ASSIGNMENT DATE	2/11/2022
STUDENT NAME	T. KIRUBA
STUDENT ROLL NUMBER	811519104056
MAXIMUM MARKS	2

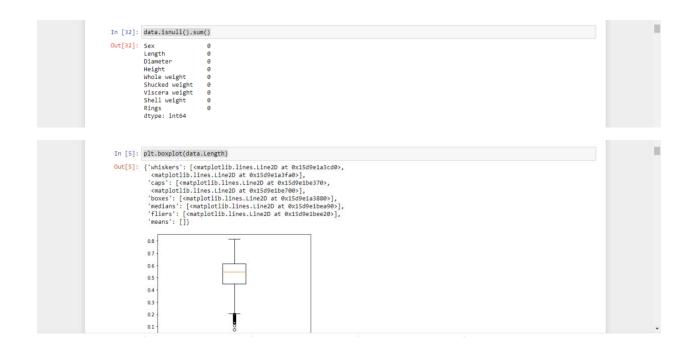
Description: - Predicting the age of abalone from physical measurements. The age of abalone is determined by cutting the shell through the cone, staining it, and counting the number of rings through a microscope -- a boring and time-consuming task. Other measurements, which are easier to obtain, are used to predict age. Further information, such as weather patterns and location (hence food availability) may be required to solve the problem.

## **Building a Regression Model**

- 1. Download the dataset: Dataset
- 2. Load the dataset into the tool.
- 3. Perform Below Visualizations.
- · Univariate Analysis
- · Bi-Variate Analysis
- · Multi-Variate Analysis
- 4. Perform descriptive statistics on the dataset.
- 5. Check for Missing values and deal with them.
- 6. Find the outliers and replace them outliers
- 7. Check for Categorical columns and perform encoding.
- 8. Split the data into dependent and independent variables.
- 9. Scale the independent variables
- 10. Split the data into training and testing
- 11. Build the Model
- 12. Train the Model
- 13. Test the Model
- 14. Measure the performance using Metrics.

```
In [28]: import pandas as pd
        import numpy as np
import seaborn as sns
         import matplotlib.pyplot as plt
In [29]: data=pd.read_csv("C:/Users/admin/Downloads/abalone.csv")
In [30]: data.head()
Out[30]:
          Sex Length Diameter Height Whole weight Shucked weight Viscera weight Shell weight Rings
         0 M 0.455 0.365 0.095
                                     0.5140
                                                     0.2245
                                                                0 1010
                                                                            0.150 15
                                         0.2255
                                                     0.0995
                                                                            0.070
         1 M 0.350 0.265 0.090
                                                                 0.0485
                                                                                    7
                                                                            0.210 9
         2 F 0.530 0.420 0.135
                                        0.6770
                                                     0.2565
                                                                 0.1415
         3 M 0.440 0.365 0.125
                                         0.5160
                                                     0.2155
                                                                 0.1140
                                                                            0.155
                                                                                   10
         4 I 0.330 0.255 0.080
                                       0.2050
                                                     0.0895
                                                                 0.0395
                                                                           0.055 7
```

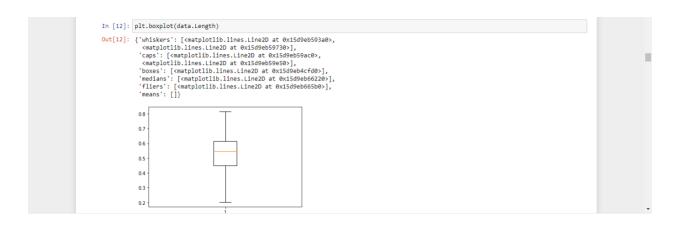
In [31]:	data.d	escribe()								
Out[31]:		Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	Rings	
	count	4177.000000	4177.000000	4177.000000	4177.000000	4177.000000	4177.000000	4177.000000	4177.000000	
	mean	0.523992	0.407881	0.139516	0.828742	0.359367	0.180594	0.238831	9.933684	
	std	0.120093	0.099240	0.041827	0.490389	0.221963	0.109614	0.139203	3.224169	
	min	0.075000	0.055000	0.000000	0.002000	0.001000	0.000500	0.001500	1.000000	
	25%	0.450000	0.350000	0.115000	0.441500	0.186000	0.093500	0.130000	8.000000	
	50%	0.545000	0.425000	0.140000	0.799500	0.336000	0.171000	0.234000	9.000000	
	75%	0.615000	0.480000	0.165000	1.153000	0.502000	0.253000	0.329000	11.000000	
	max	0.815000	0.650000	1,130000	2.825500	1.488000	0.760000	1.005000	29.000000	



