Assignment - 3
Problem Statement: Abalone Age Prediction

| ASSIGNMENT DATE | 19 OCTOBER 2022 |
|---------------------|-----------------|
| STUDENT NAME | G.NANDHINI |
| STUDENT ROLL NUMBER | 19BEC026 |
| 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 |

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