

ASSIGNMENT – 4

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1.Download the dataset

In [265]:

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

2.LOAD THE DATASET

In [266]:

```
df = pd.read_csv('abalone.csv')
```

In [267]:

```
df.head
```

Out[267]:

```
<bound method NDFrame.head of
eight \
0      M    0.455    0.365    0.095          0.5140          0.2245
1      M    0.350    0.265    0.090          0.2255          0.0995
2      F    0.530    0.420    0.135          0.6770          0.2565
3      M    0.440    0.365    0.125          0.5160          0.2155
4      I    0.330    0.255    0.080          0.2050          0.0895
...
4172   F    0.565    0.450    0.165          0.8870          0.3700
4173   M    0.590    0.440    0.135          0.9660          0.4390
4174   M    0.600    0.475    0.205          1.1760          0.5255
4175   F    0.625    0.485    0.150          1.0945          0.5310
4176   M    0.710    0.555    0.195          1.9485          0.9455
```

```
Viscera weight  Shell weight  Rings
0              0.1010      0.1500    15
1              0.0485      0.0700     7
2              0.1415      0.2100     9
3              0.1140      0.1550    10
4              0.0395      0.0550     7
...
4172           0.2390      0.2490    11
4173           0.2145      0.2605    10
4174           0.2875      0.3080     9
4175           0.2610      0.2960    10
4176           0.3765      0.4950    12
```

```
df["Age"]=Age
df=df.rename(columns = {'whole weight':'whole_weight','Shucked weight':'Shucked_weight',
'Viscera weight':'Viscera_weight','Shell weight':'Shell_weight'})
df=df.drop(columns=["Rings"],axis=1)
df.head()
```

Out [268]:

	Sex	Length	Diameter	Height	Whole weight	Shucked_weight	Viscera_weight	Shell_weight	Age
0	M	0.455	0.365	0.095	0.5140	0.2245	0.1010	0.150	16.5
1	M	0.350	0.265	0.090	0.2255	0.0995	0.0485	0.070	8.5
2	F	0.530	0.420	0.135	0.6770	0.2565	0.1415	0.210	10.5
3	M	0.440	0.365	0.125	0.5160	0.2155	0.1140	0.155	11.5
4	I	0.330	0.255	0.080	0.2050	0.0895	0.0395	0.055	8.5

In [269]:

```
df.tail()
```

Out [269]:

	Sex	Length	Diameter	Height	Whole weight	Shucked_weight	Viscera_weight	Shell_weight	Age
4172	F	0.565	0.450	0.165	0.8870	0.3700	0.2390	0.2490	12.5
4173	M	0.590	0.440	0.135	0.9660	0.4390	0.2145	0.2605	11.5
4174	M	0.600	0.475	0.205	1.1760	0.5255	0.2875	0.3080	10.5
4175	F	0.625	0.485	0.150	1.0945	0.5310	0.2610	0.2960	11.5
4176	M	0.710	0.555	0.195	1.9485	0.9455	0.3765	0.4950	13.5

3. Perform Below Visualizations

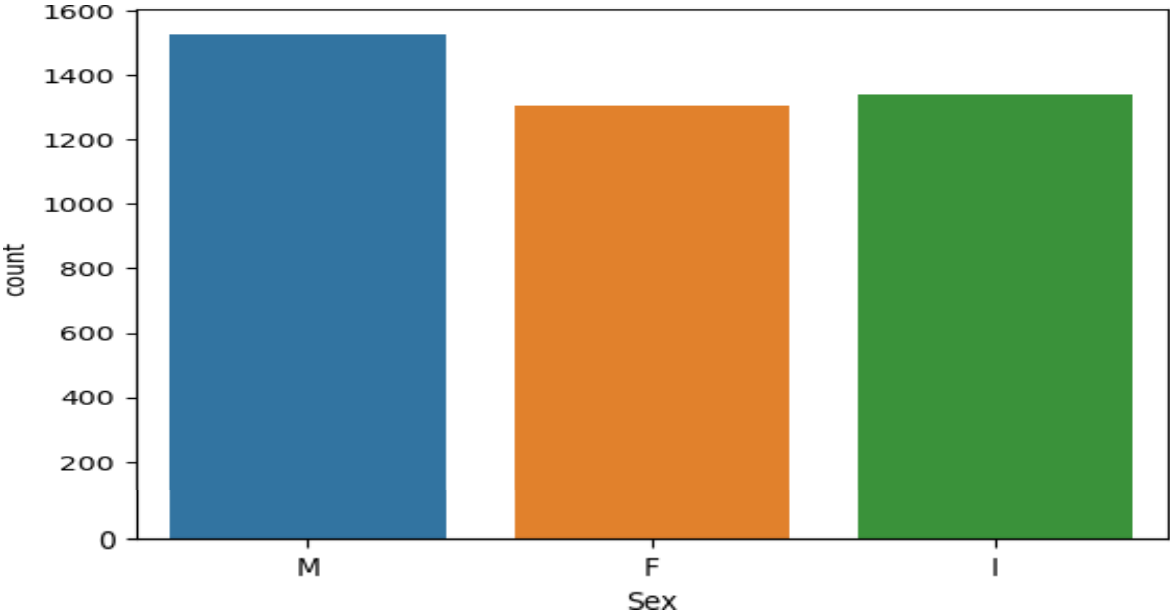
Univariate Analysis

In [270]:

```
sns.countplot(x='Sex',data=df)
```

Out [270]:

```
<AxesSubplot:xlabel='Sex', ylabel='count'>
```

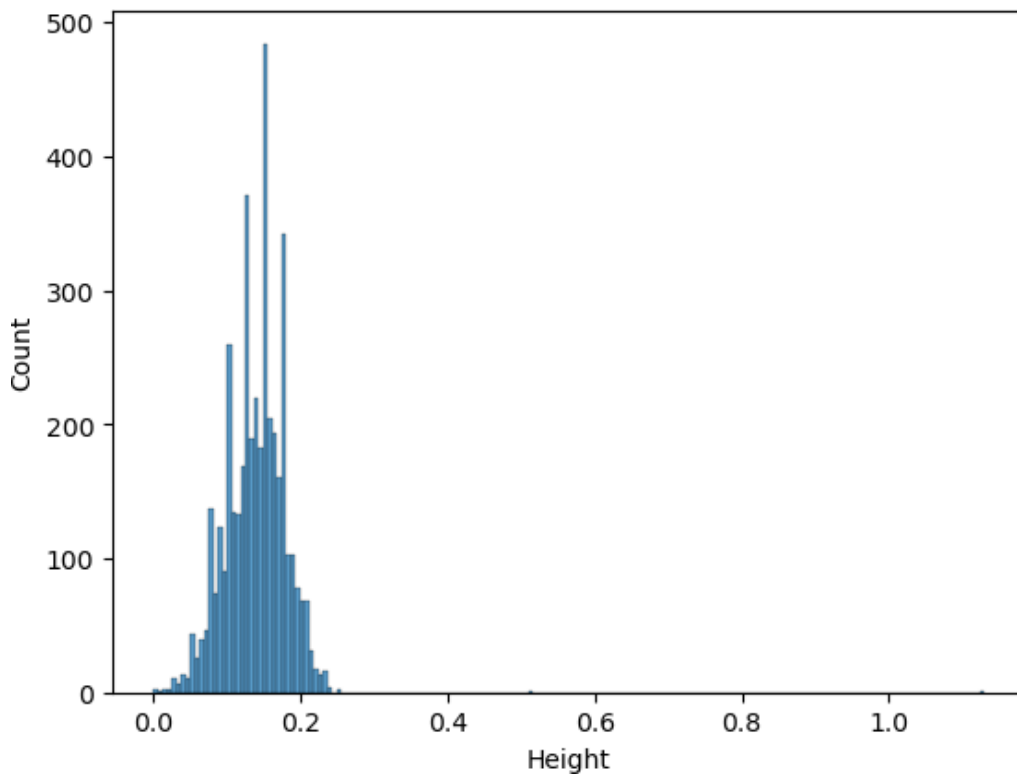


In [271]:

```
sns.histplot(df["Height"])
```

Out[271]:

