
         0
                 0
                 0
         1
         2
                 0
         3
                 0
         4
                 0
         1594
                 0
         1595
                 0
         1596
                 0
         1597
                 0
         1598
                 0
         Name: quality, Length: 1599, dtype: int64
In [20]: x_train, x_test,y_train, y_test=train_test_split(x,y,test_size=0.2,random_state=3)
In [22]: print(y.shape,y_train.shape,y_test.shape)
         (1599,) (1279,) (320,)
In [23]: model=RandomForestClassifier()
In [24]: model.fit(x_train,y_train)
Out[24]: ▼ RandomForestClassifier
         RandomForestClassifier()
         x test prediction=model.predict(x test)
In [25]:
         test_data_accuracy=accuracy_score(x_test_prediction,y_test)
In [26]: print('Accuracy:',test_data_accuracy)
         Accuracy: 0.921875
In [29]: input data= (7.3,0.65,0.0,1.2,0.065,15.0,21.0,0.9946,3.39,0.47,10.0)
         input_data_as_numpy_array=np.asarray(input_data)
         input data reshaped=input data as numpy array.reshape(1,-1)
         prediction=model.predict(input_data_reshaped)
         print(prediction)
         if (prediction[0]==1):
```

fixed acidity volatile acidity citric acid residual sugar chlorides \

```
print("Good Quality Wine")
else:
    print("Bad Quality Wine")

[1]
Good Quality Wine
    C:\ProgramData\anaconda3\lib\site-packages\sklearn\base.py:420: UserWarning: X does not have valid feature name s, but RandomForestClassifier was fitted with feature names
    warnings.warn(
In []:
```

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