

ASSIGNMENT 3

| | |
|---------------------|--------------|
| Student Name | G.sarulatha |
| Student Roll Number | 211519104142 |
| Maximum Marks | 10 Marks |

Question 1&2:

```
import numpy as np
import matplotlib.pyplot as plt
from google.colab import files
import io
import pandas as pd

data = files.upload()

[ ] data = pd.read_csv('/content/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 |

Question 3:

Perform Below Visualizations.

- Univariate Analysis
- Bi-Variate Analysis
- Multivariate Analysis

Univariate Analysis

```
import matplotlib.pyplot as plt
import seaborn as sns

data.dtypes
plt.scatter(data.index, data['Height'])
plt.show()
```

Bi-Variate Analysis

```
import seaborn as sns

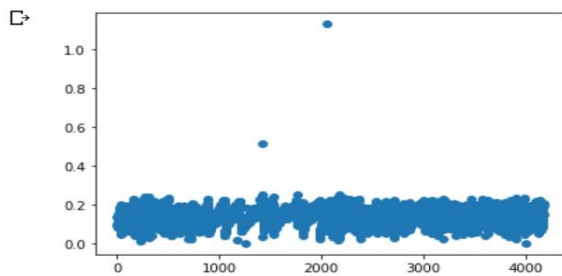
sns.boxplot(x='Sex', y='Height', data=data)
plt.show()
```

Multi-Variate Analysis

```
import seaborn as sns
sns.pairplot(data,hue="Rings",height=3)
plt.show()
```

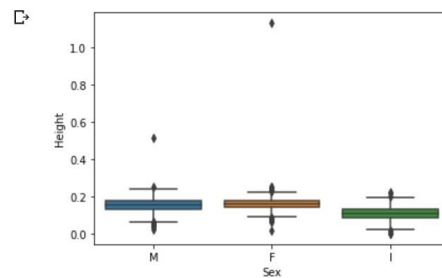
UNIVARIATE ANALYSIS

```
import matplotlib.pyplot as plt
import seaborn as sns
data.dtypes
plt.scatter(data.index,data['Height'])
plt.show()
```



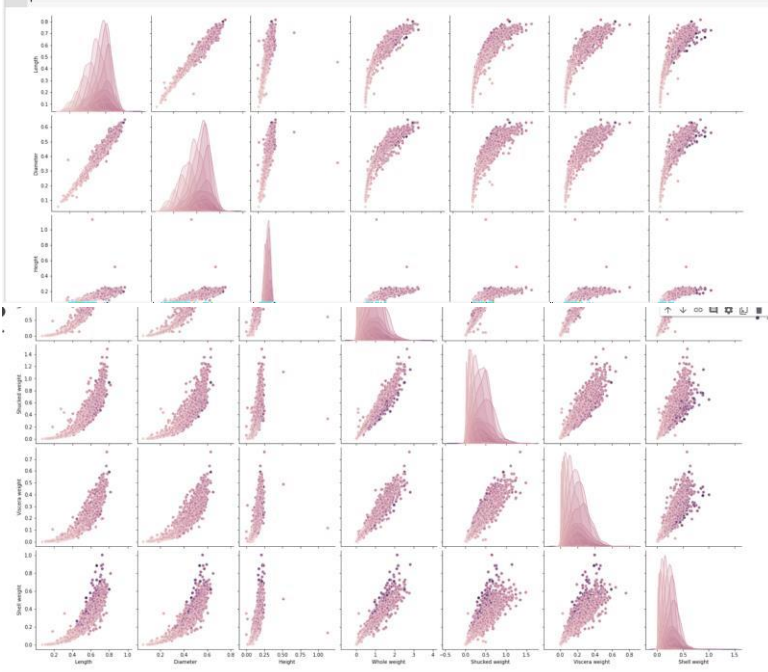
BIVARIATE ANALYSIS

```
import seaborn as sns
sns.boxplot(x='Sex',y='Height',data=data)
plt.show()
```



MULTIVARIATE ANALYSIS

```
import seaborn as sns
sns.pairplot(data,hue="Rings",height=3)
plt.show()
```



Question 4. Perform descriptive statistics on the dataset.

```
import pandas as pd
```

```
import numpy as np
df = pd.DataFrame(data)
print (df)
df.describe()
df.count()
```

```

0      M    0.455    0.365    0.095    0.5140    0.2245
1      M    0.350    0.265    0.090    0.2255    0.0995
2      F    0.530    0.420    0.135    0.6770    0.2565
3      M    0.440    0.365    0.125    0.5160    0.2155
4      I    0.330    0.255    0.080    0.2050    0.0895
...
4172   F    0.565    0.450    0.165    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
0                0.1010        0.1500    15
1                0.0485        0.0700     7
2                0.1415        0.2100     9
3                0.1140        0.1550    10
4                0.0395        0.0550     7
...
4172            0.2390        0.2490    11
4173            0.2145        0.2605    10
4174            0.2875        0.3080     9
4175            0.2610        0.2960    10
4176            0.3765        0.4950    12

```

[4177 rows x 9 columns]

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 |

df.count()

