TEAM ID	PNT2022TMID27851
STUDENT NAME	R. NIVETHA
DOMAIN NAME	HEALTH CARE
PROJECT NAME	EARLY DETECTION
	OF CHRONIC
	KIDNEY DISEASE
	USING MACHINE
	LEARNING
MAXIMUM MARKS	2 MARKS

• LOAD THE DATASET INTO THE DATASET

LOAD DATASET

```
' [1] import pandas as pd
    import numpy as np
    import matplotlib.pyplot as plt
    import seaborn as sns

' [3] df = pd.read_csv('abalone.csv')
```

• PERFORM BELOW VISUALIZATIONS

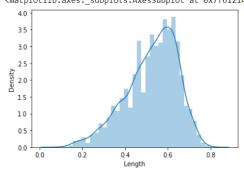
• UNIVARIANT

VISUALIZATIONS

UNIVARIANT ANALYSIS

[4] sns.distplot(df.Length)

/usr/local/lib/python3.7/dist-packages/seaborn/distributions.py:2619: FutureWarning: `distplot` is a deprecated funwarnings.warn(msg, FutureWarning)
<matplotlib.axes._subplots.AxesSubplot at 0x7f61214aa210>

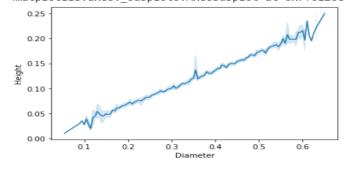


• BIVARIANT

BIVARIANT ANALYSIS

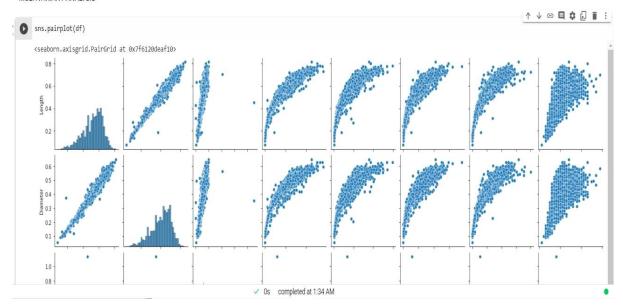
[5] sns.lineplot(df.Diameter,df.Height)

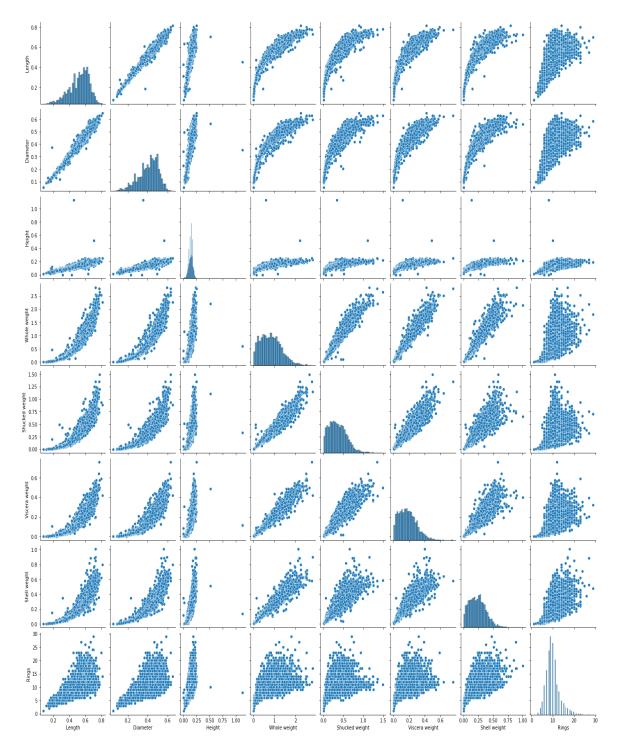
/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass FutureWarning
<matplotlib.axes._subplots.AxesSubplot at 0x7f6120e37150>



MULTIVARIANT

MULTIVARIANT ANALYSIS





• PERFORM DESCRIPTIVE STATISTICS ONTHE DATASET

DESCRIPTIVE STATISTICS

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

✓ [8] df.shape

(4177, 9)

[9] df.info()

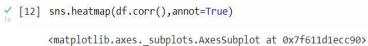
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 4177 entries, 0 to 4176
Data columns (total 9 columns):

