

▼ Importing the Dependencies

```
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
```

▼ Data Collection

```
# loading the dataset to a Pandas DataFrame
wine_dataset = pd.read_csv('/content/winequality-red (2).csv')
```

```
# number of rows & columns in the dataset
wine_dataset.shape
```

```
(1599, 12)
```

```
# first 5 rows of the dataset
wine_dataset.head()
```

| fixed acidity | volatile acidity | citric acid | residual sugar | chlorides | free sulfur dioxide | total sulfur dioxide | density | pH | sulphates | alcohol | quality |
|------------------|---------------------|----------------|-------------------|-----------|------------------------|----------------------------|---------|----|-----------|---------|---------|
|------------------|---------------------|----------------|-------------------|-----------|------------------------|----------------------------|---------|----|-----------|---------|---------|

```
# checking for missing values
```

```
wine_dataset.isnull().sum()
```

```
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
```

▼ Data Analysis and Visulaization

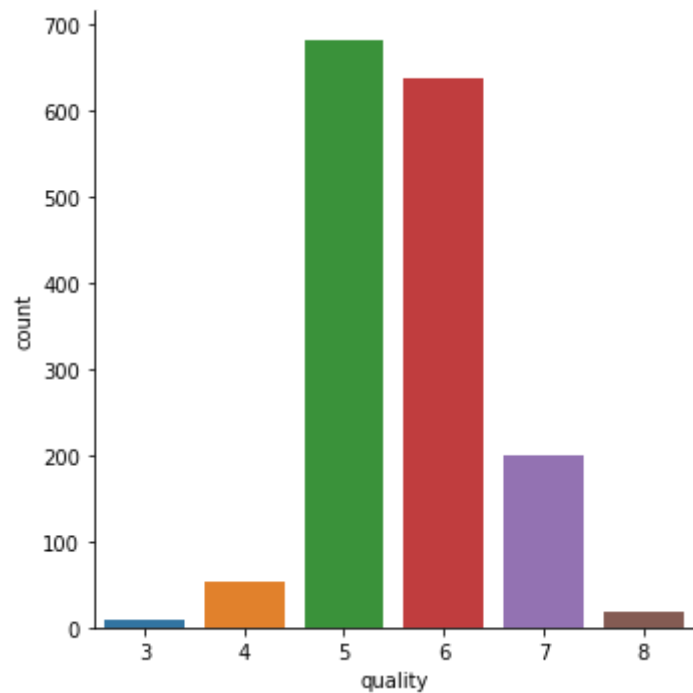
```
# statistical measures of the dataset
```

```
wine_dataset.describe()
```

| | fixed acidity | volatile acidity | citric acid | residual sugar | chlorides | free sulfur dioxide | total sulfur dioxide | density | pH | sulphate |
|--------------|------------------|---------------------|----------------|-------------------|-------------|---------------------------|----------------------------|-------------|-------------|-------------|
| count | 1599.000000 | 1599.000000 | 1599.000000 | 1599.000000 | 1599.000000 | 1599.000000 | 1599.000000 | 1599.000000 | 1599.000000 | 1599.000000 |
| mean | 8.319637 | 0.527821 | 0.270976 | 2.538806 | 0.087467 | 15.874922 | 46.467792 | 0.996747 | 3.311113 | 0.658 |

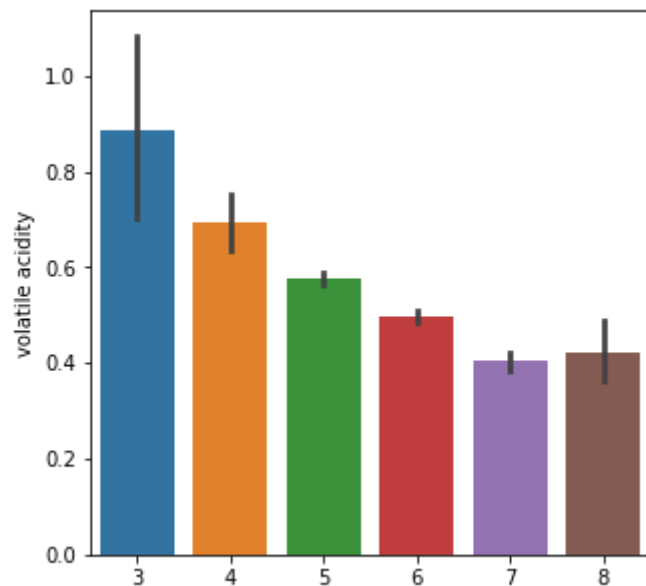
```
# number of values for each quality
sns.catplot(x='quality', data = wine_dataset, kind = 'count')
```

