# Wine Quality Prediction using Support Vector Machine

# **Import Library**

In [72]:

import pandas as pd

In [73]:

import numpy as np

# Import CSV as DataFrame

In [74]:

df = pd.read\_csv('https://github.com/YBI-Foundation/Dataset/raw/main/WhiteWineQuality.csv',

## Get the First rows of Dataframe

In [75]:

df.head()

Out[75]:

	fixed acidity	volatile acidity	citric acid	residual sugar	chlorides	free sulfur dioxide	total sulfur dioxide	density	рН	sulphates	alcoh
0	7.0	0.27	0.36	20.7	0.045	45.0	170.0	1.0010	3.00	0.45	8
1	6.3	0.30	0.34	1.6	0.049	14.0	132.0	0.9940	3.30	0.49	9
2	8.1	0.28	0.40	6.9	0.050	30.0	97.0	0.9951	3.26	0.44	10
3	7.2	0.23	0.32	8.5	0.058	47.0	186.0	0.9956	3.19	0.40	9
4	7.2	0.23	0.32	8.5	0.058	47.0	186.0	0.9956	3.19	0.40	9
1											•

## **Get Information of DataFrame**

## In [76]:

```
df.info()
```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 4898 entries, 0 to 4897
Data columns (total 12 columns):

#	Column	Non-Null Count	Dtype
0	fixed acidity	4898 non-null	float64
1	volatile acidity	4898 non-null	float64
2	citric acid	4898 non-null	float64
3	residual sugar	4898 non-null	float64
4	chlorides	4898 non-null	float64
5	free sulfur dioxide	4898 non-null	float64
6	total sulfur dioxide	4898 non-null	float64
7	density	4898 non-null	float64
8	рН	4898 non-null	float64
9	sulphates	4898 non-null	float64
10	alcohol	4898 non-null	float64
11	quality	4898 non-null	int64

dtypes: float64(11), int64(1)

memory usage: 459.3 KB

# **Get Summary Statistics**

### In [77]:

df.describe()

Out[77]:

	fixed acidity	volatile acidity	citric acid	residual sugar	chlorides	free sulfur dioxide	total su dio:
count	4898.000000	4898.000000	4898.000000	4898.000000	4898.000000	4898.000000	4898.000
mean	6.854788	0.278241	0.334192	6.391415	0.045772	35.308085	138.360
std	0.843868	0.100795	0.121020	5.072058	0.021848	17.007137	42.498
min	3.800000	0.080000	0.000000	0.600000	0.009000	2.000000	9.000
25%	6.300000	0.210000	0.270000	1.700000	0.036000	23.000000	108.000
50%	6.800000	0.260000	0.320000	5.200000	0.043000	34.000000	134.000
75%	7.300000	0.320000	0.390000	9.900000	0.050000	46.000000	167.000
max	14.200000	1.100000	1.660000	65.800000	0.346000	289.000000	440.000
4							•

## **Get Column Names**

## **Get Shape of DataFrame**

```
In [79]:

df.shape

Out[79]:
  (4898, 12)
```

# Get Unique Values (Class or Label) in y Variable

```
In [80]:
df['quality'].value_counts()
Out[80]:
6
     2198
5
     1457
7
      880
8
      175
4
      163
3
       20
9
Name: quality, dtype: int64
```

```
In [81]:
```

```
df.groupby('quality').mean()
```

## Out[81]:

		fixed acidity	volatile acidity	citric acid	residual sugar	chlorides	free sulfur dioxide	total sulfur dioxide	density	
q	uality									
	3	7.600000	0.333250	0.336000	6.392500	0.054300	53.325000	170.600000	0.994884	3.1
	4	7.129448	0.381227	0.304233	4.628221	0.050098	23.358896	125.279141	0.994277	3.1
	5	6.933974	0.302011	0.337653	7.334969	0.051546	36.432052	150.904598	0.995263	3.1
	6	6.837671	0.260564	0.338025	6.441606	0.045217	35.650591	137.047316	0.993961	3.1
	7	6.734716	0.262767	0.325625	5.186477	0.038191	34.125568	125.114773	0.992452	3.2
	8	6.657143	0.277400	0.326514	5.671429	0.038314	36.720000	126.165714	0.992236	3.2
	9	7.420000	0.298000	0.386000	4.120000	0.027400	33.400000	116.000000	0.991460	3.3

# **Define y and X**

```
In [82]:
```

```
y = df['quality']
```

```
In [83]:
```

y.shape

Out[83]:

(4898,)

```
In [84]:
У
Out[84]:
        6
0
1
        6
2
        6
3
        6
        6
4893
        6
4894
        5
4895
        6
4896
        7
4897
Name: quality, Length: 4898, dtype: int64
In [85]:
x = df.drop(['quality'], axis=1)
In [86]:
x.shape
Out[86]:
(4898, 11)
```

```
In [87]:
```

Х

### Out[87]:

	fixed acidity	volatile acidity	citric acid	residual sugar	chlorides	free sulfur dioxide	total sulfur dioxide	density	рН	sulphates	al
0	7.0	0.27	0.36	20.7	0.045	45.0	170.0	1.00100	3.00	0.45	_
1	6.3	0.30	0.34	1.6	0.049	14.0	132.0	0.99400	3.30	0.49	
2	8.1	0.28	0.40	6.9	0.050	30.0	97.0	0.99510	3.26	0.44	
3	7.2	0.23	0.32	8.5	0.058	47.0	186.0	0.99560	3.19	0.40	
4	7.2	0.23	0.32	8.5	0.058	47.0	186.0	0.99560	3.19	0.40	
4893	6.2	0.21	0.29	1.6	0.039	24.0	92.0	0.99114	3.27	0.50	
4894	6.6	0.32	0.36	8.0	0.047	57.0	168.0	0.99490	3.15	0.46	
4895	6.5	0.24	0.19	1.2	0.041	30.0	111.0	0.99254	2.99	0.46	
4896	5.5	0.29	0.30	1.1	0.022	20.0	110.0	0.98869	3.34	0.38	
4897	6.0	0.21	0.38	8.0	0.020	22.0	98.0	0.98941	3.26	0.32	
4898 rows × 11 columns											