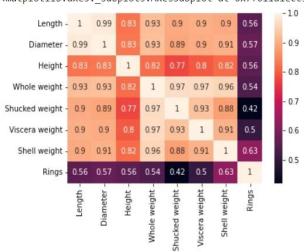
,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,	coramiio (cocar	5 001411110/1	
#	Column	Non-Null Count	Dtype
0	Sex	4177 non-null	object
1	Length	4177 non-null	float64
2	Diameter	4177 non-null	float64
3	Height	4177 non-null	float64
4	Whole weight	4177 non-null	float64
5	Shucked weight	4177 non-null	float64
6	Viscera weight	4177 non-null	float64
7	Shell weight	4177 non-null	float64
8	Rings	4177 non-null	int64
dtype	es: float64(7),	int64(1), object	(1)

memory usage: 293.8+ KB

1

	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	Rings
Length	1.000000	0.986812	0.827554	0.925261	0.897914	0.903018	0.897706	0.556720
Diameter	0.986812	1.000000	0.833684	0.925452	0.893162	0.899724	0.905330	0.574660
Height	0.827554	0.833684	1.000000	0.819221	0.774972	0.798319	0.817338	0.557467
Whole weight	0.925261	0.925452	0.819221	1.000000	0.969405	0.966375	0.955355	0.540390
Shucked weight	0.897914	0.893162	0.774972	0.969405	1.000000	0.931961	0.882617	0.420884
Viscera weight	0.903018	0.899724	0.798319	0.966375	0.931961	1.000000	0.907656	0.503819
Shell weight	0.897706	0.905330	0.817338	0.955355	0.882617	0.907656	1.000000	0.627574
Rings	0.556720	0.574660	0.557467	0.540390	0.420884	0.503819	0.627574	1.000000





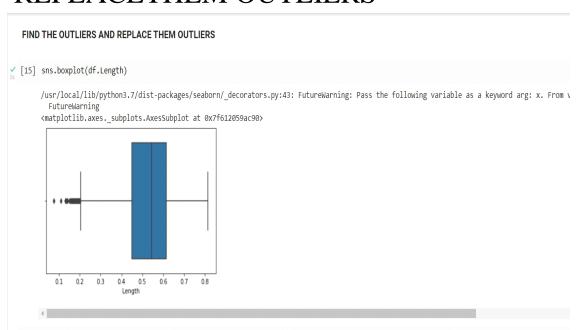
• CHECK FOR MISSING VALUES AND DEALWITH THEM

CHECK FOR MISSING VALUES AND DEAL WITH THEM

```
Sex False
Length False
Diameter False
Height False
Whole weight False
Shucked weight False
Viscera weight False
Shell weight False
Rings False
dtype: bool

* [14] df['Length'].fillna(df['Length'].mean(),inplace=True)
```

• FIND THE OUTLIERS AND REPLACETHEM OUTLIERS



```
✓ [16] df.median()
```

/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:1: FutureWarı """Entry point for launching an IPython kernel.

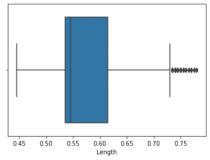
Length 0.5450 Diameter 0.4250 Height 0.1400 Whole weight 0.7995 Shucked weight 0.3360 Viscera weight 0.1710 Shell weight 0.2340 Rings 9.0000

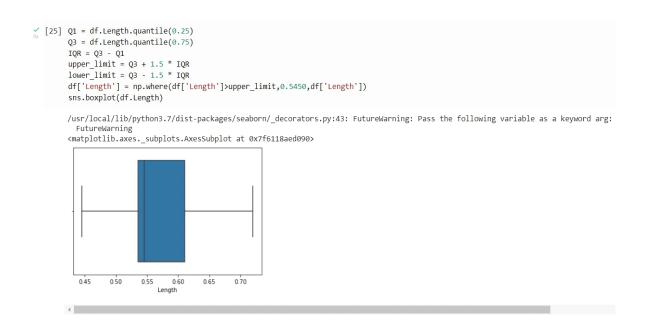
dtype: float64

```
[23] Q1 = df.Length.quantile(0.25)
      Q3 = df.Length.quantile(0.75)
     IQR = Q3 - Q1
     upper_limit = Q3 + 1.5 * IQR
lower_limit = Q3 - 1.5 * IQR
      df['Length'] = np.where(df['Length']<lower_limit,0.5450,df['Length'])</pre>
      sns.boxplot(df.Length)
```

/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variable as a ke FutureWarning

