sptone-project-aboh-michael-27a-2

May 14, 2024

0.1 Capstone project for Basic Data Science

In this project, you would explore the Data preprocessing techniques learned. You will use, pandas, plots and the different data preprocessing techniques to explore and analyse the data. This project is on Exploratory Data Analysis

Step1: Import the necessary libraries

Dataset 1: exercise

```
[7]: import matplotlib.pyplot as plt
import numpy as np
import pandas as pd
import seaborn as sns
%matplotlib inline
```

```
[12]: #List the set of available datasets in seaborn
print(len(sns.get_dataset_names()))
print(sns.get_dataset_names())
```

['anagrams', 'anscombe', 'attention', 'brain_networks', 'car_crashes',
'diamonds', 'dots', 'dowjones', 'exercise', 'flights', 'fmri', 'geyser', 'glue',
'healthexp', 'iris', 'mpg', 'penguins', 'planets', 'seaice', 'taxis', 'tips',
'titanic', 'anagrams', 'anagrams', 'anscombe', 'anscombe', 'attention',
'attention', 'brain_networks', 'brain_networks', 'car_crashes', 'car_crashes',
'diamonds', 'diamonds', 'dots', 'dots', 'dowjones', 'dowjones', 'exercise',
'exercise', 'flights', 'flights', 'fmri', 'fmri', 'geyser', 'geyser', 'glue',
'glue', 'healthexp', 'healthexp', 'iris', 'iris', 'mpg', 'mpg', 'penguins',
'penguins', 'planets', 'planets', 'seaice', 'seaice', 'taxis', 'taxis', 'tips',
'tips', 'titanic', 'titanic', 'anagrams', 'anscombe', 'attention',
'brain_networks', 'car_crashes', 'diamonds', 'dots', 'dowjones', 'exercise',
'flights', 'fmri', 'geyser', 'glue', 'healthexp', 'iris', 'mpg', 'penguins',
'planets', 'seaice', 'taxis', 'tips', 'titanic']

```
[10]: #1.Write the syntax for choosing the specific dataset df=sns.load_dataset('penguins') df
```

```
[10]:
          species
                      island bill_length_mm bill_depth_mm flipper_length_mm \
           Adelie Torgersen
                                         39.1
                                                                            181.0
      0
                                                         18.7
                                         39.5
                                                         17.4
                                                                            186.0
      1
           Adelie Torgersen
      2
           Adelie
                   Torgersen
                                         40.3
                                                         18.0
                                                                            195.0
      3
           Adelie
                   Torgersen
                                          NaN
                                                          NaN
                                                                             NaN
      4
           Adelie
                   Torgersen
                                         36.7
                                                         19.3
                                                                            193.0
      . .
              •••
                       •••
      339
           Gentoo
                      Biscoe
                                          NaN
                                                          NaN
                                                                             NaN
      340 Gentoo
                      Biscoe
                                         46.8
                                                         14.3
                                                                           215.0
                                         50.4
                                                         15.7
                                                                            222.0
      341
           Gentoo
                      Biscoe
                                         45.2
      342 Gentoo
                      Biscoe
                                                         14.8
                                                                            212.0
      343 Gentoo
                      Biscoe
                                         49.9
                                                         16.1
                                                                            213.0
           body_mass_g
                            sex
      0
                3750.0
                           Male
      1
                3800.0 Female
      2
                3250.0
                        Female
      3
                   NaN
                            NaN
      4
                3450.0 Female
      339
                   {\tt NaN}
                            NaN
      340
                4850.0
                        Female
      341
                           Male
                5750.0
      342
                5200.0 Female
      343
                5400.0
                          Male
      [344 rows x 7 columns]
[13]: #2.Display the top 5 records of penguin dataset
      #3.Display the dimensionality of the dataset
      print(df.head())
      print(df.shape)
       species
                    island bill_length_mm bill_depth_mm
                                                            flipper_length_mm \
     O Adelie Torgersen
                                      39.1
                                                      18.7
                                                                         181.0
     1 Adelie
                Torgersen
                                      39.5
                                                      17.4
                                                                         186.0
     2 Adelie Torgersen
                                      40.3
                                                      18.0
                                                                         195.0
     3 Adelie Torgersen
                                       {\tt NaN}
                                                       {\tt NaN}
                                                                           NaN
     4 Adelie Torgersen
                                      36.7
                                                      19.3
                                                                         193.0
        body_mass_g
                         sex
                        Male
     0
              3750.0
     1
             3800.0 Female
     2
             3250.0 Female
     3
                         NaN
                 NaN
             3450.0 Female
     (344, 7)
```

[14]: #4.Display the datatypes of the attributes df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 344 entries, 0 to 343
Data columns (total 7 columns):

