

```
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

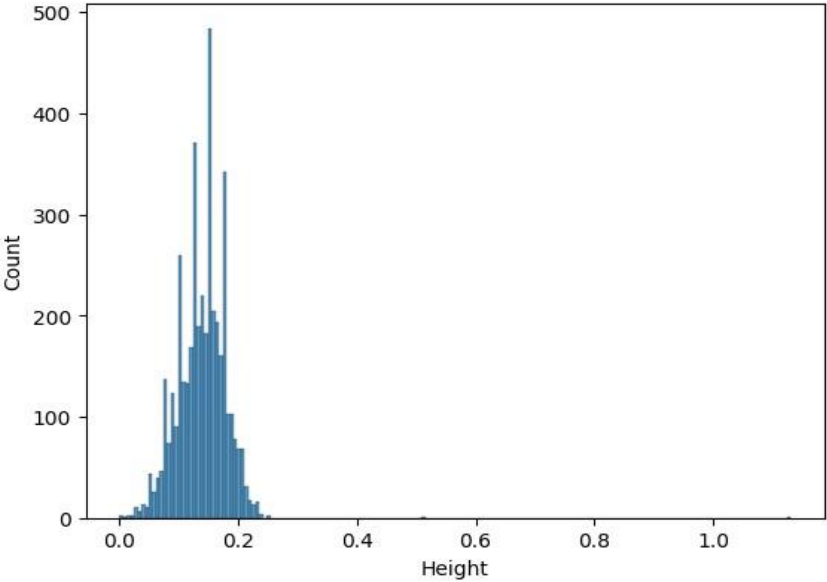
2--Load the dataset

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

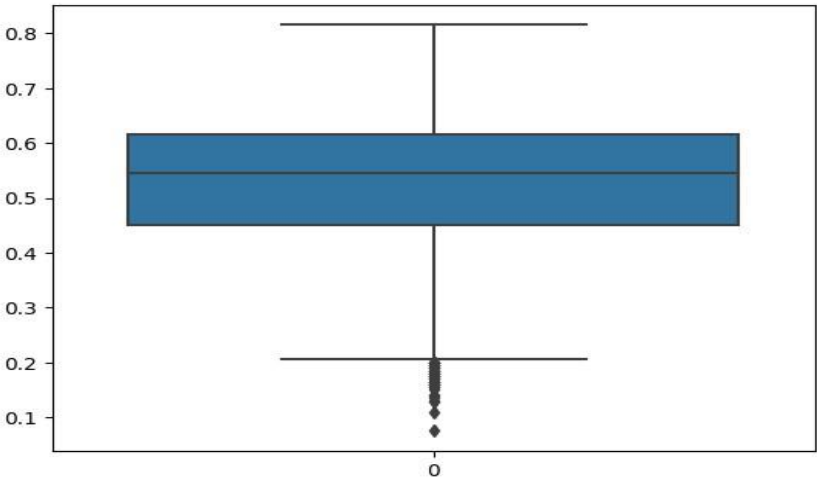
3--Visuslization

Univariate analysis

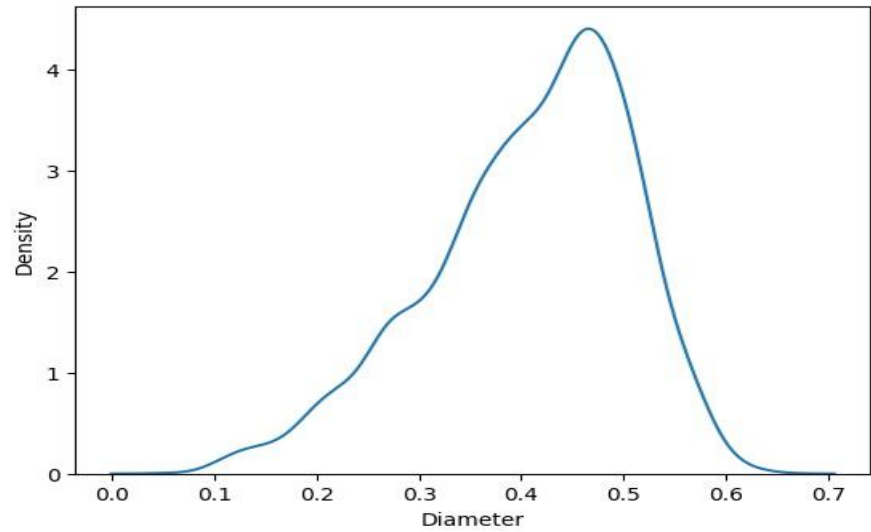
```
In [3]:sns.histplot(data['Height'])
Out[3]:<AxesSubplot: xlabel='Height', ylabel='Count'>
```



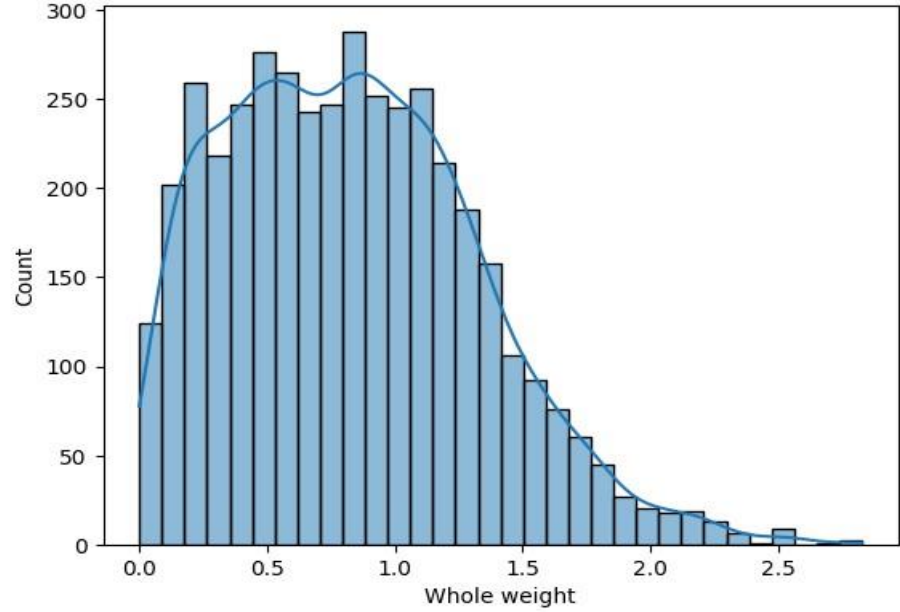
```
In [4]:sns.boxplot(data['Length'])
Out[4]:<AxesSubplot: >
```



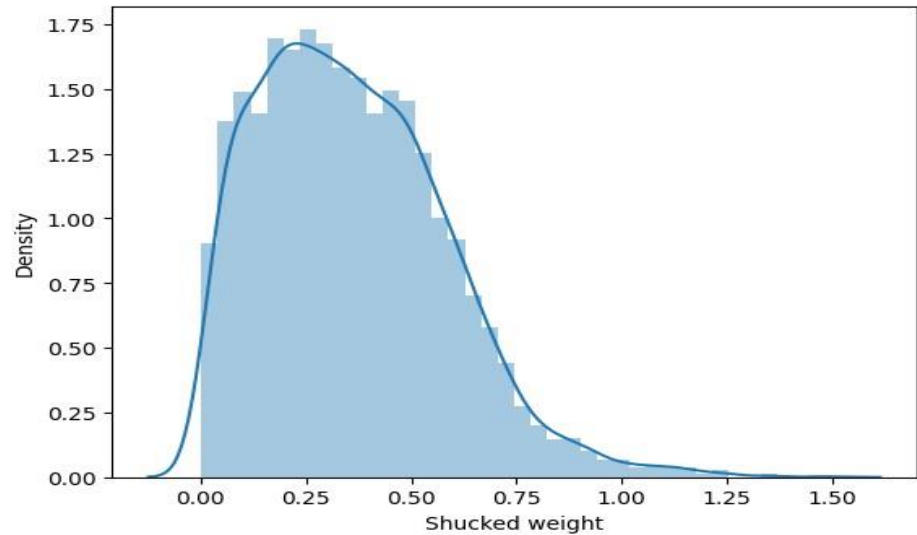
```
In [5]:sns.kdeplot(data['Diameter'])  
Out[5]:<AxesSubplot: xlabel='Diameter', ylabel='Density'>
```



