

21BCI0075 GATIK GARG

September 13, 2023

#####1. penguins_size.csv is downloaded

#####2. Load the dataset into the tool.

```
[1]: import numpy as np
import pandas as pd
```

```
[2]: df = pd.read_csv('/content/penguins_size.csv')
df.head()
```

```
[2]:  species      island  culmen_length_mm  culmen_depth_mm  flipper_length_mm  \
0  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          NaN           NaN              NaN
4  Adelie  Torgersen         36.7           19.3             193.0

      body_mass_g      sex
0         3750.0    MALE
1         3800.0  FEMALE
2         3250.0  FEMALE
3            NaN      NaN
4         3450.0  FEMALE
```

#####3.1. Perform Univariate Analysis

```
[3]: from matplotlib import rcParams
import seaborn as sns
```

```
[4]: sns.distplot(df.body_mass_g)
```

<ipython-input-4-176964dae727>:1: UserWarning:

`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

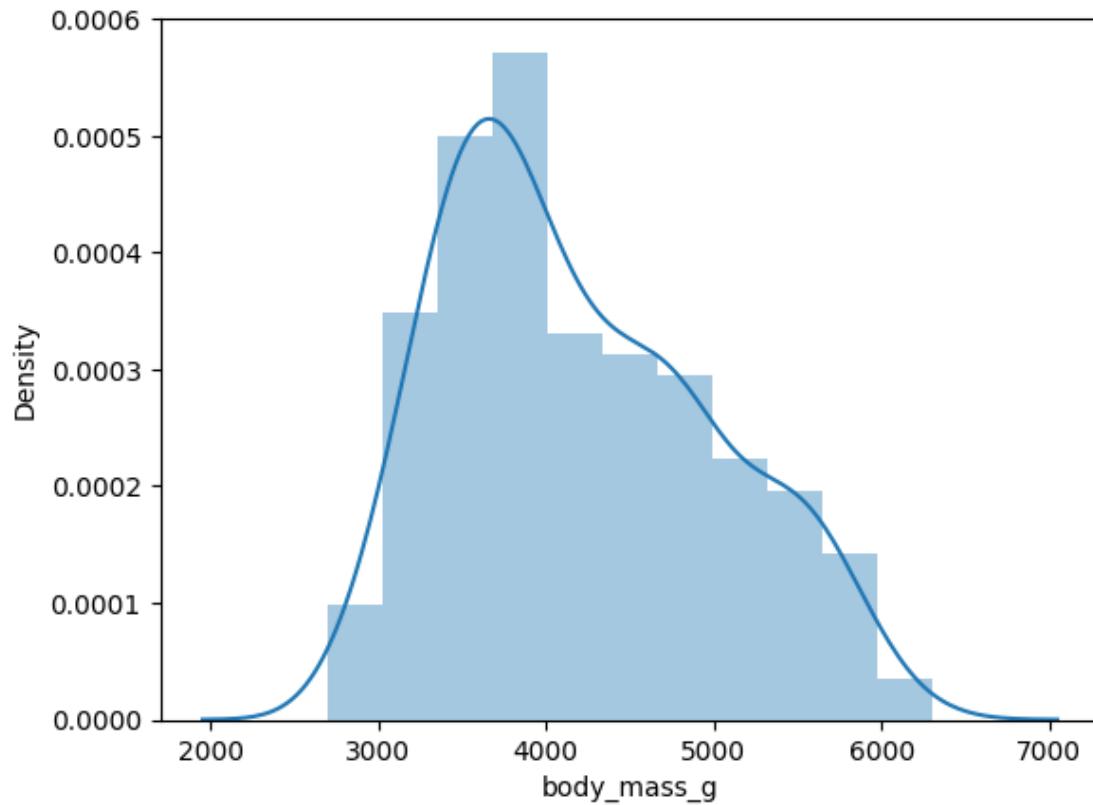
Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see

<https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751>

```
sns.distplot(df.body_mass_g)
```

