#### Assignment-4

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#### 1.Loading Dataset into tool

from google.colab import files

uploaded = files.upload()

Upload widget is only available when the cell has been executed in the current browser session. Please rerun this cell to enable.

Saving abalone.csv to abalone.csv

import pandas as pd

import numpy as np

import matplotlib.pyplot as plt

import seaborn as sns

import warnings

warnings.filterwarnings('ignore')

data = pd.read\_csv("abalone.csv")

2.Performing Visualization

sns.boxplot(data['Diameter'])

#### **Univariate Analysis**

#### data.head()

Sex	Length Diameter Rings		Height	Whole weight		Shucked weight V		t Viscera weight	Shell weight	
0	М	0.455	0.365	0.095	0.5140	0.2245	0.1010	0.150	15	
1	М	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	1	0.330	0.255	0.080	0.2050	0.0895	0.0395	0.055	7	

```
plt.hist(data['Diameter'])
(array([ 13., 66., 180., 344., 513., 812., 1017., 934., 275.,
     23.]),
array([0.055, 0.1145, 0.174, 0.2335, 0.293, 0.3525, 0.412, 0.4715,
    0.531, 0.5905, 0.65]),
)
plt.plot(data['Diameter'].head(10))
[]
plt.pie(data['Diameter'].head(),autopct='%.3f')
([,
],
[Text(0.8507215626110557, 0.6973326486753676, "),
Text(-0.32611344931648134, 1.0505474849691026, "),
Text(-1.0998053664078908, -0.02069193128747144, "),
Text(-0.08269436219656089, -1.096887251480709, "),
Text(0.9758446362287218, -0.5076684409569241, ")],
[Text(0.46402994324239394, 0.3803632629138369, '21.856'),
Text(-0.17788006326353525, 0.5730259008922377, '15.868'),
Text(-0.5998938362224858, -0.011286507974984419, '25.150'),
Text(-0.045106015743578656, -0.5983021371712958, '21.856'),
Text(0.5322788924883937, -0.2769100587037768, '15.269')])
sns.distplot(data['Diameter'].head(300))
plt.scatter(data['Diameter'].head(400),data['Length'].head(400))
```

```
plt.bar(data['Sex'].head(20),data['Rings'].head(20))
plt.title('Bar plot')
plt.xlabel('Diameter')
plt.ylabel('Rings')
Text(0, 0.5, 'Rings')
sns.barplot(data['Sex'], data['Rings'])
sns.jointplot(data['Diameter'].head(50),data['Rings'].head(100))
sns.barplot('Diameter','Rings',hue='Sex',data=data.head())
sns.lineplot(data['Diameter'].head(),data['Rings'].head())
sns.boxplot(data['Sex'].head(10),data['Diameter'].head(10),data['Rings'].head(10))
fig=plt.figure(figsize=(8,5))
sns.heatmap(data.head().corr(),annot=True)
sns.pairplot(data.head(),hue='Height')
sns.pairplot(data.head())
3. Perform Descriptive Statistics on the dataset
data.head()
       Length Diameter
                            Height Whole weight Shucked weight Viscera weight Shell weight
Sex
       Rings
0
       Μ
              1
              M
```

2	F	0.530	0.420	0.135	0.6770	0.2565	0.1415	0.210	9			
3	М	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			
data.ta	ail()											
Sex	Length Rings	Diame	ter	Height	Whole	weight	Shucke	d weight	t Viscera weight	Shell weight		
4172	F	0.565	0.450	0.165	0.8870	0.3700	0.2390	0.2490	11			
4173	М	0.590	0.440	0.135	0.9660	0.4390	0.2145	0.2605	10			
4174	М	0.600	0.475	0.205	1.1760	0.5255	0.2875	0.3080	9			
4175	F	0.625	0.485	0.150	1.0945	0.5310	0.2610	0.2960	10			
4176	М	0.710	0.555	0.195	1.9485	0.9455	0.3765	0.4950	12			
data.ir	nfo()											
Rangel	RangeIndex: 4177 entries, 0 to 4176											
Data columns (total 9 columns):												
# Column Non-Null Count Dtype												
<del></del>												
0 Sex	( 4	177 nor	n-null o	bject								
1 Ler												

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

4177 non-null int64 8 Rings

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

memory usage: 293.8+ KB

data.describe()

