

kaviyadevi 20106064

```
In [4]: #to import Libraries
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

```
In [5]: #to import dataset
data=pd.read_csv(r"C:\Users\user\Downloads\11_winequality-red - 11_winequality-re
data
```

Out[5]:

fixed dity	volatile acidity	citric acid	residual sugar	chlorides	free sulfur dioxide	total sulfur dioxide	density	pH	sulphates	alcohol	quality
7.4	0.700	0.00	1.9	0.076	11.0	34.0	0.99780	3.51	0.56	9.4	5
7.8	0.880	0.00	2.6	0.098	25.0	67.0	0.99680	3.20	0.68	9.8	5
7.8	0.760	0.04	2.3	0.092	15.0	54.0	0.99700	3.26	0.65	9.8	5
11.2	0.280	0.56	1.9	0.075	17.0	60.0	0.99800	3.16	0.58	9.8	6
7.4	0.700	0.00	1.9	0.076	11.0	34.0	0.99780	3.51	0.56	9.4	5
...
6.2	0.600	0.08	2.0	0.090	32.0	44.0	0.99490	3.45	0.58	10.5	5
5.9	0.550	0.10	2.2	0.062	39.0	51.0	0.99512	3.52	0.76	11.2	6
6.3	0.510	0.13	2.3	0.076	29.0	40.0	0.99574	3.42	0.75	11.0	6
5.9	0.645	0.12	2.0	0.075	32.0	44.0	0.99547	3.57	0.71	10.2	5
6.0	0.310	0.47	3.6	0.067	18.0	42.0	0.99549	3.39	0.66	11.0	6

; × 12 columns



DATA PREPROCESSING AND CLEANING

In [6]: data.head()

Out[6]:

	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

In [4]: data.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 158 entries, 0 to 157
Data columns (total 12 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   Country                               158 non-null    object
1   Region                                158 non-null    object
2   Happiness Rank                        158 non-null    int64
3   Happiness Score                       158 non-null    float64
4   Standard Error                       158 non-null    float64
5   Economy (GDP per Capita)             158 non-null    float64
6   Family                                158 non-null    float64
7   Health (Life Expectancy)             158 non-null    float64
8   Freedom                               158 non-null    float64
9   Trust (Government Corruption)         158 non-null    float64
10  Generosity                            158 non-null    float64
11  Dystopia Residual                     158 non-null    float64
dtypes: float64(9), int64(1), object(2)
memory usage: 14.9+ KB
```

```
In [7]: data.describe()
```

Out[7]:

	fixed acidity	volatile acidity	citric acid	residual sugar	chlorides	free sulfur dioxide	total sulfur dioxide
count	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
std	1.741096	0.179060	0.194801	1.409928	0.047065	10.460157	32.895324
min	4.600000	0.120000	0.000000	0.900000	0.012000	1.000000	6.000000
25%	7.100000	0.390000	0.090000	1.900000	0.070000	7.000000	22.000000
50%	7.900000	0.520000	0.260000	2.200000	0.079000	14.000000	38.000000
75%	9.200000	0.640000	0.420000	2.600000	0.090000	21.000000	62.000000
max	15.900000	1.580000	1.000000	15.500000	0.611000	72.000000	289.000000

```
In [8]: data.columns
```

Out[8]:

Index(['fixed acidity', 'volatile acidity', 'citric acid', 'residual sugar', 'chlorides', 'free sulfur dioxide', 'total sulfur dioxide', 'density', 'pH', 'sulphates', 'alcohol', 'quality'], dtype='object')

```
In [9]: data.isnull()
```

Out[9]:

