TEAM ID: PNT2022TMID31981

PROJECT TITLE: RETAIL STORE STOCK INVENTORY ANALYTICS

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STUDENT NAME: BHARATH KUMAR K

STUDENT ROLL NUMBER: 731619104009

1. Download the dataset

```
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</pre>
                                Sex Length Diameter Height Whole weight Shucked w
eight \
0
          0.455
                   0.365 0.095
                                         0.5140
                                                         0.2245
      Μ
1
         0.350
                    0.265 0.090
                                         0.2255
                                                        0.0995
      Μ
2
                           0.135
      F
          0.530
                    0.420
                                         0.6770
                                                        0.2565
                    0.365 0.125
3
         0.440
      Μ
                                         0.5160
                                                        0.2155
4
      I
          0.330
                    0.255
                            0.080
                                         0.2050
                                                         0.0895
                 0.450 0.165
. . .
           . . .
           0.565
                                         0.8870
                                                        0.3700
4172
     F
     M
4173
          0.590
                    0.440 0.135
                                         0.9660
                                                        0.4390
     M 0.600
                    0.475 0.205
                                         1.1760
                                                        0.5255
4174
     F
4175
          0.625
                    0.485 0.150
                                         1.0945
                                                        0.5310
                                         1.9485
4176
     M 0.710
                    0.555 0.195
                                                        0.9455
      Viscera weight Shell weight Rings
0
             0.1010
                           0.1500
                                      15
             0.0485
                           0.0700
                                       7
2
                           0.2100
                                      9
             0.1415
3
              0.1140
                           0.1550
                                      10
4
              0.0395
                           0.0550
                                       7
              0.2390
                           0.2490
4172
                                      11
4173
             0.2145
                           0.2605
                                      10
4174
                           0.3080
                                      9
              0.2875
                                      10
4175
              0.2610
                           0.2960
4176
             0.3765
                           0.4950
                                      12
[4177 \text{ rows x 9 columns}] >
```

```
In [268]:
```

Age=1.5+df.Rings

```
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	М	0.455	0.365	0.095	0.5140	0.2245	0.1010	0.150	16.5
1	М	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	М	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	М	0.590	0.440	0.135	0.9660	0.4390	0.2145	0.2605	11.5
4174	М	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	М	0.710	0.555	0.195	1.9485	0.9455	0.3765	0.4950	13.5

3. Perform Below Visualizations

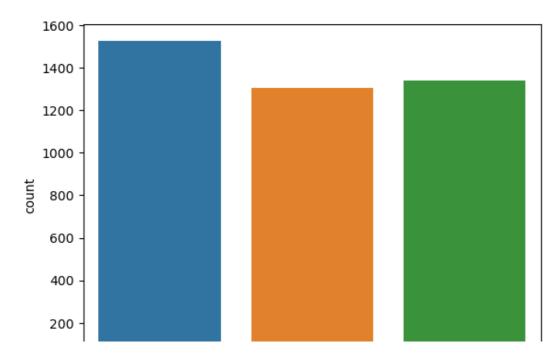
Univariate Analysis