#	Column	Non-Null Count	Dtype
0	species	344 non-null	object
1	island	344 non-null	object
2	bill_length_mm	342 non-null	float64
3	bill_depth_mm	342 non-null	float64
4	flipper_length_mm	342 non-null	float64
5	body_mass_g	342 non-null	float64
6	sex	333 non-null	object

dtypes: float64(4), object(3)

memory usage: 18.9+ KB

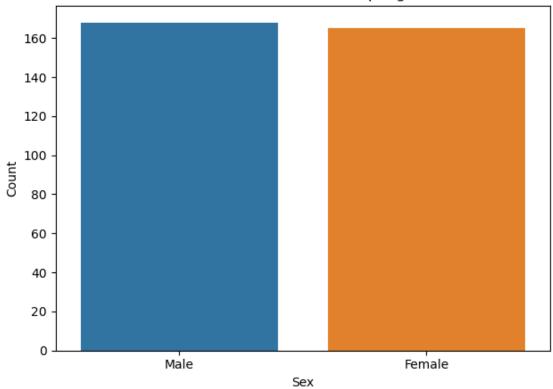
[15]: #5.Describe the dataset extensively df.describe()

```
[15]:
             bill_length_mm bill_depth_mm flipper_length_mm body_mass_g
                 342.000000
                                342.000000
                                                   342.000000
                                                                342.000000
      count
     mean
                 43.921930
                                 17.151170
                                                   200.915205 4201.754386
                   5.459584
                                  1.974793
                                                                801.954536
      std
                                                    14.061714
     min
                  32.100000
                                 13.100000
                                                   172.000000 2700.000000
     25%
                  39.225000
                                 15.600000
                                                   190.000000
                                                               3550.000000
      50%
                  44.450000
                                 17.300000
                                                   197.000000 4050.000000
      75%
                  48.500000
                                 18.700000
                                                   213.000000 4750.000000
                  59.600000
                                 21.500000
                                                   231.000000 6300.000000
     max
```

- [16]: #6.Display the number of null values in the dataset and the total count of it df.isnull().sum()
- [18]: #7. Identify the columns having null values and remove them penguins_cleaned=df.dropna()

```
[19]: # 8.check the size of dataset after cleaning and compare with the size before.
       \hookrightarrow cleaning
      print('Before cleaning:', df.shape)
      print('After cleaning:',penguins_cleaned.shape)
     Before cleaning: (344, 7)
     After cleaning: (333, 7)
[20]: #9. Find out the count of male and female penguins
      penguins_cleaned['sex'].value_counts()
[20]: sex
     Male
                168
      Female
                165
      Name: count, dtype: int64
[21]: #10.Find out the count of species
      penguins_cleaned['species'].value_counts()
[21]: species
      Adelie
                   146
      Gentoo
                   119
      Chinstrap
                    68
      Name: count, dtype: int64
[22]: #11. Use a countplot graph to display the number of male and female penguins
      #Give the plot a title "Count of male and female penguins"
      #Give a xlabel and ylabel
      sns.countplot( x="sex",data=penguins_cleaned)
      plt.title("Count of male and female penguins")
      plt.xlabel("Sex")
      plt.ylabel("Count")
      plt.tight_layout()
      plt.show()
```

Count of male and female penguins



```
[23]: #12. For all the penguins display the bill_length_mm and bill_depth_mm using_
       \hookrightarrow pairplot
      sns.pairplot(df[['bill_length_mm','bill_depth_mm']])
```

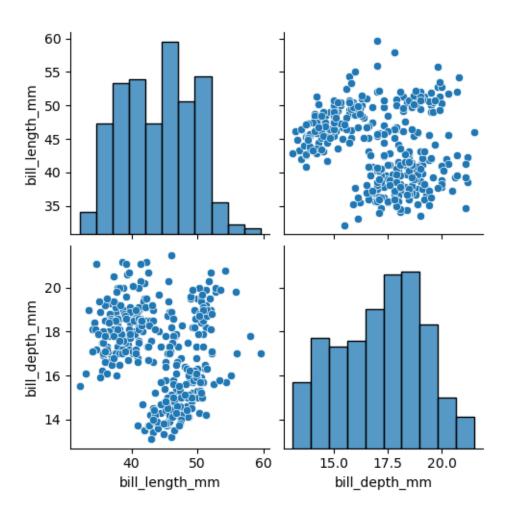
C:\Users\PC\anaconda3\Lib\site-packages\seaborn_oldcore.py:1119: FutureWarning: use_inf_as_na option is deprecated and will be removed in a future version. Convert inf values to NaN before operating instead.

with pd.option_context('mode.use_inf_as_na', True):

