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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

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

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

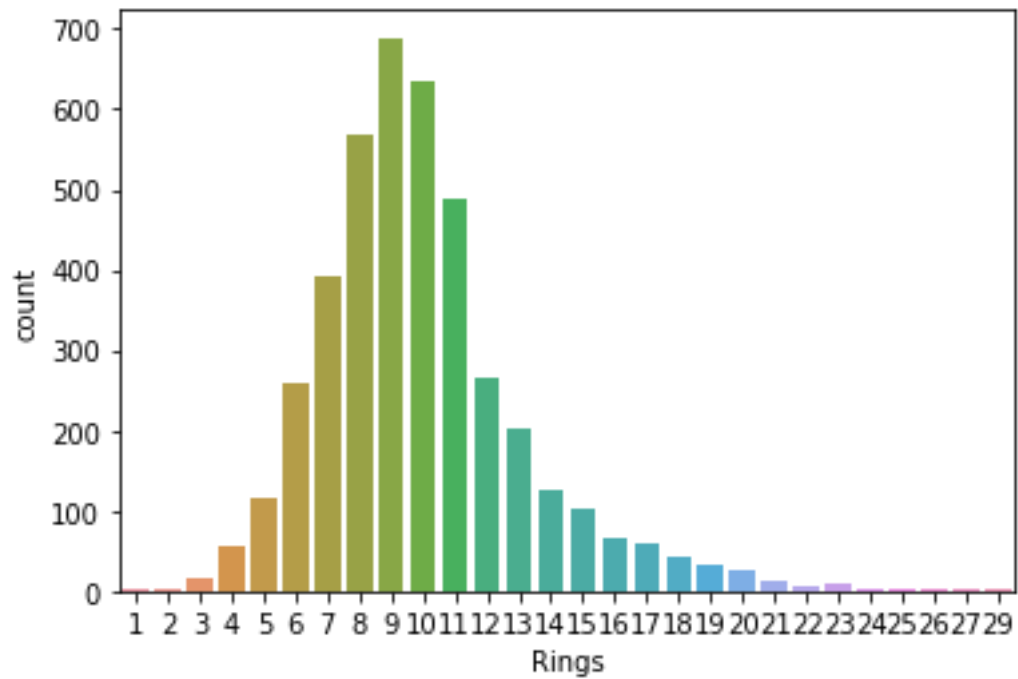
(iii)Multi-Variate Analysis

(i) UNIVARIATE

```
In [ ]:

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

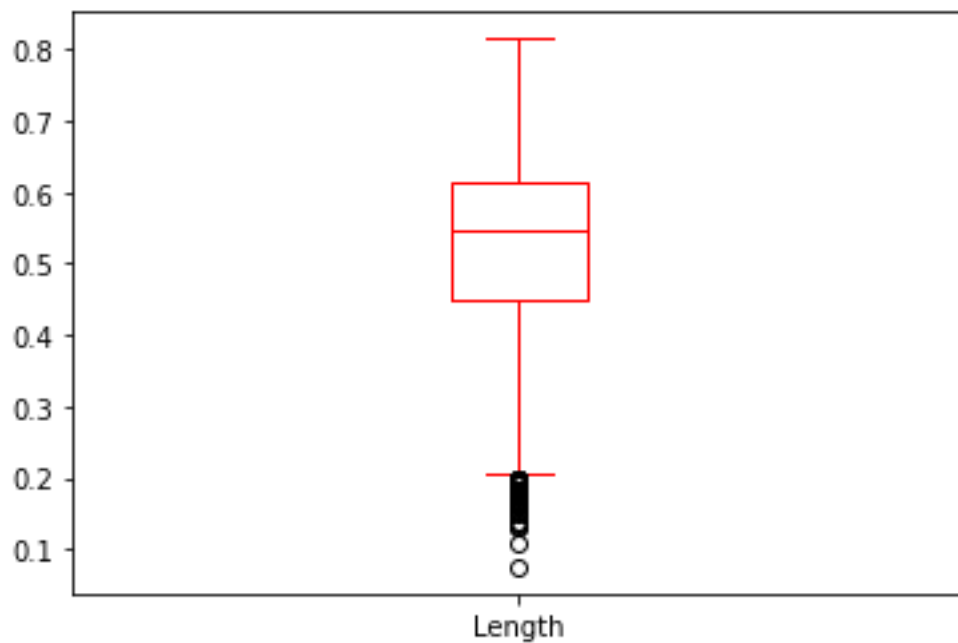
Out[]:



In []:

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

Out[]:

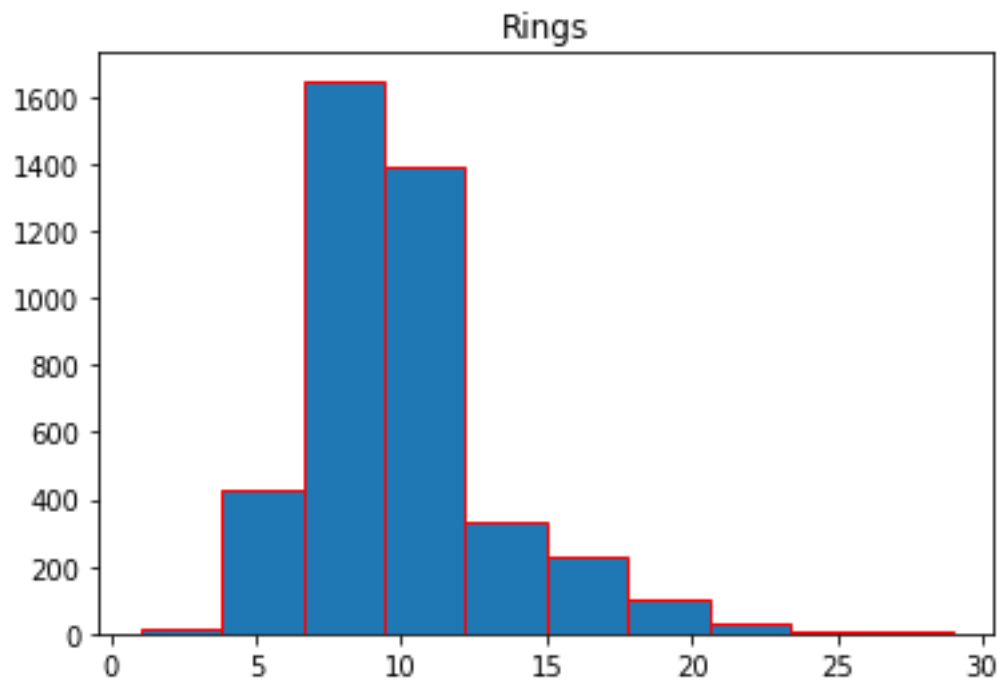


In []:

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

Out[]:

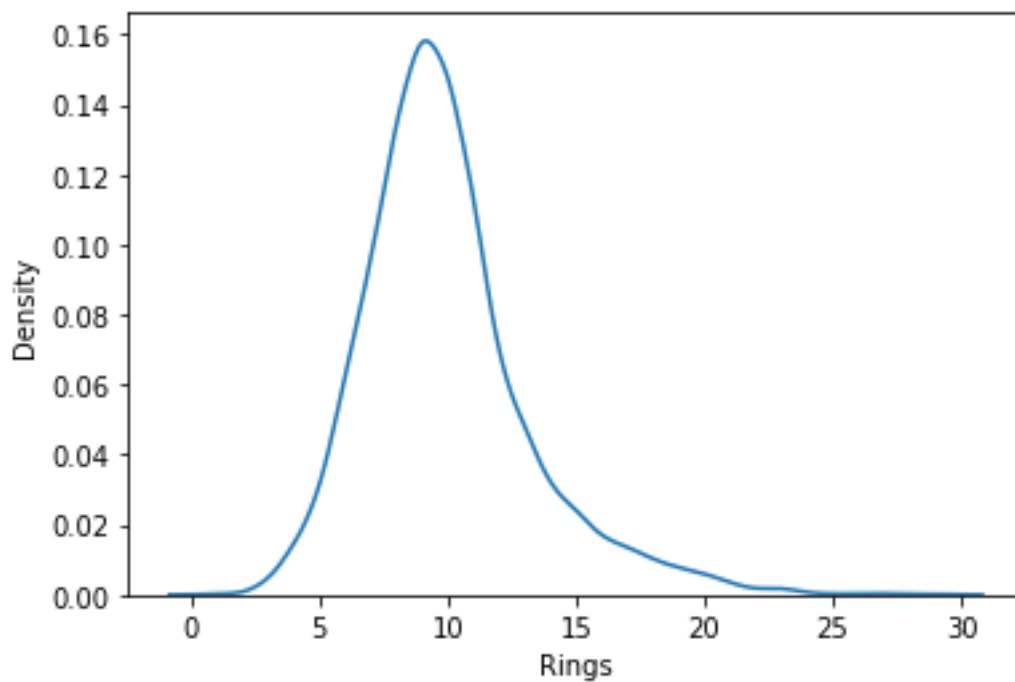
```
array([],
      dtype=object)
```



In []:

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

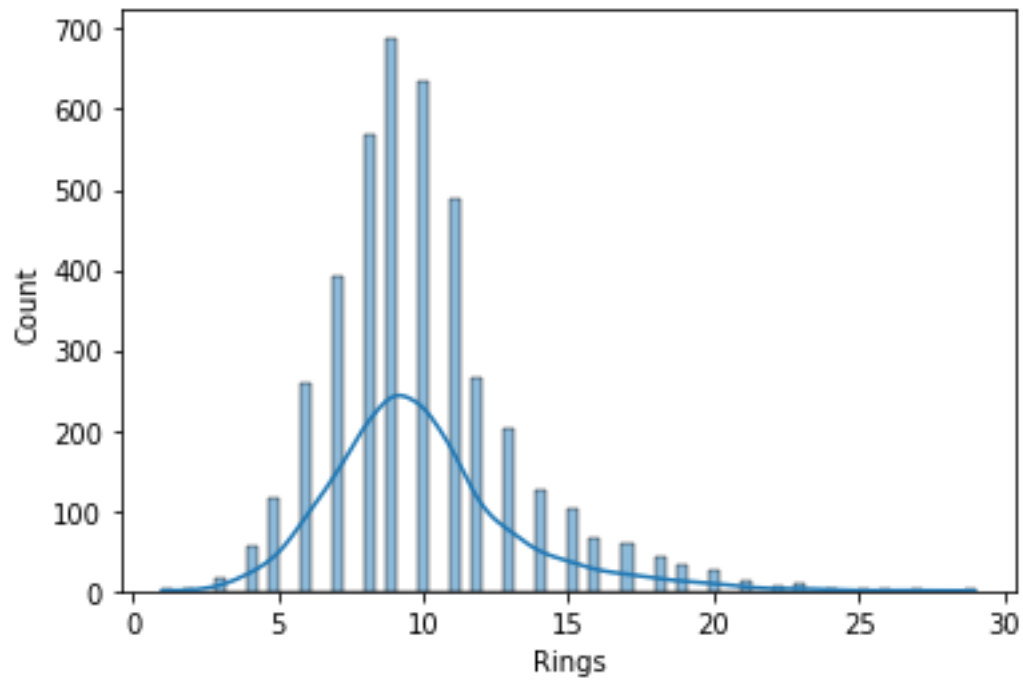
Out[]:



In []:

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

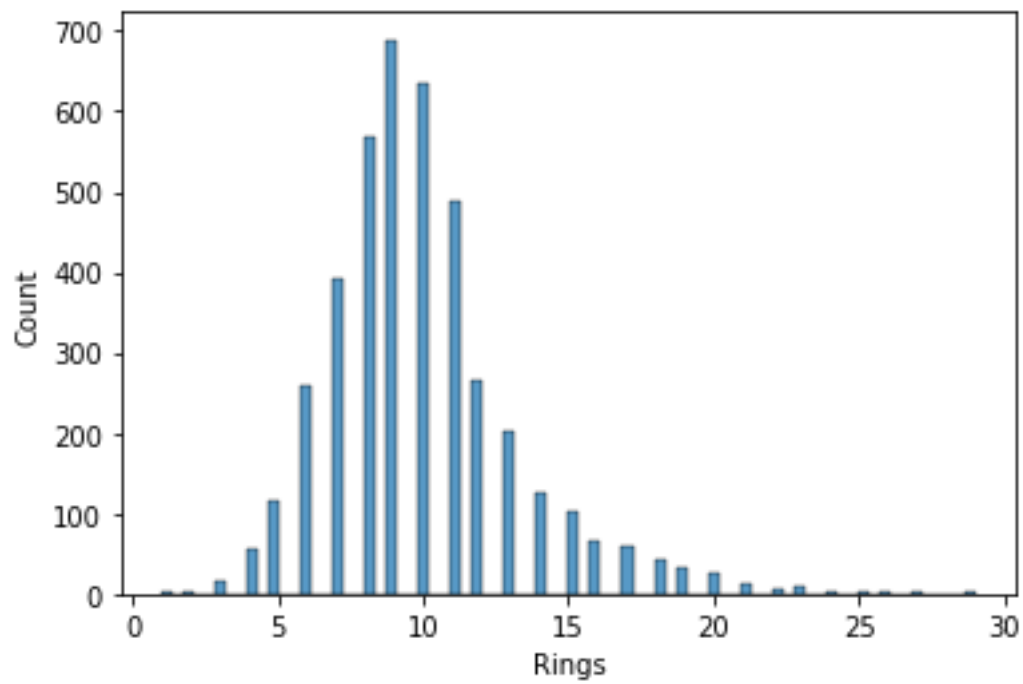
Out[]:



In []:

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

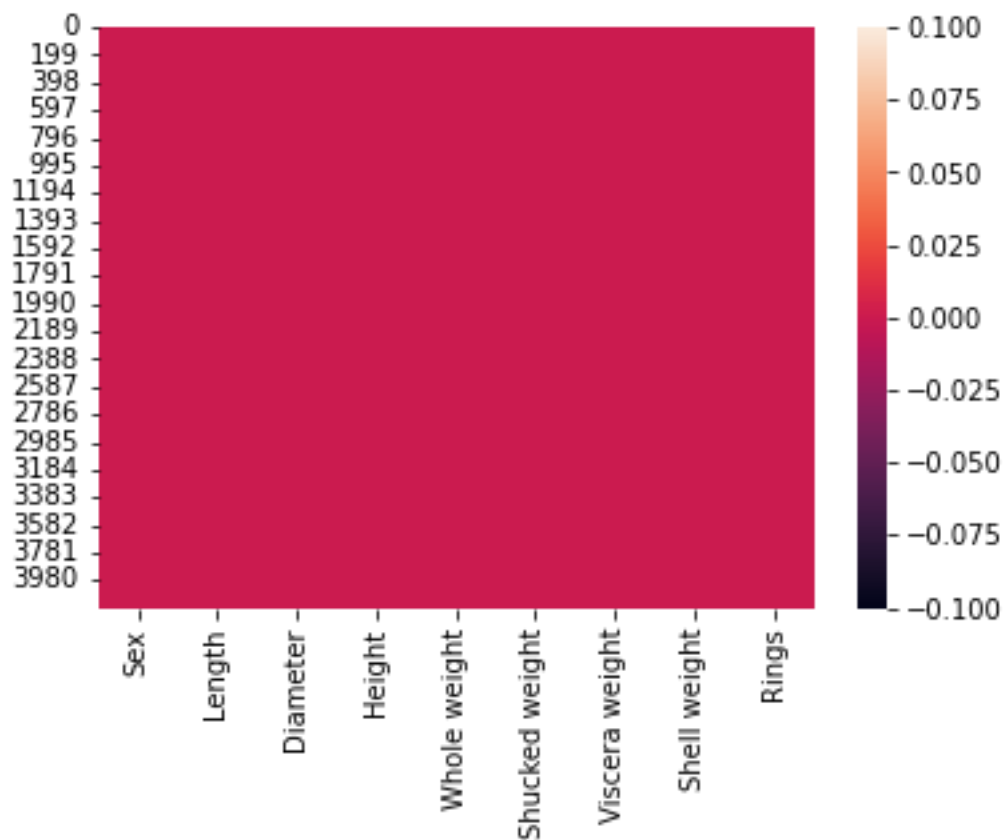
Out[]:



In []:

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

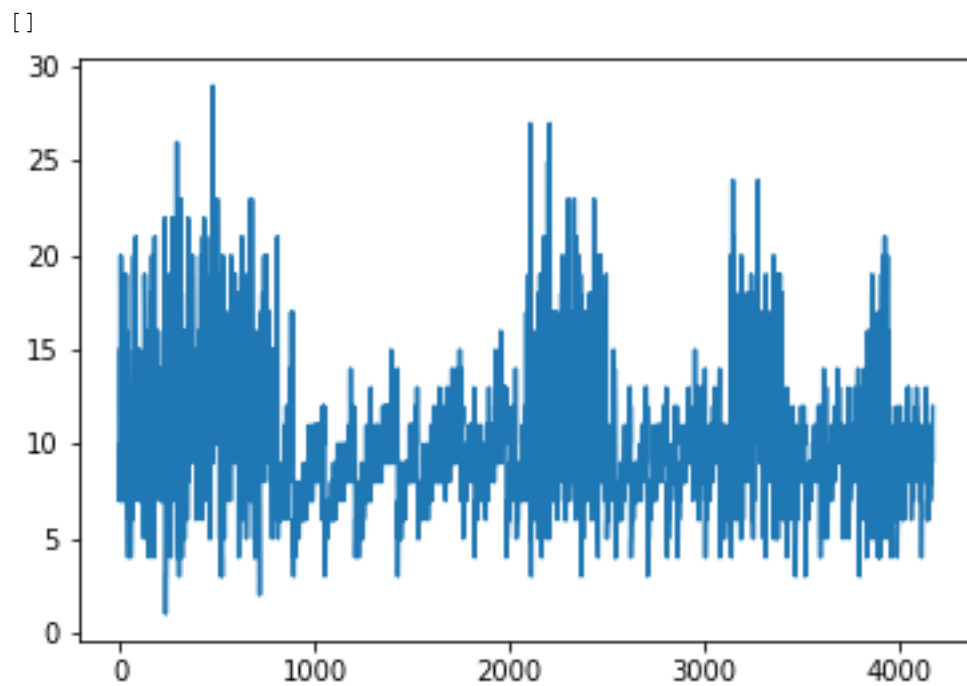
Out[]:



In []:

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

Out[]:

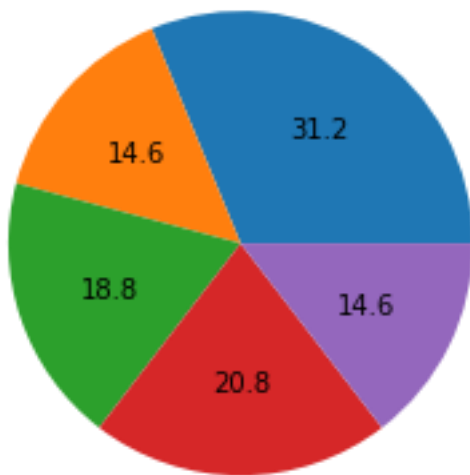


In []:

```
#piechart
plt.pie(df['Rings'].head(), autopct='%0.1f')
```

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')])
```

