

```
In [13]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from matplotlib import rcParams

In [14]: df=pd.read_csv('labalone.csv')
df.head()
```

	Sex	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	Rings
0	M	0.455	0.365	0.095	0.5140	0.2245	0.1010	0.150	15
1	M	0.350	0.265	0.090	0.2255	0.0995	0.0485	0.070	7
2	F	0.530	0.420	0.135	0.6770	0.2565	0.1415	0.210	9
3	M	0.440	0.365	0.125	0.5160	0.2155	0.1140	0.155	10
4	I	0.330	0.255	0.080	0.2050	0.0895	0.0395	0.055	7

```
In [15]: df.shape
Out[15]: (4177, 9)

In [16]: 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   Rings              4177 non-null   int64
dtypes: float64(7), int64(1), object(1)
memory usage: 293.8+ KB

In [17]: 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

In [18]: df.describe()
Out[18]:
```

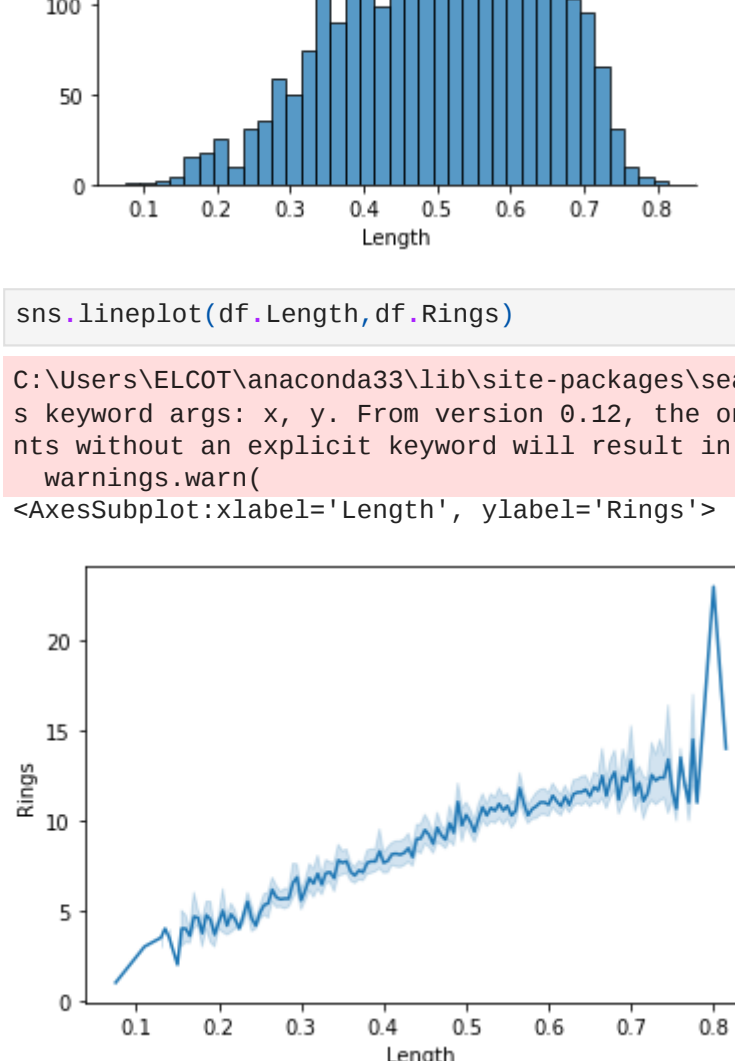
	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	Rings
count	4177.000000	4177.000000	4177.000000	4177.000000	4177.000000	4177.000000	4177.000000	4177.000000
mean	0.523992	0.407881	0.139516	0.828742	0.359367	0.180594	0.238831	9.933684