<AxesSubplot:xlabel='Height', ylabel='Count'>

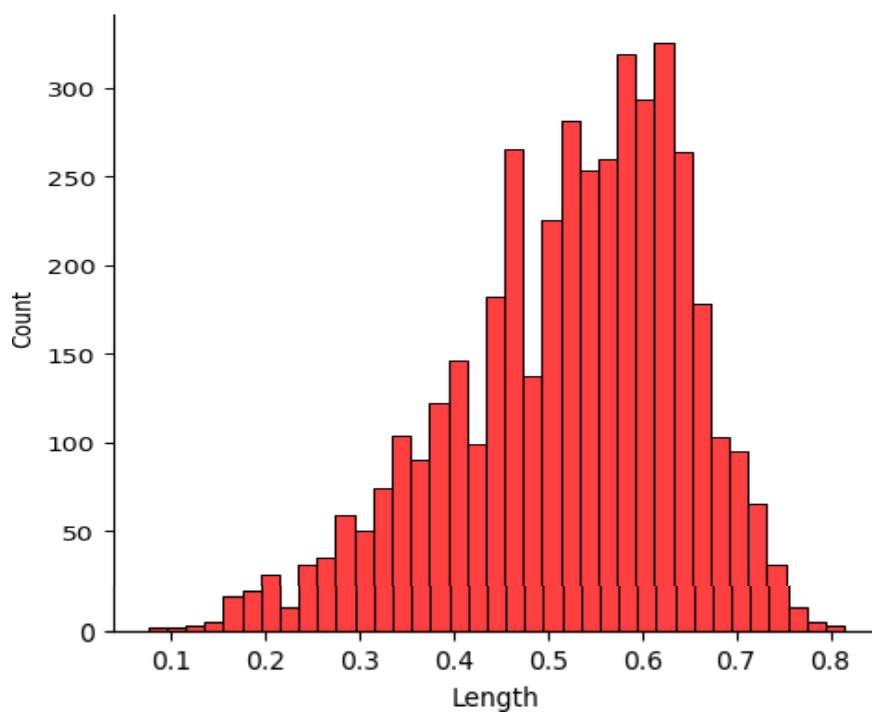


In [272]:

```
sns.displot(df["Length"], color='red')
```

Out[272]:

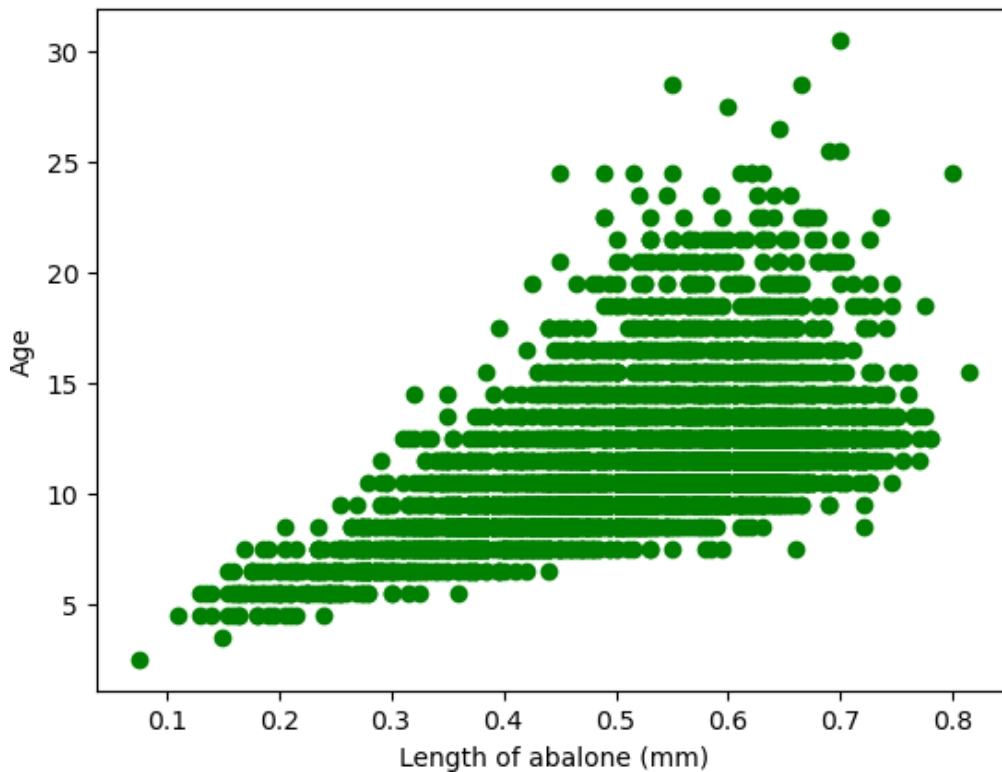
<seaborn.axisgrid.FacetGrid at 0x1af5e2f7820>



Bi-Variate Analysis

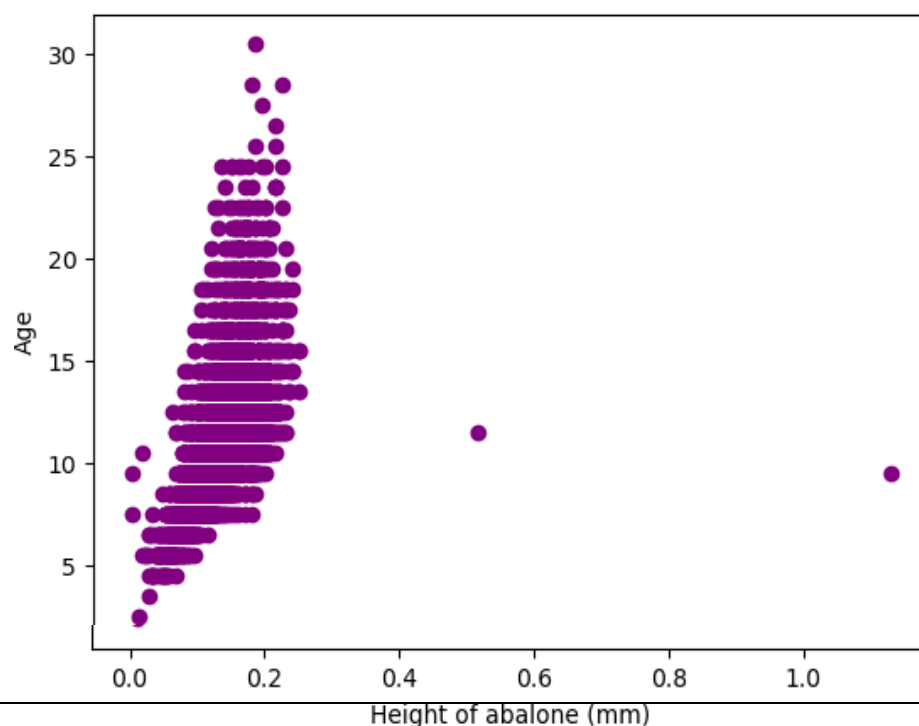
In [273]:

```
plt.scatter(df['Length'],df['Age'],c='green')  
plt.xlabel('Length of abalone (mm)')  
plt.ylabel('Age')  
plt.show()
```



