

ASSIGNMENT – 3

Abalone Age Prediction

ASSIGNMENT DATE	29-09-2022
STUDENT NAME	Asraf I
STUDENT ROLL NO.	913219104002
MAXIMUM MARK	2 Mark

1. Download the dataset

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

2. Load the dataset

```
In [8]:
data=pd.read_csv("abalone.csv")
data
```

```
Out[8]:
```

	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	0.1500	15
1	M	0.350	0.265	0.090	0.2255	0.0995	0.0485	0.0700	7
2	F	0.530	0.420	0.135	0.6770	0.2565	0.1415	0.2100	9
3	M	0.440	0.365	0.125	0.5160	0.2155	0.1140	0.1550	10
4	I	0.330	0.255	0.080	0.2050	0.0895	0.0395	0.0550	7
...
4172	F	0.565	0.450	0.165	0.8870	0.3700	0.2390	0.2490	11
4173	M	0.590	0.440	0.135	0.9660	0.4390	0.2145	0.2605	10
4174	M	0.600	0.475	0.205	1.1760	0.5255	0.2875	0.3080	9

	Sex	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	Rings
4175	F	0.625	0.485	0.150	1.0945	0.5310	0.2610	0.2960	10
4176	M	0.710	0.555	0.195	1.9485	0.9455	0.3765	0.4950	12

4177 rows × 9 columns

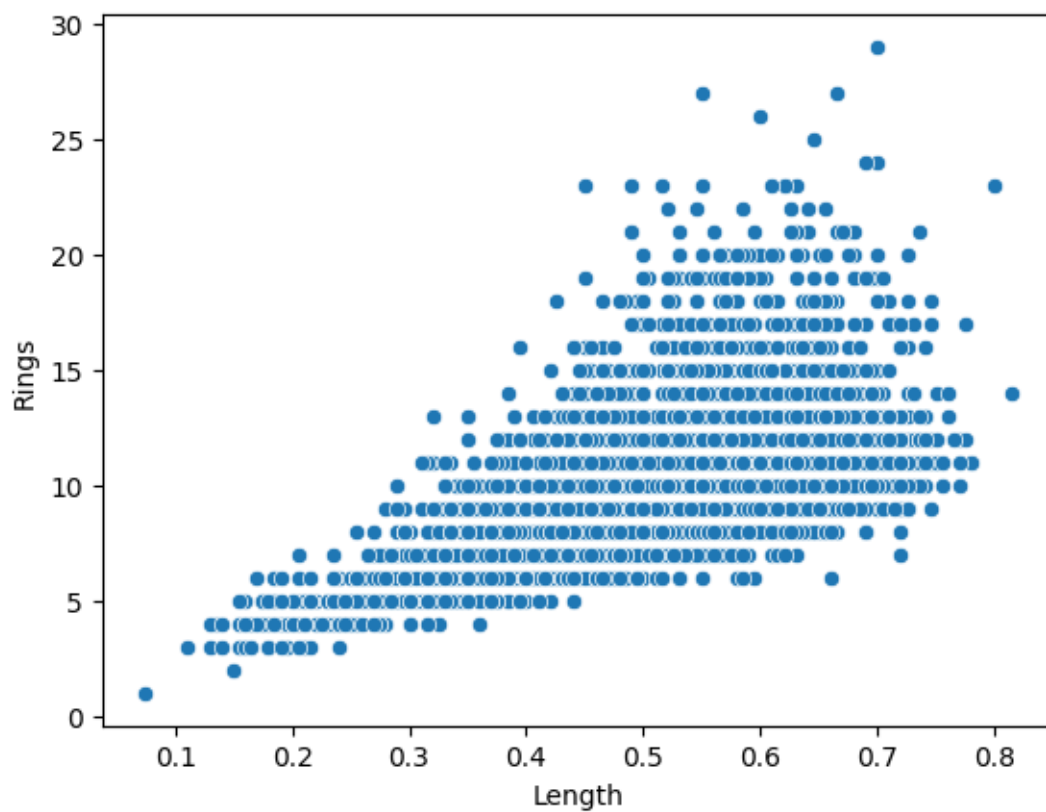
3. perform the visualizations

Bivariate analysis

```
sns.scatterplot(x=data.Length,y=data.Rings)
```

In [9]:

