

## ASSIGNMENT -4

### RETAIL STORE STOCK INVENTORY ANALYTICS

TEAM ID: PNT2022TMID08352

NAME: S.Navaneethakrishnan

**Answer the questions or complete the tasks:**

#### 1. LOADING THE DATASET

##### 1. LOADING THE DATASET

```
[ ] import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns

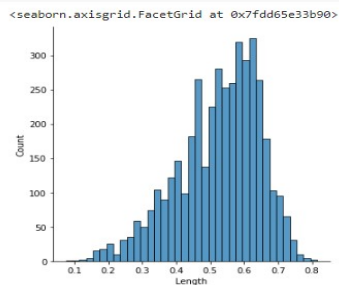
## 2.load the dataset
data = pd.read_csv('abalone.csv')
data.head()
```

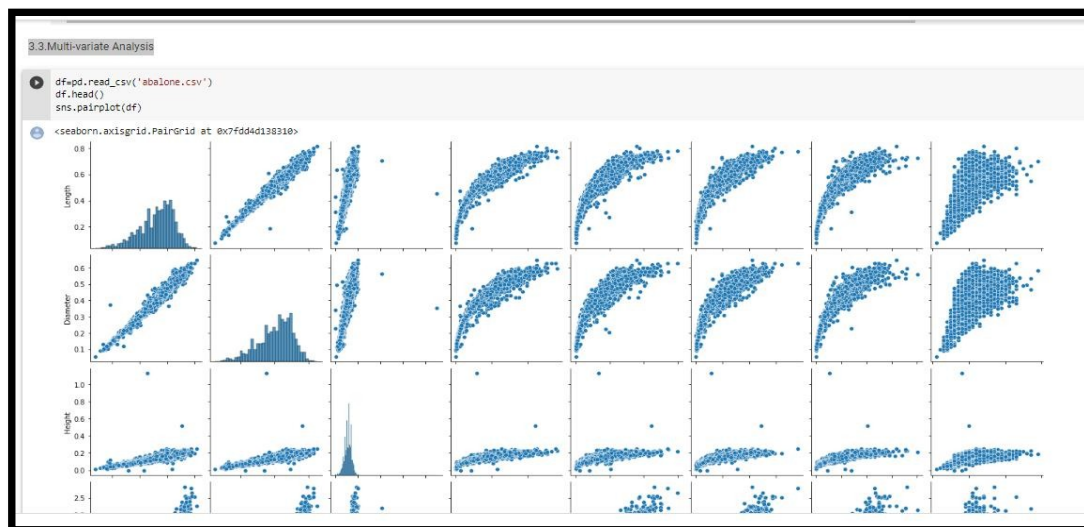
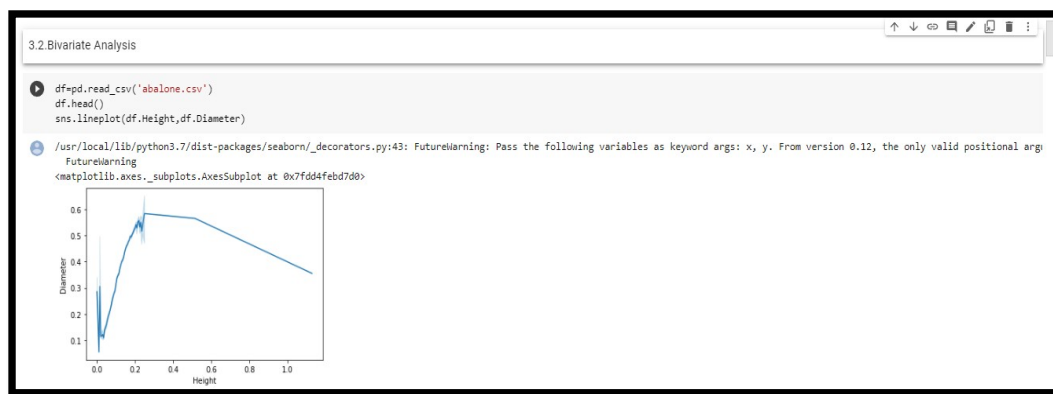
	Sex	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	Rings
0	M	0.455	0.365	0.095	0.5140	0.2245	0.1010	0.150	15
1	M	0.350	0.265	0.090	0.2255	0.0995	0.0485	0.070	7
2	F	0.530	0.420	0.135	0.6770	0.2565	0.1415	0.210	9
3	M	0.440	0.365	0.125	0.5160	0.2155	0.1140	0.155	10
4	I	0.330	0.255	0.080	0.2050	0.0895	0.0395	0.055	7