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 0.238831	0.407881 9.933684	0.139516	0.828742	0.359367	0.180594
std	0.120093 0.139203	0.099240 3.224169	0.041827	0.490389	0.221963	0.109614
min	0.075000 0.001500	0.055000 1.000000	0.000000	0.002000	0.001000	0.000500
25%	0.450000 0.130000	0.350000 8.000000	0.115000	0.441500	0.186000	0.093500
50%	0.545000 0.234000	0.425000 9.000000	0.140000	0.799500	0.336000	0.171000
75%	0.615000 0.329000	0.480000 11.000000	0.165000	1.153000	0.502000	0.253000
max	0.815000 1.005000	0.650000 29.000000	1.130000	2.825500	1.488000	0.760000

data.mode().T

0 1

Sex M NaN

Length 0.55 0.625

Diameter 0.45 NaN

Height 0.15 NaN

Whole weight 0.2225 NaN

Shucked weight 0.175 NaN

Viscera weight 0.1715 NaN

Shell weight 0.275 NaN

Rings 9.0 NaN

data.shape

(4177, 9)

data.kurt()

Length 0.064621

Diameter -0.045476

Height 76.025509

Whole weight -0.023644

Shucked weight 0.595124

Viscera weight 0.084012

Shell weight 0.531926

Rings 2.330687

dtype: float64

data.skew()

Length -0.639873

Diameter -0.609198

Height 3.128817

Whole weight 0.530959

Shucked weight 0.719098

Viscera weight 0.591852

Shell weight 0.620927

Rings 1.114102

dtype: float64

data.var()

Length 0.014422

Diameter 0.009849

Height 0.001750

Whole weight 0.240481

Shucked weight 0.049268

Viscera weight 0.012015

Shell weight 0.019377

Rings 10.395266

dtype: float64

data.nunique()

Sex 3

Length 134

Diameter 111

Height 51

Whole weight 2429

Shucked weight 1515

Viscera weight 880

Shell weight 926

Rings 28

dtype: int64

4. Check for missing values and deal with them

## data.isna()

Sex	Length Rings	Diame	ter	Height	ight Whole weight		Shucke	t Viscera weight Shell weight	
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
	•••		•••						
4172	False	False	False	False	False	False	False	False	False
4173	False	False	False	False	False	False	False	False	False
4174	False	False	False	False	False	False	False	False	False
4175	False	False	False	False	False	False	False	False	False
4176	False	False	False	False	False	False	False	False	False

# 4177 rows × 9 columns

## data.isna().any()

Sex False

Length False

Diameter False

Height False

Whole weight False

Shucked weight False

Viscera weight False

Shell weight False

```
dtype: bool
data.isna().sum()
Sex
         0
Length
          0
Diameter
           0
Height
          0
Whole weight
Shucked weight 0
Viscera weight 0
Shell weight 0
Rings
          0
dtype: int64
data.isna().any().sum()
0
5. Find the outliers and replace them outliers
sns.boxplot(data['Diameter'])
quant=data.quantile(q=[0.25,0.75])
quant
Length Diameter
                   Height Whole weight Shucked weight Viscera weight Shell weight
      Rings
0.25
      0.450 0.35
                   0.75
      0.615 0.48
                   iqr=quant.loc[0.75]-quant.loc[0.25]
iqr
Length
          0.1650
Diameter
           0.1300
Height
          0.0500
Whole weight 0.7115
```

Rings

False

Shucked weight 0.3160

Viscera weight 0.1595

Shell weight 0.1990

Rings 3.0000

dtype: float64

low=quant.loc[0.25]-(1.5\*iqr)

low

Length 0.20250

Diameter 0.15500

Height 0.04000

Whole weight -0.62575

Shucked weight -0.28800

Viscera weight -0.14575

Shell weight -0.16850

Rings 3.50000

dtype: float64

up=quant.loc[0.75]+(1.5\*iqr)

up

Length 0.86250

Diameter 0.67500

Height 0.24000

Whole weight 2.22025

Shucked weight 0.97600

Viscera weight 0.49225

Shell weight 0.62750

Rings 15.50000

dtype: float64

data['Diameter']=np.where(data['Diameter']<0.155,0.4078,data['Diameter'])

sns.boxplot(data['Diameter'])