```
In [270]:
```

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

Out[270]:

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



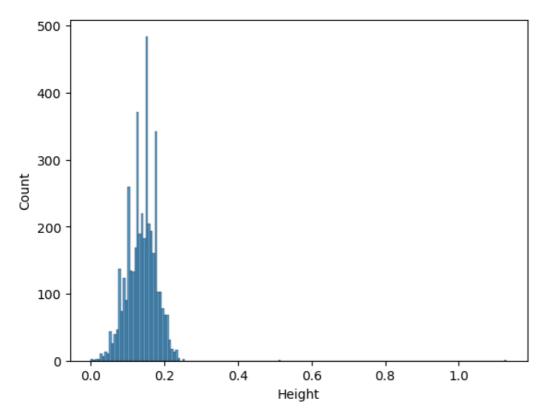
```
0 F I Sex
```

In [271]:

sns.histplot(df["Height"])

Out[271]:

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

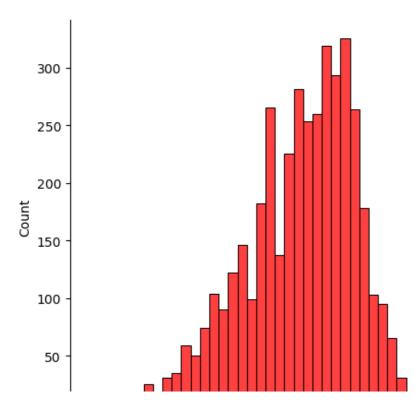


In [272]:

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

Out[272]:

<seaborn.axisgrid.FacetGrid at 0x1af5e2f7820>

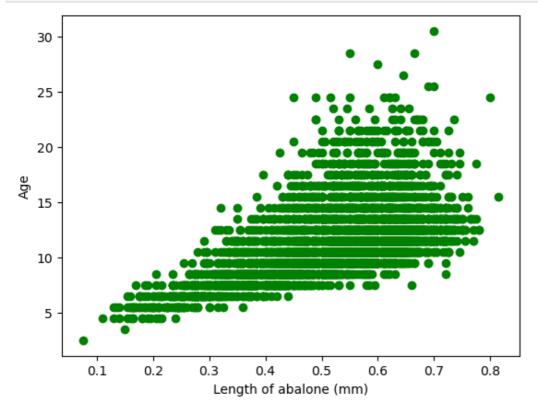


```
0.1 0.2 0.3 0.4 0.5 0.6 0.7 0.8
Length
```

Bi-Variate Analysis

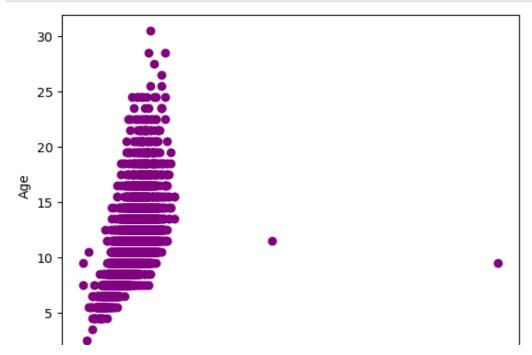
```
In [273]:
```

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



In [274]:

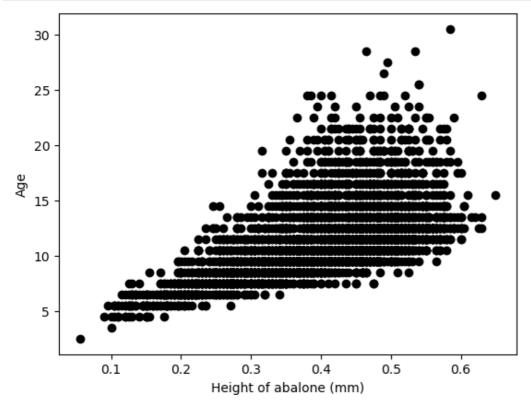
```
plt.scatter(df['Height'],df['Age'],c='purple')
plt.xlabel('Height of abalone (mm)')
plt.ylabel('Age')
plt.show()
```