std	0.120093	0.099240	0.041827	0.490389	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

```
In [19]: df.Sex.value_counts()
Out[19]:
M    1528
I     1342
F     1307
Name: Sex, dtype: int64

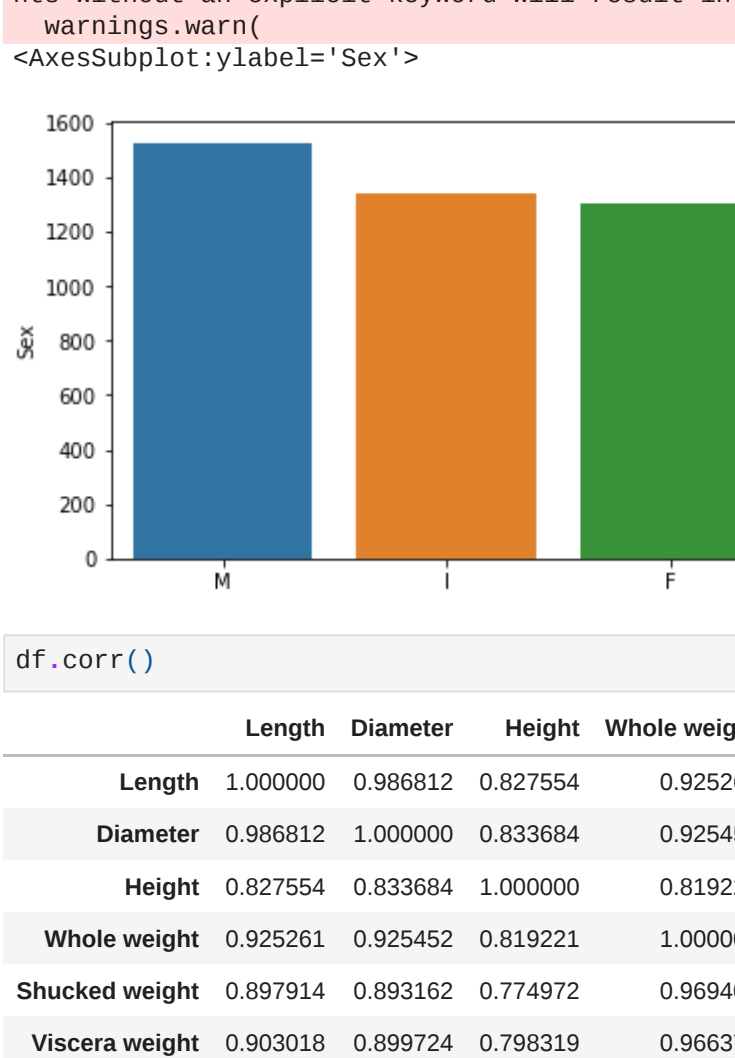
In [20]: df.Diameter.value_counts()
Out[20]:
0.450    139
0.475    129
0.490    111
0.500    110
0.470    190
...
0.610         1
0.650         1
0.620         1
0.695         1
0.615         1
Name: Diameter, dtype: int64

In [21]: df.Rings.value_counts()
Out[21]:
9      689
10     634
8      568
11     487
7      391
12     267
6      259
13     293
14     126
5      115
15     183
16      67
17      58
4       57
18      42
19      32
20      26
3       15
21      14
23       9
22       6
27       2
24       2
1         1
26       1
29       1
2         1
25       1
Name: Rings, dtype: int64

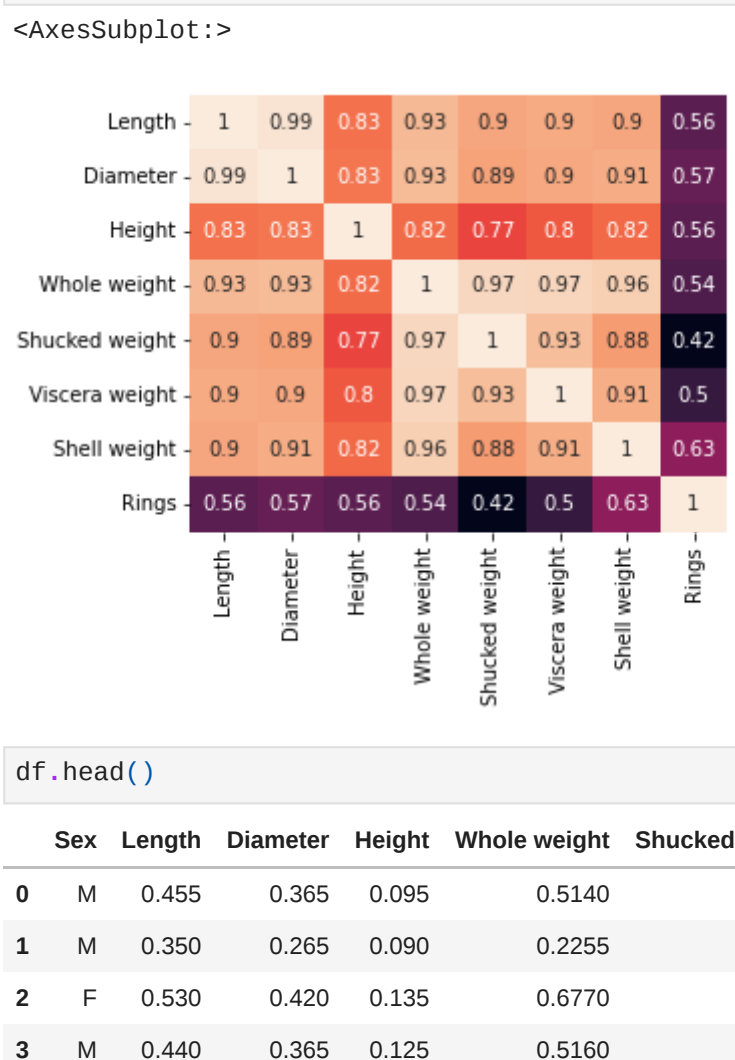
In [22]: sns.displot(df.Length)
Out[22]: <seaborn.axisgrid.FacetGrid at 0x218ae00d90>
```



```
In [23]: sns.lineplot(df.Length, df.Rings)
Out[23]: <AxesSubplot: xlabel='Length', ylabel='Rings'>
```



