In [1]:
```
import pandas as pd import
    numpy as np import
    matplotlib.pyplot as plt import
    seaborn as sns import warnings
    warnings.filterwarnings('ignore')
```

1--Dataset Downloaded

2--Load the dataset

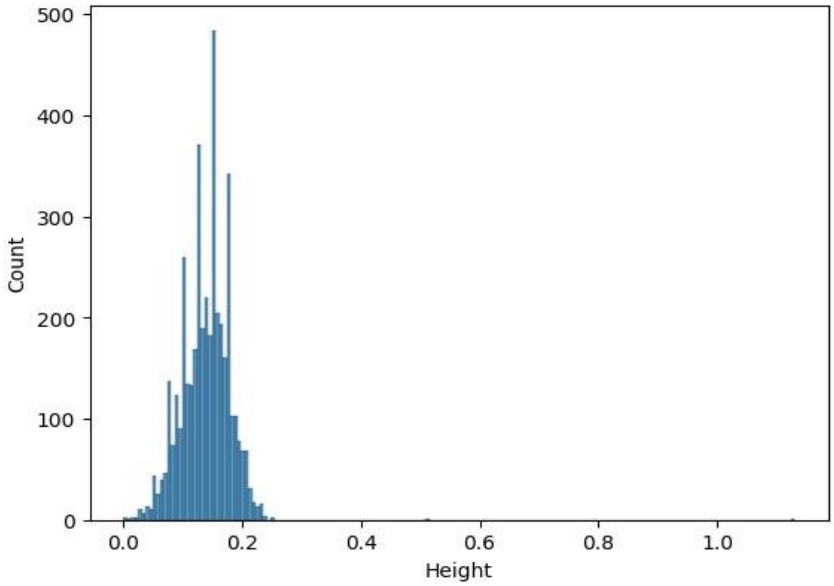
In [2]:
```
data=pd.read_csv(r"D:\Chrome_Downloads\abalone.csv")
```

3--Visuslization

Univariate analysis

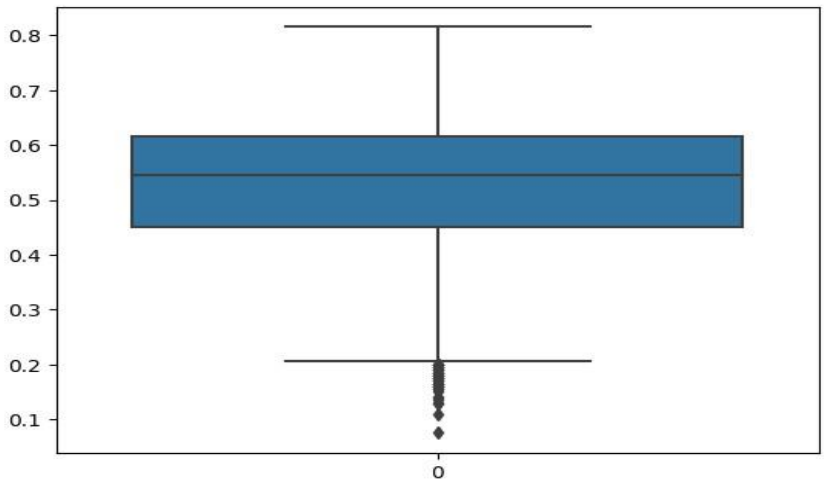
In [3]:
```
sns.histplot(data['Height'])
```

Out[3]:<AxesSubplot: xlabel='Height', ylabel='Count'>



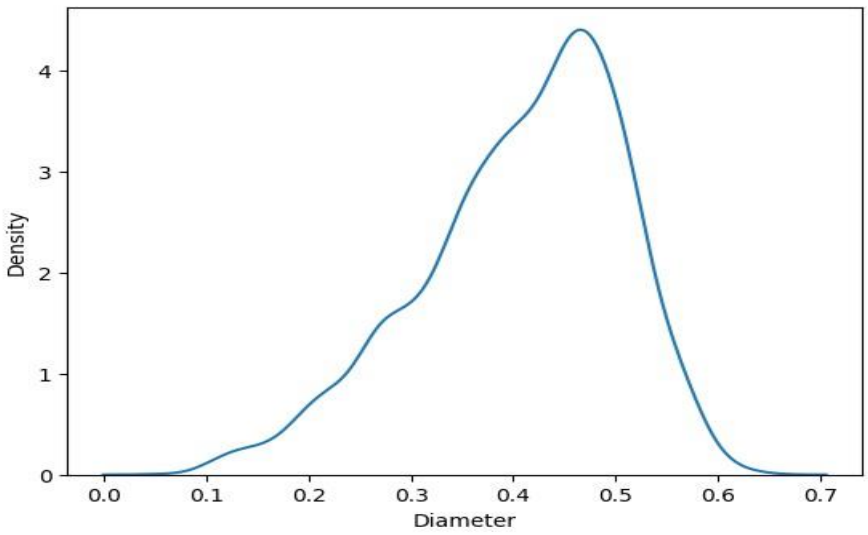
In [4]:
```
sns.boxplot(data['Length'])
```

Out[4]:<AxesSubplot: >



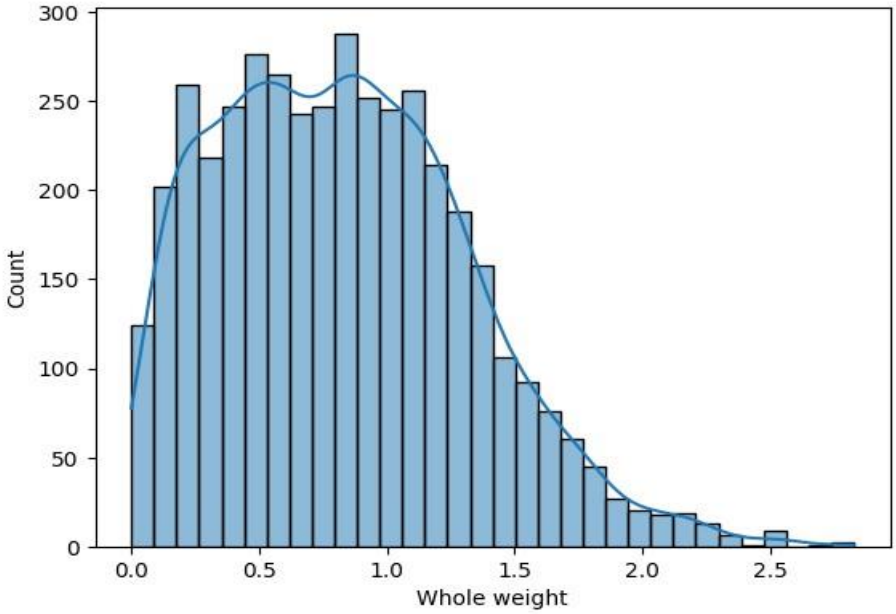
In [5]:sns.kdeplot(data['Diameter'])

