

8.05432914, 10.15371887,
13.47423856, 9.0
6825076, 6.69843582, 1
3.38213142, 9.62823486,
8.20174551, 7.7
9183041, 9.3338472, 1
1.08195328, 11.25321895,
6.11231204, 10.6
960639, 9.23348159,
7.76425036, 11.65342323,
12.6024271, 7.4
9694081, 9.71678931,
7.41119139, 6.94925679,
6.34706174, 9.9
9734923, 6.70117631, 1
0.71374432, 9.59457302,
7.07847213, 6.6
940933, 9.30356123, 1
3.66698224, 9.71369221,
17.36952958, 7.8
1225327, 8.86909973,
9.29540502, 11.03405521,
12.90720962, 13.0
3952065, 4.90843127,
9.50619996, 10.09434256,
8.67296752, 9.0
3746047, 8.33310609, 1
0.60445018, 9.66636969,
7.67351279, 8.7
4447193, 12.37470593,
7.70552082, 11.35599144,
11.25726129, 10.0
2276461, 8.01953433, 1
1.39538114, 7.92288557,
11.02588274, 7.0
2530311, 10.80014326, 1
3.22266766, 11.41469264,
7.5577235, 6.8
3654146, 6.97820486, 1
0.29150052, 9.1851768,
9.72122817, 9.2
9569276, 11.98122676,
9.87982582, 8.55374278,

```
In [ ]: from sklearn.preprocessing
x = scale(x)
x
```

```
Out[ ]: array([[ -0.0105225 , -0.67088921, -0.50179694,
..., -0.61037964,
-0.7328165 , -0.64358742],
[ -0.0105225 , -1.61376082, -1.57304487,
..., -1.22513334,
-1.24343929, -1.25742181],
[ -1.26630752,  0.00259051,  0.08738942,
..., -0.45300269,
-0.33890749, -0.18321163],
...,
[ -0.0105225 ,  0.63117159,  0.67657577,
...,  0.86994729,
1.08111018,  0.56873549],
[ -1.26630752,  0.85566483,  0.78370057,
...,  0.89699645,
0.82336724,  0.47666033],
[ -0.0105225 ,  1.61894185,  1.53357412,
...,  0.00683308,
1.94673739,  2.00357336]])
```

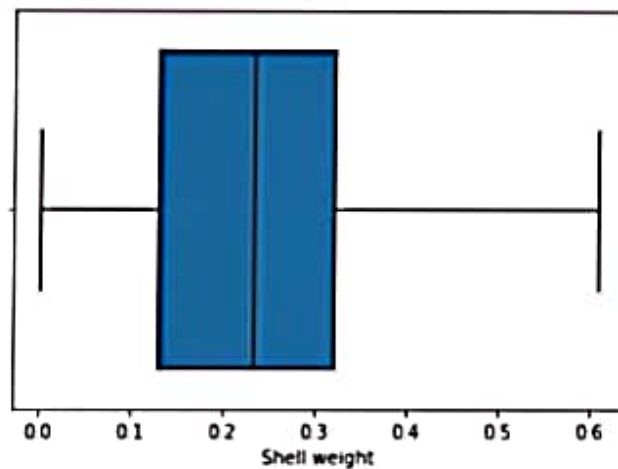
9.Split the data into training and testing

```
In [ ]: from sklearn.model_selection
```

In []:

```
data['Shell weight']=np.  
sns.boxplot(data['Shell
```

Out[]:



6.Check for Categorical columns and perform encoding.

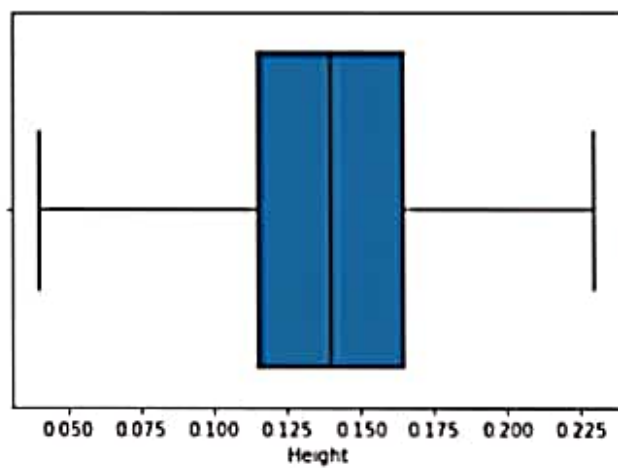
In []:

```
data['Sex'].replace({'M'  
data
```

Out[]:

	Sex	Length	Diameter	H
0	1	0.455	0.365	1
1	1	0.350	0.265	1
2	0	0.530	0.420	1
3	1	0.440	0.365	1
4	2	0.330	0.255	1
...
4172	0	0.565	0.450	1
4173	1	0.590	0.440	1
4174	1	0.600	0.475	1

