NAME : VAISHNAVI.R REG NO :512219205010 PROJECT : RETAIL STORE STOCK INVENTORY ANALYTICS ASSIGNMENT: 04 1. DOWNLOADING DATASET

Dataset Link : [Abalone]

2. LOADING THE DATASET

In [ ]:

```
import numpy as np import pandas as pd
import matplotlib.pyplot as plt
%matplotlib inline import
seaborn as sns
```

In [ ]:

```
df = pd.read_csv('/content/drive/MyDrive/abalone.csv')
```

In [ ]:

```
df.head()
```

Out[ ]:

	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

3.PERFORMING VISUALIZATION'S

(i)Univariate Analysis

(ii)Bi-Variate Analysis

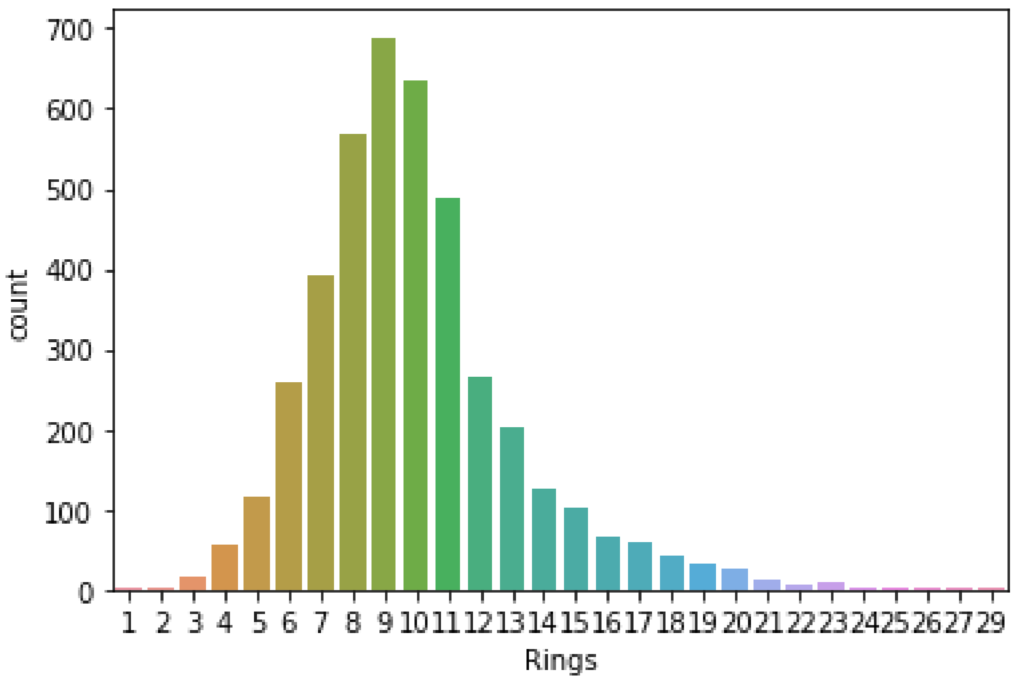
(iii)Multi-Variate Analysis

(i) UNIVARIATE

In [ ]:

```
# countplot
sns.countplot(data=df,x="Rings")
```

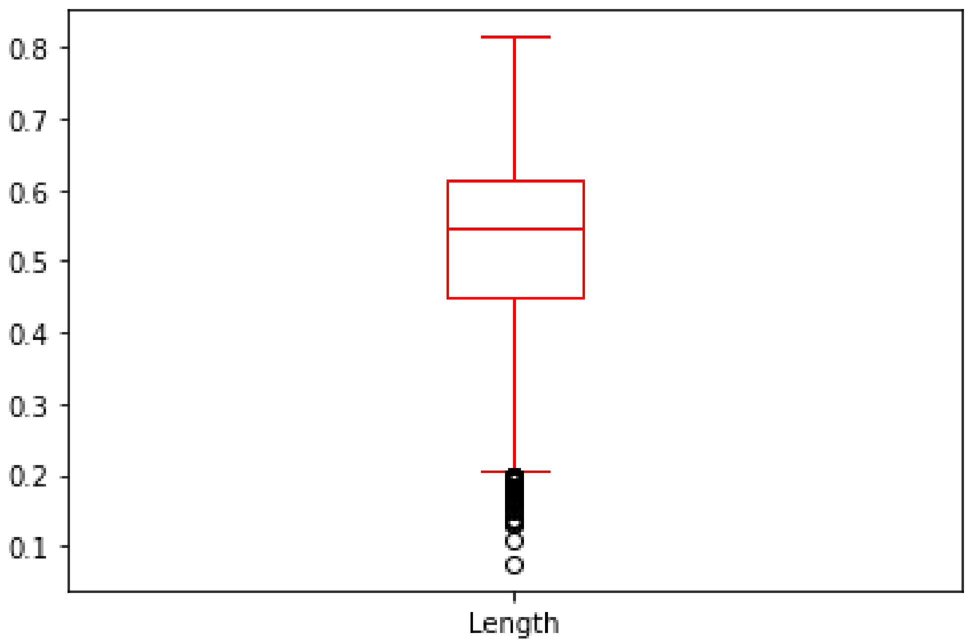
Out[ ]:



In [ ]:

```
#boxplot
df.boxplot(column=['Length'], grid=False, color='Red')
```

Out[ ]:

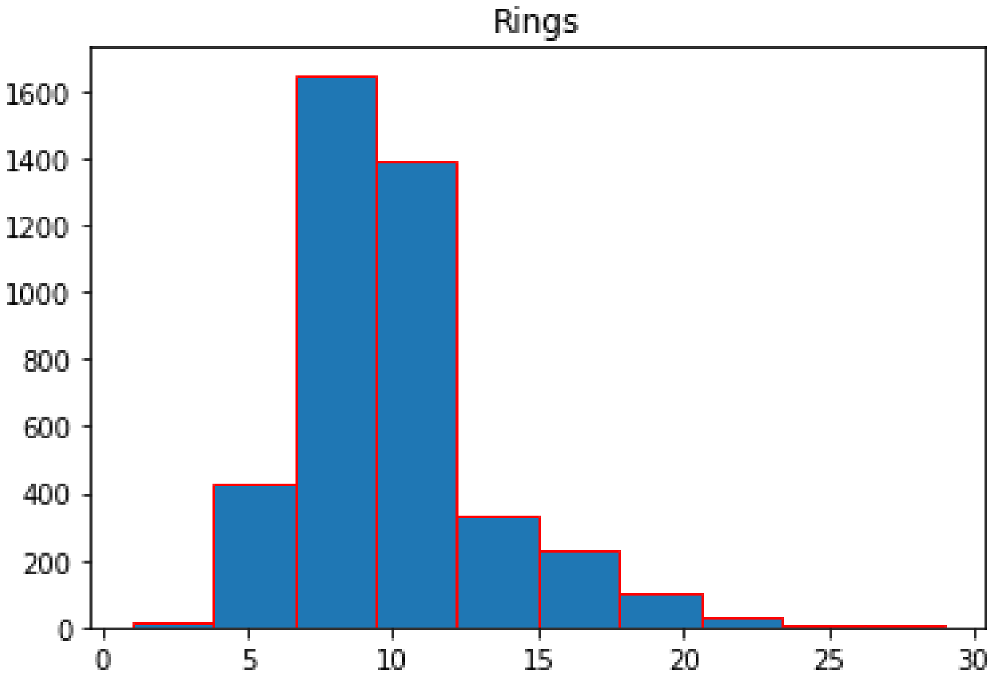


In [ ]:

```
#histogram
df.hist(column='Rings', grid=False, edgecolor='Red')
```

Out[ ]:

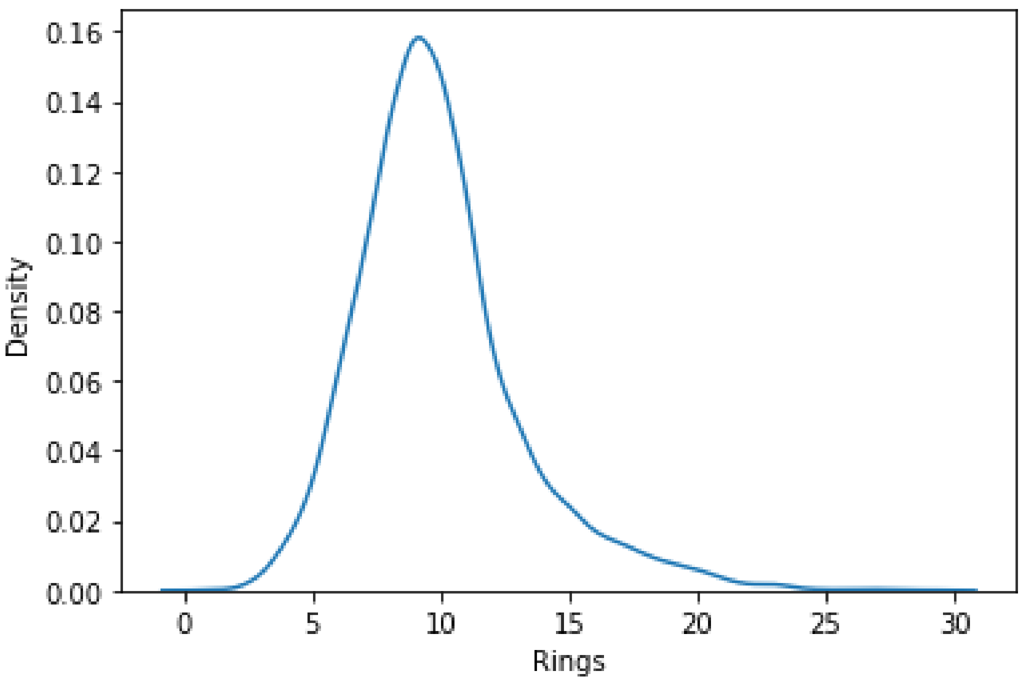
```
array([[]],       dtype=object)
```



In [ ]:

```
#kdeplot
sns.kdeplot(df['Rings'])
```

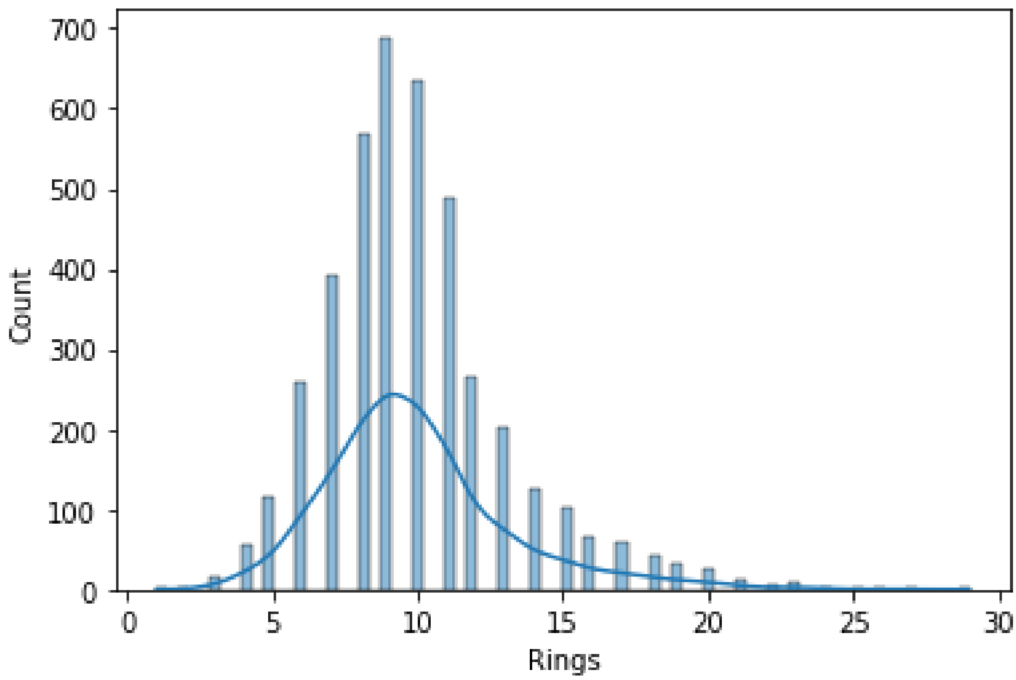
Out[ ]:



In [ ]:

```
#histplot
sns.histplot(df.Rings,kde=True)
```

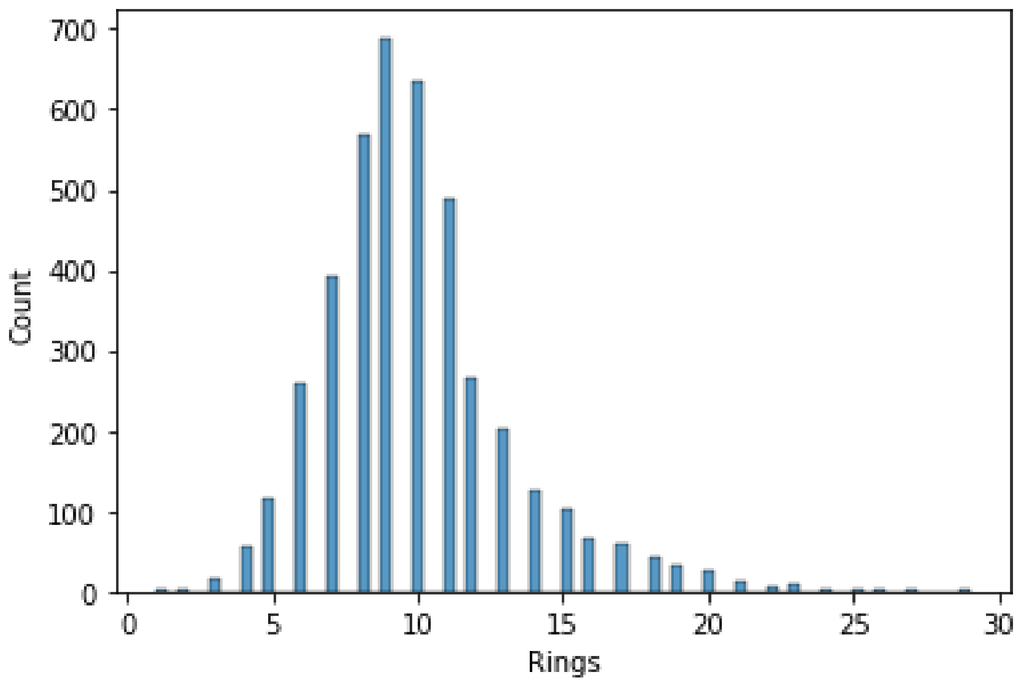
Out[ ]:



In [ ]:

```
#histplot
sns.histplot(df['Rings'])
```

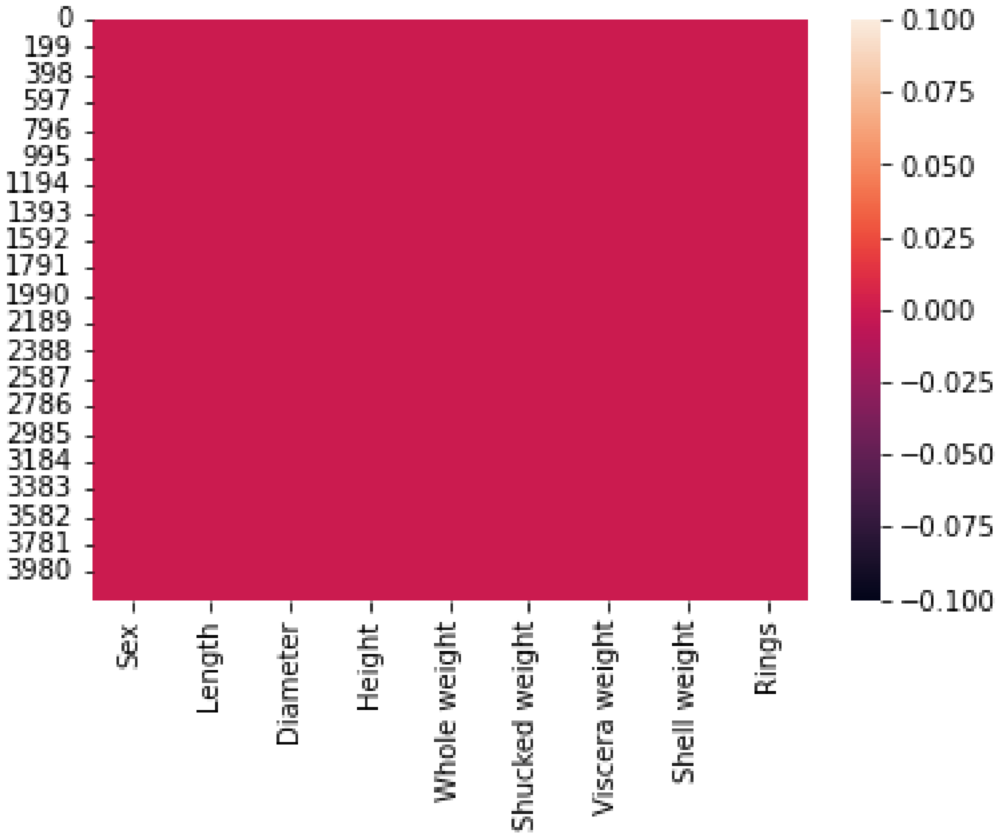
Out[ ]:



In [ ]:

```
#heatmap
sns.heatmap(df.isnull())
```

Out[ ]:

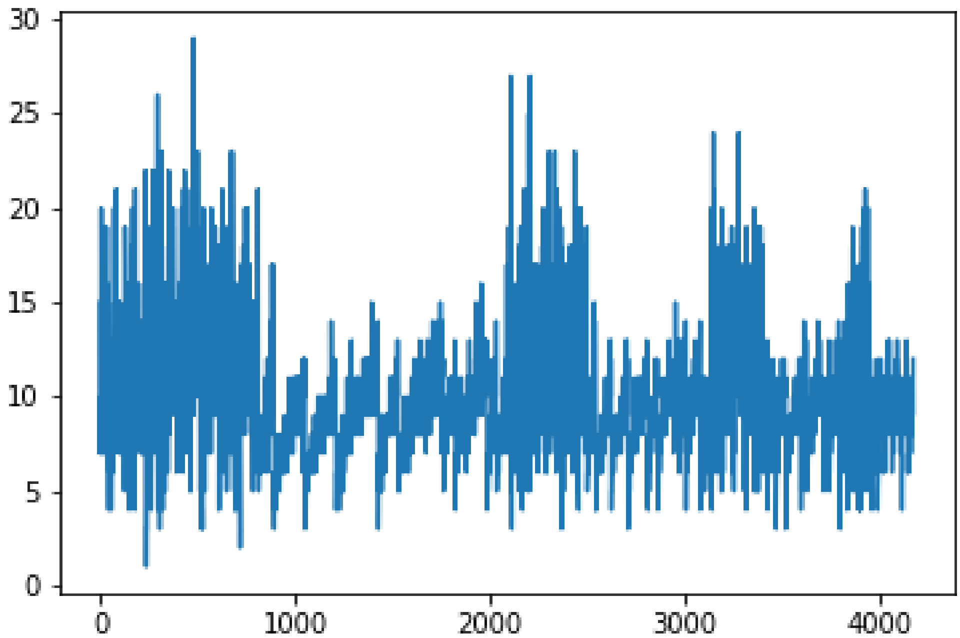


In [ ]:

```
#line plot plt.plot(df['Rings'])
```

Out[ ]:

```
[]
```

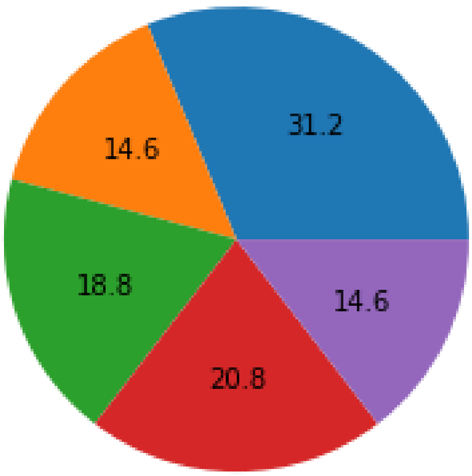


In [ ]:

```
#piechart plt.pie(df['Rings'].head(),autopct='%.1f')
```

Out[ ]:

```
([,
  ,
  ,
  ,
  ],
 [Text(0.6111272563215626, 0.9146165735327998, ''),
  Text(-0.8270237769092663, 0.725280409515335, ''),
  Text(-1.041623153479572, -0.35358337932554523, ''),
  Text(-5.149471704824549e-08,     -1.0999999999999988,     ''),
Text(0.9865599777267362, -0.4865176362145796, '')],
 [Text(0.33334213981176136, 0.4988817673815271, '31.2'),
  Text(-0.4511038783141452, 0.39560749609927365, '14.6'),
  Text(-0.5681580837161301, -0.1928636614502974, '18.8'),
  Text(-2.8088027480861175e-08,   -0.5999999999999993,   '20.8'),
Text(0.5381236242145833, -0.2653732561170434, '14.6')])
```

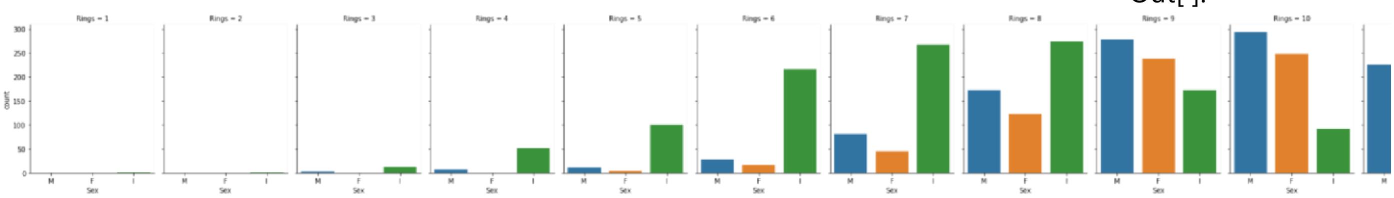


(ii) BIVARIATE

In [ ]:

```
#countplot
sns.catplot(x="Sex",col="Rings",data=df, kind="count",height=4, aspect=.7)
```

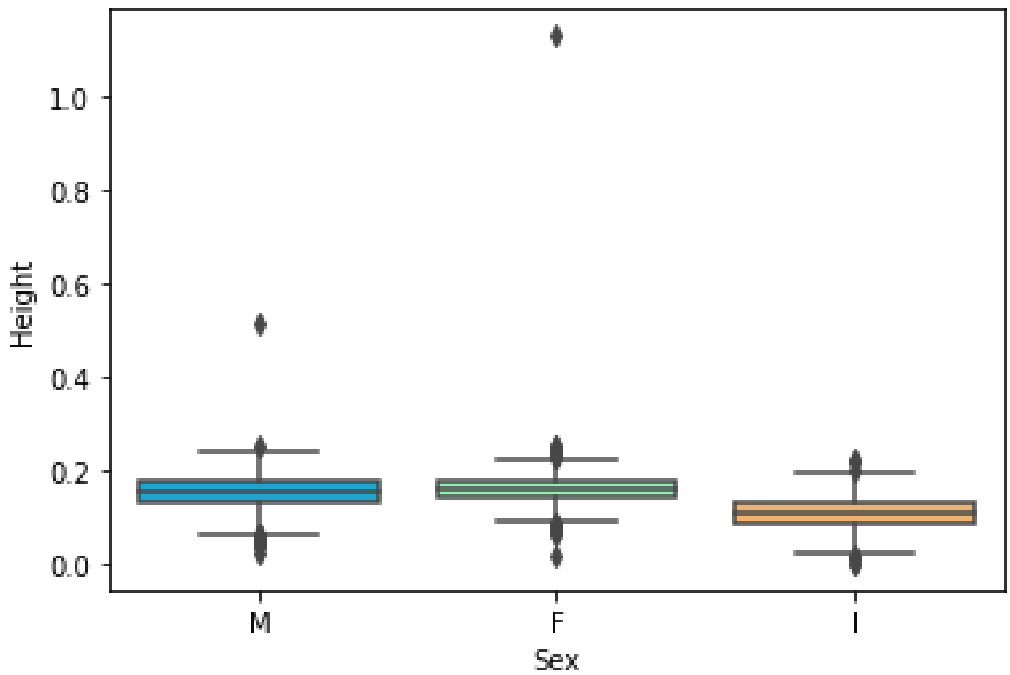
Out[ ]:



In [ ]:

```
#boxplot
sns.boxplot(x='Sex',y='Height',data=df,palette='rainbow')
```