<matplotlib.axes._subplots.AxesSubplot at 0x7f6118b5d090>





CHECK FOR CATEGORICAL COLUMNAND PERFORM ENCODING

CHECK FOR CATEGORICAL COLUMN AND PERFORM ENCODING

```
√ [30] from sklearn.preprocessing import LabelEncoder
/ [32] le = LabelEncoder()
        df.Sex = le.fit_transform(df.Sex)
        df.head()
            Sex Length Diameter Height Whole weight Shucked weight Viscera weight Shell weight Rings
                  0.455
                            0.365
                                    0.095
                                                  0.5140
                                                                  0.2245
                                                                                  0.1010
                                                                                                  0.150
                                                                                  0.0485
                                                                                                  0.070
                  0.545
                            0.265
                                    0.090
                                                  0.2255
                                                                  0.0995
                                                                                                            7
             2
                  0.530
                            0.420
                                    0.135
                                                  0.6770
                                                                  0.2565
                                                                                  0.1415
                                                                                                  0.210
                                                                                                            9
             2
                  0.545
                            0.365
                                                  0.5160
                                                                  0.2155
                                                                                  0.1140
                                                                                                  0.155
                                                                                                           10
                                    0.125
                  0.545
                            0.255
                                    0.080
                                                  0.2050
                                                                  0.0895
                                                                                  0.0395
                                                                                                  0.055
```

• SPLIT THE DATA INTO DEPENDENT ANDINDEPENDENT VARIABLES

SPLIT THE DATA INTO DEPENDENT AND INDEPENDENT VARIABLES

Name: Rings, dtype: int64

	Sex	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight
0	2	0.455	0.365	0.095	0.5140	0.2245	0.1010	0.150
1	2	0.545	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	2	0.545	0.365	0.125	0.5160	0.2155	0.1140	0.155
4	1	0.545	0.255	0.080	0.2050	0.0895	0.0395	0.055
y =		Rings	0.200	0.000	0.2000	0.0000	0.0000	0.000
0 1	15							
1 2	7							

• SCALE THE INDEPENDENT VARIABLES

SCALE THE INDEPENDENT VARIABLES

✓ [38] from sklearn.preprocessing import MinMaxScaler

	Sex	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight
0	1.0	0.036364	0.521008	0.084071	0.181335	0.150303	0.132324	0.147982
1	1.0	0.363636	0.352941	0.079646	0.079157	0.066241	0.063199	0.068261
2	0.0	0.309091	0.613445	0.119469	0.239065	0.171822	0.185648	0.207773
3	1.0	0.363636	0.521008	0.110619	0.182044	0.144250	0.149440	0.152965
4	0.5	0.363636	0.336134	0.070796	0.071897	0.059516	0.051350	0.053313
	3.0			inc		***	***	***
4172	0.0	0.436364	0.663866	0.146018	0.313441	0.248151	0.314022	0.246637
4173	1.0	0.527273	0.647059	0.119469	0.341420	0.294553	0.281764	0.258097
4174	1.0	0.563636	0.705882	0.181416	0.415796	0.352724	0.377880	0.305431
4175	0.0	0.654545	0.722689	0.132743	0.386931	0.356422	0.342989	0.293473
4176	1.0	0.963636	0.840336	0.172566	0.689393	0.635171	0.495063	0.491779
4177 rc	NAIC V	8 columns						

4177 rows × 8 columns

• SPLIT THE DATA INTO TRAINING ANDTESTING

SPLIT THE DATA INTO TRAINING AND TESTING

```
from sklearn.model_selection import train_test_split

[41] X_train,X_test,y_train,y_test = train_test_split(X_scaled,y,test_size=0.3,random_state = 0)

[42] X_train.shape
(2923, 8)

[43] X_test.shape
(1254, 8)

[44] y_train.shape
(2923,)

[45] y_test.shape
(1254,)
```

• BUILD THE MODEL

BUILD THE MODEL

```
/ [49] from sklearn.linear_model import LinearRegression
model = LinearRegression()
```

TRAIN THE MODEL

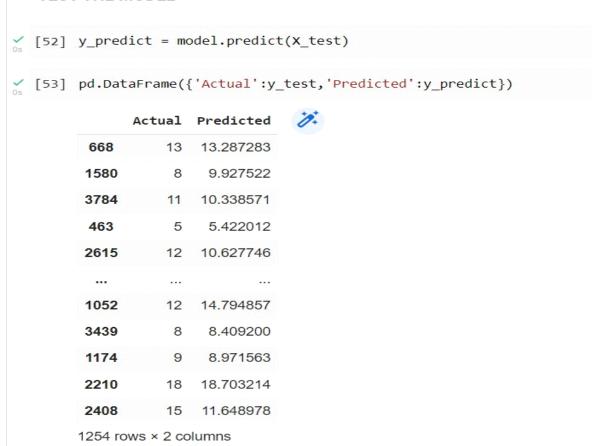
TRAIN THE MODEL

```
[51] model.fit(X_train,y_train)

LinearRegression()
```

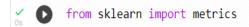
TEST THE MODEL

TEST THE MODEL



• MEASURE THE PERFORMANCE USINGMETRICS

MEASURE THE PERFORMANCE USING METRICS



vision [55] metrics.r2_score(y_test,y_predict)
vision [55] metrics

0.5202338594368163