

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

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

```
df.head()
```

	Sex	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight \
0	M	0.455	0.365	0.095	0.5140	0.2245	0.1010
1	M	0.350	0.265	0.090	0.2255	0.0995	0.0485
2	F	0.530	0.420	0.135	0.6770	0.2565	0.1415
3	M	0.440	0.365	0.125	0.5160	0.2155	0.1140
4	I	0.330	0.255	0.080	0.2050	0.0895	0.0395

	Shell weight	Rings
0	0.150	15
1	0.070	7
2	0.210	9
3	0.155	10
4	0.055	7

```
df.tail()
```

	Sex	Length	Diameter	Height	Whole weight	Shucked weight	\
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
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.isnull().any()
```

Sex	False
Length	False
Diameter	False
Height	False

```
Whole weight      False
Shucked weight    False
Viscera weight    False
Shell weight      False
Rings             False
dtype: bool
```

```
df.rename({'Rings': 'Age'}, axis=1, inplace=True)
```

```
df
```

	Sex	Length	Diameter	Height	Whole weight	Shucked weight	\
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	Age
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

```
[4177 rows x 9 columns]
```

```
hi= pd.Series(df.Age)
```

```
hi
```

0	15
1	7
2	9
3	10
4	7
...	..
4172	11
4173	10
4174	9

```

4175    10
4176    12
Name: Age, Length: 4177, dtype: int64

for i in hi:
    age=(hi+1.5)
age
0         16.5
1         8.5
2        10.5
3        11.5
4         8.5
...
4172     12.5
4173     11.5
4174     10.5
4175     11.5
4176     13.5
Name: Age, Length: 4177, dtype: float64

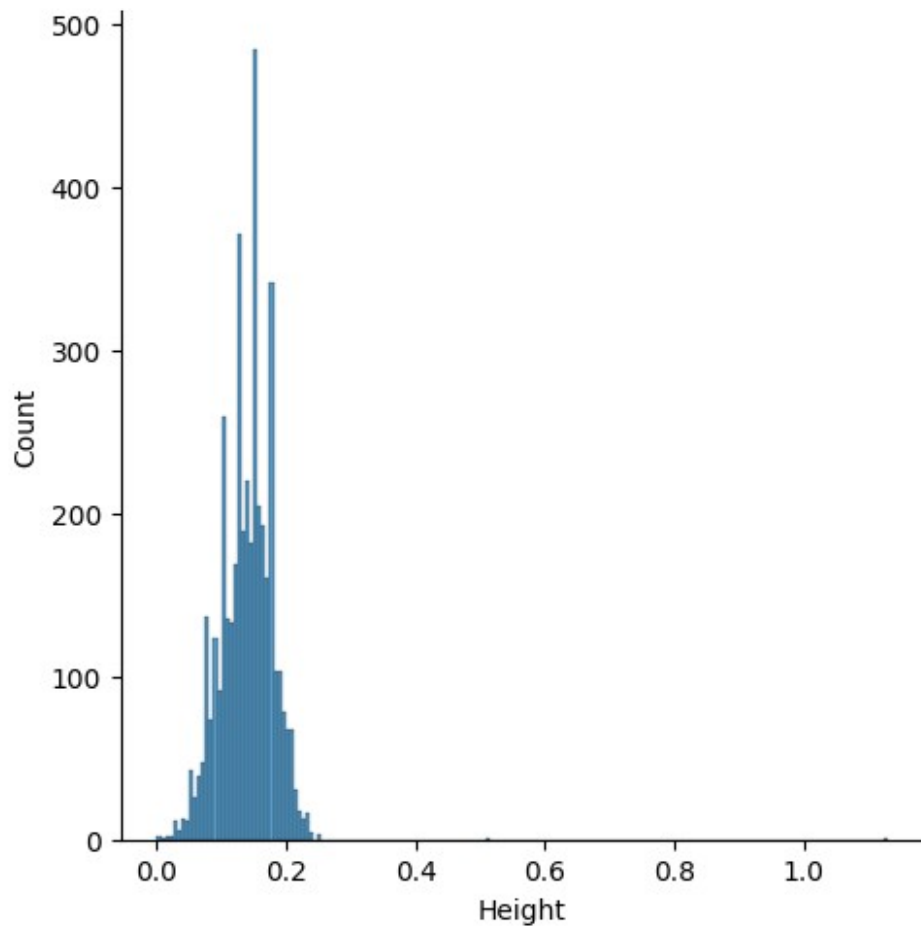
y=age
y
0         16.5
1         8.5
2        10.5
3        11.5
4         8.5
...
4172     12.5
4173     11.5
4174     10.5
4175     11.5
4176     13.5
Name: Age, Length: 4177, dtype: float64

```

## Univariate analysis

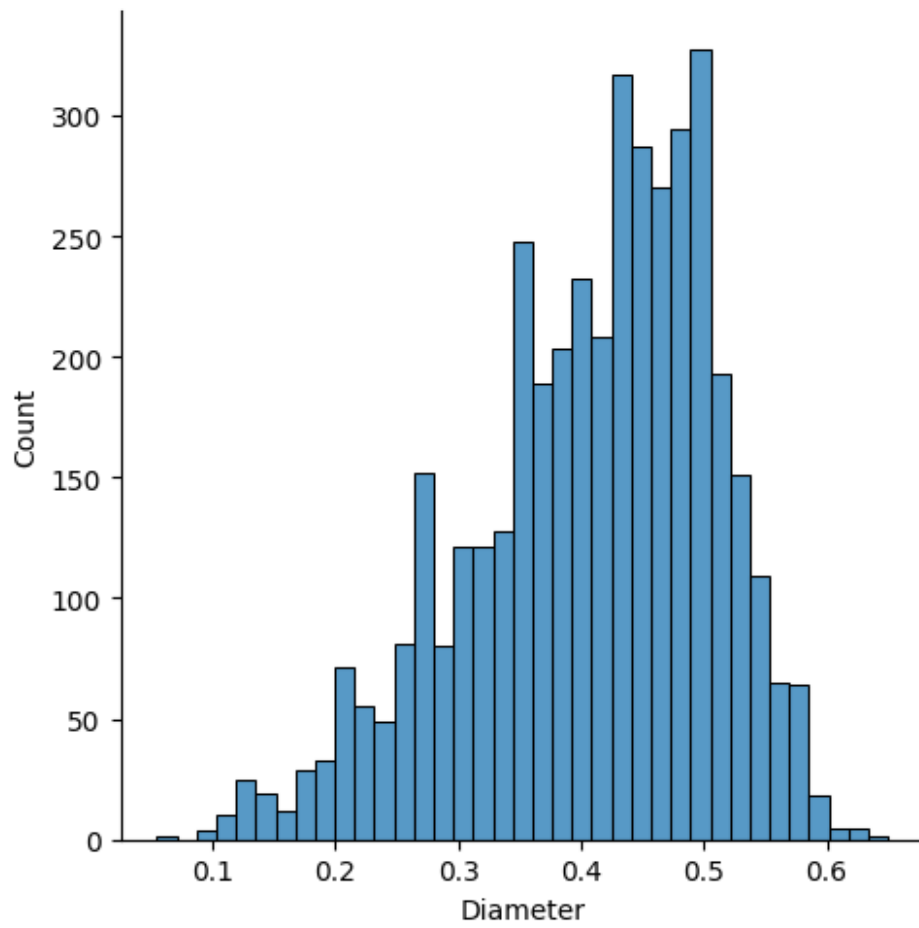
```
sns.displot(df.Height)
```