```
In [6]:sns.histplot(data['Whole weight'],kde=True)  
Out[6]:<AxesSubplot: xlabel='Whole weight', ylabel='Count'>
```

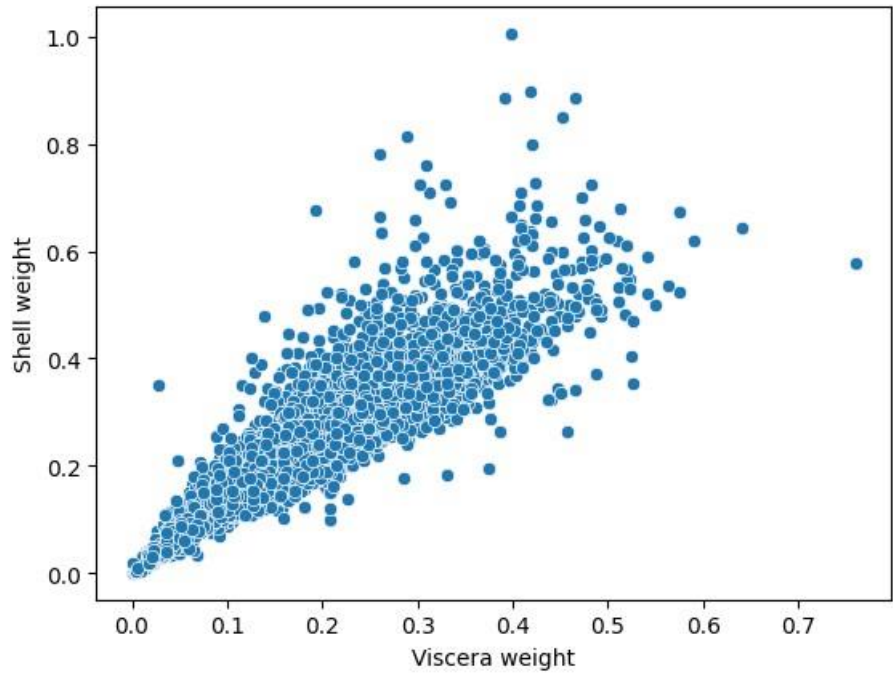


```
In [7]:sns.distplot(data['Shucked weight'],kde=True)  
Out[7]:<AxesSubplot: xlabel='Shucked weight', ylabel='Density'>
```

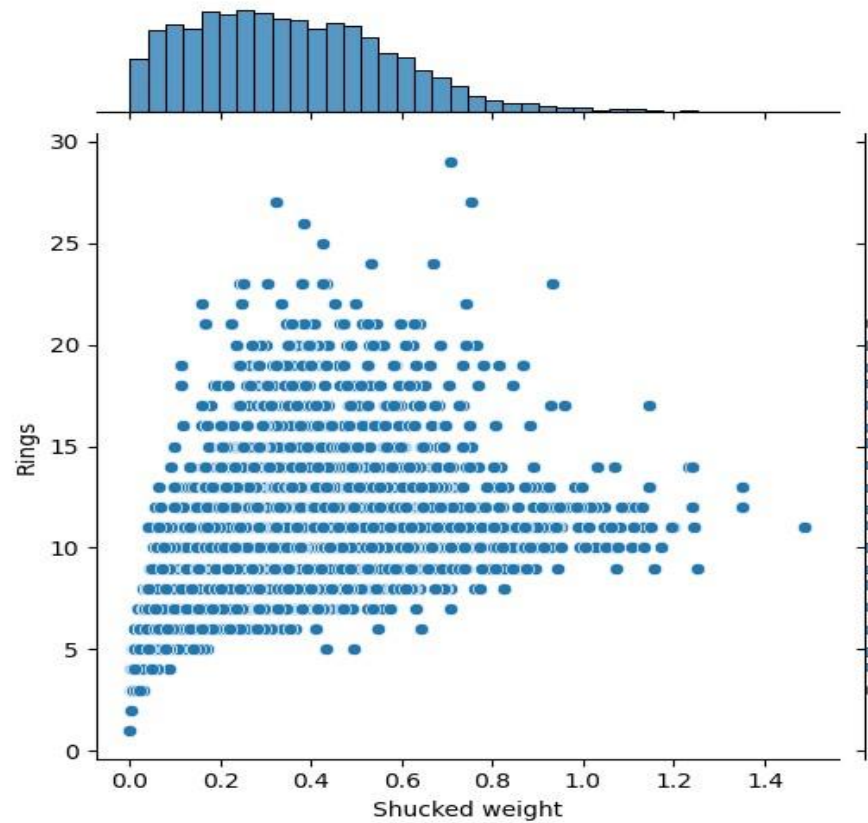


Bivariate Analysis

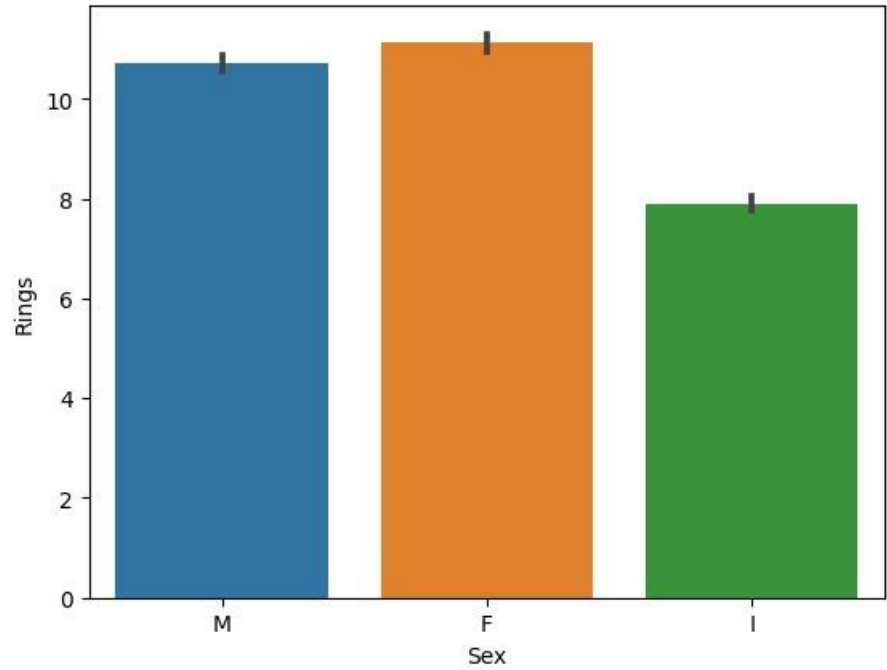
```
In [8]:sns.scatterplot(data=data,x='Viscera weight', y='Shell weight')
Out[8]:<AxesSubplot: xlabel='Viscera weight', ylabel='Shell weight'>
```