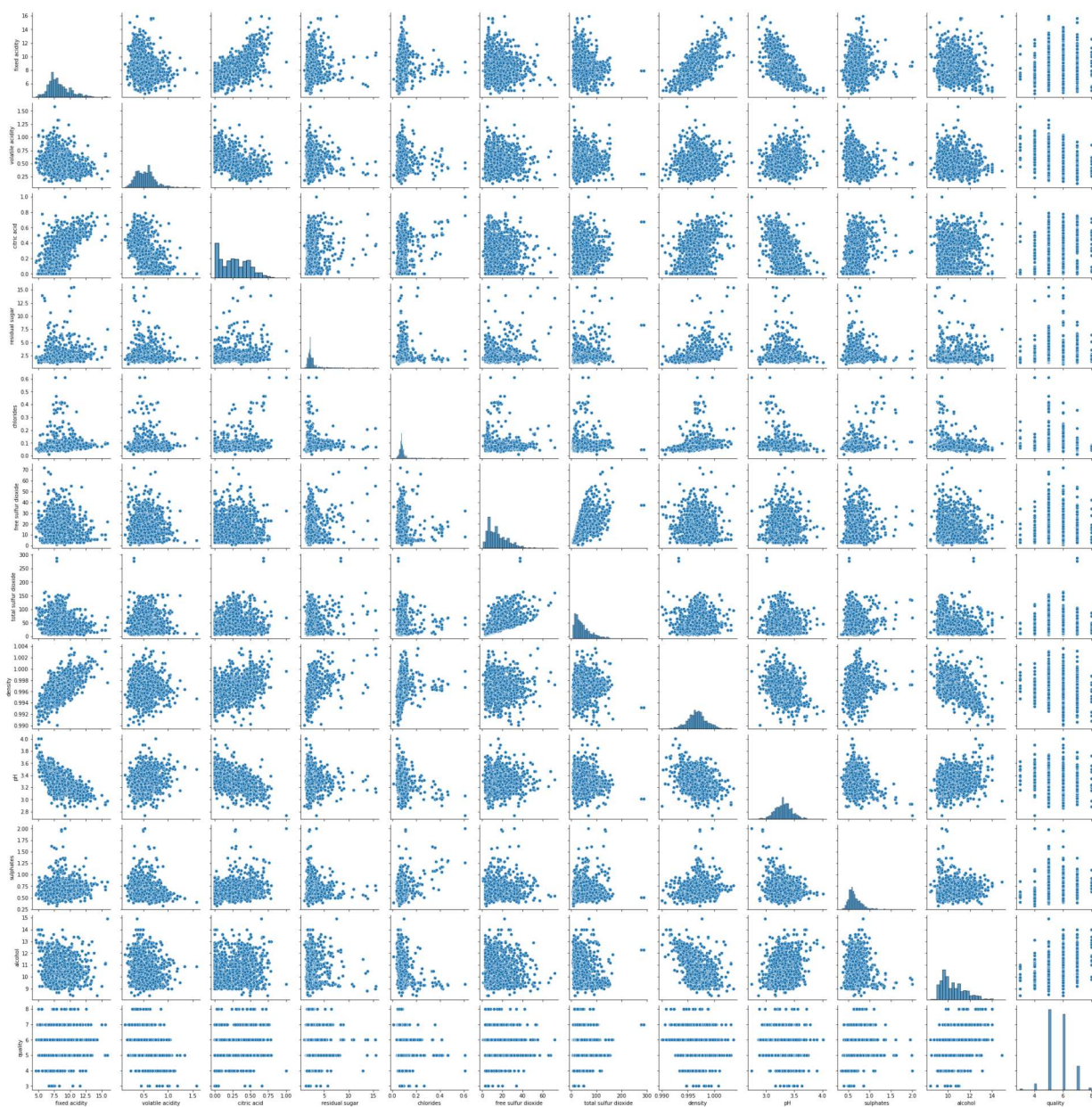
	fixed acidity	volatile acidity	citric acid	residual sugar	chlorides	free sulfur dioxide	total sulfur dioxide	density	pH	sulphates	alcohol	quality
0	False	False	False	False	False	False	False	False	False	False	False	False
1	False	False	False	False	False	False	False	False	False	False	False	False
2	False	False	False	False	False	False	False	False	False	False	False	False
3	False	False	False	False	False	False	False	False	False	False	False	False
4	False	False	False	False	False	False	False	False	False	False	False	False
...
10	False	False	False	False	False	False	False	False	False	False	False	False
11	False	False	False	False	False	False	False	False	False	False	False	False
12	False	False	False	False	False	False	False	False	False	False	False	False
13	False	False	False	False	False	False	False	False	False	False	False	False
14	False	False	False	False	False	False	False	False	False	False	False	False
...

× 12 columns

EDA and DATA VISUALIZATION

```
In [11]: sns.pairplot(data)
```

```
Out[11]: <seaborn.axisgrid.PairGrid at 0x139e0d66130>
```

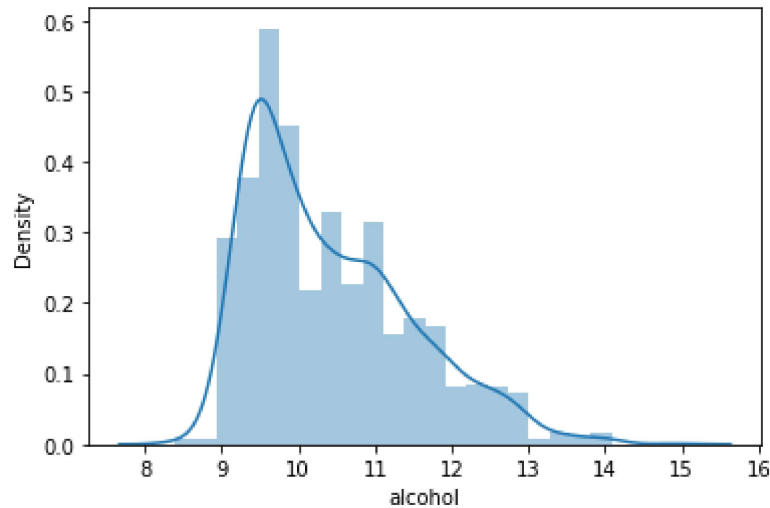


```
In [16]: sns.distplot(data['alcohol'])
```

