1. Libraries

DONE BY MADHUMITHA K

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as tlp
%matplotlib inline
import seaborn as ss
```

2. Loading the dataset

```
from google.colab import files
upload=files.upload()
a=pd.read_csv('/content/abalone.csv')
a.head()
```

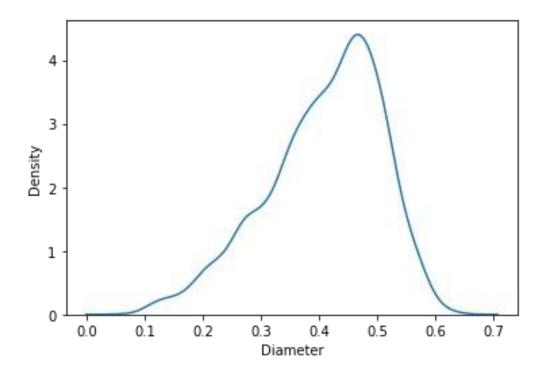
Sex	Length	Diameter	Height	Whole weight	Shucked weight	Viscera
weight	\					
0 M	0.455	0.365	0.095	0.5140	0.2245	
0.1010						
1 M	0.350	0.265	0.090	0.2255	0.0995	
0.0485						
2 F	0.530	0.420	0.135	0.6770	0.2565	
0.1415						
3 M	0.440	0.365	0.125	0.5160	0.2155	
0.1140						
4 I	0.330	0.255	0.080	0.2050	0.0895	
0.0395						

```
Shell weight Rings
0 0.150 15
1 0.070 7
2 0.210 9
3 0.155 10
4 0.055 7
```

```
a['age']=a['Rings']+1.5
a=a.drop('Rings',axis=1)
```

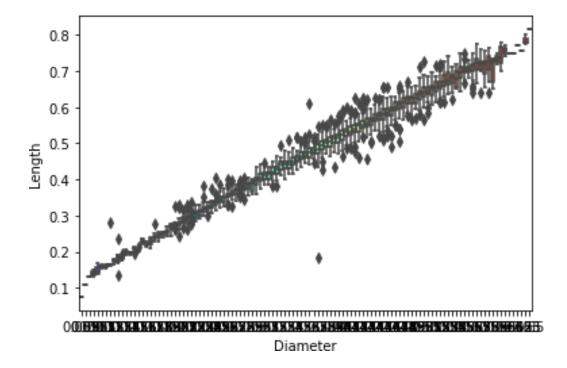
3. A. univariate Analysis

```
ss.kdeplot(a['Diameter'])
<matplotlib.axes._subplots.AxesSubplot at 0x7fbad1961d90>
```



2. Bi-Variate Analysis

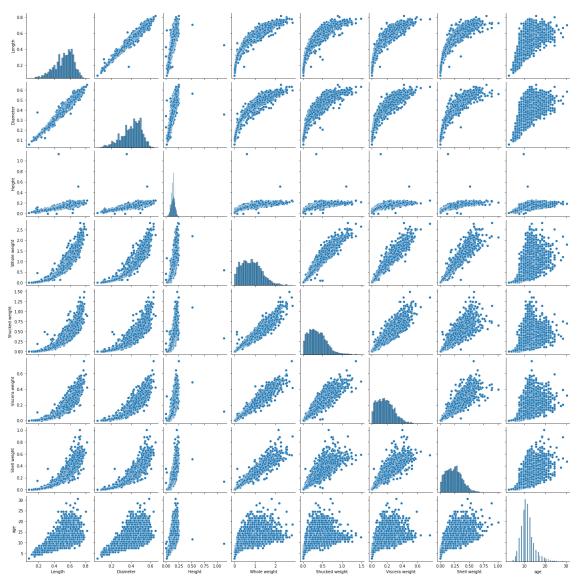
ss.boxplot(x=a.Diameter,y = a.Length, palette='rainbow')
<matplotlib.axes._subplots.AxesSubplot at 0x7fbad1844c50>



3. Multi-Variate Analysis

ss.pairplot(a)

<seaborn.axisgrid.PairGrid at 0x7fbad096df50>



4. Descriptive Statistics

a.info()

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

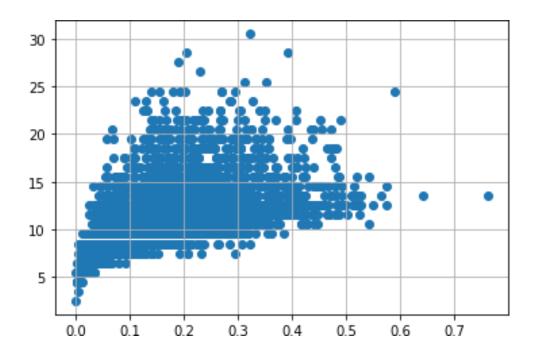
#	Column	Non-Null Count	Dtype
0	Sex	4177 non-null	object
1	Length	4177 non-null	float64
2	Diameter	4177 non-null	float64
3	Height	4177 non-null	float64

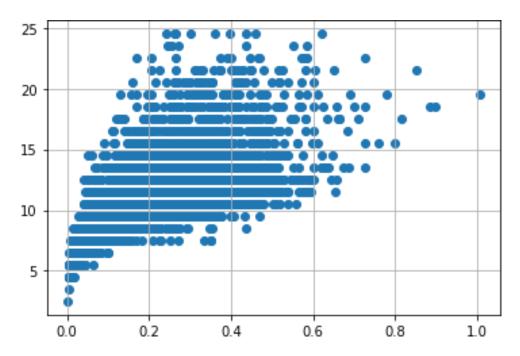
```
Whole weight
                    4177 non-null
                                    float64
 5
    Shucked weight
                    4177 non-null float64
 6
    Viscera weight
                    4177 non-null
                                   float64
 7
    Shell weight
                    4177 non-null
                                   float64
                    4177 non-null float64
 8
    age
dtypes: float64(8), object(1)
memory usage: 293.8+ KB
a['Diameter'].describe()
        4177.000000
count
           0.407881
mean
std
           0.099240
           0.055000
min
25%
           0.350000
50%
           0.425000
75%
           0.480000
           0.650000
max
Name: Diameter, dtype: float64
a['Sex'].value counts
<bound method IndexOpsMixin.value counts of 0</pre>
1
2
       F
3
       Μ
4
       Ι
       . .
4172
       F
4173
       Μ
4174
       Μ
4175
       F
4176
       Μ
Name: Sex, Length: 4177, dtype: object>
5. Checking for missing values and dealing with them
a.isnull()
       Sex Length Diameter Height Whole weight Shucked weight \
0
      False False
                       False
                              False
                                             False
                                                             False
1
      False False
                       False False
                                             False
                                                             False
2
      False False
                       False False
                                                             False
                                             False
3
      False False
                       False False
                                             False
                                                             False
4
      False False
                       False False
                                             False
                                                             False
       . . .
             . . .
                         . . .
                               . . .
                                               . . .
                                                               . . .
4172 False False
                       False False
                                             False
                                                             False
4173 False False
                       False False
                                             False
                                                            False
4174 False False
                       False False
                                             False
                                                            False
4175 False False
                       False False
                                             False
                                                            False
4176 False False
                       False False
                                             False
                                                            False
```