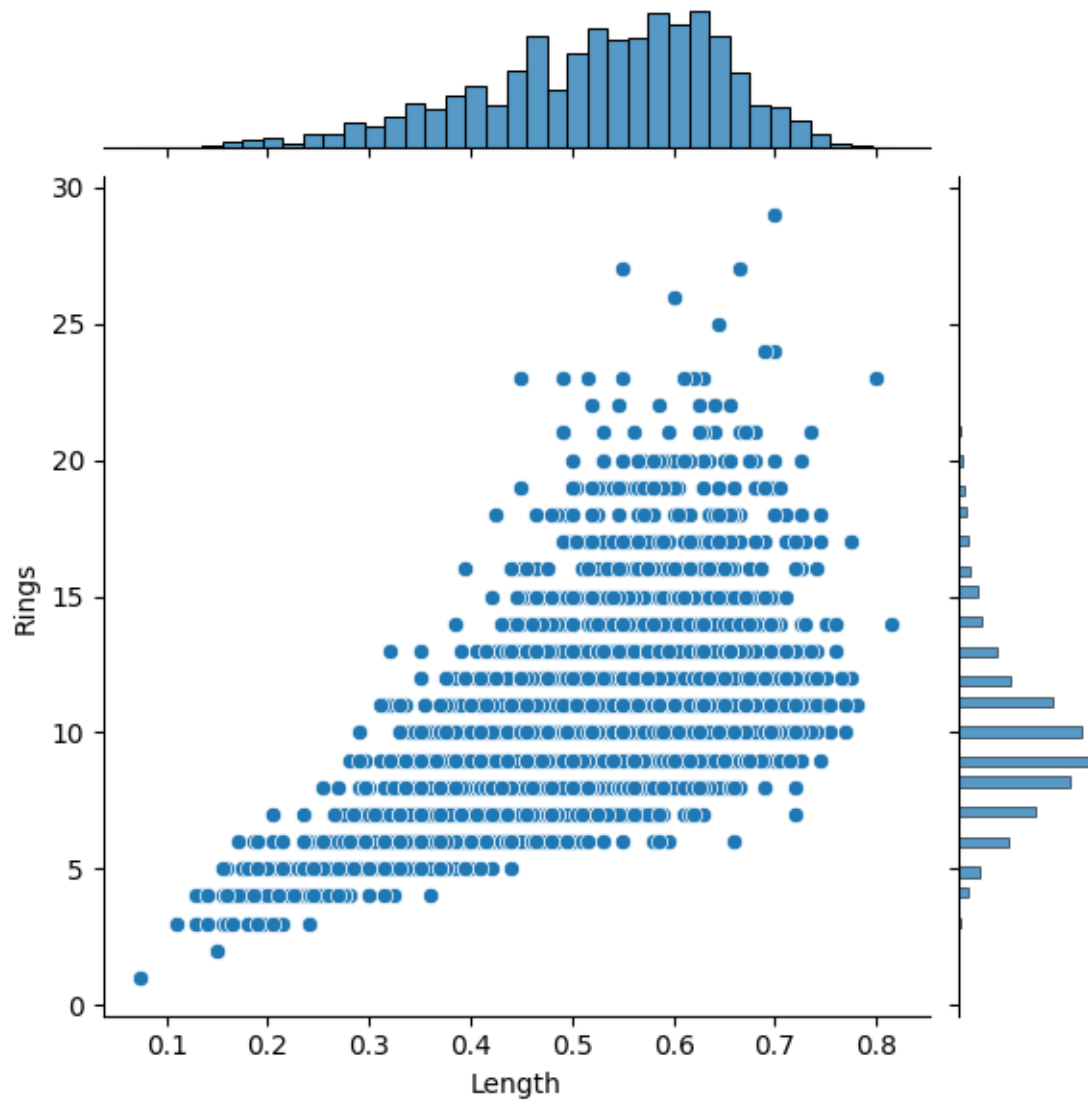
Out[9]:



```
sns.jointplot(x=data.Length,y=data.Rings)
```

In [10]:

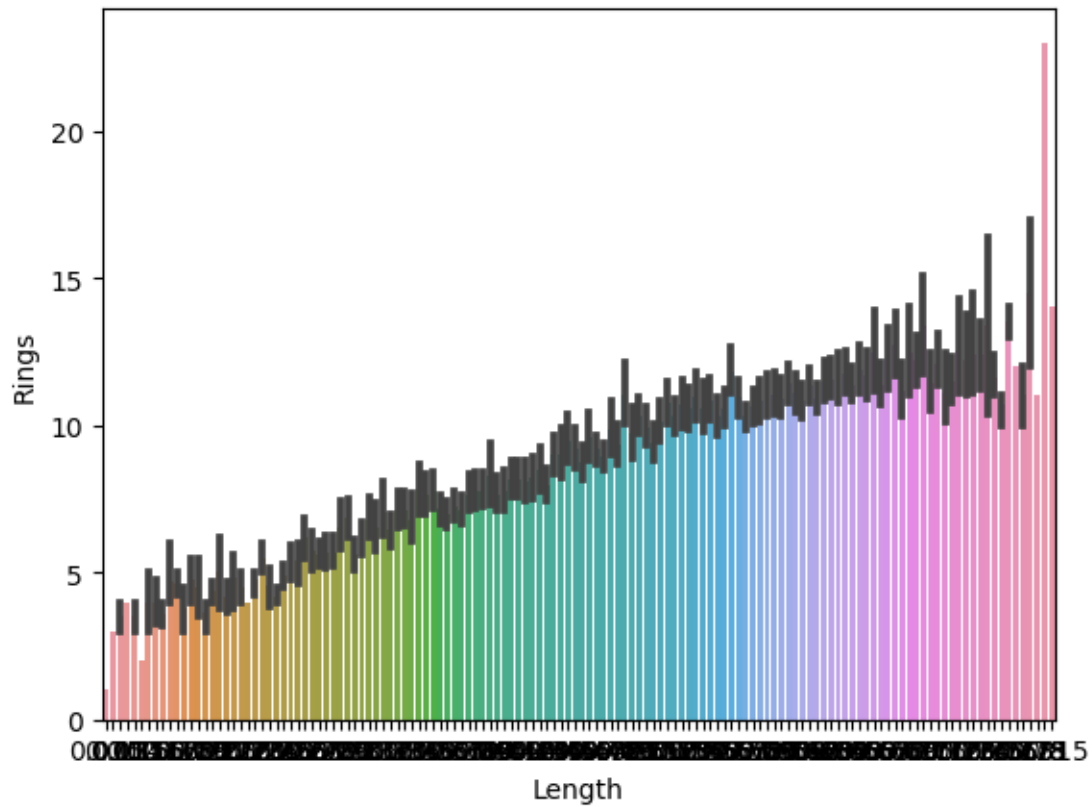
Out[10]:



```
sns.barplot(x=data.Length,y=data.Rings)
```

In [21]:

Out[21]:



Univariate analysis

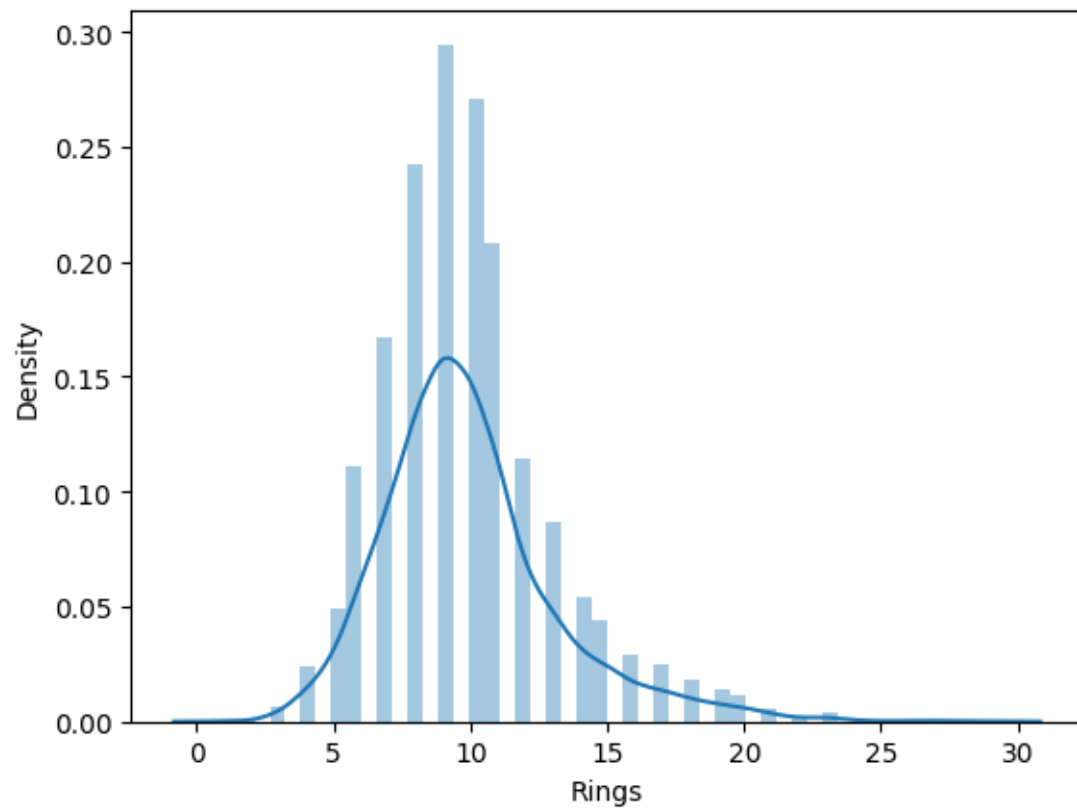
```
import warnings
warnings.filterwarnings('ignore')
```

