### 1. penguins\_size.csv is downloaded

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

```
import numpy as np
import pandas as pd

df = pd.read_csv('/content/penguins_size.csv')
df.head()
```

|   | species | island    | culmen_length_mm | culmen_depth_mm | flipper_length_mm | body_ |
|---|---------|-----------|------------------|-----------------|-------------------|-------|
| 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             |       |

### ▼ 3.1. Perform Univariate Analysis

from matplotlib import rcParams
import seaborn as sns

sns.distplot(df.body\_mass\_g)

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

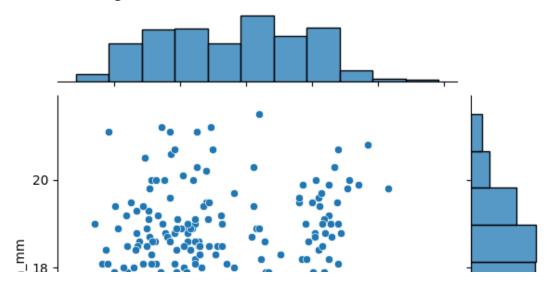
`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

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

### ▼ 3.2. Perform Bivariate Analysis

sns.jointplot(x='culmen\_length\_mm',y='culmen\_depth\_mm',data=df)

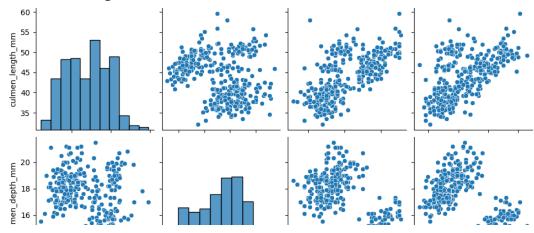
<seaborn.axisgrid.JointGrid at 0x7c313325c6a0>



### ▼ 3.3. Perform Multi-Variate Analysis

sns.pairplot(df)

### <seaborn.axisgrid.PairGrid at 0x7c31298f71f0>



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

df.describe()

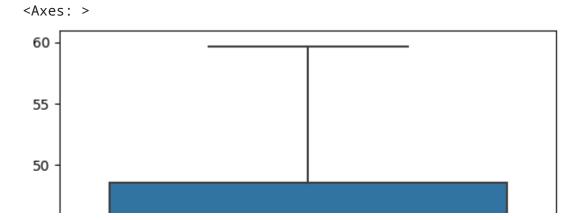
|       | culmen_length_mm | culmen_depth_mm | flipper_length_mm | body_ |
|-------|------------------|-----------------|-------------------|-------|
| count | 342.000000       | 342.000000      | 342.000000        | 342   |

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

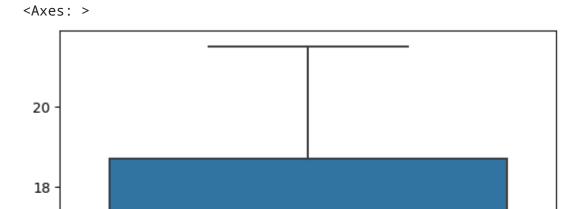
```
df.isnull().any() #Checking is there any null values in our dataset
                            False
     species
     island
                            False
     culmen_length_mm
                             True
     culmen_depth_mm
                             True
     flipper_length_mm
                             True
     body_mass_g
                             True
                             True
     sex
     dtype: bool
df.isnull().sum()
     species
                             0
     island
                             0
     culmen_length_mm
                             2
                             2
     culmen_depth_mm
     flipper_length_mm
                             2
                             2
     body_mass_g
                            10
     sex
     dtype: int64
# 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)
# Now all null values are replaced with median and mode and dealt properly.
df.isnull().any()
     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

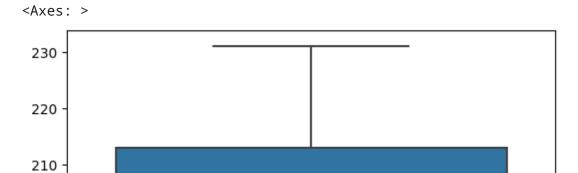
### sns.boxplot(df.culmen\_length\_mm)



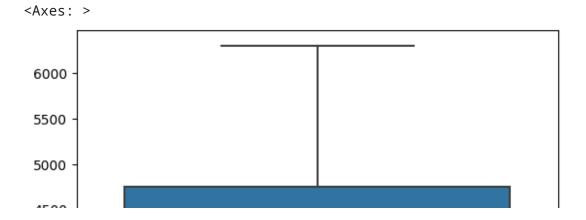
### sns.boxplot(df.culmen\_depth\_mm)



# sns.boxplot(df.flipper\_length\_mm)



### sns.boxplot(df.body\_mass\_g)



Hence there are no outliers in the dataset.

▼ 7. Check for Categorical columns and perform encoding.

```
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()
```

|   | species | island | culmen_length_mm | culmen_depth_mm | flipper_ler |
|---|---------|--------|------------------|-----------------|-------------|
| 0 | 0       | 2      | 39.10            | 18.7            |             |
| 1 | 0       | 2      | 39.50            | 17.4            |             |
| 2 | 0       | 2      | 40.30            | 18.0            |             |
| 3 | 0       | 2      | 44.45            | 17.3            |             |
| 4 | 0       | 2      | 36.70            | 19.3            |             |

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

```
df.corr().species.sort_values(ascending=False)
```

```
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

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

|   | island | culmen_length_mm | culmen_depth_mm | flipper_length_mm | k |
|---|--------|------------------|-----------------|-------------------|---|
| 0 | 2      | 39.10            | 18.7            | 181.0             |   |
| 1 | 2      | 39.50            | 17.4            | 186.0             |   |
| 2 | 2      | 40.30            | 18.0            | 195.0             |   |

Y=df['species']
Y.head()

Name: species, dtype: int64

### ▼ 10. Scaling the independent data

from sklearn.preprocessing import MinMaxScaler
scale = MinMaxScaler()
X\_scaled = pd.DataFrame(scale.fit\_transform(X),columns=X.columns)
X\_scaled.head()

|   | island | culmen_length_mm | culmen_depth_mm | flipper_length_mm k |
|---|--------|------------------|-----------------|---------------------|
| 0 | 1.0    | 0.254545         | 0.666667        | 0.152542            |
| 1 | 1.0    | 0.269091         | 0.511905        | 0.237288            |
| 2 | 1.0    | 0.298182         | 0.583333        | 0.389831            |
| 3 | 1.0    | 0.449091         | 0.500000        | 0.423729            |
| 4 | 1.0    | 0.167273         | 0.738095        | 0.355932            |

### ▼ 11. Split 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\_scaled,Y,test\_size=0.2,random\_state=0)

▼ 12. Check the training and testing data shape.

X\_train.shape (275, 6)

# The End!!!

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