```
<seaborn.axisgrid.FacetGrid at 0x21f6c00ebc0>
```



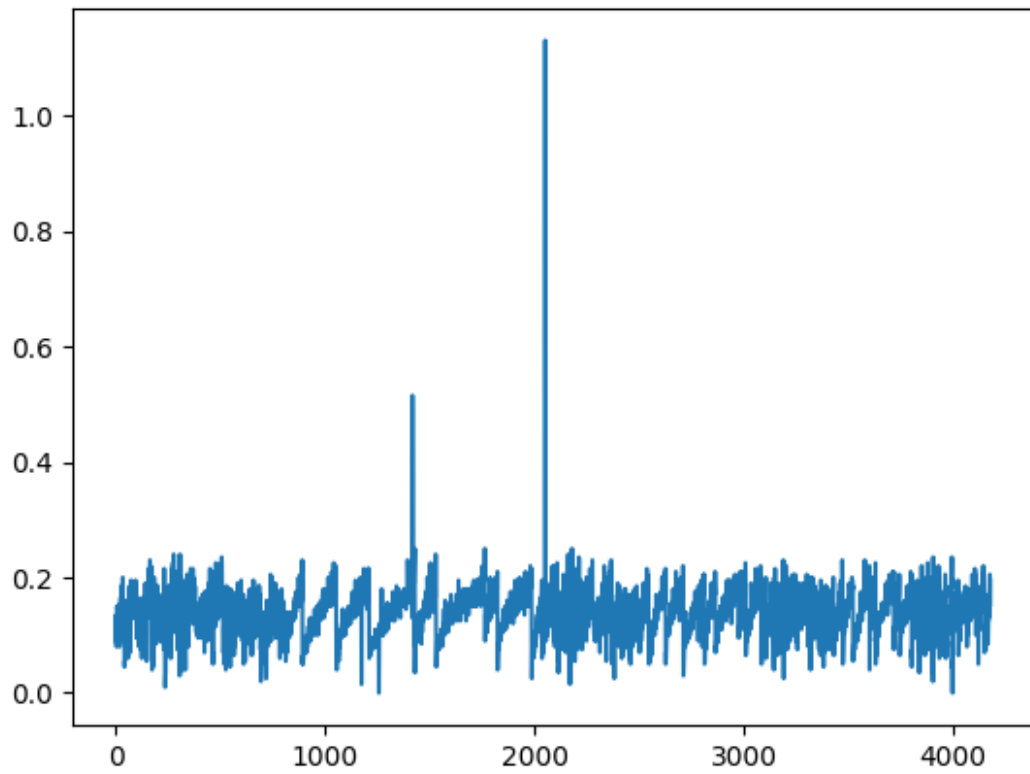
```
sns.displot(df.Diameter)
```