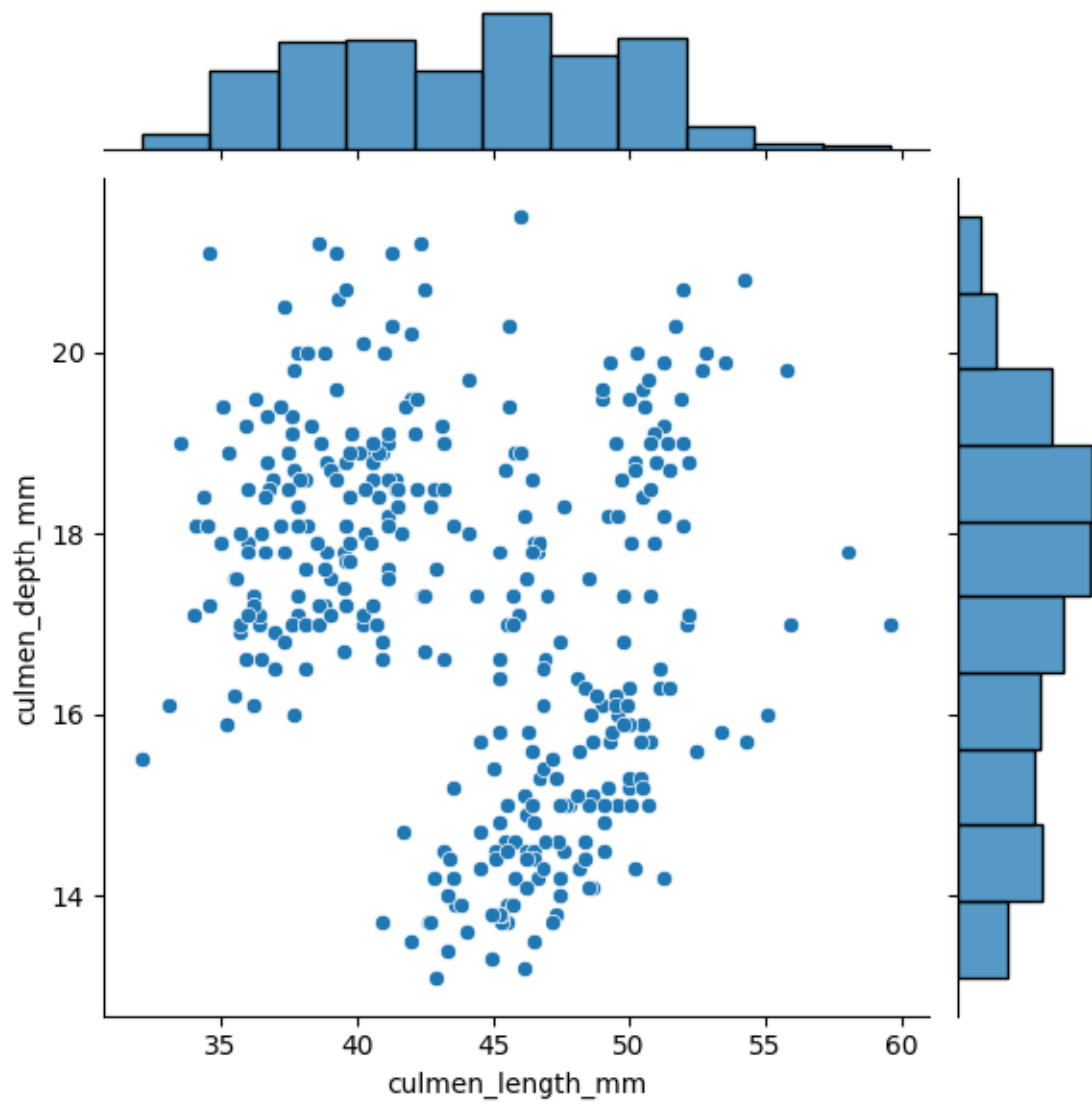
```
[4]: <Axes: xlabel='body_mass_g', ylabel='Density'>
```



#####3.2. Perform Bivariate Analysis

```
[5]: sns.jointplot(x='culmen_length_mm',y='culmen_depth_mm',data=df)
```