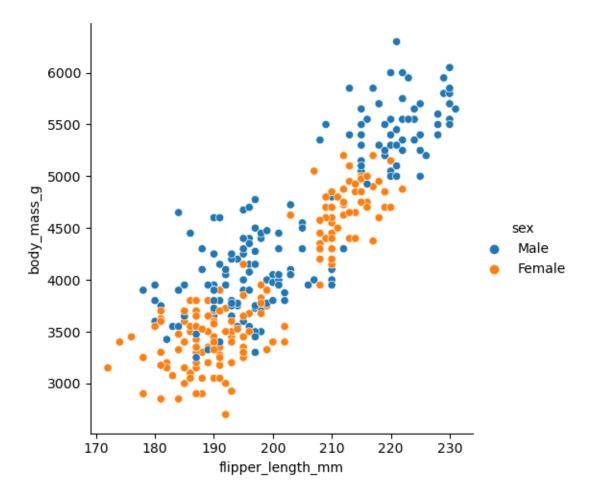
C:\Users\PC\anaconda3\Lib\site-packages\seaborn_oldcore.py:1119: FutureWarning: use_inf_as_na option is deprecated and will be removed in a future version. Convert inf values to NaN before operating instead.

with pd.option_context('mode.use_inf_as_na', True):

[23]: <seaborn.axisgrid.PairGrid at 0x226b68b16d0>



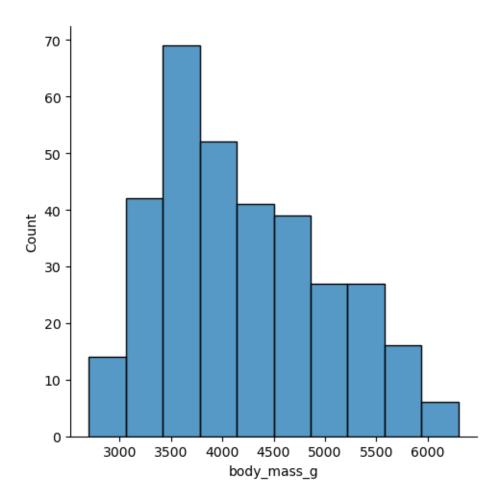
[24]: <seaborn.axisgrid.FacetGrid at 0x226b7fedf10>



```
[25]: #14. Create a histogram with the flipper_length_mm and bins = 10 sns.displot(penguins_cleaned["body_mass_g"], kde= False, bins = 10)
```

C:\Users\PC\anaconda3\Lib\site-packages\seaborn_oldcore.py:1119: FutureWarning:
use_inf_as_na option is deprecated and will be removed in a future version.
Convert inf values to NaN before operating instead.
 with pd.option_context('mode.use_inf_as_na', True):

[25]: <seaborn.axisgrid.FacetGrid at 0x226b76a82d0>

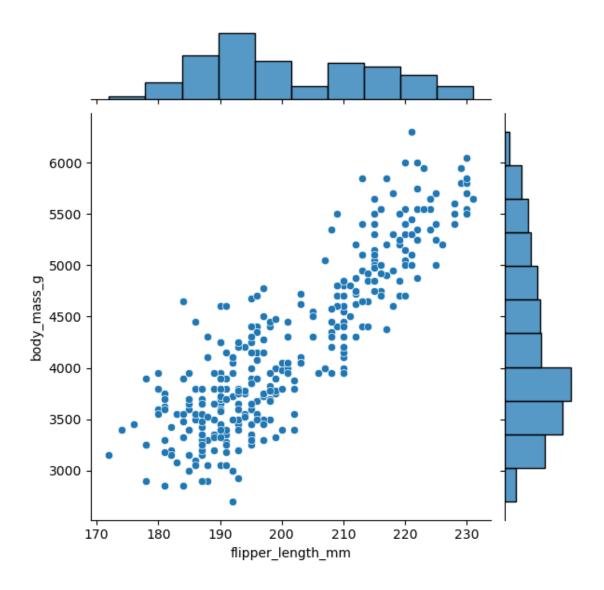


```
[26]: #15. Create a joint point with body_length_mm, body_depth_mm for → penguins_cleaned dataset

sns.jointplot(x= 'flipper_length_mm' , y= 'body_mass_g' , data = → penguins_cleaned)
```

C:\Users\PC\anaconda3\Lib\site-packages\seaborn_oldcore.py:1119: FutureWarning:
use_inf_as_na option is deprecated and will be removed in a future version.
Convert inf values to NaN before operating instead.
 with pd.option_context('mode.use_inf_as_na', True):
C:\Users\PC\anaconda3\Lib\site-packages\seaborn_oldcore.py:1119: FutureWarning:
use_inf_as_na option is deprecated and will be removed in a future version.
Convert inf values to NaN before operating instead.
 with pd.option_context('mode.use_inf_as_na', True):