In [22]:

```
sns.distplot(data.Rings)
```

In [23]:

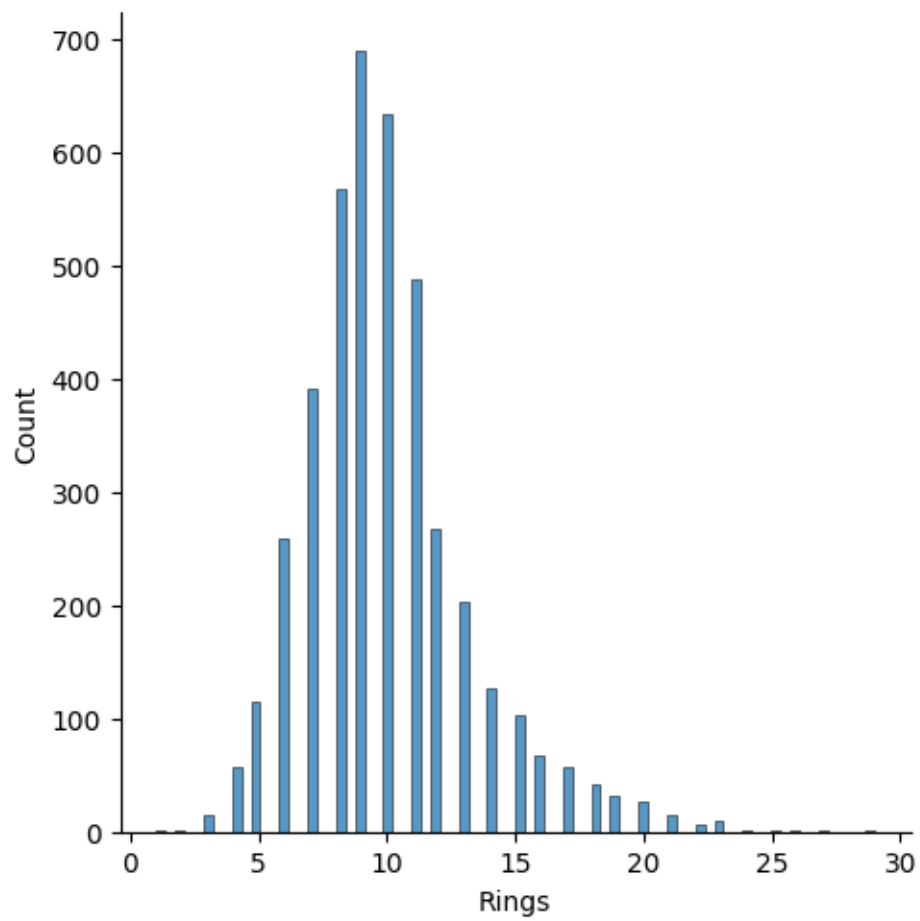
Out[23]:



```
sns.displot(data.Rings)
```

In [24]:

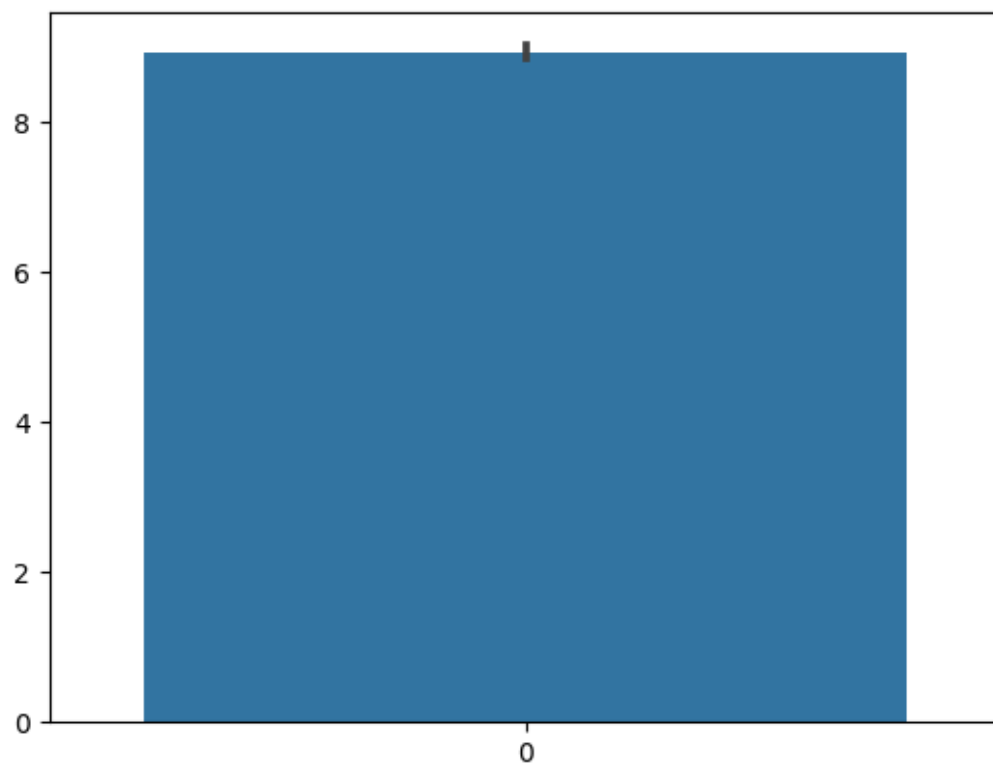
Out[24]:



```
sns.barplot(data.Rings)
```

In [111]:

Out[111]:



4. perform descriptive statistics

In [27]:

```
data.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 [67]:

```
data.tail()
```

Out[67]:

	Sex	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	Rings
4172	0	0.565	73	0.165	0.8870	0.3700	0.2390	0.2490	10
4173	2	0.590	71	0.135	0.9660	0.4390	0.2145	0.2605	9
4174	2	0.600	78	0.205	1.1760	0.5255	0.2875	0.3080	8
4175	0	0.625	80	0.150	1.0945	0.5310	0.2610	0.2960	9
4176	2	0.710	94	0.195	1.9485	0.9455	0.3765	0.4950	11

In [68]:

```
data.info()
```

RangeIndex: 4177 entries, 0 to 4176

Data columns (total 9 columns):

#	Column	Non-Null Count	Dtype
---	-----	-----	-----
0	Sex	4177 non-null	int32
1	Length	4177 non-null	float64
2	Diameter	4177 non-null	int64
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(6), int32(1), int64(2)
memory usage: 277.5 KB
```

In [69]:

```
data.shape
```

Out[69]:

```
(4177, 9)
```

measure of tendency

In [70]:

```
data.mean()
```

Out[70]:

```
Sex           1.052909
Length        0.523992
Diameter      64.576969
Height        0.139516
Whole weight  0.828742
Shucked weight 0.359367
Viscera weight 0.180594
Shell weight  0.238831
Rings         8.933445
dtype: float64
```

In [71]:

```
data.mode()
```

Out[71]:

	Sex	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	Rings
0	2.0	0.550	73.0	0.15	0.2225	0.175	0.1715	0.275	8.0
1	NaN	0.625	NaN	NaN	NaN	NaN	NaN	NaN	NaN

In [72]:

```
data.median()
```

Out[72]:

```
Sex           1.0000
Length        0.5450
Diameter      68.0000
Height        0.1400
Whole weight  0.7995
Shucked weight 0.3360
Viscera weight 0.1710
Shell weight  0.2340
Rings         8.0000
dtype: float64
```

In [73]:


```
data.skew()
```

Out[73]:

```
Sex          -0.098155
Length       -0.639873
Diameter     -0.607999
Height       3.128817
Whole weight 0.530959
Shucked weight 0.719098
Viscera weight 0.591852
Shell weight 0.620927
Rings        1.108353
dtype: float64
```

In [74]:

```
data.kurtosis()
```

Out[74]:

```
Sex          -1.514387
Length       0.064621
Diameter     -0.054859
Height      76.025509
Whole weight -0.023644
Shucked weight 0.595124
Viscera weight 0.084012
Shell weight 0.531926
Rings        2.283203
dtype: float64
```

