

## \*\* PROJECT DAY - 24\*\*

### [Wine-Quality Prediction Using ML]

#### Importing the Dependencies

```
In [141]: 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

```
In [144]: # loading the dataset to a Pandas DataFrame
wine_dataset = pd.read_csv('winequality.csv')

In [146]: wine_dataset.head()

Out[146]: fixed acidity  volatile acidity  citric acid  residual sugar  chlorides  free sulfur dioxide  total sulfur dioxide  density  pH  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 [148]: # number of rows & columns in the dataset
wine_dataset.shape
```

```
Out[148]: (1599, 12)
```

```
In [150]: # checking for missing values
wine_dataset.isnull().sum()
```

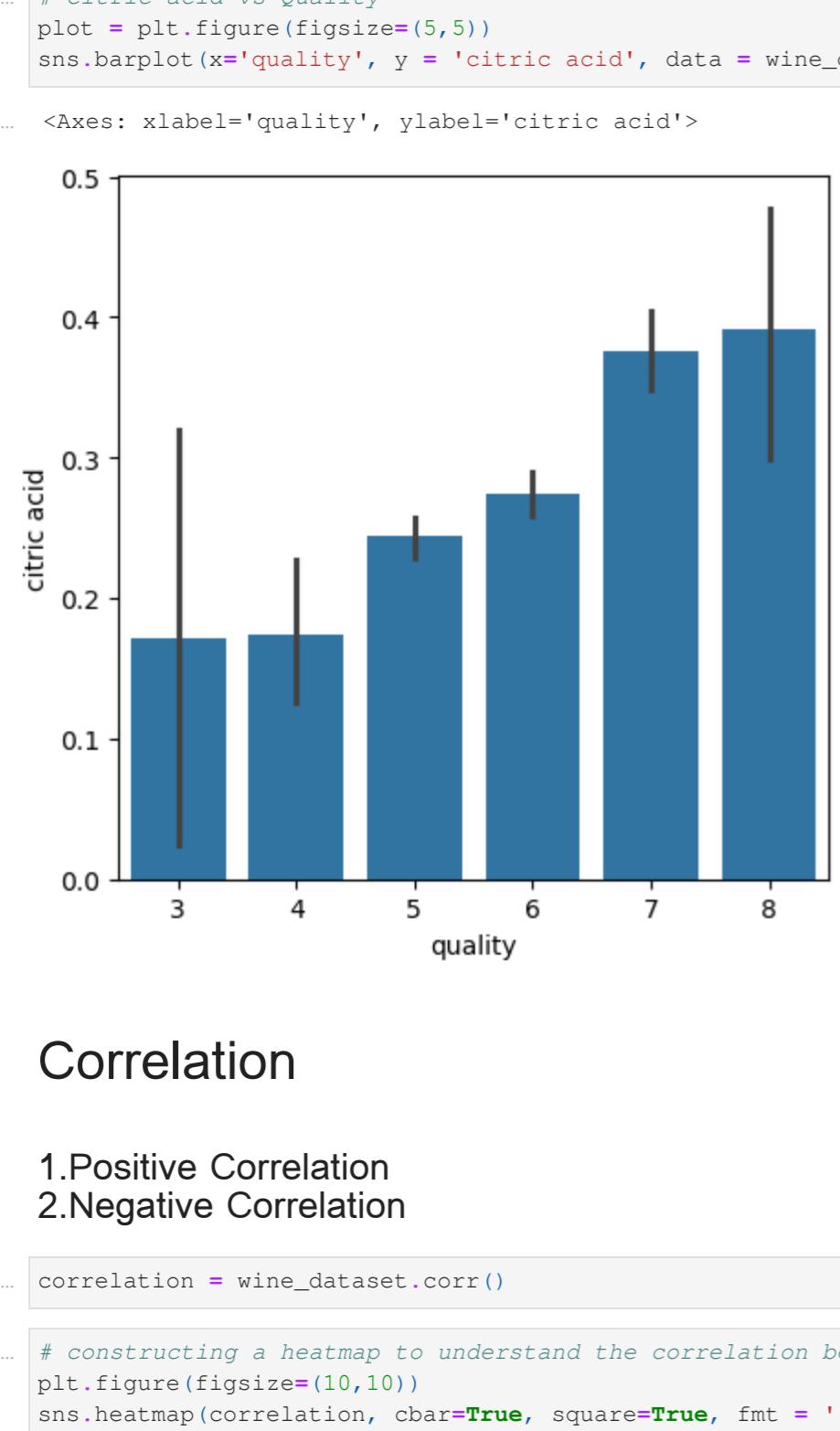
```
Out[150]: fixed acidity      0
volatile acidity      0
citric acid          0
residual sugar        0
chlorides             0
free sulfur dioxide  0
total sulfur dioxide 0
density               0
pH                    0
sulphates             0
alcohol               0
quality               0
dtype: int64
```

#### Data Analysis and Visualization

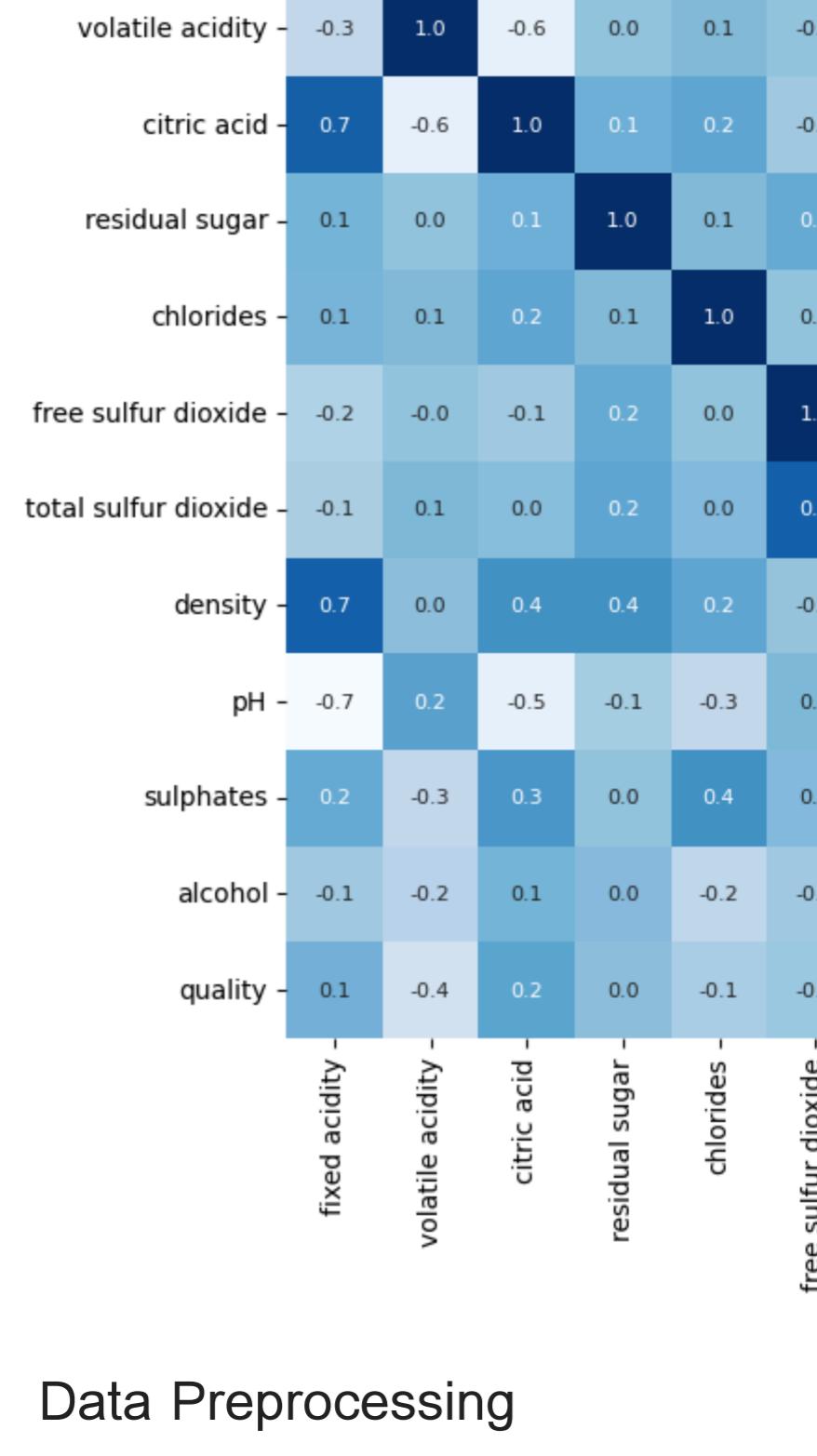
```
In [153]: # statistical measures of the dataset
wine_dataset.describe()
```

