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In [1]: #to import libraries

import numpy as np

import pandas as pd

import matplotlib.pyplot as plt

import seaborn as sns

In [2]: #to import dataset

data1=pd.read_csv(r"C:\Users\user\Downloads\11_winequality-red - 11_winequality-r
data1

Out[2]:

	fixed acidity	volatile acidity	citric acid	residual sugar	chlorides	free sulfur dioxide	total sulfur dioxide	density	рН	sulphates	alcoh
0	7.4	0.700	0.00	1.9	0.076	11.0	34.0	0.99780	3.51	0.56	9
1	7.8	0.880	0.00	2.6	0.098	25.0	67.0	0.99680	3.20	0.68	9
2	7.8	0.760	0.04	2.3	0.092	15.0	54.0	0.99700	3.26	0.65	9
3	11.2	0.280	0.56	1.9	0.075	17.0	60.0	0.99800	3.16	0.58	9
4	7.4	0.700	0.00	1.9	0.076	11.0	34.0	0.99780	3.51	0.56	9
1594	6.2	0.600	0.08	2.0	0.090	32.0	44.0	0.99490	3.45	0.58	10
1595	5.9	0.550	0.10	2.2	0.062	39.0	51.0	0.99512	3.52	0.76	11
1596	6.3	0.510	0.13	2.3	0.076	29.0	40.0	0.99574	3.42	0.75	11
1597	5.9	0.645	0.12	2.0	0.075	32.0	44.0	0.99547	3.57	0.71	10
1598	6.0	0.310	0.47	3.6	0.067	18.0	42.0	0.99549	3.39	0.66	11

1599 rows × 12 columns

data

Out[3]:

	fixed acidity	volatile acidity	citric acid	residual sugar	chlorides	free sulfur dioxide	total sulfur dioxide	density	рН	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
4											

In [4]: data.info()

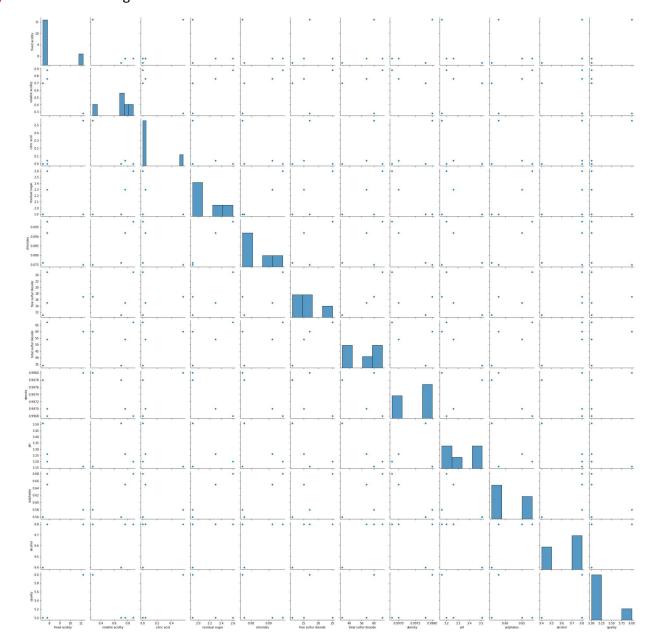
DATA CLEANING AND PREPROCESSING

```
<class 'pandas.core.frame.DataFrame'>
         RangeIndex: 5 entries, 0 to 4
         Data columns (total 12 columns):
               Column
                                        Non-Null Count
                                                          Dtype
               _____
                                        _____
                                                          ____
          0
               fixed acidity
                                        5 non-null
                                                          float64
          1
               volatile acidity
                                        5 non-null
                                                          float64
               citric acid
          2
                                        5 non-null
                                                          float64
          3
               residual sugar
                                                          float64
                                        5 non-null
          4
               chlorides
                                        5 non-null
                                                          float64
          5
               free sulfur dioxide
                                       5 non-null
                                                          float64
          6
               total sulfur dioxide 5 non-null
                                                          float64
          7
               density
                                        5 non-null
                                                          float64
          8
               рΗ
                                        5 non-null
                                                          float64
          9
               sulphates
                                        5 non-null
                                                          float64
          10
              alcohol
                                        5 non-null
                                                          float64
          11
               quality
                                        5 non-null
                                                          int64
         dtypes: float64(11), int64(1)
         memory usage: 608.0 bytes
In [5]:
         #to display summary of statistics
         data.describe()
Out[5]:
                                                                     free
                                                                               total
                     fixed
                            volatile
                                       citric
                                              residual
                                                      chlorides
                                                                    sulfur
                                                                              sulfur
                                                                                    density
                                                                                                  pН
                   acidity
                             acidity
                                        acid
                                                sugar
                                                                  dioxide
                                                                            dioxide
                 5.000000
                          5.000000
                                    5.000000
                                             5.000000
                                                       5.000000
                                                                 5.000000
                                                                           5.000000
                                                                                    5.00000 5.000000
          count
                 8.320000
                           0.664000
                                    0.120000
                                             2.120000
                                                       0.083400
                                                                15.800000
                                                                          49.800000
                                                                                    0.99748
                                                                                            3.328000
          mean
                                                       0.010807
            std
                 1.622344
                           0.226892
                                    0.246577
                                             0.319374
                                                                 5.761944
                                                                          15.139353
                                                                                    0.00054
                                                                                            0.169912
            min
                 7.400000
                          0.280000
                                    0.000000
                                             1.900000
                                                       0.075000
                                                                11.000000
                                                                          34.000000
                                                                                    0.99680
                                                                                            3.160000
           25%
                 7.400000
                           0.700000
                                    0.000000
                                             1.900000
                                                       0.076000
                                                                11.000000
                                                                          34.000000
                                                                                    0.99700
                                                                                             3.200000
           50%
                 7.800000
                           0.700000
                                    0.000000
                                             1.900000
                                                       0.076000
                                                                15.000000
                                                                          54.000000
                                                                                    0.99780
                                                                                            3.260000
           75%
                 7.800000
                           0.760000
                                    0.040000
                                             2.300000
                                                       0.092000
                                                                17.000000
                                                                          60.000000
                                                                                    0.99780
                                                                                             3.510000
           max 11,200000
                          0.880000
                                    0.560000
                                             2.600000
                                                       0.098000
                                                                25.000000
                                                                          67.000000 0.99800
                                                                                            3.510000
In [6]:
         #to display the column heading
         data.columns
Out[6]: Index(['fixed acidity', 'volatile acidity', 'citric acid', 'residual sugar',
                 'chlorides', 'free sulfur dioxide', 'total sulfur dioxide', 'density',
                 'pH', 'sulphates', 'alcohol', 'quality'],
                dtype='object')
```