```
<seaborn.axisgrid.FacetGrid at 0x21f6e2cfd0>
```



```
df.Height.plot()
```

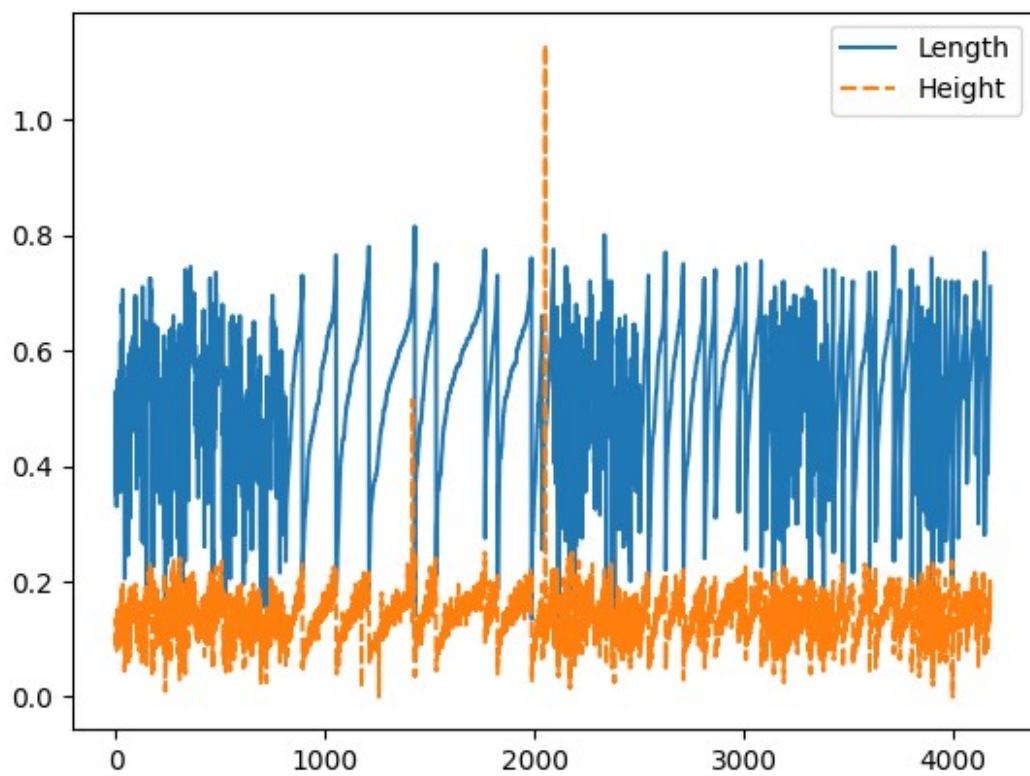
```
<AxesSubplot:>
```



### Bi-Variate Analysis

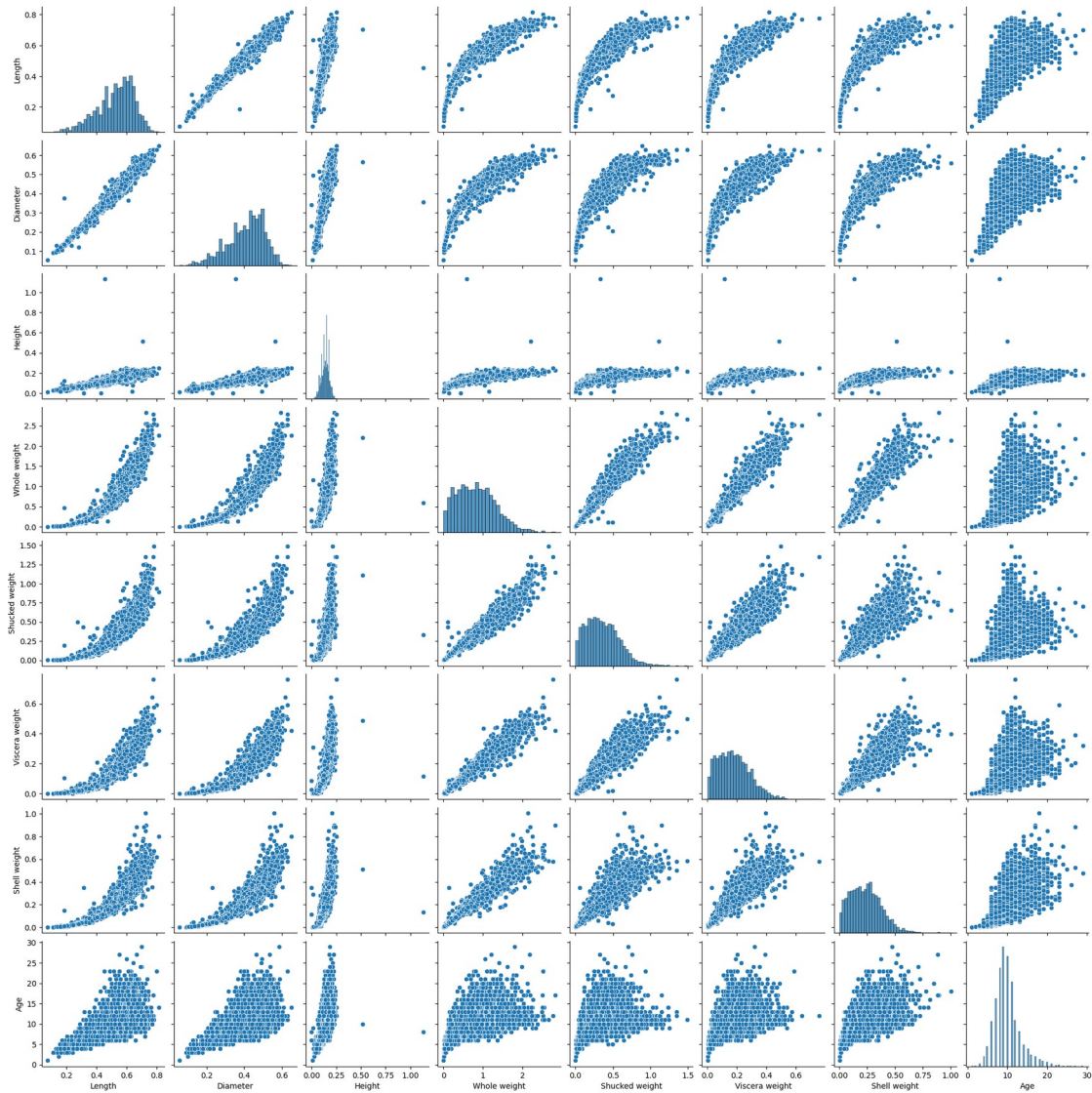
```
sns.lineplot([df.Length,df.Height])
```

<AxesSubplot:>



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

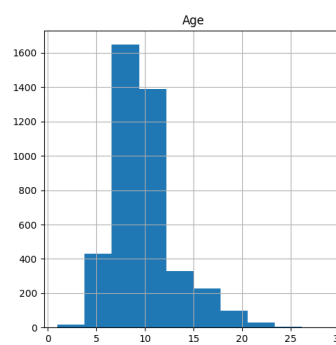
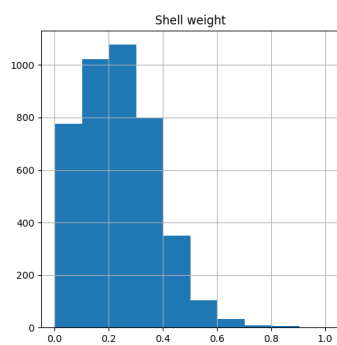
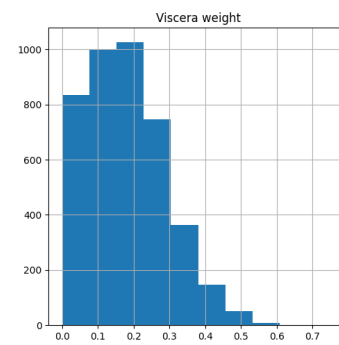
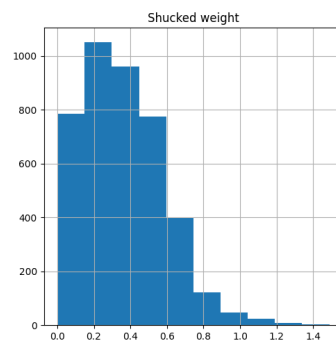
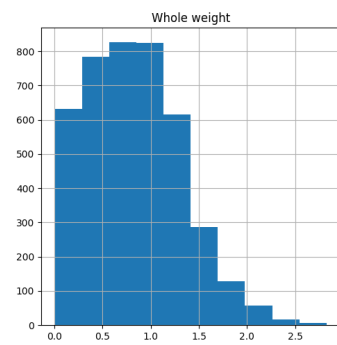
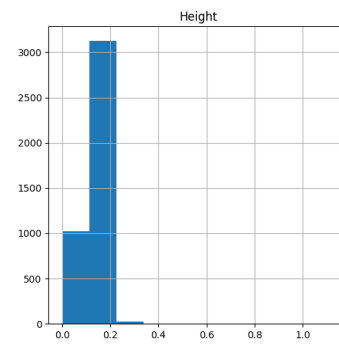
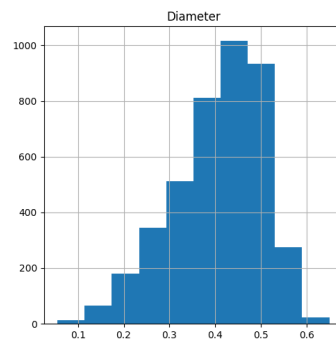
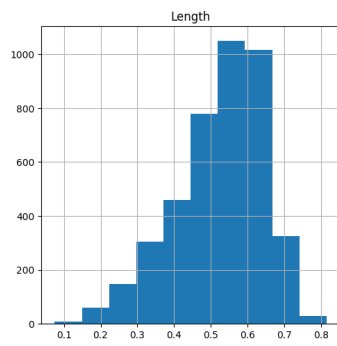
```
<seaborn.axisgrid.PairGrid at 0x21f6e61a4a0>
```