```
Out[153]: fixed acidity  volatile acidity  citric acid  residual sugar  chlorides  free sulfur dioxide  total sulfur dioxide  density  pH  sulphates  alcohol  quality
count 1599.000000  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  0.658149  10.422983  5.636023
std   1.741096  0.179060  0.194801  1.409928  0.047065  10.460157  32.895324  0.001887  0.154386  0.169507  1.065668  0.807569
min   4.600000  0.120000  0.000000  0.900000  0.012000  1.000000  6.000000  0.990070  2.740000  0.330000  8.400000  3.000000
25%   7.100000  0.390000  0.090000  1.900000  0.070000  7.000000  22.000000  0.995600  3.210000  0.550000  9.500000  5.000000
50%   7.900000  0.520000  0.260000  2.200000  0.079000  14.000000  38.000000  0.996750  3.310000  0.620000  10.200000  6.000000
75%   9.200000  0.640000  0.420000  2.600000  0.090000  21.000000  62.000000  0.997835  3.400000  0.730000  11.100000  6.000000
max   15.900000  1.580000  1.000000  15.500000  0.611000  72.000000  289.000000  1.003690  4.010000  2.000000  14.900000  8.000000
```

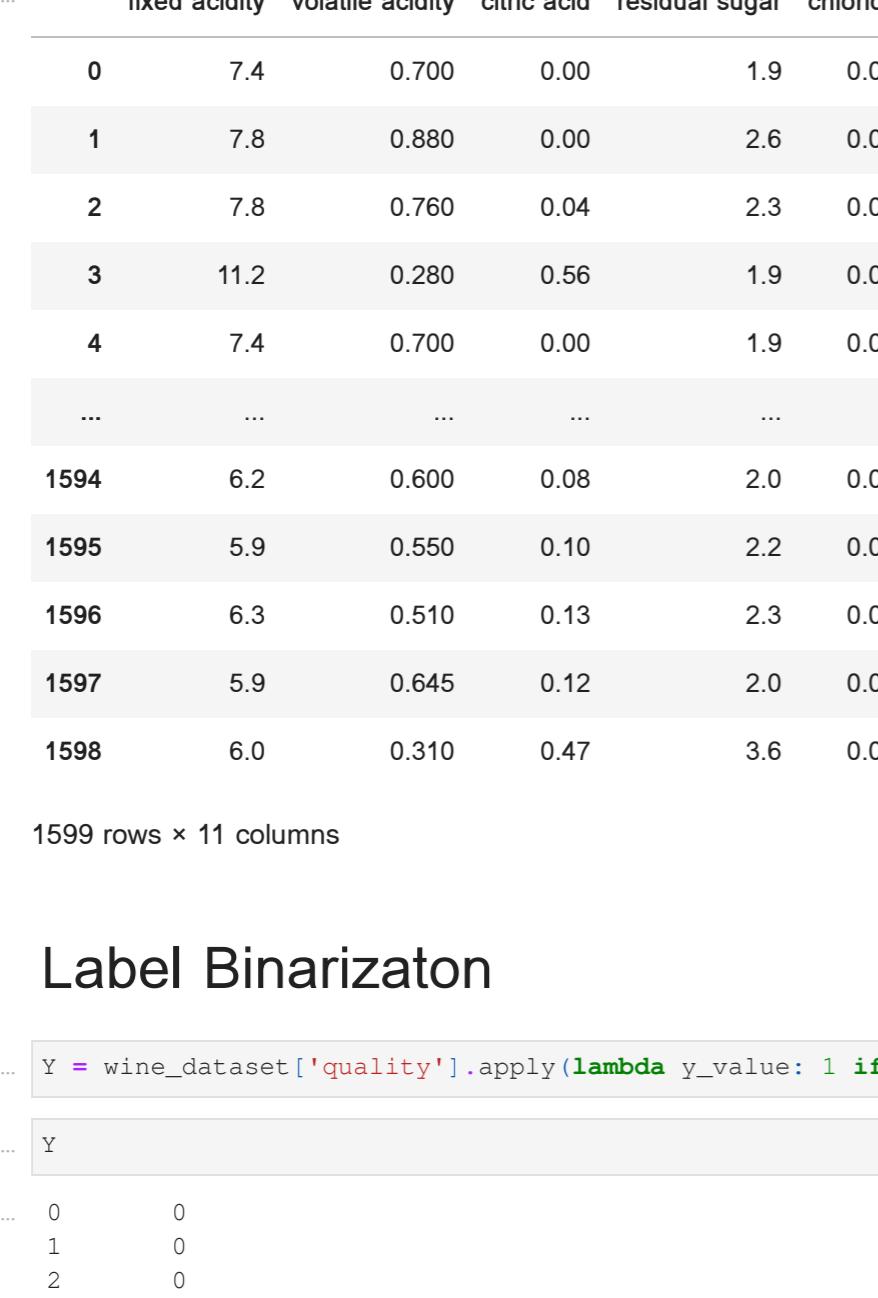
```
In [155]: # number of values for each quality
sns.countplot(x='quality', data = wine_dataset, kind = 'count')
plt.show()
```