```
Viscera weight Shell weight age
           False False False False
0
                 False False False
False False False
False False False
False False
1
2
3
4
                False False
                                   . . .
4172
4173
4174
4175
4176
[4177 rows x 9 columns]
a.isnull().sum()
Length
                      0
                     0
Diameter
Height
                      0
Whole weight 0 Shucked weight 0
Viscera weight
                     0
Shell weight
                      0
                      0
age
Sex F
                      0
                      0
Sex I
Sex M
                      0
dtype: int64
```

6. Find the outliers and replace the outliers

```
a=pd.get_dummies(a)
dummy_a=a
var='Viscera weight'
tlp.scatter(x=a[var], y=a['age'])
tlp.grid(True)
```

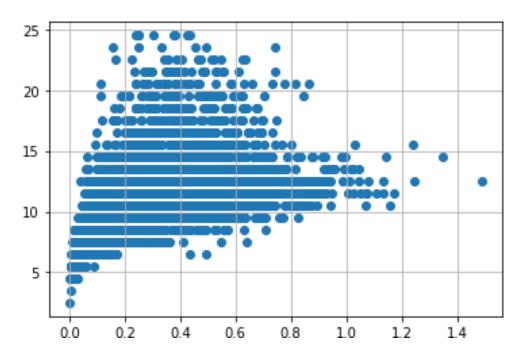


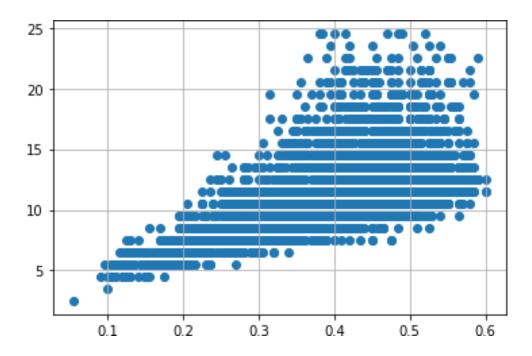


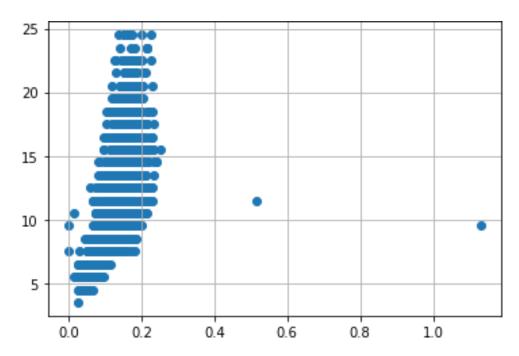
```
a.drop(a[(a['Shell weight'] > 0.6) &
          (a['age'] < 25)].index, inplace = True)</pre>
```

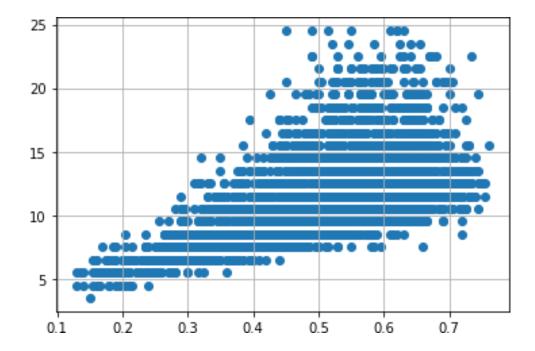
```
a.drop(a[(a['Shell weight']<0.8) & (
a['age'] > 25)].index, inplace = True)

