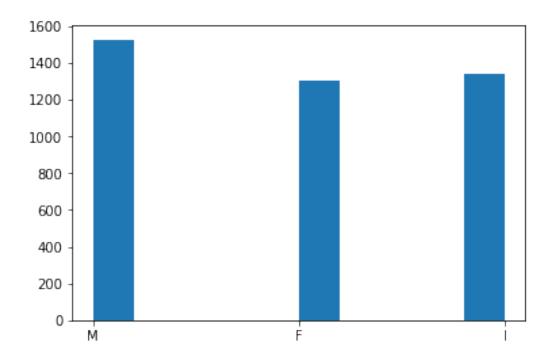
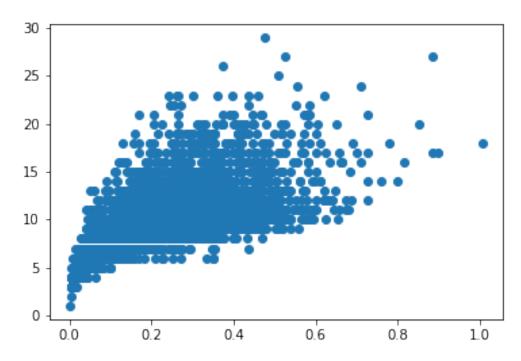
```
Assignment 3
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
2.Load Dataset
from google.colab import drive
drive.mount( '/content/drive' )
Drive already mounted at /content/drive; to attempt to forcibly
remount, call drive.mount("/content/drive", force remount=True).
#import dataset
data = pd.read_csv(r"/content/drive/MyDrive/content/abalone.csv")
data.head()
                         Height Whole weight Shucked weight
  Sex Length Diameter
weight \
        0.455
                  0.365
                          0.095
                                       0.5140
                                                        0.2245
   М
0.1010
                  0.265
   М
        0.350
                          0.090
                                       0.2255
                                                        0.0995
0.0485
                                                       0.2565
    F
        0.530
                  0.420
                          0.135
                                       0.6770
0.1415
        0.440
                  0.365
                                       0.5160
   М
                          0.125
                                                       0.2155
0.1140
        0.330
                  0.255
                          0.080
                                       0.2050
                                                        0.0895
4
    Ι
0.0395
   Shell weight
                 Rings
0
          0.150
                    15
                     7
1
          0.070
2
          0.210
                     9
3
          0.155
                    10
          0.055
                     7
     Visualization
plt.hist(data['Sex'])
                                0., 0., 1307.,
                  0.,
                         0.,
                                                     0.,
(array([1528.,
                                                             0.,
                                                                    0.,
        1342.]),
 array([0., 0.2, 0.4, 0.6, 0.8, 1., 1.2, 1.4, 1.6, 1.8, 2.]),
 <a list of 10 Patch objects>)
```

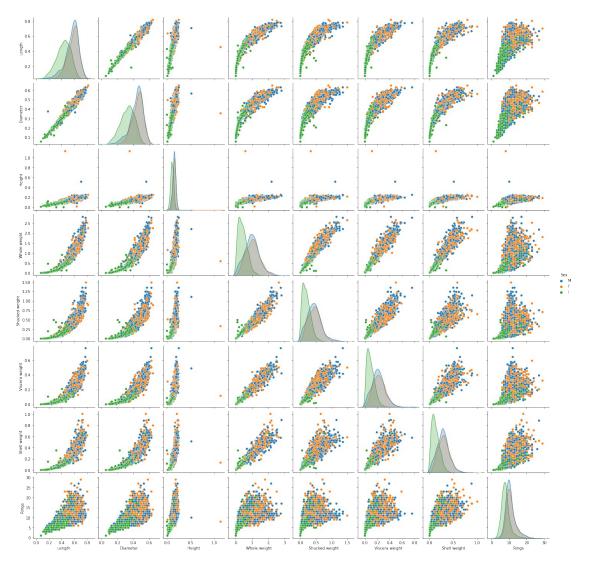


plt.scatter(data['Shell weight'],data['Rings'])
<matplotlib.collections.PathCollection at 0x7f57a1bbefd0>



sns.pairplot(data,hue='Sex')