In [75]:

```
data.std()
```

Out[75]:

```
Sex          0.822240
Length       0.120093
Diameter     19.841382
Height       0.041827
Whole weight 0.490389
Shucked weight 0.221963
Viscera weight 0.109614
Shell weight 0.139203
Rings        3.222790
dtype: float64
```

In [76]:

```
data.var()
```

Out[76]:

```
Sex          0.676079
Length       0.014422
Diameter     393.680437
Height       0.001750
Whole weight 0.240481
Shucked weight 0.049268
Viscera weight 0.012015
Shell weight 0.019377
Rings       10.386374
dtype: float64
```

5.check the missing values and deal with them

```
data.isnull().any()
```

In [77]:

Out[77]:

```
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 [78]:

```
data.isnull().sum()
```

Out[78]:

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

In [79]:

```
data.dropna()
```

Out[79]:

	Sex	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	Rings
0	2	0.455	56	0.095	0.5140	0.2245	0.1010	0.1500	14
1	2	0.350	36	0.090	0.2255	0.0995	0.0485	0.0700	6
2	0	0.530	67	0.135	0.6770	0.2565	0.1415	0.2100	8
3	2	0.440	56	0.125	0.5160	0.2155	0.1140	0.1550	9
4	1	0.330	34	0.080	0.2050	0.0895	0.0395	0.0550	6
...
4172	0	0.565	73	0.165	0.8870	0.3700	0.2390	0.2490	10

	Sex	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	Rings
4173	2	0.590	71	0.135	0.9660	0.4390	0.2145	0.2605	9
4174	2	0.600	78	0.205	1.1760	0.5255	0.2875	0.3080	8
4175	0	0.625	80	0.150	1.0945	0.5310	0.2610	0.2960	9
4176	2	0.710	94	0.195	1.9485	0.9455	0.3765	0.4950	11

4177 rows × 9 columns

6. Find the outliers and replace them

outliers

In [80]:

```
qnt=data.quantile(q=[0.25,0.75])
qnt
```

Out[80]:

	Sex	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	Rings
0.25	0.0	0.450	53.0	0.115	0.4415	0.186	0.0935	0.130	7.0
0.75	2.0	0.615	79.0	0.165	1.1530	0.502	0.2530	0.329	10.0

7.Check the categorical columns and perform the encoding

In [81]:

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

In [82]:

```
le=LabelEncoder()
```

In [83]:

```
data["Sex"]=le.fit_transform(data['Sex'])
data["Rings"]=le.fit_transform(data['Rings'])
data["Diameter"]=le.fit_transform(data['Diameter'])
data.head()
```

Out[83]:

	Sex	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	Rings
0	2	0.455	56	0.095	0.5140	0.2245	0.1010	0.150	14
1	2	0.350	36	0.090	0.2255	0.0995	0.0485	0.070	6
2	0	0.530	67	0.135	0.6770	0.2565	0.1415	0.210	8
3	2	0.440	56	0.125	0.5160	0.2155	0.1140	0.155	9
4	1	0.330	34	0.080	0.2050	0.0895	0.0395	0.055	6

8. split the data into dependent and independent variables

```
In [84]:
x=data.iloc[:, :-1].values
y=data.iloc[:, -1].values
x
```

```
Out[84]:
array([[2.000e+00, 4.550e-01, 5.600e+01, ..., 2.245e-01, 1.010e-01,
        1.500e-01],
       [2.000e+00, 3.500e-01, 3.600e+01, ..., 9.950e-02, 4.850e-02,
        7.000e-02],
       [0.000e+00, 5.300e-01, 6.700e+01, ..., 2.565e-01, 1.415e-01,
        2.100e-01],
       ...,
       [2.000e+00, 6.000e-01, 7.800e+01, ..., 5.255e-01, 2.875e-01,
        3.080e-01],
       [0.000e+00, 6.250e-01, 8.000e+01, ..., 5.310e-01, 2.610e-01,
        2.960e-01],
       [2.000e+00, 7.100e-01, 9.400e+01, ..., 9.455e-01, 3.765e-01,
        4.950e-01]])
```

```
In [85]:
y
```

```
Out[85]:
array([14,  6,  8, ...,  8,  9, 11], dtype=int64)
```

```
In [86]:
print(x.shape,y.shape)
(4177, 8) (4177,)
```

9. Scale the independent variables