```
In [157]: # volatile acidity vs Quality
plot = plt.figure(figsize=(5,5))
sns.barplot(x='quality', y = 'volatile acidity', data = wine_dataset)
plt.show()
```



```
In [159]: # citric acid vs Quality
plot = plt.figure(figsize=(5,5))
sns.barplot(x='quality', y = 'citric acid', data = wine_dataset)
plt.show()
```



#### Correlation

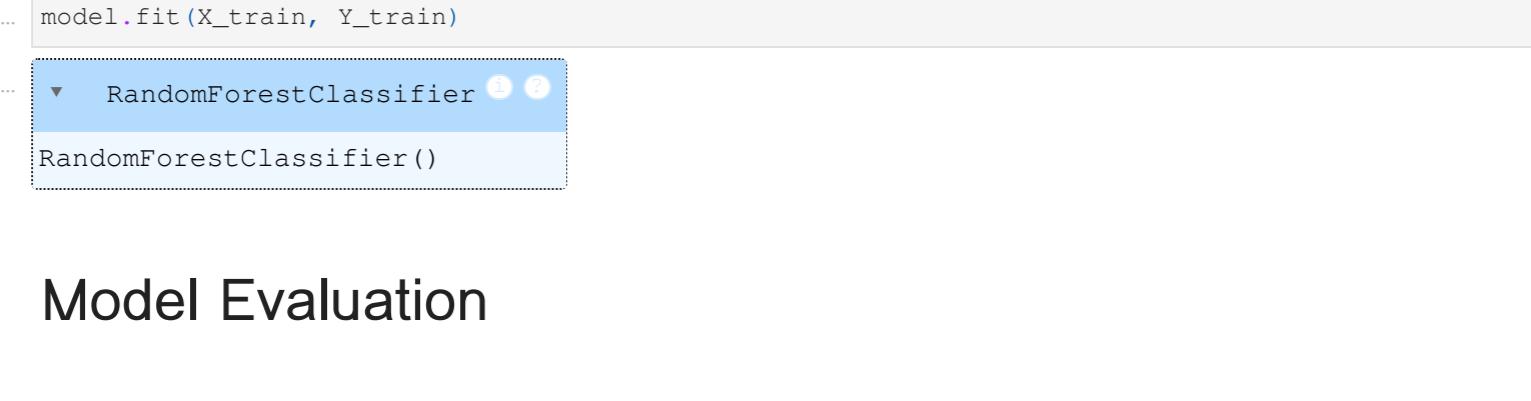
##### 1. Positive Correlation

##### 2. Negative Correlation

```
In [163]: correlation = wine_dataset.corr()
```

```
In [165]: # constructing a heatmap to understand the correlation between the columns
plt.figure(figsize=(10,10))
sns.heatmap(correlation, cmap='Blues', annot=True, annot_kws={'size':8})
```

```
plt.show()
```



#### Data Preprocessing

```
In [168]: # separate the data and Label
X = wine_dataset.drop('quality',axis=1)
```

```
In [170]: X
```

```
Out[170]: fixed acidity  volatile acidity  citric acid  residual sugar  chlorides  free sulfur dioxide  total sulfur dioxide  density  pH  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
..    ...
1594   6.2            0.60        0.08         2.0     0.090          32.0           44.0   0.99490  3.45     0.58    10.5
1595   5.9            0.550       0.10         2.2     0.062          39.0           51.0   0.99512  3.52     0.76    11.2
1596   6.3            0.510       0.13         2.3     0.076          29.0           40.0   0.99574  3.42     0.75    11.0
1597   5.9            0.645       0.12         2.0     0.075          32.0           44.0   0.99547  3.57     0.71    10.2
1598   6.0            0.310       0.47         3.6     0.067          18.0           42.0   0.99549  3.39     0.66    11.0
```

1599 rows x 11 columns

#### Label Binarization

```
In [173]: Y = wine_dataset['quality'].apply(lambda y_value: 1 if y_value>=7 else 0)
```

```
In [175]: Y
```

```
Out[175]: 0      0
1      1
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

```
In [178]: X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.2, random_state=3)
```

```
In [180]: X.shape, Y_train.shape, Y_test.shape
```

```
Out[180]: (1159,), (1279,), (320,)
```

#### Model Training

##### Random Forest Classifier

```
In [184]: model = RandomForestClassifier()
```

```
In [186]: model.fit(X_train, Y_train)
```

```
Out[186]: RandomForestClassifier()
```

Bad Quality Wine ✗

Good Quality Wine ☀

Bad Quality Wine ✗