#### 2. PERFORM THE VISUALIZATION

##### 3.1. Univariate Analysis

```
[ ] df=pd.read_csv('abalone.csv')
df.head()
sns.displot(df.Length)
```





### 3. PERFORM DESCRIPTIVE STATISTICS ON THE DATASET

#### 4.PERFORM DESCRIPTIVE STATISTICS ON THE DATASET

```
[ ] 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

#### 4. CHECK FOR MISSING VALUES AND DEAL WITH THEM

```
6.FIND THE OUTLIERS AND REPLACE THEM OUTLIERS

data=pd.read_csv('abalone.csv')
data.head()
Q1=data.length.quantile(0.25)
Q3=data.length.quantile(0.75)
Q1,Q3

(0.45, 0.615)

[ ] IQR=Q3-Q1
IQR
lower_limit = Q1-1.5*IQR
upper_limit = Q3+1.5*IQR
lower_limit, upper_limit
data_no_outlier = data[(data.Length>lower_limit)&(data.Length<upper_limit)]
data_no_outlier
```

#### 5. FIND THE OUTLIERS AND REPLACE THEM OUTLIERS

```
6.FIND THE OUTLIERS AND REPLACE THEM OUTLIERS

[ ] data=pd.read_csv('abalone.csv')
data.head()
Q1=data.length.quantile(0.25)
Q3=data.length.quantile(0.75)
Q1,Q3

(0.45, 0.615)

IQR=Q3-Q1
IQR
lower_limit = Q1-1.5*IQR
upper_limit = Q3+1.5*IQR
lower_limit, upper_limit
data_no_outlier = data[(data.Length>lower_limit)&(data.Length<upper_limit)]
data_no_outlier
```

	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	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	F	0.710	0.555	0.195	1.9485	0.9455	0.3765	0.4950	12

#### 6. CHECK FOR CATEGORICAL COLUMNS AND PERFORM ENCODING

```
7.CHECK FOR CATEGORICAL COLUMNS AND PERFORM ENCODING

data['Sex'].replace({'M':1,'F':0,'I':2},inplace=True)
data
```

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

4177 rows x 9 columns

## 7. SPLIT THE DATA INTO DEPENDENT AND INDEPENDENT VARIABLES

8. SPLIT THE DATA INTO DEPENDENT AND INDEPENDENT VARIABLES

```
x=data.drop(columns= ['Rings'])
y=data['Rings']
x
```

	Sex	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight
0	1	0.455	0.365	0.095	0.5140	0.2245	0.1010	0.1500
1	1	0.350	0.265	0.090	0.2255	0.0995	0.0485	0.0700
2	0	0.530	0.420	0.135	0.6770	0.2565	0.1415	0.2100
3	1	0.440	0.365	0.125	0.5160	0.2155	0.1140	0.1550
4	2	0.330	0.255	0.080	0.2050	0.0895	0.0395	0.0550
...	...	...	...	...	...	...	...	...
4172	0	0.565	0.450	0.165	0.8870	0.3700	0.2390	0.2490
4173	1	0.590	0.440	0.135	0.9660	0.4390	0.2145	0.2605
4174	1	0.600	0.475	0.205	1.1760	0.5255	0.2875	0.3080
4175	0	0.625	0.485	0.150	1.0945	0.5310	0.2610	0.2960
4176	1	0.710	0.555	0.195	1.9485	0.9455	0.3765	0.4950

4177 rows x 8 columns

```
[ ] y
0      15
1       7
2       9
3      10
4       7
..
4172    11
4173    10
4174     9
4175    10
4176    12
Name: Rings, Length: 4177, dtype: int64
```

## 8. SCALE THE INDEPENDENT VARIABLES

9. SCALE THE INDEPENDENT VARIABLES