EDA and DATA VISUALIZATION

In [7]: sns.pairplot(data)

Out[7]: <seaborn.axisgrid.PairGrid at 0x1e8eebb1b50>

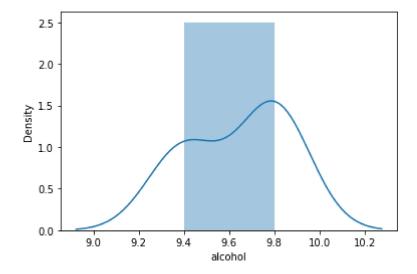


In [8]: | sns.distplot(data['alcohol'])

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

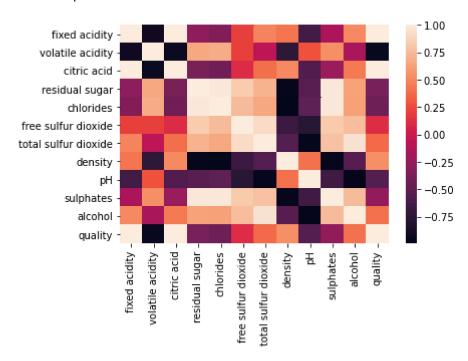
warnings.warn(msg, FutureWarning)

Out[8]: <AxesSubplot:xlabel='alcohol', ylabel='Density'>



```
In [10]: sns.heatmap(df.corr())
```

Out[10]: <AxesSubplot:>



TRAINING MODEL

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

from sklearn.model_selection import train_test_split

In [13]: from sklearn.linear_model import LinearRegression

```
lr=LinearRegression()
          lr.fit(x_train,y_train)
Out[13]: LinearRegression()
In [14]: #to find intercept
          print(lr.intercept_)
          [5.]
In [15]:
          prediction = lr.predict(x_test)
          plt.scatter(y_test,prediction)
Out[15]: <matplotlib.collections.PathCollection at 0x1e8f6ce71f0>
           5.2
           5.1
           5.0
           4.9
           4.8
               5.0
                        5.2
                                5.4
                                         5.6
                                                 5.8
                                                          6.0
```

In [16]: print(lr.score(x_test,y_test)) -1.0

RIDGE AND LASSO REGRESSION

```
In [20]: la=Lasso(alpha=10)
         la.fit(x_train,y_train)
         C:\ProgramData\Anaconda3\lib\site-packages\sklearn\linear_model\_coordinate_des
         cent.py:530: ConvergenceWarning: Objective did not converge. You might want to
         increase the number of iterations. Duality gap: 0.0, tolerance: 0.0
           model = cd_fast.enet_coordinate_descent(
Out[20]: Lasso(alpha=10)
In [21]: la.score(x_test,y_test)
Out[21]: -1.0
In [22]: | from sklearn.linear_model import ElasticNet
         en=ElasticNet()
         en.fit(x_train,y_train)
         C:\ProgramData\Anaconda3\lib\site-packages\sklearn\linear_model\_coordinate_des
         cent.py:530: ConvergenceWarning: Objective did not converge. You might want to
         increase the number of iterations. Duality gap: 0.0, tolerance: 0.0
           model = cd fast.enet coordinate descent(
Out[22]: ElasticNet()
In [23]: print(en.coef )
         [0. 0. 0. 0. 0. 0. 0. 0. 0. 0.]
In [24]: |print(en.predict(x_test))
         [5.5.]
In [25]: print(en.score(x_test,y_test))
         -1.0
In [26]: from sklearn import metrics
In [27]: print("Mean Absolute error", metrics.mean_absolute_error(y_test, prediction))
         Mean Absolute error 0.5
In [28]: print("Mean Squared error", metrics.mean_squared_error(y_test, prediction))
         Mean Squared error 0.5
In [29]: print("Root Mean Absolute error", np.sqrt(metrics.mean_squared_error(y_test, predic
         Root Mean Absolute error 0.7071067811865476
```