Assignment Date	06-10-2022
Student Name	Mr. jeyaprasanna
Student Roll Number	923819104017
Maximum Marks	2 Mark

## Question-1:

## 1. Importing Required Package

#### **Solution:**

```
import pandas as pd
import seaborn as sns
import numpy as np
from matplotlib import pyplot as plt
%matplotlib inline
```

### Question-2:

## 2. Loading the Dataset

#### **Solution:**

```
df = pd.read_csv("abalone.csv")
df
```

	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	7
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	- 1	0.330	0.255	0.080	0.2050	0.0895	0.0395	0.0550	7
	000	5235	ELECTR		1999			777	
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
4177 rd	ws ×	9 columns	S						

### 3. Visualizations

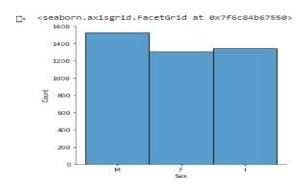
Question-3:

# 3.1 Univariate Analysis

#### **Solution:**

sns.displot(df.Sex)

## **Output:**



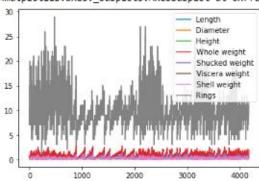
## 3.2 Bi-Variate Analysis

#### **Solution:**

df.plot.line()

## **Output:**

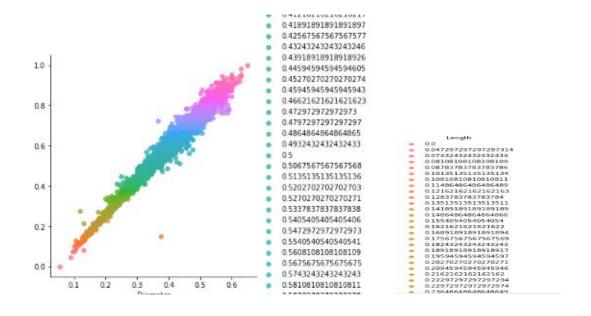
<matplotlib.axes.\_subplots.AxesSubplot at 0x7f6c81d866d0>



# 3.3 Multi - Variate Analysis

#### **Solution:**

```
sns.lmplot("Diameter", "Length", df, hue="Length", fit_reg=False);
```



### 4. Perform descriptive statistics on the dataset.

Question-4:

#### **Solution:**

df.describe()

	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

## **5. Handle the Missing values.**

Question-5:

### **Solution:**

```
data = pd.read_csv("abalone.csv")
pd.isnull(data["Sex"])
```

```
0 False
1 False
2 False
3 False
4 False
4 False
4172 False
4173 False
4174 False
4175 False
4176 False
Name: Sex, Length: 4177, dtype: bool
```

## Question-6:

### 6. Find the outliers and replace the outliers.

#### **Solution:**

```
df["Rings"] = np.where(df["Rings"] >10, np.median,df["Rings"])
df["Rings"]
```

### **Output:**

## Question-7:

### 7. Check for Categorical columns and perform encoding.

### **Solution:**

D	iameter	Height	weight	weight	viscera weight	weight	Rings	Length_F	Length_I	Length_	м.	•	Sex_0.745	Sex_0.75	Sex_0.755	Sex_0.76	Sex_0.765	Sex_0.77	Sex_0.775	Sex_0.78	Sex_0.8
0	0.365	0.095	0.5140	0.2245	0.1010	0.150	<function at<br="" median="">0x7f6c9fd64cb0&gt;</function>	0	0		1		0	0	0	0	0	0	0	0	0
1	0.265	0.090	0.2255	0.0995	0.0485	0.070	7	0	D		1		0	0	0	0	0	0	0	0	0
2	0.420	0.135	0.6770	0.2565	0.1415	0.210	9	1	0		0		0	0	0	0	0	0	0	0	0
3	0.365	0.125	0.5160	0.2155	0.1140	0.155	10	0	0		1		0	0	0	0	0	0	0	. 0	0
4	0.255	0.080	0.2050	0.0895	0.0395	0.055	7	0	1		0		0	0	0	0	0	0	0	0	0
5 rows	× 144 col	lumns																			

## Question-8:

### 8. Split the data into dependent and independent variables

## 8.1 Split the data into Independent variables.

### **Solution:**

```
X = df.iloc[:, :-2].values
print(X)
```

```
[['M' 0.455 0.365 ... 0.514 0.2245 0.101]
['M' 0.35 0.265 ... 0.2255 0.0995 0.0485]
['F' 0.53 0.42 ... 0.677 0.2565 0.1415]
...
['M' 0.6 0.475 ... 1.176 0.5255 0.2875]
['F' 0.625 0.485 ... 1.0945 0.531 0.261]
['M' 0.71 0.555 ... 1.9485 0.9455 0.3765]]
```