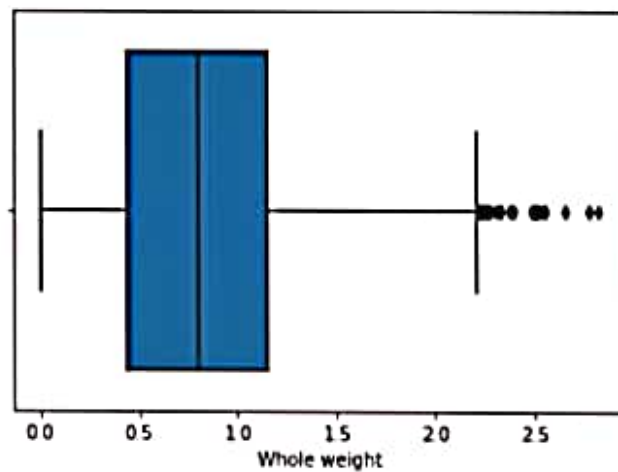
Out[]:



In []:

```
sns.boxplot(data['Whole
```

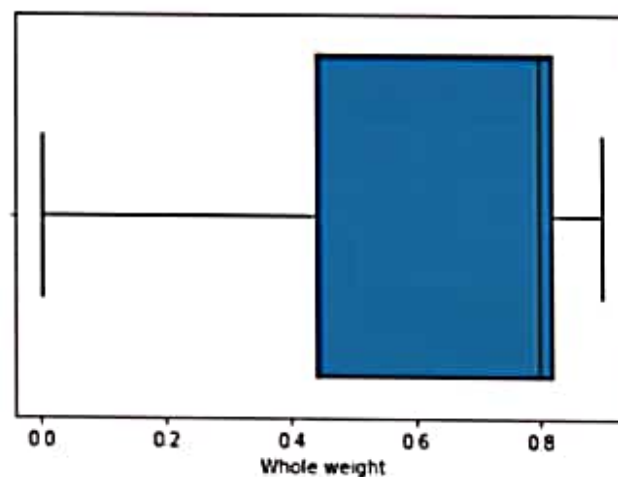
Out[]:



In []:

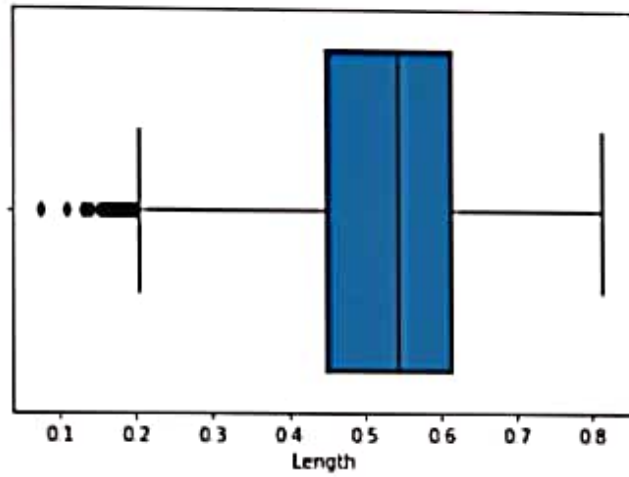
```
data['Whole weight']=np.  
sns.boxplot(data['Whole
```

Out[]:



In []:

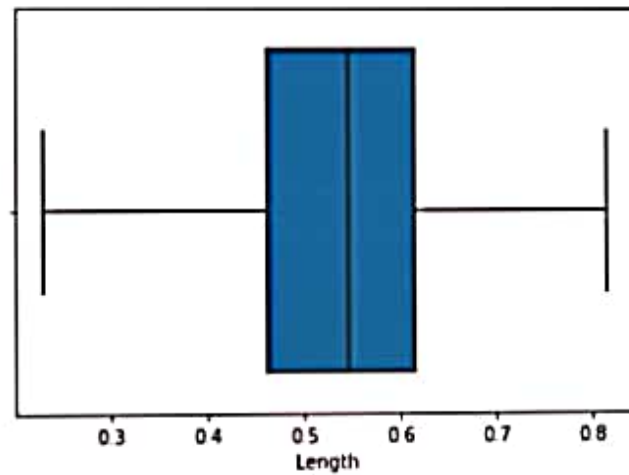
Out[]:



In []:

```
data['Length']=np.where(  
sns.boxplot(data['Length'
```

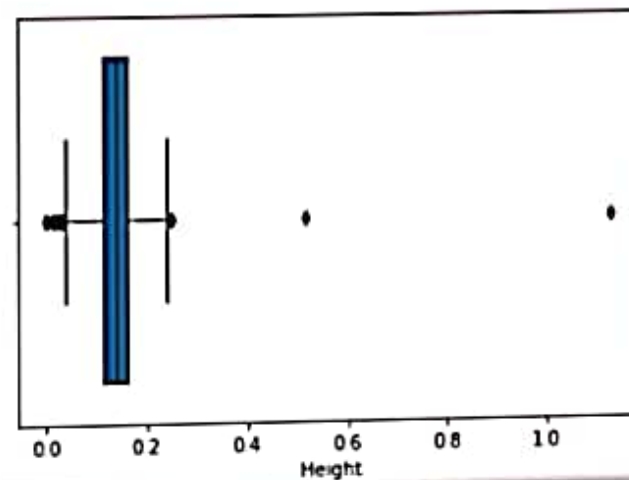
Out[]:



In []:

```
sns.boxplot(data['Height'
```

Out[]:



Out[]:

	Length	Diameter	Height
0.25	0.450	0.35	0.115
0.75	0.615	0.48	0.165

In []:

```
iqr=quant.loc[0.75]-quant.loc[0.25]
iqr
```

Out[]:

```
Length      0.1650
Diameter    0.1300
Height      0.0500
Whole weight 0.7115
Shucked weight 0.3160
Viscera weight 0.1595
Shell weight 0.1990
Rings       3.0000
dtype: float64
```

In []:

```
low=quant.loc[0.25]-(1.5*iqr)
low
```

Out[]:

```
Length      0.2025
0
Diameter    0.1550
0
Height      0.0400
0
Whole weight -0.6257
5
Shucked weight -0.2880
0
Viscera weight -0.1457
5
Shell weight  -0.1685
0
Rings       3.5000
0
dtype: float64
```

In []:

```
data.isna().sum()
```

Out[]:

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

In []:

```
data.isna().any().sum()
```

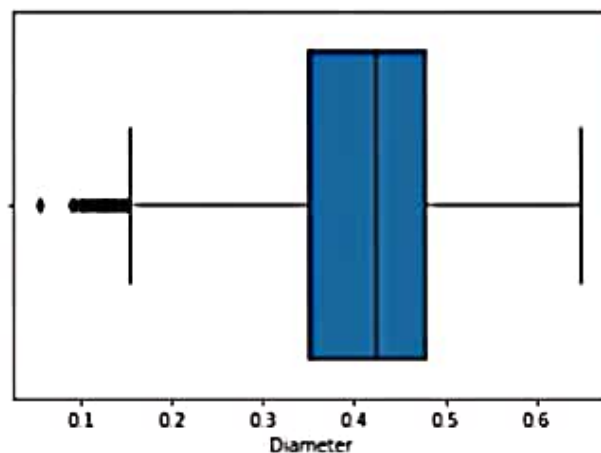
Out[]: 0

5. Find the outliers and replace them outliers

In []:

```
sns.boxplot(data['Diameter'])
```

Out[]:



In []:

```
quant=data.quantile(q=[0.01, 0.99])
data.replace(data[quant[0]:quant[1]], quant[0])
```


In []:

```
data.isna()
```

Out[]:

	Sex	Length	Diameter
0	False	False	False
1	False	False	False
2	False	False	False
3	False	False	False
4	False	False	False
...
4172	False	False	False
4173	False	False	False
4174	False	False	False
4175	False	False	False
4176	False	False	False

4177 rows x 9 columns

In []:

```
data.isna().any()
```

Out[]:

Sex	False
Length	False
Diameter	False
Height	False
Whole weight	False
Shucked weight	False
Viscera weight	False
Shell weight	False
Rings	False
dtype:	bool

In []:

```
data.var()
```

Out[]:

Length	0.014
422	
Diameter	0.009
849	
Height	0.001
750	
Whole weight	0.240
481	
Shucked weight	0.049
268	
Viscera weight	0.012
015	
Shell weight	0.019
377	
Rings	10.395
266	
dtype:	float64

In []:

```
data.nunique()
```

Out[]:

Sex	3
Length	134
Diameter	111
Height	51
Whole weight	2429
Shucked weight	1515
Viscera weight	880
Shell weight	926
Rings	28
dtype:	int64

4.Check for missing values and deal with them

In []:

```
data.isna()
```

In []:

```
data.describe()
```

Out[]:

	Length	Diameter
count	4177.000000	4177.000000
mean	0.523992	0.40788
std	0.120093	0.09924
min	0.075000	0.05500
25%	0.450000	0.35000
50%	0.545000	0.42500
75%	0.615000	0.48000
max	0.815000	0.65000

In []:

```
data.mode().T
```

Out[]:

	0	1
Sex	M	NaN
Length	0.55	0.625
Diameter	0.45	NaN
Height	0.15	NaN
Whole weight	0.2225	NaN
Shucked weight	0.175	NaN
Viscera weight	0.1715	NaN
Shell weight	0.275	NaN
Rings	9.0	NaN

In []:

```
data.head()
```

Out[]:

	Sex	Length	Diameter	Heigl
0	M	0.455	0.365	0.09
1	M	0.350	0.265	0.09
2	F	0.530	0.420	0.13
3	M	0.440	0.365	0.12
4	I	0.330	0.255	0.08

In []:

```
data.tail()
```

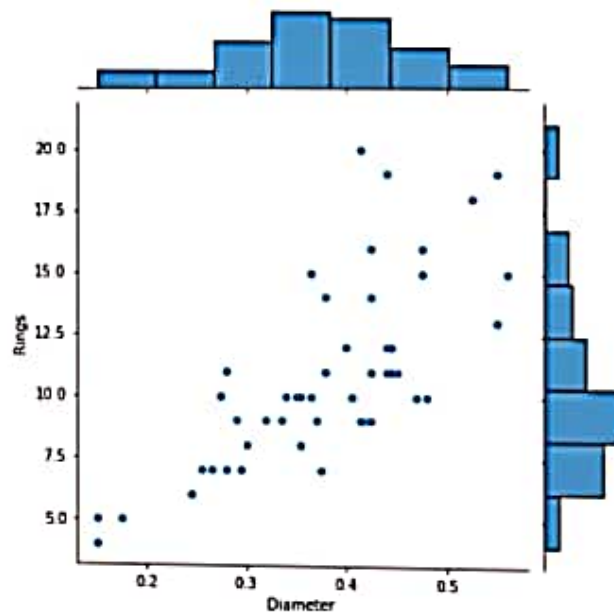
Out[]:

	Sex	Length	Diameter	H
4172	F	0.565	0.450	
4173	M	0.590	0.440	
4174	M	0.600	0.475	
4175	F	0.625	0.485	
4176	M	0.710	0.555	

In []:

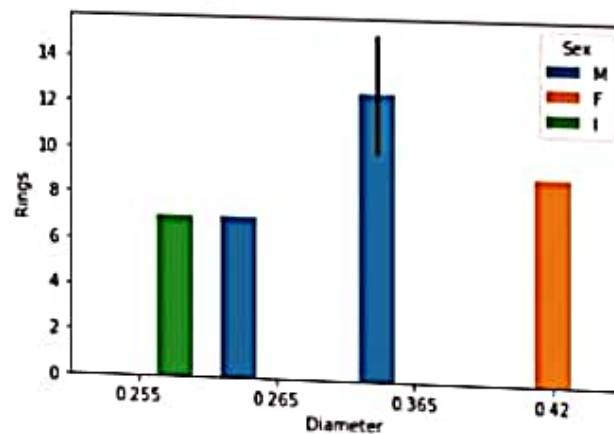
```
data.info()
```

```
RangeIndex: 4177 entries, 0 to 4176
Data columns (total 9 columns):
#   Column              Non
-Null Count  Dtype  ---
-----
```



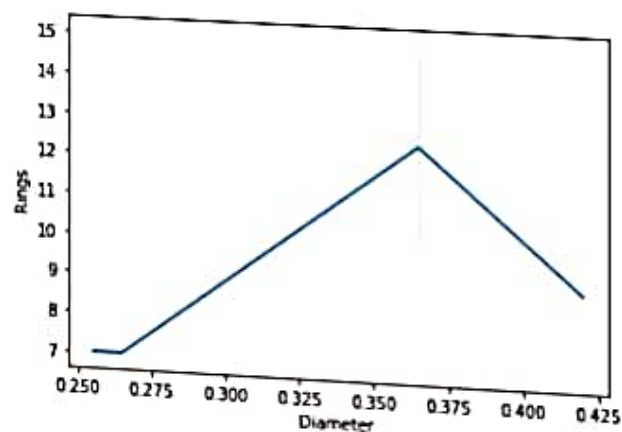
```
In [ ]: sns.barplot('Diameter', 'Rings')
```

```
Out[ ]:
```



```
In [ ]: sns.lineplot(data['Diameter'], 'Rings')
```

```
Out[ ]:
```

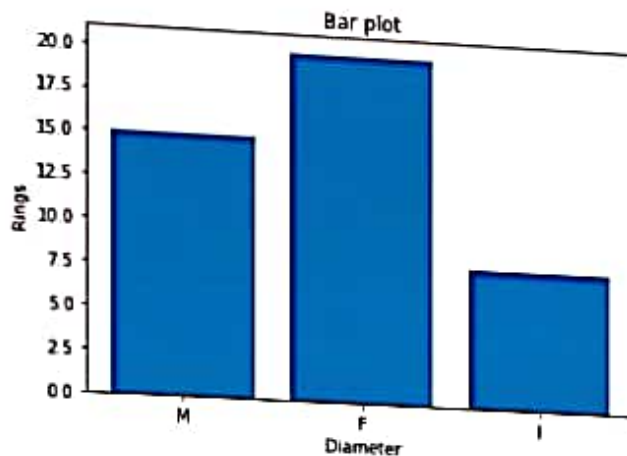


```
In [ ]:
```

In []:

```
plt.bar(data['Sex'].head(3))
plt.title('Bar plot')
plt.xlabel('Diameter')
plt.ylabel('Rings')
```

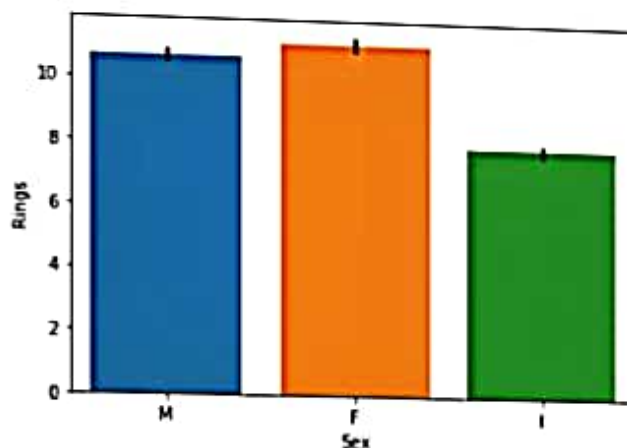
Out[]: Text(0, 0.5, 'Rings')



In []:

```
sns.barplot(data['Sex'],
```

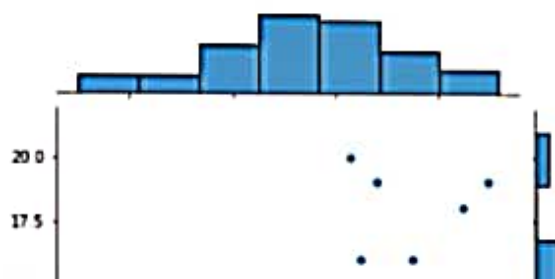
Out[]:



In []:

```
sns.jointplot(data['Dian
```

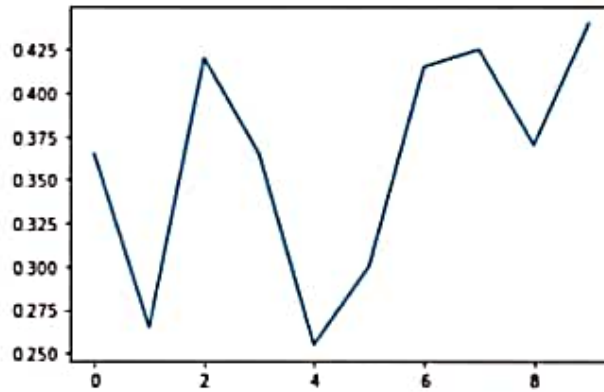
Out[]:



In []:

```
plt.plot(data['Diameter']
```

Out[]: []



In []:

```
plt.pie(data['Diameter']
```

Out[]: ([,

,

,

,

],

```
[Text(0.850721562611055
7, 0.6973326486753676,
''),
```

```
Text(-0.32611344931648
134, 1.0505474849691026,
''),
```

```
Text(-1.09980536640789
08, -0.0206919312874714
4, ''),
```

```
Text(-0.08269436219656
089, -1.096887251480709,
''),
```

```
Text(0.975844636228721
8, -0.5076684409569241,
'')],
```

```
[Text(0.464029943242393
94, 0.3803632629138369,
'21.856'),
```

```
Text(-0.17788006326353
```

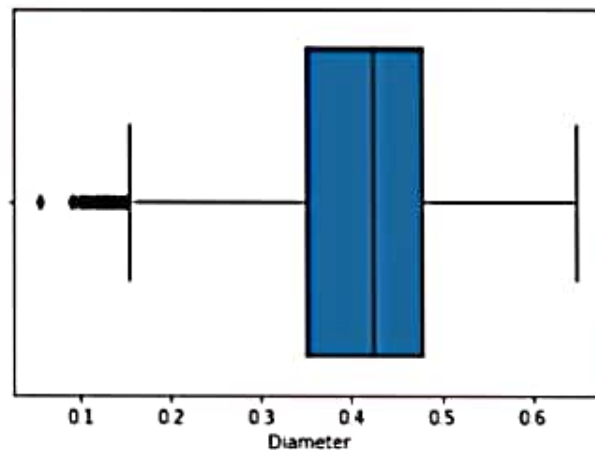
Out[]:

	Sex	Length	Diameter	Heigl
0	M	0.455	0.365	0.09
1	M	0.350	0.265	0.09
2	F	0.530	0.420	0.13
3	M	0.440	0.365	0.12
4	I	0.330	0.255	0.08

In []:

```
sns.boxplot(data['Diameter'
```

Out[]:



In []:

```
plt.hist(data['Diameter'
```

```
Out[ ]: (array([ 13.,  66.,  1
80., 344., 513., 81
2., 1017., 934., 275.,
          23.]),
        array([0.055 , 0.1145,
0.174 , 0.2335, 0.293 ,
0.3525, 0.412 , 0.4715,
          0.531 , 0.5905,
0.65 ]),
        )
```



1.Loading Dataset into tool

```
In [ ]: from google.colab import  
        uploaded = files.upload()
```

Upload widget is only available when the cell has been executed in the current browser session. Please rerun this cell to enable.
Saving abalone.csv to abalone.csv

```
In [ ]: import pandas as pd  
        import numpy as np  
        import matplotlib.pyplot  
        import seaborn as sns  
        import warnings  
        warnings.filterwarnings(
```

```
In [ ]: data = pd.read_csv("aba]
```

2.Performing Visualization

Univariate Analysis

```
In [ ]: data.head()
```

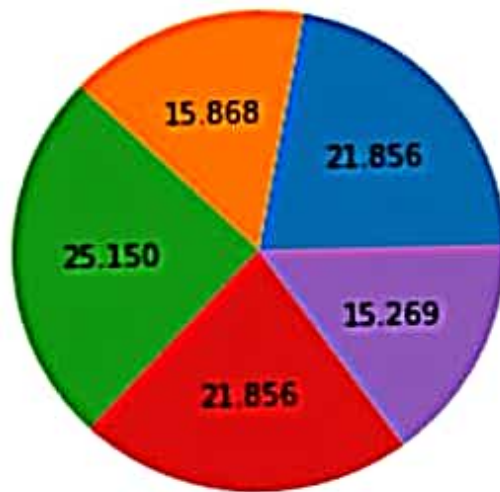
```
Out[ ]:
```

	Sex	Length	Diameter	Heigl
--	-----	--------	----------	-------

0	M	0.455	0.365	0.09
---	---	-------	-------	------

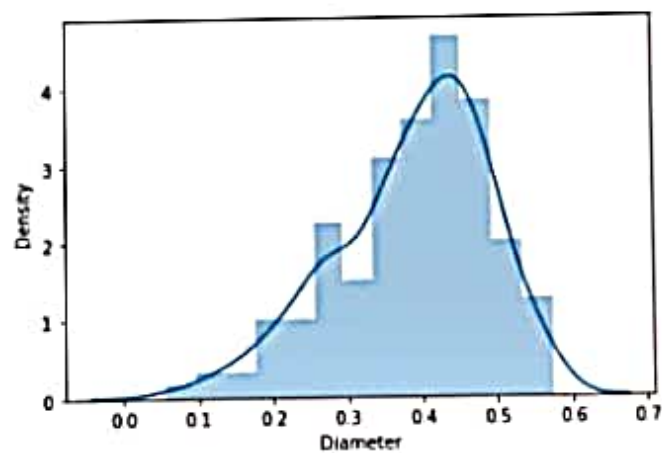
1	M	0.350	0.265	0.09
---	---	-------	-------	------