sns.boxplot(data['Length'])

```
data['Length']=np.where(data['Length']<0.23,0.52, data['Length'])
sns.boxplot(data['Length'])
sns.boxplot(data['Height'])
data['Height']=np.where(data['Height']<0.04,0.139, data['Height'])
data['Height']=np.where(data['Height']>0.23,0.139, data['Height'])
sns.boxplot(data['Height'])
sns.boxplot(data['Whole weight'])
data['Whole weight']=np.where(data['Whole weight']>0.9,0.82, data['Whole weight'])
sns.boxplot(data['Whole weight'])
sns.boxplot(data['Shucked weight'])
data['Shucked weight']=np.where(data['Shucked weight']>0.93,0.35, data['Shucked weight'])
sns.boxplot(data['Shucked weight'])
sns.boxplot(data['Viscera weight'])
data['Viscera weight']=np.where(data['Viscera weight']>0.46,0.18, data['Viscera weight'])
sns.boxplot(data['Viscera weight'])
sns.boxplot(data['Shell weight'])
data['Shell weight']=np.where(data['Shell weight']>0.61,0.2388, data['Shell weight'])
sns.boxplot(data['Shell weight'])
6. Check for Categorical columns and perform encoding.
```

## data['Sex'].replace({'M':1,'F':0,'l':2},inplace=True)

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Sex	Length Rings	Diamet	ter	Height	Whole weight		Shucked weight Viscera w		Viscera weight	Shell weight
0	1	0.455	0.365	0.095	0.5140	0.2245	0.1010	0.1500	15	
1	1	0.350	0.265	0.090	0.2255	0.0995	0.0485	0.0700	7	
2	0	0.530	0.420	0.135	0.6770	0.2565	0.1415	0.2100	9	
3	1	0.440	0.365	0.125	0.5160	0.2155	0.1140	0.1550	10	
4	2	0.330	0.255	0.080	0.2050	0.0895	0.0395	0.0550	7	
4172	0	0.565	0.450	0.165	0.8870	0.3700	0.2390	0.2490	11	
4173	1	0.590	0.440	0.135	0.8200	0.4390	0.2145	0.2605	10	
4174	1	0.600	0.475	0.205	0.8200	0.5255	0.2875	0.3080	9	
4175	0	0.625	0.485	0.150	0.8200	0.5310	0.2610	0.2960	10	
4176	1	0.710	0.555	0.195	0.8200	0.3500	0.3765	0.4950	12	

4177 rows × 9 columns

7. Split the data into dependent and independent variables.

```
x=data.drop(columns= ['Rings'])
```

y=data['Rings']

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Sex	Length	Diamet	er	Height	Whole	weight	Shucke	d weight Viscera weight	Shell weight
0	1	0.455	0.365	0.095	0.5140	0.2245	0.1010	0.1500	
1	1	0.350	0.265	0.090	0.2255	0.0995	0.0485	0.0700	
2	0	0.530	0.420	0.135	0.6770	0.2565	0.1415	0.2100	
3	1	0.440	0.365	0.125	0.5160	0.2155	0.1140	0.1550	
4	2	0.330	0.255	0.080	0.2050	0.0895	0.0395	0.0550	
4172	0	0.565	0.450	0.165	0.8870	0.3700	0.2390	0.2490	

```
4173 1
              4174 1
              0.600 \quad 0.475 \quad 0.205 \quad 0.8200 \ 0.5255 \ 0.2875 \ 0.3080
4175 0
              0.625 \quad 0.485 \quad 0.150 \quad 0.8200 \ 0.5310 \ 0.2610 \ 0.2960
4176 1
              0.710 \quad 0.555 \quad 0.195 \quad 0.8200 \ 0.3500 \ 0.3765 \ 0.4950
4177 rows × 8 columns
У
0
    15
1
     7
2
     9
3
    10
4
    7
4172 11
4173 10
4174 9
4175 10
4176 12
Name: Rings, Length: 4177, dtype: int64
8. Scale the independent variables
from sklearn.preprocessing import scale
x = scale(x)
array([[-0.0105225, -0.67088921, -0.50179694, ..., -0.61037964,
    -0.7328165, -0.64358742],
   [-0.0105225, -1.61376082, -1.57304487, ..., -1.22513334,
    -1.24343929, -1.25742181],
   [-1.26630752, 0.00259051, 0.08738942, ..., -0.45300269,
    -0.33890749, -0.18321163],
```

```
[-0.0105225, 0.63117159, 0.67657577, ..., 0.86994729,
    1.08111018, 0.56873549],
   [-1.26630752, 0.85566483, 0.78370057, ..., 0.89699645,
    0.82336724, 0.47666033],
   [-0.0105225, 1.61894185, 1.53357412, ..., 0.00683308,
    1.94673739, 2.00357336]])
9. 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)
print(x_train.shape, x_test.shape)
(3341, 8) (836, 8)
10.Build the Model
from sklearn.linear_model import LinearRegression
MLR=LinearRegression()
11.Train the model
MLR.fit(x_train,y_train)
LinearRegression()
12.Test the model
y_pred=MLR.predict(x_test)
y pred
array([6.3204331, 10.41671748, 13.91911179, 12.29316277, 8.7273177,
   11.04369928, 12.40210281, 11.6992544, 12.01785949, 6.57983392,
   11.91353764, 10.79661591, 11.56560952, 10.14326497, 13.16762604,
    9.34621768, 10.76904478, 11.88283609, 9.34461447, 10.08802992,
   12.80140942, 9.58177975, 11.20908126, 10.3662699, 10.0168299,
   15.92815446, 15.93700213, 7.36066362, 13.2889134, 10.1579858,
   11.62833855, 11.08597007, 11.60253151, 11.74194458, 9.75151497,
```