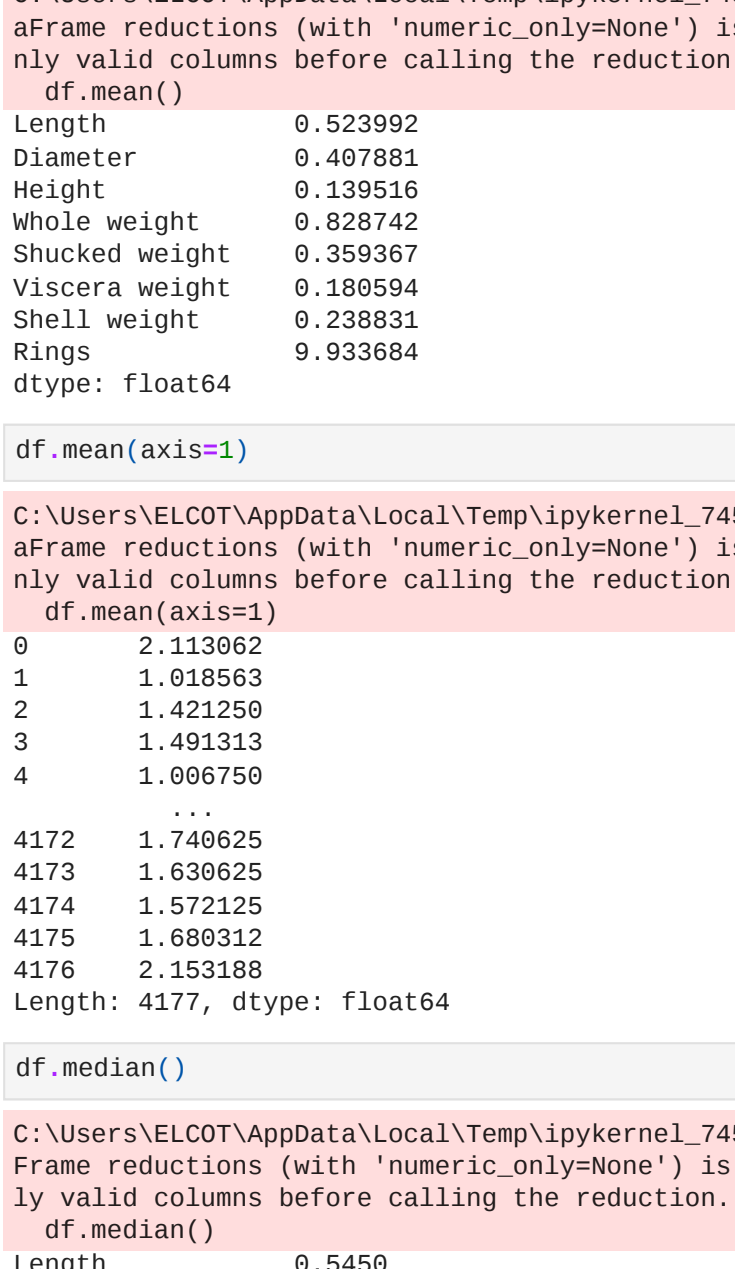
```
In [24]: sns.barplot(df.Sex.value_counts().index, df.Sex.value_counts())
Out[24]: <AxesSubplot: ylabel='Sex'>
```



```
In [25]: df.corr()
Out[25]:
```