2	F	0.533	0.433	0.12
---	---	-------	-------	------



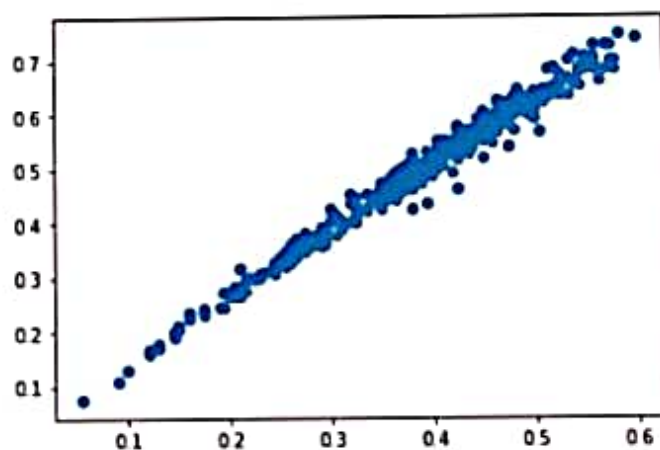
In []: `sns.distplot(data['Diamet`

Out[]:



In []: `plt.scatter(data['Diamet`

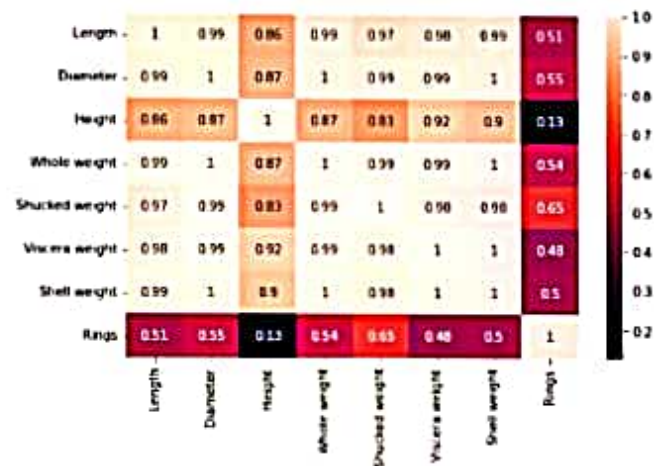
Out[]:



In []:

```
fig=plt.figure(figsize=(  
sns.heatmap(data.head()).
```

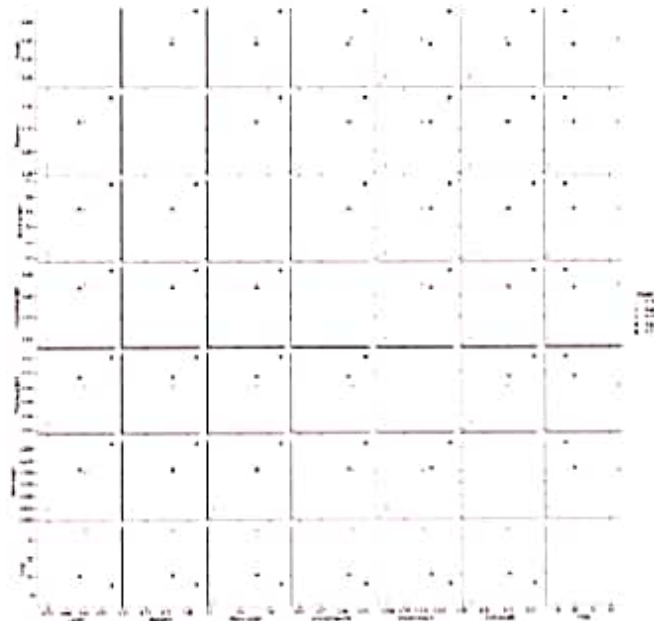
Out[]:



In []:

```
sns.pairplot(data.head()
```

Out[]:



In []:

```
sns.pairplot(data.head()
```

Out[]:



```
In [ ]: data.shape
```

```
Out[ ]: (4177, 9)
```

```
In [ ]: data.kurt()
```

```
Out[ ]: Length      0.064
        621
        Diameter    -0.045
        476
        Height      76.025
        509
        Whole weight -0.023
        644
        Shucked weight 0.595
        124
        Viscera weight 0.084
        012
        Shell weight  0.531
        926
        Rings        2.330
        687
        dtype: float64
```

```
In [ ]: data.skew()
```