9.16685512, 7.93960537, 10.04563481, 10.81773394, 10.55133893, 7.19389026, 9.30303442, 10.83957317, 10.63432914, 10.19371808, 13.47423856, 9.06825076, 6.69843582, 13.38213142, 9.62823486, 8.20174551, 7.79183041, 9.3338472, 11.08195328, 11.25321895, 6.11231204, 10.6960639, 9.23348159, 7.76425036, 11.65342323,12.6024271, 7.49694081, 9.71678931, 7.41119139, 6.94925679, 6.34706174, 9.99734923, 6.70117631, 10.71374432, 9.59457302, 7.07847213, 6.6940933, 9.30356123, 13.66698224, 9.71369221, 17.36952958, 7.81225327, 8.86909973, 9.29540502, 11.03405521, 12.90720962, 13.03952065, 4.90843127, 9.50619996, 10.09434256, 8.67296752, 9.03746047, 8.33310609, 10.60445018, 9.66636969, 7.67351279, 8.74447193, 12.37470593, 7.70552082, 11.35599144, 11.25726129, 10.02276461, 8.01953433, 11.39538114, 7.92288557, 11.02588274, 7.02530311, 10.80014326, 13.22266766, 11.41469264, 7.5577235 , 6.83654146 , 6.97820486 , 10.29150052 , 9.1851768 , 9.72122817, 9.29569276, 11.98122676, 9.87982582, 8.55374278, 7.67912597, 10.93152036, 11.90656204, 11.93625854, 12.59760271, 11.87092001, 5.99671728, 9.20248712, 11.18185068, 11.13316757, 12.85726928, 9.50993961, 9.39438115, 10.55793101, 8.61221838, 7.12344177, 7.0075169, 7.56528442, 14.02672909, 13.39176121, 10.27099354, 13.04124533, 9.72264547, 11.63284409, 3.06922786, 8.60297955, 10.80917425, 11.27118306, 6.4757245, 10.27830248, 10.17409116, 10.39178358, 6.11330216, 8.27295199, 12.0413644, 10.43536813, 11.12820999, 10.56478101, 12.12900686, 9.0459273, 6.50569617, 8.65471113, 11.17391657, 4.17641665, 6.45933408, 10.94174559, 10.56404265, 7.32806471, 10.90718067, 8.76983179, 9.54866214, 9.71969088, 9.19215908, 11.19107958, 9.95023994, 10.33050587, 11.98860703, 5.76011208, 8.82560871, 8.26963359, 6.41006108, 7.62776781, 7.77958091, 10.53587014, 8.89399096, 11.50322847, 6.46552063, 6.62035734, 11.27313616, 8.28747988, 12.05544015, 11.6973709, 12.73972343, 11.36996234, 7.97256548,

9.42073857, 11.25296103, 8.05208624, 10.99827477, 8.28671759, 11.9443616, 11.82872121, 9.74400382, 8.90145486, 8.57310105, 7.40827472, 11.17489105, 10.0697987, 9.82070981, 7.33964403, 14.9428325, 7.76026974, 12.77292992, 6.50073351, 11.29473941, 11.88889387, 7.67192672, 11.10156897, 12.84247625, 6.80849608,