<seaborn.axisgrid.PairGrid at 0x7f57a1b8ab10>



4.Descriptive statistics

## data.describe()

Length	Diameter	Height	Whole weight	Shucked
7.000000	4177.000000	4177.000000	4177.000000	
0.523992	0.407881	0.139516	0.828742	
0.120093	0.099240	0.041827	0.490389	
0.075000	0.055000	0.000000	0.002000	
0.450000	0.350000	0.115000	0.441500	
0.545000	0.425000	0.140000	0.799500	
	7.000000 0 0.523992 0.120093 0.075000 0.450000	7.000000 4177.000000 0.523992 0.407881 0.120093 0.099240 0.075000 0.055000 0.450000 0.350000	7.000000 4177.000000 4177.000000 0.523992 0.407881 0.139516 0.120093 0.099240 0.041827 0.075000 0.055000 0.000000 0.450000 0.350000 0.115000	7.000000 4177.000000 4177.000000 4177.000000 00 0.523992 0.407881 0.139516 0.828742 0.120093 0.099240 0.041827 0.490389 0.075000 0.055000 0.000000 0.002000 0.450000 0.350000 0.115000 0.441500

75% 0.502000	0.615000	0.480000	0.165000	1.153000
max 1 488000	0.815000	0.650000	1.130000	2.825500

	Viscera weight	Shell weight	Rings
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

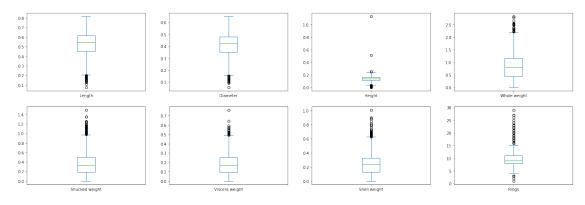
### 1. Missing Values

data.isna().sum() #No missing values

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

### 6.Outliers

data.plot(kind="box", subplots=True, layout=(7,4), figsize=(25,30));



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

Length Diameter Height Whole weight Shucked weight Viscera weight \ 0.25 0.450 0.35 0.115 0.4415 0.186 0.0935

```
0.75
       0.615
                  0.48
                          0.165
                                                         0.502
                                        1.1530
0.2530
      Shell weight
                     Rings
0.25
             0.130
                       8.0
0.75
             0.329
                      11.0
igr=qnt.loc[0.75]-qnt.loc[0.25]
igr
Length
                   0.1650
Diameter
                  0.1300
Height
                  0.0500
Whole weight
                  0.7115
                  0.3160
Shucked weight
Viscera weight
                  0.1595
                  0.1990
Shell weight
                   3.0000
Rings
dtype: float64
lower=qnt.loc[0.25]-(1.5*iqr)
lower
Length
                  0.20250
Diameter
                  0.15500
Heiaht
                  0.04000
Whole weight
                  -0.62575
Shucked weight
                 -0.28800
Viscera weight
                 -0.14575
Shell weight
                  -0.16850
                  3.50000
Rings
dtype: float64
upper=qnt.loc[0.75]+(1.5*iqr)
upper
                    0.86250
Length
Diameter
                    0.67500
Height
                    0.24000
Whole weight
                    2.22025
Shucked weight
                    0.97600
Viscera weight
                    0.49225
Shell weight
                    0.62750
                   15.50000
Rings
dtype: float64
data.mean()
```

/usr/local/lib/python3.7/dist-packages/ipykernel\_launcher.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

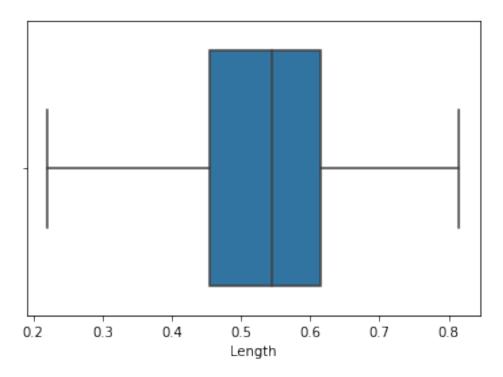
```
reduction.
"""Entry point for launching an IPython kernel.
```