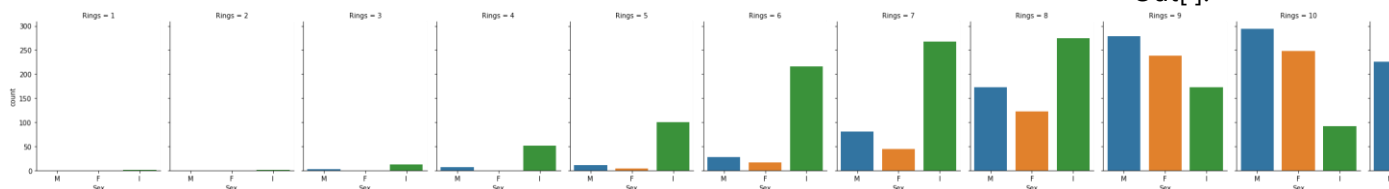


(ii) BIVARIATE

In []:

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

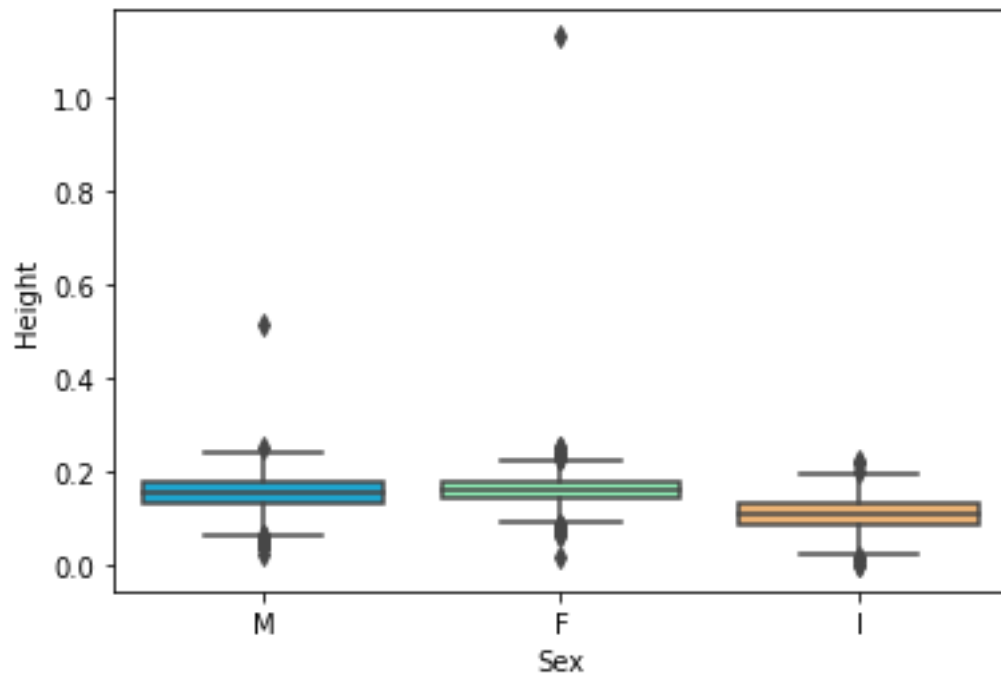
Out[]:



In []:

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

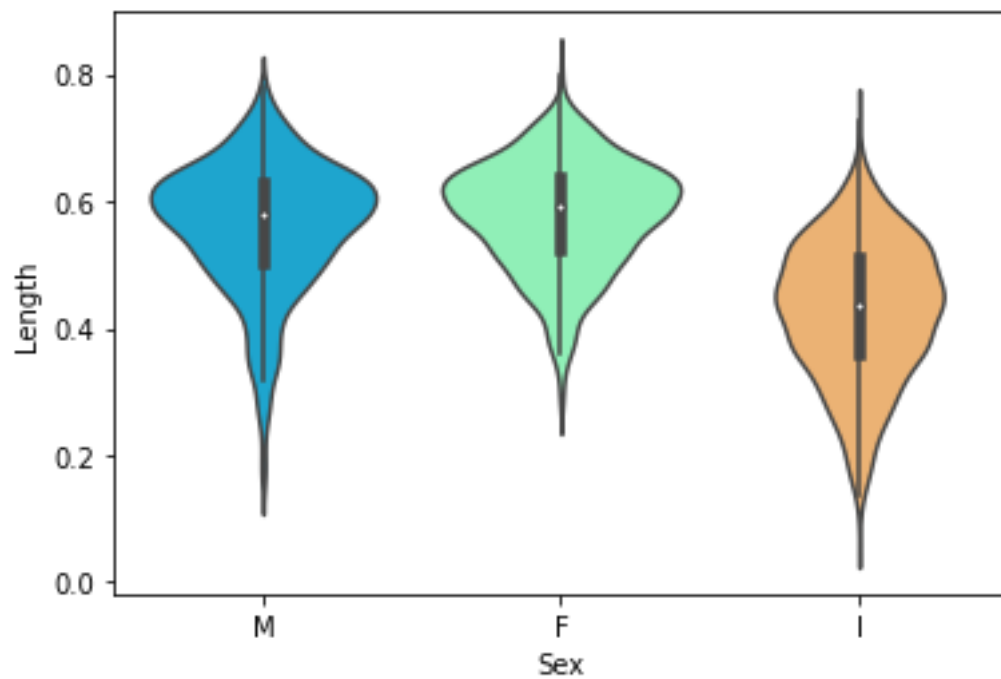
Out[]:



In []:

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

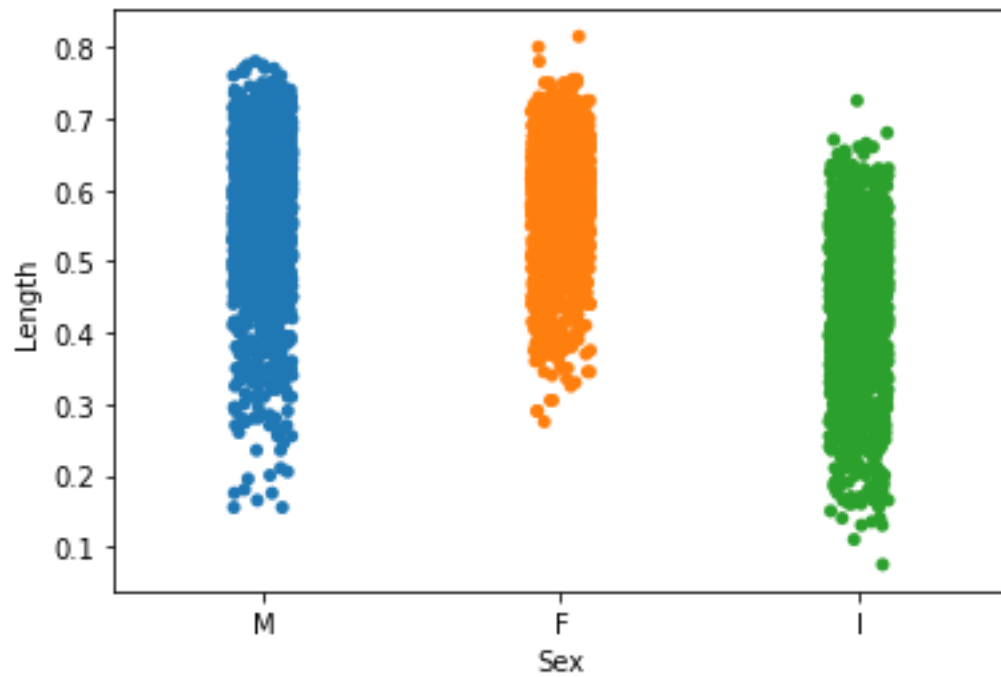
Out[]:



In []:

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

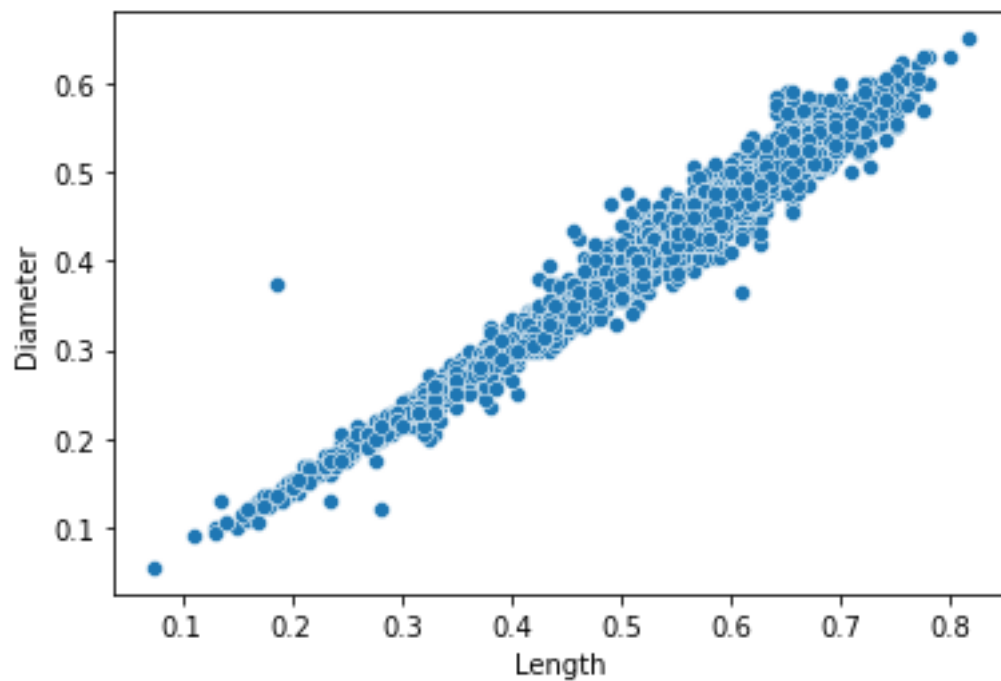
Out[]:



In []:

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

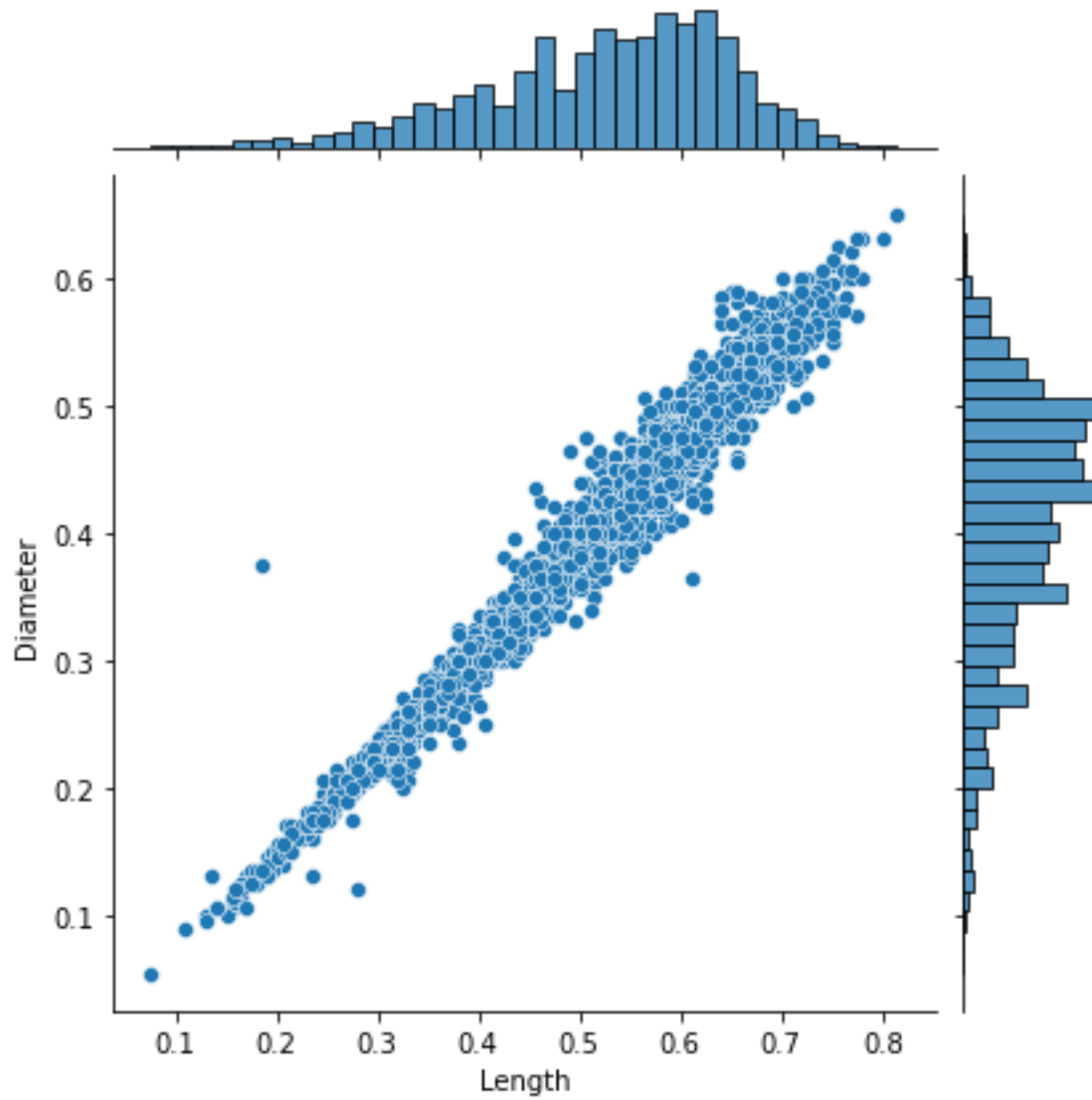
Out[]:



In []:

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

Out[]:



(III) MULTI-VARIATE

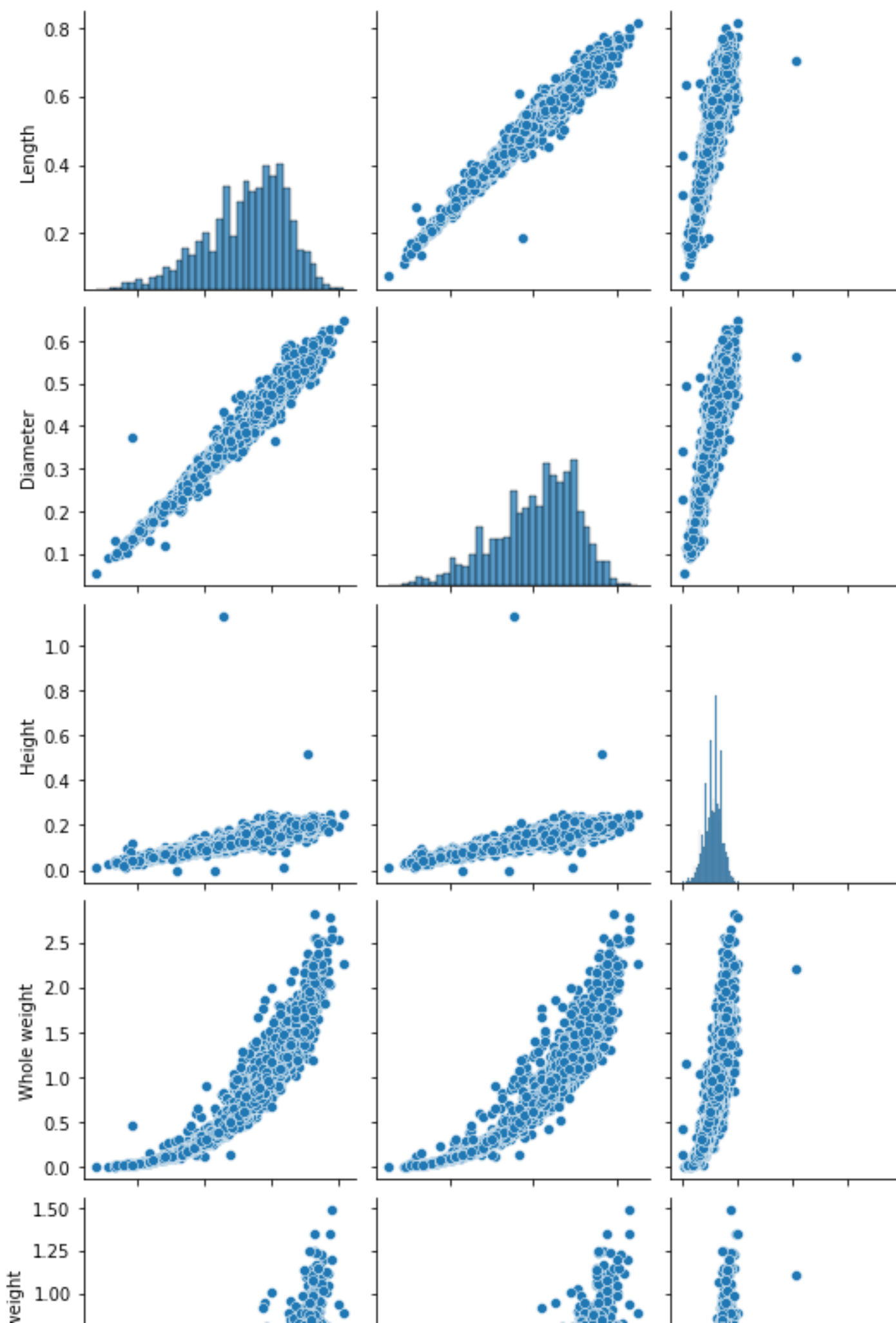
```
#Boxplot
fig, ax1 = plt.subplots(figsize=(8,5))
testPlot = sns.boxplot(ax=ax1, x='Length', y='Diameter', hue='Sex',
data=df)
```

In []:

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

In []:

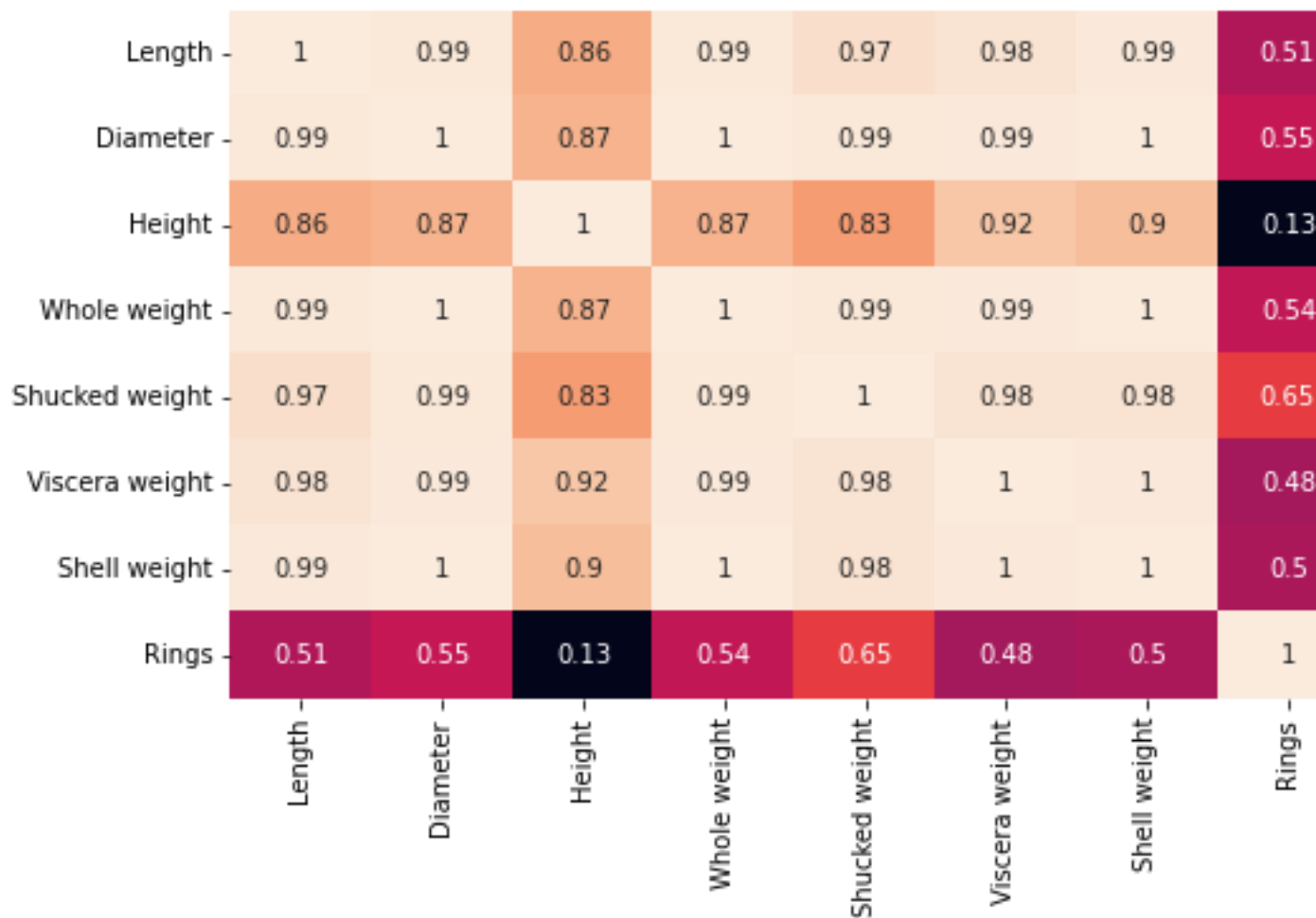
Out[]:



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

In []:

Out[]:



4. Perform descriptive statistics on the dataset

```
df
```

In []:

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
```

In []:

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

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
```

In []:

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

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

In []:

```
df.isnull()
```

Out[]:

[illegible]

	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
```

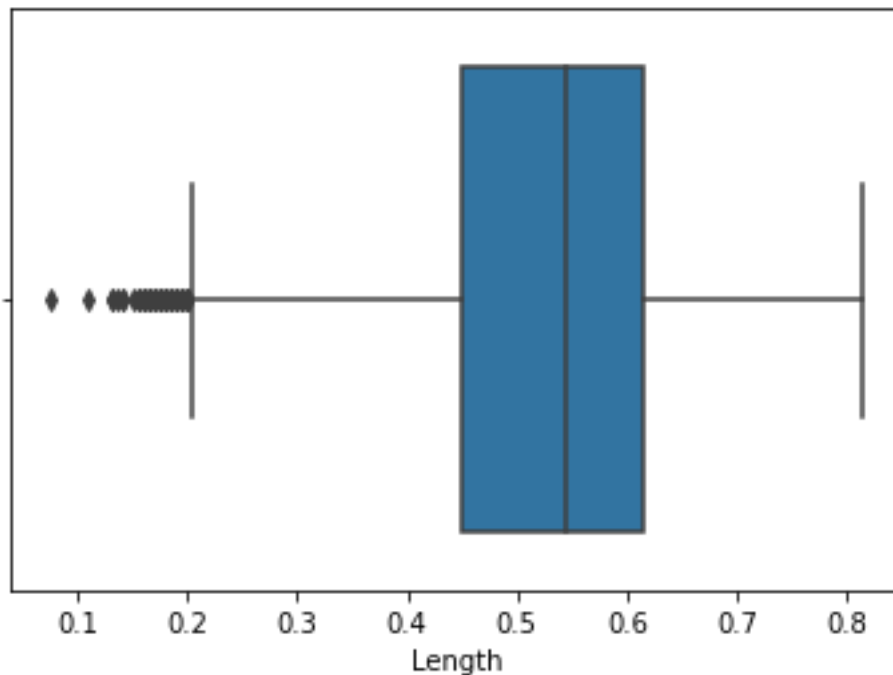
```
Rings          0  
dtype: int64
```

6. Find the outliers and replace them outliers

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

In []:

Out[]:



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

In []:

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[]:

```
(0.202500000000000004, 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 []:

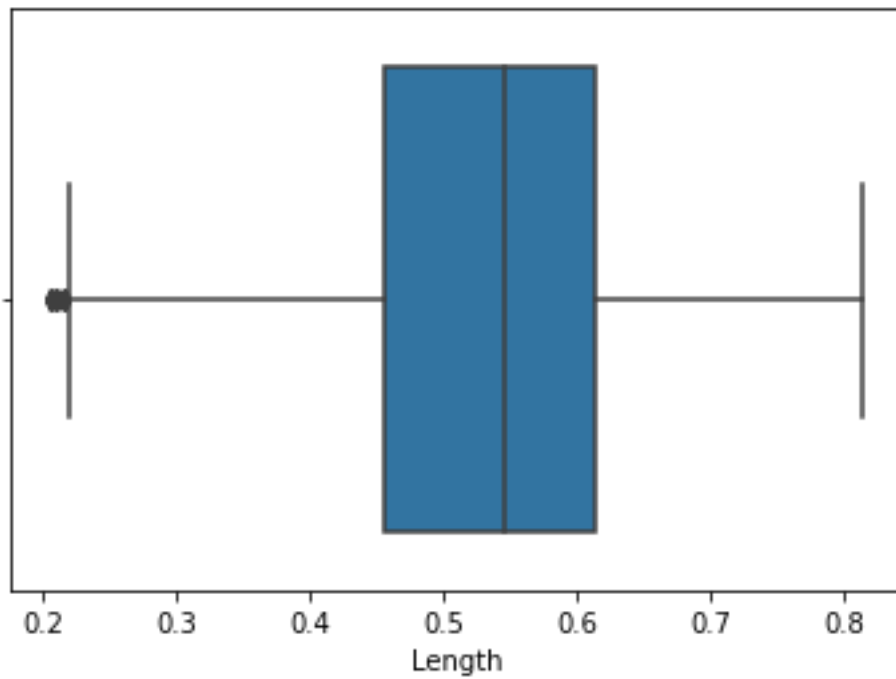
```
new_df = df.copy()  
new_df.loc[(new_df['Length']>upper_limit), 'Length'] = upper_limit
```

```
new_df.loc[(new_df['Length']<lower_limit), 'Length'] = lower_limit
```

In []:

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

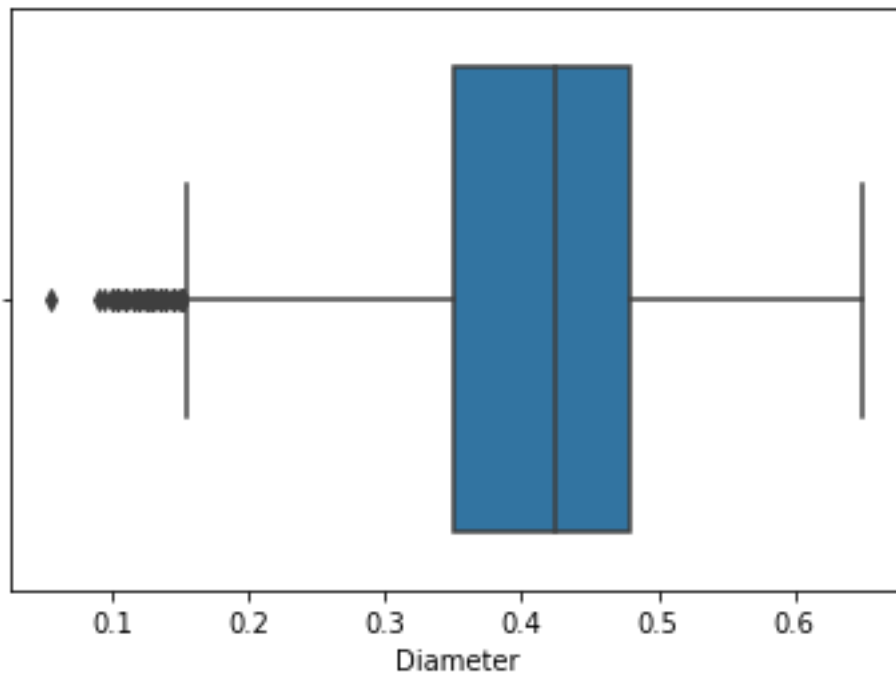
Out[]:



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

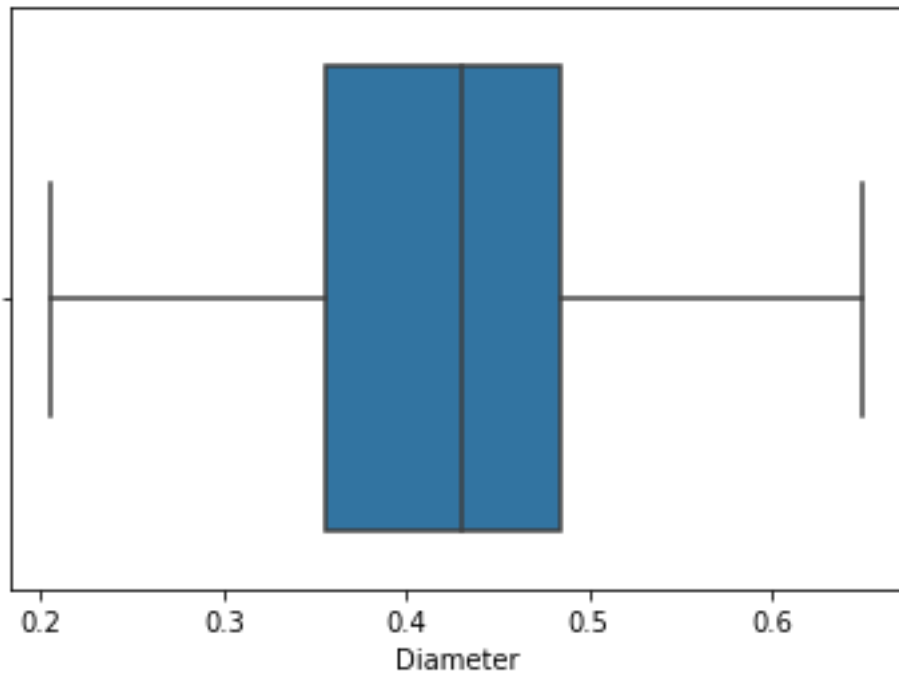
/usr/local/lib/python3.7/dist-packages/pandas/core/indexing.py:1817: Setting
WithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