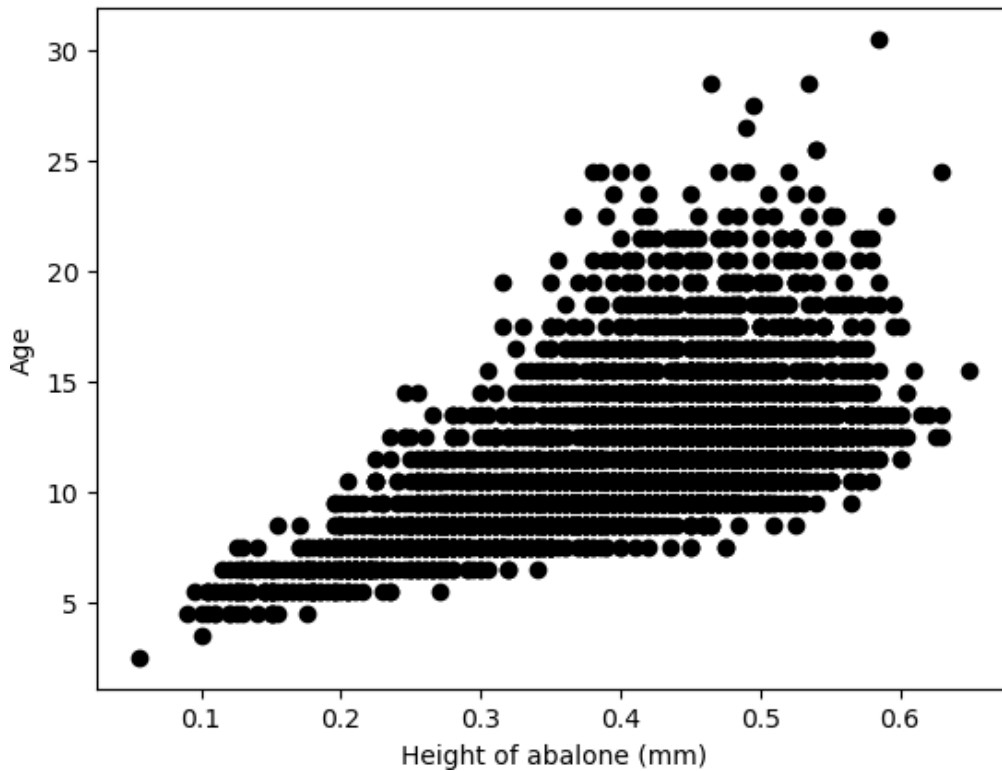
In [274]:

```
plt.scatter(df['Height'],df['Age'],c='purple')  
plt.xlabel('Height of abalone (mm)')  
plt.ylabel('Age')  
plt.show()
```



In [275]:

```
plt.scatter(df['Diameter'],df['Age'],c='black')
plt.xlabel('Height of abalone (mm)')
plt.ylabel('Age')
plt.show()
```



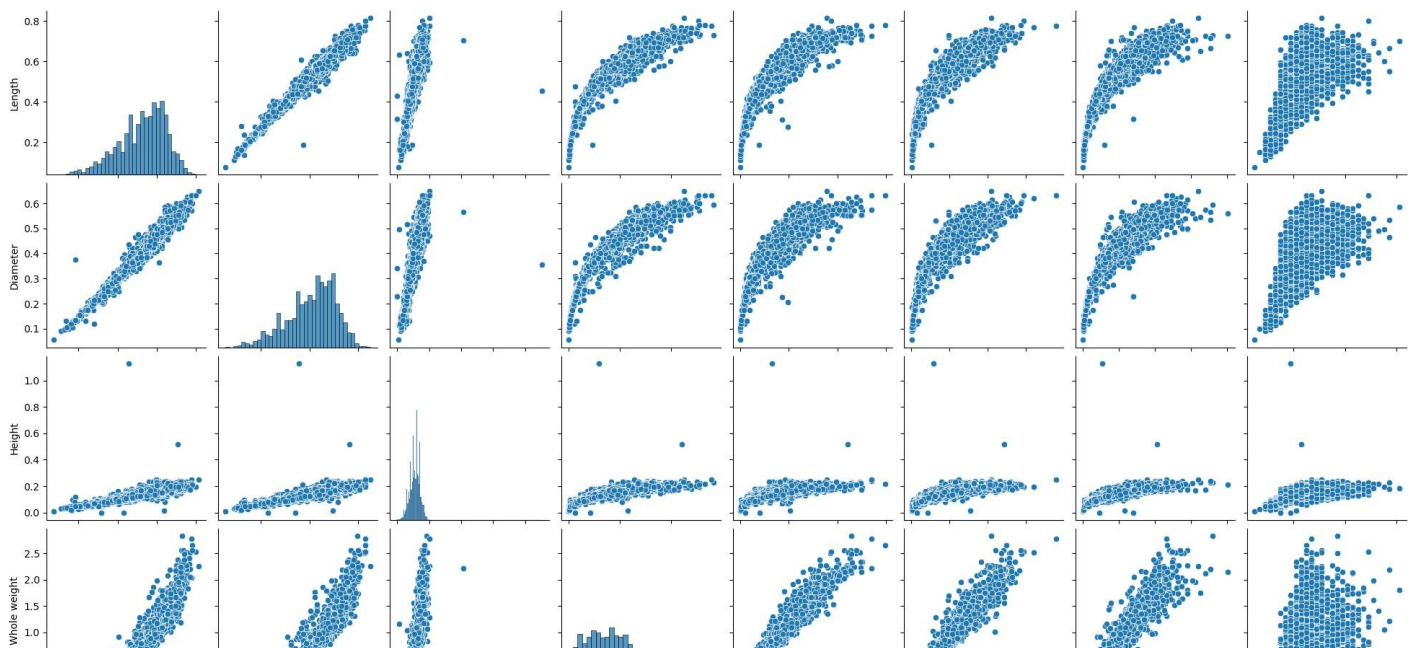
Multi-Variate Analysis

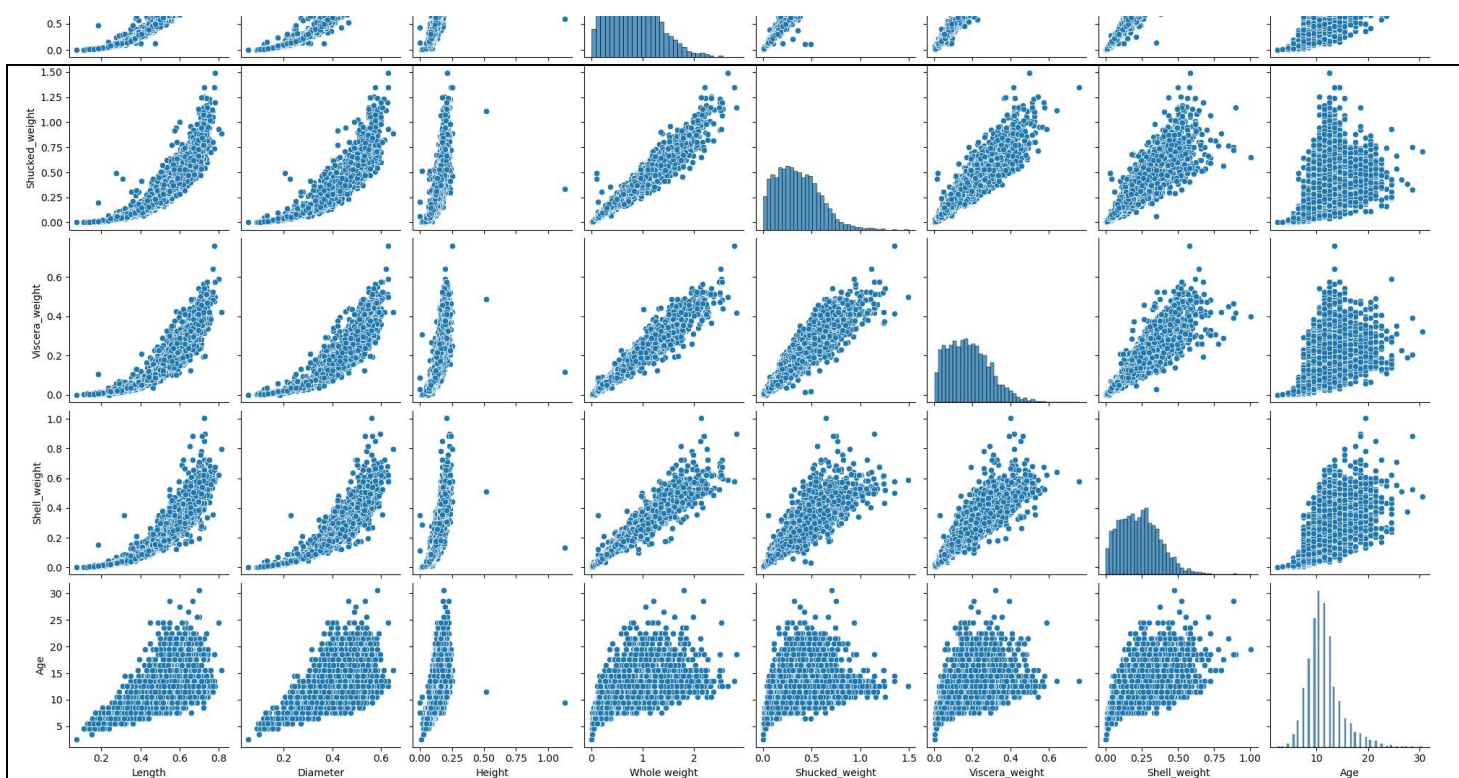
In [276]:

```
numerical_features = df.select_dtypes(include = [np.number]).columns
sns.pairplot(df[numerical_features])
```