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

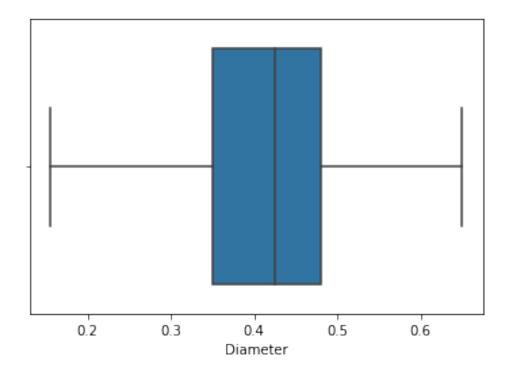
dtype: float64

```
data['Length']=np.where(data['Length']< 0.22,0.53,data['Length'])
sns.boxplot(x=data['Length'])</pre>
```

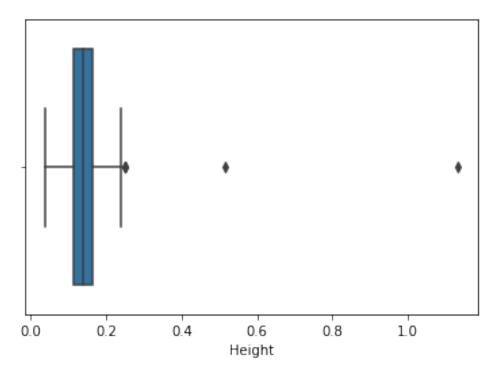
<matplotlib.axes.\_subplots.AxesSubplot at 0x7f579b444310>



```
data['Diameter']=np.where(data['Diameter']<
0.155,0.407,data['Diameter'])
sns.boxplot(x=data['Diameter'])
<matplotlib.axes._subplots.AxesSubplot at 0x7f579b4305d0>
```

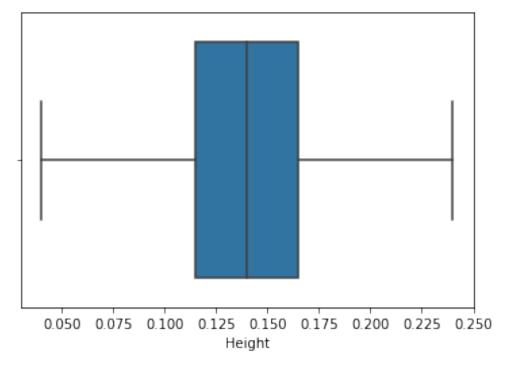


data['Height']=np.where(data['Height']< 0.04,0.14,data['Height'])
sns.boxplot(x=data['Height'])
<matplotlib.axes.\_subplots.AxesSubplot at 0x7f579d96a210>



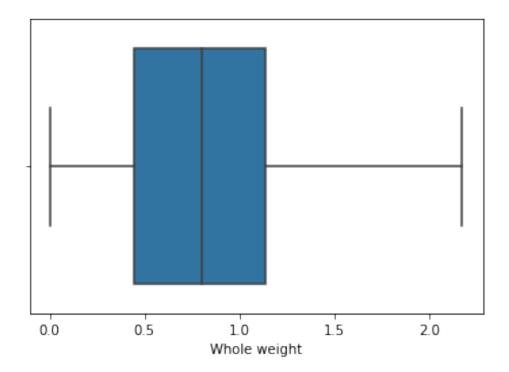
data['Height']=np.where(data['Height']>0.24,0.14,data['Height'])
sns.boxplot(x=data['Height'])