<seaborn.axisgrid.FacetGrid at 0x7ff9f38ac590>



```
# volatile acidity vs Quality
plot = plt.figure(figsize=(5,5))
sns.barplot(x='quality', y = 'volatile acidity', data = wine_dataset)
```

<matplotlib.axes._subplots.AxesSubplot at 0x7ff9efac2990>



▼ Correlation

1. Positive Correlation
2. Negative Correlation

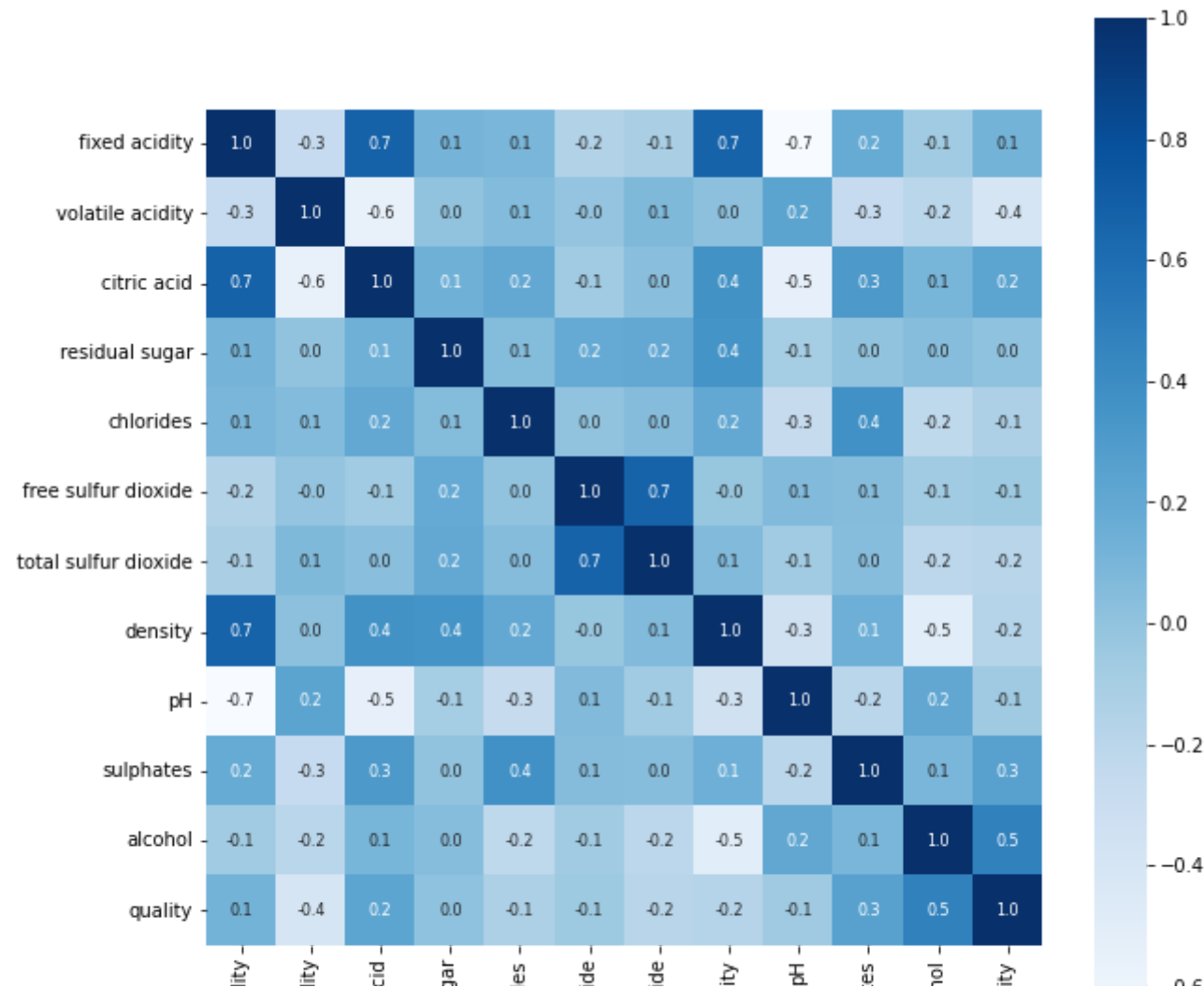
```
correlation = wine_dataset.corr()
```

```
# 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':8}, cmap = 'Blues')
```