Out[5]:<AxesSubplot: xlabel='Diameter', ylabel='Density'>



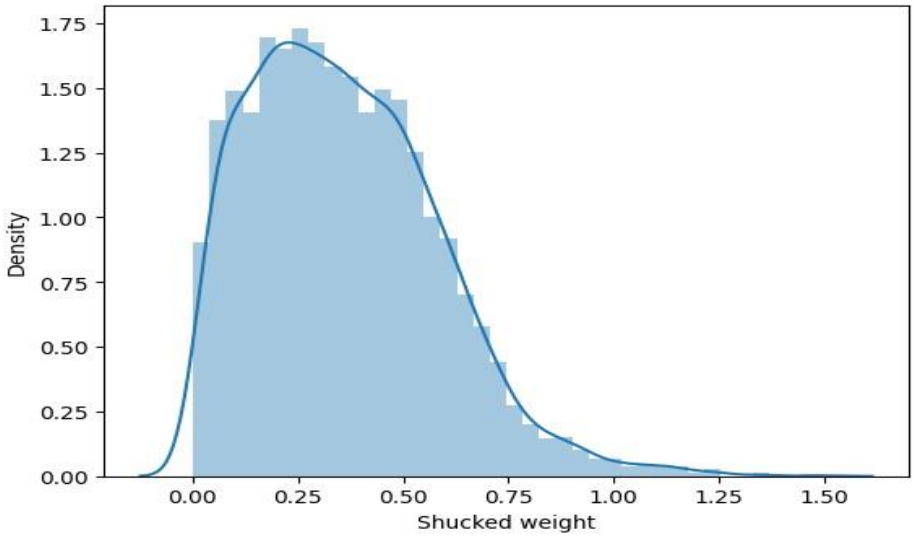
In [6]:sns.histplot(data['Whole weight'],kde=**True**)

Out[6]:<AxesSubplot: xlabel='Whole weight', ylabel='Count'>



In [7]:sns.distplot(data['Shucked weight'],kde=**True**)

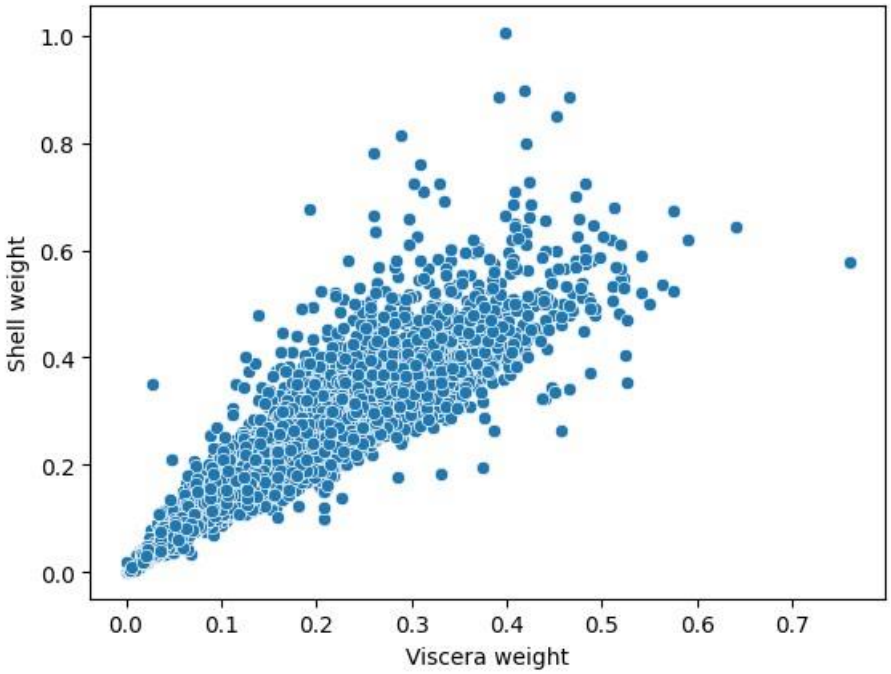
Out[7]:<AxesSubplot: xlabel='Shucked weight', ylabel='Density'>



Bivariate Analysis

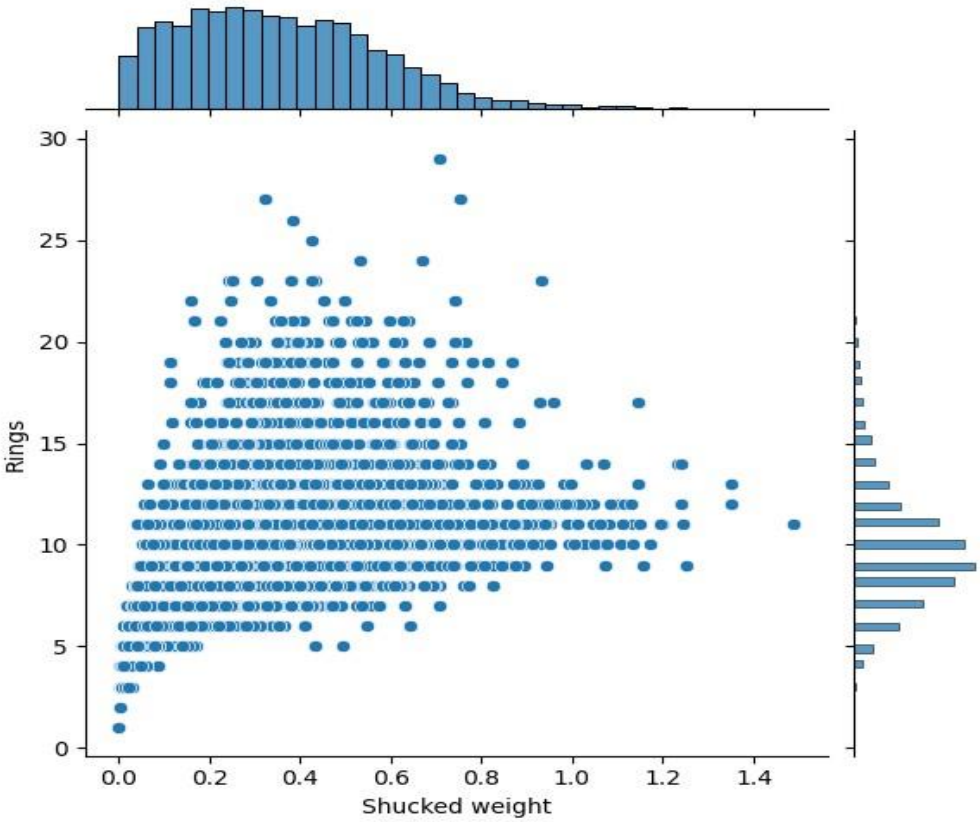
In [8]:sns.scatterplot(data=data,x='Viscera weight', y='Shell weight')

