Assignment - 3
Problem Statement: Abalone Age Prediction

ASSIGNMENT DATE	19 OCTOBER 2022
STUDENT NAME	M.TARUNIKA
STUDENT ROLL NUMBER	19BEC020
MAXIMUM MARKS	2 MARKS

#1.Download the dataset

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

#2. Load the dataset into the tool

df=pd.read_csv("abalone.csv") df.head()

15
7
9
10
7
)

df.tail()

Whol

	Shuc	ke Vis	cer Shell
Se Lengt Diamete Heigh e Ring x	ır t weigh d a	weigh s	

					t	weight	weight t	
417	F	0.565	0.450	0.165	0.8870	0.3700	0.2390 0.2490	11
417	M	0.590	0.440	0.135	0.9660	0.4390	0.2145 0.2605	10
417	M	0.600	0.475	0.205	1.1760	0.5255	0.2875 0.3080	9
417	F	0.625	0.485	0.150	1.0945	0.5310	0.2610 0.2960	10
417 6	M	0.710	0.555	0.195	1.9485	0.9455	0.3765 0.4950	12

df.shape

(4177, 9) df.info()

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 #3. Perform Below Visualizations

#Univariate Analysis sns.boxplot(x=df['Height'])

#Bi-Variate Analysis sns.lineplot(df['Sex'],df['Length'])

C:\Users\shire\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pass t he following variables as keyword args: x, y. From version 0.12, the only valid positional arg ument will be `data`, and passing other arguments without an explicit keyword will result in a n error or misinterpretation.

warnings.warn(

#Multi-Variate Analysis sns.heatmap(df.corr(),annot=**True**)

#4. Perform descriptive statistics on the dataset df.describe()

				Shucke			
	Diamet		Whole		Viscera	Shell	
Length	er	Height	weight	d	weight	weight	Rings
				weight			

	Diamet		Whole		Viscera	Shell	
Length	er	Height	weight	d	weight	weight	Rings
				weight			

me 0.52399 0.40788 0.13951 0.82874 0.35936 0.18059 0.23883 9.93368 an 2 1 6 2 7 4 1 4

	0.12009	0.09924	0.04182	0.49038	0.22196	0.10961	0.13920	3.22416				
std	3	0	7	9	3	4	3	9				
mi	mi 0.07500 0.05500 0.00000 0.00200 0.00100 0.00050 0.00150 1.00000 n 0 0 0 0 0 0 0											
25	0.45000	0.35000	0.11500	0.44150	0.18600	0.09350	0.13000	8.00000				
%	0	0	0	0	0	0	0	0				
50	0.54500	0.42500	0.14000	0.79950	0.33600	0.17100	0.23400	9.00000				
%	0	0	0	0	0	0	0	0				
75	0.61500	0.48000	0.16500	1.15300	0.50200	0.25300	0.32900	11.0000				
%	0	0	0	0	0	0	0	00				

ma 0.81500 0.65000 1.13000 2.82550 1.48800 0.76000 1.00500 29.0000 **x** 0 0 0 0 0 0 0 0 0 df.describe().T

	count	mean	std	min	25%	50%	75%	max
Length	4177.0	0.523992	0.120093	0.0750	0.4500	0.5450	0.615	0.8150
	count	mean	std	min	25%	50%	75%	max
Diameter	4177.0	0.407881	0.099240	0.0550	0.3500	0.4250	0.480	0.6500
	4177.0	0.139516	0.041827	0.0000	0.1150	0.1400	0.165	1.1300
Height								

Whole	4177.0	0.828742	0.490389	0.0020	0.4415	0.7995	1.153	2.8255
weight Shucked								

	4177.0	0.359367	0.221963	0.0010	0.1860	0.3360	0.502	1.4880
weight Viscera	4177.0	0.180594	0.109614	0.0005	0.0935	0.1710	0.253	0.7600
weight Shell weight	4177.0	0.238831	0.139203	0.0015	0.1300	0.2340	0.329	1.0050

Rings 4177.0 9.933684 3.224169 1.0000 8.0000 9.0000 11.000 29.0000 #5. Check for Missing values and deal with them

df.isna().sum()

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. Find the outliers and replace them outliers df['Sex'].replace({'M':1, 'F':0, 'I':-1},inplace=**True**) df.head()