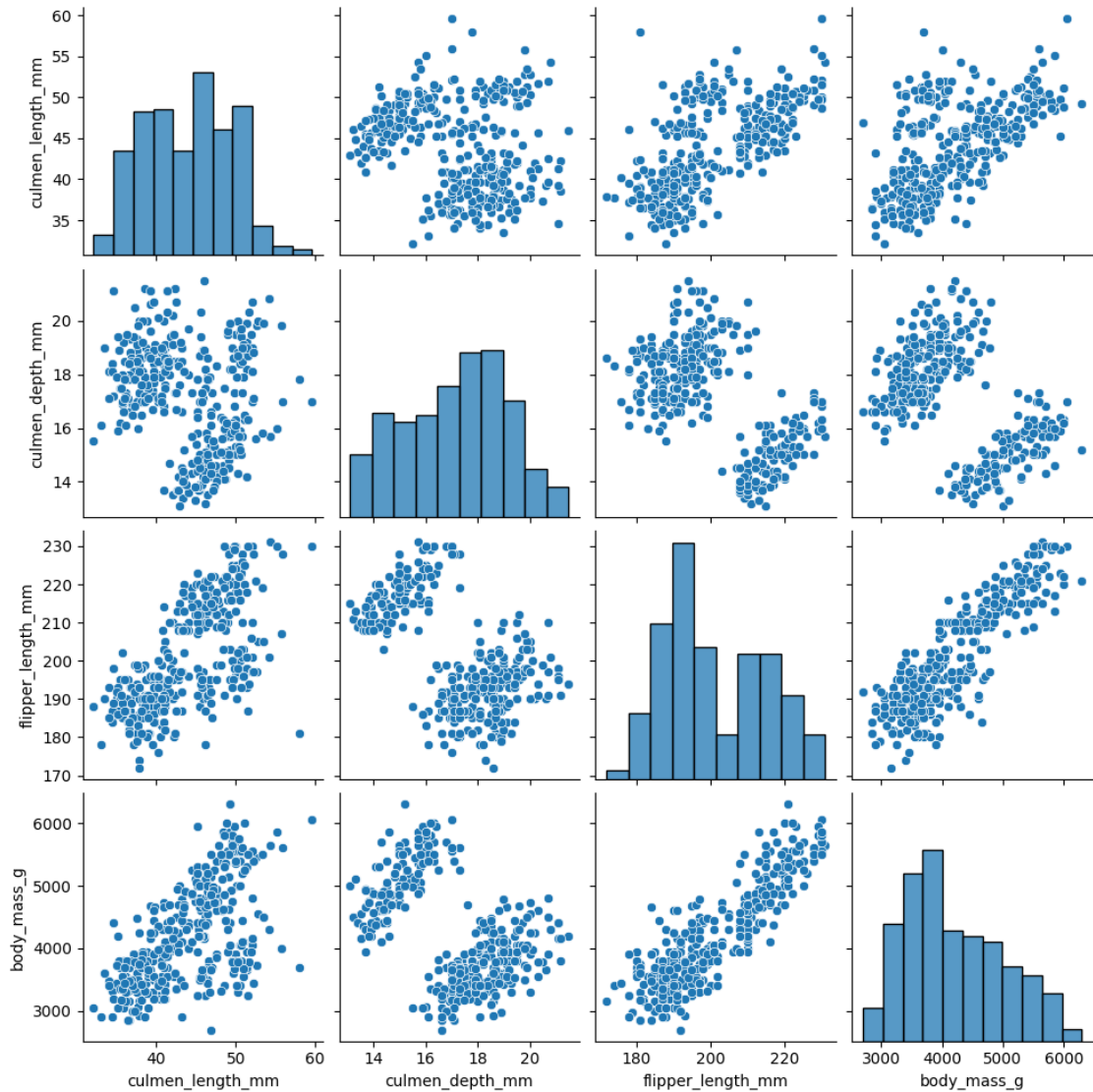
```
[5]: <seaborn.axisgrid.JointGrid at 0x7c313325c6a0>
```



#####3.3. Perform Multi-Variate Analysis

```
[6]: sns.pairplot(df)
```

```
[6]: <seaborn.axisgrid.PairGrid at 0x7c31298f71f0>
```