Out[276]:

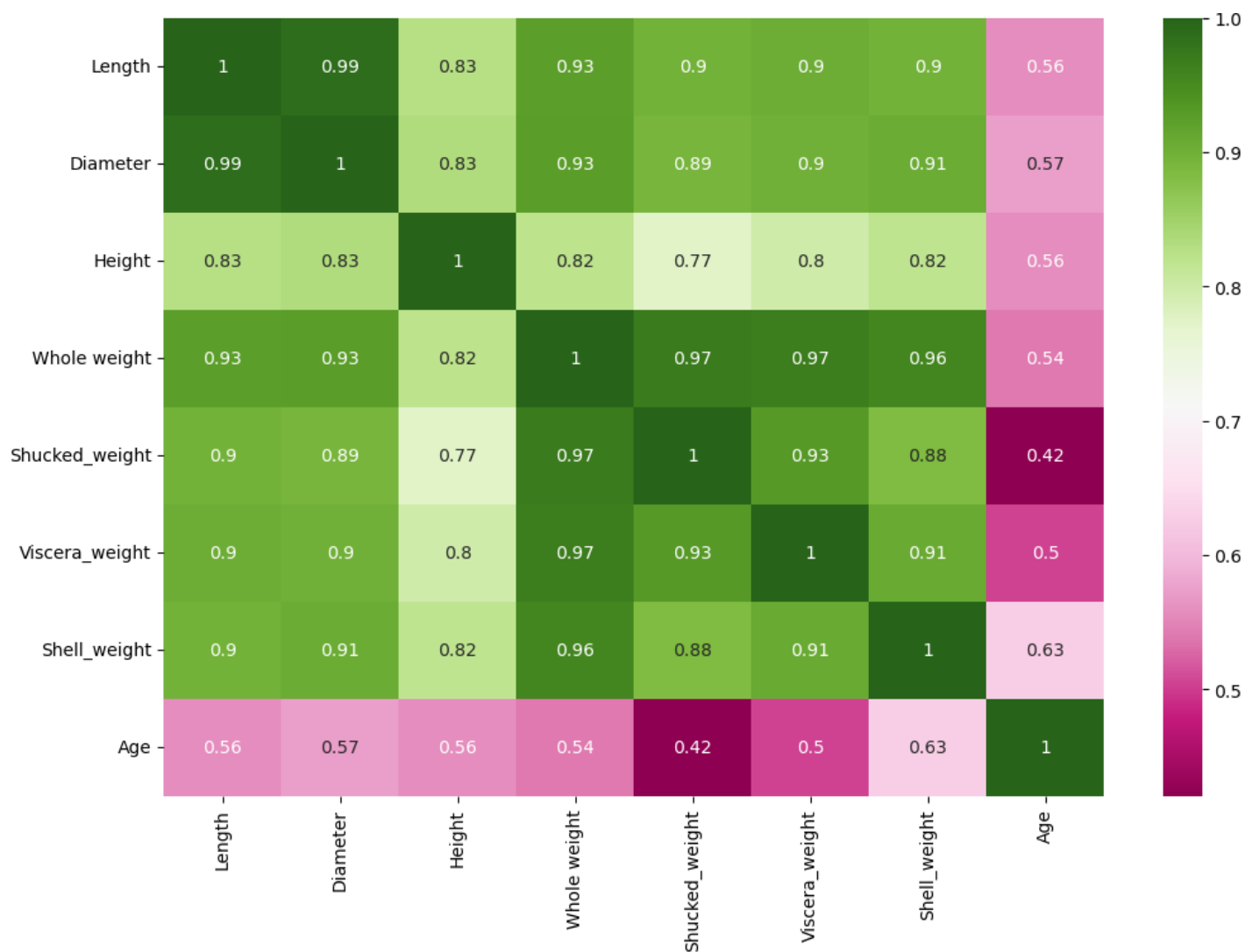
<seaborn.axisgrid.PairGrid at 0x1af61732d00>





In [277]:

```
plt.figure(figsize=(12,8));
sns.heatmap(df.corr(),cmap='PiYG',annot=True);
```



4. Perform descriptive statistics on the dataset

In [278]:

```
df.describe()
```

Out[278]:

	Length	Diameter	Height	Whole weight	Shucked_weight	Viscera_weight	Shell_weight	Age
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	11.433684
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	2.500000
25%	0.450000	0.350000	0.115000	0.441500	0.186000	0.093500	0.130000	9.500000
50%	0.545000	0.425000	0.140000	0.799500	0.336000	0.171000	0.234000	10.500000
75%	0.615000	0.480000	0.165000	1.153000	0.502000	0.253000	0.329000	12.500000
max	0.815000	0.650000	1.130000	2.825500	1.488000	0.760000	1.005000	30.500000

4.Check for Missing values and deal with them

In [279]:

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

Out[279]:

```
Sex          0
Length       0
Diameter     0
Height       0
Whole weight 0
Shucked_weight 0
Viscera_weight 0
Shell_weight 0
Age          0
dtype: int64
```

5.Find the outliers and replace them outliers

In [280]:

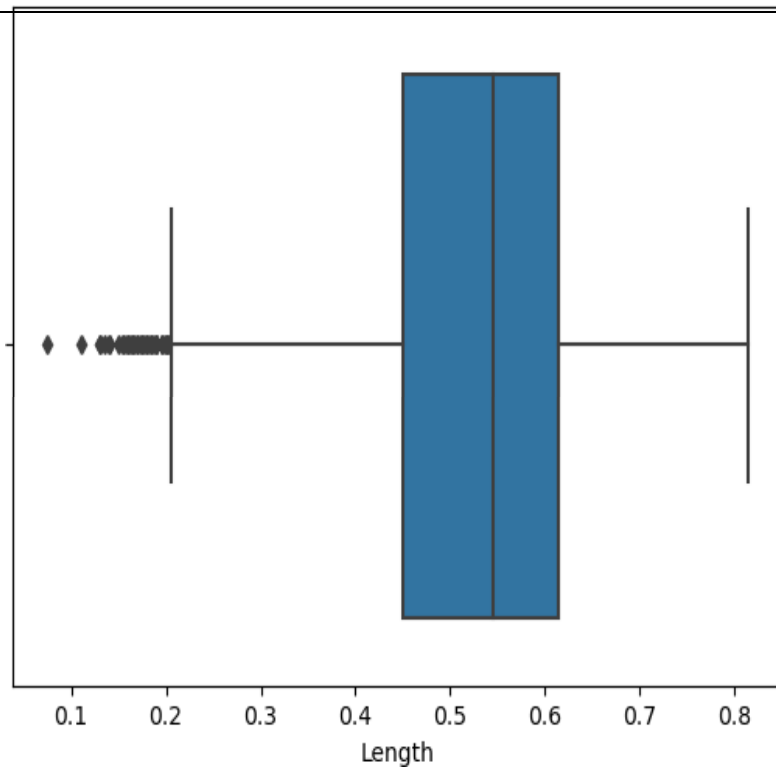
```
sns.boxplot(df['Length'])
```

C:\Users\ELCOT\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

```
warnings.warn(
```