<matplotlib.axes.\_subplots.AxesSubplot at 0x7f579b6e25d0>



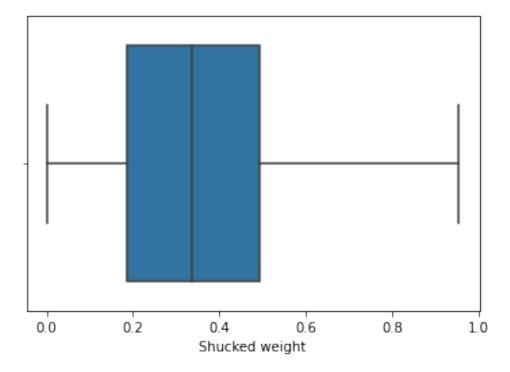
data['Whole weight']=np.where(data['Whole
weight']>2.18,0.83,data['Whole weight'])
sns.boxplot(x=data['Whole weight'])

<matplotlib.axes.\_subplots.AxesSubplot at 0x7f579d84d550>



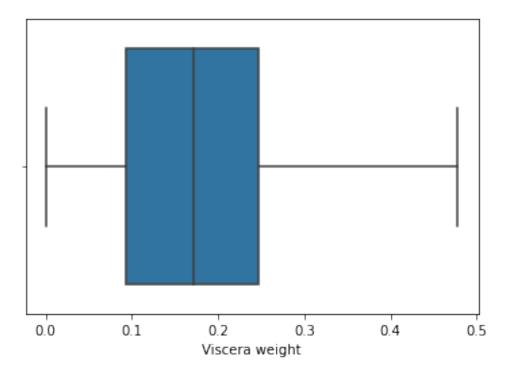
```
data['Shucked weight']=np.where(data['Shucked
weight']>0.958,0.359367,data['Shucked weight'])
sns.boxplot(x=data['Shucked weight'])
```

<matplotlib.axes. subplots.AxesSubplot at 0x7f579b560ad0>



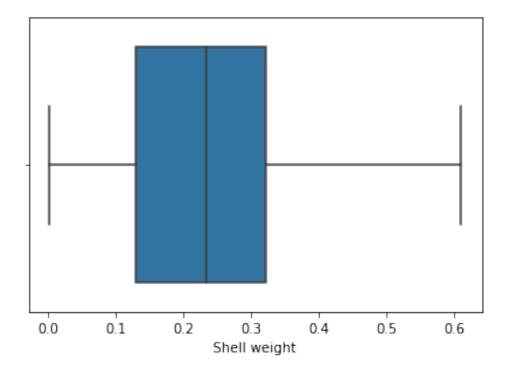
data['Viscera weight']=np.where(data['Viscera
weight']>0.478,0.18,data['Viscera weight'])
sns.boxplot(x=data['Viscera weight'])

<matplotlib.axes.\_subplots.AxesSubplot at 0x7f579d76fa10>



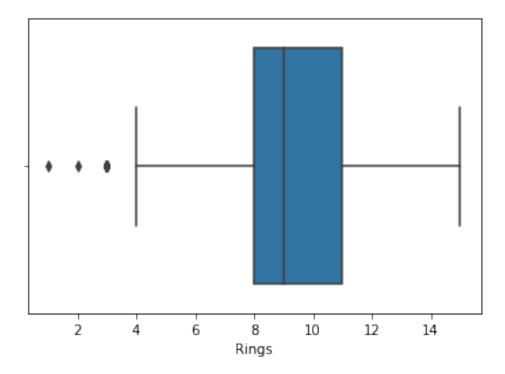
```
data['Shell weight']=np.where(data['Shell
weight']>0.61,0.238831,data['Shell weight'])
sns.boxplot(x=data['Shell weight'])
```

<matplotlib.axes.\_subplots.AxesSubplot at 0x7f579b75fc90>



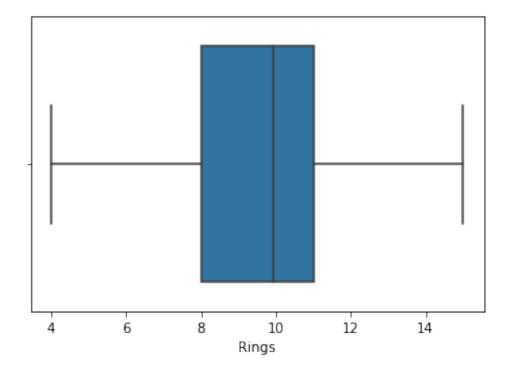
data['Rings']=np.where(data['Rings']>15.5,9.933684,data['Rings'])
sns.boxplot(x=data['Rings'])