```
df.hist(figsize=(20,20))
```

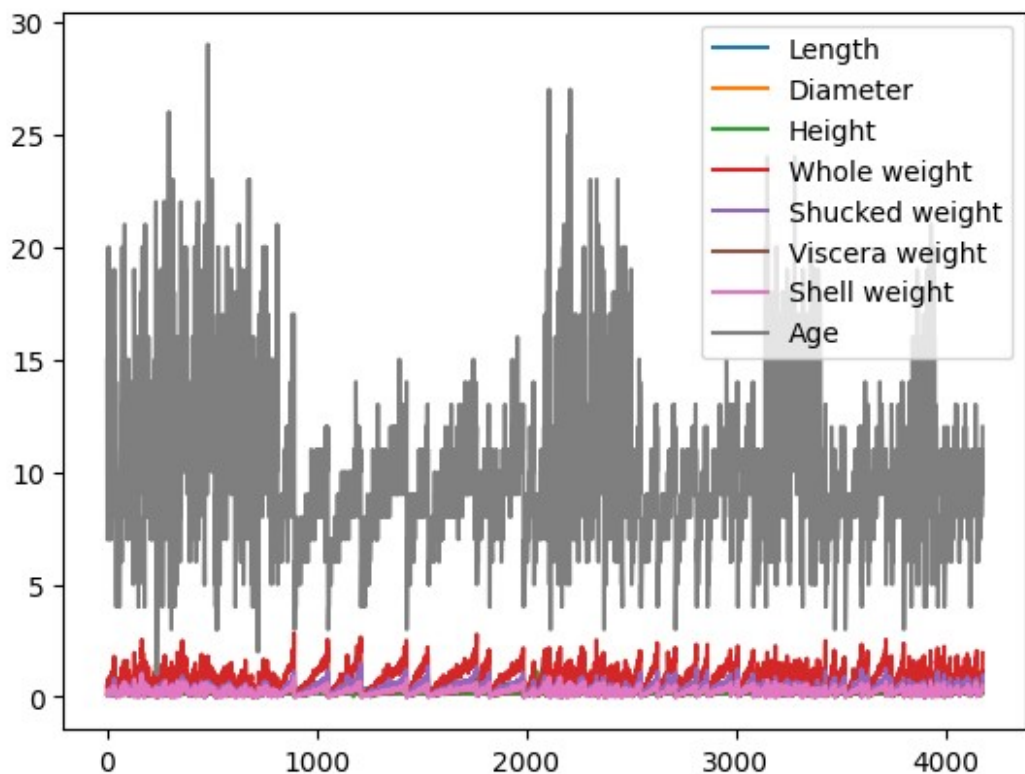
```
array([[<AxesSubplot:title={'center':'Length'}>,
        <AxesSubplot:title={'center':'Diameter'}>,
        <AxesSubplot:title={'center':'Height'}>],
       [<AxesSubplot:title={'center':'Whole weight'}>,
        <AxesSubplot:title={'center':'Shucked weight'}>,
        <AxesSubplot:title={'center':'Viscera weight'}>],
       [<AxesSubplot:title={'center':'Shell weight'}>,
        <AxesSubplot:title={'center':'Age'}>, <AxesSubplot:>]],
      dtype=object)
```