Out[8]:<AxesSubplot: xlabel='Viscera weight', ylabel='Shell weight'>



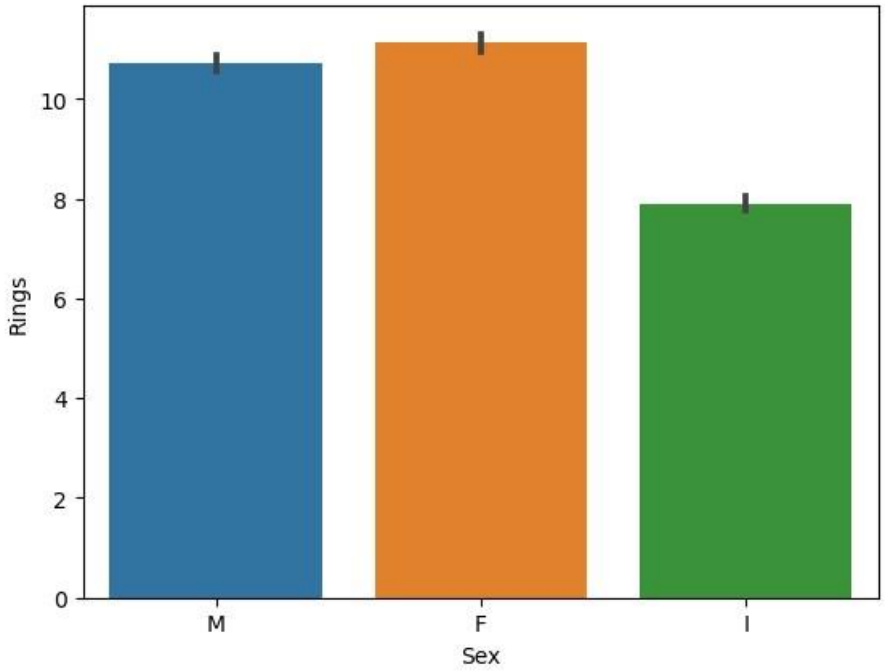
In [9]:sns.jointplot(data=data,x='Shucked weight',y='Rings')

Out[9]:<seaborn.axisgrid.JointGrid at 0x26d098190c0>



In [10]:sns.barplot(data=data,x='Sex',y='Rings')

Out[10]:<AxesSubplot: xlabel='Sex', ylabel='Rings'>



Multivariate Analysis

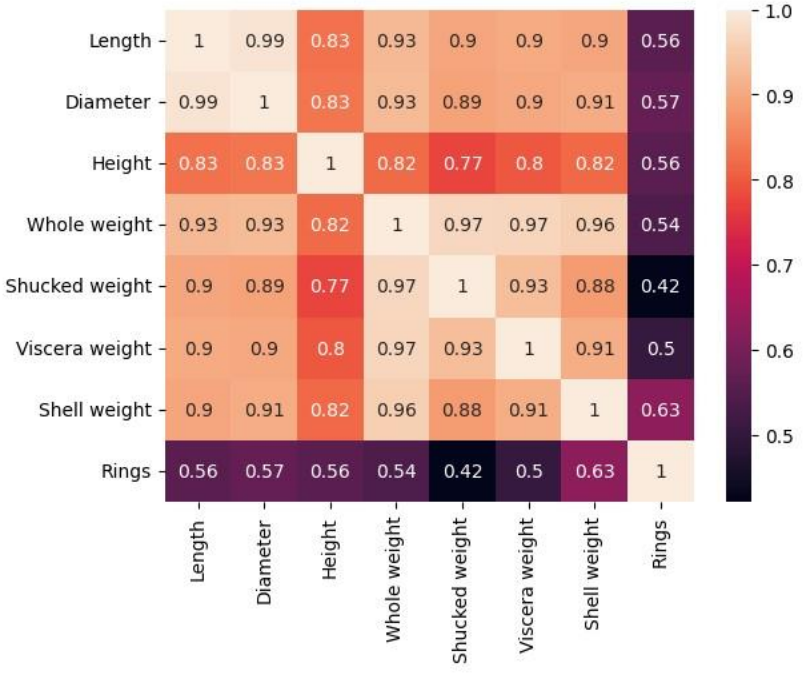
In [11]:data.corr()

Out[11]:

	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	Rings
Length	1.000000	0.986812	0.827554	0.925261	0.897914	0.903018	0.897706	0.556720
Diameter	0.986812	1.000000	0.833684	0.925452	0.893162	0.899724	0.905330	0.574660
Height	0.827554	0.833684	1.000000	0.819221	0.774972	0.798319	0.817338	0.557467
Whole weight	0.925261	0.925452	0.819221	1.000000	0.969405	0.966375	0.955355	0.540390
Shucked weight	0.897914	0.893162	0.774972	0.969405	1.000000	0.931961	0.882617	0.420884
Viscera weight	0.903018	0.899724	0.798319	0.966375	0.931961	1.000000	0.907656	0.503819
Shell weight	0.897706	0.905330	0.817338	0.955355	0.882617	0.907656	1.000000	0.627574
Rings	0.556720	0.574660	0.557467	0.540390	0.420884	0.503819	0.627574	1.000000

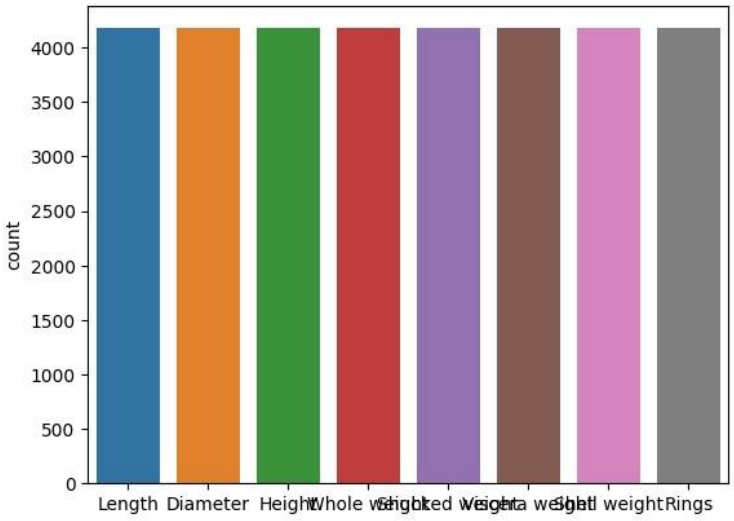
In [12]:sns.heatmap(data.corr(),annot=**True**)