```
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

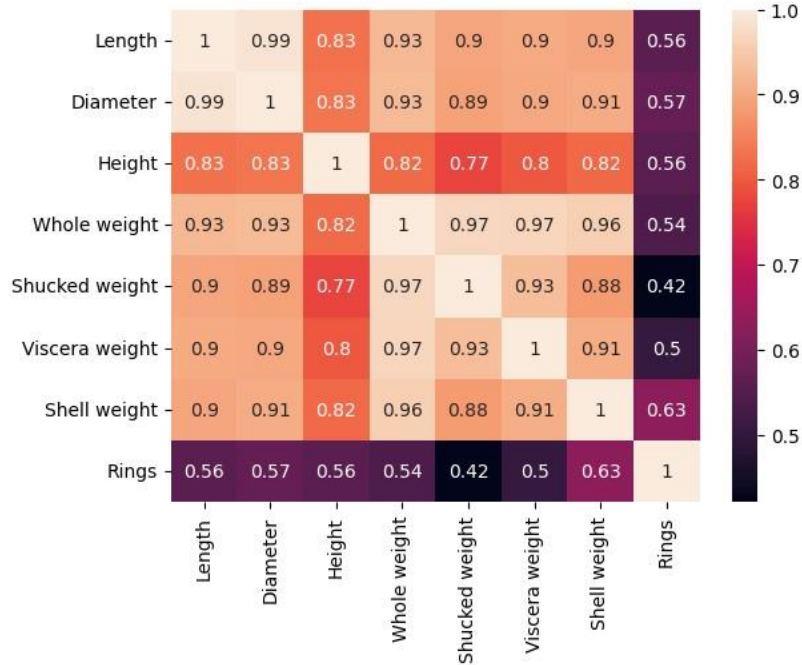
In [11]:data.corr()

Out[11]:

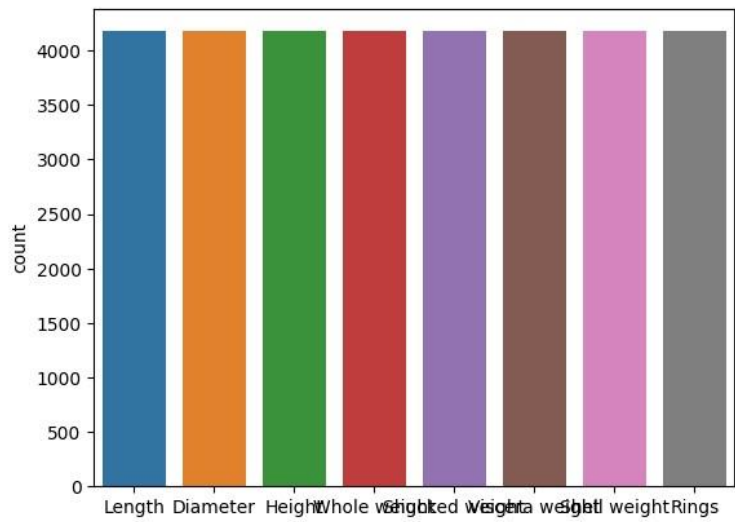
	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	Rings
Length	1.000000	0.986812	0.827554	0.925261	0.897914	0.903018	0.897706	0.556720
Diameter	0.986812	1.000000	0.833684	0.925452	0.893162	0.899724	0.905330	0.574660
Height	0.827554	0.833684	1.000000	0.819221	0.774972	0.798319	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)