	Sex	Length	Diameter	Height	Whole weight	Shucked weight		Shell weight	Rings
0	1	0.455	0.365	0.095	0.5140	0.2245	0.1010	0.150	15
1	1	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
       1
            0.440
                        0.365
                                  0.125
                                          0.5160
                                                      0.2155
                                                                0.1140
                                                                           0.155
                                                                                      10
                                                                                       7
                        0.255
                                  0.080
                                          0.2050
                                                      0.0895
                                                                0.0395
                                                                           0.055
4
      -1
            0.330
df.Sex.unique()
array([ 1, 0, -1], dtype=int64)
sns.boxplot(x=df["Sex"])
sns.boxplot(x=df["Length"])
sns.boxplot(x=df["Diameter"])
sns.boxplot(x=df["Height"])
sns.boxplot(x=df["Whole weight"])
sns.boxplot(x=df["Shucked weight"])
sns.boxplot(x=df["Viscera weight"])
sns.boxplot(x=df["Shell weight"])
sns.boxplot(x=df["Rings"])
#handle outlier
qnt=df-quantile(q=[0.25,0.75]) qnt
```

	Se	Lengt	Diamete	Heigh	Whol	Shucke	Viscer	Shell	Ring
	X	h	r	t	e	d	a weight	weigh	s
					weigh	weight	weight	t	
					t				
0.2 5	1.0	0.450	0.35	0.115	0.4415	0.186	0.0935	0.130	8.0
0.7 5	1.0	0.615	0.48	0.165	1.1530	0.502	0.2530	0.329	11.0

iqr=qnt.loc[0.75]-qnt.loc[0.25]iqr

Sex 2.0000

Length 0.1650

Diameter 0.1300

Height 0.0500

Whole weight 0.7115

Shucked weight 0.3160

Viscera weight 0.1595

Shell weight 0.1990 Rings

3.0000 dtype: float64 #lower

limit lower=qnt.loc[0.25]-

(1.5*iqr) lower

Sex -4.00000

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
```

upper=qnt.loc[0.75]+(1.5*iqr)

upper

Sex 4.00000

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 df.mean()

Sex 0.044530

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

#replace outlier

df['Length']=np.where(df['Length']<0.22,0.52,df['Length'])

df['Diameter']=np.where(df['Diameter']<0.155,0.407,df['Diameter'])

df['Height']=np.where(df['Height']<0.04,0.13,df['Height'])

df['Height']=np.where(df['Height']>0.24,0.13,df['Height'])

df['Whole weight']=np.where(df['Whole weight']>2.18,0.83,df['Whole weight'])

df['Shucked weight']=np.where(df['Shucked weight']>0.958,0.359367,df['Shucked weight'])

df['Viscera weight']=np.where(df['Viscera weight']>0.478,0.18,df['Viscera weight'])

df['Shell weight']=np.where(df['Shell weight']>0.61,0.238831,df['Shell weight'])

df['Rings']=np.where(df['Rings']<3.5,9.93,df['Rings'])

df['Rings']=np.where(df['Rings']>15.5,9.93,df['Rings']) sns.boxplot(df['Length'])

C:\Users\shire\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pass t he following variable as a keyword arg: x. From version 0.12, the only valid positional argum ent will be `data`, and passing other arguments without an explicit keyword will result in an e rror or misinterpretation.

warnings.warn(

sns.boxplot(df['Diameter'])

C:\Users\shire\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pass t he following variable as a keyword arg: x. From version 0.12, the only valid positional argum ent will be `data`, and passing other arguments without an explicit keyword will result in an e rror or misinterpretation.

warnings.warn(

sns.boxplot(df['Height'])

C:\Users\shire\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pass t he following variable as a keyword arg: x. From version 0.12, the only valid positional argum ent will be `data`, and passing other arguments without an explicit keyword will result in an e rror or misinterpretation.

warnings.warn(

sns.boxplot(df['Whole weight'])

C:\Users\shire\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pass t he following variable as a keyword arg: x. From version 0.12, the only valid positional argum ent will be `data`, and passing other arguments without an explicit keyword will result in an e rror or misinterpretation.

warnings.warn(

sns.boxplot(df['Shucked weight'])