Out[12]:<AxesSubplot: >



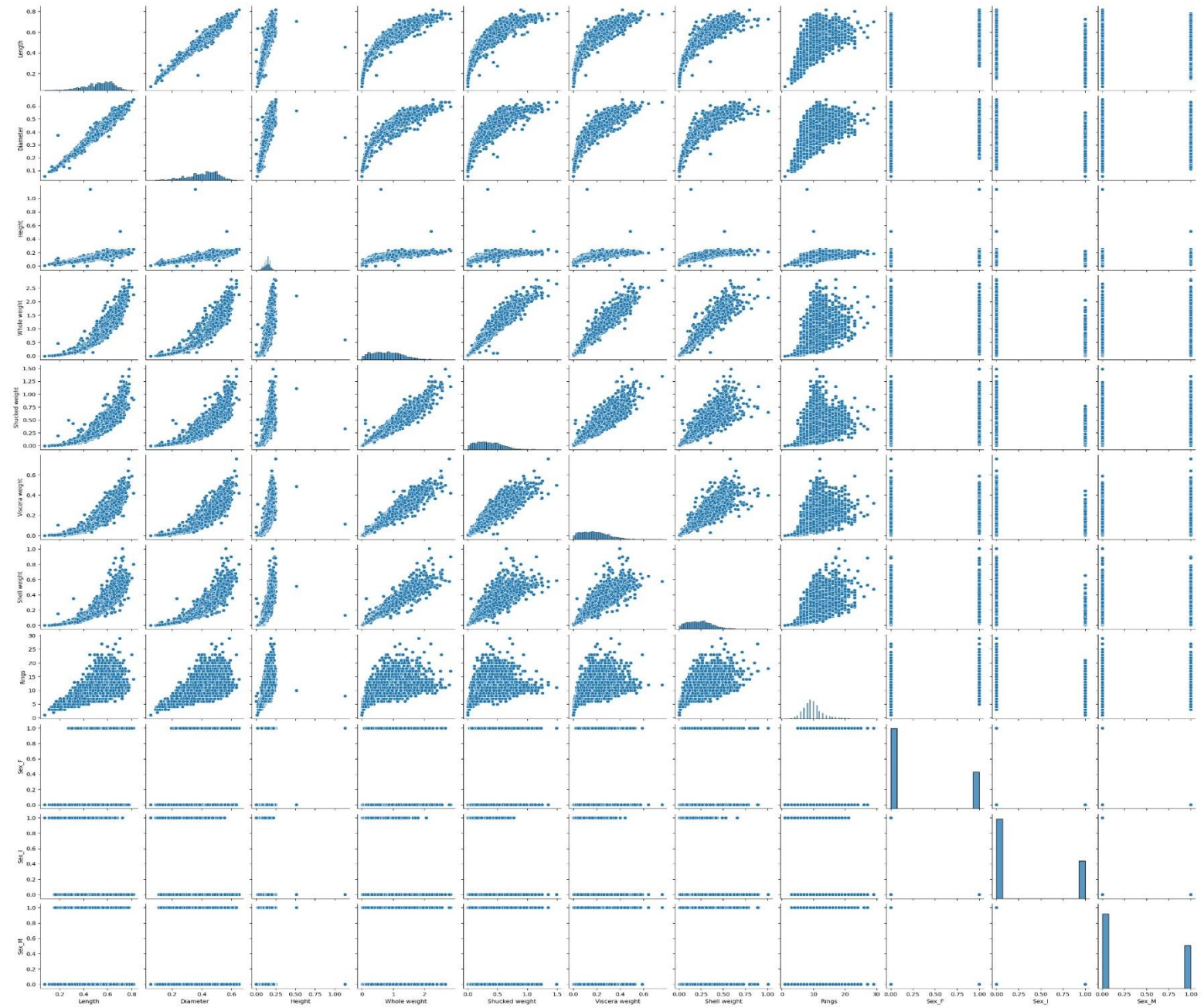
In [13]:sns.countplot(data)

Out[13]:<AxesSubplot: ylabel='count'>



In [13]:sns.pairplot(data)

Out[13]:<seaborn.axisgrid.PairGrid at 0x250074f0ee0>



4--Descriptive Statistics

In [14]:data.info()

```
<class 'pandas.core.frame.DataFrame'>
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 [15]:data.describe()

Out[15]:

	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

5--Handling missing values

In [16]:data.isna().sum()

```
Out[16]:Sex             0
        Length          0
        Diameter        0
        Height          0
        Whole weight    0
        Shucked weight  0
        Viscera weight  0
        Shell weight    0 Rings
        0 dtype: int64
```

There is no missing values in this dataset

6--Handling Outliers

IQR

In [3]:quant=data.quantile(q=[0.75,0.25]) quant

Out[3]:

	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	Rings
0.75	0.615	0.48	0.165	1.1530	0.502	0.2530	0.329	11.0
0.25	0.450	0.35	0.115	0.4415	0.186	0.0935	0.130	8.0

In [4]:iqr=quant.loc[0.75]-quant.loc[0.25] iqr

```
Out[4]: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

UPPER BOUND

In [5]:upper=quant.loc[0.75]+(1.5*iqr) print(upper)

Length 0.86250
Diameter 0.67500
Height 0.24000
Whole weight 2.22025
Shucked weight 0.97600
Viscera weight 0.49225
Shell weight 0.62750 Rings
15.50000 dtype: float64

LOWER BOUND

In [6]:lower=quant.loc[0.25]-(1.5*iqr) lower

Out[6]:Length 0.20250
Diameter 0.15500
Height 0.04000
Whole weight -0.62575
Shucked weight -0.28800
Viscera weight -0.14575
Shell weight -0.16850 Rings
3.50000 dtype: float64

MEAN

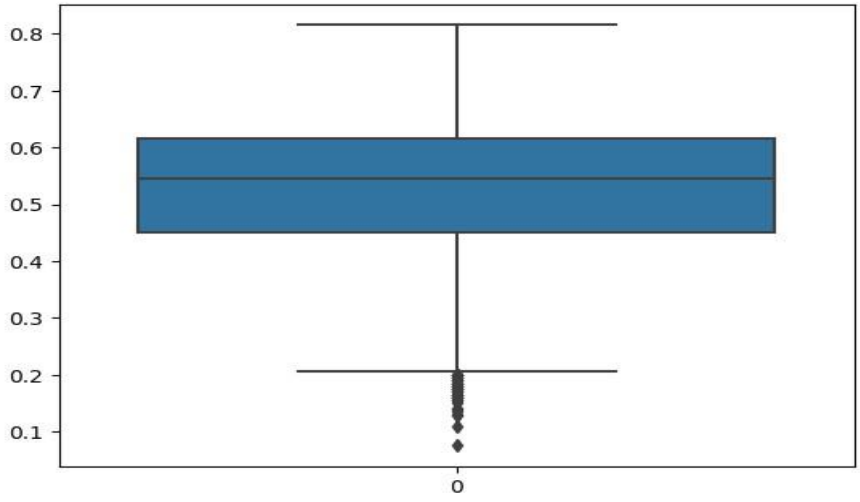
In [7]:data.mean()

Out[7]: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

Removing outlier

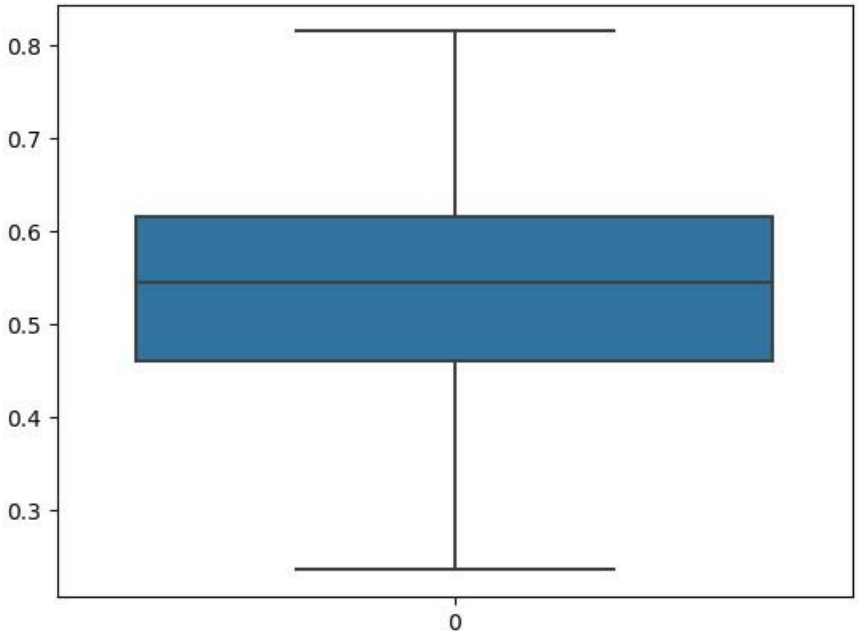
In [17]:sns.boxplot(data['Length'])