	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 [26]: sns.heatmap(df.corr(), annot=True)
Out[26]: <AxesSubplot: >
```



```
In [27]: df.head()
Out[27]:
```

	Sex	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	Rings
0	M	0.455	0.365	0.095	0.5140	0.2245	0.1010	0.150	15
1	M	0.350	0.265	0.090	0.2255	0.0995	0.0485	0.070	7
2	F	0.530	0.420	0.135	0.6770	0.2565	0.1415	0.210	9
3	M	0.440	0.365	0.125	0.5160	0.2155	0.1140	0.155	10
4	I	0.330	0.255	0.080	0.2050	0.0895	0.0395	0.055	7

```
In [28]: df.mean()
Out[28]:
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

In [29]: df.mean(axis=1)
Out[29]:
0      2.113062
1      1.018563
2      1.421250
3      1.493133
4      1.006750
...
4172    1.740625
4173    1.630625
4174    1.572125
4175    1.680312
4176    2.153188
Length: 4177, dtype: float64

In [30]: df.median()
Out[30]:
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
Rings      9.0000
dtype: float64

In [31]: df.mode()
Out[31]:
```

	Sex	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	Rings
0	M	0.550	0.45	0.15	0.2225	0.175	0.1715	0.275	9.0
1	NaN	0.625	NaN	NaN	NaN	NaN	NaN	NaN	NaN

```
In [32]: df[["Rings"]].std()
Out[32]: 3.2241699320681133

In [33]: df[["Rings"]].var()
Out[33]: 10.395265947347835

In [34]: df[["Rings"]].skew()
Out[34]: 1.114101898355677

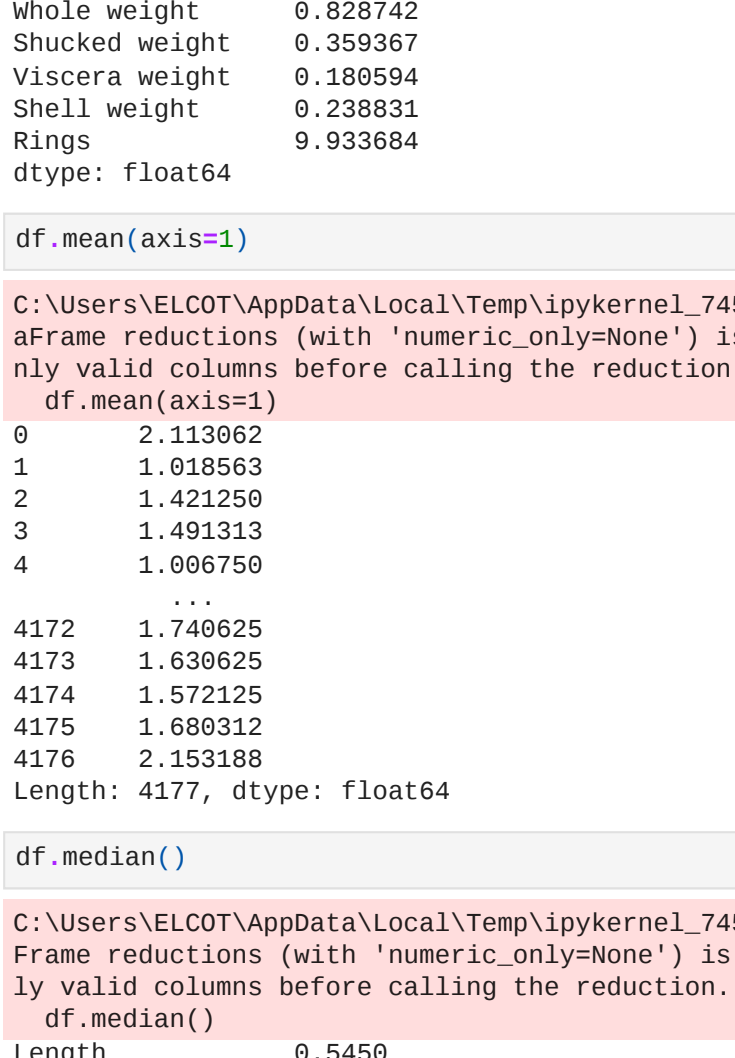
In [35]: df[["Rings"]].kurt()
Out[35]: 2.3306874266535847