Out[280]:

```
<AxesSubplot:xlabel='Length'>
```



In [281]:

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

Out[281]:

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

In [282]:

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

Out[282]:

```
(0.20250000000000004, 0.8624999999999999)
```

In [283]:

```
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 [284]:

```
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 [285]:

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

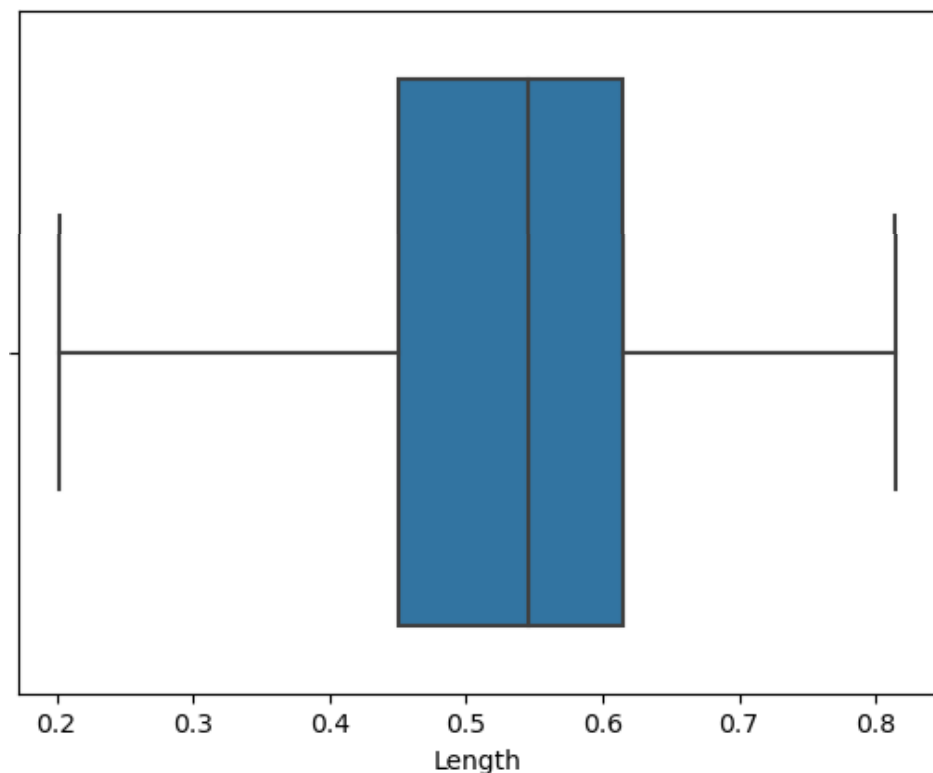
C:\Users\ELCOT\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

```
warnings.warn(
```



```
Out[285]:
```

```
<AxesSubplot:xlabel='Length'>
```



```
In [286]:
```

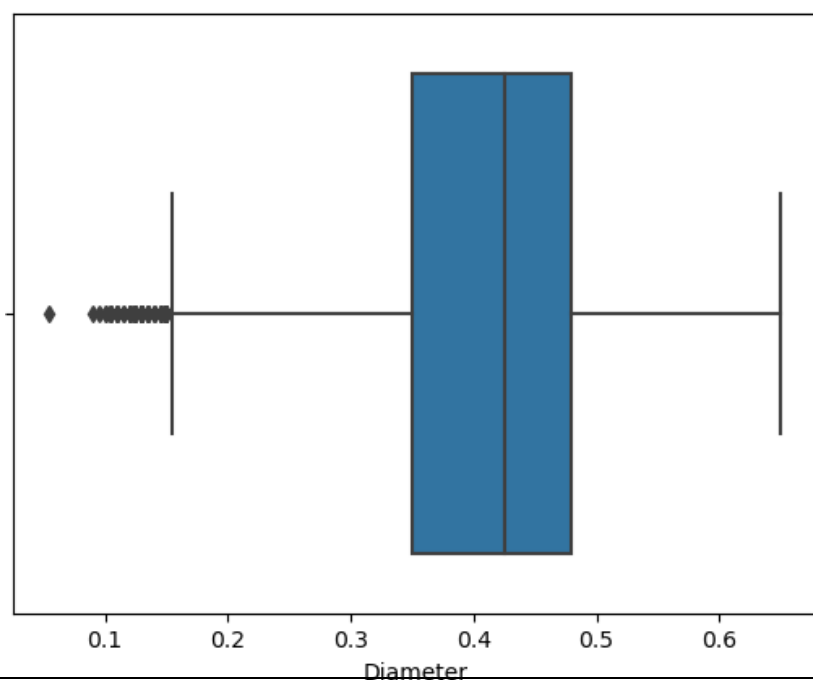
```
sns.boxplot(df['Diameter'])
```

C:\Users\ELCOT\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

```
warnings.warn(
```

```
Out[286]:
```

```
<AxesSubplot:xlabel='Diameter'>
```



In [287]:

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

Out[287]:

```
(0.35, 0.48, 0.13)
```

In [288]:

```
upper_limit = q2 (1.5*iqr)
lower_limit = q1 (1.5*iqr)
lower limit, upper limit
```

Out[288]:

```
(0.15499999999999997, 0.675)
```

In [289]:

```
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 : 4118
outliers : 59
```

In [290]:

```
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
```

In [291]:

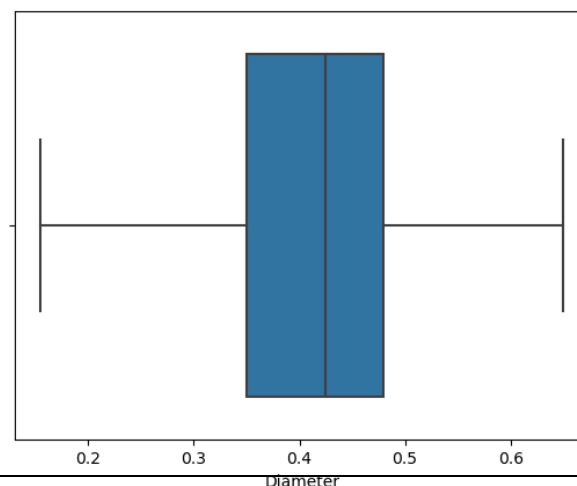
```
sns.boxplot(new_df['Diameter'])
```

C:\Users\ELCOT\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

```
warnings.warn(
```

Out[291]:

<AxesSubplot:xlabel='Diameter'>



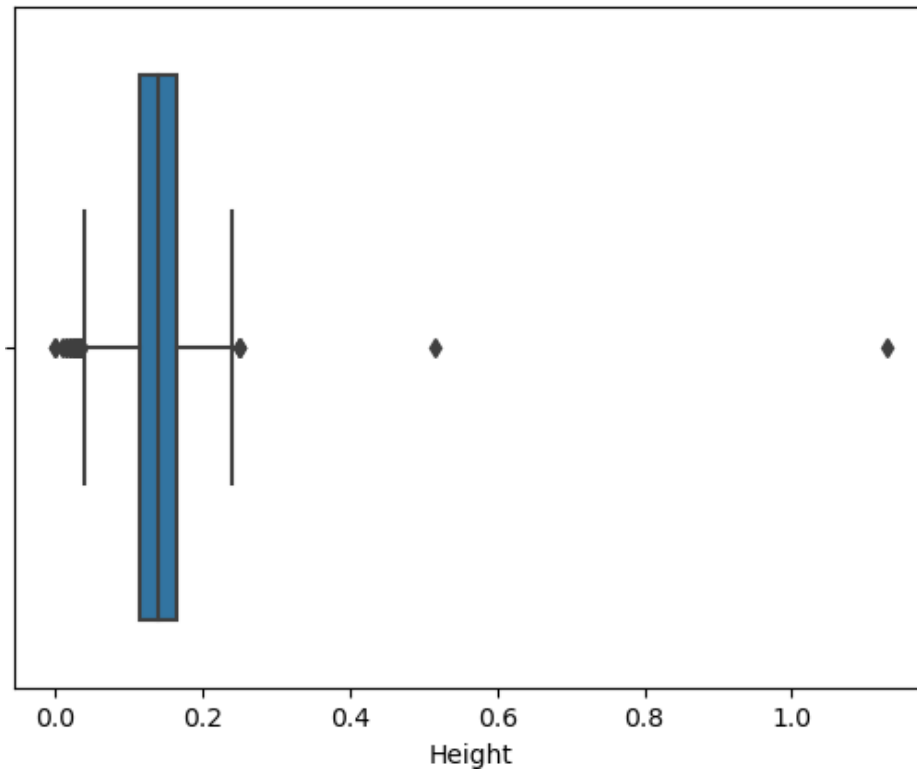
In [292]:

```
sns.boxplot(df['Height'])
```

