Assignment -3
Problem Statement: Abalone Age Prediction

ASSIGNMENT DATE	19 OCTOBER 2022
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STUDENT ROLL NUMBER	CS19036
MAXIMUM MARKS	2 mark

#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()

	Sex 1	Length	Diame	eter H	eight we	Whole Shucl	ked Viscer weight weight		Shell
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 df.tail()

Whol

Shucke Viscer Shell

## Se Lengt Diamete Heigh e Ring x h r t weigh d a weigh s

			weight	weight t 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 5	F	0.625 0.485 0.150 1.0945	0.5310	0.2610 0.2960	10
<b>417</b> 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 wei	ght 4177 non-null float64
5	Shucked w	eight 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	Shel	l	
Length Rings	er	Height	weight weight	d	weight	weight

	Diamet	Whole	Viscera	Shel	1	
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

std

 $\mathbf{mi}\ 0.07500\ 0.05500\ 0.00000\ 0.00200\ 0.00100\ 0.00050\ 0.00150\ 1.00000\ \mathbf{n}\ 0\ 0\ 0\ 0\ 0\ 0\ 0$ 

<b>25</b> 0.4500 0.130		0.35 8.00		0.11	500	0.44	150	0.18600	0.09350
<b>%</b> 0	0	0	0	0	0	0	0		
<b>50</b> 0.5450 0.234		0.42 9.00		0.14	000	0.79	950	0.33600	0.17100
<b>%</b> 0	0	0	0	0	0	0	0		
<b>75</b> 0.6150	_	0.48 11.0		0.16	500	1.15	300	0.50200	0.25300
<b>%</b> 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

df.describe().T

		count mean	ı std	min 25%	50% 75%	max
<b>Length</b> 417 0.5450	7.0 0.5 0.615 0.8		20093	0.0750	0.4500	
cour Diameter 4177.	nt mean	std	min 0.3500	<b>25% 50%</b> 0.4250	<b>75%</b> 0.480	<b>max</b> 0.6500
Height	4177.0 0.13	9516 0.041827	0.0000	0.1150 0.1400	0.165	1.1300
Whole						
	4177.0 0.82	8742 0.490389	0.0020	0.4415 0.7995	1.153	2.8255
weight Shucked						

**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()

	L	Sex Length	Diameter	Height	Whole weight	Shucked weight	Viscera 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
                        0.255
                                  0.080
                                          0.2050
                                                      0.0895
                                                                0.0395
                                                                           0.055
                                                                                       7
4
              0.330
       -1
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 d weigh s weigh weight t r weight t t 0.450 0.35 0.115 0.4415 0.186 0.0935 8.0 0.130 0.2 1.0 5 0.7 1.0 0.48 0.615 0.165 1.1530 0.502 0.2530 0.329 11.0 **5** 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

		Who	ole Shucked	Viscera	Shell
Sex	Length	Diameter	Height	weight	weight
		weight	weight Ring	gs	

```
0.455 0.365 0.095 0.5140
                                                                 0.1010
                                                                                0.150 15.0
 0
       1
                                                   0.2245
 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
              0.440 0.365 0.125 0.5160
                                                                                0.155 10.0
       1
                                                   0.2155
                                                                 0.1140
              0.330 0.255 0.080 0.2050
                                                   0.0895
                                                                 0.0395
                                                                                0.055 7.0
       -1
#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],
, 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, ..., 0.5255, 0.2875, 0.308],
       [0. , 0.625, 0.485, ..., 0.531, 0.261, 0.296],
       [1. , 0.71 , 0.555 , ..., 0.9455, 0.3765, 0.495 ]])
#dependent variable
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,
    -1.23612645, -1.25742425],
    [-0.05379815, 0.00638264, 0.08750996, ..., -0.45614178,
    -0.34370929, -0.18321418],
    ...,
```

```
[ 1.15434629, 0.63158191, 0.67669293, ..., 0.85210986,
     1.05728969, 0.56873287],
    [-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,
```

```
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()

				Whole	e Shucked	Viscera	Shell
		Sex Le	ength	Diameter	Height	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 7.0

2	0	0.530 0.420 0.135 0.6770	0.2565	0.1415	0.210 9.0
3	1	0.440 0.365 0.125 0.5160	0.2155	0.1140	0.155 10.0
4	-1	0.330 0.255 0.080 0.2050	0.0895	0.0395	0.055 7.0

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