```
In [87]:
from sklearn.preprocessing import scale
```

```
x=scale(x)
x
```

In [88]:

Out[88]:

```
array([[ 1.15198011, -0.57455813, -0.43232856, ..., -0.60768536,
        -0.72621157, -0.63821689],
       [ 1.15198011, -1.44898585, -1.44044354, ..., -1.17090984,
        -1.20522124, -1.21298732],
       [-1.28068972,  0.05003309,  0.12213469, ..., -0.4634999 ,
        -0.35668983, -0.20713907],
       ...,
       [ 1.15198011,  0.6329849 ,  0.67659793, ...,  0.74855917,
        0.97541324,  0.49695471],
       [-1.28068972,  0.84118198,  0.77740943, ...,  0.77334105,
        0.73362741,  0.41073914],
       [ 1.15198011,  1.54905203,  1.48308992, ...,  2.64099341,
        1.78744868,  1.84048058]])
```

10. Split the data into training and testing

```
from sklearn.model_selection import train_test_split
```

In [89]:

```
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2,random_state=0)
```

In [90]:

```
x_train.shape
```

In [91]:

```
(3341, 8)
```

Out[91]:

```
x_test.shape
```

In [92]:

```
(836, 8)
```

Out[92]:

11. Build the model

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

In [93]:

```
regressor=LinearRegression()
```

In [94]:

```
regressor.fit(x_train,y_train)
```

Out[94]:

```
LinearRegression()
```

In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.

On GitHub, the HTML representation is unable to render, please try loading this page with nbviewer.org.

```
from sklearn.tree import DecisionTreeClassifier
```

In [95]:

```
model=DecisionTreeClassifier()
```

In [96]:

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

In [97]:

DecisionTreeClassifier()

Out[97]:

In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.

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12. Train and the model

```
from sklearn.neighbors import KNeighborsClassifier
```

In [98]:

```
knn=KNeighborsClassifier(n_neighbors=5)
```

In [99]:

```
knn.fit(x_train,y_train)
knn.fit(x_test,y_test)
```

In [100]:

KNeighborsClassifier()

Out[100]:

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14. Measure the performance using metrics

```
from sklearn.naive_bayes import GaussianNB
```

In [101]:

```
nb=GaussianNB()
nb.fit(x_train,y_train)
```

In [102]:

GaussianNB()

Out[102]:

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```
from sklearn.metrics import accuracy_score, confusion_matrix
```

In [103]:

```
pred=nb.predict(x_test)
pred
```

In [104]:

Out[104]:

```
array([[ 8,  7, 10,  3, 10, 10,  6,  7,  5, 10,  6,  4,  6,  7,  3,  9,  5,
        9, 10,  6,  5,  4,  6,  5,  8,  8,  2, 10,  8,  8,  5,  2,  9, 10,
        6,  7,  6,  9,  6, 10,  9,  7,  8,  8,  9,  6,  9, 10,  8,  5,  8,
```

```
5, 6, 10, 7, 6, 3, 5, 5, 5, 9, 8, 8, 6, 5, 10, 10, 10,
7, 10, 7, 26, 10, 8, 8, 7, 8, 7, 10, 7, 9, 4, 5, 9, 8,
6, 9, 10, 2, 5, 8, 8, 6, 5, 8, 4, 7, 5, 9, 10, 10, 7,
10, 9, 3, 10, 7, 5, 5, 9, 10, 10, 7, 8, 8, 10, 10, 9, 4,
7, 5, 8, 5, 6, 8, 9, 10, 10, 5, 10, 10, 7, 7, 10, 10, 9,
7, 7, 5, 10, 7, 7, 5, 8, 9, 8, 6, 7, 6, 4, 7, 23, 5,
10, 5, 5, 9, 5, 9, 5, 8, 4, 7, 9, 3, 9, 5, 4, 8, 8,
8, 2, 6, 9, 10, 10, 3, 6, 6, 8, 7, 9, 8, 7, 8, 8, 9,
7, 4, 7, 8, 7, 6, 9, 8, 7, 5, 7, 2, 10, 6, 7, 10, 8,
10, 9, 3, 10, 7, 7, 3, 8, 8, 8, 9, 8, 8, 10, 3, 10, 8,
10, 6, 9, 10, 9, 10, 7, 11, 9, 8, 7, 4, 10, 7, 9, 7, 3,
7, 8, 7, 9, 8, 10, 8, 8, 10, 6, 9, 10, 10, 6, 10, 10, 8,
7, 5, 7, 8, 10, 10, 7, 9, 4, 7, 6, 6, 6, 5, 5, 8, 6,
5, 6, 4, 10, 10, 6, 9, 7, 4, 10, 5, 10, 10, 8, 6, 10, 8,
9, 7, 7, 4, 7, 5, 7, 4, 3, 8, 10, 6, 6, 23, 7, 6, 6,
8, 8, 8, 8, 10, 10, 6, 8, 10, 9, 7, 4, 6, 8, 10, 8, 6,
10, 9, 7, 7, 8, 10, 10, 7, 23, 2, 10, 8, 10, 6, 7, 10, 8,
6, 7, 10, 9, 8, 6, 4, 8, 9, 5, 4, 8, 10, 10, 10, 7, 10,
10, 6, 7, 7, 8, 6, 9, 10, 3, 4, 8, 8, 7, 9, 5, 9, 6,
10, 9, 10, 6, 7, 2, 7, 8, 9, 10, 3, 10, 10, 7, 6, 4, 10,
8, 7, 7, 19, 5, 10, 6, 8, 7, 6, 8, 5, 10, 6, 10, 3, 10,
10, 7, 7, 6, 10, 8, 10, 8, 9, 9, 10, 10, 7, 7, 8, 9, 4,
8, 8, 10, 9, 9, 7, 10, 7, 8, 6, 6, 10, 8, 8, 10, 6, 4,
9, 6, 5, 5, 9, 3, 7, 9, 2, 4, 9, 4, 5, 6, 9, 8, 6,
9, 6, 10, 7, 8, 7, 10, 4, 19, 6, 7, 8, 8, 6, 10, 7, 6,
6, 6, 5, 10, 7, 10, 6, 8, 10, 10, 8, 7, 7, 10, 2, 8, 8,
8, 4, 5, 6, 4, 9, 9, 6, 9, 5, 9, 9, 7, 10, 5, 4,
7, 9, 10, 9, 6, 6, 9, 10, 10, 8, 8, 8, 10, 9, 4, 10, 6,
7, 7, 7, 6, 6, 6, 8, 3, 8, 10, 10, 9, 10, 6, 8, 6, 9,
5, 8, 7, 6, 9, 9, 8, 10, 6, 10, 9, 6, 7, 8, 9, 8, 8,
9, 10, 8, 10, 3, 6, 8, 8, 6, 8, 8, 7, 8, 7, 6, 8, 8,
10, 2, 9, 4, 3, 4, 7, 8, 26, 9, 7, 10, 4, 6, 10, 5, 9,
10, 2, 6, 4, 10, 7, 10, 8, 10, 10, 7, 9, 4, 8, 10, 6, 10,
8, 7, 8, 10, 7, 5, 10, 8, 7, 10, 5, 8, 9, 8, 6, 9, 6,
8, 6, 5, 6, 5, 6, 8, 3, 10, 6, 8, 7, 10, 10, 9, 8, 7,
5, 8, 10, 7, 6, 6, 6, 8, 5, 10, 6, 6, 9, 9, 7, 6, 9,
8, 7, 6, 6, 4, 4, 8, 8, 10, 10, 10, 3, 6, 6, 4, 8, 10,
9, 8, 9, 7, 7, 6, 8, 9, 7, 10, 9, 9, 8, 10, 7, 10, 10,
8, 7, 6, 5, 8, 9, 9, 7, 8, 7, 6, 10, 9, 4, 4, 4, 8,
5, 7, 9, 7, 6, 6, 9, 3, 7, 5, 9, 4, 10, 7, 8, 9, 8,
10, 9, 10, 7, 8, 7, 10, 6, 4, 8, 8, 4, 6, 8, 5, 9, 8,
7, 6, 8, 8, 5, 10, 6, 10, 7, 5, 8, 9, 5, 4, 5, 4, 8,
8, 10, 9, 8, 6, 4, 6, 8, 10, 8, 7, 5, 7, 5, 8, 10, 8,
6, 5, 6, 8, 7, 6, 9, 8, 8, 10, 10, 8, 9, 8, 8, 8, 2,
7, 6, 6, 6, 4, 4, 6, 7, 10, 6, 7, 6, 7, 8, 8, 23, 7,
8, 9, 8, 7, 4, 8, 10, 4, 6, 8, 9, 7, 4, 4, 8, 9, 6,
6, 9, 3], dtype=int64)
```

In [105]:

```
accuracy_score(y_test,pred)
```

Out[105]:

```
0.24162679425837322
```