```
df.plot()
```

```
<AxesSubplot:>
```



```
df.describe()
```

	Length	Diameter	Height	Whole weight	Shucked
count	4177.000000	4177.000000	4177.000000	4177.000000	
mean	0.523992	0.407881	0.139516	0.828742	
std	0.120093	0.099240	0.041827	0.490389	
min	0.075000	0.055000	0.000000	0.002000	
25%	0.450000	0.350000	0.115000	0.441500	
50%	0.545000	0.425000	0.140000	0.799500	
75%	0.615000	0.480000	0.165000	1.153000	
max	0.815000	0.650000	1.130000	2.825500	

	Viscera weight	Shell weight	Age
count	4177.000000	4177.000000	4177.000000
mean	0.180594	0.238831	9.933684
std	0.109614	0.139203	3.224169
min	0.000500	0.001500	1.000000

25%	0.093500	0.130000	8.000000
50%	0.171000	0.234000	9.000000
75%	0.253000	0.329000	11.000000
max	0.760000	1.005000	29.000000

```
df.isnull().any()
```

```
Sex          False
Length       False
Diameter     False
Height       False
Whole weight False
Shucked weight False
Viscera weight False
Shell weight False
Age          False
dtype: bool
```

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

```
Sex          0
Length       0
Diameter     0
Height       0
Whole weight 0
Shucked weight 0
Viscera weight 0
Shell weight 0
Age          0
dtype: int64
```

```
df.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	Age	4177 non-null	int64

```
dtypes: float64(7), int64(1), object(1)
```

```
memory usage: 293.8+ KB
```

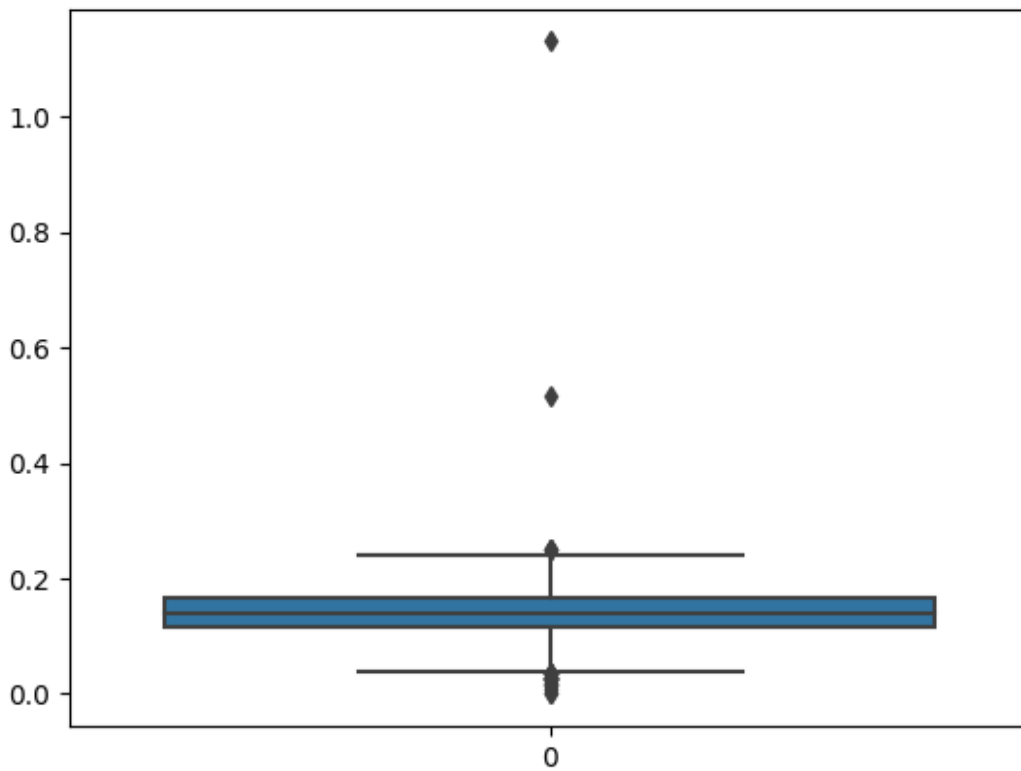
## Outlier Detection

```
df.shape
```

```
(4177, 9)
```

```
sns.boxplot(df.Height)
```

```
<AxesSubplot:>
```



```
q1=df.Height.quantile(0.25)
```

```
q3=df.Height.quantile(0.75)
```

```
IQR=q3-q1
```

```
upper_limit= q3 + 1.5*IQR
```

```
lower_limit= q1 - 1.5*IQR
```

```
upper_limit
```

```
0.24000000000000002
```

```
lower_limit
```

```
0.039999999999999994
```

```
df.median()
```