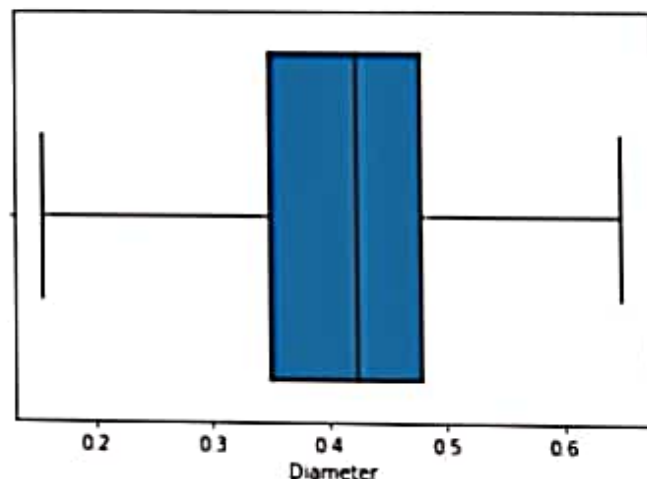
```
Out[ ]: Length      -0.6398
        73
        Diameter    -0.6091
        98
        Height      3.1288
        17
        Whole weight 0.5309
        59
        Shucked weight 0.7190
        98
        Viscera weight 0.5918
        52
```

```
In [ ]: up=quant.loc[0.75]+(1.5*  
up
```

```
Out[ ]: Length      0.862  
50  
Diameter      0.675  
00  
Height      0.240  
00  
Whole weight  2.220  
25  
Shucked weight 0.976  
00  
Viscera weight 0.492  
25  
Shell weight  0.627  
50  
Rings      15.500  
00  
dtype: float64
```

```
In [ ]: data['Diameter']=np.where  
sns.boxplot(data['Diameter
```

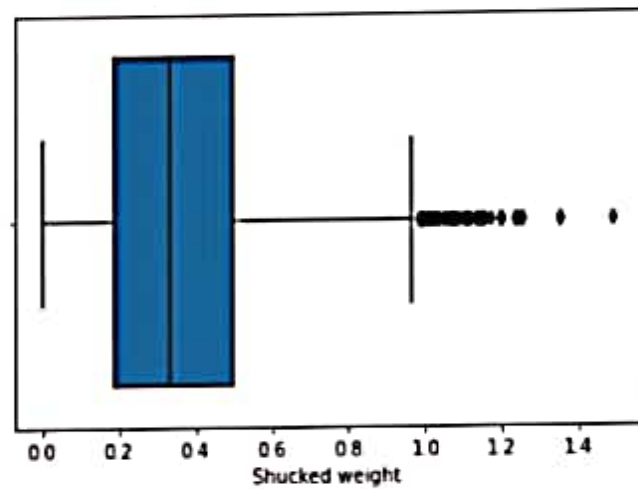
```
Out[ ]:
```



```
In [ ]: sns.boxplot(data['Length
```

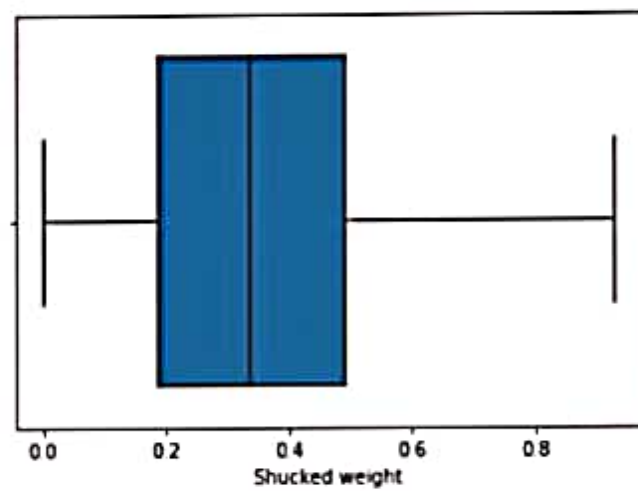
```
In [ ]: sns.boxplot(data['Shucked weight'])
```

Out[]:



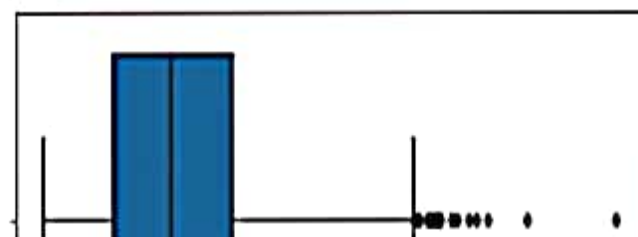
```
In [ ]: data['Shucked weight'] = r  
sns.boxplot(data['Shucked weight'])
```

Out[]:

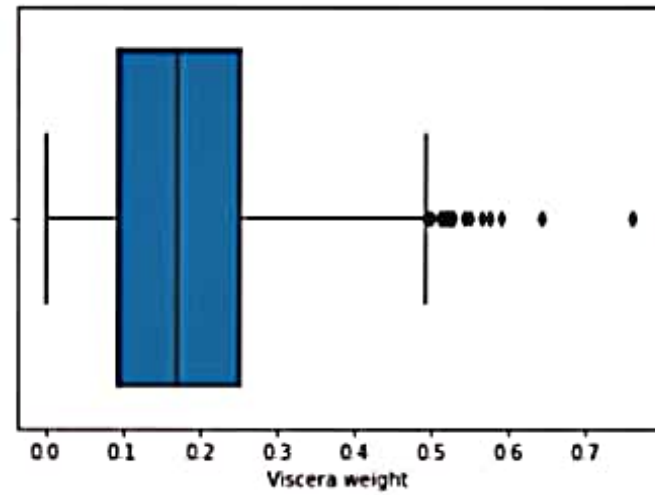


```
In [ ]: sns.boxplot(data['Viscera weight'])
```