#####4. Perform descriptive statistics on the dataset.

```
[7]: df.describe()
```

```
[7]:
```

| | culmen_length_mm | culmen_depth_mm | flipper_length_mm | body_mass_g |
|-------|------------------|-----------------|-------------------|-------------|
| count | 342.000000 | 342.000000 | 342.000000 | 342.000000 |
| mean | 43.921930 | 17.151170 | 200.915205 | 4201.754386 |
| std | 5.459584 | 1.974793 | 14.061714 | 801.954536 |
| 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 |
| max | 59.600000 | 21.500000 | 231.000000 | 6300.000000 |

#####5. Check for Missing values and deal with them.

```
[8]: df.isnull().any() #Checking is there any null values in our dataset
```

```
[8]: species           False
     island            False
     culmen_length_mm   True
     culmen_depth_mm    True
     flipper_length_mm  True
     body_mass_g        True
     sex                True
     dtype: bool
```

```
[9]: df.isnull().sum()
```

```
[9]: species           0
     island            0
     culmen_length_mm   2
     culmen_depth_mm    2
     flipper_length_mm  2
     body_mass_g        2
     sex               10
     dtype: int64
```

```
[25]: # Code to replace null values in numerical columns with MEDIAN
df['culmen_length_mm'].fillna(df['culmen_length_mm'].median(),inplace=True)
df['culmen_depth_mm'].fillna(df['culmen_depth_mm'].median(),inplace=True)
df['flipper_length_mm'].fillna(df['flipper_length_mm'].median(),inplace=True)
df['body_mass_g'].fillna(df['body_mass_g'].median(),inplace=True)

# Code to replace null values in categorical column with MODE
df['sex'].fillna(df['sex'].mode().iloc[0],inplace=True)
```

```
[26]: # Now all null values are replaced with median and mode and dealt properly.
```

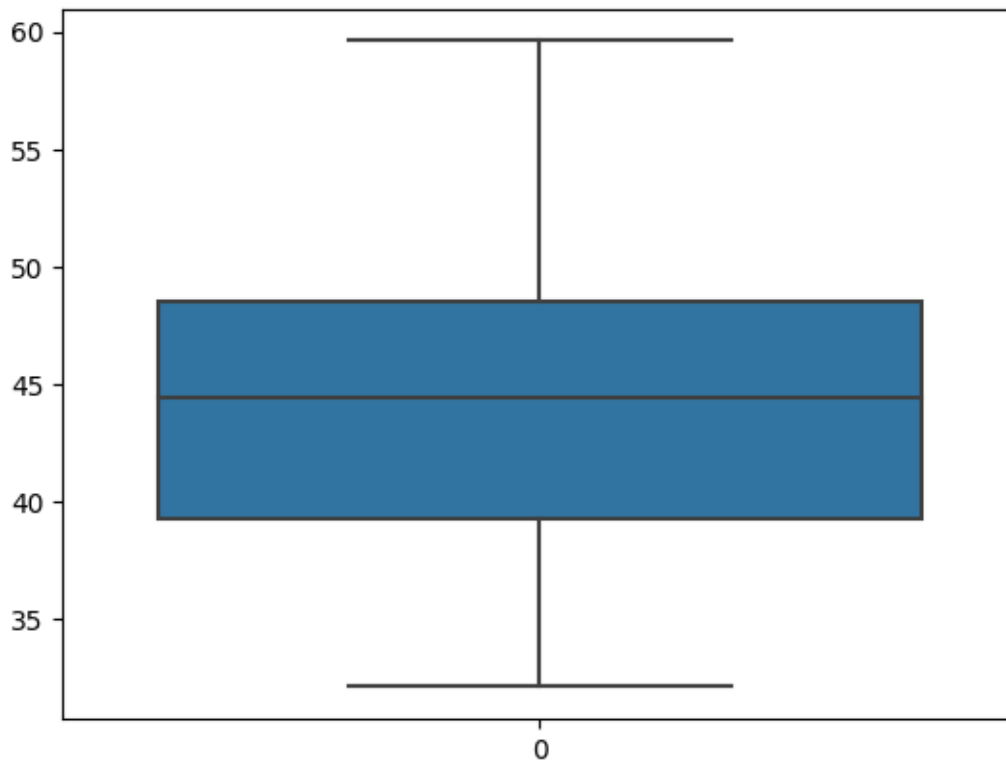
```
df.isnull().any()
```

```
[26]: species           False
     island            False
     culmen_length_mm   False
     culmen_depth_mm    False
     flipper_length_mm  False
     body_mass_g        False
     sex                False
     dtype: bool
```