<matplotlib.axes.\_subplots.AxesSubplot at 0x7f579b6a6b10>

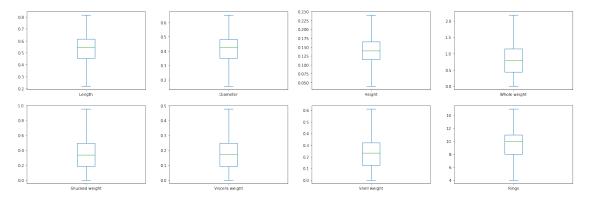


data['Rings']=np.where(data['Rings']<3.5,9.933684,data['Rings'])
sns.boxplot(x=data['Rings'])</pre>

<matplotlib.axes.\_subplots.AxesSubplot at 0x7f579b6e1dd0>



data.plot(kind="box", subplots=True, layout=(7,4), figsize=(25,30));



# Categorical value and encoding data.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 4177 entries, 0 to 4176
Data columns (total 9 columns):
 #
     Column
                     Non-Null Count
                                      Dtype
     -----
 0
                     4177 non-null
     Sex
                                      object
 1
     Length
                     4177 non-null
                                      float64
 2
                     4177 non-null
                                      float64
     Diameter
 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
                     4177 non-null
                                      float64
 8
     Rings
dtypes: float64(8), object(1)
memory usage: 293.8+ KB
data['Sex'].unique()
array(['M', 'F', 'I'], dtype=object)
#sex is categorical
df=pd.get dummies(data,columns=['Sex'])
# data['Sex'].replace({'I':2,'M':1,'F':0})
df.head()
```

Length weight \	Diameter	Height	Whole weight	Shucked weight	Viscera
0 0.455 0.1010	0.365	0.095	0.5140	0.2245	
1 0.350	0.265	0.090	0.2255	0.0995	
0.0485 2 0.530	0.420	0.135	0.6770	0.2565	
0.1415 3 0.440 0.1140	0.365	0.125	0.5160	0.2155	
0.1140					

```
0.0895
    0.330
              0.255
                       0.080
                                     0.2050
0.0395
   Shell weight
                  Rings
                                Sex_I
                                        Sex_M
                         Sex F
0
          0.150
                   15.0
                             0
                                     0
                                             1
1
          0.070
                             0
                                     0
                                            1
                    7.0
2
                                            0
          0.210
                    9.0
                             1
                                     0
3
                                            1
          0.155
                   10.0
                             0
                                     0
                                     1
4
          0.055
                             0
                                            0
                    7.0
     Split data into dependent and independent
# dependent variable
y=df['Rings'].values
У
array([15., 7., 9., ..., 9., 10., 12.])
#independent variable
x=df.drop(columns=['Rings'],axis=1).values
array([[0.455, 0.365, 0.095, ..., 0.
                                                 , 1.
                                         , 0.
                                                        ],
                                         , 0.
                                                 , 1.
       [0.35, 0.265, 0.09, ..., 0.
                                                        ],
       [0.53, 0.42, 0.135, \ldots, 1.
                                                 , 0.
                                         , 0.
                                                        ],
       [0.6 , 0.475, 0.205, ..., 0.
                                         , 0.
                                                 , 1.
                                                        ],
       [0.625, 0.485, 0.15, ..., 1.
                                         , 0.
                                                 , 0.
                                                        ],
                                                 , 1.
       [0.71, 0.555, 0.195, \ldots, 0.
                                         , 0.
                                                        ]])
9. Scale the independent
from sklearn.preprocessing import scale
x=scale(x)
Χ
array([[-0.66489959, -0.50167301, -1.19856285, ..., -0.67483383,
        -0.68801788.
                       1.31667716],
       [-1.60274931, -1.57291477, -1.332413, \ldots, -0.67483383,
                       1.31667716],
        -0.68801788,
                       0.08750996, -0.12776168, ..., 1.48184628,
       [ 0.00499306,
        -0.68801788, -0.75948762],
       [ 0.6302262 ,
                                     1.74614038, ..., -0.67483383,
                       0.67669293,
        -0.68801788,
                       1.316677161,
                                     0.27378876, ..., 1.48184628,
       [ 0.85352375,
                       0.78381711,
        -0.68801788,
                      -0.759487621,
       [ 1.61273542,
                       1.53368634,
                                     1.47844008, ..., -0.67483383,
        -0.68801788,
                       1.31667716]])
```