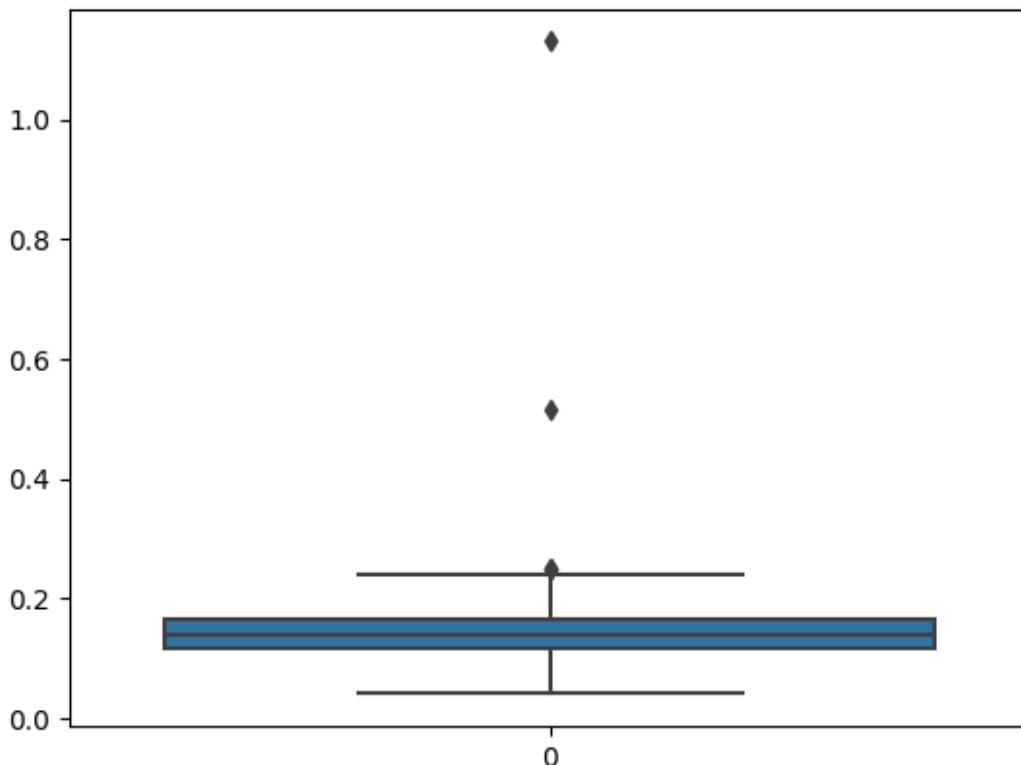
```
C:\Users\nojma\AppData\Local\Temp\ipykernel_21444\530051474.py:1:
FutureWarning: Dropping of nuisance columns in DataFrame reductions
(with 'numeric_only=None') is deprecated; in a future version this
will raise TypeError. Select only valid columns before calling the
reduction.
```

```
df.median()
```

```
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
Age              9.0000
dtype: float64
```

```
df['Height']=
np.where(df['Height']<lower_limit,0.039999999999999994 ,df['Height'])
sns.boxplot(df.Height)
```