Out[17]:<AxesSubplot: >



In [18]:data['Length']=np.where(data['Length']< 0.23350,0.523992,data['Length']) sns.boxplot(data['Length'])

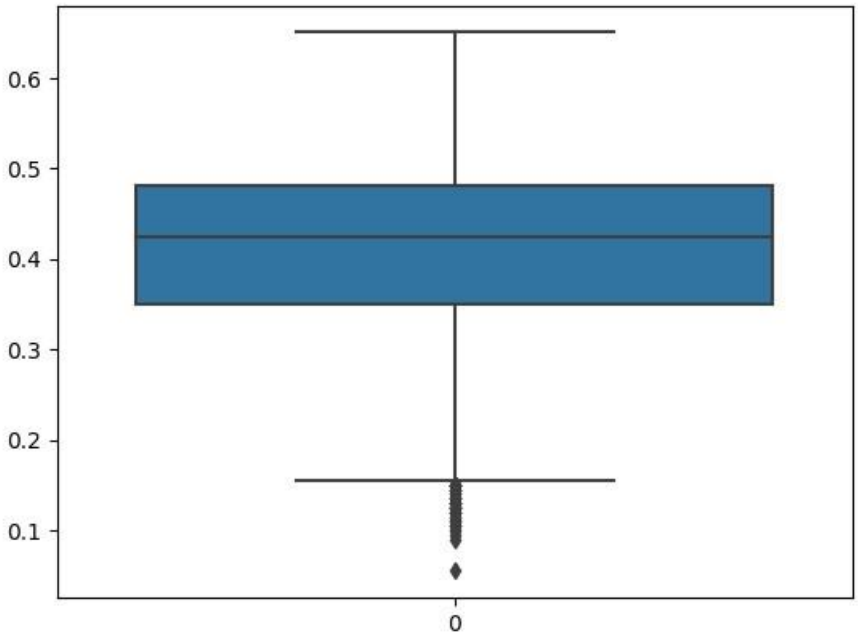
Out[18]:



Outlier removed for Length

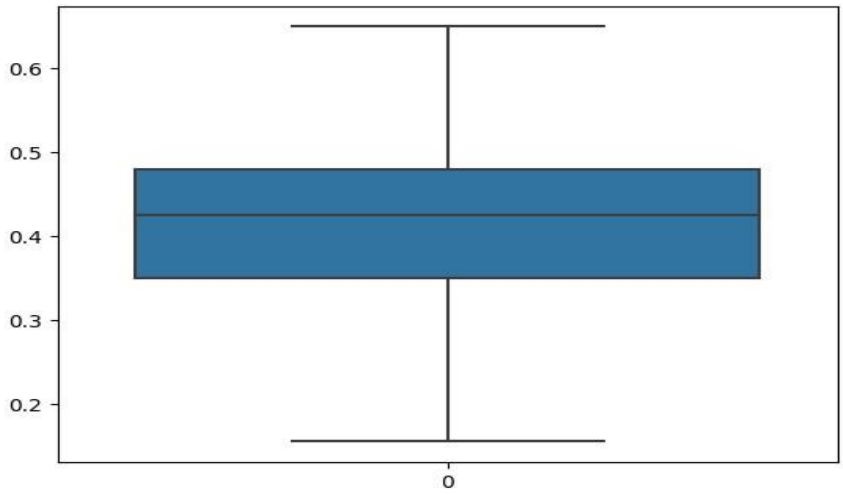
In [19]:sns.boxplot(data['Diameter'])

Out[19]:<AxesSubplot: >



In [20]:data['Diameter']=np.where(data['Diameter']<0.15500,0.407881,data['Diameter'])
sns.boxplot(data['Diameter'])

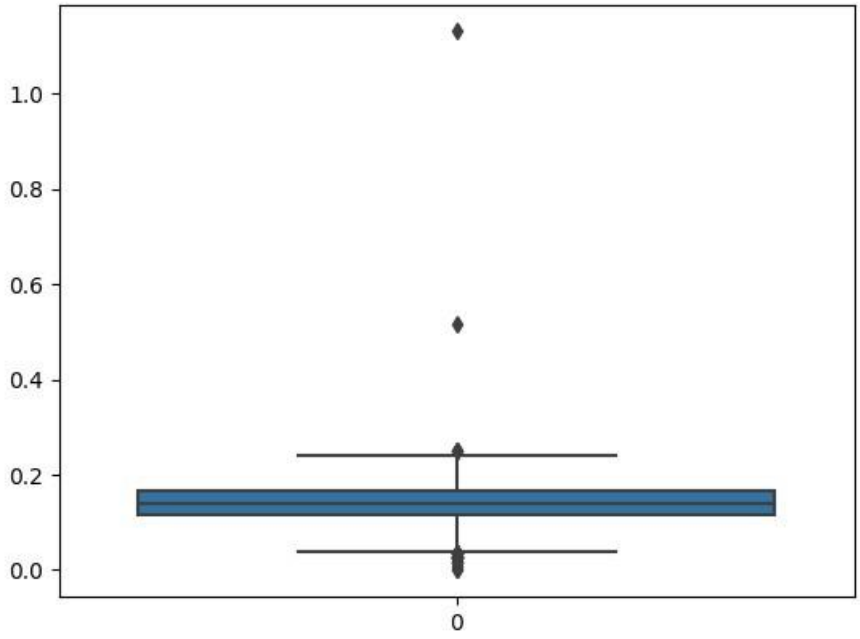
Out[20]:



Outlier removed for diameter

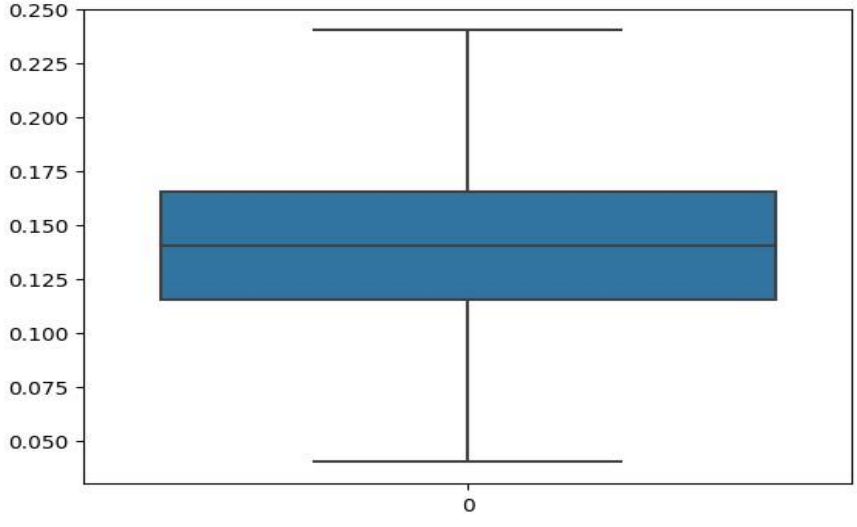
In [21]:sns.boxplot(data['Height'])

Out[21]:<AxesSubplot: >



In [22]:data['Height']=np.where(data['Height']< 0.04000,0.139516,data['Height'])
data['Height']=np.where(data['Height']> 0.24000,0.139516,data['Height']) sns.boxplot(data['Height'])