## **Get X Variables Standardized**

## In [88]:

from sklearn.preprocessing import StandardScaler

## In [89]:

ss = StandardScaler()

### In [90]:

x = ss.fit\_transform(x)

## **Get Train Test Split**

```
In [92]:
from sklearn.model_selection import train_test_split

In [93]:

x_train, x_test, y_train, y_test = train_test_split(x,y, test_size = 0.3, stratify=y, rando

In [94]:

x_train.shape, x_test.shape, y_train.shape, y_test.shape

Out[94]:
((3428, 11), (1470, 11), (3428,), (1470,))
```

## **Get Model Train**

```
In [95]:
from sklearn.svm import SVC
In [96]:
svc = SVC()
```

```
In [97]:
svc.fit(x_train, y_train)
Out[97]:
SVC()
```

## **Model Prediction**

```
In [98]:

y_pred = svc.predict(x_test)

In [99]:

y_pred.shape

Out[99]:
(1470,)

In [100]:

y_pred

Out[100]:

array([5, 7, 5, ..., 5, 5], dtype=int64)
```

## **Model Evaluation**

```
In [101]:
```

```
from sklearn.metrics import confusion_matrix, classification_report
```

```
In [102]:
```

```
print(confusion_matrix(y_test, y_pred))
```

```
[[
   0
        0
            1
                5
                    0
                         0
                             0]
   0
        2 25
              22
                    0
                         0
                             0]
[
   0
        3 273 160
                             0]
                   1
                        0
   0
        0 122 515
                   23
                             0]
            6 191
   0
        0
                   67
                         0
                             0]
            0
               39
                   14
                             0]
   0
            0
                             0]]
                0
                    1
```

### In [104]:

```
print(classification_report(y_test,y_pred))
```

	precision	recall	f1-score	support
_				_
3	0.00	0.00	0.00	6
4	0.40	0.04	0.07	49
5	0.64	0.62	0.63	437
6	0.55	0.78	0.65	660
7	0.63	0.25	0.36	264
8	0.00	0.00	0.00	53
9	0.00	0.00	0.00	1
accuracy			0.58	1470
macro avg	0.32	0.24	0.25	1470
weighted avg	0.57	0.58	0.55	1470

C:\Users\91814\New folder\lib\site-packages\sklearn\metrics\\_classification. py:1318: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero\_division` parameter to control this behavior.

\_warn\_prf(average, modifier, msg\_start, len(result))

C:\Users\91814\New folder\lib\site-packages\sklearn\metrics\\_classification. py:1318: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero\_division` par ameter to control this behavior.

\_warn\_prf(average, modifier, msg\_start, len(result))

C:\Users\91814\New folder\lib\site-packages\sklearn\metrics\\_classification. py:1318: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero\_division` parameter to control this behavior.

\_warn\_prf(average, modifier, msg\_start, len(result))

# Model Re-run with two class created for wine quality

```
In [105]:
```

```
y = df['quality'].apply(lambda y_value: 1 if y_value>=6 else 0)
```

#### In [106]:

```
y.value_counts()
```

### Out[106]:

1 3258 0 1640

Name: quality, dtype: int64

## Train test split

```
In [107]:
from sklearn.model_selection import train_test_split

In [108]:
x_train, x_test, y_train, y_test = train_test_split(x,y ,test_size = 0.3, stratify= y, rand

In [109]:
x_train.shape, x_test.shape, y_train.shape, y_test.shape

Out[109]:
((3428, 11), (1470, 11), (3428,), (1470,))
```

## **Model Train**

```
In [110]:
from sklearn.svm import SVC

In [111]:
svc = SVC()

In [112]:
svc.fit(x_train, y_train)

Out[112]:
SVC()
```

## **Model Prediction**

```
In [113]:
y_pred = svc.predict(x_test)

In [114]:
y_pred.shape

Out[114]:
(1470,)

In [115]:
y_pred

Out[115]:
array([0, 1, 1, ..., 1, 1, 1], dtype=int64)
```

## **Model Evaluation**

```
In [116]:
```

from sklearn.metrics import confusion\_matrix, classification\_report

### In [117]:

```
print(confusion_matrix(y_test, y_pred))
```

[[289 203] [124 854]]

### In [118]:

print(classification\_report(y\_test,y\_pred))

	precision	recall	f1-score	support
0	0.70	0.59	0.64	492
1	0.81	0.87	0.84	978
accuracy			0.78	1470
macro avg	0.75	0.73	0.74	1470
weighted avg	0.77	0.78	0.77	1470

## **Future Predictions**

### In [119]:

```
df_new = df.sample(1)
```

### In [120]:

```
df_new
```

### Out[120]:

		fixed acidity	volatile acidity	citric acid	residual sugar	chlorides		total sulfur dioxide	density	рН	sulphates	al
•	4611	5.9	0.42	0.36	2.4	0.034	19.0	77.0	0.99184	3.25	0.48	
	<b>√</b>											<b>•</b>

### In [121]:

df\_new.shape

## Out[121]:

(1, 12)

```
In [122]:
```

```
x_new = df_new.drop(['quality'], axis = 1)
```

## In [123]:

```
x_new = ss.fit_transform(x_new)
```

### In [124]:

```
y_pred_new = svc.predict(x_new)
```

## In [125]:

```
y_pred_new
```

### Out[125]:

array([1], dtype=int64)