In [36]: null=pd.isnull(df)
null.head()
```

	Sex	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	Rings
0	False	False	False	False	False	False	False	False	False
1	False	False	False	False	False	False	False	False	False
2	False	False	False	False	False	False	False	False	False
3	False	False	False	False	False	False	False	False	False
4	False	False	False	False	False	False	False	False	False

```
In [37]: pd.isnull(df).sum().sum()
Out[37]: 0

In [38]: sns.boxplot(df.Rings)
Out[38]: <seaborn.axisgrid.FacetGrid at 0x218ae00d90>
```



```
In [39]: q1=df.Rings.quantile(0.25)
q3=df.Rings.quantile(0.75)

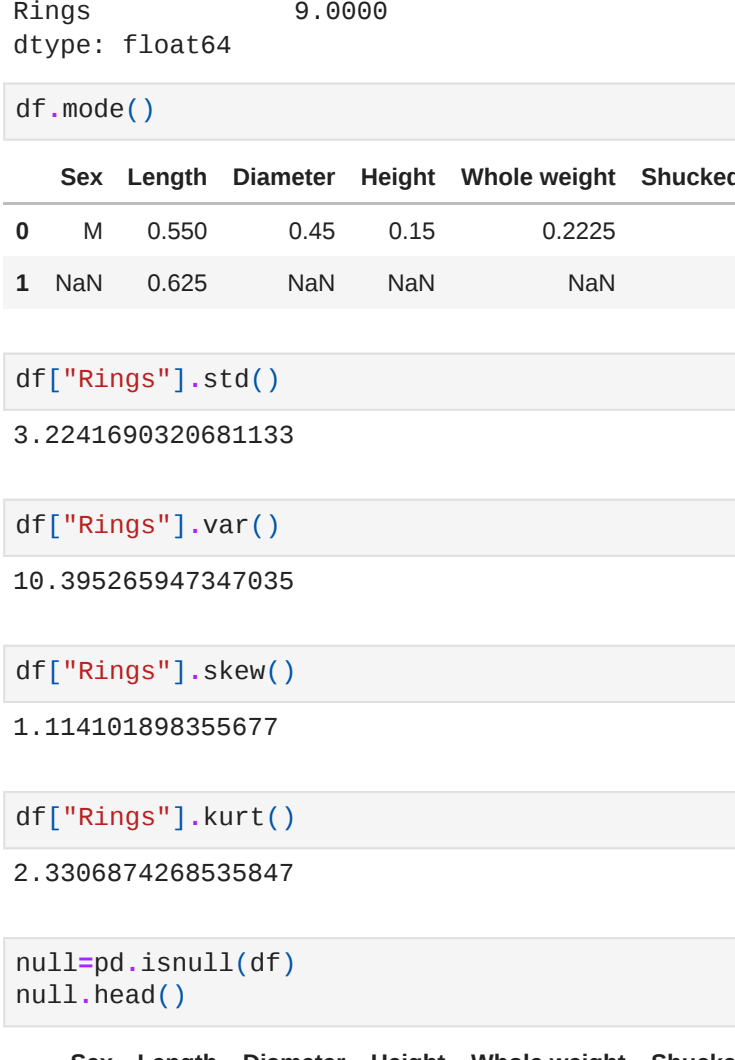
In [40]: IQR=q3-q1

In [41]: upper_limit=q3+1.5*IQR
lower_limit=q1-1.5*IQR

In [42]: upper_limit
Out[42]: 15.5

In [43]: df=df[df.Rings<upper_limit]