```
0.0 0.2 0.4 0.6 0.8 1.0
Height of abalone (mm)
```

In [275]:

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



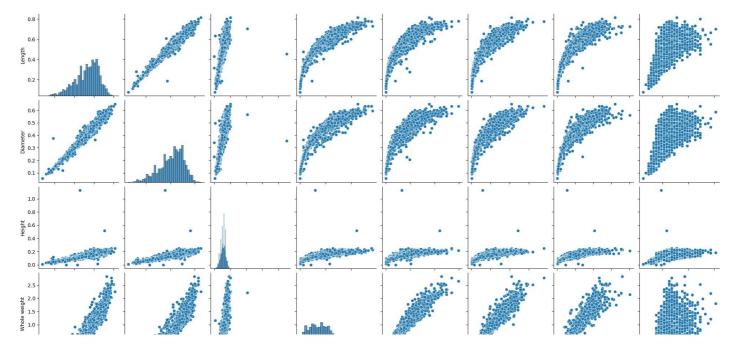
Multi-Variate Analysis

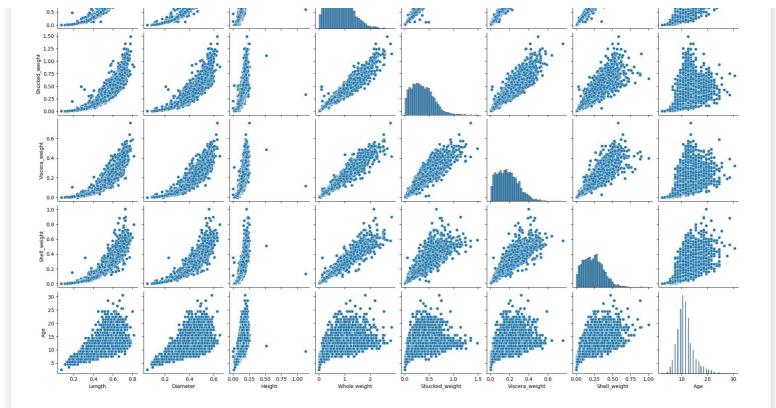
In [276]:

```
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 std min	0.523992	0.407881	0.139516	0.828742	0.359367	0.180594	0.238831	11.433684
	0.120093	0.099240	0.041827	0.490389	0.221963	0.109614	0.139203	3.224169
	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

5. Check for Missing values and deal with them

```
In [279]:
df.isnull().sum()
Out [279]:
Sex
Length
                   0
Diameter
                   0
Height
Whole weight
                   0
Shucked weight
                   0
Viscera weight
                   0
Shell weight
                   0
Age
dtype: int64
```

6. Find the outliers and replace them outliers

```
In [280]:
```

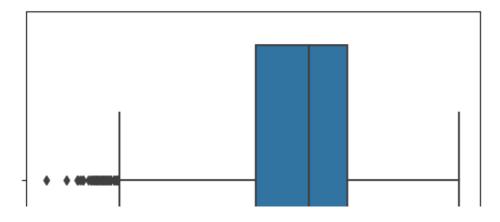
sns.boxplot(df['Length'])

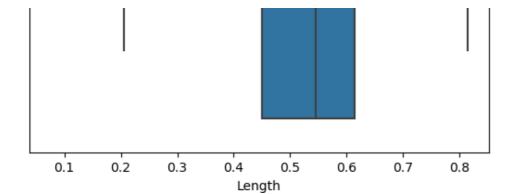
C:\Users\ELCOT\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

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, 0.16499999999999999)

In [282]:

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

Out[282]:

(0.20250000000000004, 0.862499999999999)

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</pre>
```

In [285]:

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

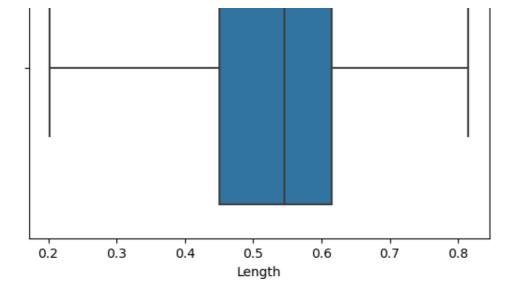
C:\Users\ELCOT\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

warnings.warn(

Out[285]:

<AxesSubplot:xlabel='Length'>





In [286]:

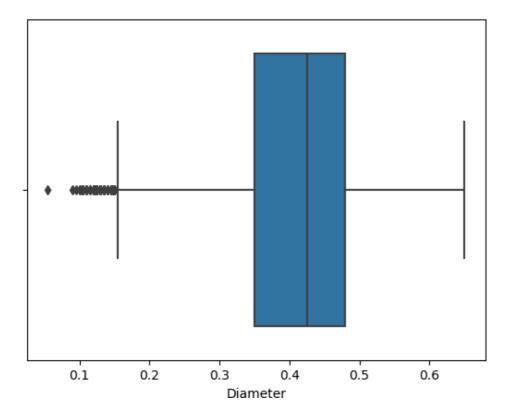
sns.boxplot(df['Diameter'])

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]:

(0.1549999999999997, 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</pre>
```

