Assignment – 3

Abalone Age Prediction

Assignment Date	30 September 2022		
Student Name	Paul Jabez Talakayala		
Student Roll Number	195002079		
Maximum Marks	2 Marks		

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import LabelEncoder
from sklearn.preprocessing import scale
from sklearn.linear_model import LinearRegression
from sklearn.metrics import mean_squared_error
```

- Task 1 & 2 - Downloading and Loading the Dataset

```
/ [3] path = '/content/drive/MyDrive/IBM/Assignment 3/abalone.csv'
       df = pd.read_csv(path)
[4] 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
                                                                                                        15
                 0.350
                                   0.090
                                                0.2255
                                                                0.0995
                                                                                0.0485
                                                                                               0.070
                                                                                                         7
            M
                           0.265
                 0.530
                           0.420
                                   0.135
                                                0.6770
                                                                0.2565
                                                                                0.1415
                                                                                               0.210
                                                                                                         9
                 0.440
                                   0.125
                                                0.5160
                                                                0.2155
                                                                                0.1140
                                                                                               0.155
                                                                                                        10
                           0.365
                 0.330
                           0.255
                                   0.080
                                                0.2050
                                                                0.0895
                                                                                0.0395
                                                                                               0.055
                                                                                                         7
[5] df.shape
       (4177, 9)
```

```
[6] df.info()
     <class 'pandas.core.frame.DataFrame'>
     RangeIndex: 4177 entries, 0 to 4176
     Data columns (total 9 columns):
         Column
                        Non-Null Count Dtype
     --- -----
                                        ----
      0
                         4177 non-null
                                         object
          Sex
                         4177 non-null float64
      1
         Length
                         4177 non-null float64
      2
         Diameter
      3
         Height
                         4177 non-null
                                       float64
                                       float64
         Whole weight
                         4177 non-null
          Shucked weight 4177 non-null
                                       float64
          Viscera weight 4177 non-null
                                         float64
      7
          Shell weight
                         4177 non-null
                                        float64
                         4177 non-null
                                         int64
          Rings
     dtypes: float64(7), int64(1), object(1)
```

Task - 3 Visualizing the Analysis

memory usage: 293.8+ KB

3.1 Univariate Analysis

0.1

0.3

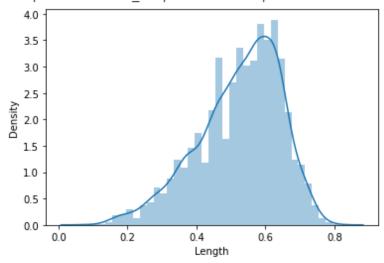
0.4

0.5

```
[7] plt.hist(df['Diameter'])
       (array([ 13., 66., 180., 344., 513., 812., 1017., 934., 275.,
                 23.]),
        array([0.055, 0.1145, 0.174, 0.2335, 0.293, 0.3525, 0.412, 0.4715,
               0.531 , 0.5905, 0.65 ]),
        <a list of 10 Patch objects>)
        1000
         800
         600
         400
         200
           0
                       0.2
```

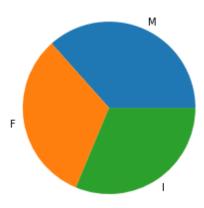
/usr/local/lib/python3.7/dist-packages/seaborn/distributions.py:2619: warnings.warn(msg, FutureWarning)

<matplotlib.axes._subplots.AxesSubplot at 0x7fa744495750>



4 │

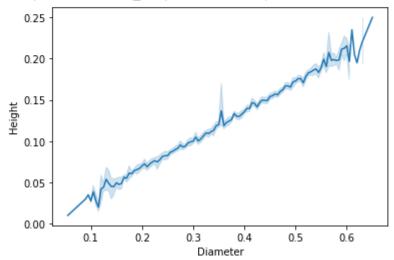
[9] plt.pie(df['Sex'].value_counts(), labels=df['Sex'].unique())
 plt.show()



▼ 3.2 Bi - Variate Analysis

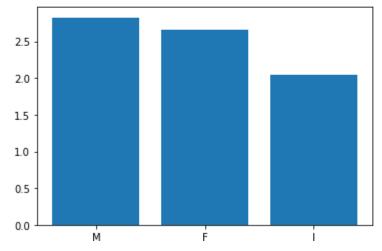
[10] sns.lineplot(x=df.Diameter,y=df.Height)

<matplotlib.axes._subplots.AxesSubplot at 0x7fa7443800d0>

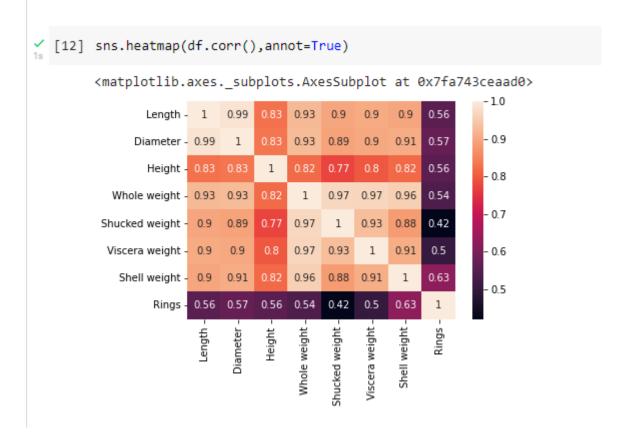


[11] plt.bar(df.Sex, df['Whole weight'])

<BarContainer object of 4177 artists>



▼ 3.3 Multi - Variate Analysis



- Task 4 - Descriptive Statistics

	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

Task 5 - Handle Missing Values