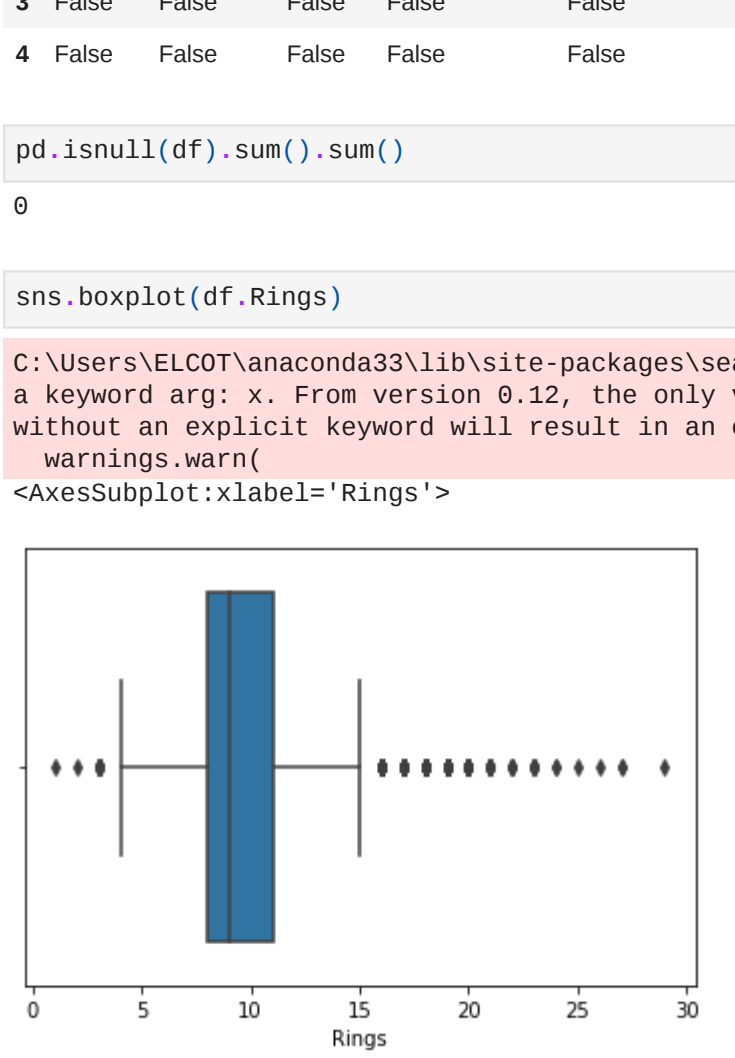
In [44]: sns.boxplot(df.Rings)
Out[44]: <seaborn.axisgrid.FacetGrid at 0x218ae00d90>
```



```
In [45]: lower_limit
Out[45]: 3.5

In [46]: df=df[df.Rings>lower_limit]

In [47]: sns.boxplot(df.Rings)
Out[47]: <seaborn.axisgrid.FacetGrid at 0x218ae00d90>
```



```
In [48]: df.shape
Out[48]: (3899, 9)

In [49]: df.head()
Out[49]:
```

	Sex	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	Rings
0	M	0.455	0.365	0.095	0.5140	0.2245	0.1010	0.150	15
1	M	0.350	0.265	0.090	0.2255	0.0995	0.0485	0.070	7
2	F	0.530	0.420	0.135	0.6770	0.2565	0.1415	0.210	9
3	M	0.440	0.365	0.125	0.5160	0.2155	0.1140	0.155	10
4	I	0.330	0.255	0.080	0.2050	0.0895	0.0395	0.055	7

```
In [53]: from sklearn.preprocessing import LabelEncoder

In [54]: le=LabelEncoder()

In [55]: df.Sex=le.fit_transform(df.Sex)
df.head()
```

	Sex	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	Rings
0	2	0.455	0.365	0.095	0.5140	0.2245	0.1010	0.150	15
1	2	0.350	0.265	0.090	0.2255	0.0995	0.0485	0.070	7
2	0	0.530	0.420	0.135	0.6770	0.2565	0.1415	0.210	9
3	2	0.440	0.365	0.125	0.5160	0.2155	0.1140	0.155	10
4	1	0.330	0.255	0.080	0.2050	0.0895	0.0395	0.055	7

```
In [59]: y=df[["Rings"]]
y
Out[59]:
0      15
1       7
2       9
3      10
4       7
...
4172    11
4173    10
4174     9
4175    10
4176    12
Name: Rings, Length: 3899, dtype: int64

In [60]: X=df.drop(columns=["Rings"], axis=1)
X.head()
```

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

```
In [61]: from sklearn.preprocessing import scale

In [62]: X_scaled=pd.DataFrame(scale(X), columns=X.columns)
X_scaled.head()
```

	Sex	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight
0	1.157932	-0.551043	-0.404229	-1.035557	-0.608006	-0.585646	-0.697589	-0.604476
1	1.157932	-1.433200	-1.423098	-1.156858	-1.206406	-1.146009	-1.179895	-1.213621
2	-1.299758	0.079070	0.156149	-0.065146	-0.269916	-0.442193	-0.325524	-0.147618
3	1.157932	-0.677065	-0.404229	-0.307749	-0.603858	-0.625992	-0.578161	-0.566405
4	-0.070913	-1.601230	-1.524985	-1.399461	-1.248927	-1.190638	-1.262576	-1.327835