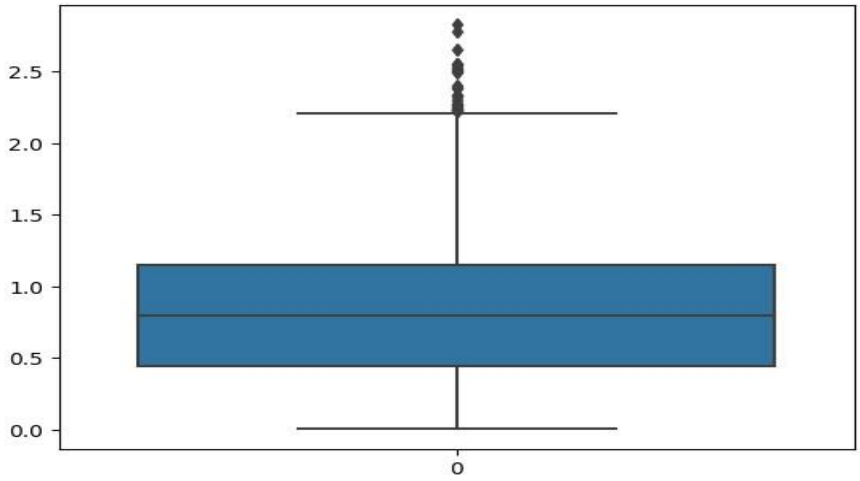
Out[22]:



Outlier removed for Height

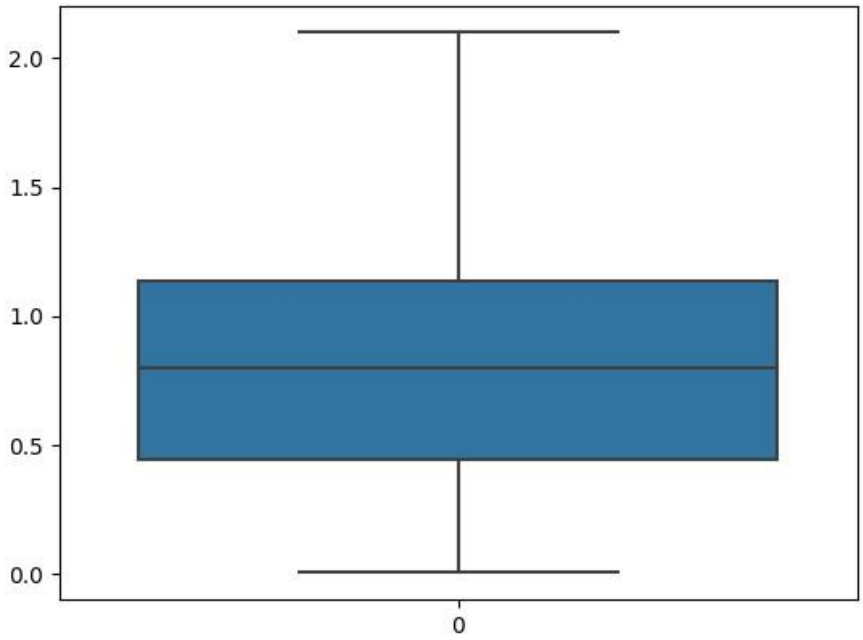
In [23]:sns.boxplot(data['Whole weight'])

Out[23]:<AxesSubplot: >



In [24]:data['Whole weight']=np.where(data['Whole weight']> 2.10022,0.828742,data['Whole weight'])
sns.boxplot(data['Whole weight'])

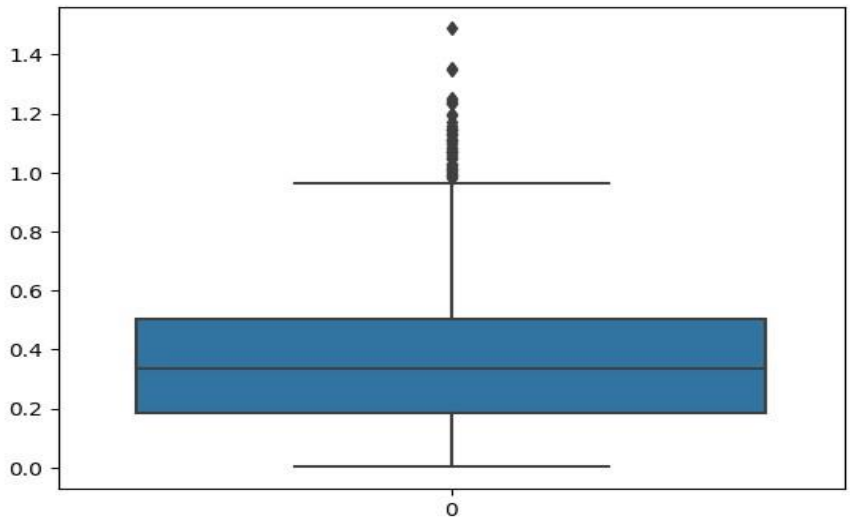
Out[24]:



In [25]:*# Outlier removed for Whole weightt*

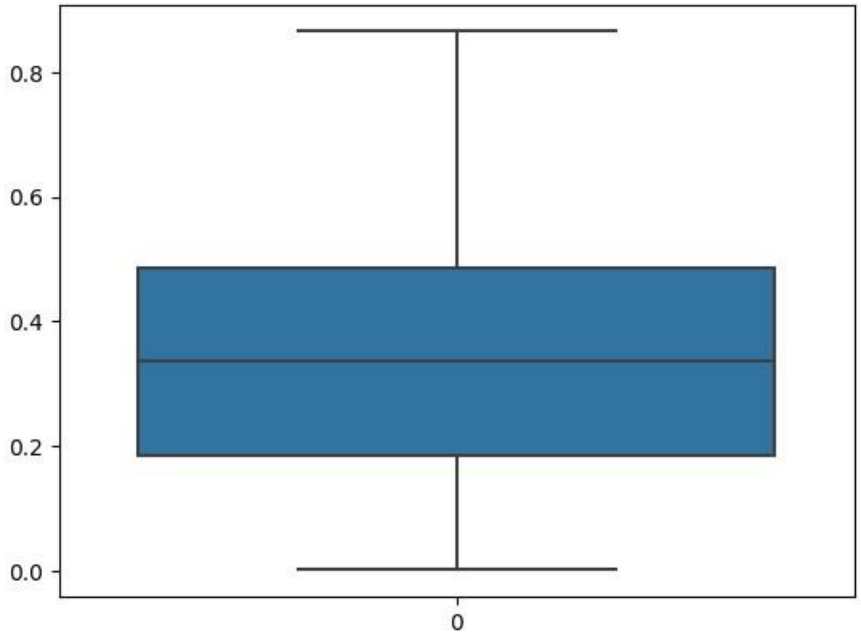
In [26]:sns.boxplot(data['Shucked weight'])

Out[26]:<AxesSubplot: >



In [27]:
```
data['Shucked weight']=np.where(data['Shucked weight']> 0.86600,0.3593676,data['Shucked weight'])
sns.boxplot(data['Shucked weight'])
```

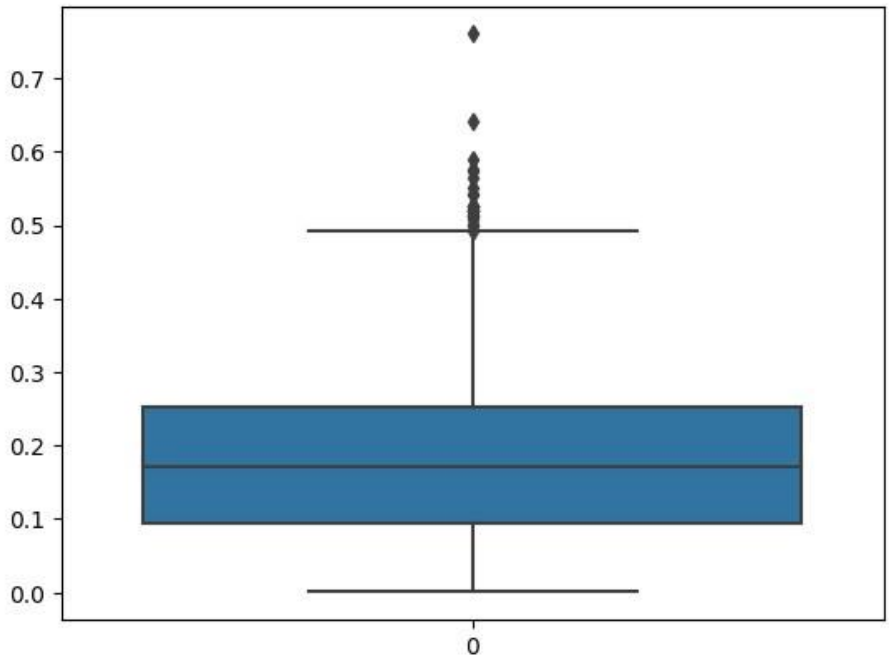
Out[27]:<AxesSubplot: >



In [28]:
```
# Outlier removed for Shucked weight
```

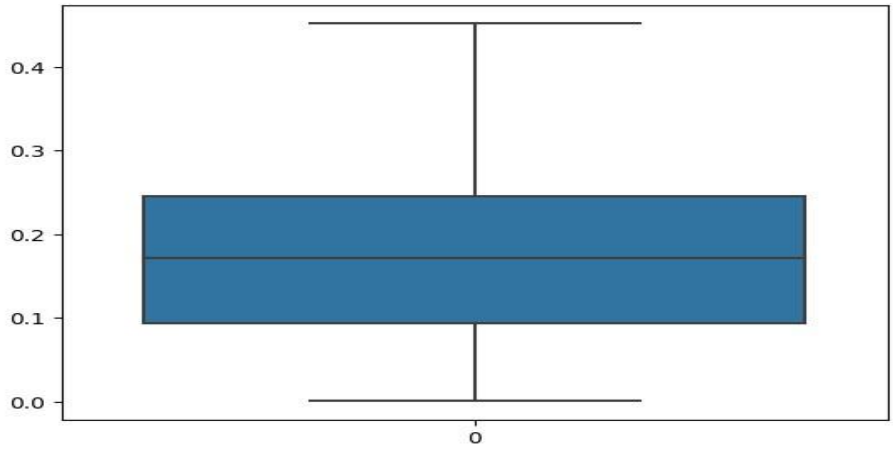
In [29]:
```
sns.boxplot(data['Viscera weight'])
```

Out[29]:<AxesSubplot: >



In [30]:data['Viscera weight']=np.where(data['Viscera weight']> 0.45225,0.180594,data['Viscera weight'])
sns.boxplot(data['Viscera weight'])

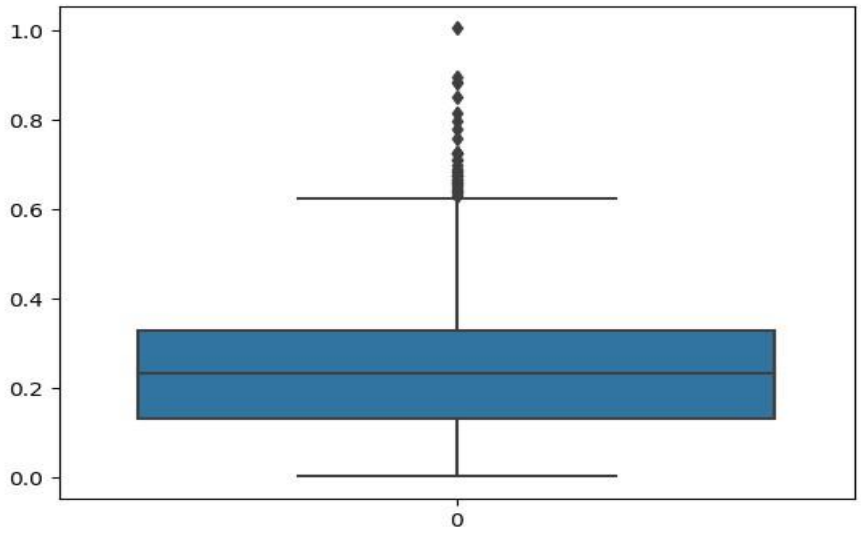
Out[30]:<AxesSubplot: >



In [31]:*# Outlier removed for Viscera weight*

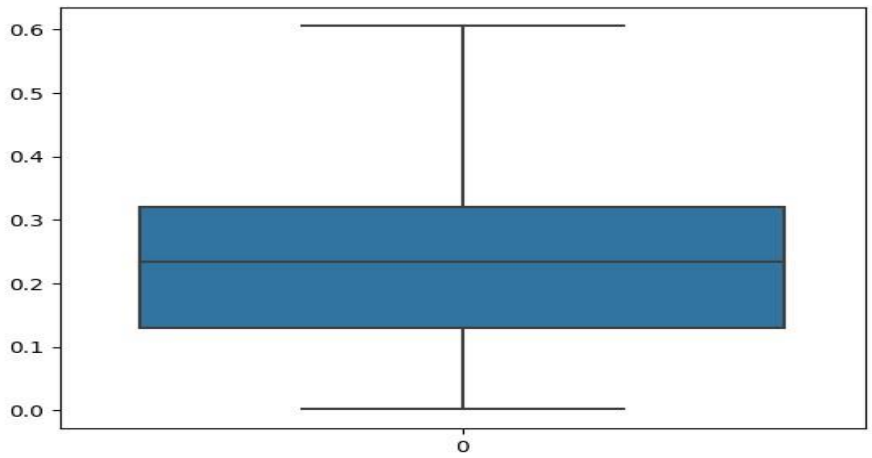
In [32]:sns.boxplot(data['Shell weight'])

Out[32]:<AxesSubplot: >



In [33]:data['Shell weight']=np.where(data['Shell weight']> 0.60750,0.238831,data['Shell weight'])
sns.boxplot(data['Shell weight'])