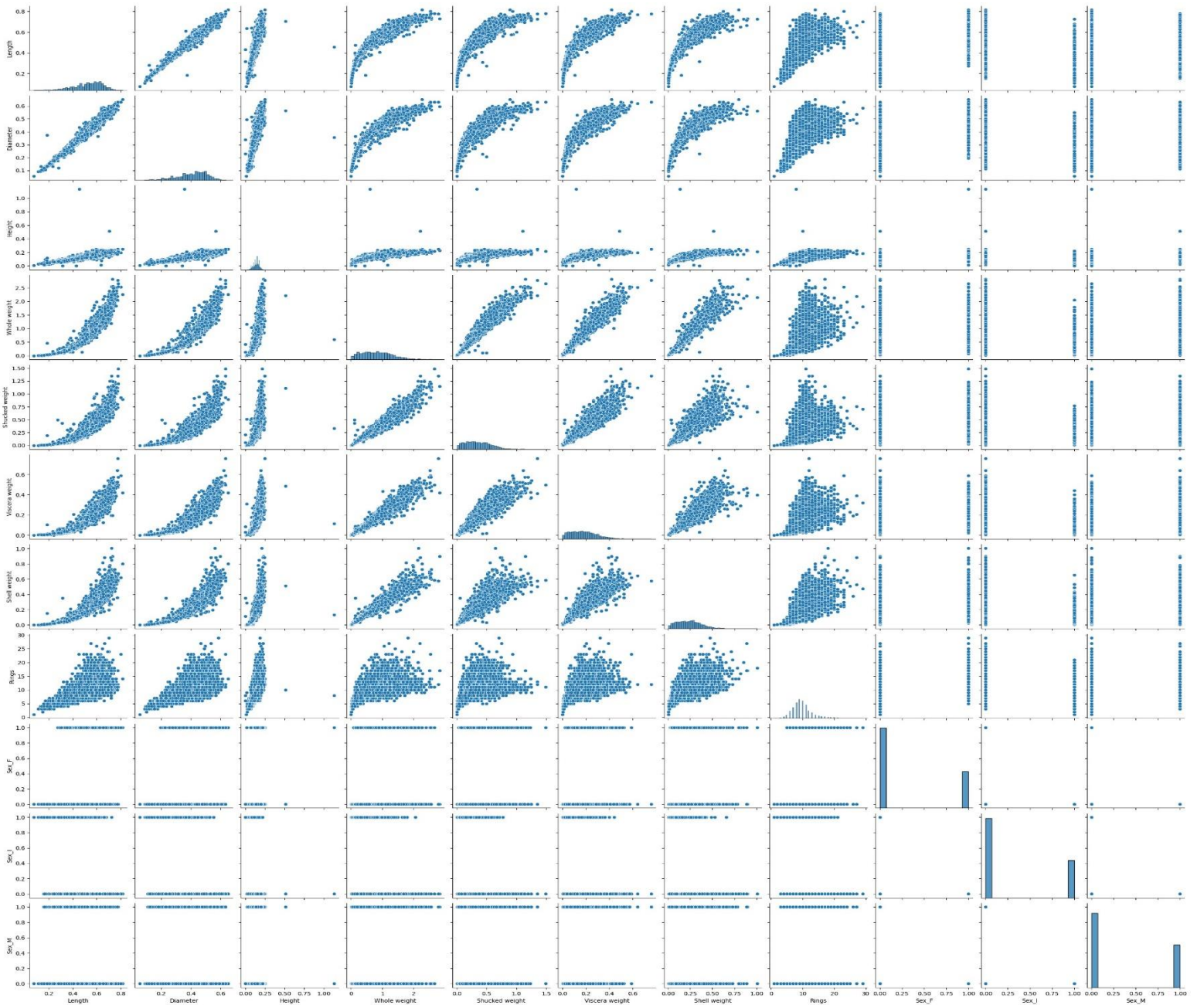
Out[12]:<AxesSubplot: >



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

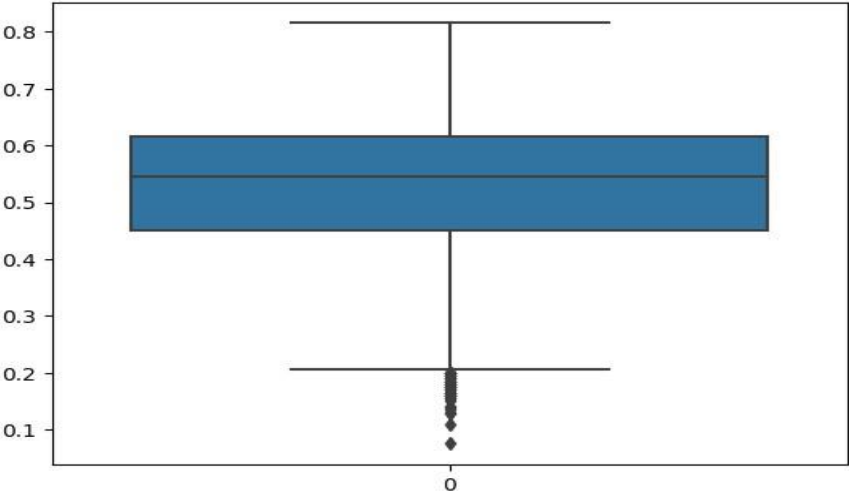
```
In [6]:lower=quant.loc[0.25]-(1.5*iqr) lower
Out[6]:Length      0.20250
Diameter     0.15500
Height       0.04000
Whole weight -0.62575
Shucked weight -0.28800
Viscera weight -0.14575
Shell weight -0.16850 Rings
3.50000 dtype: float64
```

MEAN

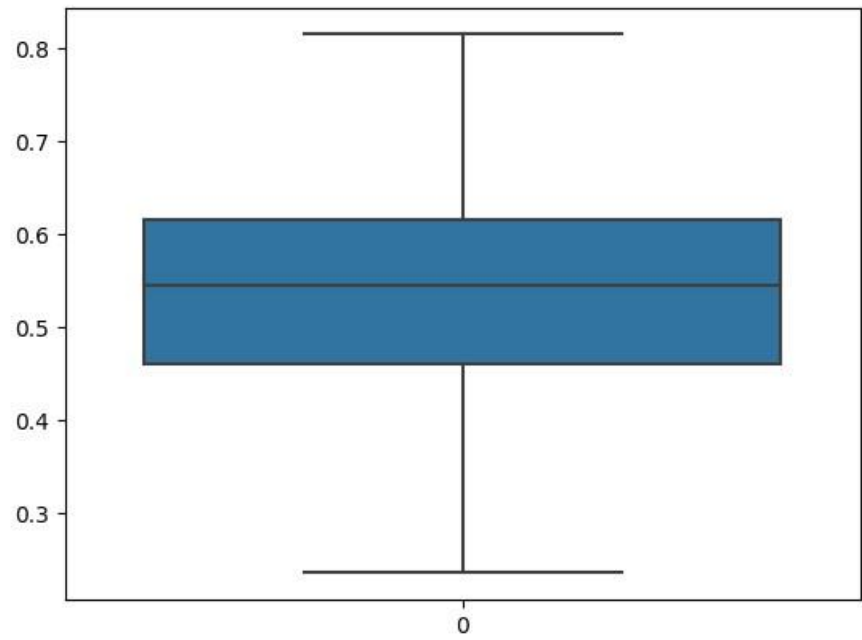
```
In [7]:data.mean()
Out[7]:Length      0.523992
Diameter     0.407881
Height       0.139516
Whole weight  0.828742
Shucked weight  0.359367
Viscera weight  0.180594
Shell weight  0.238831 Rings
9.933684 dtype: float64
```

Removing outlier

```
In [17]:sns.boxplot(data['Length'])
Out[17]:<AxesSubplot: >
```