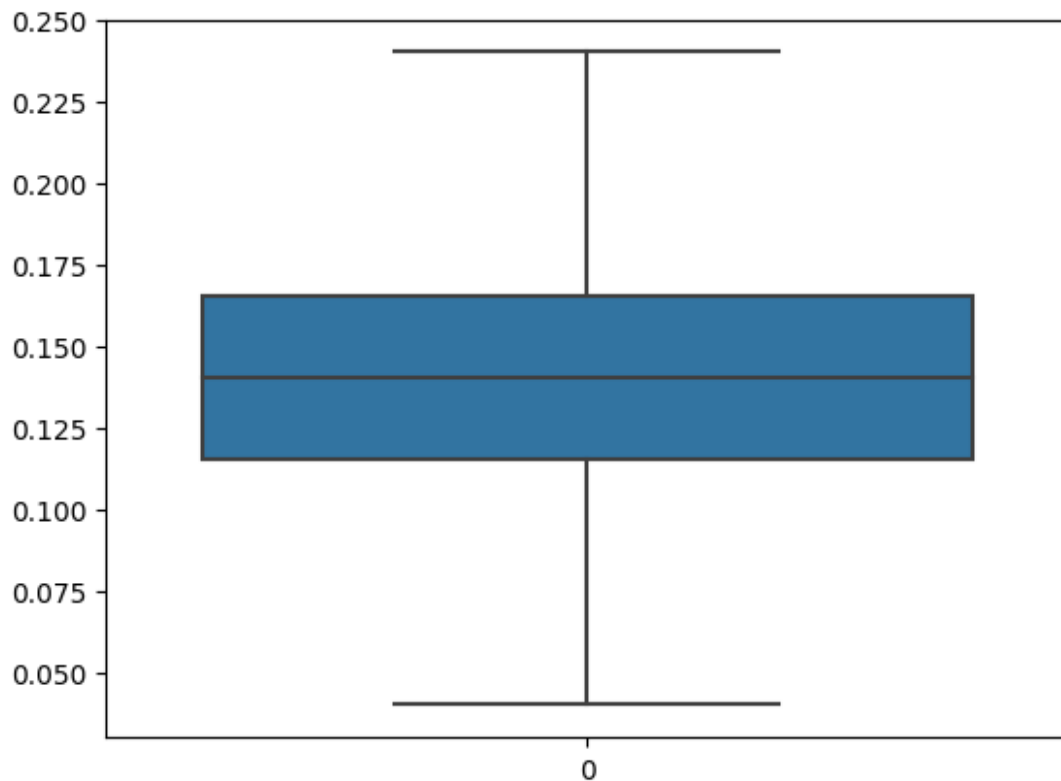
```
<AxesSubplot:>
```



```
df['Height']=
np.where(df['Height']>upper_limit,0.24000000000000002 ,df['Height'])
```

```
sns.boxplot(df.Height)
```

```
<AxesSubplot:>
```



```
df.shape
```

```
(4177, 9)
```

## The Categorical columns and perform Encoding.

```
df.head()
```

	Sex	Length	Diameter	Height	Whole weight	Shucked weight	Viscera
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	

Shell weight Age

```
0      0.150    15
1      0.070     7
2      0.210     9
3      0.155    10
4      0.055     7
```

```
df.Sex.value_counts()
```

```
M      1528
I      1342
F      1307
Name: Sex, dtype: int64
```

```
from sklearn.preprocessing import LabelEncoder
```

```
le=LabelEncoder()
```

```
df.Sex=le.fit_transform(df.Sex)
```

```
df.head()
```

	Sex	Length	Diameter	Height	Whole weight	Shucked weight	\
0	2	0.455	0.365	0.095	0.5140	0.2245	
1	2	0.350	0.265	0.090	0.2255	0.0995	
2	0	0.530	0.420	0.135	0.6770	0.2565	
3	2	0.440	0.365	0.125	0.5160	0.2155	
4	1	0.330	0.255	0.080	0.2050	0.0895	

	Viscera weight	Shell weight	Age
0	0.1010	0.150	15
1	0.0485	0.070	7
2	0.1415	0.210	9
3	0.1140	0.155	10
4	0.0395	0.055	7

```
df.tail()
```

	Sex	Length	Diameter	Height	Whole weight	Shucked weight	\
4172	0	0.565	0.450	0.165	0.8870	0.3700	
4173	2	0.590	0.440	0.135	0.9660	0.4390	
4174	2	0.600	0.475	0.205	1.1760	0.5255	
4175	0	0.625	0.485	0.150	1.0945	0.5310	
4176	2	0.710	0.555	0.195	1.9485	0.9455	

	Viscera weight	Shell weight	Age
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

## Split the Data into Dependent and Independent variables.

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

x

	Sex	Length	Diameter	Height	Whole weight	Shucked weight	\
0	2	0.455	0.365	0.095	0.5140	0.2245	
1	2	0.350	0.265	0.090	0.2255	0.0995	
2	0	0.530	0.420	0.135	0.6770	0.2565	
3	2	0.440	0.365	0.125	0.5160	0.2155	
4	1	0.330	0.255	0.080	0.2050	0.0895	
...	...	...	...	...	...	...	
4172	0	0.565	0.450	0.165	0.8870	0.3700	
4173	2	0.590	0.440	0.135	0.9660	0.4390	
4174	2	0.600	0.475	0.205	1.1760	0.5255	
4175	0	0.625	0.485	0.150	1.0945	0.5310	
4176	2	0.710	0.555	0.195	1.9485	0.9455	

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

[4177 rows x 8 columns]

y

0	16.5
1	8.5
2	10.5
3	11.5
4	8.5
...	...
4172	12.5
4173	11.5
4174	10.5
4175	11.5
4176	13.5

Name: Age, Length: 4177, dtype: float64

## Scale the independent variables

from sklearn.preprocessing import scale



```
x_scaled=pd.DataFrame(scale(x),columns=x.columns)
x_scaled.head()
```