```

Sex      4177
Length   4177
Diameter  4177
Height   4177
Whole weight  4177
Shucked weight  4177
Viscera weight  4177
Shell weight  4177
Rings    4177
dtype: int64

```

Question 5. Check for Missing values and deal with them.

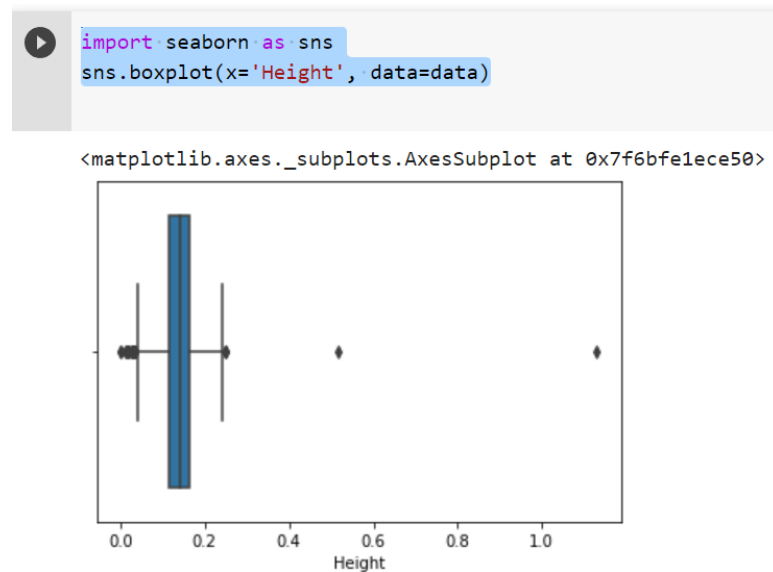
```
df.isnull().sum()
```

```
df.isnull().sum()
```

| | |
|----------------|---|
| Sex | 0 |
| Length | 0 |
| Diameter | 0 |
| Height | 0 |
| Whole weight | 0 |
| Shucked weight | 0 |
| Viscera weight | 0 |
| Shell weight | 0 |
| Rings | 0 |
| dtype: int64 | |

Question 6 Find the outliers and replace them outliers

```
import seaborn as sns
sns.boxplot(x='Height', data=data)
```



Question 7. Check for Categorical columns and perform encoding.

```
df = pd.DataFrame(data)
import pandas as pd
x=df.iloc[:,3:13].values
y=df.iloc[:,13:14].values
x.shape
```

CATEGORICAL COLUMNS

```
df = pd.DataFrame(data)
import pandas as pd
x=df.iloc[:,3:13].values
y=df.iloc[:,13:14].values
x.shape
```

➡ (4177, 6)

Question 8 Split the data into dependent and independent variables.

```
x=df.iloc[:,3:13].values
y=df.iloc[:,13:14].values
x.shape
```

DEPENDENT AND INDEPENDENT VARIABLES

```
x=df.iloc[:,3:13].values
y=df.iloc[:,13:14].values
x.shape
```

(4177, 6)

Question 9. Scale the independent variables

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

SCALE INDEPENDENT VARIABLES

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

```
[['M' 0.455 0.365 ... 0.2245 0.101 0.15]
 ['M' 0.35 0.265 ... 0.0995 0.0485 0.07]
 ['F' 0.53 0.42 ... 0.2565 0.1415 0.21]
 ...
 ['M' 0.6 0.475 ... 0.5255 0.2875 0.308]
 ['F' 0.625 0.485 ... 0.531 0.261 0.296]
 ['M' 0.71 0.555 ... 0.9455 0.3765 0.495]]
```

Question 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,random_state=0)
x_train.shape
x_test.shape
x_train.shape

```

```

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=0)
x_train.shape

```

```

(3341, 6)

```

```

[ ] x_test.shape

```

```

(836, 6)

```

```

[ ] x_train.shape

```

```

(3341, 6)

```

Question 11. Build the Model

```

my_dict=pd.read_csv("/content/abalone.csv")

```

```

df = pd.DataFrame(my_dict)

```

```

print(df)

```

```

0    Sex  Length  Diameter  Height  Whole weight  Shucked weight \
1      M   0.455    0.365    0.095    0.5140    0.2245
2      F   0.350    0.265    0.090    0.2255    0.0995
3      F   0.530    0.420    0.135    0.6770    0.2565
4      M   0.440    0.365    0.125    0.5160    0.2155
...    ...    ...    ...    ...    ...    ...
4172   F   0.565    0.450    0.165    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
0              0.1010    0.1500    15
1              0.0485    0.0700     7
2              0.1415    0.2100     9
3              0.1140    0.1550    10
4              0.0395    0.0550     7
...          ...    ...    ...
4172           0.2390    0.2490    11
4173           0.2145    0.2605    10
4174           0.2875    0.3080     9
4175           0.2610    0.2960    10
4176           0.3765    0.4950    12