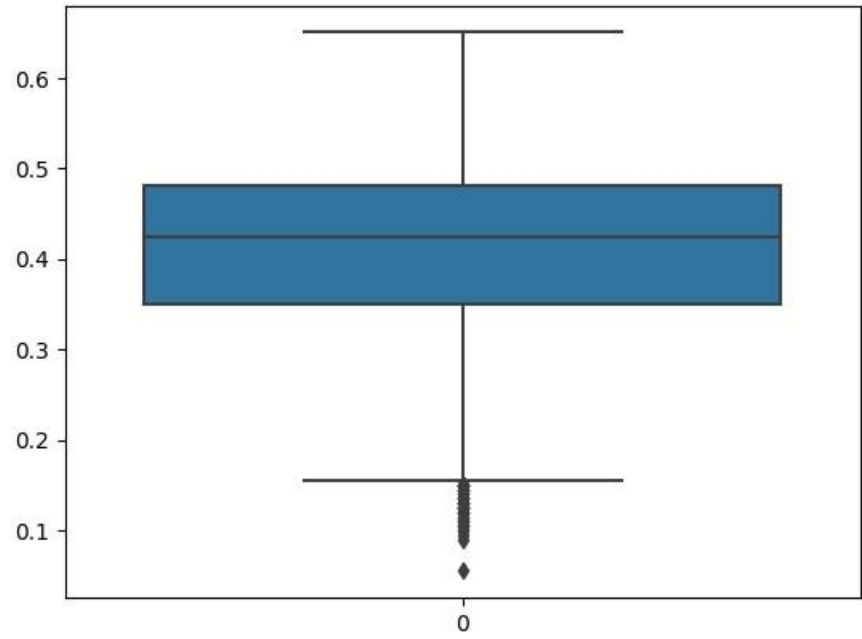


```
In [18]:data['Length']=np.where(data['Length']< 0.23350,0.523992,data['Length']) sns.boxplot(data['Length'])
Out[18]:
```

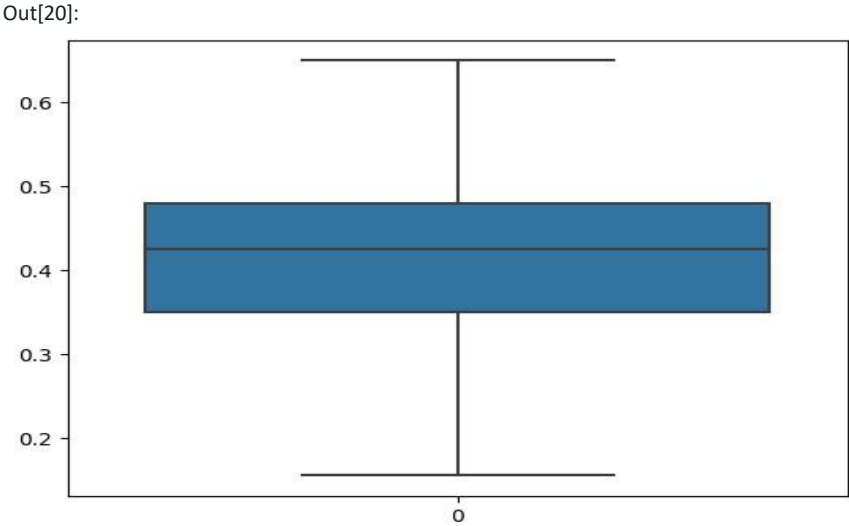


Outlier removed for Length

```
In [19]:sns.boxplot(data['Diameter'])
Out[19]:<AxesSubplot: >
```

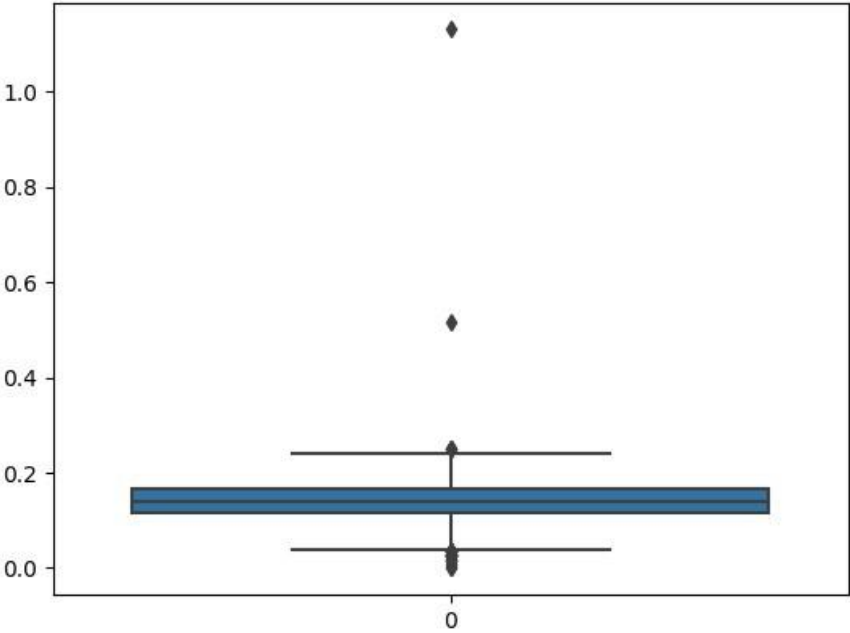



```
In [20]:data['Diameter']=np.where(data['Diameter']<0.15500,0.407881,data['Diameter'])
sns.boxplot(data['Diameter'])
```

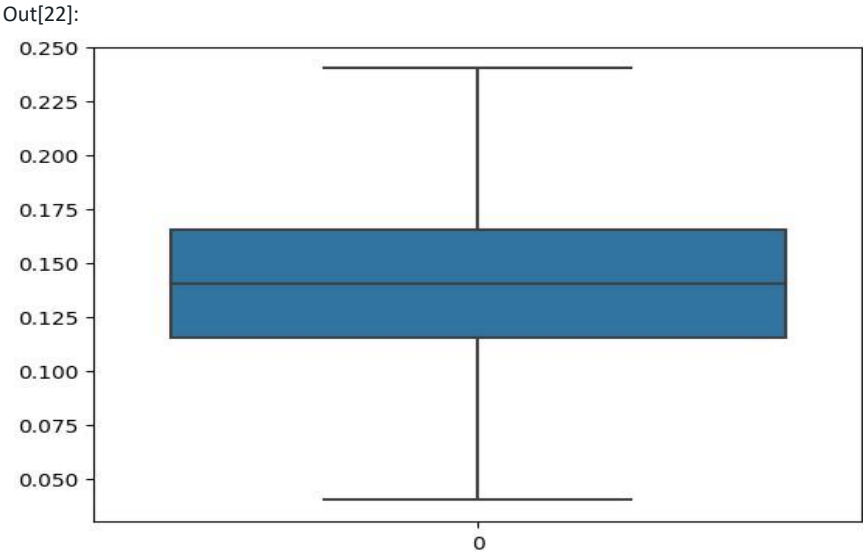


Outlier removed for diameter

```
In [21]:sns.boxplot(data['Height'])
Out[21]:<AxesSubplot: >
```