### 8.2 Split the data into Dependent variables.

#### **Solution:**

```
Y = df.iloc[:, -1].values
print(Y)
```

## **Output:**

```
[<function median at 0x7f6c9fd64cb0> 7 9 ... 9 10 
<function median at 0x7f6c9fd64cb0>]
```

## Question-9:

## 9. Scale the independent variables

#### **Solution:**

```
import pandas as pd
from sklearn.preprocessing import MinMaxScaler
scaler = MinMaxScaler()
df[["Length"]] = scaler.fit_transform(df[["Length"]])
print(df)
```

```
Length Diameter Height Whole weight Shucked weight \
   Sex
     M 0.513514 0.365 0.095 0.5140
                                            0.2245
    M 0.371622 0.265 0.090 0.2255
F 0.614865 0.420 0.135 0.6770
M 0.493243 0.365 0.125 0.5160
                                                  0.0995
1
2
                                                  0.2565
     I 0.344595 0.255 0.080
                                   0.2050
                                                  0.0895
0.3700
                                                  0.4390
                                                  0.5255
4175 F 0.743243 0.485 0.150
                                    1.0945
                                                  0.5310
4176 M 0.858108 0.555 0.195
                                    1.9485
                                                  0.9455
     Viscera weight Shell weight
                       0.1500 <function median at 0x7f6c9fd64cb0>
0
           0.1010
1
           0.0485
                       0.0700
           0.1415
2
                       0.2100
3
           0.1140
                       0.1550
                                                           10
4
           0.0395
                       0.0550
                                                           7
                      0.2490 <function median at 0x7f6c9fd64cb0>
           0.2390
4172
4173
           0.2145
                       0.2605
4174
           0.2875
                       0.3080
4175
           0.2610
                       0.2960
                                                           10
4176
           0.3765
                       0.4950 <function median at 0x7f6c9fd64cb0>
[4177 rows x 9 columns]
```

#### Question-10:

#### 10. Split the data into training and testing

#### Solution:

```
from sklearn.model selection import train test split
train size=0.8
X = df.drop(columns = ['Sex']).copy()
y = df['Sex']
X train, X rem, y train, y rem = train test split(X,y, train size=0.8)
test size = 0.5
X valid, X_test, y_valid, y_test = train_test_split(X_rem,y_rem, test
size=0.5)
print(X train.shape), print(y train.shape)
print(X valid.shape), print(y_valid.shape)
print(X test.shape), print(y test.shape)
```

## **Output:**

```
(3341, 8)
(3341,)
(418, 8)
(418,)
(418, 8)
(418,)
(None, None)
```

#### 11.Build the Model

```
test_size = 0.33
seed = 7
X_train, X_test, y_train, y_test = train_test_split(X, y, test_s
ize=test size, random state=seed)
```

#### 12.Train the model

X train

	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	Rings
4131	0.682432	0.425	0.145	0.8300	0.3790	0.1605	0.2575	<function 0x7f6c9fd64cb0="" at="" median=""></function>
3204	0.797297	0.530	0.185	1.3955	0.4560	0.3205	0.4900	<function 0x7f6c9fd64cb0="" at="" median=""></function>
2622	0.844595	0.575	0.205	1.7975	0.7295	0.3935	0.5165	<function 0x7f6c9fd64cb0="" at="" median=""></function>
2114	0.074324	0.095	0.035	0.0105	0.0050	0.0065	0.0035	4
1422	0.871622	0.575	0.215	2.1730	0.9515	0.5640	0.5365	<function 0x7f6c9fd64cb0="" at="" median=""></function>
	****		150			1000	150	
1372	0.729730	0.475	0.165	1.0230	0.4905	0.1955	0.3035	<function 0x7f6c9fd64cb0="" at="" median=""></function>
919	0.452703	0.310	0.090	0.3335	0.1635	0.0610	0.0910	6
2550	0.277027	0.220	0.080	0.1315	0.0660	0.0240	0.0300	5
537	0.290541	0.230	0.075	0.1165	0.0430	0.0255	0.0400	7
1220	0.344595	0.250	0.095	0.2085	0.1020	0.0395	0.0520	7

```
y_train
```

# Output:

```
4131 I
3204 F
2622 F
2114 I
1422 M
...
1372 F
919 I
2550 I
537 M
1220 I
Name: Sex, Length: 2798, dtype: object
```

### 13.Test the model:

X\_test

## **Output:**

	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	Rings	6
1157	0.716216	0.470	0.165	1.1775	0.6110	0.2275	0.2920	9	
1125	0.641892	0.425	0.150	0.8315	0.4110	0.1765	0.2165	10	
2053	0.520270	0.345	0.110	0.4595	0.2350	0.0885	0.1160	7	
3591	0.777027	0.475	0.165	1.3875	0.5800	0.3485	0.3095	9	
455	0.675676	0.470	0.140	0.8375	0.3485	0.1735	0.2400	<function 0x7f6c9fd64cb0="" at="" median=""></function>	
	;			1944	***	14.7	Since		
3150	0.783784	0.505	0.165	1.3670	0.5835	0.3515	0.3960	10	
3037	0.655405	0.450	0.145	0.8940	0.3885	0.2095	0.2640	9	
2050	0.506757	0.350	0.130	0.4655	0.2075	0.1045	0.1350	8	
1690	0.743243	0.500	0.170	1.0985	0.4645	0.2200	0.3540	9	
253	0.675676	0.460	0.185	1.0940	0.4485	0.2170	0.3450	<function 0x7f6c9fd64cb0="" at="" median=""></function>	

y\_test

```
1157
1125
      14
     М
2053
3591
      M
455
3150
3037
      M
     M
2050
1690 M
253
Name: Sex, Length: 1379, dtype: object
```

#### 14. Measure the performance using Metrics

```
from sklearn.metrics import r2_score
from sklearn.metrics import mean_absolute_error
from sklearn.metrics import mean_squared_error
X_train = [5, -1, 2, 10]
y_test = [3.5, -0.9, 2, 9.9]
print ('R Squared =', r2_score(X_train, y_test))
print ('MAE =', mean_absolute_error(X_train, y_test))
print ('MSE =', mean_squared_error(X_train, y_test))
```

```
R Squared = 0.9656060606060606
MAE = 0.424999999999999
MSE = 0.56749999999999
```