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'])
```

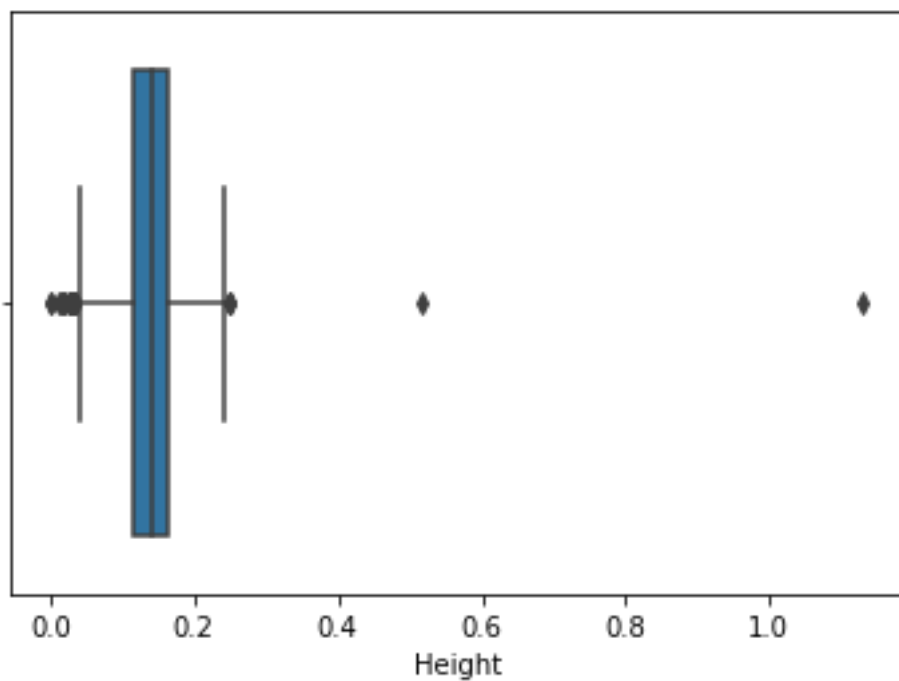
Out[]:



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 []:

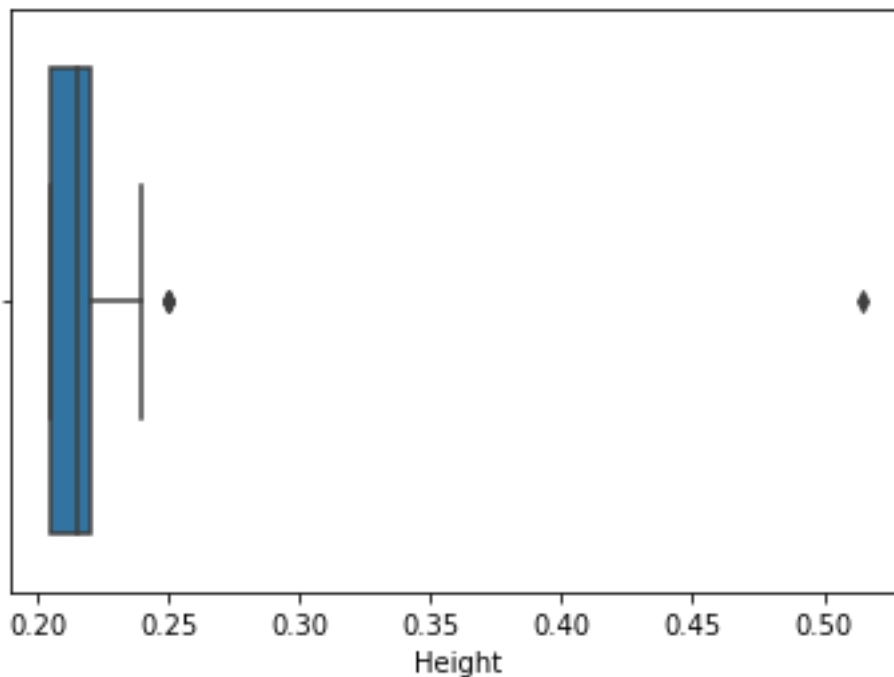
```
new_df = df.copy()
new_df.loc[(new_df['Height']>upper_limit), 'Height'] = upper_limit
new_df.loc[(new_df['Height']<lower_limit), 'Height'] = lower_limit
/usr/local/lib/python3.7/dist-packages/pandas/core/indexing.py:1817: Setting
WithCopyWarning:
A value is trying to be set on a copy of a slice from 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'])
```

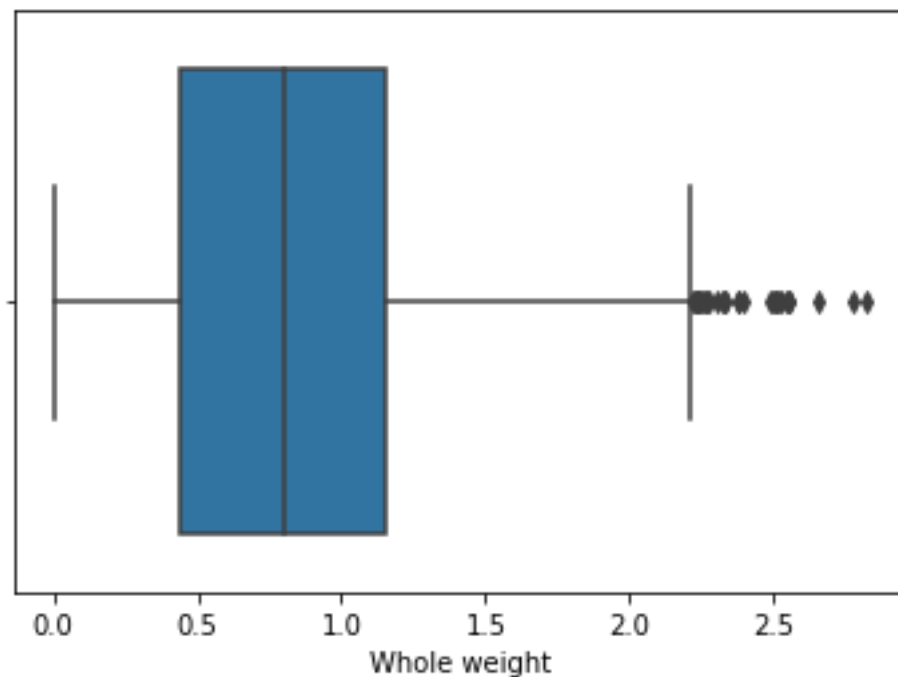
Out[]:



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: Setting
WithCopyWarning:
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
```

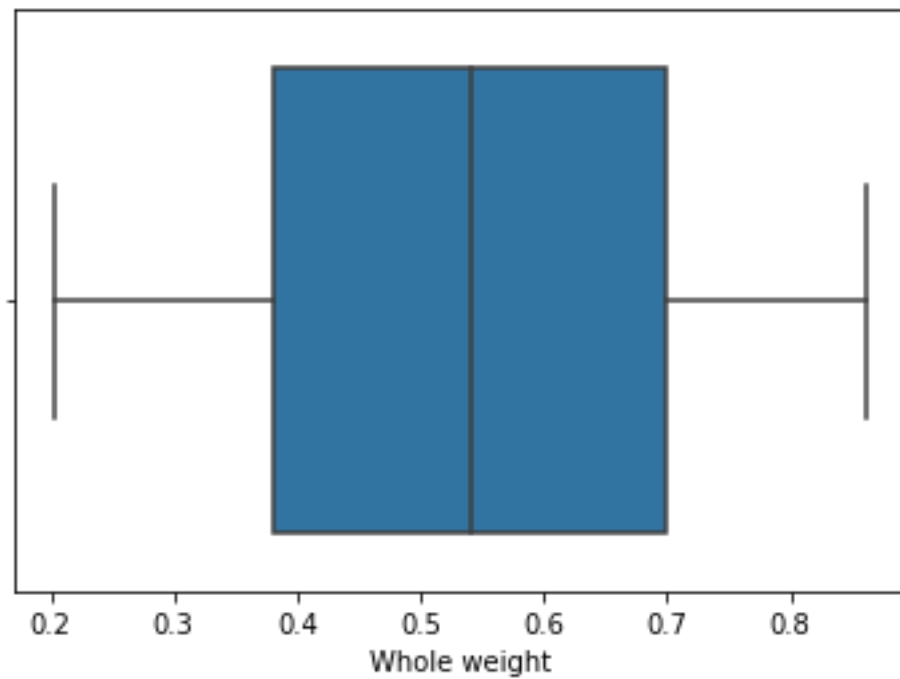
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy

```
self._setitem_single_column(loc, value, pi)
```

In []:

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

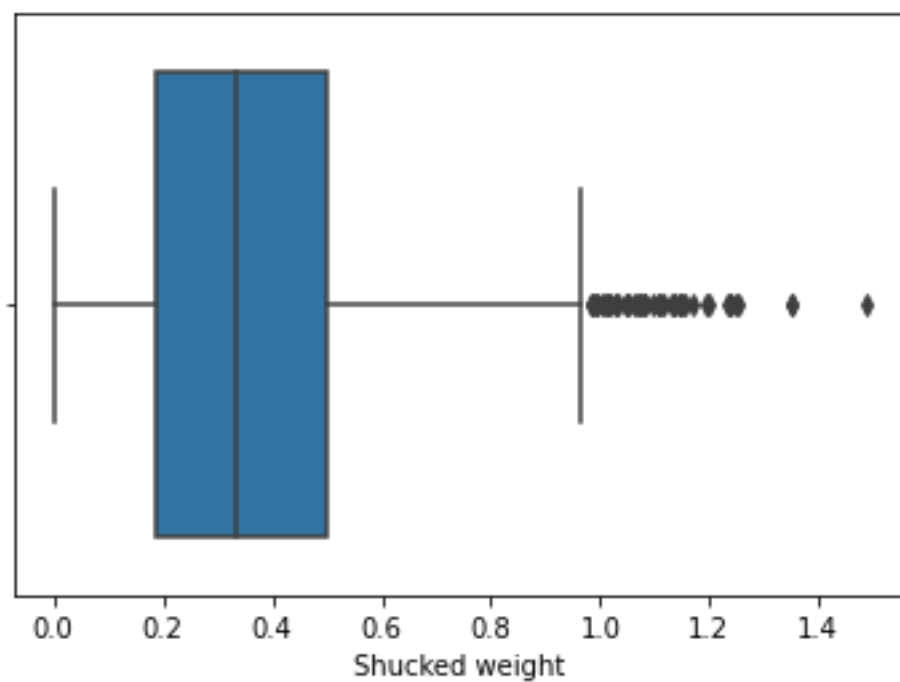
Out[]:



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

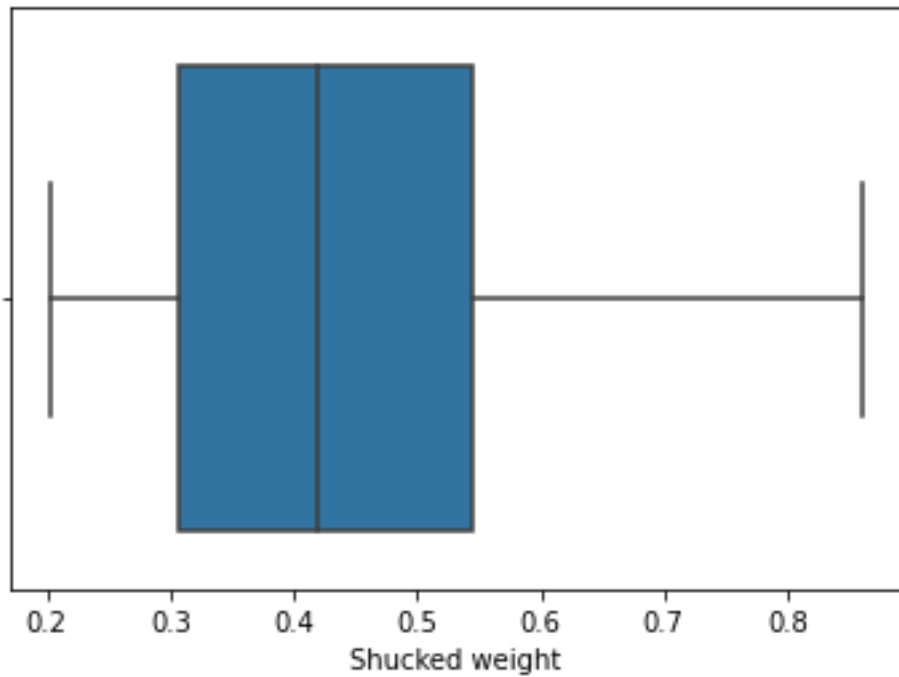
/usr/local/lib/python3.7/dist-packages/pandas/core/indexing.py:1817: Settin
gWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

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['Shucked weight'])
```

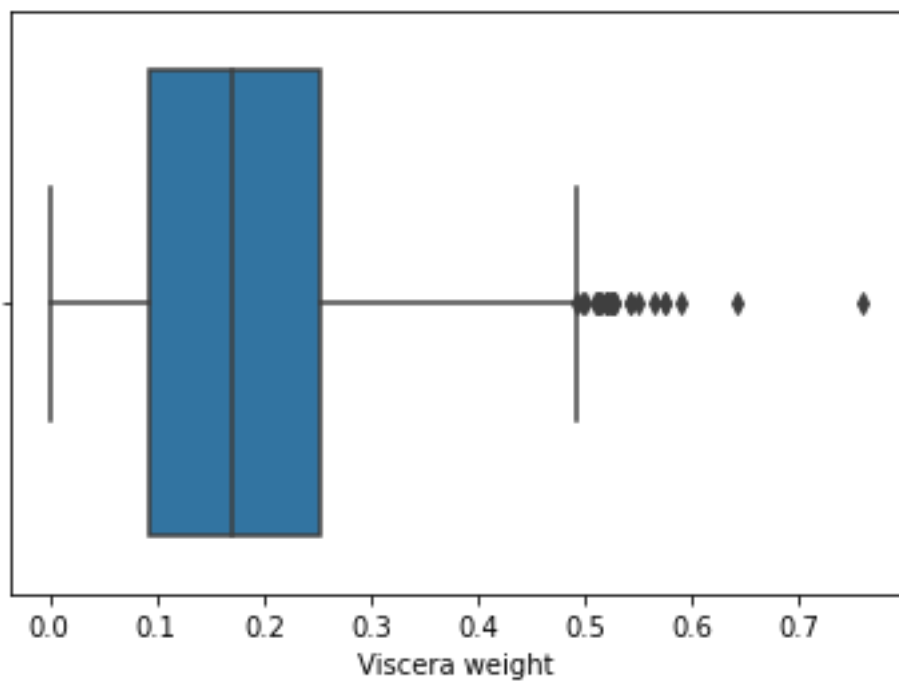
Out[]:



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

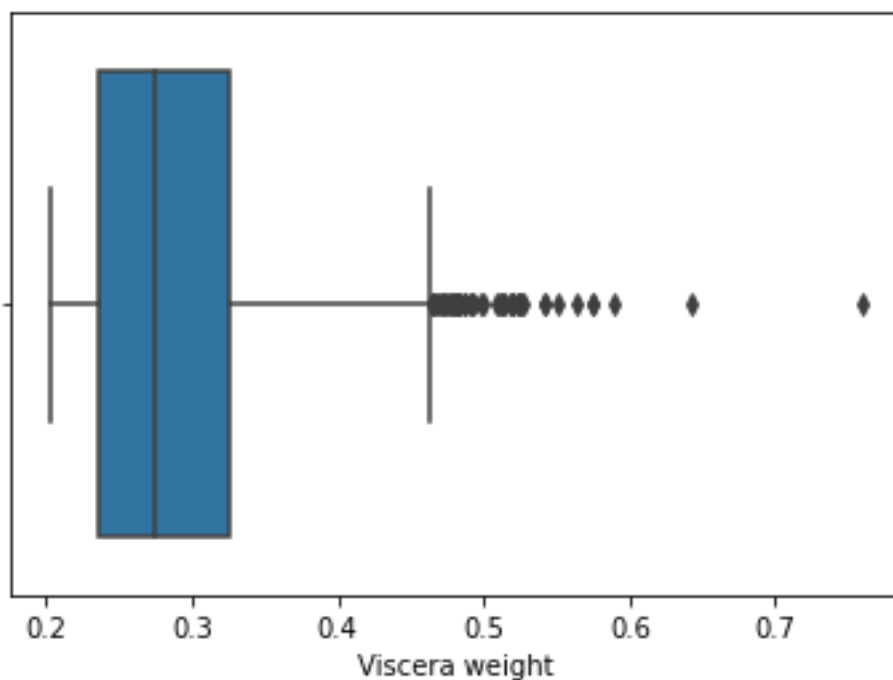
/usr/local/lib/python3.7/dist-packages/pandas/core/indexing.py:1817: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead
```

```
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['Viscera weight'])
```

Out[]:

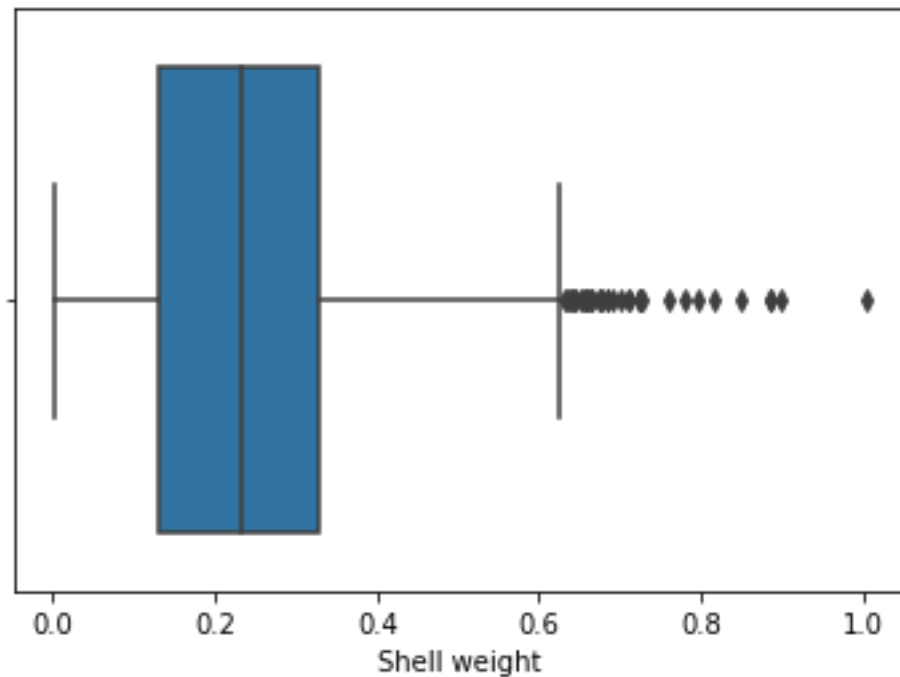


In []:

```
#shell weight
```

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

Out[]:



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

In []:

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

Out[]:

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

In []:

```
(0.202500000000000004, 0.8624999999999999)
```

Out[]:

```
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

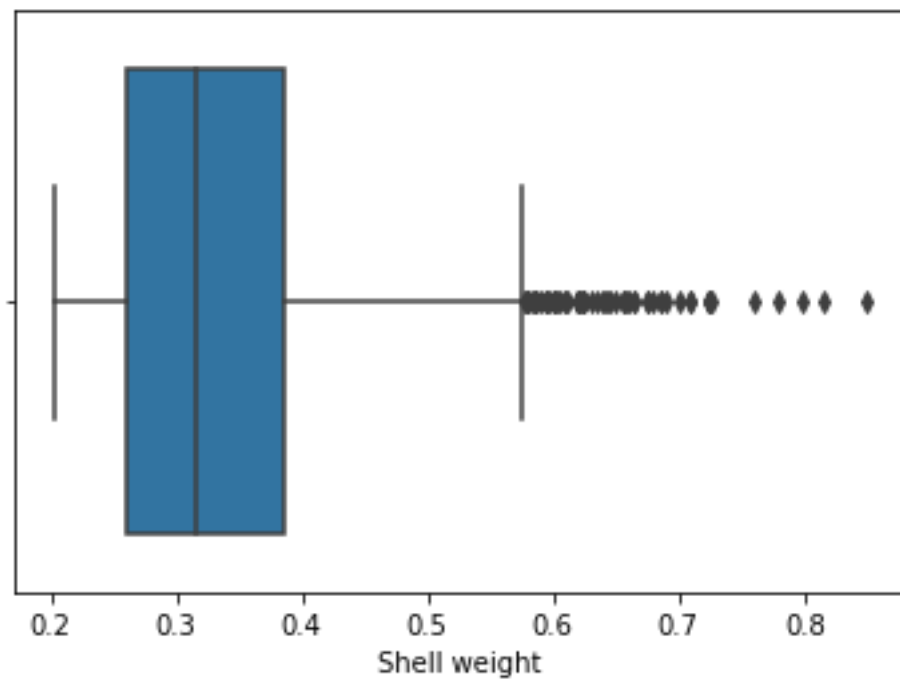
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy

```
self._setitem_single_column(loc, value, pi)
```

In []:

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

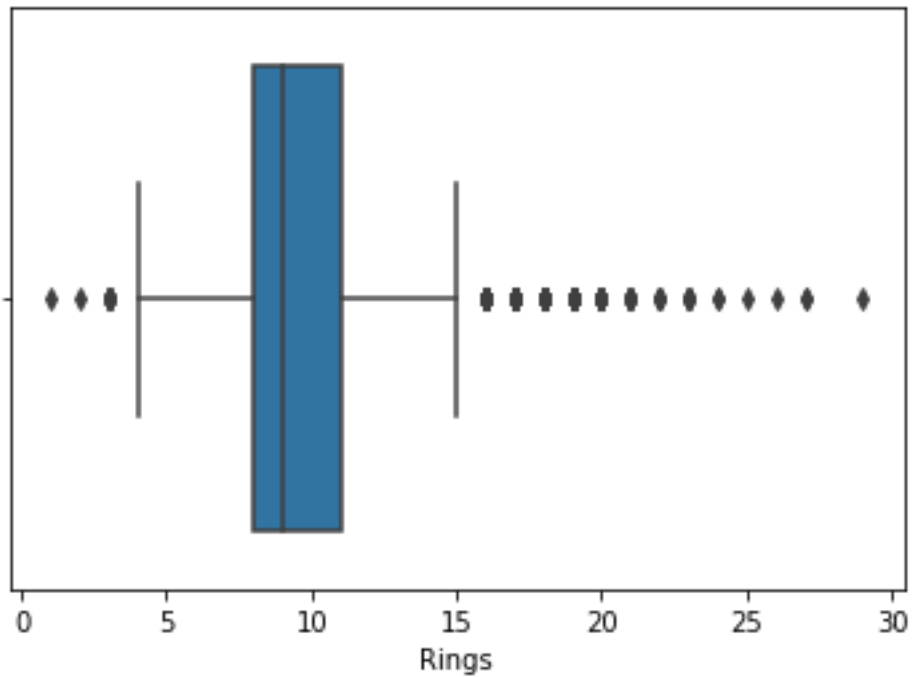
Out[]:



In []:

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

Out[]:



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

In []:

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

Out[]:

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

In []:

```
(0.202500000000000004, 0.8624999999999999)
```

Out[]:

```
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 []:

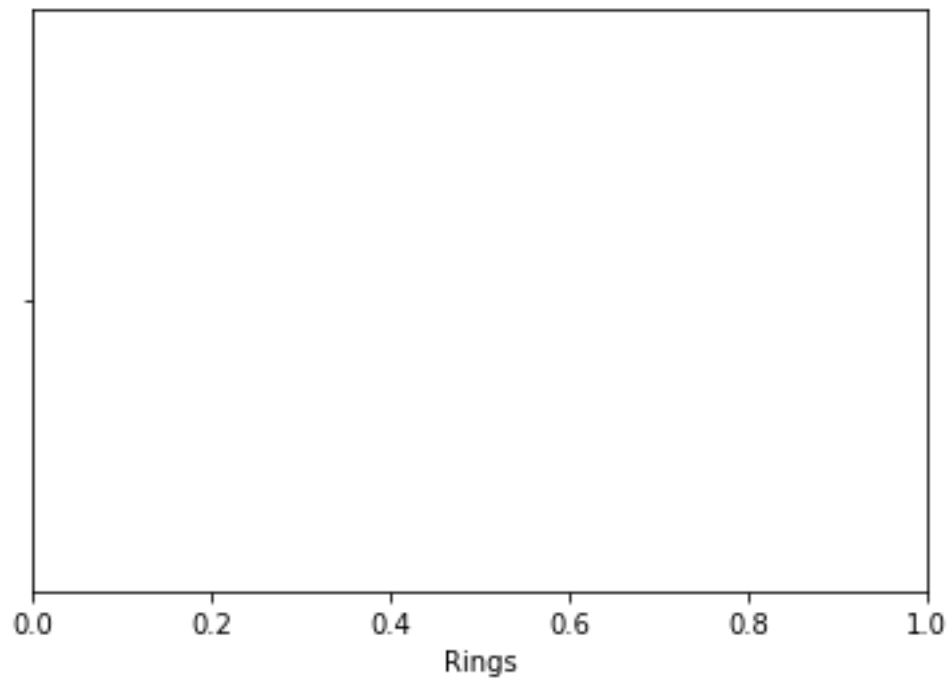
```
new_df = df.copy()
new_df.loc[(new_df['Rings']>upper_limit), 'Rings'] = upper_limit
new_df.loc[(new_df['Rings']<lower_limit), 'Rings'] = lower_limit
```

In []:

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

In [219]:

Out[219]:



7. Check for Categorical columns and perform encoding

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

In []:

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

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

```

	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
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

	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight
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

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

In []:

```
0      15
1       7
2       9
3      10
4       7
..
4172    11
4173    10
4174     9
4175    10
4176    12
```

Out[]:

Name: Rings, Length: 4177, dtype: int64

9. Scale the independent variables

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

In []:

```
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

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

In []:

```
In [ ]:
x_train, x_test, y_train, y_test = train_test_split(x,y, test_size = 0.2)

In [ ]:
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 [ ]:
from sklearn.linear_model import LinearRegression

In [ ]:
linearmodel = LinearRegression()
```

12. Train the Model

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

Out[ ]:
LinearRegression()
```

13. Test the Model

```
In [ ]:
y_train_pred = linearmodel.predict(x_train)
y_test_pred = linearmodel.predict(x_test)
```

```
In [ ]:
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,
       10.57414181, 11.41222985,  7.2778873 ,  6.70117135,  9.80118094,
        7.10712737,  6.50572362, 11.70751204,  8.76972122,  6.96654951,
        6.47886711,  6.93951927,  7.458727 , 11.01554321, 10.36006982,
        7.61624674,  9.16579141,  8.04761481, 12.29038942,  6.20248568,
       10.8963334 , 10.28472972, 14.8411801 , 13.2740434 ,  7.49258174,
       13.25230426, 10.63162313, 11.48753783,  9.12967209,  9.03018002,
       10.45067022,  8.82633807,  8.36808156,  8.1866462 , 10.24960421,
       10.21056392, 10.72719629,  9.7728592 ,  7.38834192,  8.51824041,
       10.75769027, 11.85414492, 12.87270151,  9.16270242,  9.16333672,
        7.67461895, 11.76874831, 10.16330431, 10.05005681,  9.51670308,
        9.62637671,  9.1479746 , 10.34868794, 10.84331115,  7.357215 ,
       12.41355527, 14.38136393, 11.14666246, 12.38530452,  7.54358072,
        9.40835468,  9.88315837, 12.11159133,  7.94462274,  7.47162288,
        8.90032367, 10.43913282,  9.07962681,  9.73850771,  5.83375835,
        8.36362646, 10.9307715 ,  6.82609483,  8.29060331,  9.69949734,
        8.89587123,  5.05902985, 12.66596704,  9.28710853, 12.5326091 ,
```

8.79671925, 9.49859385, 11.96474424, 11.52223318, 9.60091683,
8.53619584, 15.32591088, 14.00301768, 8.97484674, 13.05508044,
6.2030136 , 9.21207997, 7.936351 , 10.16988917, 11.47842671,
7.46631576, 9.75499741, 10.27332107, 9.23986585, 5.9322595 ,
12.28408611, 7.92151313, 15.25635443, 8.21425868, 5.09784281,
8.64249299, 11.10284085, 7.78901197, 8.83090234, 9.1646292 ,
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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)

```

```
0.4497726034378102
```

Out[]:

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 []:

```
#Measure 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
```