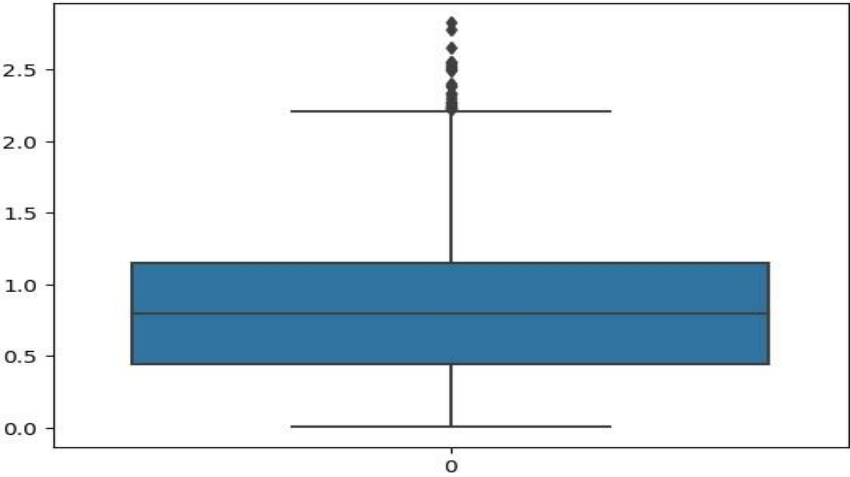
```
In [22]:data['Height']=np.where(data['Height']< 0.04000,0.139516,data['Height'])
data['Height']=np.where(data['Height']> 0.24000,0.139516,data['Height']) sns.boxplot(data['Height'])
```



Outlier removed for Height

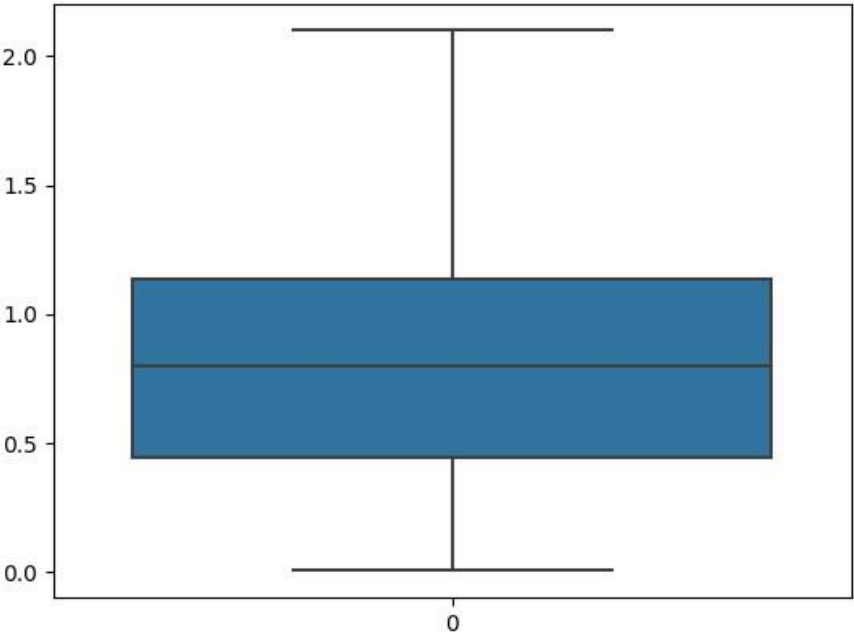
In [23]:sns.boxplot(data['Whole weight'])

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'])

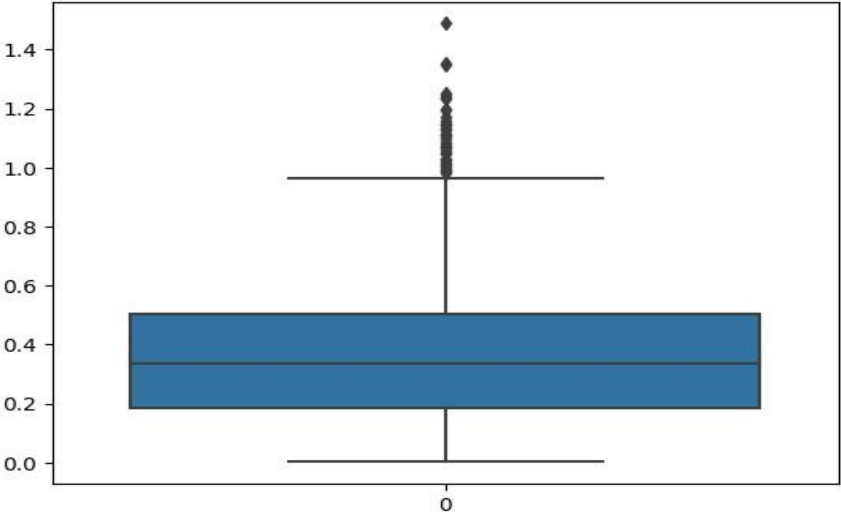
Out[24]:



In [25]:# Outlier removed for Whole weightt

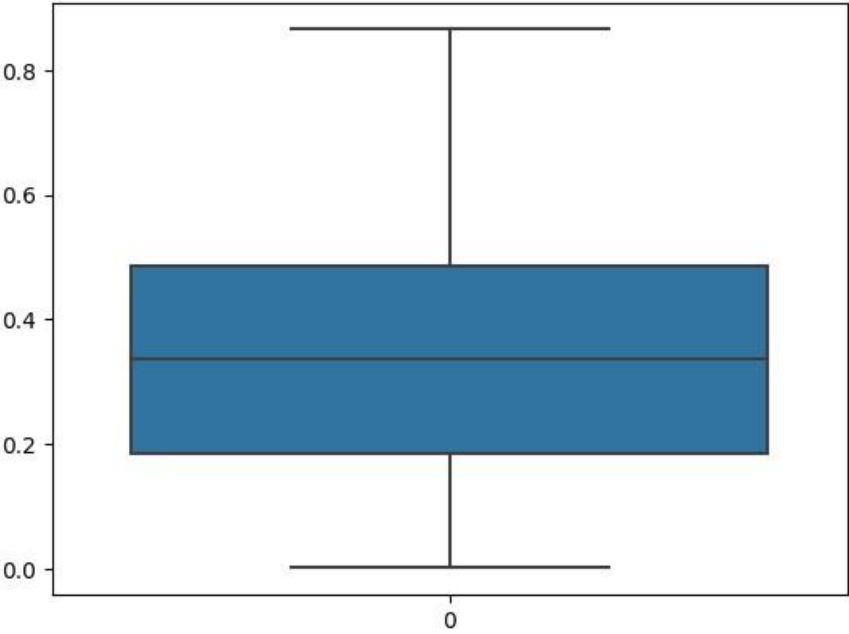
In [26]:sns.boxplot(data['Shucked weight'])