```
/ [14] df.isnull().any()
                       False
       Sex
       Length
                       False
                      False
       Diameter
       Height
                      False
       Whole weight
                     False
       Shucked weight False
       Viscera weight False
       Shell weight
                       False
       Rings
                       False
       dtype: bool
()
()
()
()
()
()
()
       Sex
       Length
                       0
       Diameter
       Height
       Whole weight
       Shucked weight 0
       Viscera weight 0
       Shell weight
                       0
       Rings
       dtype: int64
```

Task 6 - Find Outliers and Replacing

10

Rings

```
(16] sns.boxplot(x=df['Rings'])

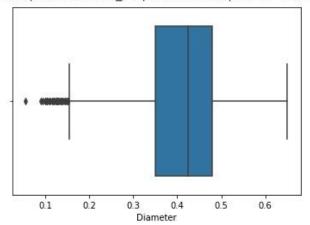
<matplotlib.axes._subplots.AxesSubplot at 0x7fa73e668a10>
```

20

30

```
[17] sns.boxplot(x=df['Diameter'])
```

<matplotlib.axes. subplots.AxesSubplot at 0x7fa73e668a50>



```
[18] for col in df.columns:
    if (((df[col].dtype)=='float64') | ((df[col].dtype)=='int64')):

        q2 = df[col].quantile(0.25)
        q3 = df[col].quantile(0.75)
        IQR = q3 - q2
        upper_limit = q3 + 1.5*IQR
        lower_limit = q2 - 1.5*IQR

        df[col] = np.where(df[col] < lower_limit, lower_limit, df[col])
        df[col] = np.where(df[col] > upper_limit, upper_limit, df[col])
```

- Task - 7 Categorical Columns and Perform Encoding

```
/ [19] 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
                                                                                                         15.0
                                                                                 0.0485
                                                                                                 0.070
                                                                                                          7.0
             M
                 0.350
                            0.265
                                    0.090
                                                 0.2255
                                                                 0.0995
                  0.530
                                    0.135
                                                 0.6770
                                                                 0.2565
                                                                                  0.1415
                                                                                                 0.210
                                                                                                          9.0
                            0.420
                                                                                                         10.0
                  0.440
                            0.365
                                    0.125
                                                 0.5160
                                                                 0.2155
                                                                                  0.1140
                                                                                                 0.155
                 0.330
                            0.255
                                    0.080
                                                 0.2050
                                                                 0.0895
                                                                                 0.0395
                                                                                                 0.055
                                                                                                          7.0
[20] print(df.Sex.unique())
```

```
['M' 'F' 'I']
```

```
[21] le = LabelEncoder()
        df.Sex = le.fit_transform(df.Sex)
/ [22] 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
                                                                                                           15.0
                  0.350
                            0.265
                                     0.090
                                                  0.2255
                                                                   0.0995
                                                                                   0.0485
                                                                                                   0.070
                                                                                                            7.0
                  0.530
                             0.420
                                     0.135
                                                  0.6770
                                                                   0.2565
                                                                                    0.1415
                                                                                                   0.210
                                                                                                            9.0
              2
                            0.365
                                     0.125
                                                  0.5160
                                                                   0.2155
                                                                                    0.1140
                                                                                                            10.0
                  0.440
                                                                                                   0.155
                  0.330
                            0.255
                                     0.080
                                                  0.2050
                                                                   0.0895
                                                                                    0.0395
                                                                                                   0.055
```

- Task - 8 Split the Data into Dependent and Independent Variables

