```
import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import accuracy_score

wine_dataset = pd.read_csv('/content/winequality-red (1).csv')
wine_dataset.shape

> (1599, 12)
wine_dataset.head()
```

→		fixed acidity	volatile acidity	citric acid	residual sugar	chlorides	free sulfur dioxide		density	рН	sulph
	0	7.4	0.70	0.00	1.9	0.076	11.0	34.0	0.9978	3.51	
	1	7.8	0.88	0.00	2.6	0.098	25.0	67.0	0.9968	3.20	
	2	7.8	0.76	0.04	2.3	0.092	15.0	54.0	0.9970	3.26	
	3	11.2	0.28	0.56	1.9	0.075	17.0	60.0	0.9980	3.16	
	4										+

Next steps: Generate code wine_dataset plots

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wine_dataset.describe()



	fixed acidity	volatile acidity	citric acid	residual sugar	chlorides	free sulfur dioxide	s ib
count	1599.000000	1599.000000	1599.000000	1599.000000	1599.000000	1599.000000	1599.0
mean	8.319637	0.527821	0.270976	2.538806	0.087467	15.874922	46.4
std	1.741096	0.179060	0.194801	1.409928	0.047065	10.460157	32.8
min	4.600000	0.120000	0.000000	0.900000	0.012000	1.000000	6.0
25%	7.100000	0.390000	0.090000	1.900000	0.070000	7.000000	22.0
50%	7.900000	0.520000	0.260000	2.200000	0.079000	14.000000	38.0
75%	9.200000	0.640000	0.420000	2.600000	0.090000	21.000000	62.0
max	15.900000	1.580000	1.000000	15.500000	0.611000	72.000000	289.0

wine_dataset.info()

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

Column	Non-Null Count	Dtype
fixed acidity	1599 non-null	float64
volatile acidity	1599 non-null	float64
citric acid	1599 non-null	float64
residual sugar	1599 non-null	float64
chlorides	1599 non-null	float64
free sulfur dioxide	1599 non-null	float64
total sulfur dioxide	1599 non-null	float64
density	1599 non-null	float64
рН	1599 non-null	float64
sulphates	1599 non-null	float64
alcohol	1599 non-null	float64
quality	1599 non-null	int64
	fixed acidity volatile acidity citric acid residual sugar chlorides free sulfur dioxide total sulfur dioxide density pH sulphates alcohol	fixed acidity volatile acidity citric acid residual sugar chlorides free sulfur dioxide total sulfur dioxide density pH sulphates alcohol 1599 non-null

dtypes: float64(11), int64(1)

memory usage: 150.0 KB

wine_dataset.isnull().sum()



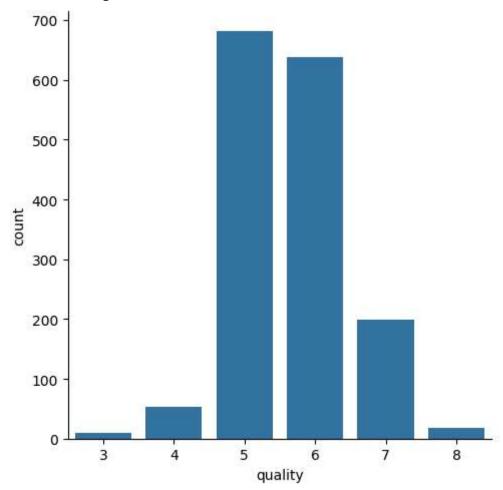
	0
fixed acidity	0
volatile acidity	0
citric acid	0
residual sugar	0
chlorides	0
free sulfur dioxide	0
total sulfur dioxide	0
density	0
рН	0
sulphates	0
alcohol	0
quality	0

dtype: int64

sns.catplot(x='quality', data = wine_dataset, kind = 'count')



<seaborn.axisgrid.FacetGrid at 0x7839bab03be0>



```
fig = plt.figure(figsize=(15,10))
plt.subplot(3,4,1)
sns.barplot(x='quality',y='fixed acidity',data=wine_dataset)

plt.subplot(3,4,2)
sns.barplot(x='quality',y='volatile acidity',data=wine_dataset)

plt.subplot(3,4,3)
sns.barplot(x='quality',y='citric acid',data=wine_dataset)

plt.subplot(3,4,4)
sns.barplot(x='quality',y='residual sugar',data=wine_dataset)

plt.subplot(3,4,5)
sns.barplot(x='quality',y='chlorides',data=wine_dataset)

plt.subplot(3,4,6)
sns.barplot(x='quality',y='free sulfur dioxide',data=wine_dataset)

plt.subplot(3,4,7)
sns.barplot(x='quality',y='total sulfur dioxide',data=wine_dataset)
```

```
plt.subplot(3,4,8)
sns.barplot(x='quality',y='density',data=wine_dataset)

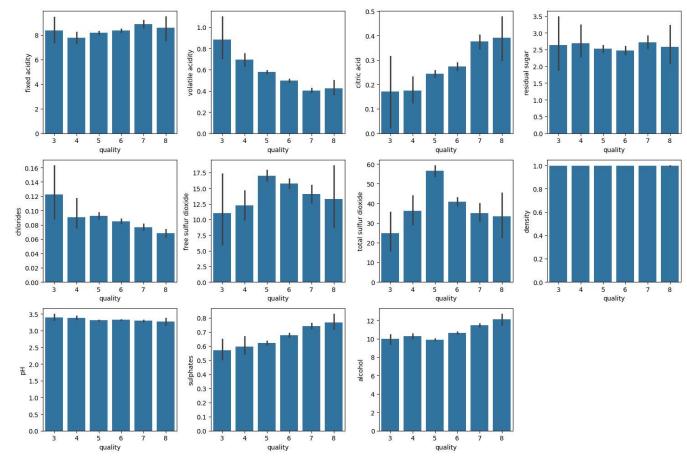
plt.subplot(3,4,9)
sns.barplot(x='quality',y='pH',data=wine_dataset)

plt.subplot(3,4,10)
sns.barplot(x='quality',y='sulphates',data=wine_dataset)

plt.subplot(3,4,11)
sns.barplot(x='quality',y='alcohol',data=wine_dataset)

plt.tight_layout()
```

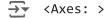


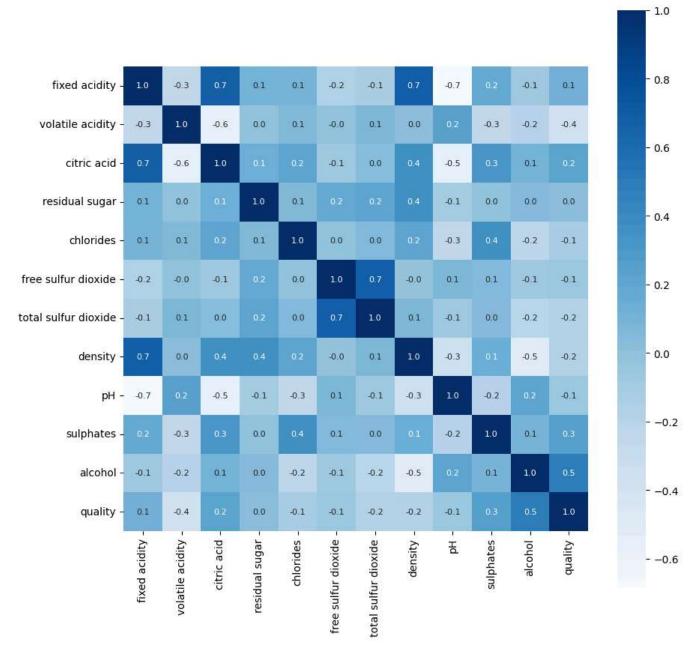


correlation = wine_dataset.corr()

constructing a heatmap to understand the correlation between the columns
plt.figure(figsize=(10,10))

sns.heatmap(correlation, cbar=True, square=True, fmt = '.1f', annot = True, annot_kws={'size





```
X = wine_dataset.drop('quality',axis=1)
x
```

•

		fixed acidity	volatile acidity	citric acid	residual sugar	chlorides	free sulfur dioxide	total sulfur dioxide	density	рН	su
	0	7.4	0.700	0.00	1.9	0.076	11.0	34.0	0.99780	3.51	
	1	7.8	0.880	0.00	2.6	0.098	25.0	67.0	0.99680	3.20	
	2	7.8	0.760	0.04	2.3	0.092	15.0	54.0	0.99700	3.26	
	3	11.2	0.280	0.56	1.9	0.075	17.0	60.0	0.99800	3.16	
	4	7.4	0.700	0.00	1.9	0.076	11.0	34.0	0.99780	3.51	
1	594	6.2	0.600	0.08	2.0	0.090	32.0	44.0	0.99490	3.45	
1	595	5.9	0.550	0.10	2.2	0.062	39.0	51.0	0.99512	3.52	
1	596	6.3	0.510	0.13	2.3	0.076	29.0	40.0	0.99574	3.42	
1	597	5.9	0.645	0.12	2.0	0.075	32.0	44.0	0.99547	3.57	
4	598	6.0	0.310	0.47	3.6	0.067	18.0	42.0	0.99549	3.39	•

Next steps:

Generate code with X

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```
Y = wine_dataset['quality'].apply(lambda y_value: 1 if y_value>=7 else 0 )
Y
```

→ quality