#### **ASSIGNMENT - 4**

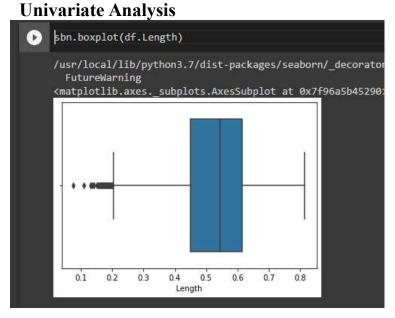
Assignment Date	25 October 2022
Student Name	DHANALAKSHMI K
Student Roll Number	822719104012
Maximum Marks	2 Marks

**Dataset**: <a href="https://drive.google.com/file/d/1sIv-7x7CE0zAPAt0Uv-6pbO2ST2LVp5u/view">https://drive.google.com/file/d/1sIv-7x7CE0zAPAt0Uv-6pbO2ST2LVp5u/view</a>

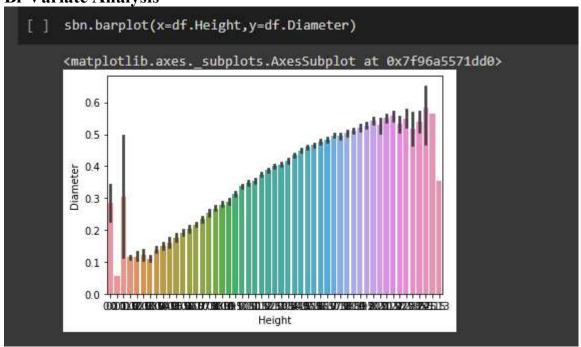
### **Loading the dataset:**

	mport i	pandas a numpy as matplotl seaborn	np ib.pyplot	as plt						
	f.head	()	"abalone.c		Whole weight	Shucked weight	Viscera weight	Shell weight	Rings	7
(	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	2 F	0.530	0.420	0.135	0.6770	0.2565	0.1415	0.210		
4	3 M	0.440	0.365	0.125	0.5160	0.2155	0.1140	0.155	10	

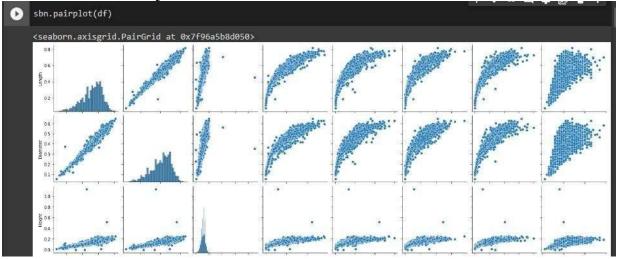
# Perform Below Visualizations.



### **Bi-Variate Analysis**



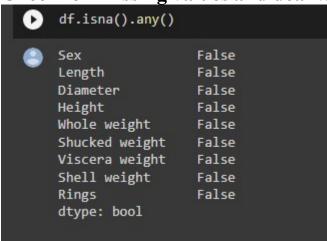




Perform descriptive analytics on the dataset

```
[ ] df['Height'].mean()
    0.13951639932966242
[ ] df['Diameter'].median()
    0.425
[ ] df['Length'].mode()
    0
         0.550
         0.625
    dtype: float64
  df.max()
      Sex
      Length
                        0.815
      Diameter
                         0.65
      Height
                         1.13
      Whole weight
                       2.8255
      Shucked weight
                       1.488
      Viscera weight
                        0.76
      Shell weight
                        1.005
      Rings
                           29
      dtype: object
 [ ] df.min()
      Sex
      Length
                        0.075
      Diameter
                        0.055
      Height
                          0.0
      Whole weight
                        0.002
      Shucked weight
                        0.001
      Viscera weight
                       0.0005
      Shell weight
                       0.0015
      Rings
                            1
```

Check for Missing values and deal with them.