```
In [63]: from sklearn.model_selection import train_test_split
X_train,X_test,y_train,y_test=train_test_split(X_scaled,y,test_size=0.3, random_state=0)

In [64]: X_train.shape
Out[64]: (2729, 8)

In [65]: y_train.shape
Out[65]: (2729, )

In [66]: X_test.shape
Out[66]: (1170, 8)

In [67]: y_test.shape
Out[67]: (1170, )

In [72]: from sklearn.tree import DecisionTreeClassifier
model=DecisionTreeClassifier(max_depth=3)

In [73]: model.fit(X_train,y_train)

Out[73]: DecisionTreeClassifier(max_depth=3)

In [74]: test_pred=model.predict(X_test)
test_pred
Out[74]: array([11,  8,  6, ..., 11, 10,  6], dtype=int64)

In [75]: train_pred=model.predict(X_train)

In [76]: from sklearn.metrics import accuracy_score,classification_report,confusion_matrix

In [77]: print('Training Accuracy:',accuracy_score(y_train,train_pred))
Training Accuracy: 0.3023085379259802

In [78]: print('Testing Accuracy:',accuracy_score(y_test,test_pred))
Testing Accuracy: 0.28034188034188035

In [79]: pd.crosstab(y_test,test_pred)
Out[79]:
```

	col_0	4	6	7	8	9	10	11
Rings	4	9	7	0	0	0	0	0
	5	6	25	8	0	0	0	0
	6	0	22	47	10	0	2	0
	7	1	13	60	42	3	0	0
	8	0	6	30	64	51	9	3
	9	0	4	13	54	79	44	7
	10	0	0	7	21	58	68	25
	11	0	1	6	20	41	50	26
	12	0	0	4	11	28	24	14
	13	0	0	1	8	20	21	12
	14	0	0	1	3	17	11	12
	15	0	0	0	5	8	18	10

```
In [81]: print(classification_report(y_test,test_pred))
```

```

              precision    recall  f1-score   support

     4         0.50         0.56         0.56         16
     5         0.55         0.60         0.60         39
     6         0.28         0.27         0.28         81
     7         0.34         0.50         0.41        119
     8         0.27         0.39         0.32        163
     9         0.26         0.39         0.31        201
    10         0.28         0.38         0.32        179
    11         0.24         0.19         0.21        144
    12         0.00         0.00         0.00         81
    13         0.00         0.00         0.00         62
    14         0.00         0.00         0.00         44
    15         0.00         0.00         0.00         41

   accuracy          0.19         0.22         0.20        1170
  macro avg          0.19         0.22         0.20        1170
 weighted avg          0.22         0.28         0.24        1170

C:\Users\ELCOT\anaconda3\lib\site-packages\seaborn\metrics\classification.py:1318: UndefinedMetricWarning: Precision for class 0 is ill-defined and being set to 0.0 in labels with no predicted samples. Use 'zero_division' parameter to control this behavior.
  warn_prf(average, modifier, msg_start, len(result))
C:\Users\ELCOT\anaconda3\lib\site-packages\seaborn\metrics\classification.py:1318: UndefinedMetricWarning: Precision for class 0 is ill-defined and being set to 0.0 in labels with no predicted samples. Use 'zero_division' parameter to control this behavior.
  warn_prf(average, modifier, msg_start, len(result))
C:\Users\ELCOT\anaconda3\lib\site-packages\seaborn\metrics\classification.py:1318: UndefinedMetricWarning: Precision for class 0 is ill-defined and being set to 0.0 in labels with no predicted samples. Use 'zero_division' parameter to control this behavior.
  warn_prf(average, modifier, msg_start, len(result))
C:\Users\ELCOT\anaconda3\lib\site-packages\seaborn\metrics\classification.py:1318: UndefinedMetricWarning: Precision for class 0 is ill-defined and being set to 0.0 in labels with no predicted samples. Use 'zero_division' parameter to control this behavior.
  warn_prf(average, modifier, msg_start, len(result))

In [ ]:

In [ ]:
```