var = 'Shucked weight'
tlp.scatter(x = a[var], y =a['age'])
tlp.grid(True)
```









7. Checking for categorical columns

```
numerical_features = a.select_dtypes(include = [np.number]).columns
categorical features = a.select dtypes(include = [np.object]).columns
```

/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:2: DeprecationWarning: `np.object` is a deprecated alias for the builtin `object`. To silence this warning, use `object` by itself. Doing this will not modify any behavior and is safe.

Deprecated in NumPy 1.20; for more details and guidance: https://numpy.org/devdocs/release/1.20.0-notes.html#deprecations

```
Index(['Length', 'Diameter', 'Height', 'Whole weight', 'Shucked
weight',
       'Viscera weight', 'Shell weight', 'age', 'Sex F', 'Sex I',
'Sex M'],
     dtype='object')
categorical features
Index([], dtype='object')
Encoding
from sklearn.preprocessing import LabelEncoder
le=LabelEncoder()
print(a.Length.value counts())
0.550
        93
0.575
        93
0.625
        93
0.580
        92
0.600
       86
        . .
0.755
0.220
        2
0.150
        1
0.135
         1
0.760
         1
Name: Length, Length: 126, dtype: int64
x=a.iloc[:,:5]
Х
      Length Diameter Height Whole weight Shucked weight
0
      0.455
             0.365 0.095
                                     0.5140
                                                     0.2245
1
      0.350
               0.265 0.090
                                     0.2255
                                                     0.0995
2
      0.530
               0.420 0.135
                                     0.6770
                                                     0.2565
             0.420 0.133
0.365 0.125
0.255 0.080
3
      0.440
                                     0.5160
                                                     0.2155
      0.330
                                     0.2050
                                                    0.0895
                         . . .
       . . .
                 . . .
4172 0.565
              0.450 0.165
                                     0.8870
                                                   0.3700
             0.440 0.135
4173 0.590
                                     0.9660
                                                    0.4390
4174 0.600
               0.475 0.205
                                                    0.5255
                                     1.1760
4175 0.625
               0.485 0.150
                                     1.0945
                                                    0.5310
4176 0.710
               0.555 0.195
                                     1.9485
                                                    0.9455
[4096 rows x 5 columns]
y=a.iloc[:,:5]
У
      Length Diameter Height Whole weight Shucked weight
                                     0.5140
0
      0.455
                0.365 0.095
                                                     0.2245
```

1	0.350	0.265	0.090	0.2255	0.0995
2	0.530	0.420	0.135	0.6770	0.2565
3	0.440	0.365	0.125	0.5160	0.2155
4	0.330	0.255	0.080	0.2050	0.0895
				• • •	
4172	0.565	0.450	0.165	0.8870	0.3700
4173	0.590	0.440	0.135	0.9660	0.4390
4174	0.600	0.475	0.205	1.1760	0.5255
4175	0.625	0.485	0.150	1.0945	0.5310
4176	0.710	0.555	0.195	1.9485	0.9455

[4096 rows x 5 columns]

9. Spliting 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)
```

10. Building the model

```
from sklearn.linear_model import LinearRegression
mlr=LinearRegression()
mlr.fit(x_train,y_train)
```

LinearRegression()

11.Training the model

12. Testingthe model

x_test[0:5]

	Length	Diameter	Height	Whole weight	Shucked weight
2358	0.610	0.485	0.210	1.3445	0.5350
723	0.525	0.410	0.130	0.9900	0.3865
2535	0.640	0.500	0.180	1.4995	0.5930
2717	0.345	0.255	0.095	0.1830	0.0750
29	0.575	0.425	0.140	0.8635	0.3930

y test[0:5]

	Length	Diameter	Height	Whole weight	Shucked weight
2358	0.610	0.485	0.210	1.3445	0.5350
723	0.525	0.410	0.130	0.9900	0.3865
2535	0.640	0.500	0.180	1.4995	0.5930
2717	0.345	0.255	0.095	0.1830	0.0750
29	0.575	0.425	0.140	0.8635	0.3930

13. Scaling the independent variables

14. Measuring the performance using metrics

```
from sklearn.metrics import r2_score
r2_score(mlr.predict(x_test),y_test)
```

1.0