C:\Users\shire\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pass t he following variable as a keyword arg: x. From version 0.12, the only valid positional argum ent will be `data`, and passing other arguments without an explicit keyword will result in an e rror or misinterpretation.

warnings.warn(

sns.boxplot(df['Viscera weight'])

C:\Users\shire\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pass t he following variable as a keyword arg: x. From version 0.12, the only valid positional argum ent will be `data`, and passing other arguments without an explicit keyword will result in an e rror or misinterpretation.

warnings.warn(

sns.boxplot(df['Shell weight'])

C:\Users\shire\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pass t he following variable as a keyword arg: x. From version 0.12, the only valid positional argum ent will be `data`, and passing other arguments without an explicit keyword will result in an e rror or misinterpretation.

warnings.warn(

sns.boxplot(df['Rings'])

C:\Users\shire\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pass t he following variable as a keyword arg: x. From version 0.12, the only valid positional argum ent will be `data`, and passing other arguments without an explicit keyword will result in an e rror or misinterpretation.

warnings.warn(

#7. Check for Categorical columns and perform encoding df.head()

#sex is categorical and encoding is performed

						Whole	Shucked	Viscera	Shell
	Sex	Length	Diamet	er He	eight	weight	weight	weight	weight Rings
0	1	0.455	0.365	0.095	0.5140)	0.2245	0.1010	0.150
		15.0							

1	1	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	1	0.440 10.0	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			
#8. Split the data into dependent and independent variables #independent variable											
x=df.drop(columns=['Rings'],axis=1).values x											
array([[1. , 0.455 , 0.365 ,, 0.2245, 0.101 , 0.15], [1. , 0.35 , 0.265 ,, 0.0995, 0.0485, 0.07], [0. , 0.53 , 0.42 ,, 0.2565, 0.1415, 0.21],											
•	·· · ,										
	[1.	, 0.6 ,	0.475,	, 0.52	55, 0.2875, 0.3	308],					
[0. , 0.625, 0.485,, 0.531, 0.261, 0.296],											
[1. , 0.71 , 0.555 ,, 0.9455, 0.3765, 0.495]])											
_	endent vo										
y=df['Rings'].values y											
array([15., 7., 9.,, 9., 10., 12.]) #9.											
Scale the independent variables from											
sklearn.preprocessing import scale											
x=scale(x) x											
array([[1.15434629, -0.66347373, -0.50167301,, -0.61177023,											
-0.73234257, -0.64358992],											
[1.15434629, -1.60127264, -1.57291477,, -1.21969385,											

7.0

9.0

7.0

 $[\ 1.15434629,\ 0.63158191,\ 0.67669293,...,\ 0.85210986,$

 $[-0.05379815,\ 0.00638264,\ 0.08750996,...,\ -0.45614178,$

-1.23612645, -1.25742425],

-0.34370929, -0.18321418],

...,

```
[-0.05379815, 0.85486737, 0.78381711, ..., 0.8788585,
     0.80299878, 0.47665772],
    [ 1.15434629, 1.61403792, 1.53368634, ..., 2.89473324,
     1.91132331, 2.0035706 ]]) #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)
#11. Build the Model
from sklearn.linear model import LinearRegression linreg=LinearRegression()
#12. Train the Model linreg.fit(x_train,y_train)
LinearRegression() #13. Test
the Model
test_pred=linreg.predict(x_test)
test_pred
array([10.49406044, 14.63071197, 9.5052661, 7.12175027, 9.59508595,
9.4343576, 8.77992046, 10.17130406, 7.45502774, 9.87604313,
    10.98645479, 7.5538718, 8.87927518, 9.7638226, 8.54256728,
    10.42387201, 9.10033819, 9.87903278, 11.42897247, 7.06763663,
    10.57223182, 9.76975285, 12.30738965, 8.55382376, 9.52374863,
    8.21301289, 6.25183627, 7.12742482, 9.74741373, 10.3017582,
    9.82726168, 9.84749184, 10.4108395, 10.3081998, 10.08572396,
    8.30245647, 7.235845, 6.74452118, 10.42584137, 7.64274971,
    7.14405667, 9.16150599, 8.70935569, 10.74880185, 9.86452375,
    12.88609365, 6.57858505, 9.5398517, 6.81250209, 10.60088961,
    10.58682023, 10.59758934, 10.63987208, 10.60373354, 9.03578911,
    8.62103663, 9.90652623, 7.02963956, 9.84641914, 8.62932278,
    7.71223792, 11.69923451, 11.10448696, 8.06123754, 8.55513658,
    13.39968976, 8.26727764, 9.52753025, 9.09315656, 12.58339768,
    9.99703469, 10.24999324, 9.29384572, 10.84986883, 9.23432613,
```