#####6. Find the outliers and replace the outliers

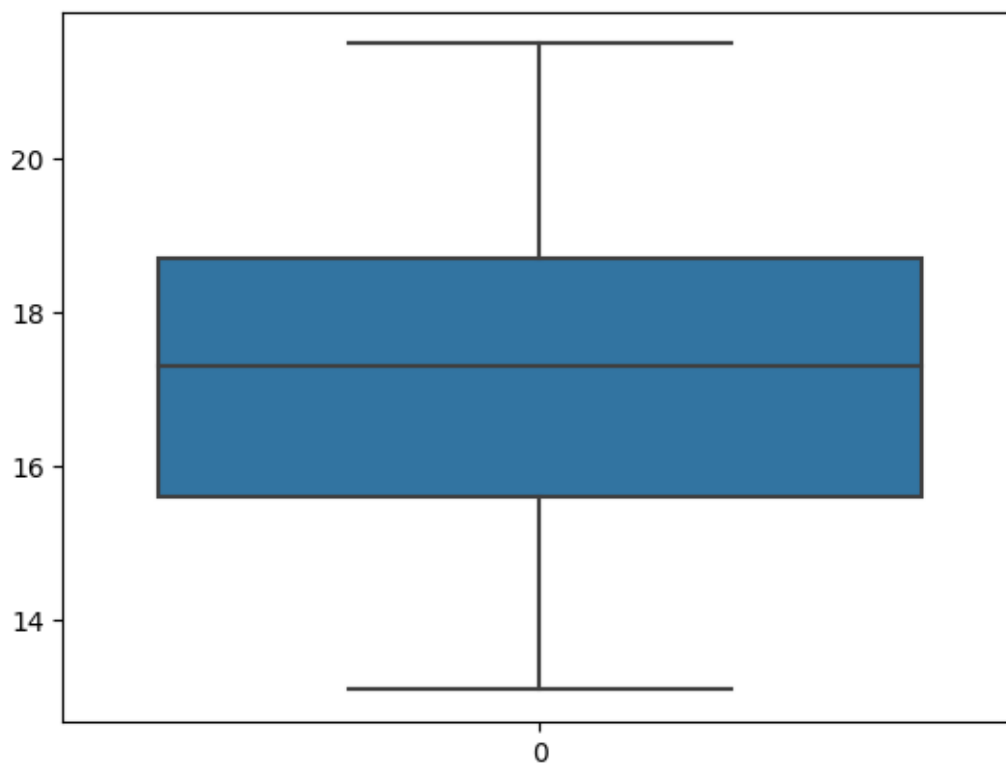
```
[29]: sns.boxplot(df.culmen_length_mm)
```

```
[29]: <Axes: >
```



