DATA COLLECTION

In [1]: |# import libraries import numpy as np import pandas as pd

import matplotlib.pyplot as plt

import seaborn as sns

In [2]: # To Import Dataset sd=pd.read_csv(r"c:\Users\user\Downloads\11_winequality-red.csv")

Out[2]:

	fixed acidity	volatile acidity	citric acid	residual sugar	chlorides	free sulfur dioxide	total sulfur dioxide	density	рН	sulphates	alcol
0	7.4	0.700	0.00	1.9	0.076	11.0	34.0	0.99780	3.51	0.56	
1	7.8	0.880	0.00	2.6	0.098	25.0	67.0	0.99680	3.20	0.68	
2	7.8	0.760	0.04	2.3	0.092	15.0	54.0	0.99700	3.26	0.65	
3	11.2	0.280	0.56	1.9	0.075	17.0	60.0	0.99800	3.16	0.58	
4	7.4	0.700	0.00	1.9	0.076	11.0	34.0	0.99780	3.51	0.56	
1594	6.2	0.600	80.0	2.0	0.090	32.0	44.0	0.99490	3.45	0.58	1
1595	5.9	0.550	0.10	2.2	0.062	39.0	51.0	0.99512	3.52	0.76	1
1596	6.3	0.510	0.13	2.3	0.076	29.0	40.0	0.99574	3.42	0.75	1
1597	5.9	0.645	0.12	2.0	0.075	32.0	44.0	0.99547	3.57	0.71	1
1598	6.0	0.310	0.47	3.6	0.067	18.0	42.0	0.99549	3.39	0.66	1

1599 rows × 12 columns

In [3]: # to display top 10 rows
sd.head(10)

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
5	7.4	0.66	0.00	1.8	0.075	13.0	40.0	0.9978	3.51	0.56	9.4
6	7.9	0.60	0.06	1.6	0.069	15.0	59.0	0.9964	3.30	0.46	9.4
7	7.3	0.65	0.00	1.2	0.065	15.0	21.0	0.9946	3.39	0.47	10.0
8	7.8	0.58	0.02	2.0	0.073	9.0	18.0	0.9968	3.36	0.57	9.5
9	7.5	0.50	0.36	6.1	0.071	17.0	102.0	0.9978	3.35	0.80	10.5
4.0	_	_	_	_		_	_	_	_		•

DATA CLEANING AND PRE_PROCESSING

In [4]: sd.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1599 entries, 0 to 1598
Data columns (total 12 columns):

_		(0000 = 000	······································	
ŧ	#	Column	Non-Null Count	Dtype
-				
(9	fixed acidity	1599 non-null	float64
:	1	volatile acidity	1599 non-null	float64
:	2	citric acid	1599 non-null	float64
:	3	residual sugar	1599 non-null	float64
4	4	chlorides	1599 non-null	float64
	5	free sulfur dioxide	1599 non-null	float64
(6	total sulfur dioxide	1599 non-null	float64
-	7	density	1599 non-null	float64
8	8	рН	1599 non-null	float64
9	9	sulphates	1599 non-null	float64
:	10	alcohol	1599 non-null	float64
:	11	quality	1599 non-null	int64

dtypes: float64(11), int64(1)

memory usage: 150.0 KB

```
In [5]: # to display summary of statistics
sd.describe()
```

Out[5]:

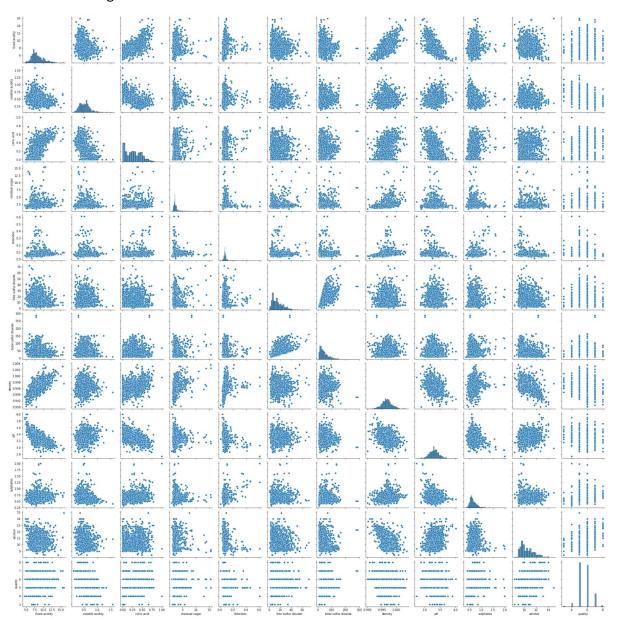
	fixed acidity	volatile acidity	citric acid	residual sugar	chlorides	free sulfur dioxide	total sulfu dioxid
count	1599.000000	1599.000000	1599.000000	1599.000000	1599.000000	1599.000000	1599.00000
mean	8.319637	0.527821	0.270976	2.538806	0.087467	15.874922	46.46779
std	1.741096	0.179060	0.194801	1.409928	0.047065	10.460157	32.89532
min	4.600000	0.120000	0.000000	0.900000	0.012000	1.000000	6.00000
25%	7.100000	0.390000	0.090000	1.900000	0.070000	7.000000	22.00000
50%	7.900000	0.520000	0.260000	2.200000	0.079000	14.000000	38.00000
75%	9.200000	0.640000	0.420000	2.600000	0.090000	21.000000	62.00000
max	15.900000	1.580000	1.000000	15.500000	0.611000	72.000000	289.00000

```
In [6]: #to display colums heading
sd.columns
```