Out[26]:<AxesSubplot: >



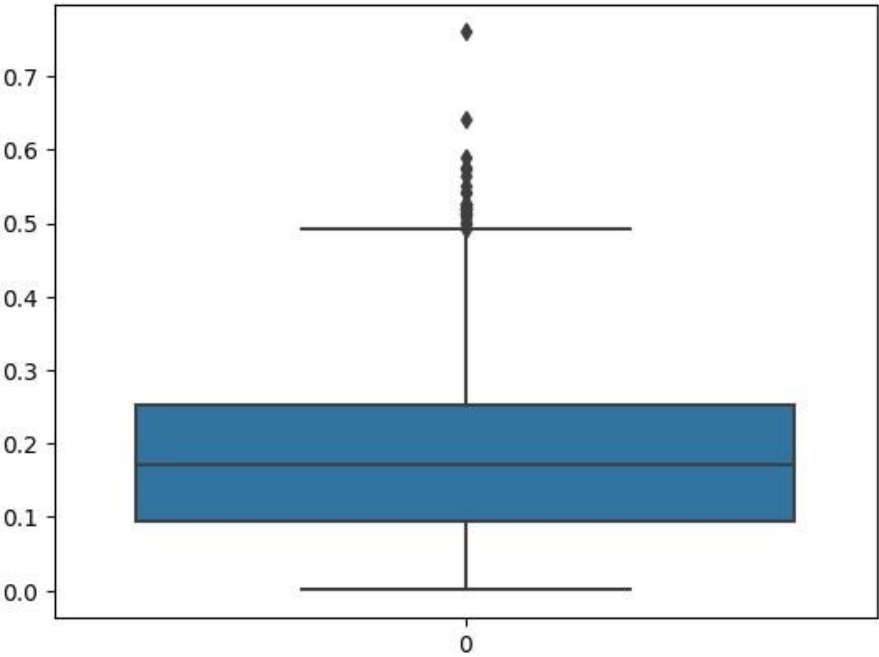
```
In [27]:data['Shucked weight']=np.where(data['Shucked weight']> 0.86600,0.3593676,data['Shucked weight'])
sns.boxplot(data['Shucked weight'])
```

Out[27]:<AxesSubplot: >

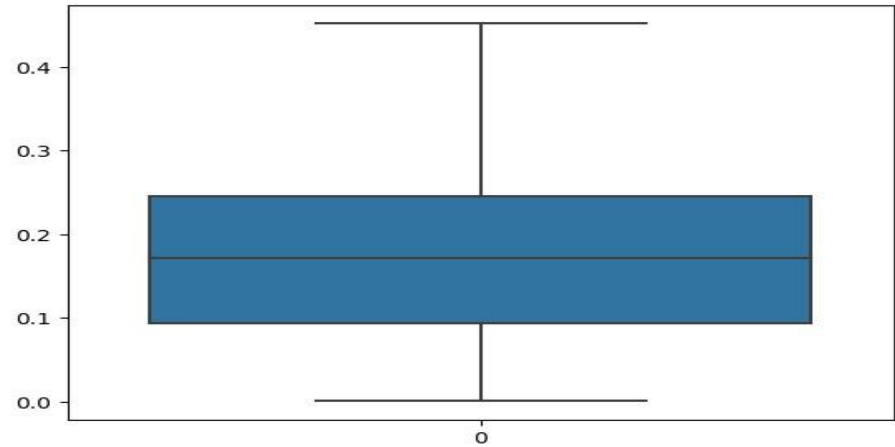


```
In [28]:# Outlier removed for Shucked weight
In [29]:sns.boxplot(data['Viscera weight'])
```

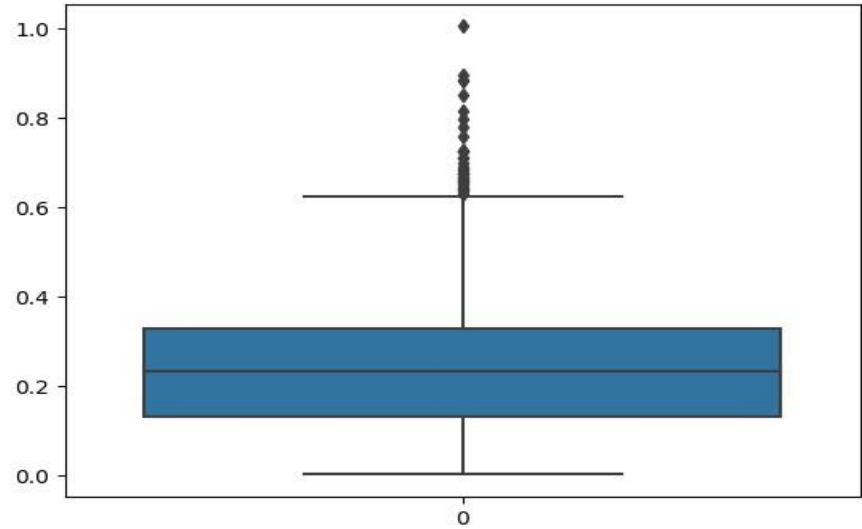
Out[29]:<AxesSubplot: >



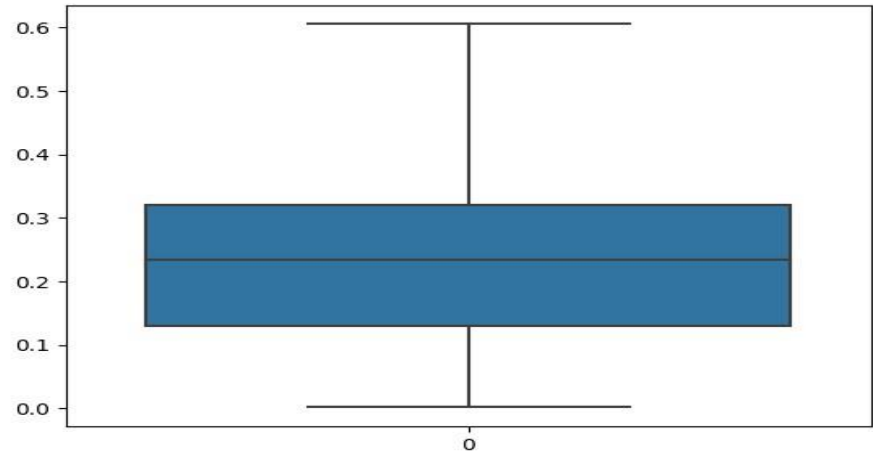
```
In [30]:data['Viscera weight']=np.where(data['Viscera weight']> 0.45225,0.180594,data['Viscera weight'])
sns.boxplot(data['Viscera weight'])
Out[30]:<AxesSubplot: >
```



```
In [31]:# Outlier removed for Viscera weight
In [32]:sns.boxplot(data['Shell weight'])
Out[32]:<AxesSubplot: >
```