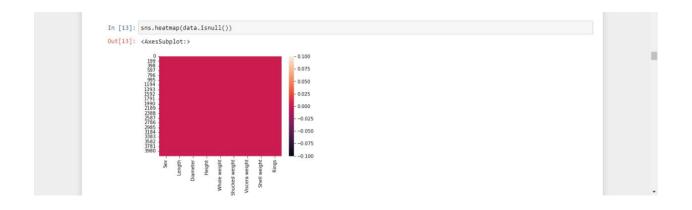
## 0.2025000000000000004 In [10]: data[(data.Length<ue)&(data.Length>le)] Out[10]: Sex Length Diameter Height Whole weight Shucked weight Viscera weight Shell weight Rings **0** M 0.455 0.365 0.095 0.5140 0.2245 0.1010 0.1500 15 1 M 0.350 0.265 0.090 0.2255 0.0995 0.0485 0.0700 **2** F 0.530 0.420 0.135 0.6770 0.2565 0.1415 0.2100 9 3 M 0.440 0.365 0.125 0.5160 0.2155 0.1140 0.1550 10 4 I 0.330 0.255 0.080 0.2050 0.0895 0.0395 0.0550 7 **4172** F 0.565 0.450 0.165 0.8870 0.3700 0.2390 0.2490 11 4173 M 0.590 0.440 0.135 0.9660 0.4390 0.2145 0.2605 10 4174 M 0.600 0.475 0.205 1.1760 0.5255 0.2875 0.3080 9 4175 F 0.625 0.485 0.150 1.0945 0.5310 0.2610 0.2960 10 4176 M 0.710 0.555 0.195 1.9485 0.9455 0.3765 0.4950 12