```
 (23) y = df['Rings']
         X = df.drop('Rings', axis=1)

√ [24] y

         0
                 15.0
         1
                  7.0
         2
         3
                 10.0
         4
                  7.0
        4172
                 11.0
         4173
                 10.0
         4174
        4175
                 10.0
         4176
                 12.0
        Name: Rings, Length: 4177, dtype: float64

✓ [25] X

               Sex Length Diameter Height Whole weight Shucked weight Viscera weight Shell weight
          0
                     0.455
                                0.365
                                        0.095
                                                      0.5140
                                                                      0.2245
                                                                                       0.1010
                                                                                                     0.1500
                 2
                                                                                                     0.0700
          1
                     0.350
                                0.265
                                        0.090
                                                     0.2255
                                                                      0.0995
                                                                                       0.0485
          2
                 0
                     0.530
                                0.420
                                        0.135
                                                                      0.2565
                                                                                       0.1415
                                                                                                     0.2100
                                                     0.6770
          3
                     0.440
                                0.365
                                        0.125
                                                      0.5160
                                                                      0.2155
                                                                                       0.1140
                                                                                                     0.1550
          4
                     0.330
                                0.255
                                        0.080
                                                      0.2050
                                                                      0.0895
                                                                                       0.0395
                                                                                                     0.0550
         4172
                 0
                     0.565
                                0.450
                                        0.165
                                                     0.8870
                                                                      0.3700
                                                                                       0.2390
                                                                                                     0.2490
```

4177 rows x 8 columns

0

0.590

0.600

0.625

0.710

0.440

0.475

0.485

0.555

0.135

0.205

0.150

0.195

0.9660

1.1760

1.0945

1.9485

0.4390

0.5255

0.5310

0.9455

0.2145

0.2875

0.2610

0.3765

0.2605

0.3080

0.2960

0.4950

4173

4174

4175

4176

- Task - 9 Scale the Independent Variables

	Sex	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight
0	1.151980	-0.583117	-0.440884	-1.158093	-0.644740	-0.614985	-0.730304	-0.645184
1	1.151980	-1.465694	-1.459762	-1.288751	-1.238208	-1. <mark>191637</mark>	-1.213890	-1.231390
2	-1.280690	0.047295	0.119499	-0.112828	-0.309436	-0.467362	-0.357253	-0.205531
3	1.151980	-0.709200	-0.440884	-0.374145	-0.640626	-0.656504	-0.610559	-0.608547
4	-0.064355	-1.633804	-1.561650	-1.550067	-1.280378	-1.237770	-1.296790	-1.341303
	925	2003	1020	1000	222	26.2	222	0.00
4172	-1.280690	0.341487	0.425163	0.671120	0.122550	0.056238	0.540835	0.080244
4173	1.151980	0.551624	0.323275	-0.112828	0.285059	0.374550	0.315162	0.164511
4174	1.151980	0.635679	0.679882	1.716385	0.717046	0.773593	0.987576	0.512571
4175	-1.280690	0.845817	0.781770	0.279146	0.549394	0.798966	0.743480	0.424640
4176	1.151980	1.560284	1.494985	1.455069	2.306139	2.711144	1.807369	1.882825

4177 rows × 8 columns

- Task - 10 Split the Data into Training and Testing

```
    [27] x_train, x_test, y_train, y_test = train_test_split(scaled_X, y, test_size=0.3, random_state=0)

    [28] print(x_train.shape, x_test.shape, y_train.shape, y_test.shape)
    (2923, 8) (1254, 8) (2923,) (1254,)
```

- Task - 11 Build the Model

```
v [29] lr = LinearRegression()
```

→ Task - 12 Train the Model

```
[30] lr.fit(x_train, y_train)
LinearRegression()
```

→ Task - 13 Test the Model

```
y_pred = lr.predict(x_test)
y_pred

array([12.55446979, 9.47758181, 10.25463091, ..., 7.9886076,
17.03108659, 11.51225618])
```