Out[]:



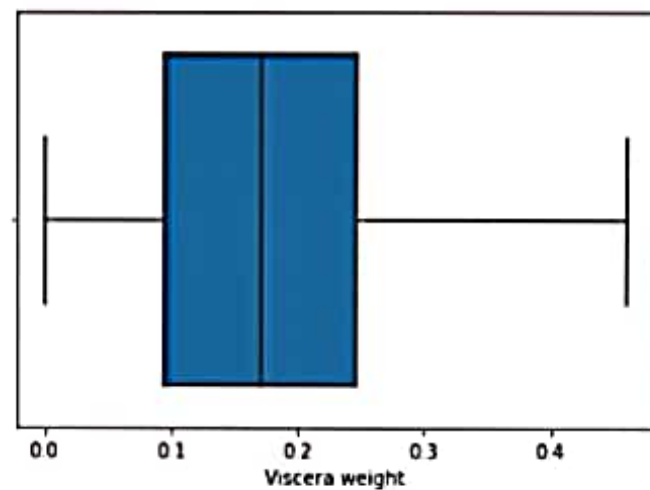
Out[]:



In []:

```
data['Viscera weight'] =  
sns.boxplot(data['Viscer
```

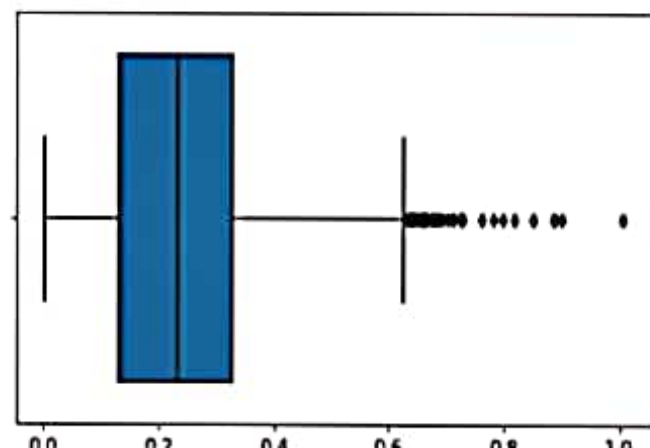
Out[]:



In []:

```
sns.boxplot(data['Shell
```

Out[]:




```
In [ ]: x=data.drop(columns= ['F
y=data['Rings']
x
```

```
Out[ ]:
```

	Sex	Length	Diameter	H
0	1	0.455	0.365	1
1	1	0.350	0.265	1
2	0	0.530	0.420	1
3	1	0.440	0.365	1
4	2	0.330	0.255	1
...
4172	0	0.565	0.450	1
4173	1	0.590	0.440	1
4174	1	0.600	0.475	1
4175	0	0.625	0.485	1
4176	1	0.710	0.555	1

4177 rows x 8 columns

```
In [ ]: y
```

```
Out[ ]:
```

0	15
1	7
2	9
3	10
4	7
...	...
4172	11
4173	10
4174	9
4175	10

10.Build the Model

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

11.Train the model

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

```
Out[ ]: LinearRegression()
```

12.Test the model

```
In [ ]: y_pred=MLR.predict(x_test)  
        y_pred
```

```
Out[ ]: array([ 6.3204331 , 10.4  
1671748, 13.91911179, 1  
2.29316277, 8.7273177 ,  
11.04369928, 12.4  
0210281, 11.6992544 , 1  
2.01785949, 6.57983392,  
11.91353764, 10.7  
9661591, 11.56560952, 1  
0.14326497, 13.16762604,  
9.34621768, 10.7  
6904478, 11.88283609,  
9.34461447, 10.08802992,  
12.80140942, 9.5  
8177975, 11.20908126, 1  
0.3662699 , 10.0168299 ,  
15.92815446, 15.9  
3700213, 7.36066362, 1  
3.2889134 , 10.1579858 ,  
11.62833855, 11.0  
2507007, 11.60253151, 1
```

8.27235155, 12.0413044, 1
10.43536813, 11.1
2820999, 10.56478101, 1
2.12900686, 9.0459273, 8.6
6.50569617, 8.6
5471113, 11.17391657,
4.17641665, 6.45933408,
10.94174559, 10.5
6404265, 7.32806471, 1
0.90718067, 8.76983179,
9.54866214, 9.7
1969088, 9.19215908, 1
1.19107958, 9.95023994,
10.33050587, 11.9
8860703, 5.76011208,
8.82560871, 8.26963359,
6.41006108, 7.6
2776781, 7.77958091, 1
0.53587014, 8.89399096,
11.50322847, 6.4
6552063, 6.62035734, 1
1.27313616, 8.28747988,
12.05544015, 11.6
973709, 12.73972343, 1
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