4128 rows x 9 columns

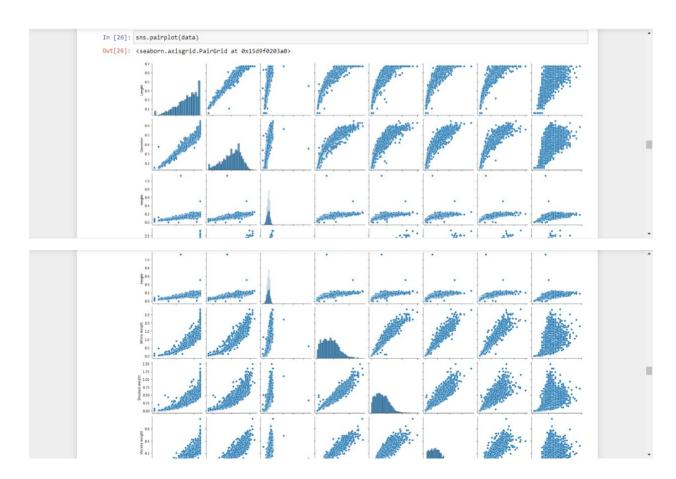


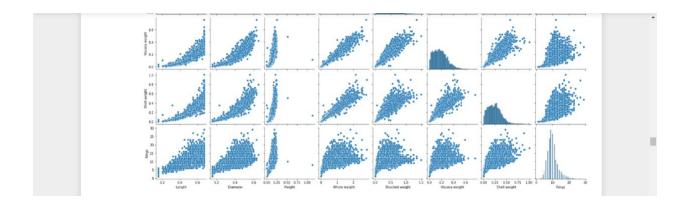




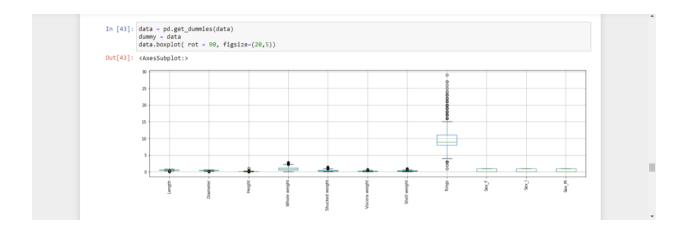


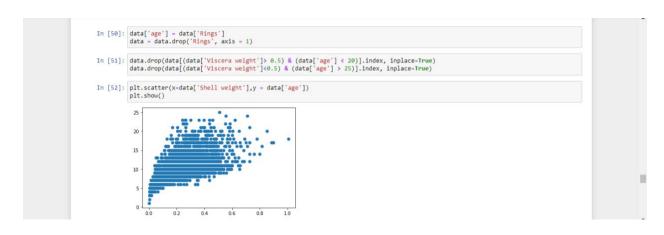






```
In [36]: data[data.Height == 0]
Out[36]: Sex Length Diameter Height Whole weight Shucked weight Viscera weight Shell weight Rings
        1257 I 0.430 0.34 0.0 0.428 0.2065 0.0860 0.1150 8
         3996 I 0.315
                         0.23 0.0
                                          0.134
                                                     0.0575
                                                                0.0285
                                                                          0.3505
In [37]: data['Shucked weight'].kurtosis()
Out[37]: 0.5951236783694207
In [38]: data['Diameter'].median()
Out[38]: 0.425
In [39]: data['Shucked weight'].skew()
                                                                                                                                          Out[39]: 0.7190979217612694
```





```
In [54]: numerical_features = data.select_dtypes(include = [np.number]).columns

categorical_features = data.select_dtypes(include = [np.object]).columns

C:\Users\admin\AppData\Local\Temp/ipykernel_4524/3364968343.py:2: DeprecationWarning: `np.object` is a deprecated alias for the builtin 'object'. To silence this warning, use 'object' by itself. Doing this will not modify any behavior and is safe.

Deprecated in NumPy 1.20; for more details and guidance: https://numpy.org/devdocs/release/1.20.0-notes.html#deprecations categorical_features = data.select_dtypes(include = [np.object]).columns

In [55]: numerical_features

categorical_features

categoric
```

```
In [58]: abalone_numeric.head()
Out[58]:
               Length Diameter Height Whole weight Shucked weight Viscera weight Shell weight age Sex_F Sex_I Sex_M
              0 0.455 0.365 0.095 0.5140 0.2245 0.1010 0.150 15 0 0 1
              1 0.350
                             0.265 0.090
                                                       0.2255
                                                                          0.0995
                                                                                           0.0485
                                                                                                           0.070 7
                                                                                                                              0
             2 0.530 0.420 0.135 0.6770 0.2565 0.1415 0.210 9 1 0 0

        3
        0.440
        0.365
        0.125
        0.5160
        0.2155
        0.1140
        0.155
        10
        0
        0
        1

        4
        0.330
        0.255
        0.080
        0.2050
        0.0895
        0.0395
        0.055
        7
        0
        1
        0

In [59]: from sklearn.preprocessing import MinMaxScaler
scaler=MinMaxScaler()
model=scaler.fit(data)
             scaled data-model.transform(data)
             print(scaled_data)
             [[0.51351351 0.5210084 0.0840708 ... 0.
[0.37162162 0.35294118 0.07964602 ... 0.
[0.61486486 0.61344538 0.11946903 ... 0.
                                                                                                   0.583333331
                                                                                                   0.25 ]
0.333333333]
                                                                                                                                                                                                                         [0.70945946 0.70588235 0.18141593 ... 0.
                                                                                                   0.33333333]
               [0.74324324 0.72268908 0.13274336 ... 0.
[0.85810811 0.84033613 0.17256637 ... 0.
                                                                                                   0.45833333]]
```

```
In [62]: from sklearn.model_selection import train_test_split
train_x,test_x,train_y,test_y-train_test_split(x,y,test_size=0.2,random_state=25)

In [63]: from sklearn.linear_model import LinearRegression
model_LinearRegression()

In [64]: model.fit(train_x,train_y)

Out[64]: LinearRegression()

In [65]: model.predict(test_x)

Out[65]: array([ 9.94801644, 12.46888676, 6.53380794, 4.46644662, 9.63517262,
8.74456903, 9.09747127, 13.29464626, 9.38111887, 9.21741905,
5.665421 , 7.99255495, 9.87593558, 9.63938237, 9.79547
10.63946681, 9.47755474, 11.71758488, 8.4292209, 9.85388942,
7.60136481, 11.03319821, 12.66507999, 11.94267943, 9.21941446,
10.74556065, 10.802506121, 8.64506228, 12.26465815, 8.0870857,
8.76986532, 9.80191058, 11.21690827, 12.31902844, 6.6344747,
7.76173894, 13.15952026, 11.2086834, 11.1937116, 11.01660578,
8.94430032, 8.48179814, 8.27284612, 10.0837559, 9.81437691,
11.44122216, 8.10659319, 13.41844322, 9.13833438, 7.16260984,
10.66778019, 12.060996671, 9.64265261, 11.4447366, 9.10260988,
9.07829066, 7.7929642, 10.90121456, 8.95506527, 5.66173825,
7.70229579, 9.67405012, 7.85663108, 12.02022154, 10.442759406,
```

```
In [66]: predicted_y=model.predict(test_x)

In [67]: model.score(test_x,test_y)

Out[67]: 0.4907507202674961

In [68]: from sklearn.metrics import mean_squared_error

In [69]: mean_squared_error(test_y,predicted_y)
Out[69]: 4.72907775645871
```