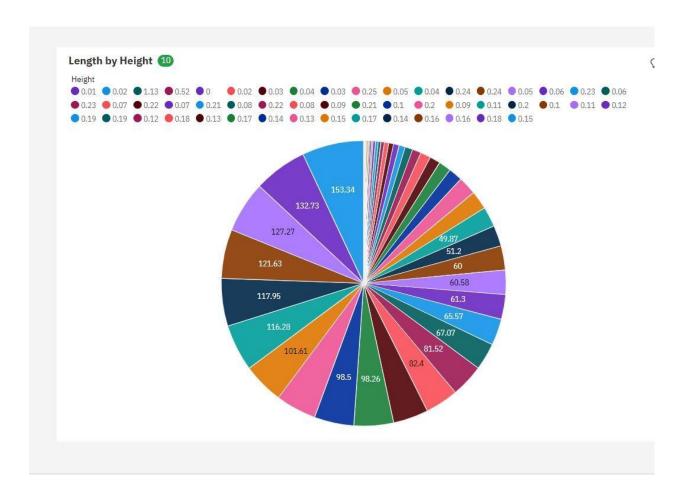
## **ASSIGNMENT 4**

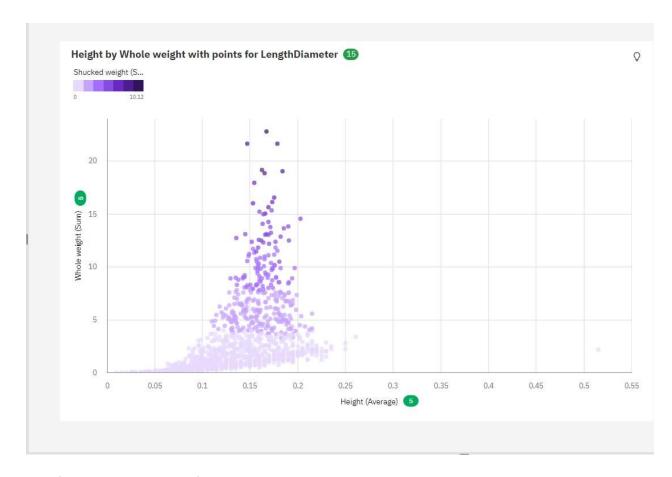
Assignment Date	29 oct 2022
Student Name	S.Dhanalakshmi
Student Roll Number	420619104012
Maximum Mark	2

- 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

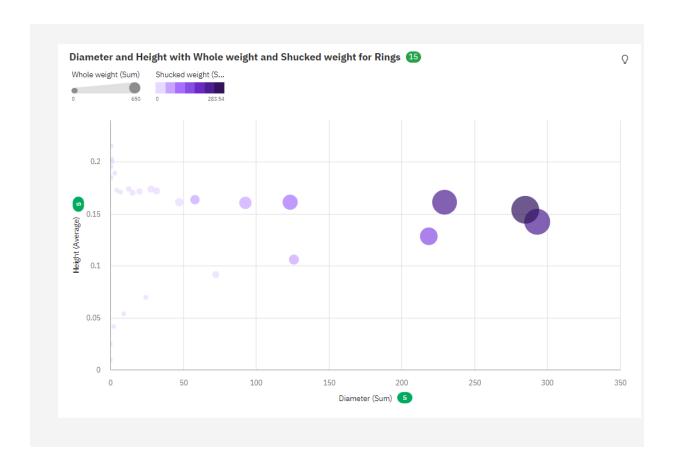
Univariate Analysis



Bi variate Analysis



Multi Variate Analysis



```
import pandas as pd
import numpy as np
import sklearn as sk
#loading the data
data = pd.read csv('/content/abalone.csv')
data.head()
 Sex Length Diameter Height Whole weight Shucked weight Viscera
weight \
      0.455
               0.365
O M
                     0.095
                                  0.5140
                                                0.2245
0.1010
      0.350 0.265
                      0.090
                                  0.2255
                                                0.0995
1 M
0.0485
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
4 I 0.330 0.255 0.080
                                 0.2050
                                                0.0895
0.0395
  Shell weight Rings
0
        0.150
                 15
1
        0.070
                 7
2
        0.210
                 9
3
        0.155
                 10
4
        0.055
                 7
data.tail()
    Sex Length Diameter Height Whole weight Shucked weight \
        0.565
                  0.450 0.165
4172
    F
                                    0.8870
                                                   0.3700
4173 M 0.590
                  0.440 0.135
                                     0.9660
                                                   0.4390
4174 M 0.600
                  0.475 0.205
                                     1.1760
                                                   0.5255
4175 F 0.625
                  0.485 0.150
                                     1.0945
                                                   0.5310
4176 M 0.710
                 0.555 0.195
                                    1.9485
                                                   0.9455
     Viscera weight Shell weight Rings
4172
            0.2390
                        0.2490
                                  11
```