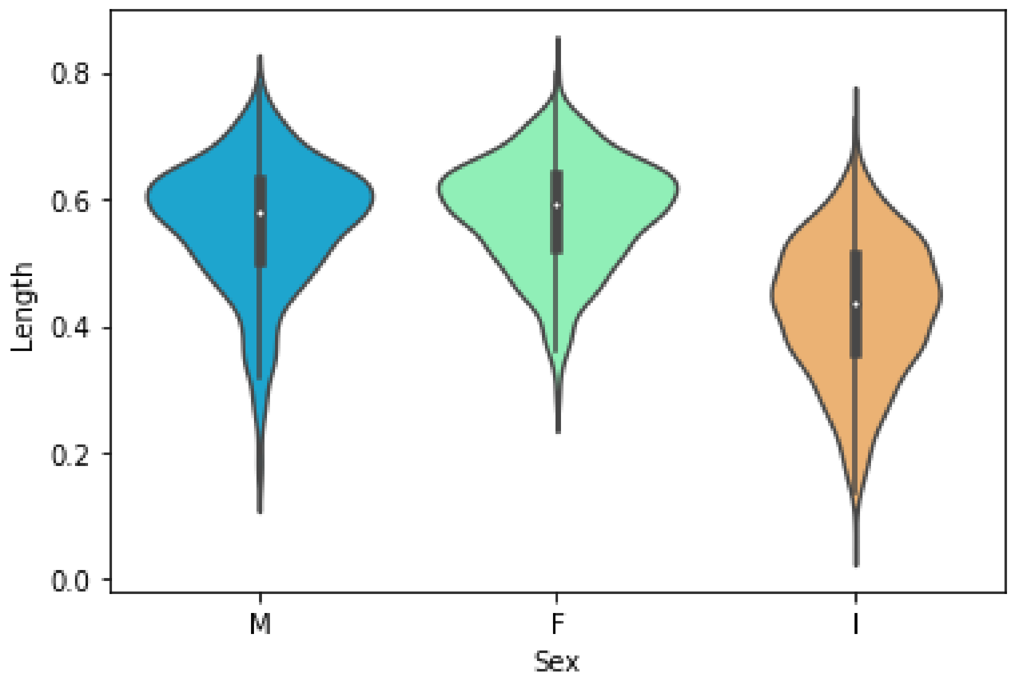
Out[ ]:



In [ ]:

```
#violin plot
sns.violinplot(x="Sex", y="Length", data=df,palette='rainbow')
```

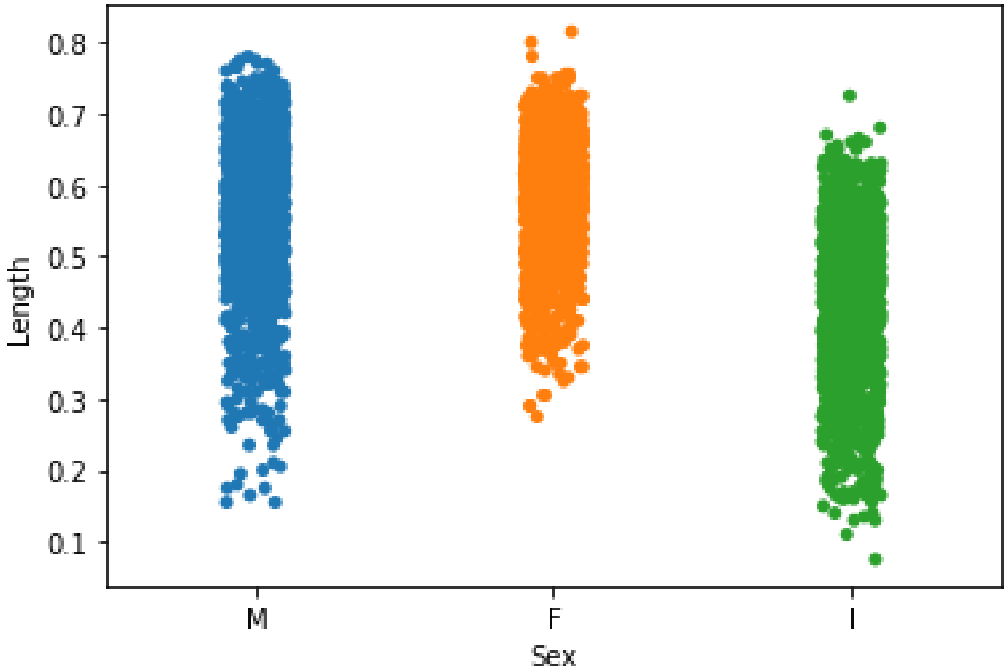
Out[ ]:



In [ ]:

```
#strip plot
sns.stripplot(x="Sex", y="Length", data=df)
```

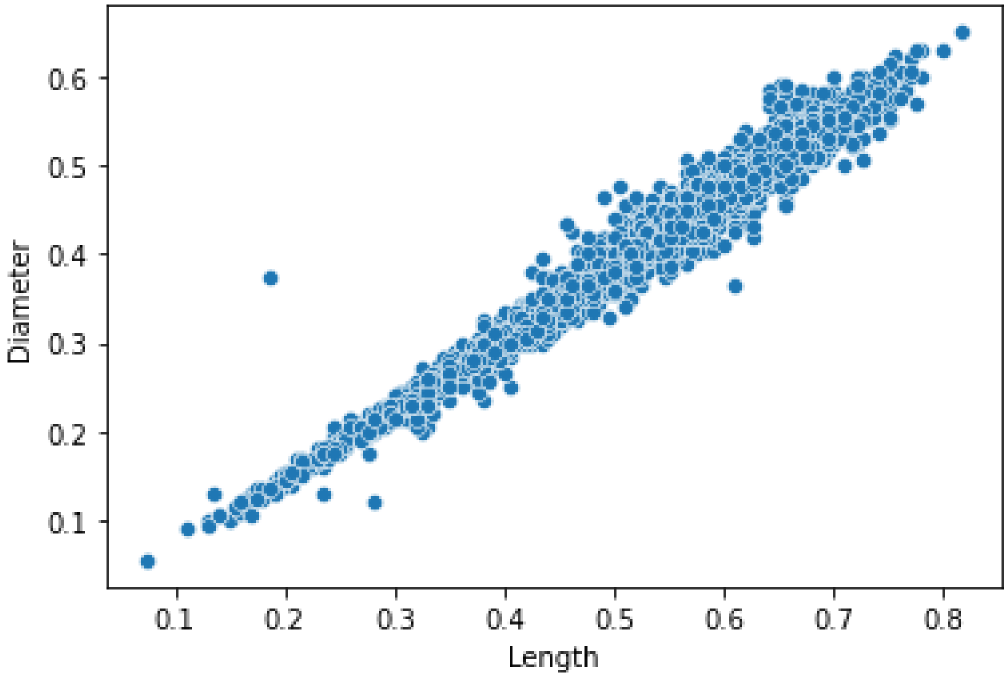
Out[ ]:



In [ ]:

```
#scatter plot
sns.scatterplot(x = df["Length"],y = df["Diameter"])
```

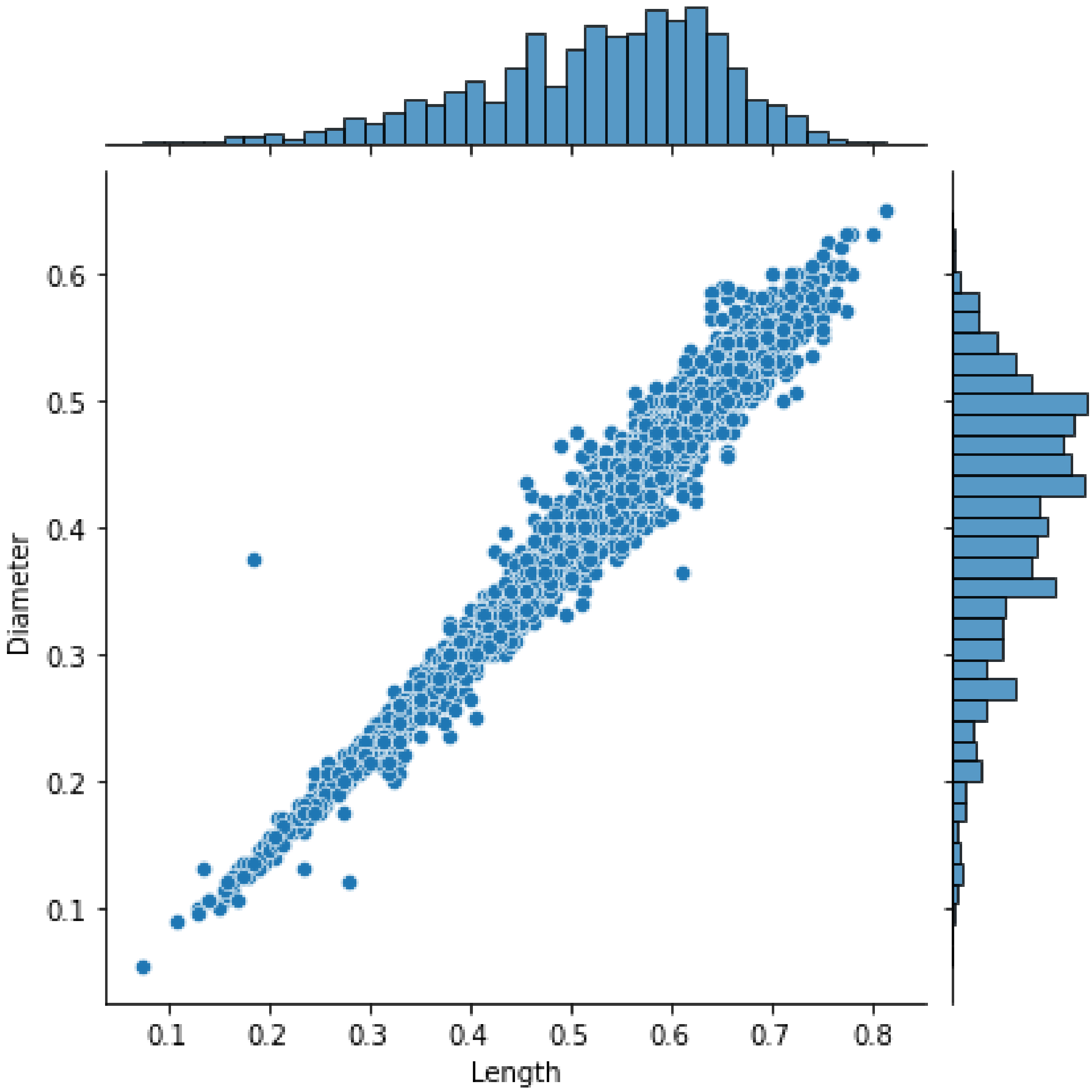
Out[ ]:



In [ ]:

```
#joint_plot
sns.jointplot(x="Length",y="Diameter",data=df)
```

Out[ ]:



(III) MULTI-VARIATE

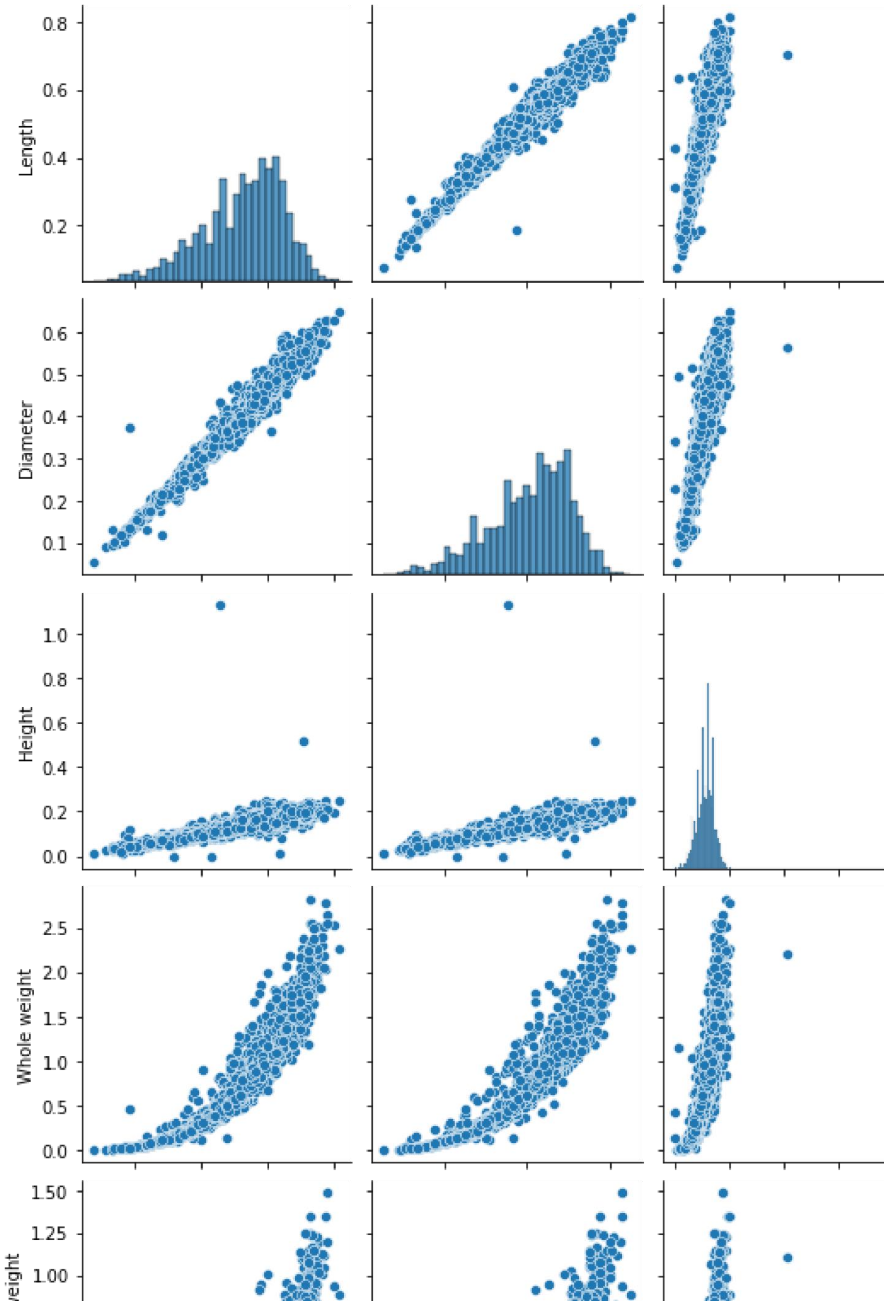
In [ ]:

```
#Boxplot fig, ax1 = plt.subplots(figsize=(8,5)) testPlot =
sns.boxplot(ax=ax1, x='Length', y='Diameter', hue='Sex', data=df)
```