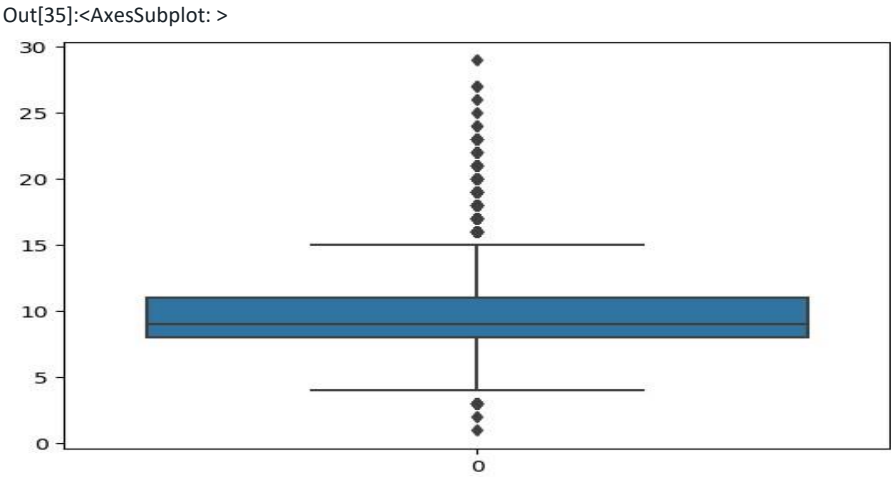


```
In [33]:data['Shell weight']=np.where(data['Shell weight']> 0.60750,0.238831,data['Shell weight'])
sns.boxplot(data['Shell weight'])
Out[33]:<AxesSubplot: >
```

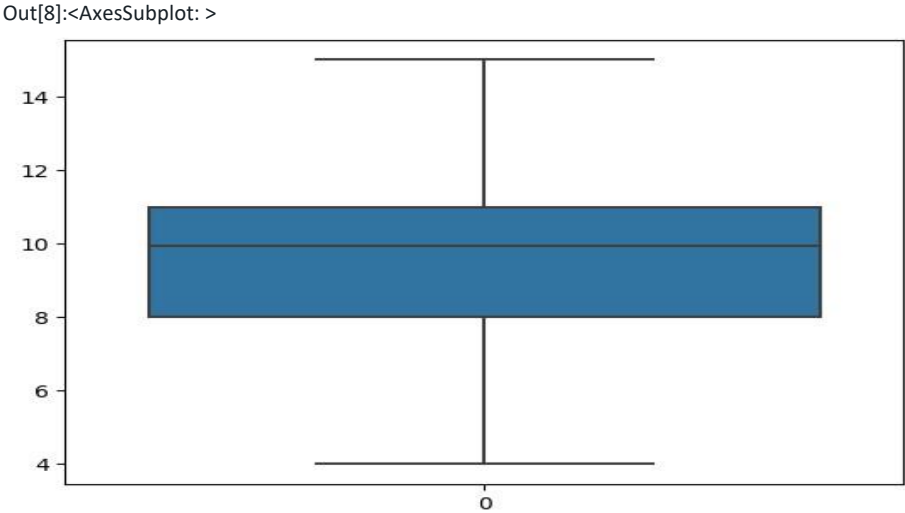


```
In [34]:# Outlier removed for Shell weight
```

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



```
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'])
```



```
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										
1	0.350	0.365	0.095	0.5140	0.2245	0.1010	0.150	15.0	0	0	1
2	0.530	0.265	0.090	0.2255	0.0995	0.0485	0.070	7.0	0	0	1
3	0.440	0.420	0.135	0.6770	0.2565	0.1415	0.210	9.0	1	0	0
4	0.330	0.365	0.125	0.5160	0.2155	0.1140	0.155	10.0	0	0	1
		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. ], ...,
              [0.42 , 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]:
```

LinearRegression

LinearRegression()

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,
        8.22228608,  6.26339268,  8.4766956 ,  9.30261309,  5.70767624,
        9.45939464,  8.32687786, 13.85702971, 11.09090306,  7.77985409,
        7.44406172,  6.94780239,  8.88503259,  7.9654523 ,  9.51969794,
        11.56414001,  5.59707836, 13.11441854, 10.10204559, 11.59519979,
        8.45138755,  4.66462354, 11.30148049, 12.76604688,  7.15709165,
        8.63988457,  8.5200469 , 10.25020425,  8.16602336, 11.70289969,
        11.60141065,  9.56418323, 12.04986801, 11.89787512, 12.49372844,
        9.76276335,  9.2051602 , 11.83360392, 11.42857215,  7.90264847,
```

11.93961315, 7.13983251, 9.40292052, 13.71258569, 9.35573394,
8.09482073, 6.88614845, 7.8807578, 7.3433024, 7.10793707,
9.7911508, 9.14297012, 10.43650314, 7.60734099, 8.01110075,
12.3951208, 12.398778, 12.34386704, 8.53070865, 14.31136822,
9.164063, 18.80234388, 10.85486878, 10.62353632, 10.09018517,
8.90345615, 10.26383911, 10.1508343, 11.30742361, 8.5655046,
9.62245825, 6.12382531, 7.17042395, 12.21957718, 9.45685753,
8.02722065, 10.11019131, 12.4580311, 5.03646764, 7.00715717,
9.71770007, 10.62680397, 8.37143, 2.50393825, 11.98492521,
6.25260615, 10.0485812, 8.14000991, 14.19423453, 9.54563988,
10.01505727, 11.24031021, 10.00665829, 11.15604241, 5.65409689,
10.3605766, 7.32736447, 6.994029, 7.51308546, 13.5993784,
9.06757741, 11.00923226, 11.24404651, 8.60387213, 14.38612508,
10.2577704, 11.7119004, 13.23129132, 5.21937368, 9.28352027,
7.20128047, 11.72547647, 6.87725702, 9.13839505, 11.03963916,
13.27387874, 10.61175995, 10.79014221, 7.7722332, 10.716316,
10.45506787, 8.41743889, 9.84185901, 11.2842196, 11.41959736,
11.0808743, 10.49707272, 8.82320871, 7.20728334, 13.88165762,
10.87036059, 11.54690512, 7.16806741, 8.53621066, 11.51270352,
10.27343561, 9.12768361, 7.27469893, 8.0704399, 6.65863828,
8.7182523, 17.15062041, 7.1208007, 11.09105929, 8.21477651,
6.68016316, 10.76651969, 7.03804216, 13.17375097, 8.47241615,
10.41950876, 7.86988347, 9.82248106, 10.80970665, 5.38752726,
12.27147414, 7.11609533, 6.50365269, 11.30095645, 9.35235169,
8.30256607, 6.16510119, 8.75977964, 10.50787414, 12.77043643,
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```

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 metrics

```
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%
```