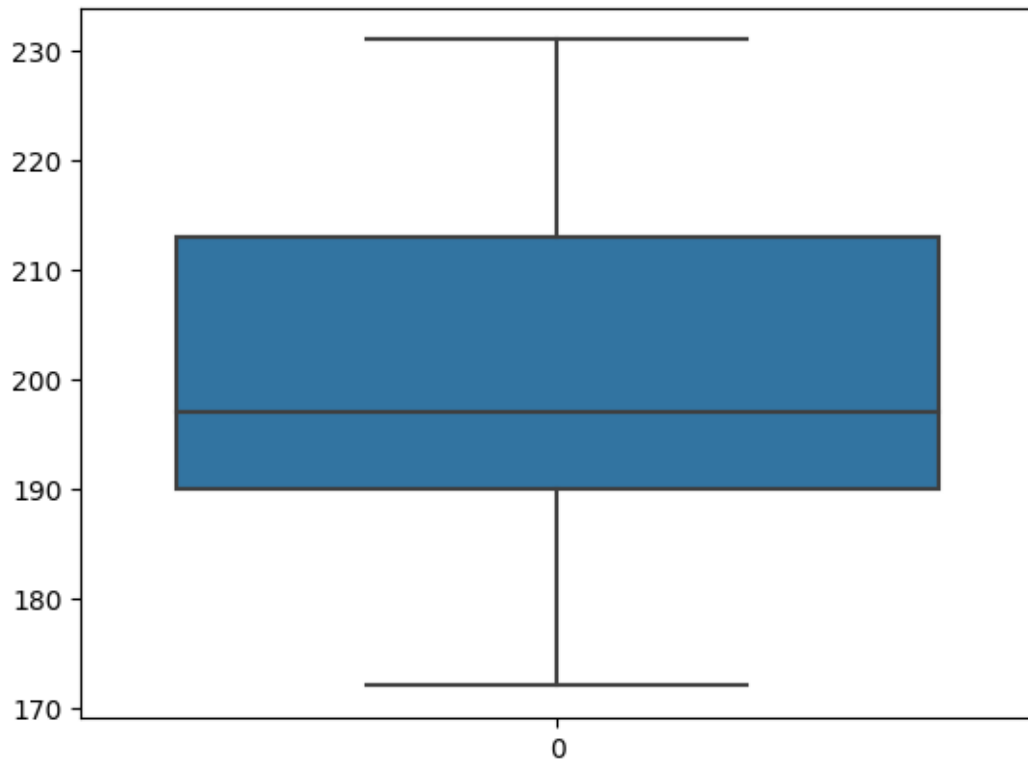
```
[30]: sns.boxplot(df.culmen_depth_mm)
```

```
[30]: <Axes: >
```



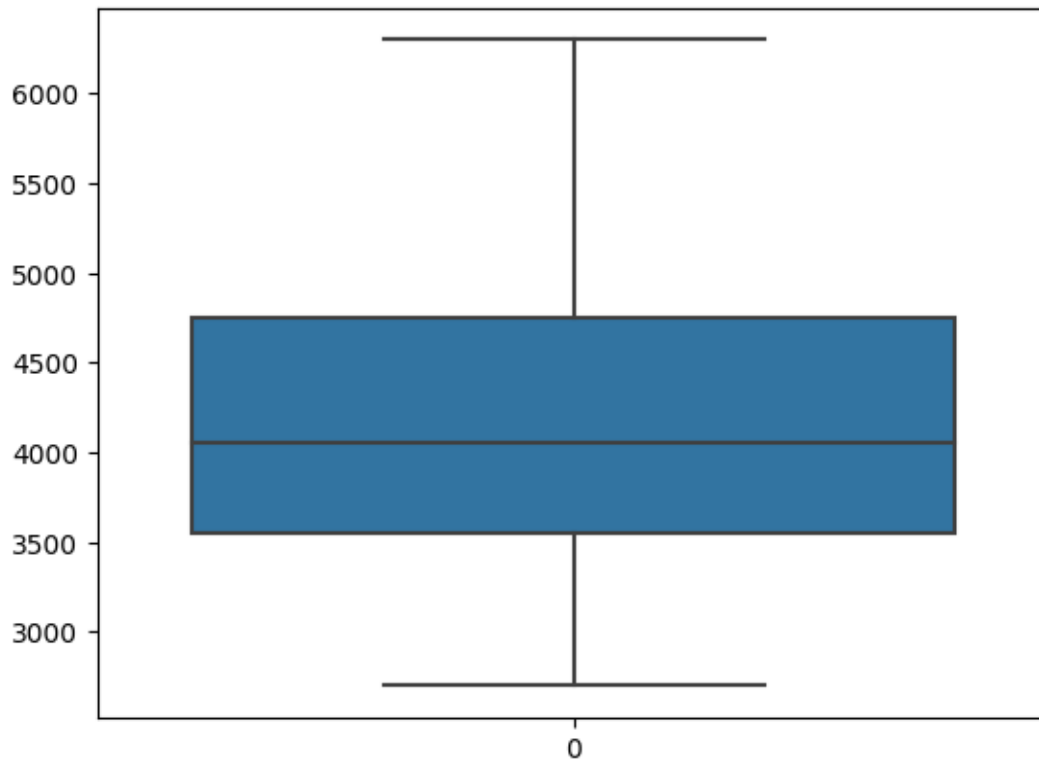
```
[31]: sns.boxplot(df.flipper_length_mm)
```

```
[31]: <Axes: >
```



```
[32]: sns.boxplot(df.body_mass_g)
```

```
[32]: <Axes: >
```

#####Hence there are no outliers in the dataset.