[26]: <seaborn.axisgrid.JointGrid at 0x226b76af810>



[27]: #16. Use a pairwise plot to display all the numerical values of the dataset

→penguins_cleaned

sns.pairplot(penguins_cleaned)

C:\Users\PC\anaconda3\Lib\site-packages\seaborn_oldcore.py:1119: FutureWarning: use_inf_as_na option is deprecated and will be removed in a future version. Convert inf values to NaN before operating instead.

with pd.option_context('mode.use_inf_as_na', True):

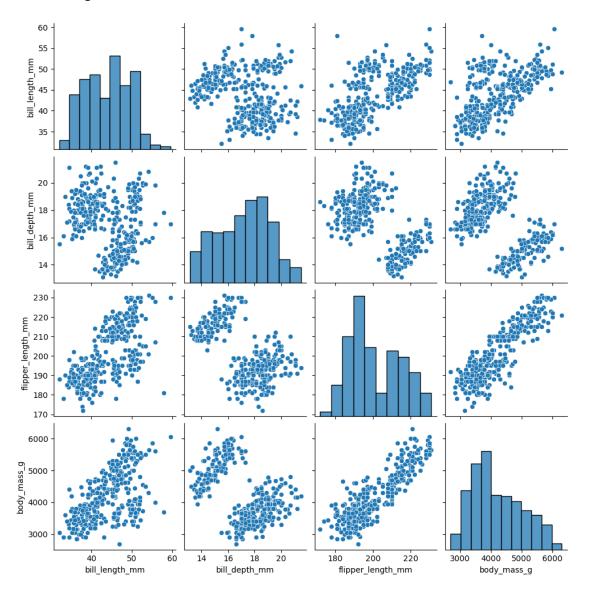
C:\Users\PC\anaconda3\Lib\site-packages\seaborn_oldcore.py:1119: FutureWarning: use_inf_as_na option is deprecated and will be removed in a future version. Convert inf values to NaN before operating instead.

with pd.option_context('mode.use_inf_as_na', True):

C:\Users\PC\anaconda3\Lib\site-packages\seaborn_oldcore.py:1119: FutureWarning: use_inf_as_na option is deprecated and will be removed in a future version.

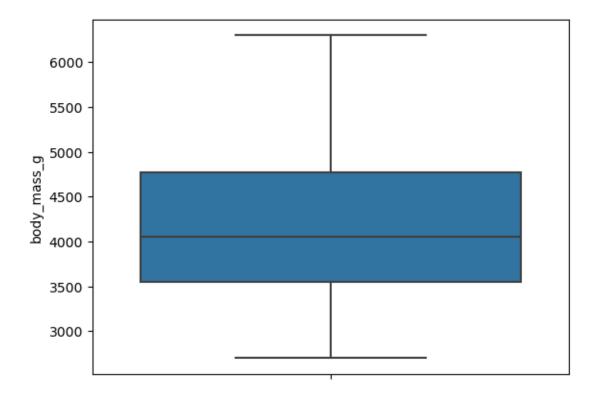
Convert inf values to NaN before operating instead.
 with pd.option_context('mode.use_inf_as_na', True):
C:\Users\PC\anaconda3\Lib\site-packages\seaborn_oldcore.py:1119: FutureWarning:
use_inf_as_na option is deprecated and will be removed in a future version.
Convert inf values to NaN before operating instead.
 with pd.option_context('mode.use_inf_as_na', True):

[27]: <seaborn.axisgrid.PairGrid at 0x226b80e2e50>

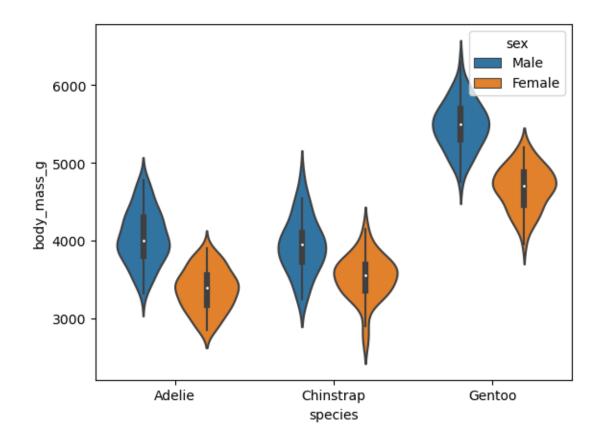


[28]: #17.Check if there exists any outliers based on the boxplot based on numerical values of penguins_cleaned dataset sns.boxplot(y='body_mass_g', data= penguins_cleaned)

```
[28]: <Axes: ylabel='body_mass_g'>
```



[30]: <Axes: xlabel='species', ylabel='body_mass_g'>



[]: