Assignment-3

Python Programming

Assignment Date	7 October 2022			
Student Name	M.Shehha			
Student Roll Number	812419106042			
Maximum Marks	2 Marks			

```
In [121]: import os
          import pandas as pd
           import numpy as np
           import matplotlib.pyplot as plt
          import seaborn as sns
          from sklearn.model_selection import train_test_split
          from sklearn.preprocessing import LabelEncoder
          from sklearn.linear_model import LinearRegression
          from sklearn import metrics
           %matplotlib inline
In [122]: os.getcwd()
Out[122]: 'C:\\Users\\pc'
In [123]: path='C:\\Users\\pc\\downloads\\'
          data=pd.read_csv(path+'abalone.csv')
          data
Out[123]:
                 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
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                      0.350
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4973	М	0.550	0.440	0.135	0.9550	0.4300	0.2145	0.2605	10
4174	М	0.800	0.475	0.205	1,1760	0.5255	0.2875	0.3080	9
4175	Γ	0.625	0.485	0.158	1.0945	0.5010	0.2010	0.2960	10
4176	И	0.710	0.555	0.195	1,9485	0.9455	0.3765	0.4950	12

4177 mes × 9 columns

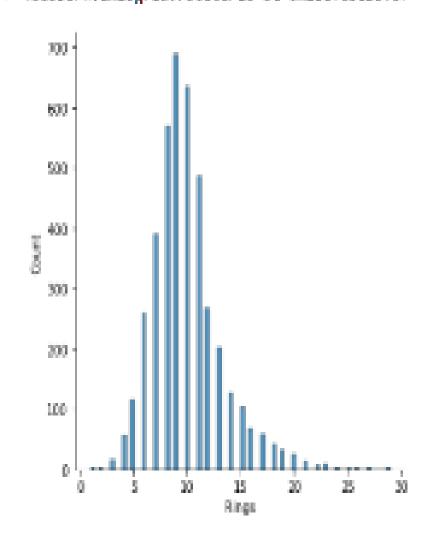
```
In [124]: data.shape
```

Out[124]: (4177, 4)

In [125]: ana.displot(data['Rings'])

 $0 = \{129\} \leftarrow \texttt{cseaborn.exisgrid.FacetGrid at 0x2bet651b370} \}$

Out[125]) (seeborn.axisgrid.FacetOrid at 0x2bef651b370)



In [128]: ana.berplot(x='Sex',y='Rings',dete=dete)

Out[126]: chxesSubplot:xlabel='Sev', ylabel='Rings'>



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Ovt[125]: 
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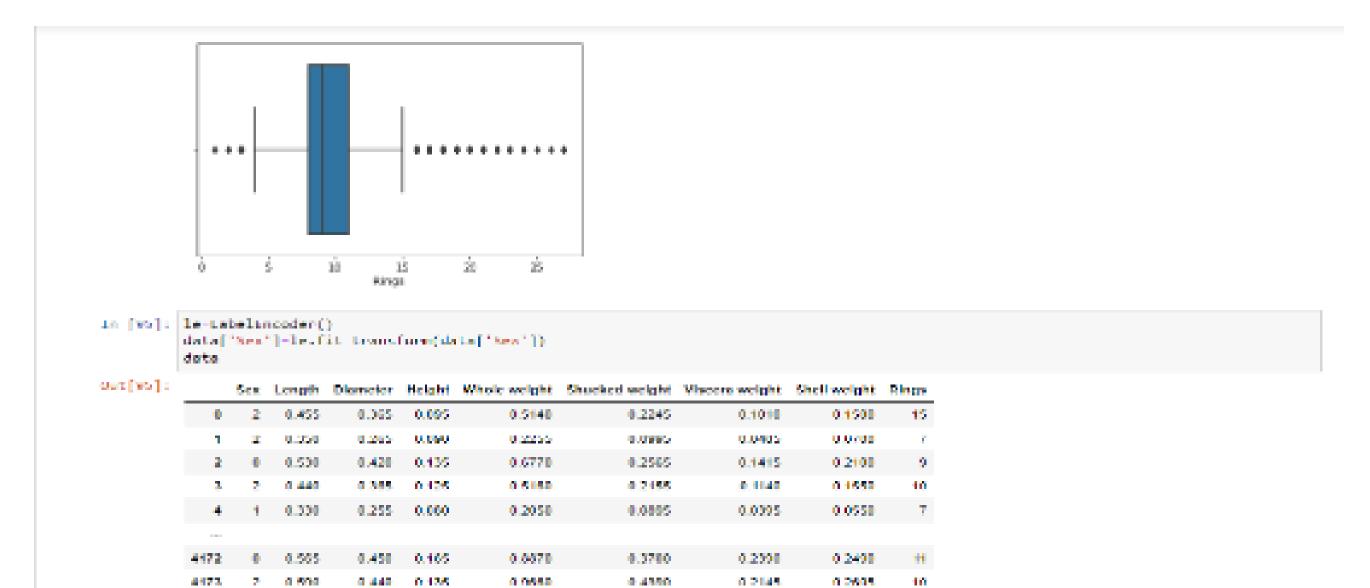
In [125]: 
impurt sarrings
warnings.simplefilter(action-'ignore', category-futurekerning)# to avoid warning
```

```
In [129]: impact secology.
          warnings.simplefilter(action='ignore', category=futureWarning)# to avoid warning
 in [55]: | snc.pairplot(data- data [['bex', 'Length', 'Diameter', 'Height', 'Whole Weight', 'Kings']], hue-'Kings')
          c:\ucerc\pc\anacondas\lib\cite-packagec\ceaborn\distributionc.py:set: ucermanning: uataset has a variance; skipping density a
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          otimate.
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          whiteaches.
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            warmings.warm(msg, UserWarming)
          Caldisers \palanamandat\lib\sitespackages\sealoon\distributions.pga 98%. Excellandings. Balaset, has 8-variance; skipping density e. ....
In [130]: data.isnull().sum()
Out[130]: Sex
          Longth
                            0
          Oismeter
                             9
          Height
                             -
```

```
Out[130]: Sex
           Length
           Diameter
           Height
                               0
           Whole weight
                               Shucked weight
           Viscera weight
           Shell weight
           Rings
           dtype: int64
In [151]: data.describe()
Out[131]:
                                               Height Whole weight Shucked weight Viscers weight Shell weight
                       Length
                                 Diameter
                                                                                                                Rings
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            count 4177.000000 4177.000000 4177.000000
                                                                     4177.000000
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                                             1.130000
                                                          2.025500
                                                                        1.488000
                                                                                      0.760000
                                                                                                            29,000000
             1133
In [132]: data.skew()
                               0.639873
Out[132]: Length
                              -0.689198
           Diameter
           Height
                               3.128817
           Diameter
                              0.600198
           Height
                              3,128817
                              0.530959
           Whole weight:
           Shocked weight
                              0.719098
                              0.591852
           Viscera weight
           Shell weight
                              0.629927
           Rings
                              1.114182
           dtype: float84
In [128]: |ans.boxplot(date['Rings'])
           TypeError
                                                        Traceback (most recent call last)
           <ipython-input-120-0fiddebe65al> in <module>
           ----> 1 sns.boxplot(data['Kings'])
           TypeError: list indices must be integers or slices, not str.
In [134]: q1-data['Rings'] . describe()['258']
           q3=data['Rings'] . describe()['75%']
Out[134]: 8.8
           Type Markdown and LaTeX: \alpha^2
 In [84]: qd
```