EDA and visualization

In [7]: sns.pairplot(sd)

Out[7]: <seaborn.axisgrid.PairGrid at 0x224a82e2e50>

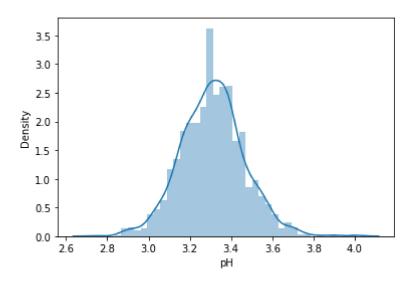


In [8]: sns.distplot(sd['pH'])

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

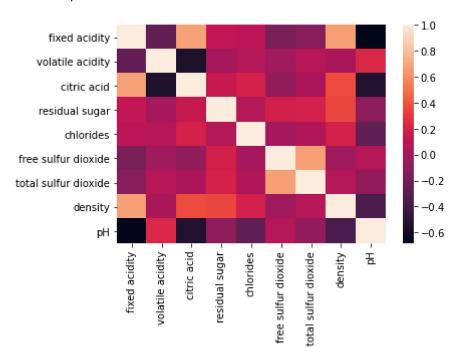
warnings.warn(msg, FutureWarning)

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



In [10]: | sns.heatmap(sd1.corr())

Out[10]: <AxesSubplot:>



TO TRAIN THE MODEL MODEL BUILDING

we are goint train Liner Regression model; we need to split out the data into two varibles x and y where x is independent on x (output) and y is dependent on x(output) adress coloumn as it is not required our model

```
In [11]: x= sd1[['fixed acidity', 'volatile acidity', 'citric acid', 'residual sugar',
                  chlorides', 'free sulfur dioxide', 'total sulfur dioxide', 'density']]
          y=sd1['pH']
In [12]: # To split my dataset into training data and test 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 [13]: from sklearn.linear_model import LinearRegression
          lr=LinearRegression()
          lr.fit(x train,y train)
Out[13]: LinearRegression()
In [14]: | from sklearn.linear_model import LinearRegression
          lr=LinearRegression()
          lr.fit(x_train,y_train)
Out[14]: LinearRegression()
In [15]: |print(lr.intercept_)
          -25.316802604094057
          coeff= pd.DataFrame(lr.coef_,x.columns,columns=['Co-efficient'])
In [16]:
          coeff
Out[16]:
                           Co-efficient
                fixed acidity
                             -0.082906
              volatile acidity
                             0.075282
                  citric acid
                             0.077434
              residual sugar
                             -0.009203
                  chlorides
                             -0.875926
           free sulfur dioxide
                             0.001638
           total sulfur dioxide
                             -0.001125
                    density
                             29.480031
```

```
In [17]: | prediction = lr.predict(x_test)
         plt.scatter(y_test,prediction)
Out[17]: <matplotlib.collections.PathCollection at 0x224b1bd3220>
          3.4
          3.2
          3.0
          2.8
                            3.2
                    3.0
                                    3.4
                                            3.6
                                                    3.8
In [18]: |print(lr.score(x_test,y_test))
         0.5730167122886737
In [19]: |lr.score(x_test,y_test)
Out[19]: 0.5730167122886737
In [20]: |lr.score(x_train,y_train)
Out[20]: 0.5923644193297555
In [21]: from sklearn.linear_model import Ridge,Lasso
In [22]: dr=Ridge(alpha=10)
         dr.fit(x_train,y_train)
Out[22]: Ridge(alpha=10)
In [23]: |dr.score(x_test,y_test)
Out[23]: 0.5370667836824576
In [24]: | dr.score(x_train,y_train)
Out[24]: 0.5080184263006466
In [25]: la=Lasso(alpha=10)
         la.fit(x_train,y_train)
Out[25]: Lasso(alpha=10)
```

```
In [26]: la.score(x_test,y_test)
Out[26]: -0.00023321142972210218
In [27]: la.score(x_train,y_train)
Out[27]: 0.0
In []:
```