Find the outliers and replace them outliers

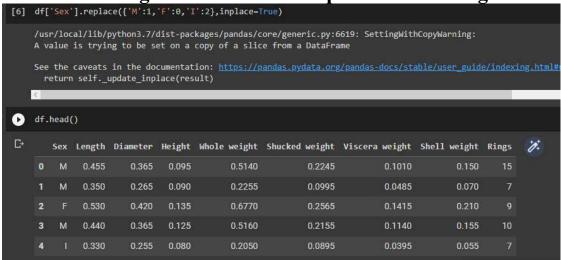
```
[3] q1=df.Rings.quantile(0.25)
    q3=df.Rings.quantile(0.75)
    iqr=q3-q1

[4] print(iqr)

[5] 3.0

df=df[~((df.Rings<(q1-1.5*iqr))|(df.Rings>(q3+1.5*iqr)))]
```

Check for Categorical columns and perform encoding.



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

```
x=df.iloc[:, :-1].values

y=df.iloc[:, -1].values

y=df.iloc[:, -1].values
```

## Scale the independent variables

```
[39] from sklearn.preprocessing import StandardScaler
     std=StandardScaler()
     x=std.fit transform(x)
     array([[-0.03822742, -0.55104264, -0.40422906, ..., -0.58564588,
             -0.69758868, -0.60447624],
                                   , -1.42309849, ..., -1.14600915,
            [-0.03822742, -1.4332
             -1.17989471, -1.21362086],
            [-1.2907376 , 0.07906976, 0.15614912, ..., -0.44219288,
             -0.32552403, -0.14761778],
            [-0.03822742, 0.66717467, 0.71652731, ..., 0.76370889,
              1.01574608, 0.59858438],
            [-1.2907376 , 0.87721213, 0.81841425, ..., 0.78836487,
              0.77229637, 0.50721269],
            [-0.03822742, 1.59133952, 1.53162285, ..., 2.64652949,
             1.83336964, 2.02245992]])
```

Split the data into training and testing

```
[60] 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,random_state=42)
[61] x_train
      array([[0.
                      , 0.695 , 0.53 , ..., 0.75 , 0.4195, 0.6095],
                     , 0.525 , 0.41 , ..., 0.4065, 0.198 , 0.177 ],
                     , 0.64 , 0.485 , ..., 0.456 , 0.2245, 0.2835],
                     , 0.595 , 0.47 , ..., 0.4515, 0.178 , 0.155 ],
              [1.
                     , 0.555 , 0.46 , ..., 0.3345, 0.1935, 0.275 ],
                     , 0.36 , 0.27 , ..., 0.097 , 0.0405, 0.065 ]])
[62] y_train
      array([14, 8, 9, ..., 11, 10, 6])
[63] x_test
       array([[1.
                          , 0.7 , 0.565 , ..., 0.895 , 0.3355, 0.446 ],
                          , 0.735 , 0.6 , ..., 1.1335, 0.44 , 0.6
                [0.
                          , 0.61 , 0.495 , ..., 0.3705, 0.3135, 0.33
                [0.
                                                                                     ],
                [0.
                          , 0.66 , 0.53 , ..., 0.493 , 0.245 , 0.49
                          , 0.555 , 0.435 , ..., 0.341 , 0.1645, 0.214 ],
                [1.
                          , 0.505 , 0.39 , ..., 0.2595, 0.18 , 0.19 ]])
                [1.
[64] y_test
 array([ 9, 11, 12, 15, 9, 7, 9, 9, 9, 11, 10, 9, 7, 11, 8, 12, 10,
              6, 10, 8, 11, 6, 11, 10, 10, 10, 7, 14, 11, 8, 9, 10, 15, 9,
              9, 11, 15, 8, 10, 8, 15, 10, 14, 12, 9, 10, 14, 9, 10, 5, 7, 10, 11, 13, 9, 9, 13, 7, 11, 9, 10, 10, 13, 8, 9, 8, 9, 7, 7, 8, 11, 8, 4, 11, 7, 9, 8, 11, 10, 10, 14, 6, 6, 4, 11, 10, 8, 7, 6, 12, 12, 11, 13, 11, 10, 10, 12, 5, 11, 13, 9, 12, 10, 10, 11, 10, 9, 8, 11, 14, 11, 9, 6, 7, 9, 7, 6, 11, 9,
              13, 5, 10, 9, 9, 9, 12, 9, 9, 8, 11, 11, 10, 7, 11, 8, 11,
              11, 11, 12, 12, 5, 9, 9, 11, 8, 6, 10, 9, 11, 9, 7, 7, 10, 12, 8, 11, 9, 12, 11, 8, 11, 10, 12, 9, 9, 10, 9, 9, 15, 4, 14, 9, 7, 10, 11, 5, 9, 8, 8, 8, 10, 12, 13, 12, 11, 10, 15, 9, 9, 9, 9, 13, 6, 8, 11, 11, 11, 9, 8, 9, 10, 7, 9,
               5, 8, 12, 11, 9, 8, 9, 10, 11, 7, 6, 4, 12, 9, 6, 7, 8,
              13, 12, 12, 10, 14, 10, 12, 9, 9, 13, 9, 10, 13, 8, 15, 8, 10,
              12, 11, 10, 5, 11, 11, 15, 14, 13, 12, 7, 11, 10, 13, 9, 6, 15,
```