Out[84]: 11.0

```
In [85]: [qr=q3-q1
         1qr
Out[85]: 5.8
In [86]: a=q1-(1.5*q1)
         b=q3+(1.5*q3)
         print(a, b)
         -4.0 27.5
In [87]: data[data['Rings']<a]</pre>
Out[87]:
           Sex Length Diameter Height Whole weight Shucked weight Viscors weight Shell weight Rings
In [88]: data[data['Rings']>b].head()
Out[88]:
              Sex Length Diameter Height Whole weight Shucked weight Viscera weight Shell weight Rings.
                                                                        0.3215
          480 F 0.7 0.585 0.105
                                              1.0075
                                                            0.7055
                                                                                    0.475
In [89]: outlier_list-list(data[data['Rings']>b]['Rings'])
         print(outlier_list)
         [29]
In [90]: data['Rings']=data['Rings'].replace(outlier_list)
         ans.boxplot(data['Ringa'])
Out[90]: <AxesSubplot:xlabel='Rings'>
```



```
0.475 0.205
                                               1.1760
                                                             0.5255
                                                                          0.2875
           4174 2 0.000
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                              0.405 0.150
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                      0.625
                                                1.0945
                                                             0.5318
                                                                          6.3610
                                                                                              19
           4176 2 0.710
                              0.555 - 0.195
                                                1.9485
                                                             0.9455
                                                                          0.3765
                                                                                     0.4950
                                                                                              12
          4177 rows × 9 columns
 In [98] : X = data.iloc[r, r-1].values
          y - data.iloc|: -1| values
In [97]: train_X,val_X,train_y,val_y = train_test_split(X, y, test_size = 8.2, random_state = 8)
          print("Shape of training & t", train_&.shape)
          print("Shape of Validation X : ", val X.shape)
          Shape of Training X : (3341, 8)
          Shape of Validation X : (836, 8)
In [98]: Ir = LinearRegression()
          Ir.fit(train_X, train_y)
          print('Attempting to fit timear Regressor')
          Attempting to fit Linear Regressor
In [105]: %%time:
          y_pred_val_le = le.predict(val_x)
          print('MAE on Validation set : ',metrics.mean absolute error(val y, y pred val lr))
          print('MSE on Validation set :',metrics.mean_squared_error(val_y, y_pred_val_lr));
          print('RMSE on Validation set s'.mp.sqrt(metrics.mean_absolute_error(val_y, y_pred_val_lr)));
          peint("\a").
          print('R) Score on Validation set : '.metrics.r2 score(val v. v pred val lr))
          |print('ww score on Validation set :',metrics.ru_score(val_y, y_pred_val_ir))|
           print("\n")
           MAC on Validation set : 1.5700045000902012
           MSE on Validation set : 4.7449599677459635
           RMAR on Validation set : 1.3564971464638662
           R2 house on Validation set : 0.5466386689286487
           tall time: 6 mc
```