In [291]:

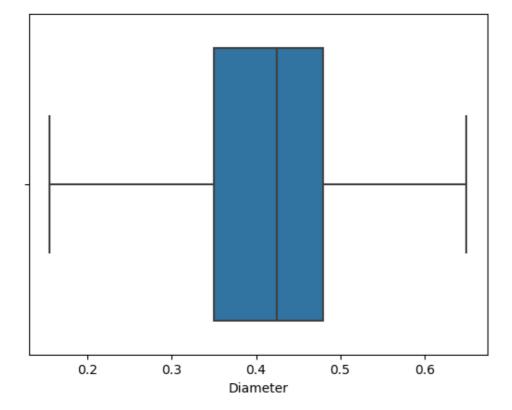
```
sns.boxplot(new df['Diameter'])
```

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 res

```
ult in an error or misinterpretation.
 warnings.warn(
Out[292]:
<AxesSubplot:xlabel='Height'>
   0.0
           0.2
                    0.4
                             0.6
                                     0.8
                                              1.0
                          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]:
In [295]:
```

new df = df.loc[(df['Height'] <= upper limit) & (df['Height'] >= lower limit)]

new_df.loc[(new_df['Height']>upper_limit), 'Height'] = upper_limit
new_df.loc[(new_df['Height']<lower_limit), 'Height'] = lower_limit</pre>

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 : 4148

outliers: 29

new df = df.copy()

In [296]:

In [297]:

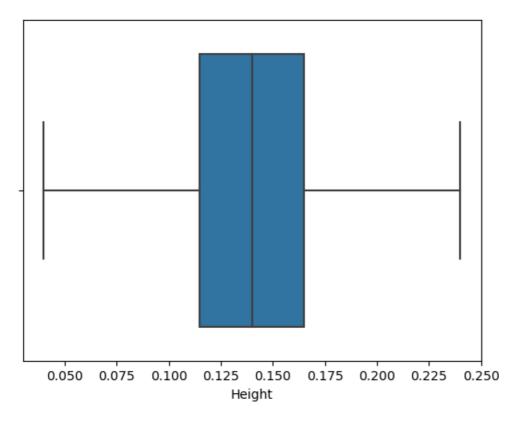
```
sns.boxplot(new_df['Height'])
```

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]:

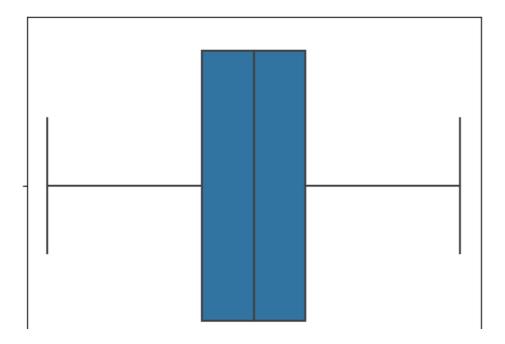
sns.boxplot(new df['Height'])

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.

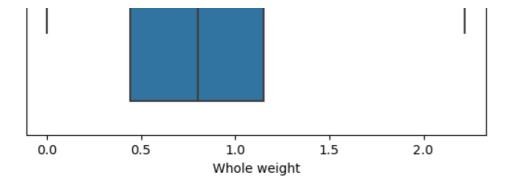
warnings.warn(

Out[298]:

<AxesSubplot:xlabel='Height'>



```
In [299]:
q1 = df['Whole weight'].quantile(0.25)
q2 = df['Whole weight'].quantile(0.75)
iqr = q2-q1
q1, q2, iqr
Out[299]:
(0.4415, 1.153, 0.7115)
In [300]:
upper limit = q2 + (1.5*iqr)
lower limit = q1 - (1.5*iqr)
lower_limit, upper_limit
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 :', len(new df))
print('outliers :', len(df)-len(new_df))
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</pre>
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 res
ult in an error or misinterpretation.
  warnings.warn(
Out[303]:
<AxesSubplot:xlabel='Whole weight'>
```



