DAY 10:

winequality Dataset

In [1]:

```
#to import libraries
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

In [2]:

df=pd.read_csv(r"C:\Users\user\Downloads\11_winequality-red.csv")[0:500]
df

Out[2]:

| _ | | fixed acidity | volatile acidity | citric acid | residual sugar | chlorides | free sulfur dioxide | total sulfur dioxide | density | рН | sulphates | al |
|---|-----|------------------|---------------------|----------------|-------------------|-----------|---------------------------|----------------------------|---------|------|-----------|----|
| | 0 | 7.4 | 0.70 | 0.00 | 1.9 | 0.076 | 11.0 | 34.0 | 0.9978 | 3.51 | 0.56 | |
| | 1 | 7.8 | 0.88 | 0.00 | 2.6 | 0.098 | 25.0 | 67.0 | 0.9968 | 3.20 | 0.68 | |
| | 2 | 7.8 | 0.76 | 0.04 | 2.3 | 0.092 | 15.0 | 54.0 | 0.9970 | 3.26 | 0.65 | |
| | 3 | 11.2 | 0.28 | 0.56 | 1.9 | 0.075 | 17.0 | 60.0 | 0.9980 | 3.16 | 0.58 | |
| | 4 | 7.4 | 0.70 | 0.00 | 1.9 | 0.076 | 11.0 | 34.0 | 0.9978 | 3.51 | 0.56 | |
| | | | | | | | | | | | | |
| | 495 | 10.7 | 0.35 | 0.53 | 2.6 | 0.070 | 5.0 | 16.0 | 0.9972 | 3.15 | 0.65 | |
| | 496 | 7.8 | 0.52 | 0.25 | 1.9 | 0.081 | 14.0 | 38.0 | 0.9984 | 3.43 | 0.65 | |
| | 497 | 7.2 | 0.34 | 0.32 | 2.5 | 0.090 | 43.0 | 113.0 | 0.9966 | 3.32 | 0.79 | |
| | 498 | 10.7 | 0.35 | 0.53 | 2.6 | 0.070 | 5.0 | 16.0 | 0.9972 | 3.15 | 0.65 | |
| | 499 | 8.7 | 0.69 | 0.31 | 3.0 | 0.086 | 23.0 | 81.0 | 1.0002 | 3.48 | 0.74 | |
| | | | | | | | | | | | | |

500 rows × 12 columns

```
In [3]:
```

```
df.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 500 entries, 0 to 499
Data columns (total 12 columns):
     Column
                           Non-Null Count Dtype
     _____
     fixed acidity
                                           float64
 0
                           500 non-null
 1
     volatile acidity
                           500 non-null
                                           float64
 2
     citric acid
                           500 non-null
                                           float64
 3
    residual sugar
                           500 non-null
                                           float64
    chlorides
                                           float64
 4
                           500 non-null
 5
     free sulfur dioxide
                           500 non-null
                                           float64
    total sulfur dioxide 500 non-null
                                           float64
    density
                           500 non-null
                                           float64
 7
 8
     рΗ
                           500 non-null
                                           float64
                                           float64
 9
     sulphates
                           500 non-null
10 alcohol
                           500 non-null
                                           float64
                           500 non-null
                                           int64
 11 quality
dtypes: float64(11), int64(1)
memory usage: 47.0 KB
In [4]:
df.columns
Out[4]:
Index(['fixed acidity', 'volatile acidity', 'citric acid', 'residual suga
       'chlorides', 'free sulfur dioxide', 'total sulfur dioxide', 'densit
у',
       'pH', 'sulphates', 'alcohol', 'quality'],
      dtype='object')
Linear Regression
In [6]:
x=df[['fixed acidity', 'volatile acidity', 'citric acid', 'residual sugar',
       'chlorides', 'free sulfur dioxide']]
y=df[ 'quality']
```

```
In [7]:
```

```
# to split my dataset into test and train data
from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.3)
```

```
In [8]:
```

```
from sklearn.linear_model import LinearRegression
lr=LinearRegression()
lr.fit(x_train,y_train)
```

Out[8]:

LinearRegression()

In [9]:

```
print(lr.score(x_test,y_test))
```

0.16091846851024572

In [10]:

```
lr.score(x_train,y_train)
```

Out[10]:

0.1504275401993429

Ridge Regression

In [11]:

```
from sklearn.linear_model import Ridge,Lasso
```

In [12]:

```
rr=Ridge(alpha=10)
rr.fit(x_train,y_train)
rr.score(x_test,y_test)
```

Out[12]:

0.14179907392676827

Lasso Regression

In [13]:

```
la=Lasso(alpha=10)
la.fit(x_train,y_train)
```

Out[13]:

Lasso(alpha=10)

```
In [14]:
la.score(x_test,y_test)
Out[14]:
-0.008272415229287011
```

Elastic regression

```
In [15]:
from sklearn.linear_model import ElasticNet
en=ElasticNet()
en.fit(x_train,y_train)

Out[15]:
ElasticNet()

In [16]:
print(en.intercept_)

5.660540623598547

In [17]:
predict=(en.predict(x_test))

In [18]:
print(en.score(x_test,y_test))
```

0.013717829632268197

Evalution matrics

```
In [19]:
from sklearn import metrics
print("Mean Absolute Error:",metrics.mean_absolute_error(y_test,predict))

Mean Absolute Error: 0.6536684801802912

In [20]:
print("Mean Square Error:",metrics.mean_squared_error(y_test,predict))

Mean Square Error: 0.5606027856370186

In [21]:
print("Root Mean Square Error:",np.sqrt(metrics.mean_squared_error(y_test,predict)))

Root Mean Square Error: 0.7487341221268192
```