C:\Users\ELCOT\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation

Out[292]:

<AxesSubplot:xlabel='Height'>



In [293]:

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

Out[293]:

(0.115, 0.165, 0.05)

In [294]:

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

Out[294]:

(0.039999999999999994, 0.24000000000000002)

In [295]:

```
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))
```

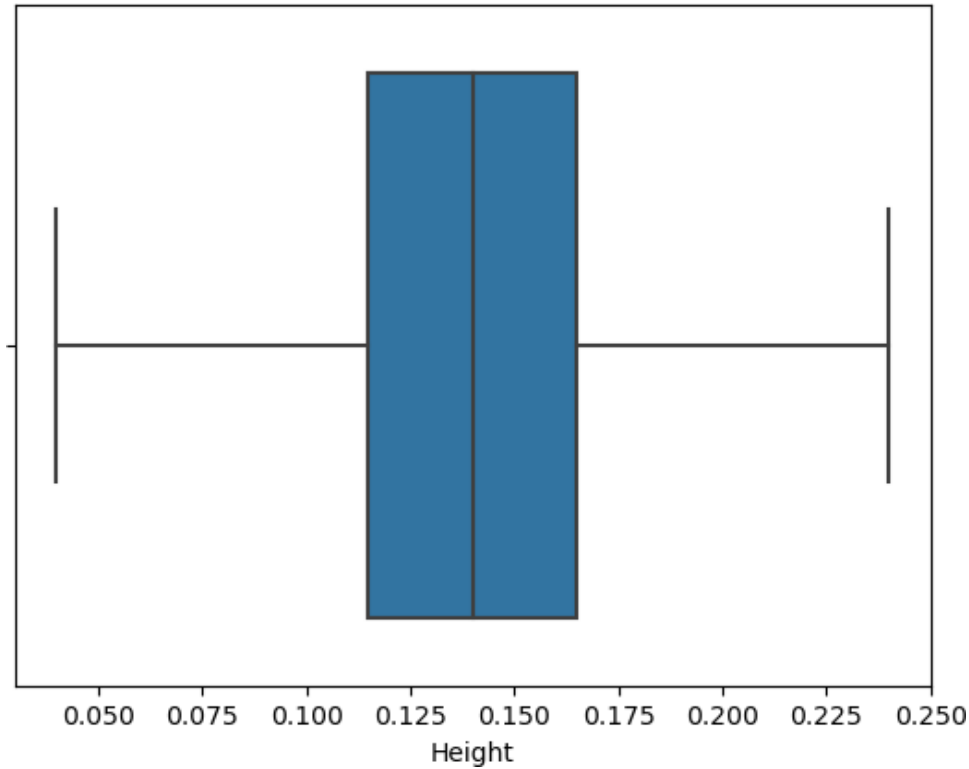
```
sns.boxplot(new_df['Height'])
```

C:\Users\ELCOT\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

```
warnings.warn(
```

Out[297]:

<AxesSubplot:xlabel='Height'>



In [298]:

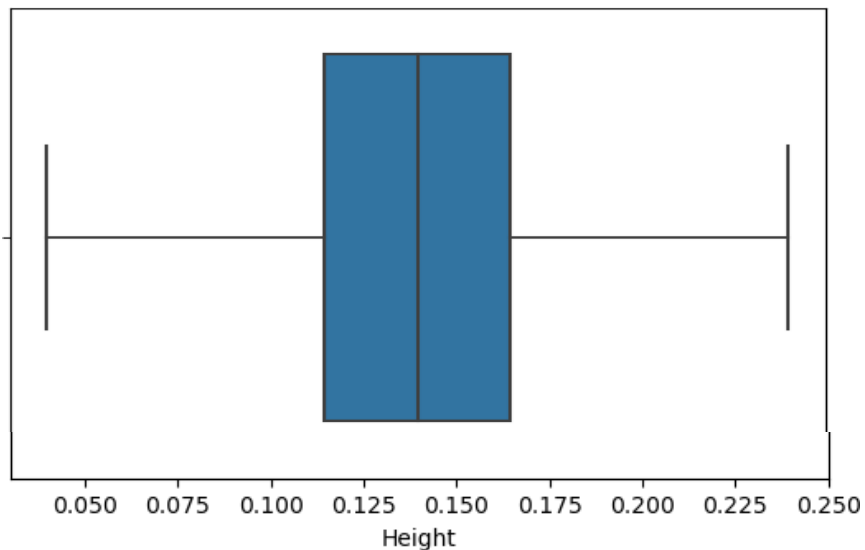
```
sns.boxplot(new_df['Height'])
```

C:\Users\ELCOT\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

```
warnings.warn(
```

Out[298]:

<AxesSubplot:xlabel='Height'>



In [299]:

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

Out[299]:

```
(0.4415, 1.153, 0.7115)
```

In [300]:

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

Out[300]:

```
(-0.62575, 2.22025)
```

In [301]:

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

```
before removing outliers : 4177
after removing outliers : 4147
outliers : 30
```

In [302]:

```
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
```

In [303]:

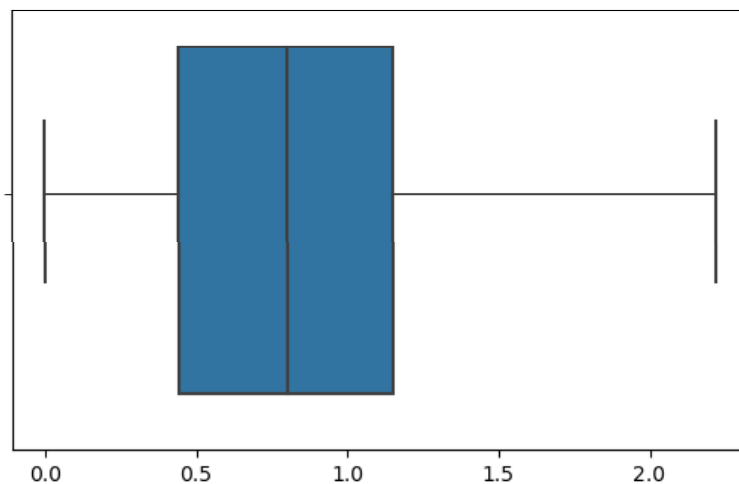
```
sns.boxplot(new_df['Whole weight'])
```

C:\Users\ELCOT\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

```
warnings.warn(
```

Out[303]:

```
<AxesSubplot:xlabel='Whole weight'>
```



In [304]:

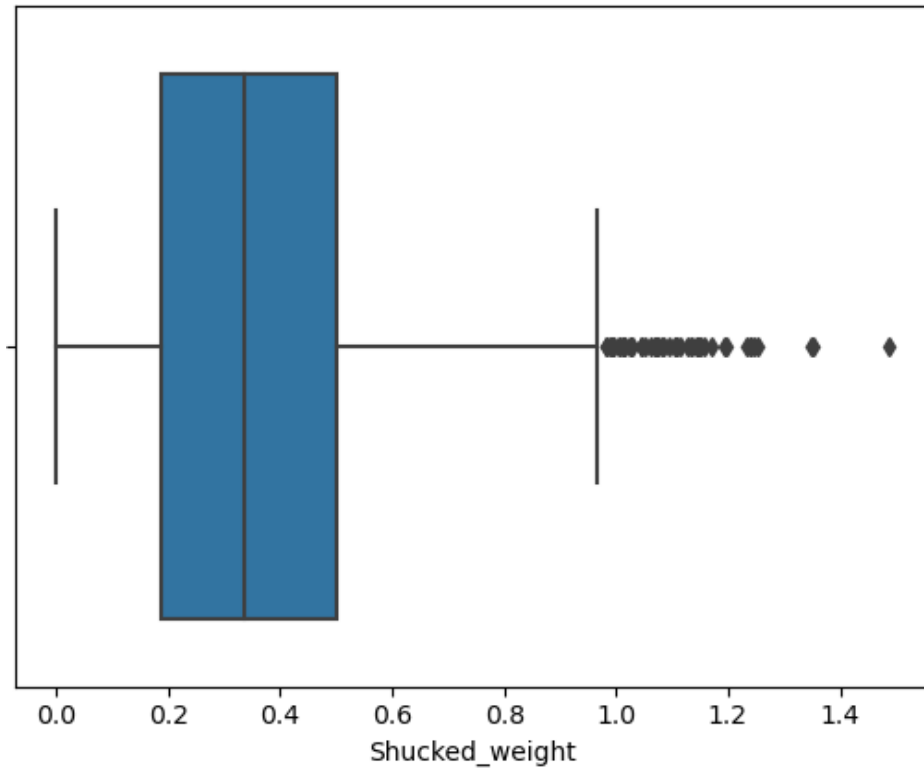
```
sns.boxplot(df['Shucked_weight'])
```

C:\Users\ELCOT\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

```
warnings.warn(
```

Out[304]:

<AxesSubplot:xlabel='Shucked_weight'>



In [305]:

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

Out[305]:

(0.186, 0.502, 0.316)

In [306]:

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

Out[306]:

(-0.288, 0.976)

In [307]:

```
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 : 4129  
outliers : 48
```

In [308]:

```
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
```

In [309]:

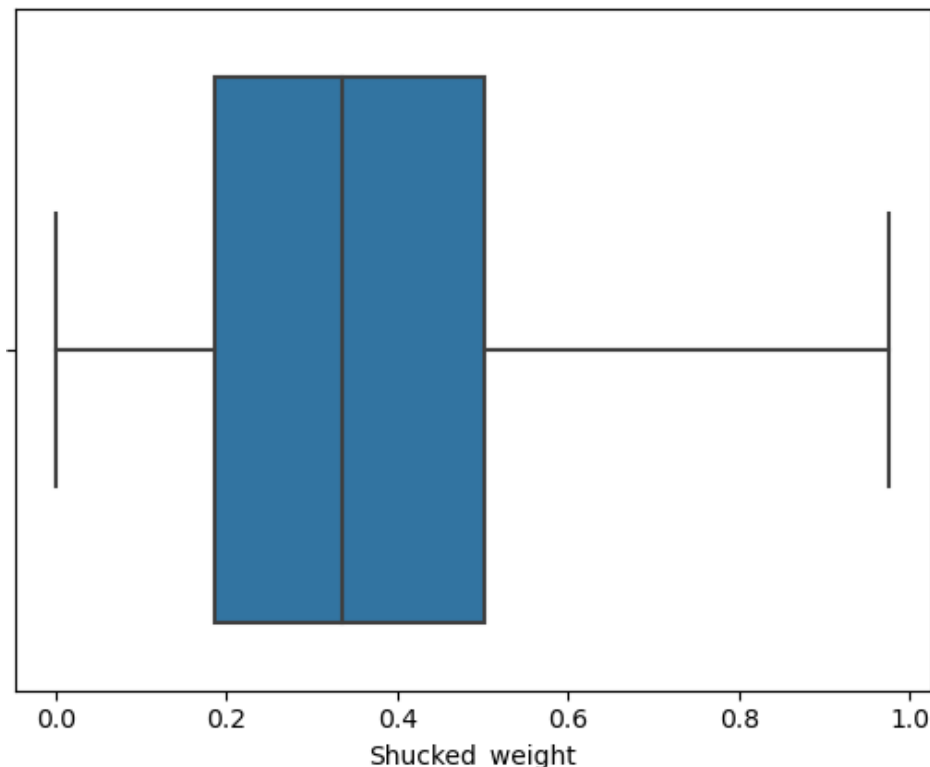
```
sns.boxplot(new_df['Shucked_weight'])
```

C:\Users\ELCOT\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

```
warnings.warn(
```

Out[309]:

<AxesSubplot:xlabel='Shucked_weight'>



In [310]:

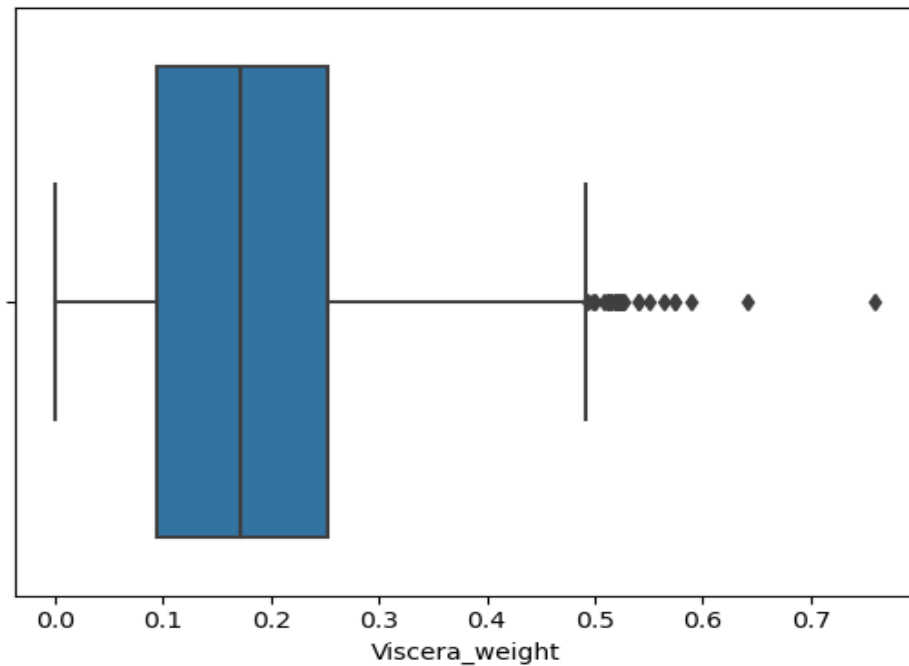
```
sns.boxplot(df['Viscera_weight'])
```

C:\Users\ELCOT\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

```
warnings.warn(
```

Out[310]:

<AxesSubplot:xlabel='Viscera_weight'>



6. Check for Categorical columns and perform encoding

In [311]:

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

Out[311]:

	Sex	Length	Diameter	Height	Whole weight	Shucked_weight	Viscera_weight	Shell_weight	Age
0	1	0.455	0.365	0.095	0.5140	0.2245	0.1010	0.1500	16.5
1	1	0.350	0.265	0.090	0.2255	0.0995	0.0485	0.0700	8.5
2	0	0.530	0.420	0.135	0.6770	0.2565	0.1415	0.2100	10.5
3	1	0.440	0.365	0.125	0.5160	0.2155	0.1140	0.1550	11.5
4	2	0.330	0.255	0.080	0.2050	0.0895	0.0395	0.0550	8.5
...
4172	0	0.565	0.450	0.165	0.8870	0.3700	0.2390	0.2490	12.5
4173	1	0.590	0.440	0.135	0.9660	0.4390	0.2145	0.2605	11.5
4174	1	0.600	0.475	0.205	1.1760	0.5255	0.2875	0.3080	10.5
4175	0	0.625	0.485	0.150	1.0945	0.5310	0.2610	0.2960	11.5
4176	1	0.710	0.555	0.195	1.9485	0.9455	0.3765	0.4950	13.5

