In [1]: 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("winequalitydata.csv")

In [3]: wine\_dataset.head()

## 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
4											•

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

## Out[4]:

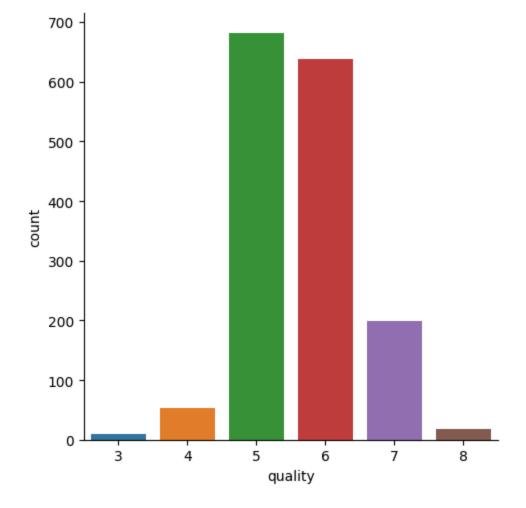
	fixed acidity	volatile acidity	citric acid	residual sugar	chlorides	free sulfur dioxide	total sulf dioxi
count	1599.000000	1599.000000	1599.000000	1599.000000	1599.000000	1599.000000	1599.0000
mean	8.319637	0.527821	0.270976	2.538806	0.087467	15.874922	46.4677
std	1.741096	0.179060	0.194801	1.409928	0.047065	10.460157	32.8953
min	4.600000	0.120000	0.000000	0.900000	0.012000	1.000000	6.0000
25%	7.100000	0.390000	0.090000	1.900000	0.070000	7.000000	22.0000
50%	7.900000	0.520000	0.260000	2.200000	0.079000	14.000000	38.0000
75%	9.200000	0.640000	0.420000	2.600000	0.090000	21.000000	62.0000
max	15.900000	1.580000	1.000000	15.500000	0.611000	72.000000	289.0000
4							<b>&gt;</b>

n [5]:	<pre>wine_dataset.isnull().</pre>	sum()	
out[5]:	fixed acidity	0	
	volatile acidity	0	
	citric acid	0	
	residual sugar	0	
	chlorides	0	
	free sulfur dioxide	0	
	total sulfur dioxide	0	
	density	0	
	pH	0	
	sulphates	0	
	alcohol	0	
	quality	0	
	dtype: int64		

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

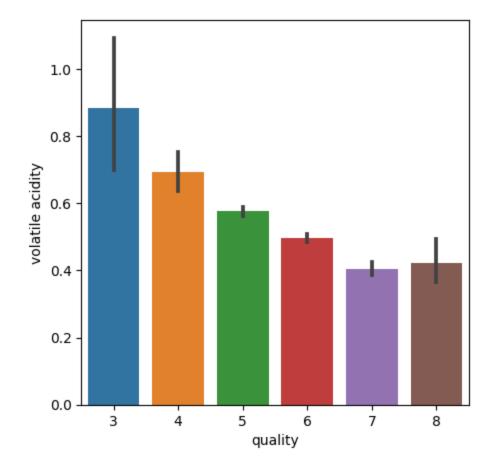
C:\Users\HP\AppData\Local\Programs\Python\Python311\Lib\site-packages\seaborn
\axisgrid.py:118: UserWarning: The figure layout has changed to tight
self.\_figure.tight\_layout(\*args, \*\*kwargs)

Out[6]: <seaborn.axisgrid.FacetGrid at 0x1a91a6ef050>



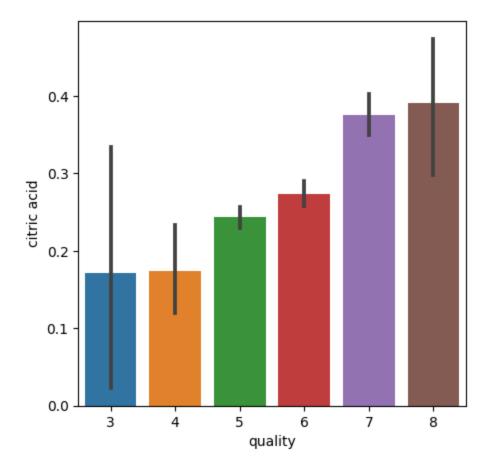
```
In [8]: plot=plt.figure(figsize=(5,5))
sns.barplot(x="quality",y="volatile acidity",data=wine_dataset)
```

Out[8]: <Axes: xlabel='quality', ylabel='volatile acidity'>



```
In [9]: plot=plt.figure(figsize=(5,5))
sns.barplot(x="quality",y="citric acid",data=wine_dataset)
```

Out[9]: <Axes: xlabel='quality', ylabel='citric acid'>



```
In [10]: x=wine_dataset.drop("quality",axis=1)
```

In [11]: y=wine\_dataset["quality"].apply(lambda y\_value:1 if y\_value>=7 else 0 )

```
In [12]: print(x)
print(y)
```

```
fixed acidity volatile acidity citric acid residual sugar
                                                                             chlorides
\
0
                  7.4
                                    0.700
                                                                       1.9
                                                    0.00
                                                                                 0.076
                                                                       2.6
1
                  7.8
                                    0.880
                                                    0.00
                                                                                 0.098
2
                  7.8
                                    0.760
                                                    0.04
                                                                       2.3
                                                                                 0.092
3
                 11.2
                                    0.280
                                                    0.56
                                                                       1.9
                                                                                 0.075
4
                  7.4
                                    0.700
                                                    0.00
                                                                       1.9
                                                                                 0.076
                  . . .
                                      . . .
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                                                                                    . . .
. . .
1594
                  6.2
                                    0.600
                                                    0.08
                                                                       2.0
                                                                                 0.090
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
                  6.0
                                    0.310
                                                    0.47
                                                                       3.6
                                                                                 0.067
      free sulfur dioxide total sulfur dioxide density
                                                                    pH sulphates
0
                        11.0
                                                34.0
                                                       0.99780
                                                                 3.51
                                                                              0.56
1
                       25.0
                                                67.0
                                                       0.99680
                                                                 3.20
                                                                              0.68
2
                       15.0
                                                54.0
                                                       0.99700
                                                                 3.26
                                                                              0.65
3
                        17.0
                                                60.0
                                                       0.99800
                                                                 3.16
                                                                              0.58
4
                       11.0
                                                34.0
                                                       0.99780
                                                                 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
1596
                       29.0
                                                40.0
                                                       0.99574
                                                                 3.42
                                                                              0.75
                       32.0
1597
                                                44.0
                                                       0.99547
                                                                 3.57
                                                                              0.71
1598
                       18.0
                                                42.0 0.99549
                                                                 3.39
                                                                              0.66
      alcohol
0
           9.4
           9.8
1
2
           9.8
3
           9.8
4
           9.4
           . . .
. . .
1594
          10.5
1595
          11.2
1596
          11.0
1597
          10.2
          11.0
1598
[1599 rows x 11 columns]
0
         0
         0
1
2
         0
3
         0
4
         0
        . .
1594
         0
1595
         0
         0
1596
1597
         0
1598
```

Name: quality, Length: 1599, dtype: int64

```
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2,stratify=y,ra
In [16]:
         model=RandomForestClassifier()
In [17]:
In [18]: model.fit(x_train,y_train)
Out[18]: RandomForestClassifier()
         In a Jupyter environment, please rerun this cell to show the HTML representation or trust
         the notebook.
         On GitHub, the HTML representation is unable to render, please try loading this page
         with nbviewer.org.
In [19]: | test_data_prediction=model.predict(x_test)
         test_data_accuracy=accuracy_score(test_data_prediction,y_test)
In [20]: test_data_accuracy
Out[20]: 0.946875
 In [ ]: input_data=input()
         input_data_numpy_array=np.asarray(input_data)
         input_data_reshape=input_data_numpy_array.reshape(1,-1)
         prediction=model.prediction(input_data_reshape)
```