```
from sklearn.preprocessing import scale
x = scale(x)
x
```

```
array([[ -0.0105225, -0.57455813, -0.43214879, ..., -0.60768536,
        -0.72621157, -0.63821689],
       [ -0.0105225, -1.44898585, -1.439929, ..., -1.17090984,
        -1.20522124, -1.21298732],
       [ -1.26630752,  0.05003309,  0.12213032, ..., -0.4634999,
        -0.35668983, -0.20713907],
       ...,
       [ -0.0105225,  0.6329849,  0.67640943, ...,  0.74855917,
        0.97541324,  0.49695471],
       [ -1.26630752,  0.84118198,  0.77718745, ...,  0.77334105,
        0.73362741,  0.41073914],
       [ -0.0105225,  1.54905203,  1.48263359, ...,  2.64099341,
        1.78744868,  1.84048058]])
```

## 9. SPLIT THE DATA INTO TRAINING AND TESTING

10. SPLIT THE DATA INTO TRAINING AND TESTING

```
[ ] from sklearn.model_selection import train_test_split
x_train, x_test, y_train, y_test = train_test_split(x,y, test_size = 0.2)
print(x_train.shape, x_test.shape)
```

```
(3341, 8) (836, 8)
```

## 10. BUILD THE MODEL

### 11. BUILD THE MODEL

```
[ ] from sklearn.linear_model import LinearRegression
    MLR=LinearRegression()
```

## 11. TRAIN THE MODEL

### 12. TRAIN THE MODEL

```
[ ] MLR.fit(x_train,y_train)
```

```
LinearRegression()
```

## 12. TEST THE MODEL

### 13. TEST THE MODEL

```
y_pred=MLR.predict(x_test)
y_pred
```

```
array([14.44666767,  7.40745222, 10.78252097,  6.67673552,  8.36060517,
        9.93173721,  7.5379131,  8.27735969, 12.06608319, 12.29583434,
        8.29574382, 10.00474082, 10.12118388, 20.41663106, 18.40287657,
        9.01410145,  8.15149249, 11.17728349,  7.6817167,  9.23627153,
        7.22475815, 10.37839064,  9.63344681, 10.47074632, 11.53950746,
        10.95434036, 10.05460627,  8.66984804,  8.32478841,  8.9477472,
        11.28292888, 11.88502862,  7.77967711,  9.23899781,  8.48185467,
        6.59484556,  9.63756273,  9.01573937, 11.92456149,  9.52820927,
        10.14646238,  9.962085,  7.37530868, 11.06812272,  9.44867903,
        9.44418147,  7.34762047, 11.16558831, 13.92827424,  9.22663544,
        11.99434293, 10.43123452, 10.32767995,  6.11983576,  6.79521823,
        10.05528337,  8.53284124,  8.04305494,  9.92171955,  7.89210892,
        6.07840864,  9.9446442,  9.85338799,  6.68752504,  8.05979124,
        10.96333799,  7.27673389,  7.60319848,  6.64672865, 13.12749717,
        7.90163657, 12.47507082,  7.31744113,  6.50778142,  8.11198339,
        8.64057198,  6.53963896,  6.45664839,  8.75598234, 10.4890979,
        9.76947248,  8.90006123,  9.31761142,  9.35957542,  7.34115295,
        13.43047766,  7.39556291, 15.7252517,  6.65972367,  8.2367926,
        8.74726516, 10.15130117, 10.33317009,  8.690774,  7.56742459,
        11.28527425, 13.60151677, 13.68633873, 11.75750974,  9.79457312,
        10.48453644,  6.6401399,  8.25499507,  7.43644897,  7.91321069,
        11.3251912,  8.46296412,  4.64548299,  6.12181089,  8.25718621,
        7.49029784,  8.57641168,  7.99613531, 10.81766123,  9.42265379,
        9.08713187, 10.09224284,  9.50202631, 13.15222633,  7.80309981,
        10.17247184, 10.20195302, 12.61893172,  9.0512437, 10.06718884,
```