In [ ]:

```
sns.pairplot(df)
```

Out[ ]:

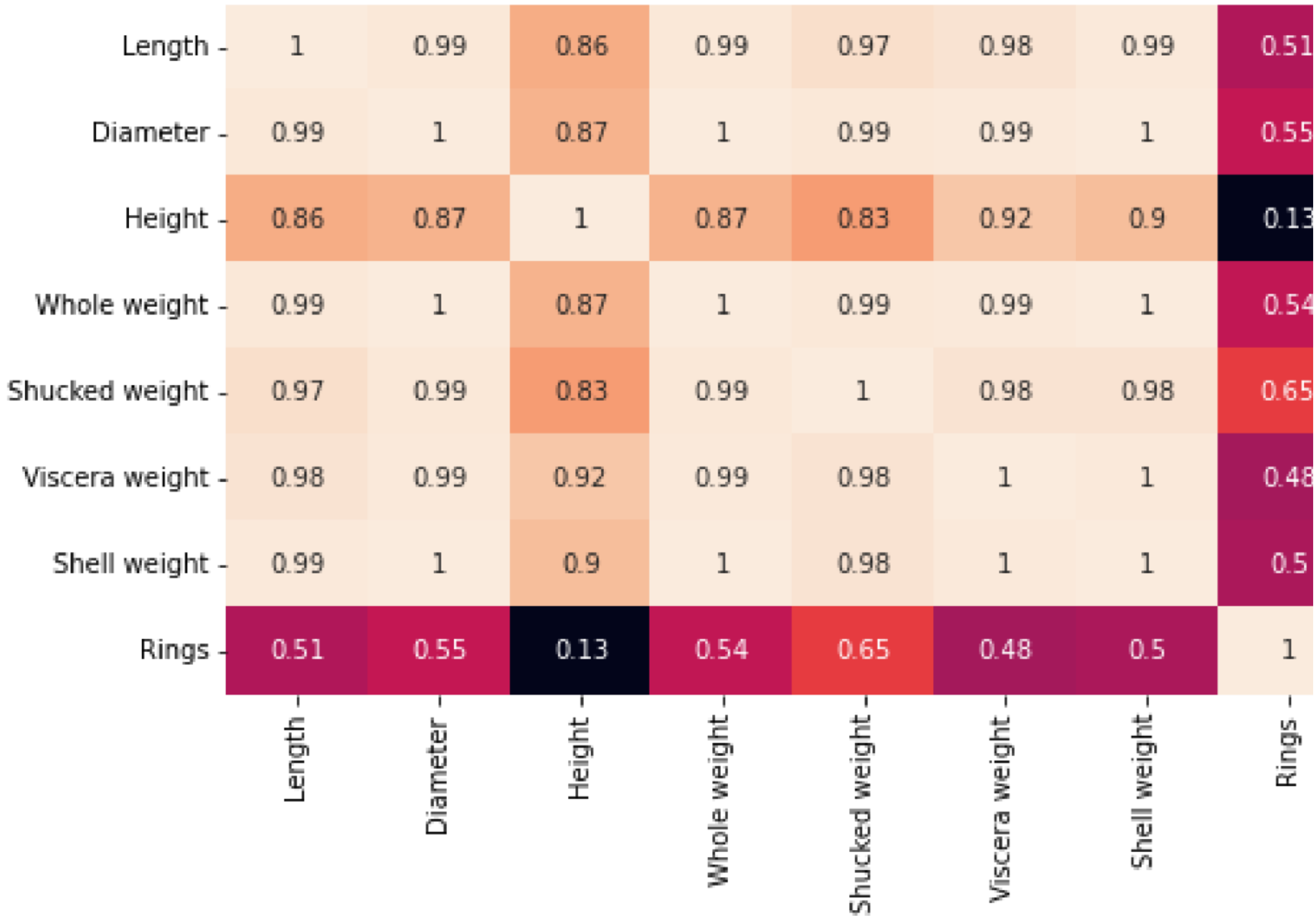


Length
Diameter
Height
Whole weight
weight

In [ ]:

```
fig=plt.figure(figsize=(10,5)) sns.heatmap(df.head().corr(),annot=True)
```

Out[ ]:



4. Perform descriptive statistics on the dataset

In [ ]:

```
df
```

Out[ ]:

	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

	Sex	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	Rings

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	M	0.710	0.555	0.195	1.9485	0.9455	0.3765	0.4950	12

4177 rows × 9 columns

In [ ]:

```
df.head()
```

Out[ ]:

	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

In [ ]:

```
df.info()
```

```
RangeIndex: 4177 entries, 0 to 4176 Data
columns (total 9 columns):
 #   Column          Non-Null Count  Dtype
---  ------          --------------  -----
0   Sex             4177 non-null   object
1   Length          4177 non-null   float64
2   Diameter        4177 non-null   float64
3   Height          4177 non-null   float64
4   Whole weight    4177 non-null   float64
5   Shucked weight  4177 non-null   float64
6   Viscera weight  4177 non-null   float64
```

```
7   Shell weight    4177 non-null   float64  8   Rings          4177 non-
    null    int64    dtypes: float64(7), int64(1), object(1) memory usage:
    293.8+ KB
```

In [ ]:

```
df.describe()
```

Out[ ]:

	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

In [ ]:

```
numerical_features = df.select_dtypes(include = [np.number]).columns
categorical_features = df.select_dtypes(include = [object]).columns
```

In [ ]:

```
df[numerical_features].mean()
```

Out[ ]:

```
Length            0.523992
Diameter          0.407881
Height            0.139516
Whole weight      0.828742
Shucked weight    0.359367
Viscera weight    0.180594
Shell weight      0.238831 Rings
9.933684 dtype: float64
```

In [ ]:

```
df[numerical_features].median()
```

Out[ ]:

```
Length            0.5450
```

```
Diameter          0.4250
Height            0.1400
Whole weight      0.7995
Shucked weight    0.3360
Viscera weight    0.1710
Shell weight      0.2340 Rings
9.0000 dtype: float64
```

In [ ]:

```
percentage = [df[numerical_features].quantile(0),
df[numerical_features].quantile(0.25),
df[numerical_features].quantile(0.50),
df[numerical_features].quantile(0.75),
df[numerical_features].quantile(1)] percentage
```

Out[ ]:

```
[Length            0.0750
 Diameter          0.0550
 Height            0.0000
 Whole weight      0.0020
 Shucked weight    0.0010
 Viscera weight    0.0005
 Shell weight      0.0015
 Rings             1.0000
 Name: 0.0, dtype: float64, Length            0.4500
 Diameter          0.3500
 Height            0.1150
 Whole weight      0.4415
 Shucked weight    0.1860
 Viscera weight    0.0935
 Shell weight      0.1300
 Rings             8.0000
 Name: 0.25, dtype: float64, Length            0.5450
 Diameter          0.4250
 Height            0.1400
 Whole weight      0.7995
 Shucked weight    0.3360
 Viscera weight    0.1710
 Shell weight      0.2340
 Rings             9.0000
 Name: 0.5, dtype: float64, Length             0.615
 Diameter           0.480
 Height             0.165
 Whole weight       1.153
 Shucked weight     0.502
 Viscera weight     0.253
 Shell weight       0.329
 Rings             11.000
 Name: 0.75, dtype: float64, Length             0.8150
 Diameter           0.6500
 Height             1.1300
 Whole weight       2.8255
 Shucked weight     1.4880
 Viscera weight     0.7600
 Shell weight       1.0050
 Rings             29.0000
```

```
Name: 1.0, dtype: float64]
```

In [ ]:

```
df[numerical_features].value_counts()
```

Out[ ]:

```
Length  Diameter  Height  Whole weight  Shucked weight  Viscera weight  She
ll weight  Rings
0.075   0.055     0.010   0.0020        0.0010          0.0005          0.0
015        1        1
0.590   0.465     0.155   1.1360        0.5245          0.2615          0.2
750        11       1
                  0.165   1.1150        0.5165          0.2730          0.2
750        10       1
                  0.170   1.0425        0.4635          0.2400          0.2
700        10       1
                  0.195   1.0885        0.3685          0.1870          0.3
750        17       1

..
0.485   0.370     0.155   0.9680        0.4190          0.2455          0.2
365        9        1
        0.375     0.110   0.4640        0.2015          0.0900          0.1
490        8        1
                  0.125   0.5620        0.2505          0.1345          0.1
525        8        1
                  0.130   0.5535        0.2660          0.1120          0.1
570        8        1
0.815   0.650     0.250   2.2550        0.8905          0.4200          0.7
975        14       1
Length: 4177, dtype: int64
```

In [ ]:

```
df[numerical_features].mode()
```

Out[ ]:

	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	Rings
0	0.550	0.45	0.15	0.2225	0.175	0.1715	0.275	9.0
1	0.625	NaN	NaN	NaN	NaN	NaN	NaN	NaN

In [ ]:

```
df[numerical_features].std()
```

Out[ ]:

```
Length            0.120093
Diameter          0.099240
Height            0.041827
Whole weight      0.490389
Shucked weight    0.221963
Viscera weight    0.109614
Shell weight      0.139203
Rings             3.224169 dtype:
float64
```

In [ ]:

```
df[numerical_features].var()
```

Out[ ]:

```
Length             0.014422
Diameter           0.009849
Height             0.001750
Whole weight       0.240481
Shucked weight     0.049268
Viscera weight     0.012015
Shell weight       0.019377
Rings             10.395266
dtype: float64
```

```
df[numerical_features].skew()
```

In [ ]:

Out[ ]:

```
Length           -0.639873
Diameter         -0.609198
Height            3.128817
Whole weight      0.530959
Shucked weight    0.719098
Viscera weight    0.591852
Shell weight      0.620927
Rings             1.114102
dtype: float64
```

```
df[numerical_features].kurt()
```

In [ ]:

Out[ ]:

```
Length             0.064621
Diameter          -0.045476
Height            76.025509
Whole weight      -0.023644
Shucked weight     0.595124
Viscera weight     0.084012
Shell weight       0.531926
Rings              2.330687
dtype: float64
```

5. Check for Missing values and deal with them

```
df.isnull()
```

In [ ]:

Out[ ]:

	Sex	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	Rings
0	False	False	False	False	False	False	False	False	False
1	False	False	False	False	False	False	False	False	False

	Sex	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	Rings
2	False	False	False	False	False	False	False	False	False

3	False	False	False	False	False	False	False	False	False
4	False	False	False	False	False	False	False	False	False
...	...	...	...	...	...	...	...	...	...
4172	False	False	False	False	False	False	False	False	False
4173	False	False	False	False	False	False	False	False	False
4174	False	False	False	False	False	False	False	False	False
4175	False	False	False	False	False	False	False	False	False
4176	False	False	False	False	False	False	False	False	False