1.05728969, 0.56873287],

```
7.71248702, 10.82510489, 9.74110842, 10.18617001, 11.15757814,
8.15589364, 7.74042932, 6.90572983, 10.00216891, 12.35623317,
9.2594473, 9.83903046, 8.79445305, 9.98771476, 10.72074918,
5.76586411, 8.83952495, 7.82141633, 9.27397291, 10.08449131,
7.97368561, 8.13133341, 10.58531402, 8.54116758, 8.87592087,
10.27752815, 9.91826533, 7.35190815, 10.30758392, 7.30769068,
10.3549833, 10.69101603, 10.1181462, 10.39559027, 11.9945787,
10.05265786, 12.85497306, 11.33865314, 10.6160416, 10.86643523,
9.98776731, 10.07059534, 7.51087688, 8.81450733, 10.76394848,
9.23449231, 8.9394567, 11.35528501, 7.02952734, 8.22981655,
7.39038626, 7.16648403, 7.72492669, 6.96924802, 7.78201642,
7.17710403, 9.82222011, 9.56803182, 8.40217156, 8.3040808,
9.19097285, 7.27282145, 8.7291546, 8.02818234, 9.6287928,
9.17367559, 10.67429449, 10.83594529, 10.03487667, 7.01082421,
8.22106326, 9.52078398, 12.01200605, 7.0664238, 7.02545033,
6.38664272, 9.03716991, 9.89980919, 9.54143876, 10.48601031,
7.89737086, 10.57993475, 12.60549688, 8.9722634, 8.86375281,
10.58737471, 8.23508559, 9.16831774, 11.32643922, 11.72162036,
7.35637849, 7.57148604, 7.1648948, 10.85620295, 9.55486626,
10.68453461, 10.42003548, 9.94733416, 11.13891581, 9.01364719,
7.82060141, 10.78208786, 7.46904197, 9.32761963, 7.78647994,
10.75827275, 8.09475084, 9.26765508, 9.58812949, 7.26964315,
8.97532078, 8.90396235, 6.62637508, 7.78750708, 8.243058,
9.46740388, 8.01654749, 8.84610761, 12.06376478, 11.18458934,
7.95791777, 8.73139889, 7.63438426, 10.19784773, 10.19657975,
9.88547762, 8.18847269, 7.75134569, 7.93222173, 8.53043085,
11.47767482, 11.63701859, 9.67054006, 7.15334679, 11.58254568,
10.91672544, 10.65123953, 11.30462744, 8.01570854, 8.691925,
6.99630889, 10.45505798, 11.08400844, 7.84853522, 7.89503444,
10.36775292, 9.29193168, 8.45869519, 9.40891292, 8.71995183,
10.41488943, 9.80584287, 9.40871844, 10.47585472, 6.77413109,
10.07855451, 9.36989613, 12.40825012, 8.71057984, 9.97974427,
9.26533226, 10.63083868, 9.49615866, 10.23657265, 11.25380255,
10.65503119, 7.22469252, 10.23933921, 11.66614343, 7.52501383,
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#14. Measure the performance using Metrics.

from sklearn import metrics from sklearn.metrics import mean_squared_error metrics.r2_score(y_test,test_pred) 0.4166836799902973 df.head()

	Sex	Length	Diamete	r He		Whole weight	Shucked weight	Viscera weight	Shell weight Rings
0	1	0.455 15.0	0.365 (0.095	0.5140		0.2245	0.1010	0.150
1	1	0.350	0.265	0.090	0.2255		0.0995	0.0485	0.070 7.0

2	0	0.530	0.420	0.135	0.6770	0.2565	0.1415	0.210	9.0
3	1	0.440 10.0	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	7.0

 $linreg. predict([[0.455, 0.365, 0.095, 0.5140, 0.2245, 0.1010, 0.150, 15.0]])\ array([21.53400745])$