#####7. Check for Categorical columns and perform encoding.

```
[37]: from sklearn.preprocessing import LabelEncoder
le = LabelEncoder()
df['sex'] = le.fit_transform(df['sex'])
df['species'] = le.fit_transform(df['species'])
df['island'] = le.fit_transform(df['island'])
df.head()
```

```
[37]:
```

| | species | island | culmen_length_mm | culmen_depth_mm | flipper_length_mm | \ |
|---|---------|--------|------------------|-----------------|-------------------|---|
| 0 | 0 | 2 | 39.10 | 18.7 | 181.0 | |
| 1 | 0 | 2 | 39.50 | 17.4 | 186.0 | |
| 2 | 0 | 2 | 40.30 | 18.0 | 195.0 | |
| 3 | 0 | 2 | 44.45 | 17.3 | 197.0 | |
| 4 | 0 | 2 | 36.70 | 19.3 | 193.0 | |

| | body_mass_g | sex |
|---|-------------|-----|
| 0 | 3750.0 | 2 |
| 1 | 3800.0 | 1 |
| 2 | 3250.0 | 1 |
| 3 | 4050.0 | 2 |

```
4          3450.0      1
```

#####8. Check the correlation of independent variables with the target (TARGET IS SPECIES and remaining are independent)

```
[38]: df.corr().species.sort_values(ascending=False)
```

```
[38]: species          1.000000
flipper_length_mm    0.850819
body_mass_g          0.747547
culmen_length_mm     0.728706
sex                 -0.003823
island              -0.635659
culmen_depth_mm     -0.741282
Name: species, dtype: float64
```

#####9. Split the data into dependent and independent variables

```
[40]: X=df.drop(columns=['species'],axis=1)
X.head()
```

```
[40]:   island  culmen_length_mm  culmen_depth_mm  flipper_length_mm  body_mass_g  \
0        2          39.10           18.7           181.0        3750.0
1        2          39.50           17.4           186.0        3800.0
2        2          40.30           18.0           195.0        3250.0
3        2          44.45           17.3           197.0        4050.0
4        2          36.70           19.3           193.0        3450.0

      sex
0      2
1      1
2      1
3      2
4      1
```

```
[41]: Y=df['species']
Y.head()
```

```
[41]: 0      0
1      0
2      0
3      0
4      0
Name: species, dtype: int64
```

#####10. Scaling the data

```
[42]: from sklearn.preprocessing import MinMaxScaler
scale = MinMaxScaler()
X_scaled = pd.DataFrame(scale.fit_transform(X), columns=X.columns)
X_scaled.head()
```

```
[42]:      island  culmen_length_mm  culmen_depth_mm  flipper_length_mm  body_mass_g  \
0      1.0      0.254545      0.666667      0.152542      0.291667
1      1.0      0.269091      0.511905      0.237288      0.305556
2      1.0      0.298182      0.583333      0.389831      0.152778
3      1.0      0.449091      0.500000      0.423729      0.375000
4      1.0      0.167273      0.738095      0.355932      0.208333
```

```
      sex
0  1.0
1  0.5
2  0.5
3  1.0
4  0.5
```

#####11. Split the data into training and testing

```
[48]: from sklearn.model_selection import train_test_split
X_train,X_test,Y_train,Y_test = train_test_split(X_scaled,Y,test_size=0.
↪2,random_state=0)
```

#####12. Check the training and testing data shape.

```
[49]: X_train.shape
```

```
[49]: (275, 6)
```

```
[50]: X_test.shape
```

```
[50]: (69, 6)
```

```
[51]: Y_train.shape
```

```
[51]: (275,)
```

```
[52]: Y_test.shape
```

```
[52]: (69,)
```

#The End!!!