#### **Build the Model**

from sklearn.ensemble import RandomForestRegressor model = RandomForestRegressor(n\_estimators = 1000, oob\_score = True,n\_jobs=-1,min\_samples\_split = 6, min\_samples\_leaf= 4, max\_features = 'sqrt', max\_depth= 120, bootstrap=True)

```
from sklearn.ensemble import RandomForestRegressor
model = RandomForestRegressor(n_estimators = 1000, oob_score = True,n_jobs=-1,min_samples_split = 6, min_samples_lear
```

#### Train the Model

#### **Test the Model**

```
predictions=model.predict(x_test)
    predictions
           9.23052686, 6.76527568, 6.27337663, 9.7808718, 10.46575533,
\Box
          10.39856318, 9.92302597, 7.03874443, 9.28506128, 4.8144354,
           8.51898345, 9.44591446, 10.50450779, 10.28790825, 10.1401078,
           7.95223754, 5.30119942, 9.96964081, 6.82311145, 6.29814986,
           8.68373737, 8.21113623, 10.6245237, 10.77857176, 11.17060581,
           9.16360497, 10.28201394, 6.6367132, 10.49952107, 8.41476732,
           9.11490296, 10.11751273, 8.49518805, 4.88652692, 10.28148647,
          10.94575126, 11.71629647, 9.46380019, 9.44207265, 10.21271332,
           9.14684877, 9.86565957, 8.92327854, 10.88901169, 10.58669074,
           8.954949 , 12.25015427, 10.70193653, 11.64170245, 8.81236519,
           8.06411968, 5.5665906, 8.73177525, 11.59118191, 10.65204263,
           9.18393415, 11.58186427, 6.54125027, 10.43332356, 6.94692004,
          11.27852383, 9.31304977, 8.40214749, 6.02948651, 12.03950182,
           6.58799368, 11.31287941, 11.37077235, 4.7255203 , 11.15012629,
          10.0408263 , 7.73944001, 6.9423391 , 4.90132305, 10.40211536,
          10.04235146, 6.96710608, 11.05620166, 11.35397795, 10.22259343,
          11.63211032, 9.39309664, 8.88237849, 10.83092528, 6.6303001,
          11.52583068, 10.787237 , 9.93738872, 11.74766958, 10.45900969,
           7.60619186, 9.82836881, 9.69601129, 10.5296791, 9.20391431,
           9.00121742, 9.79719374, 10.45730253, 8.39235724, 7.41134463,
```

**Measure the performance using Metrics.** 

```
[93] from sklearn.metrics import r2_score
acc=r2_score(y_test,predictions)
acc
0.5902139902351261
```