4177 rows × 9 columns

In [312]:

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

In [313]:

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

Out[313]:

	Sex	Length	Diameter	Height	Whole weight	Shucked_weight	Viscera_weight	Shell_weight	Age
0	1	0.455	0.365	0.095	0.5140	0.2245	0.1010	0.1500	16.5
1	1	0.350	0.265	0.090	0.2255	0.0995	0.0485	0.0700	8.5
2	0	0.530	0.420	0.135	0.6770	0.2565	0.1415	0.2100	10.5
3	1	0.440	0.365	0.125	0.5160	0.2155	0.1140	0.1550	11.5
4	2	0.330	0.255	0.080	0.2050	0.0895	0.0395	0.0550	8.5
...
4172	0	0.565	0.450	0.165	0.8870	0.3700	0.2390	0.2490	12.5
4173	1	0.590	0.440	0.135	0.9660	0.4390	0.2145	0.2605	11.5
4174	1	0.600	0.475	0.205	1.1760	0.5255	0.2875	0.3080	10.5
4175	0	0.625	0.485	0.150	1.0945	0.5310	0.2610	0.2960	11.5
4176	1	0.710	0.555	0.195	1.9485	0.9455	0.3765	0.4950	13.5

4177 rows × 9 columns

In [314]:

```
enc = OneHotEncoder(drop='first')

enc_df = pd.DataFrame(enc.fit_transform(df[['Sex']]).toarray())

df = df.join(enc_df)

df.head()
```

Out[314]:

	Sex	Length	Diameter	Height	Whole weight	Shucked_weight	Viscera_weight	Shell_weight	Age	0	1
0	1	0.455	0.365	0.095	0.5140	0.2245	0.1010	0.150	16.5	1.0	0.0
1	1	0.350	0.265	0.090	0.2255	0.0995	0.0485	0.070	8.5	1.0	0.0
2	0	0.530	0.420	0.135	0.6770	0.2565	0.1415	0.210	10.5	0.0	0.0
3	1	0.440	0.365	0.125	0.5160	0.2155	0.1140	0.155	11.5	1.0	0.0
4	2	0.330	0.255	0.080	0.2050	0.0895	0.0395	0.055	8.5	0.0	1.0

7.Split the data into dependent and independent variables

In [315]:

```
x.info()
```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 4177 entries, 0 to 4176
Data columns (total 11 columns):
Column Non-Null Count Dtype
----- -----
0 Length 4177 non-null float64
1 Diameter 4177 non-null float64
2 Height 4177 non-null float64
3 Whole weight 4177 non-null float64
4 Shucked_weight 4177 non-null float64
5 Viscera_weight 4177 non-null float64
6 Shell_weight 4177 non-null float64
7 Age 4177 non-null float64
8 Sex_F 4177 non-null uint8
9 Sex_I 4177 non-null uint8
10 Sex M 4177 non-null uint8

dtypes: float64(8), uint8(3)

memory usage: 273.4 KB

```
In [316]:
```

```
X = x.drop(['Age'], axis = 1)
```

```
In [317]:
```

```
X.head(2)
```

```
Out[317]:
```

	Length	Diameter	Height	Whole weight	Shucked_weight	Viscera_weight	Shell_weight	Sex_F	Sex_I	Sex_M
0	0.455	0.365	0.095	0.5140	0.2245	0.1010	0.15	0	0	1
1	0.350	0.265	0.090	0.2255	0.0995	0.0485	0.07	0	0	1

```
In [318]:
```

```
y = x['Age']
```

```
In [319]:
```

```
y.head(2)
```

```
Out[319]:
```

```
0    16.5
1     8.5
Name: Age, dtype: float64
```

8. Scale the independent variables

```
In [320]:
```

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

```
print(scaledX)
```

```
[[-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]]
```

9. Split the data into training and testing

```
In [321]:
```

```
X.shape, y.shape
```

```
Out[321]:
```

```
((4177, 10), (4177,))
```

```
In [322]:
```

```
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=42)
```

In [323]:

```
print(' x_tain.shape : ',x_train.shape)
print(' y_tain.shape : ',y_train.shape)
print(' x_test.shape : ',x_test.shape)
print(' y_test.shape : ',y_test.shape)
```

```
x_tain.shape : (3341, 10)
y_tain.shape : (3341,)
x_test.shape : (836, 10)
y_test.shape : (836,)
```

10. Build the Model, 11. Train the Model , 12.Test the Model

In [324]:

```
from sklearn.linear_model import LinearRegression
lr = LinearRegression()
lr.fit(x_train, y_train)
lr_pred = lr.predict(x_test)
```

In [325]:

```
from sklearn.ensemble import RandomForestRegressor
from sklearn.metrics import mean_squared_error,make_scorer
from sklearn.model_selection import RandomizedSearchCV

rf = RandomForestRegressor()

param = {
    'max_depth':[3,6,9,12,15],
    'n_estimators':[10,50,100,150,200]
}
rf_search = RandomizedSearchCV(rf,param_distributions=param,n_iter=5,scoring=make_scorer
(mean_squared_error),n_jobs=-1,cv=5,verbose=3)
rf_search.fit(x_train, y_train)
```

Fitting 5 folds for each of 5 candidates, totalling 25 fits

Out[325]:

```
RandomizedSearchCV(cv=5, estimator=RandomForestRegressor(), n_iter=5, n_jobs=-1,
    param_distributions={'max_depth': [3, 6, 9, 12, 15],
        'n_estimators': [10, 50, 100, 150,
            200]},
    scoring=make_scorer(mean_squared_error), verbose=3)
```

In [326]:

```
means = rf_search.cv_results_['mean_test_score']
params = rf_search.cv_results_['params']
for mean, param in zip(means, params):
    print("%f with: %r" % (mean,param))
    if mean == min(means):
        print('Best parameters with the minimum Mean Square Error are:' ,param)

4.664623 with: {'n_estimators': 200, 'max_depth': 6}
4.618707 with: {'n_estimators': 100, 'max_depth': 15}
4.644619 with: {'n_estimators': 200, 'max_depth': 15}
5.677870 with: {'n_estimators': 150, 'max_depth': 3}
4.581780 with: {'n_estimators': 100, 'max_depth': 9}
Best parameters with the minimum Mean Square Error are: {'n_estimators': 100, 'max_depth': 9}
```

In [327]:

```
rf = RandomForestRegressor(n_estimators=50,max_depth=6)
rf.fit(x_train,y_train)

rf_pred = rf.predict(x_test)
```

14. Measure the performance using Metrics

In [328]:

```
from sklearn import metrics
print('Linear Regression :')
print('-----')
print('MAE:',metrics.mean_absolute_error(y_test, lr_pred))
print('MSE:',metrics.mean_squared_error(y_test, lr_pred))
print('RMSE:',np.sqrt(metrics.mean_squared_error(y_test, lr_pred)))
print('R2 Score:',metrics.r2_score(y_test,lr_pred))
print('\n\n')
```

Linear Regression :

MAE: 1.5944508821770336

MSE: 4.892375672262822

RMSE: 2.211871531591024

R2 Score: 0.5480572061259404

In [329]:

```
from sklearn import metrics
print('Random Forest Contains:')
print('-----')
print('MAE:',metrics.mean_absolute_error(y_test, rf_pred))
print('MSE:',metrics.mean_squared_error(y_test, rf_pred))
print('RMSE:',np.sqrt(metrics.mean_squared_error(y_test, rf_pred)))
print('R2 Score:',metrics.r2_score(y_test,rf_pred))
```

Random Forest Contains:

MAE: 1.5580369509719958

MSE: 5.025592967383406

RMSE: 2.241783434541215

R2 Score: 0.535750997326301