```

```

[4177 rows x 9 columns]

```

```

import csv
with open ("/content/abalone.csv") as csv_file:
    csv_reader = csv.reader(csv_file)
    df = pd.DataFrame([csv_reader], index = None)
for val in list(df[1]):

```

```
print(val)
```

```
[31] import csv
with open ("/content/abalone.csv") as csv_file:
    csv_reader = csv.reader(csv_file)
    df = pd.DataFrame([csv_reader], index = None)
for val in list(df[1]):
    print(val)
```

```
['M', '0.455', '0.365', '0.095', '0.514', '0.2245', '0.101', '0.15', '15']
```

Question 12 & 13 Train and Test the Model

```
from sklearn.model_selection import train_test_split
```

```
train, test = train_test_split(df, test_size=0.2)
```

```
print(train)
```

```
print(test)
```

```

  Sex  Length  Diameter  Height  Whole weight  Shucked weight \
3614  M    0.615    0.475    0.155    1.0735    0.4375
493   M    0.655    0.530    0.175    1.2635    0.4860
2183  M    0.495    0.400    0.155    0.8085    0.2345
446   M    0.565    0.435    0.185    0.9815    0.3290
3689  F    0.630    0.505    0.195    1.3060    0.5160
...   ...    ...    ...    ...    ...    ...
2028  F    0.570    0.435    0.150    0.8295    0.3875
2719  I    0.360    0.260    0.080    0.1795    0.0740
2703  M    0.680    0.530    0.180    1.5290    0.7635
3632  I    0.300    0.225    0.075    0.1345    0.0570
184   F    0.645    0.510    0.200    1.5675    0.6210

  Viscera weight  Shell weight  Rings
3614    0.2585    0.3100    11
493     0.2635    0.4150    15
2183    0.1155    0.3500    6
446     0.1360    0.3900    13
3689    0.3305    0.3750    9
...     ...    ...    ...
2028    0.1560    0.2450    10
2719    0.0315    0.0600    5
2703    0.3115    0.4025    11
3632    0.0280    0.0440    5
184     0.3670    0.4600    12

[3341 rows x 9 columns]
  Sex  Length  Diameter  Height  Whole weight  Shucked weight \
1744  F    0.685    0.565    0.175    1.6380    0.7775
1104  I    0.510    0.405    0.125    0.6795    0.3465
1755  F    0.720    0.525    0.180    1.4450    0.6310
384   M    0.545    0.425    0.135    0.8445    0.3730
1696  M    0.630    0.490    0.170    1.1745    0.5255
...   ...    ...    ...    ...    ...
3211  F    0.600    0.480    0.165    1.1345    0.4535
418   F    0.630    0.500    0.155    1.0050    0.3670
```

```
[3341 rows x 9 columns]
```

| | Sex | Length | Diameter | Height | Whole weight | Shucked weight | \ |
|------|-----|--------|----------|--------|--------------|----------------|---|
| 1744 | F | 0.685 | 0.565 | 0.175 | 1.6380 | 0.7775 | |
| 1104 | I | 0.510 | 0.405 | 0.125 | 0.6795 | 0.3465 | |
| 1755 | F | 0.720 | 0.525 | 0.180 | 1.4450 | 0.6310 | |
| 384 | M | 0.545 | 0.425 | 0.135 | 0.8445 | 0.3730 | |
| 1696 | M | 0.630 | 0.490 | 0.170 | 1.1745 | 0.5255 | |
| ... | .. | ... | ... | ... | ... | ... | |
| 3211 | F | 0.600 | 0.480 | 0.165 | 1.1345 | 0.4535 | |
| 418 | F | 0.630 | 0.500 | 0.155 | 1.0050 | 0.3670 | |
| 2880 | I | 0.475 | 0.375 | 0.110 | 0.4940 | 0.2110 | |
| 3239 | F | 0.690 | 0.540 | 0.185 | 1.5715 | 0.6935 | |
| 264 | M | 0.270 | 0.200 | 0.080 | 0.1205 | 0.0465 | |

| | Viscera weight | Shell weight | Rings |
|------|----------------|--------------|-------|
| 1744 | 0.3750 | 0.4380 | 11 |
| 1104 | 0.1395 | 0.1820 | 8 |
| 1755 | 0.3215 | 0.4350 | 7 |
| 384 | 0.2100 | 0.2350 | 10 |
| 1696 | 0.2730 | 0.3390 | 11 |
| ... | ... | ... | ... |
| 3211 | 0.2700 | 0.3350 | 10 |
| 418 | 0.1990 | 0.3600 | 16 |
| 2880 | 0.1090 | 0.1545 | 8 |
| 3239 | 0.3180 | 0.4700 | 15 |
| 264 | 0.0280 | 0.0400 | 6 |

Question 14. Measure the performance using Metrics.

```
pd.crosstab(Y_test,y_predict)

print(classification_report(Y_test,y_predict))
```