C:\ProgramData\Anaconda3\lib\site-packages\seaborn\distributions.py:2557: FutureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

```
warnings.warn(msg, FutureWarning)
```

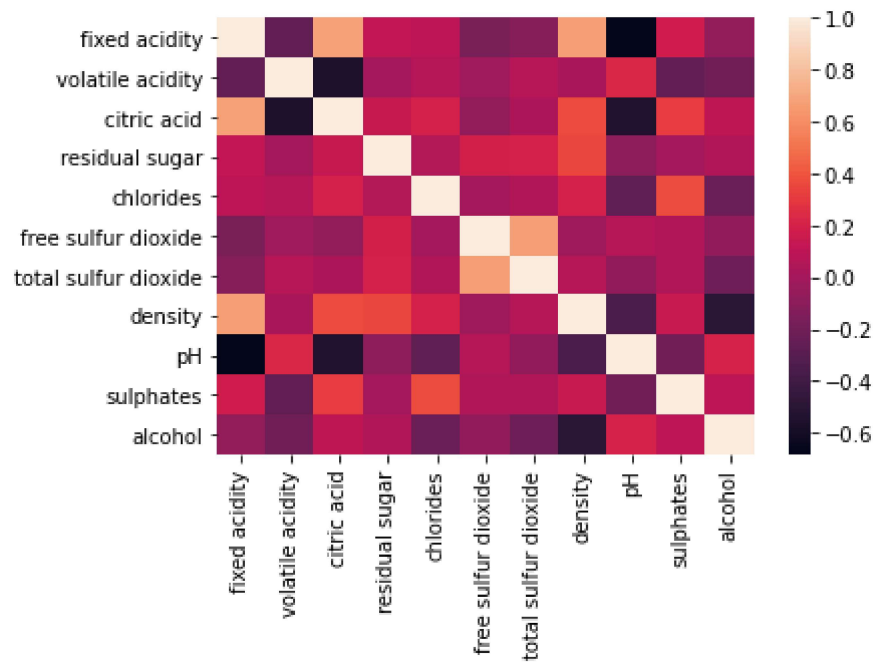
```
Out[16]: <AxesSubplot:xlabel='alcohol', ylabel='Density'>
```



```
In [22]: df=data[['fixed acidity', 'volatile acidity', 'citric acid', 'residual sugar',  
                 'chlorides', 'free sulfur dioxide', 'total sulfur dioxide', 'density',  
                 'pH', 'sulphates', 'alcohol']]
```

```
In [23]: sns.heatmap(df.corr())
```

```
Out[23]: <AxesSubplot:>
```



MODEL TRAINING

```
In [24]: x=df[['fixed acidity', 'volatile acidity', 'citric acid', 'residual sugar',  
              'chlorides', 'free sulfur dioxide', 'total sulfur dioxide', 'density',  
              'pH', 'sulphates']]  
y=df[['alcohol']]
```

```
In [25]: #to split my dataset into training and test  
  
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 [26]: from sklearn.linear_model import LinearRegression

lr=LinearRegression()
lr.fit(x_train,y_train)
```

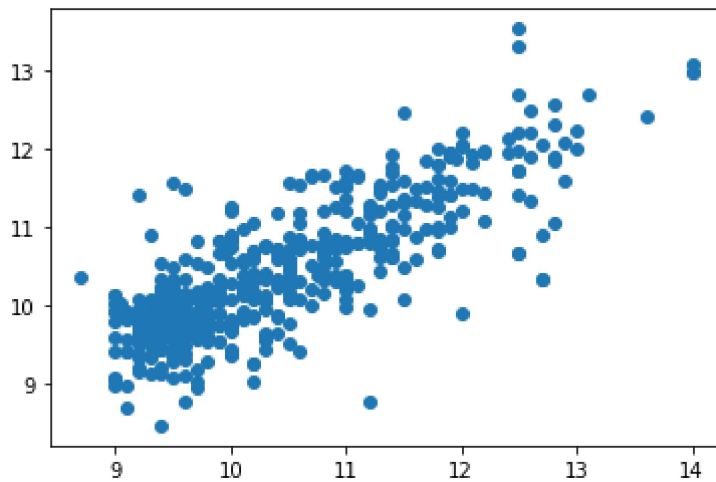
Out[26]: LinearRegression()

```
In [30]: #to find intercept
print(lr.intercept_)
```

[607.11340693]

```
In [32]: prediction = lr.predict(x_test)
plt.scatter(y_test,prediction)
```

Out[32]: <matplotlib.collections.PathCollection at 0x139f213bbb0>



```
In [33]: print(lr.score(x_test,y_test))
```

0.6631183286873545