In [304]:

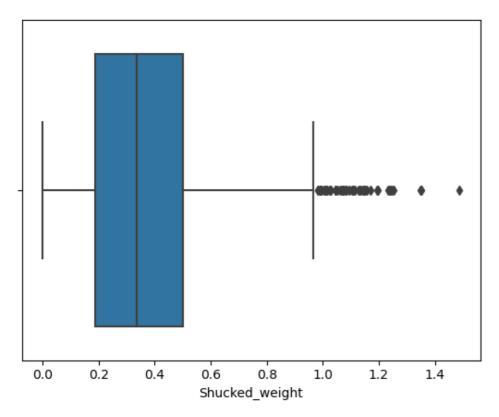
```
sns.boxplot(df['Shucked_weight'])
```

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_l
imit)]
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</pre>
```

In [309]:

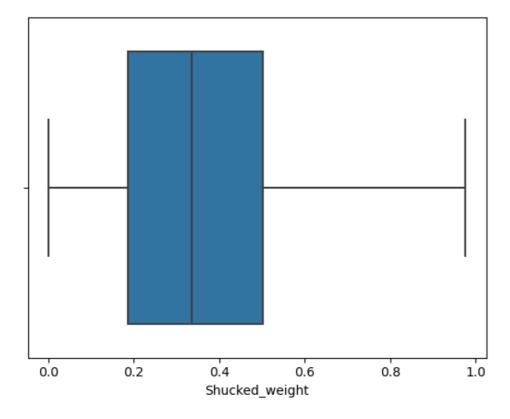
```
sns.boxplot(new df['Shucked weight'])
```

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]:

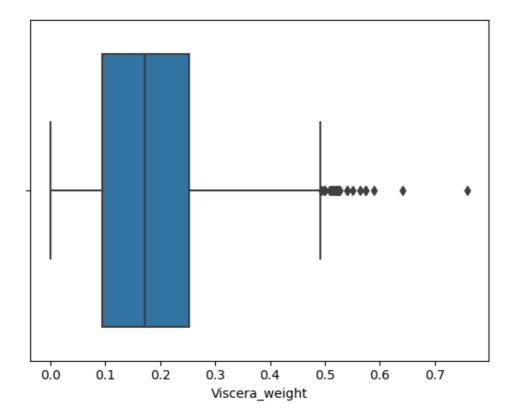
```
sns.boxplot(df['Viscera weight'])
```

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'>
```



7. 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

memory usage: 273.4 KB

```
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

8. 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
                    4177 non-null
 Length
                                    float64
  Diameter
                   4177 non-null float64
1
  Height
                   4177 non-null float64
2
                    4177 non-null float64
  Whole weight
  Shucked_weight 4177 non-null float64
Viscera_weight 4177 non-null float64
  Shell_weight 4177 non-null float64
6
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)
```

```
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
   0.455
              0.365
                    0.095
                                0.5140
                                               0.2245
                                                             0.1010
                                                                           0.15
                                                                                                 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]:
\cap
     16.5
1
      8.5
Name: Age, dtype: float64
```

9. Scale the independent variables

```
In [320]:
scale = StandardScaler()
scaledX = scale.fit transform(x)
print(scaledX)
[-0.57455813 - 0.43214879 - 1.06442415 \dots -0.67483383 - 0.68801788]
 1.31667716]
                -1.18397831 ... -0.67483383 -0.68801788
[-1.44898585 -1.439929
 1.316677161
-0.759487621
          0.67640943 1.56576738 ... -0.67483383 -0.68801788
[ 0.6329849
  1.31667716]
-0.759487621
1.31667716]]
```

10. 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]:
```

Linear Regression:

MAE: 1.5944508821770336 MSE: 4.892375672262822 RMSE: 2.211871531591024 R2 Score: 0.5480572061259404

In [329]:

Random Forest Contains:

MAE: 1.5580369509719958 MSE: 5.025592967383406 RMSE: 2.241783434541215 R2 Score: 0.535750997326301