```
10 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,y,test size=0.2,rando)
m state=\overline{0}
  1. Build the model
from sklearn.linear model import LinearRegression
MLR = LinearRegression()
  1. Train the model
MLR.fit(x_train,y_train)
LinearRegression()
  1. Test the model
y pred = MLR.predict(x test)
y pred
                     8.53128844, 10.31763804, 6.01778772,
array([11.02365129,
10.51663617,
       10.89512079,
                     8.53196211,
                                   9.54586451, 7.95152418,
10.72178073,
                     6.65294931, 8.75326108, 9.27460362,
        8.19238901,
6.18153918,
        9.70376591,
                     8.18452763, 11.71357511, 10.74267823,
7.72965036,
        7.43397439,
                     6.82788969, 8.22939127, 8.2348197,
8.82476593,
                     9.59497196, 11.19988831, 9.37302799,
       10.51345867,
10.08604571,
                     8.66800526, 10.88798522, 11.84005218,
        8.45324439,
7.5373072
        9.27974088,
                     8.34202688, 10.38254851, 8.08254611,
10.65089215.
                     9.45949769, 10.41139418, 10.48701828,
       10.88811422,
11.13070906,
                     9.80512449, 11.2951419, 10.12506148,
        9.36162378,
7.62164999,
                     7.26624166, 9.08538073, 11.25316736,
       10.35728491,
9.474184
        8.52830245,
                     7.27348366,
                                   8.04099155, 7.26550445,
7.0395887
        9.89253159,
                     9.66439031, 10.09193427, 7.97386002,
7.64935223,
       10.96582784, 11.41070407, 11.35895017, 8.58799886,
11.94575588,
        8.74620319, 12.50735682, 10.74835336, 9.64759271,
```

9.83530772,

```
8.41026344, 9.40644191, 9.52895378, 10.56541533,
8.7013532
        9.7670024 , 6.40487946, 7.45598827, 11.39853523,
9.15789339.
        7.83546821, 9.96684249, 11.41990891, 7.11742083,
7.24436966.
        9.16285502. 10.47904748. 8.79231556. 5.9082895.
10.90752403,
        6.47414787, 9.06052315, 8.30465427, 11.86601677,
10.33232021.
       10.2744247 , 9.66334047, 10.22898588, 10.40454048,
5.39520268,
       10.55472615, 7.68430725, 6.89129645, 7.5155344,
10.83233917,
       13.45300333, 10.70567597, 10.40656464,
                                               9.00164862,
11.15303693,
       10.34722953, 11.12933409, 10.68415957,
                                               8.12100617,
9.02072389,
        7.11136852, 10.95197738, 6.94468787,
                                               8.46111397,
10.22042457.
       10.75440176, 10.57852592, 10.93541514, 7.61916515,
10.40572274,
       10.12603322, 8.70338674, 9.60530608, 10.64631183,
10.84270895,
       10.4761154 , 9.83340183,
                                 8.4058751 , 7.39202714,
11.86734229,
        9.95434057, 10.14417203,
                                  7.23838717, 9.15980009,
10.90326059,
       10.1232996 , 8.33240338,
                                  7.71050908, 7.71583601,
6.94324823,
        8.80850702, 11.94417741,
                                  7.12856184, 10.48419937,
8.14500812,
                                 7.1814626 , 11.22338772,
        7.08252022, 9.96347253,
8.21143832,
       10.05274103,
                    7.80970804.
                                  9.12979101, 10.54498862,
5.84913363,
       10.15530829,
                    7.80237104,
                                  6.86143402, 10.72872808,
8.90643711.
                                  8.84069972, 10.49156114,
        8.65106054,
                    8.02190199.
11.06367457,
                                  8.24247075, 8.29980727,
       10.15616176, 6.25121954,
10.19620653,
        8.88402839, 10.41495808,
                                  8.93468817, 8.57335993,
10.18536169,
        9.81124297, 9.75113181,
                                 8.26582913, 6.84376733,
9.10167901,
       10.06842574,
                    9.7056066 ,
                                 8.50473395, 10.04221647,
10.21483885,
        9.34117335, 7.78326258, 9.017291 , 7.2227604 ,
11.44924959,
```