<matplotlib.axes._subplots.AxesSubplot at 0x7ff9ee2dd550>



```
# separate the data and Label
```

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

```
print(X)
```

```
fixed acidity  volatile acidity  citric acid  ...  pH  sulphates  alcohol
0            7.4            0.700      0.00  ...  3.51      0.56      9.4
```

| | | | | | | | |
|------|------|-------|------|-----|------|------|------|
| 1 | 7.8 | 0.880 | 0.00 | ... | 3.20 | 0.68 | 9.8 |
| 2 | 7.8 | 0.760 | 0.04 | ... | 3.26 | 0.65 | 9.8 |
| 3 | 11.2 | 0.280 | 0.56 | ... | 3.16 | 0.58 | 9.8 |
| 4 | 7.4 | 0.700 | 0.00 | ... | 3.51 | 0.56 | 9.4 |
| ... | ... | ... | ... | ... | ... | ... | ... |
| 1594 | 6.2 | 0.600 | 0.08 | ... | 3.45 | 0.58 | 10.5 |
| 1595 | 5.9 | 0.550 | 0.10 | ... | 3.52 | 0.76 | 11.2 |
| 1596 | 6.3 | 0.510 | 0.13 | ... | 3.42 | 0.75 | 11.0 |
| 1597 | 5.9 | 0.645 | 0.12 | ... | 3.57 | 0.71 | 10.2 |
| 1598 | 6.0 | 0.310 | 0.47 | ... | 3.39 | 0.66 | 11.0 |

[1599 rows x 11 columns]

▼ Label Binarization

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

```
print(Y)
```

| | |
|------|----|
| 0 | 0 |
| 1 | 0 |
| 2 | 0 |
| 3 | 0 |
| 4 | 0 |
| ... | .. |
| 1594 | 0 |
| 1595 | 0 |
| 1596 | 0 |
| 1597 | 0 |
| 1598 | 0 |

Name: quality, Length: 1599, dtype: int64

▼ Train & Test Split

```
X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.2, random_state=3)
```

```
print(Y.shape, Y_train.shape, Y_test.shape)
```

```
(1599,) (1279,) (320,)
```

▼ Model Training:

Random Forest Classifier

```
model = RandomForestClassifier()
```

```
model.fit(X_train, Y_train)
```

```
RandomForestClassifier(bootstrap=True, ccp_alpha=0.0, class_weight=None,  
                        criterion='gini', max_depth=None, max_features='auto',  
                        max_leaf_nodes=None, max_samples=None,  
                        min_impurity_decrease=0.0, min_impurity_split=None,  
                        min_samples_leaf=1, min_samples_split=2,  
                        min_weight_fraction_leaf=0.0, n_estimators=100,  
                        n_jobs=None, oob_score=False, random_state=None,  
                        verbose=0, warm_start=False)
```

Model Evaluation

Accuracy Score

```
# accuracy on test data  
X_test_prediction = model.predict(X_test)  
test_data_accuracy = accuracy_score(X_test_prediction, Y_test)
```

```
print('Accuracy : ', test_data_accuracy)
```

```
Accuracy : 0.921875
```

Building a Predictive System

```
input_data = (7.5,0.5,0.36,6.1,0.071,17.0,102.0,0.9978,3.35,0.8,10.5)
```

```
# changing the input data to a numpy array
```

```
input_data_as_numpy_array = np.asarray(input_data)
```

```
# reshape the data as we are predicting the label for only one instance
```

```
input_data_reshaped = input_data_as_numpy_array.reshape(1,-1)
```

```
prediction = model.predict(input_data_reshaped)
```

```
print(prediction)
```

```
if (prediction[0]==1):
```

```
    print('Good Quality Wine')
```

```
else:
```

```
    print('Bad Quality Wine')
```

```
[0]
```

```
Bad Quality Wine
```

<https://colab.research.google.com/drive/1RBvAg2uHQbNnCmENnqWLZXveAuB0EQBr#scrollTo=hnXyd-XbyTis>

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