0.2605

0.3080

0.2960

0.4950

10

10

12

data.shape

0.2145

0.2875

0.2610

0.3765

4173

4174

4175

4176

## data preprocessing

```
#missing values
data.isnull().sum()
Sex
                 0
Length
Diameter
Height
Whole weight
Shucked weight
Viscera weight
                 0
Shell weight
                 0
Rings
                 0
dtype: int64
# remove unwanted columns
data = data.drop(columns = ['Sex'],axis = 1)
data.head()
  Length Diameter Height Whole weight Shucked weight Viscera
weight \
0 0.455
             0.365 0.095
                                 0.5140
                                                 0.2245
0.1010
1 0.350
             0.265 0.090
                                 0.2255
                                                 0.0995
0.0485
2 0.530
             0.420 0.135
                                 0.6770
                                                 0.2565
0.1415
3 0.440
             0.365 0.125
                                 0.5160
                                                 0.2155
0.1140
4 0.330
             0.255 0.080
                                 0.2050
                                                 0.0895
0.0395
   Shell weight Rings
0
         0.150
                   15
1
         0.070
                   7
                   9
2
         0.210
3
         0.155
                   10
         0.055
                   7
```

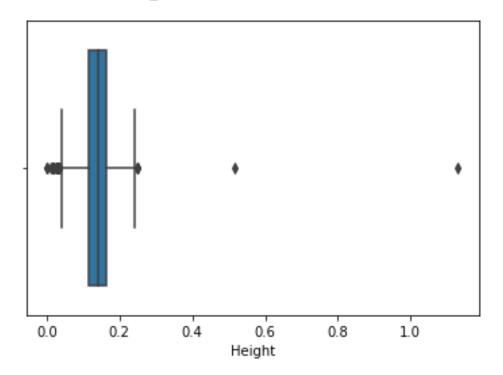
## # deal with outlier

import seaborn as sns
sns.boxplot(data.Height)

/usr/local/lib/python3.7/dist-packages/seaborn/\_decorators.py:43: FutureWarning: Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an

error or misinterpretation.
FutureWarning

<matplotlib.axes.\_subplots.AxesSubplot at 0x7f7f76a2a810>



# Encoding
pd.get\_dummies(data['Height'])

	0.000	0.010	0.015	0.020	0.025	0.030	0.035	0.040	0.045
0.050	\								
0	0	0	0	0	0	0	0	0	0
0									
1	0	0	0	0	0	0	0	0	0
0									
2	0	0	0	0	0	0	0	0	0
0									
3	0	0	0	0	0	0	0	0	0
0									
4	0	0	0	0	0	0	0	0	0
0									
• • •	• • •	• • •	• • •	• • •	• • •	• • •	• • •	• • •	• • •
	•		•		0		•	•	0
4172	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
4173	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
4174	0	U	U	U	0	0	U	U	U

0 4175		0	0	0	0	0	0	0	0	0
0 4176 0		0	0	0	0	0	0	0	0	0
0.515		0.210	0.215	0.220	0.225	0.230	0.235	0.240	0.250	
0 0	• • •	0	0	0	0	0	0	0	0	
1		0	0	0	0	0	0	0	0	
0 2		0	0	0	0	0	0	0	0	
0 3 0		0	0	0	0	0	0	0	0	
4		0	0	0	0	0	0	0	0	
•••										
4172		0	0	0	0	0	0	0	0	
0 4173		0	0	0	0	0	0	0	0	
0 4174		0	0	0	0	0	0	0	0	
0 4175		0	0	0	0	0	0	0	0	
0 4176 0		0	0	0	0	0	0	0	0	
0 1 2 3 4  4172 4173 4174 4175 4176	1.13	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0								
[4177 rows x 51 columns]										

# scaling
from sklearn.preprocessing import MinMaxScaler

scale = MinMaxScaler(feature\_range=(0,1))

```
y = data['Rings']
x = data.drop(columns=['Rings'],axis = 1)
names = x.columns
names
Index(['Length', 'Diameter', 'Height', 'Whole weight', 'Shucked
weight',
       'Viscera weight', 'Shell weight'],
      dtype='object')
x = scale.fit transform(x)
array([[0.51351351, 0.5210084 , 0.0840708 , ..., 0.15030262, 0.1323239
        0.147982061,
       [0.37162162, 0.35294118, 0.07964602, ..., 0.06624075,
0.06319947,
        0.068261091,
       [0.61486486, 0.61344538, 0.11946903, ..., 0.17182246,
0.18564845,
        0.2077728 ],
       [0.70945946, 0.70588235, 0.18141593, ..., 0.3527236,
0.37788018,
        0.30543099],
       [0.74324324, 0.72268908, 0.13274336, ..., 0.35642233,
0.34298881,
        0.29347285],
       [0.85810811, 0.84033613, 0.17256637, ..., 0.63517149,
0.49506254,
        0.4917787711)
#train and test
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)
from sklearn.linear model import LogisticRegression
model = LogisticRegression()
model.fit(x train, y train)
/usr/local/lib/python3.7/dist-packages/sklearn/linear model/
logistic.py:818: ConvergenceWarning: lbfgs failed to converge
(status=1):
STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
Increase the number of iterations (max iter) or scale the data as
```

```
shown in:
    https://scikit-learn.org/stable/modules/preprocessing.html
Please also refer to the documentation for alternative solver options:
https://scikit-learn.org/stable/modules/linear_model.html#logistic-
regression
 extra warning msg= LOGISTIC SOLVER CONVERGENCE MSG,
LogisticRegression()
# test with train data
pred = model.predict(x train)
pred
array([ 6, 8, 10, ..., 9, 8, 8])
from sklearn import metrics
metrics.accuracy score(pred, y train)
0.26967973660580663
pred = model.predict(x_test)
metrics.accuracy score(pred,y test)
0.25239234449760767
```