```
9.68196466, 12.64732542,
        9.43605705,
                                              9.74052333,
9.83311448,
       10.95300262,
                    7.34164072, 11.25166692,
                                              8.92675143,
10.74264592.
                    9.05196494, 10.41558489,
        6.24157975,
                                              9.79361244,
10.64192093,
        8.98627249. 8.8530722. 10.36484886.
                                              6.18337987.
11.39209837,
       10.35167091, 11.28579772, 7.99744479, 9.79469752,
11.06964569,
       11.5038125 , 11.03731587,
                                 9.4371465 , 11.78481308,
8.5962956
        9.55800658, 8.97520047, 6.33582782, 10.73275444,
8.49003424,
       10.9768094 ,
                    7.79014651, 6.07252956, 8.30392393,
9.79660595,
        8.98549151, 9.81318203, 10.27121809, 10.21460751,
9.06363434,
        9.13332595, 10.56526963, 7.84827343, 10.27044904,
11.8371726 .
       11.24843427, 8.56546181, 11.28604499, 13.66644576,
9.31242148,
       10.43484148, 7.33753379, 8.67965132, 9.11132755,
11.90739446,
                    9.10800624, 10.57926192, 7.8924746,
       10.21549923,
9.70098758,
        8.17759292,
                    8.98154753, 7.78026489,
                                              7.51876621,
8.04839563,
        9.82187447, 8.87420721,
                                 8.60991464,
                                              8.60260248,
6.66932669,
       10.71413731, 11.21991745, 9.33000812, 10.12575586,
9.84505524,
        7.3608023 , 11.19687773 , 7.56371831 , 9.70991817 ,
9.90636044,
       10.13989273,
                    7.69655559, 10.17666427, 10.279817
10.98951057,
        7.44950848,
                    8.43525714, 6.17616945, 8.72001621,
7.331762
        8.70683904,
                    6.94464974, 6.58569497, 8.92298619,
12.16954851,
        9.25512937, 8.60560346, 11.28932034, 8.85188983,
9.25388696,
        8.21442742, 9.80449809, 9.99583257, 10.45040867,
9.60706078,
       10.17462466, 10.51429966,
                                 7.82529387, 9.05475689,
12.13310757,
       10.9884888 , 8.78452478,
                                 6.88948856, 7.879859 ,
10.05785367,
       10.46400889, 9.71902958, 9.17043486, 10.49160165,
11.09421192,
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```

1. Measure the performance using metrics
from sklearn.metrics import r2\_score
acc = r2\_score(y\_test,y\_pred)

acc

### 0.48342896299561555

### df.head()

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

	Shell	weight	Rings	Sex_F	Sex_I	Sex_M
0		0.150	15.0	_0	_0	_1
1		0.070	7.0	0	0	1
2		0.210	9.0	1	0	0
3		0.155	10.0	0	0	1
4		0.055	7.0	0	1	0

MLR.predict([[0.455,0.365,0.095,0.5140,0.2245,0.101,0.150,0,0,1]]) array([9.79146941])