	Sex	Length	Diameter	Height	Whole weight	Shucked
weight \						
0	1.151980	-0.574558	-0.432149	-1.158093	-0.641898	-
0.607685						
1	1.151980	-1.448986	-1.439929	-1.288751	-1.230277	-
1.170910						
2	-1.280690	0.050033	0.122130	-0.112828	-0.309469	-
0.463500						
3	1.151980	-0.699476	-0.432149	-0.374145	-0.637819	-
0.648238						
4	-0.064355	-1.615544	-1.540707	-1.550067	-1.272086	-
1.215968						

	Viscera weight	Shell weight
0	-0.726212	-0.638217
1	-1.205221	-1.212987
2	-0.356690	-0.207139
3	-0.607600	-0.602294
4	-1.287337	-1.320757

## Split the data into training and testing

```
from sklearn.model_selection import train_test_split
X_train,X_test,y_train,y_test =
train_test_split(x_scaled,y,test_size=0.2,random_state=0)
```

```
X_train.shape
```

```
(3341, 8)
```

```
y_train.shape
```

```
(3341,)
```

```
X_test.shape
```

```
(836, 8)
```

```
y_test.shape
```

```
(836,)
```

## Model building

```
from sklearn.linear_model import LinearRegression
```

```
model=LinearRegression()
```

```
model.fit(X_train,y_train)
```

```
LinearRegression()
```

## Testing

```
pred_test=model.predict(X_test)
```

```
pred_test
```

```
array([14.53244534, 11.0078981 , 11.70625188,  7.02813239,  
12.18242535,  
    13.35584542,  9.28710687, 11.42212394, 10.12653741,  
13.51180524,  
    10.27844483,  7.98154103,  9.59514319, 10.44626088,  
7.22370756,  
    10.68720552,  9.24091016, 15.75581247, 12.64239931,  
9.57397868,  
    9.21366024,  8.56592733, 10.72833811,  8.97344174,  
11.55892361,  
    13.04994427,  6.25948905, 14.78324032, 12.25371328,  
12.67496945,  
    9.47615295,  6.19887287, 12.80393235, 14.62478044,  
9.06495082,  
    10.18031923, 10.66594652, 11.78933452, 10.18198705,  
12.97094919,  
    13.42820172, 10.72507517, 13.5003741 , 13.48852193,  
13.96119858,  
    10.96541652, 10.64272246, 13.17887707, 13.85346053,  
9.69107776,  
    12.98216095,  8.91133409, 10.40251492, 15.06229965,  
10.60235007,  
    9.05207143,  7.62747477,  8.75816059,  9.02348271,  
8.76280894,  
    11.08732622, 10.56218347, 11.79619028,  9.63045715,  
9.82486158,  
    13.78750076, 14.09087368, 13.99322601, 10.64057633,  
15.83697549,  
    11.2933687 , 20.47350024, 12.43234857, 11.59001976,  
11.34988504,  
    10.82440164, 11.15896341, 11.20527522, 12.69403663,  
9.52758487,  
    10.85907235,  7.67114484,  9.09760133, 13.93585886,  
11.71409318,  
    9.85798751, 11.43937101, 14.11714606,  6.51672826,  
8.87371877,  
    11.87465528, 12.12501048,  9.52022006,  4.00964382,  
13.76887416,  
    7.8371364 , 12.21012604,  9.17569268, 16.11145446,  
10.98158594,  
    11.34284257, 13.69497776, 11.32164788, 12.42320552,  
7.15386018,
```

11.888862 , 9.18449066, 8.53466049, 9.38646006,  
14.97023344,  
10.5747346 , 12.5980386 , 12.81432213, 9.67064342,  
15.60552497,  
11.6603456 , 13.27570247, 14.7189357 , 6.84372814,  
11.60998243,  
8.86782202, 13.3099358 , 8.47605783, 11.09172386,  
12.3761562 ,  
14.25380571, 12.08405342, 12.32891244, 9.57995662,  
12.05635419,  
11.75046698, 9.42634719, 11.32432279, 12.7506026 ,  
12.73823653,  
12.50471301, 11.83921499, 10.92067913, 9.0738368 ,  
15.50115667,  
12.06204485, 13.03217064, 8.94525188, 9.67189824,  
13.03917867,  
11.75642295, 10.97080939, 9.14406821, 9.75675431,  
8.47569441,  
10.98926348, 18.92776394, 8.83827004, 12.32715664,  
9.21560814,  
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```

```

pred_train = model.predict(X_train)
pred_train

```

```

array([ 7.19974119,  6.84237887, 15.18412216, ..., 10.94694691,
        13.11797394,  9.98079993])

```



```
Age= pd.DataFrame({'Actual_Age':y_test,'pred_Age':pred_test})
Age
```

	Actual_Age	pred_Age
668	14.5	14.532445
1580	9.5	11.007898
3784	12.5	11.706252
463	6.5	7.028132
2615	13.5	12.182425
...	...	...
575	12.5	11.692160
3231	13.5	10.314746
1084	8.5	9.962249
290	18.5	13.896461
2713	5.5	7.202037

```
[836 rows x 2 columns]
```

## Performance using Metrics

```
from sklearn import metrics
```

```
# r2score
```

```
print(metrics.r2_score(y_test,pred_test)) # test accuracy
```

```
0.5418733297497837
```

```
print(metrics.r2_score(y_train,pred_train)) # train accuracy
```

```
0.531977913488948
```

```
#MSE
```

```
print(metrics.mean_squared_error(y_test,pred_test)) # test accuracy
```

```
4.975398170943831
```

```
print(metrics.mean_squared_error(y_train,pred_train)) # train
accuracy
```

```
4.807868180891602
```