4177 rows × 9 columns

In [ ]:

```
df.isnull().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 [ ]:

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

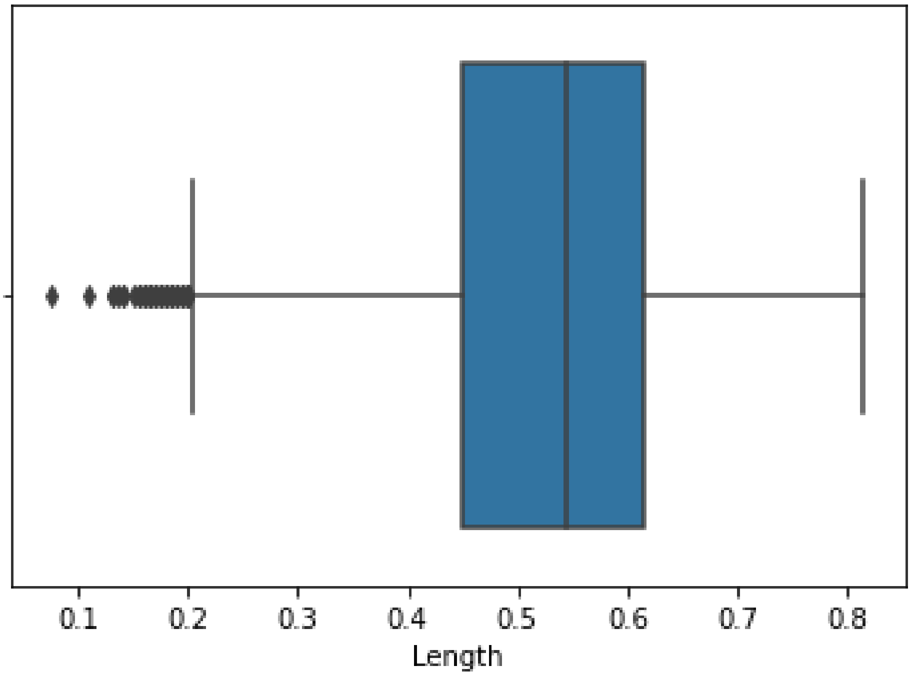
6. Find the outliers and replace them outliers

In [ ]:

```
#length
```

```
sns.boxplot(x=df['Length'])
```

Out[ ]:



In [ ]:

```
q1 = df['Length'].quantile(0.25) q2 = df['Length'].quantile(0.75) iqr = q2-
q1 q1, q2, iqr
```

Out[ ]:

```
(0.45, 0.615, 0.16499999999999998)
```

In [ ]:

```
upper_limit = q2 + (1.5 * iqr) lower_limit = q1 - (1.5 * iqr) lower_limit,
upper_limit
```

Out[ ]:

```
(0.20250000000000004, 0.8624999999999999)
```

In [ ]:

```
new_df = df.loc[(df['Length'] <= upper_limit) & (df['Length'] >=
lower_limit)]
print('before removing outliers:', len(df))
print('after removing outliers:',len(new_df))
print('outliers:', len(df)-len(new_df))
before removing outliers: 4177 after removing
outliers: 4128 outliers: 49
```

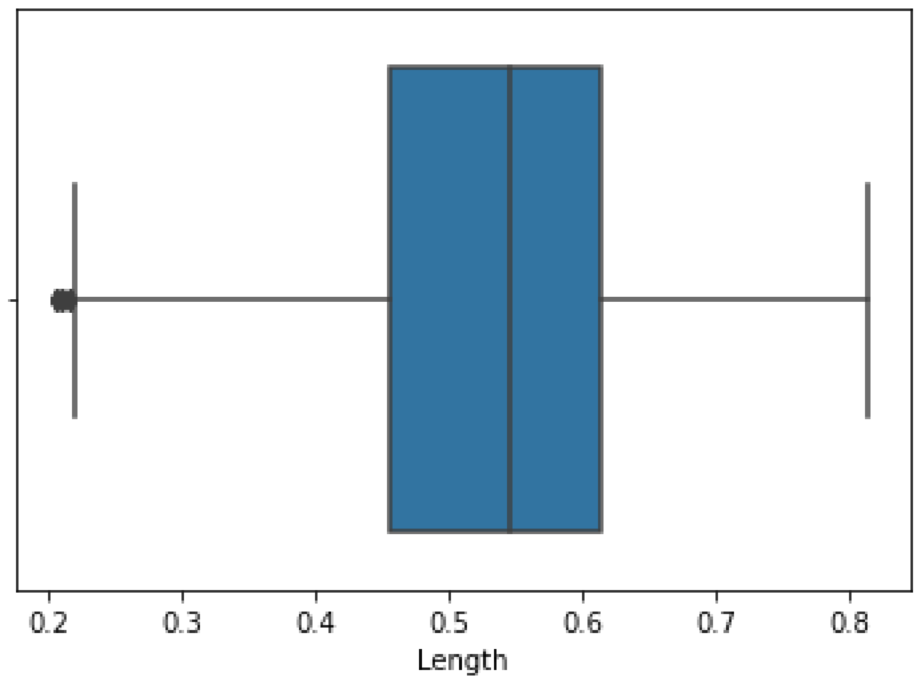
In [ ]:

```
new_df = df.copy() new_df.loc[(new_df['Length']>upper_limit), 'Length'] =
upper_limit new_df.loc[(new_df['Length']<lower_limit), 'Length'] =
lower_limit
```

In [ ]:

```
sns.boxplot(x=new_df['Length'])
```

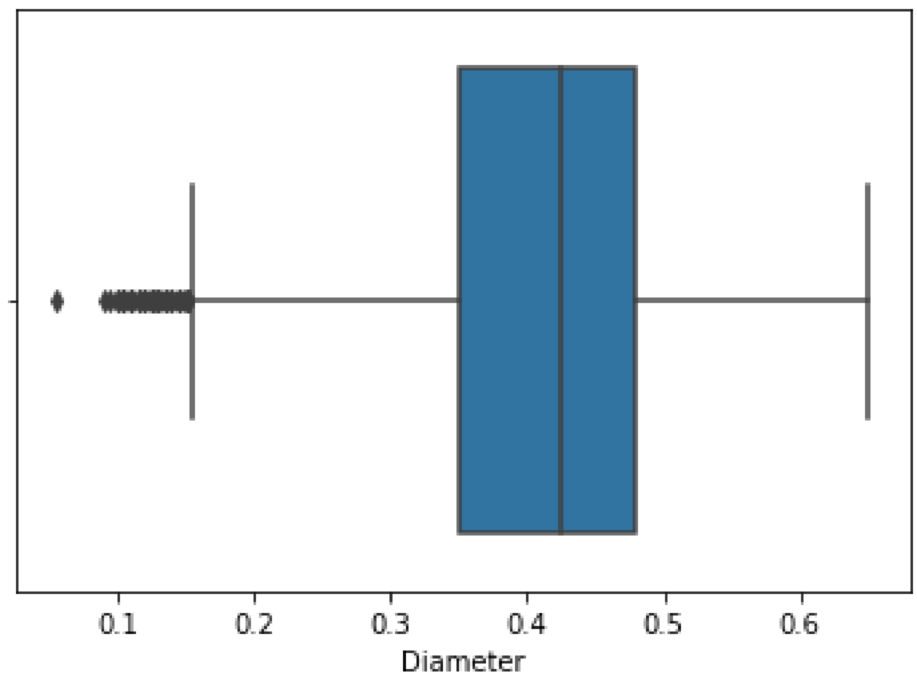
Out[ ]:



In [ ]:

```
#Diameter
sns.boxplot(x=df['Diameter'])
```

Out[ ]:



In [ ]:

```
q1 = df['Diameter'].quantile(0.25) q2 = df['Diameter'].quantile(0.75) iqr =
q2-q1
q1, q2, iqr
```

Out[ ]:

```
(0.45, 0.615, 0.16499999999999998)
```

In [ ]:

```
upper_limit = q2 + (1.5 * iqr) lower_limit = q1 - (1.5 * iqr) lower_limit, upper_limit
```

Out[ ]:

```
(0.20250000000000004, 0.8624999999999999)
```

In [ ]:

```
new_df = df.loc[(df['Diameter'] <= upper_limit) & (df['Diameter'] >= lower_limit)] print('before removing outliers:', len(df)) print('after removing outliers:',len(new_df)) print('outliers:', len(df)-len(new_df))
before removing outliers: 4177 after removing outliers: 4027 outliers: 150
```

In [ ]:

```
new_df = df.copy() new_df.loc[(new_df['Diameter']>upper_limit), 'Diameter'] = upper_limit new_df.loc[(new_df['Diameter']<lower_limit), 'Diameter'] = lower_limit
```

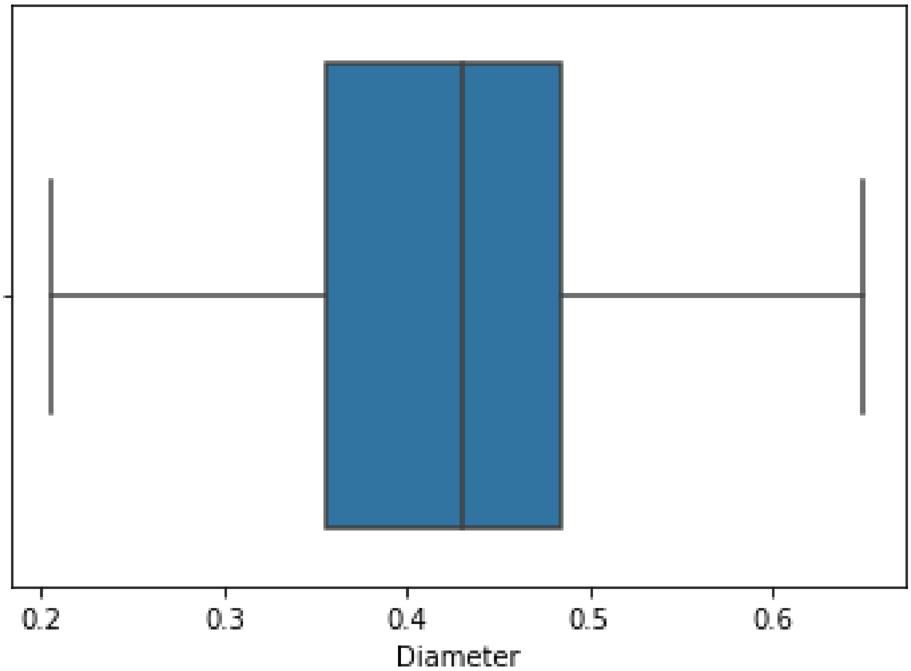
```
/usr/local/lib/python3.7/dist-packages/pandas/core/indexing.py:1817: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
self._setitem_single_column(loc, value, pi)
```

In [ ]:

```
sns.boxplot(x=new_df['Diameter'])
```

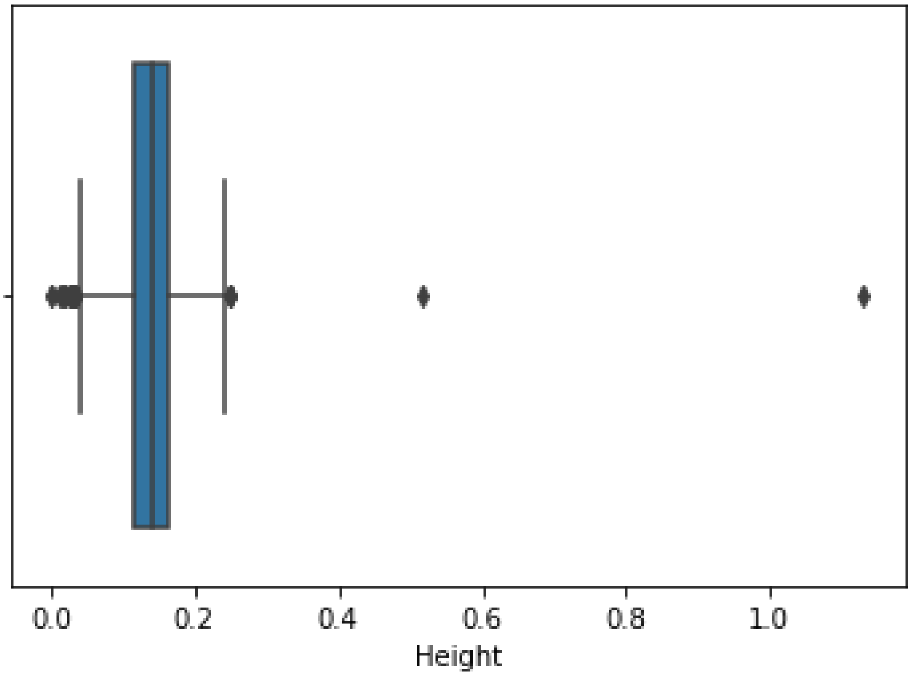
Out[ ]:



In [ ]:

```
#Height
sns.boxplot(x=df['Height'])
```

Out[ ]:



In [ ]:

```
q1 = df['Height'].quantile(0.25) q2 = df['Height'].quantile(0.75) iqr = q2-
q1 q1, q2, iqr
```

Out[ ]:

```
(0.45, 0.615, 0.16499999999999998)
```

In [ ]:

```
upper_limit = q2 + (1.5 * iqr) lower_limit
= q1 - (1.5 * iqr) lower_limit,
upper_limit
```

Out[ ]:

```
(0.20250000000000004, 0.8624999999999999)
```

In [ ]:

```
new_df = df.loc[(df['Height'] <= upper_limit) & (df['Height'] >=
lower_limit)] print('before removing outliers:', len(df)) print('after
removing outliers:',len(new_df)) print('outliers:', len(df)-len(new_df))
before removing outliers: 4177 after removing outliers: 153 outliers: 4024
```

In [ ]:

```
new_df = df.copy() new_df.loc[(new_df['Height']>upper_limit), 'Height'] =
upper_limit new_df.loc[(new_df['Height']<lower_limit), 'Height'] =
lower_limit
```

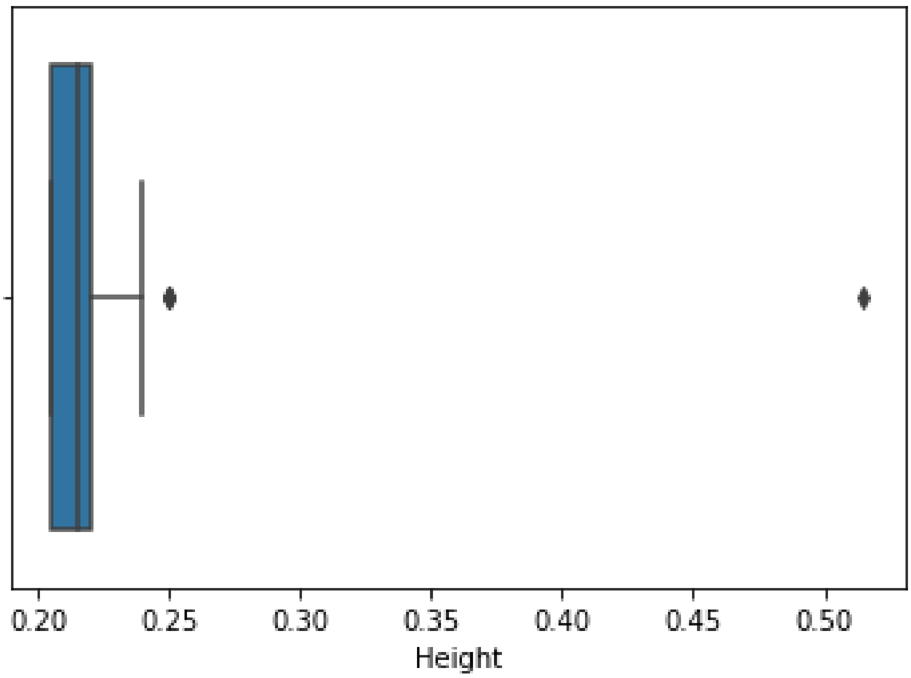
```
/usr/local/lib/python3.7/dist-packages/pandas/core/indexing.py:1817: Settin
gWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs
/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
self._setitem_single_column(loc, value, pi)
```

In [ ]:

```
sns.boxplot(x=new_df['Height'])
```

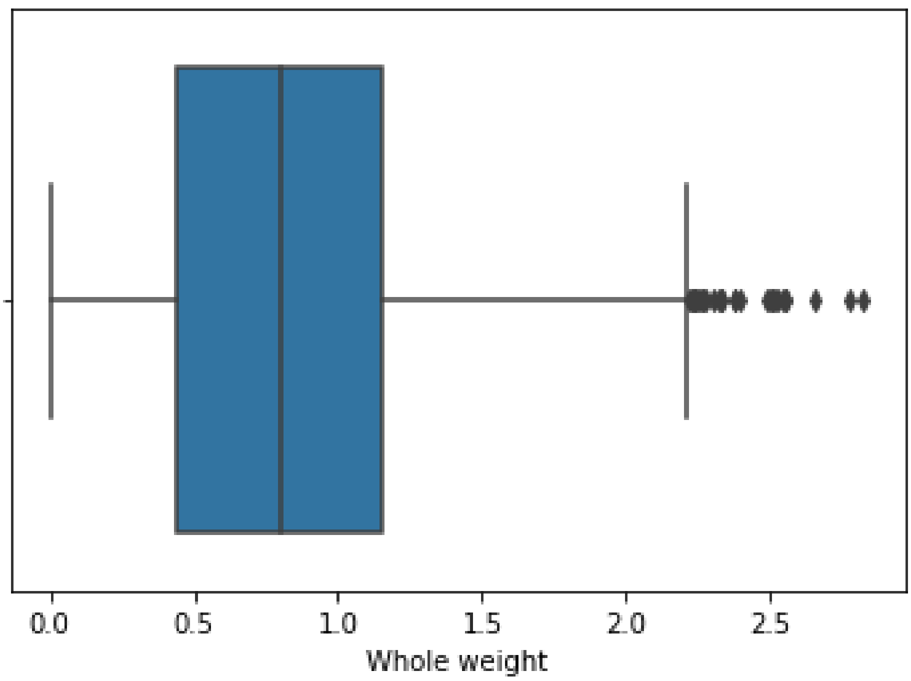
Out[ ]:



In [ ]:

```
#Whole Weight sns.boxplot(x=df['Whole
weight'])
```

Out[ ]:



In [ ]:

```
q1 = df['Whole weight'].quantile(0.25) q2 = df['Whole weight'].quantile(0.75)
iqr = q2-q1 q1, q2, iqr
```

Out[ ]:

```
(0.45, 0.615, 0.16499999999999998)
```

In [ ]:

```
upper_limit = q2 + (1.5 * iqr) lower_limit = q1 - (1.5 * iqr) lower_limit,
upper_limit
```

Out[ ]:

```
(0.20250000000000004, 0.8624999999999999)
```

In [ ]:

```
new_df = df.loc[(df['Whole weight'] <= upper_limit) & (df['Whole weight']
>= lower_limit)] print('before removing
outliers:', len(df)) print('after removing
outliers:',len(new_df)) print('outliers:',
len(df)-len(new_df)) before removing
outliers: 4177 after removing outliers: 1872
outliers: 2305
```

In [ ]:

```
new_df = df.copy() new_df.loc[(new_df['Whole weight']>upper_limit), 'Whole
weight'] = upper_limit new_df.loc[(new_df['Whole weight']<lower_limit),
'Whole weight'] = lower_limit
```

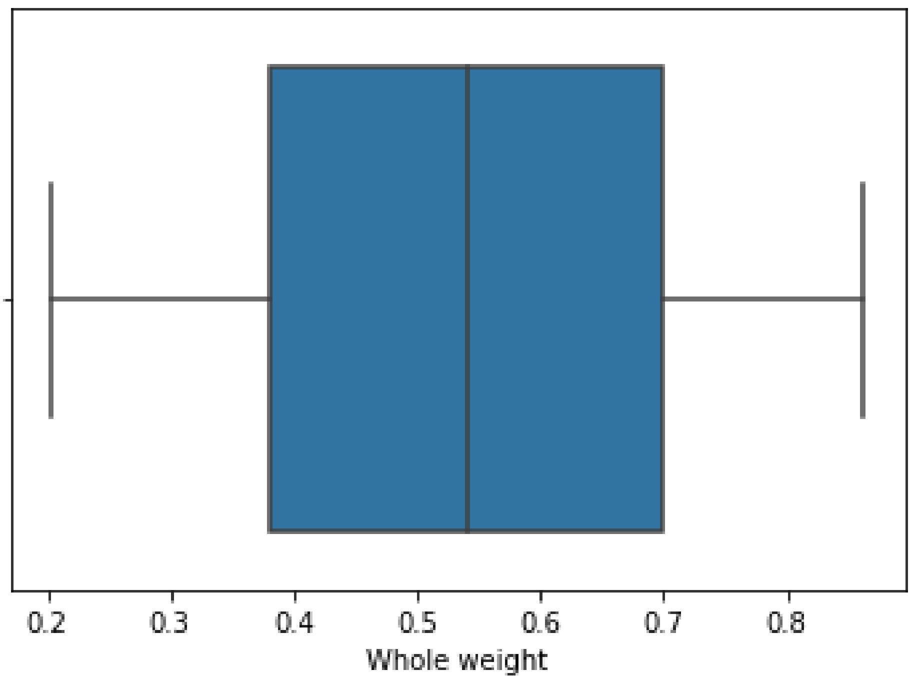
```
/usr/local/lib/python3.7/dist-packages/pandas/core/indexing.py:1817: Settin
gWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame. Try
using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs
/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
self._setitem_single_column(loc, value, pi)
```

In [ ]:

```
sns.boxplot(x=new_df['Whole weight'])
```

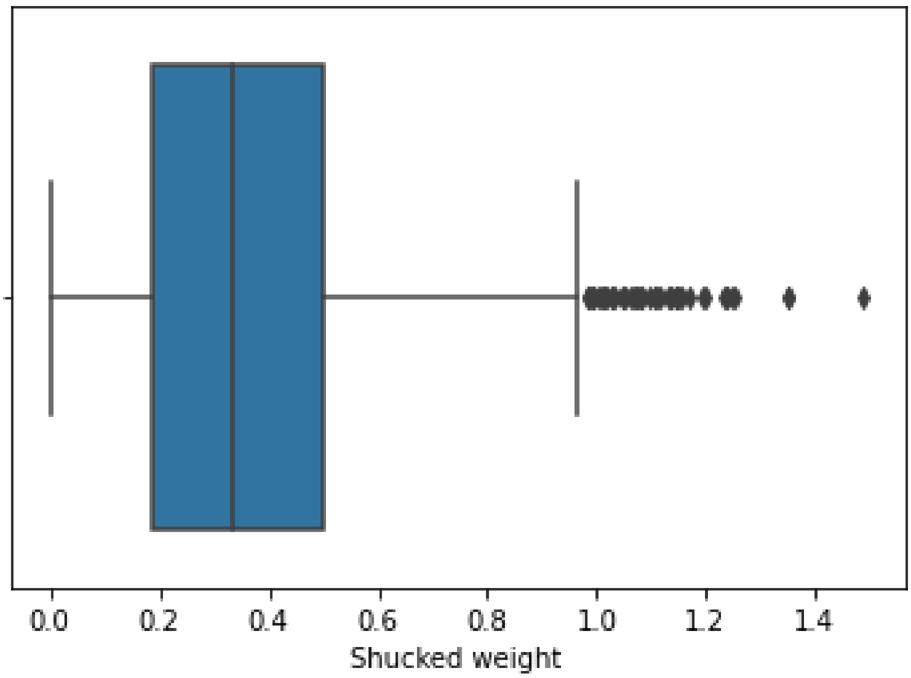
Out[ ]:



In [ ]:

```
#Shucked weight
sns.boxplot(x=df['Shucked weight'])
```

Out[ ]:



In [ ]:

```
q1 = df['Shucked weight'].quantile(0.25)
q2 = df['Shucked weight'].quantile(0.75)
iqr = q2-q1 q1, q2, iqr
```

Out[ ]:

```
(0.45, 0.615, 0.16499999999999998)
```

In [ ]:

```
upper_limit = q2 + (1.5 * iqr) lower_limit = q1 - (1.5 * iqr) lower_limit, upper_limit
```

Out[ ]:

```
(0.20250000000000004, 0.8624999999999999)
```

In [ ]:

```
new_df = df.loc[(df['Shucked weight'] <= upper_limit) & (df['Shucked weight'] >= lower_limit)] print('before removing outliers:', len(df)) print('after removing outliers:',len(new_df)) print('outliers:', len(df)-len(new_df)) before removing outliers: 4177 after removing outliers: 2900 outliers: 1277
```

In [ ]:

```
new_df = df.copy() new_df.loc[(new_df['Shucked weight']>upper_limit), 'Shucked weight'] = upper_limit new_df.loc[(new_df['Shucked weight']<lower_limit), 'Shucked weight'] = lower_limit
```

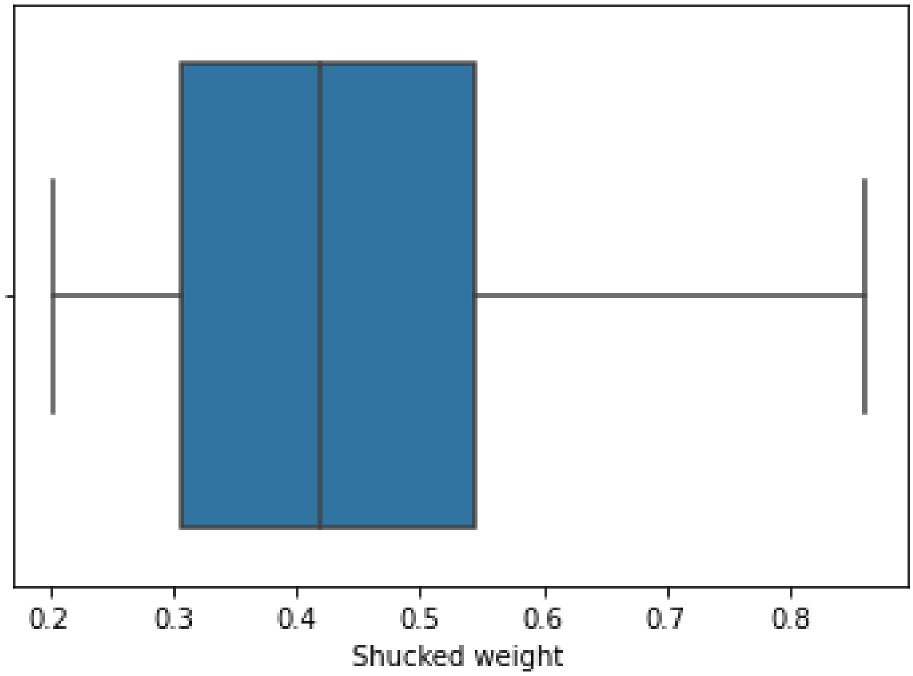
```
/usr/local/lib/python3.7/dist-packages/pandas/core/indexing.py:1817: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
self._setitem_single_column(loc, value, pi)
```

In [ ]:

```
sns.boxplot(x=new_df['Shucked weight'])
```

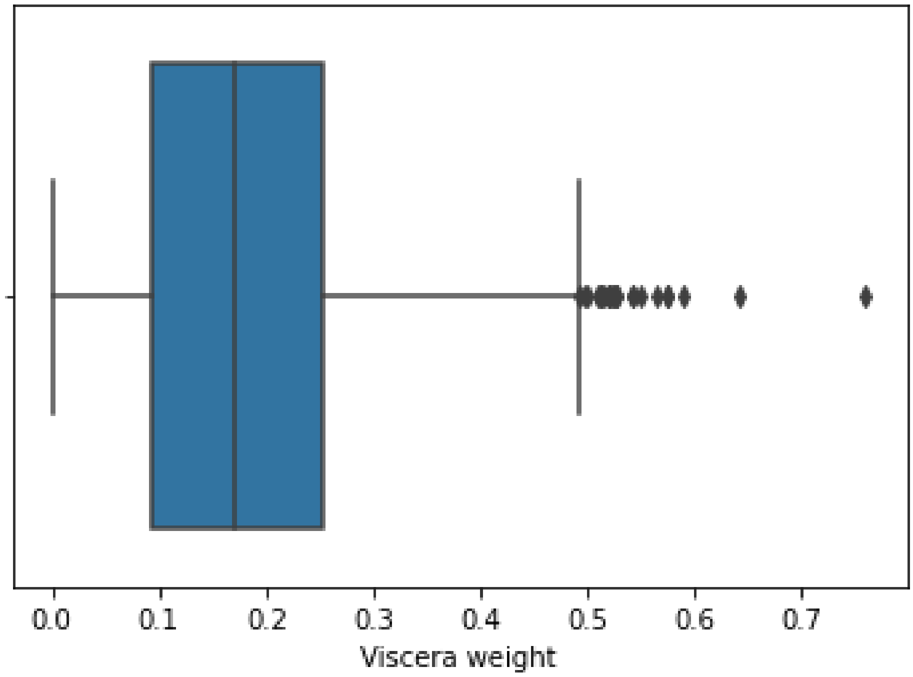
Out[ ]:



In [ ]:

```
#Viscera weight
sns.boxplot(x=df['Viscera weight'])
```

Out[ ]:



In [ ]:

```
q1 = df['Viscera weight'].quantile(0.25) q2 = df['Viscera weight'].quantile(0.75) iqr = q2-q1 q1, q2, iqr
```

Out[ ]:

```
(0.45, 0.615, 0.16499999999999998)
```

In [ ]:

```
upper_limit = q2 + (1.5 * iqr) lower_limit
= q1 - (1.5 * iqr) lower_limit,
upper_limit
```

Out[ ]:

```
(0.20250000000000004, 0.8624999999999999)
```

In [ ]:

```
new_df = df.loc[(df['Viscera weight'] <= upper_limit) & (df['Viscera weight']
>= lower_limit)] print('before removing outliers:', len(df)) print('after
removing outliers:',len(new_df)) print('outliers:', len(df)-len(new_df))
before removing outliers: 4177 after removing outliers: 1646 outliers: 2531
```

In [ ]:

```
new_df = df.copy() new_df.loc[(new_df['Viscera weight']>upper_limit),
'Viscera weight'] = upper_limit new_df.loc[(new_df['Viscera
weight']<lower_limit), 'Viscera weight'] = lower_limit
```

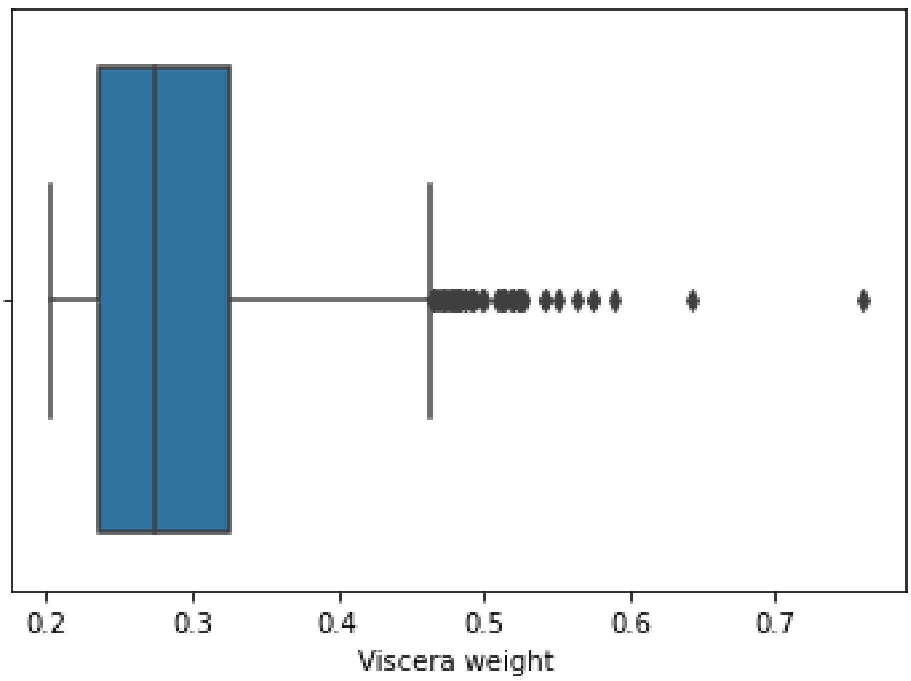
```
/usr/local/lib/python3.7/dist-packages/pandas/core/indexing.py:1817: Settin
gWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs
/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
self._setitem_single_column(loc, value, pi)
```

In [ ]:

```
sns.boxplot(x=new_df['Viscera weight'])
```

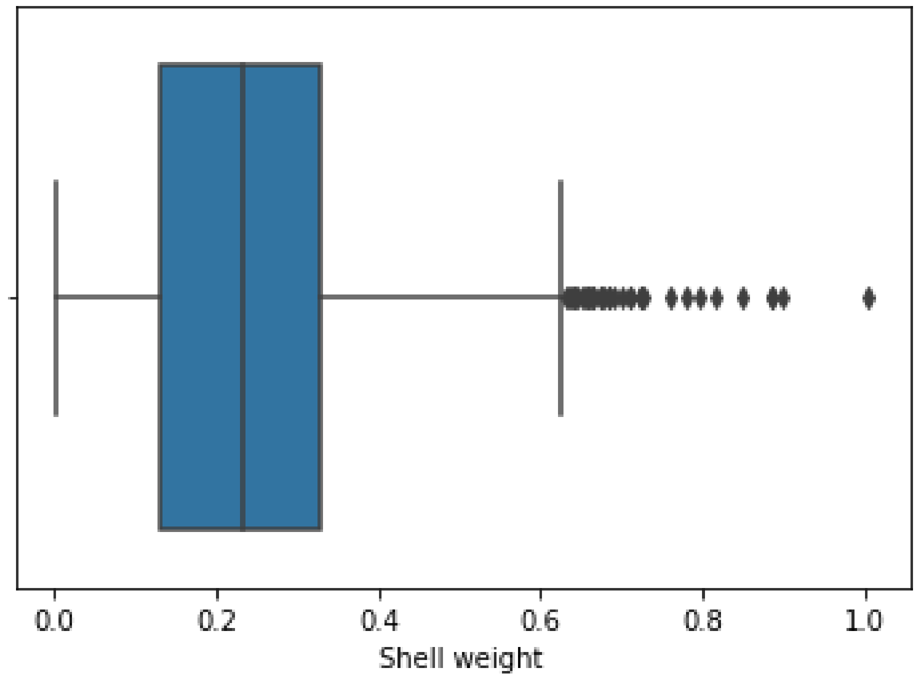
Out[ ]:



In [ ]:

```
#shell weight
sns.boxplot(x=df['Shell weight'])
```

Out[ ]:



In [ ]:

```
q1 = df['Shell weight'].quantile(0.25) q2 = df['Shell weight'].quantile(0.75)
iqr = q2-q1 q1, q2, iqr
```

Out[ ]:

```
(0.45, 0.615, 0.16499999999999998)
```

In [ ]:

```
upper_limit = q2 + (1.5 * iqr) lower_limit = q1 - (1.5 * iqr) lower_limit,
upper_limit
```

Out[ ]:

```
(0.20250000000000004, 0.8624999999999999)
```

In [ ]:

```
new_df = df.loc[(df['Shell weight'] <= upper_limit) & (df['Shell weight']
>= lower_limit)] print('before removing
outliers:', len(df)) print('after removing
outliers:',len(new_df)) print('outliers:',
len(df)-len(new_df)) before removing
outliers: 4177 after removing outliers: 2373
outliers: 1804
```

In [ ]:

```
new_df = df.copy() new_df.loc[(new_df['Shell weight']>upper_limit), 'Shell
weight'] = upper_limit new_df.loc[(new_df['Shell weight']<lower_limit),
'Shell weight'] = lower_limit
```

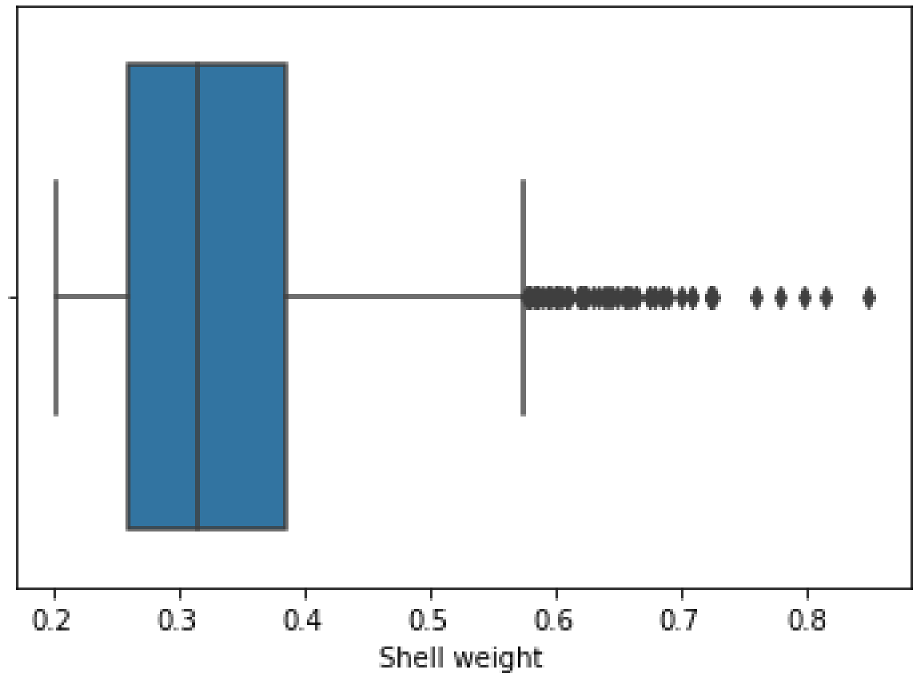
```
/usr/local/lib/python3.7/dist-packages/pandas/core/indexing.py:1817: Settin
gWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs
/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
self._setitem_single_column(loc, value, pi)
```

In [ ]:

```
sns.boxplot(x=new_df['Shell weight'])
```

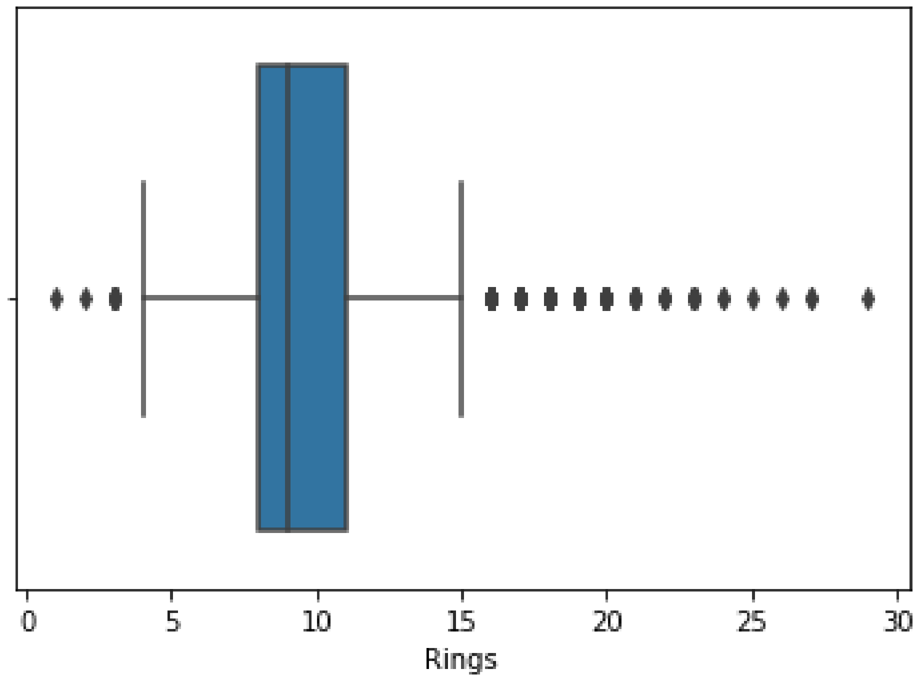
Out[ ]:



In [ ]:

```
#Rings
sns.boxplot(x=df['Rings'])
```

Out[ ]:



In [ ]:

```
q1 = df['Rings'].quantile(0.25) q2 = df['Rings'].quantile(0.75) iqr = q2-q1
q1, q2, iqr
```

Out[ ]:

```
(0.45, 0.615, 0.16499999999999998)
```

In [ ]:

```
upper_limit = q2 + (1.5 * iqr) lower_limit = q1 - (1.5 * iqr) lower_limit, upper_limit
```

Out[ ]:

```
(0.20250000000000004, 0.8624999999999999)
```

In [ ]:

```
new_df = df.loc[(df['Rings'] <= upper_limit) & (df['Rings'] >= lower_limit)] print('before removing outliers:', len(df)) print('after removing outliers:',len(new_df)) print('outliers:', len(df)-len(new_df)) before removing outliers: 4177 after removing outliers: 0 outliers: 4177
```

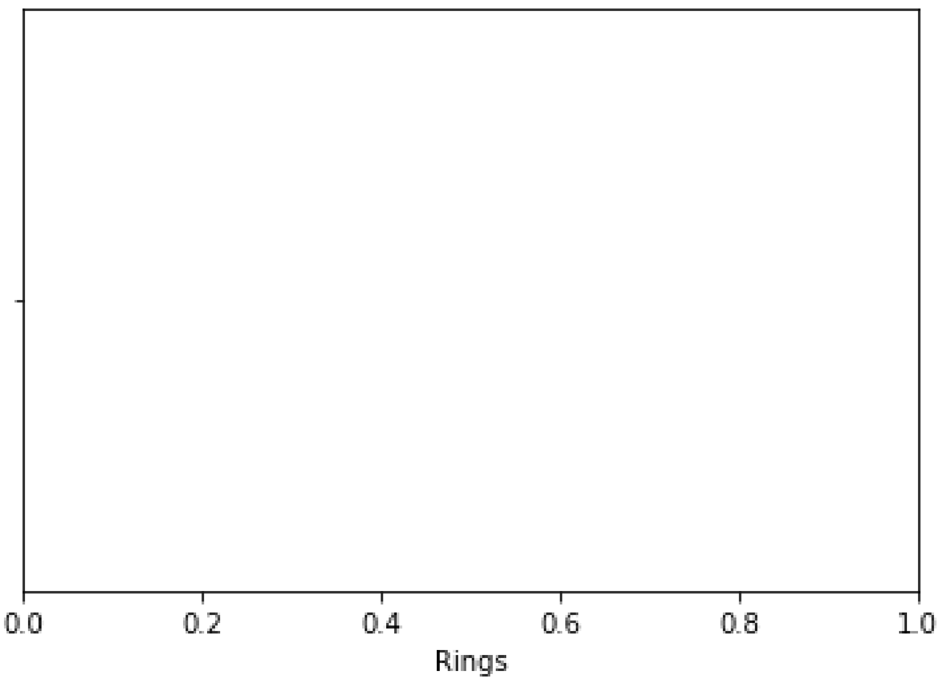
In [ ]:

```
new_df = df.copy() new_df.loc[(new_df['Rings']>upper_limit), 'Rings'] = upper_limit new_df.loc[(new_df['Rings']<lower_limit), 'Rings'] = lower_limit
```

In [219]:

```
sns.boxplot(x=new_df['Rings'])
```

Out[219]:



7. Check for Categorical columns and perform encoding

In [ ]:

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

```
df
```

Out[ ]:

	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

	Sex	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	Rings
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	M	0.710	0.555	0.195	1.9485	0.9455	0.3765	0.4950	12

4177 rows × 9 columns

In [ ]:

```
from sklearn.preprocessing import LabelEncoder,OneHotEncoder,StandardScaler
```

In [ ]:

```
label_encoder =LabelEncoder() df['Sex']=
label_encoder.fit_transform(df['Sex']) df
```

Out[ ]:

	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	M	0.710	0.555	0.195	1.9485	0.9455	0.3765	0.4950	12

4177 rows × 9 columns

In [ ]:

```
enc = OneHotEncoder(drop='first')
 enc_df                                                          =
pd.DataFrame(enc.fit_transform(df[['Sex']]).toarray())
 df
=df.join(enc_df)
df.head()
```

Out[ ]:

	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

8. Split the data into dependent and independent variables

In [ ]:

```
x= df.iloc[:,1:8] x
```

Out[ ]:

	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight
0	0.455	0.365	0.095	0.5140	0.2245	0.1010	0.1500
1	0.350	0.265	0.090	0.2255	0.0995	0.0485	0.0700
2	0.530	0.420	0.135	0.6770	0.2565	0.1415	0.2100

	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight
3	0.440	0.365	0.125	0.5160	0.2155	0.1140	0.1550
4	0.330	0.255	0.080	0.2050	0.0895	0.0395	0.0550
...	...	...	...	...	...	...	...
4172	0.565	0.450	0.165	0.8870	0.3700	0.2390	0.2490
4173	0.590	0.440	0.135	0.9660	0.4390	0.2145	0.2605
4174	0.600	0.475	0.205	1.1760	0.5255	0.2875	0.3080
4175	0.625	0.485	0.150	1.0945	0.5310	0.2610	0.2960
4176	0.710	0.555	0.195	1.9485	0.9455	0.3765	0.4950

4177 rows × 7 columns

In [ ]:

```
y=df.iloc[:,8] y
```

Out[ ]:

```
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
```

9. Scale the independent variables

In [ ]:

```
scale = StandardScaler() scaledX = scale.fit_transform(x)
 print(scaledX)
```

```
[[-0.57455813  -0.43214879  -1.06442415  ...  -0.60768536  -0.72621157     -
0.63821689]
 [-1.44898585  -1.439929     -1.18397831  ...  -1.17090984  -1.20522124     -
1.21298732]
 [  0.05003309   0.12213032  -0.10799087  ...  -0.4634999   -0.35668983     -
0.20713907]
 ...
 [ 0.6329849   0.67640943  1.56576738 ...  0.74855917  0.97541324
   0.49695471]
```

```
 [  0.84118198    0.77718745    0.25067161  ...    0.77334105    0.73362741
0.41073914]
 [ 1.54905203  1.48263359  1.32665906 ...  2.64099341  1.78744868
   1.84048058]]
```

10. Split the data into training and testing

In [ ]:

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

In [ ]:

```python
x_train, x_test, y_train, y_test = train_test_split(x,y, test_size = 0.2)
```

In [ ]:

```python
print(x.shape, x_train.shape, x_test.shape,y_train.shape, y_test.shape)
```

```
(4177, 7) (3341, 7) (836, 7) (3341,) (836,)
```

11. Build the Model

In [ ]:

```python
from sklearn.linear_model import LinearRegression
```

In [ ]:

```python
linearmodel = LinearRegression()
```

12. Train the Model

In [ ]:

```python
linearmodel.fit(x_train, y_train)
```

Out[ ]:

```
LinearRegression()
```

13. Test the Model

In [ ]:

```python
y_train_pred = linearmodel.predict(x_train) y_test_pred = linearmodel.predict(x_test)
```

In [ ]:

```python
y_test_pred
```

Out[ ]:

```
array([ 8.70365574, 10.39057789,  9.40293106, 10.68158892,  7.57464889,
4.79636131,  8.67332668,  14.02754984,  9.87864789,  7.25750569,
10.85233616,  8.50778462,  7.15078854,  9.32393986,  5.76619464,
7.49797457,  5.76688568,  6.2241946  ,  6.18696811,  9.25884721,
11.5681706  ,  12.13604097,  10.98303848,  11.69986211,  7.83702624,
9.31462136,  10.40327259,  6.96378017,  5.81839663,  12.26690446,
10.86817082,  9.02369275,  8.12760588,  8.83313399,  7.73292169,
6.91592262,  6.07309496,  7.88643423,  9.63507119,  4.7209354  ,
11.34436294,  10.57283751,  10.49665213,  12.88894543,  12.28423666,
8.12974709,  7.58999374,  9.08527348,  10.6411015  ,  5.89349286,
8.30141881,  7.24999833,  8.25347176,  9.08328759,  8.99010706,
9.10730271,  14.52308851,  9.41403346,  10.23775522,  8.0477514  ,
9.70375626,  11.03640036,  9.5435852  ,  10.8850895  ,  8.08267684,
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```

14. Measure the performance using Metrics

In [ ]:

```
from sklearn.metrics import mean_absolute_error, mean_squared_error s = mean_squared_error(y_train, y_train_pred) print('Mean Squared error of training set :%2f'%s)
```

In [ ]:

```
p = mean_squared_error(y_test, y_test_pred) print('Mean Squared error of testing set :%2f'%p)
```

```
Mean Squared error of testing set :4.869245
```

In [ ]:

```
# Build the Model
from sklearn.ensemble import RandomForestRegressor
```

In [ ]:

```
rfr = RandomForestRegressor(max_depth=2, random_state=0,
n_estimators=100)
```

In [ ]:

```
rfr.fit(x_train, y_train) rfr.fit(x_test, y_test)
```

Out[ ]:

```
RandomForestRegressor(max_depth=2, random_state=0)
```

In [ ]:

```
#Test the model y_train_pred =
rfr.predict(x_train) y_test_pred =
rfr.predict(x_test)
```

In [ ]:

```
#measure the performance using metrics rfr.score(x_test,
y_test)
```

Out[ ]:

```
0.4497726034378102
```

K Neighbors Regression

In [ ]:

```
#Build the model
from sklearn.neighbors import KNeighborsRegressor
```

In [ ]:

```
knr = KNeighborsRegressor(n_neighbors =4 )
```

In [ ]:

```
#Train the model knr.fit(x_train,
y_train) knr.fit(x_test, y_test)
```

Out[ ]:

```
KNeighborsRegressor(n_neighbors=4)
```

In [ ]:

```
#Test the model y_train_pred =
knr.predict(x_train) y_test_pred =
knr.predict(x_test)
```

In [ ]:

```
#Measure the performance using Metrics
knr.score(x_train, y_train)
```

Out[ ]:

```
0.458628955466746
```

Decision Tree Regression

In [ ]:

```
#Build the model
from sklearn.tree import DecisionTreeRegressor
```

dtr = DecisionTreeRegressor(random_state=0)

In [ ]:

```
#Train the model dtr.fit(x_test,y_test)
```

Out[ ]:

```
DecisionTreeRegressor(random_state=0)
```

Test the model

y_train_pred = dtr.predict(x_train) y_test_pred = dtr.predict(x_test)

In [ ]:

```
#Mesure the performance using Metrics dtr.score(x_train,
y_train)
```

Out[ ]:

```
0.15715160117393523
```

Lasso Regression

In [214]:

```
#Build the model
from sklearn.linear_model import Lasso
```

In [215]:

```
lr=Lasso(alpha=0.01)
```

In [216]:

```
#Train the model lr.fit(x_train,y_train)
```

Out[216]:

```
Lasso(alpha=0.01)
```

In [217]:

```
y_train_pred = lr.predict(x_train) y_test_pred = lr.predict(x_test)
```

In [218]:

```
#Measure the performance using Metrics lr.score(x_train,
y_train)
```

Out[218]: 0.5098141532900928