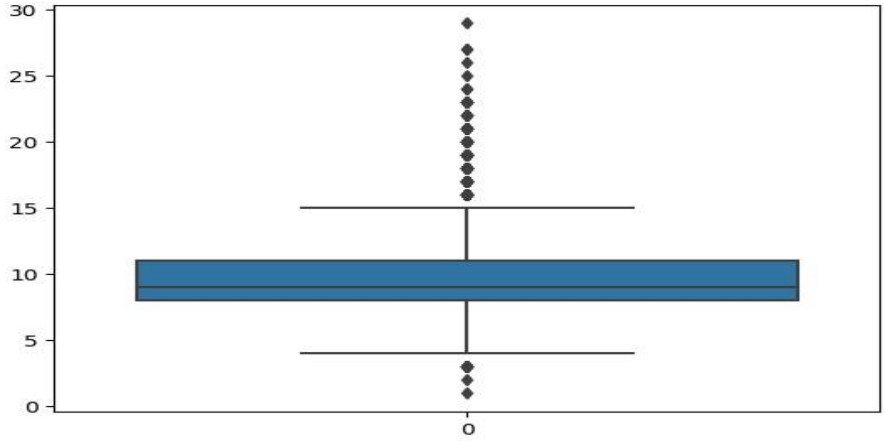
Out[33]:<AxesSubplot: >



In [34]:*# Outlier removed for Shell weight*

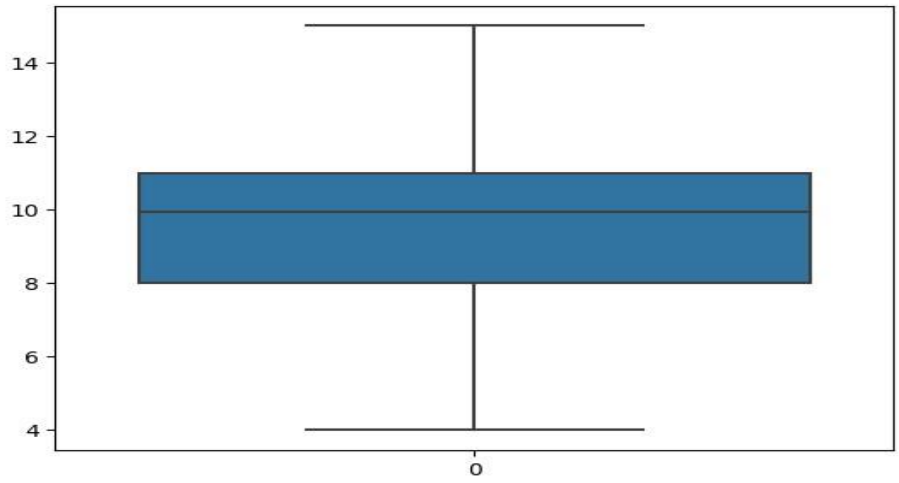
In [35]:sns.boxplot(data['Rings'])

Out[35]:<AxesSubplot: >



In [8]:data['Rings']=np.where(data['Rings']> 15.50000,9.933684,data['Rings'])
data['Rings']=np.where(data['Rings']< 3.50000,9.933684,data['Rings']) sns.boxplot(data['Rings'])

Out[8]:<AxesSubplot: >



In [37]:*# Outlier removed for Ring*

7--Encoding

In [3]:data=pd.get_dummies(data,columns=['Sex'])

In [10]:data.head()

Out[10]:

	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	Rings	Sex_F	Sex_I	Sex_M
0	0.455	0.365	0.095	0.5140	0.2245	0.1010	0.150	15.0	0	0	1
1	0.350	0.265	0.090	0.2255	0.0995	0.0485	0.070	7.0	0	0	1
2	0.530	0.420	0.135	0.6770	0.2565	0.1415	0.210	9.0	1	0	0
3	0.440	0.365	0.125	0.5160	0.2155	0.1140	0.155	10.0	0	0	1
4	0.330	0.255	0.080	0.2050	0.0895	0.0395	0.055	7.0	0	1	0

8--Splitting dependant and independant variables

In [4]:x=data.drop(columns=['Rings']).values
y=data.Rings.values x

Out[4]:array([[0.455, 0.365, 0.095, ..., 0. , 0. , 1.],
[0.35 , 0.265, 0.09 , ..., 0. , 0. , 1.], [0.53
, 0.42 , 0.135, ..., 1. , 0. , 0.],
...,
[0.6 , 0.475, 0.205, ..., 0. , 0. , 1.],
[0.625, 0.485, 0.15 , ..., 1. , 0. , 0.],
[0.71 , 0.555, 0.195, ..., 0. , 0. , 1.]])

In [12]:y

Out[12]:array([15., 7., 9., ..., 9., 10., 12.])

9--Scaling independant variables

```
In [13]:from sklearn.preprocessing import scale
         x=scale(x) x
```

```
Out[13]:array([[-0.57455813, -0.43214879, -1.06442415, ..., -0.67483383,
                -0.68801788,  1.31667716],
               [-1.44898585, -1.439929  , -1.18397831, ..., -0.67483383,
                -0.68801788,  1.31667716],
               [ 0.05003309,  0.12213032, -0.10799087, ...,  1.48184628,
                -0.68801788, -0.75948762],
               ...,
               [ 0.6329849 ,  0.67640943,  1.56576738, ..., -0.67483383,
                -0.68801788,  1.31667716],
               [ 0.84118198,  0.77718745,  0.25067161, ...,  1.48184628,
                -0.68801788, -0.75948762],
               [ 1.54905203,  1.48263359,  1.32665906, ..., -0.67483383,
                -0.68801788,  1.31667716]])
```

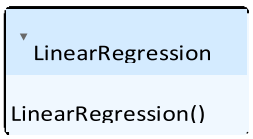
10--splitting the data into training and testing

```
In [5]: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)
In [44]:x_train.shape
Out[44]:(3341, 10)
In [45]:x_test.shape
Out[45]:(836, 10)
In [46]:x.shape
Out[46]:(4177, 10)
In [47]:y_train.shape
Out[47]:(3341,)
In [48]:y_test.shape
Out[48]:(836,)
In [49]:y_train.shape
Out[49]:(3341,)
```

11-Building the model

```
In [6]:from sklearn.linear_model import LinearRegression reg=LinearRegression()
        reg.fit(x_train,y_train)
```

Out[6]:



12--Training the model

```
In [7]:#prediction on training data

        pred=reg.predict(x_train) pred
```

```
Out[7]:array([ 5.32778396,  5.34787166, 13.87574587, ...,  9.71432748,
               11.7565958 ,  8.03917066])
In [8]:#prediction on testing data
       y_pred=reg.predict(x_test) y_pred
```

```
Out[8]:array([13.17014483,  9.10922796, 10.35985364,  5.54120008, 10.65916033,
              12.05100358,  8.26779233, 10.09973617,  8.07089377, 12.28730504,
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               9.45939464,  8.32687786, 13.85702971, 11.09090306,  7.77985409,
               7.44406172,  6.94780239,  8.88503259,  7.9654523 ,  9.51969794,
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               9.76276335,  9.2051602 , 11.83360392, 11.42857215,  7.90264847,
```

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7.96279994, 10.7008404 , 9.10178821, 8.13580849, 12.18420246,
5.69113674])

13--Testing the model

In [9]:reg.predict([[0.530,0.139516,0.135,0.6770,0.2565,0.1415,0.210,1,0,0]])

Out[9]:array([7.93478621])

14--Performance using metrices

In [10]:**from** sklearn.metrics **import** r2_score print(f'Training
accuracy:{(r2_score(y_train,pred))*100:.2f}%') print(f'Testing
accuracy:{(r2_score(y_test,y_pred))*100:.2f}%')

Training accuracy:53.70%

Testing accuracy:53.90%