

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

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

```
df.head()
```

|   | species | island    | culmen_length_mm | culmen_depth_mm | flipper_length_mm | body_mass_g |
|---|---------|-----------|------------------|-----------------|-------------------|-------------|
| 0 | Adelie  | Torgersen | 39.1             | 18.7            | 181.0             | 3750.0      |
| 1 | Adelie  | Torgersen | 39.5             | 17.4            | 186.0             | 3800.0      |
| 2 | Adelie  | Torgersen | 40.3             | 18.0            | 195.0             | 3250.0      |
| 3 | Adelie  | Torgersen | NaN              | NaN             | NaN               | NaN         |
| 4 | Adelie  | Torgersen | 36.7             | 19.3            | 193.0             | 3450.0      |

```
df.shape
```

```
(344, 7)
```

## 1.Univariate Analysis

```
sbn.distplot(df["body_mass_g"])
```

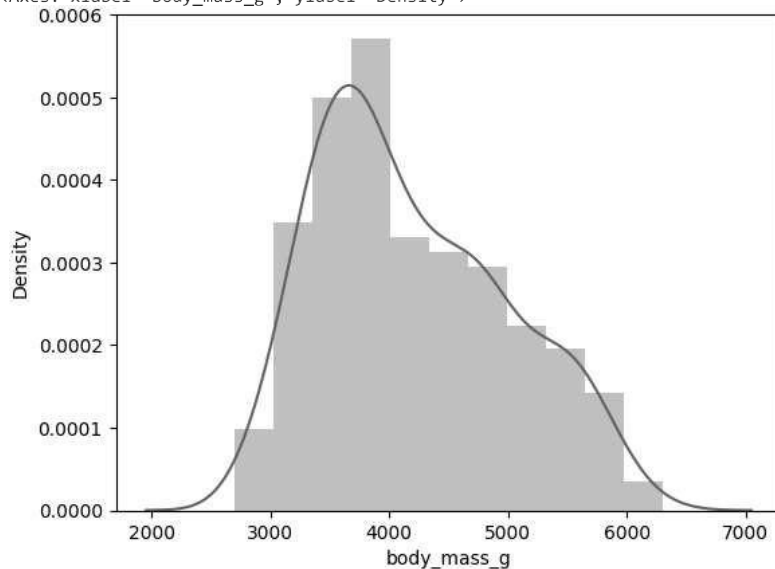
```
<ipython-input-9-9a0d592a49b0>:1: UserWarning:
```

```
`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).
```

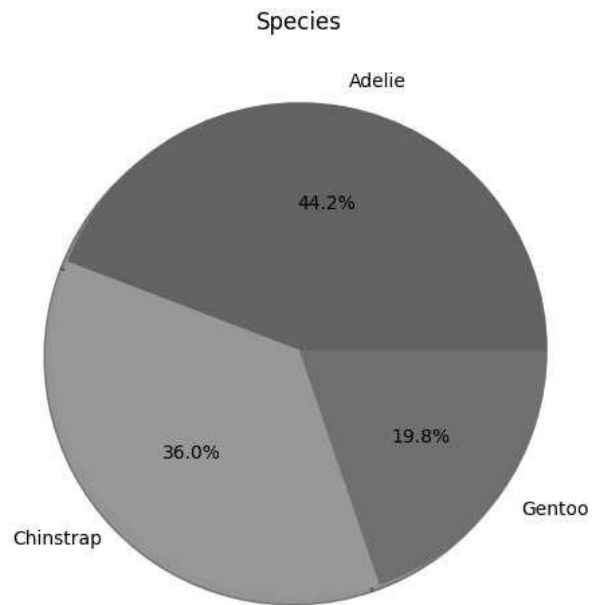
```
For a guide to updating your code to use the new functions, please see https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751
```

```
sbn.distplot(df["body_mass_g"])
<Axes: xlabel='body_mass_g', ylabel='Density'>
```



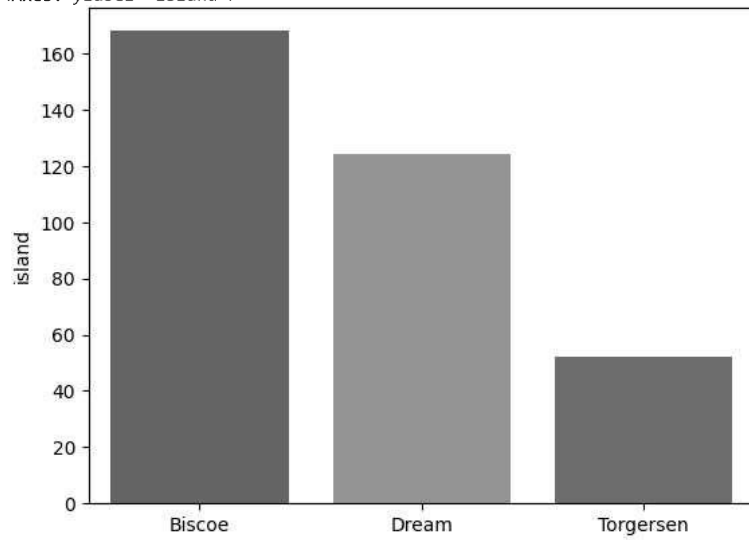
```
plt.figure(figsize=(6,6))
plt.pie(df["species"].value_counts(),labels=df["species"].unique(),autopct='%1.1f%%',shadow = True)
plt.title("Species")
plt.show
```

```
<Function matplotlib.pyplot.show(close=None, block=None)>
```



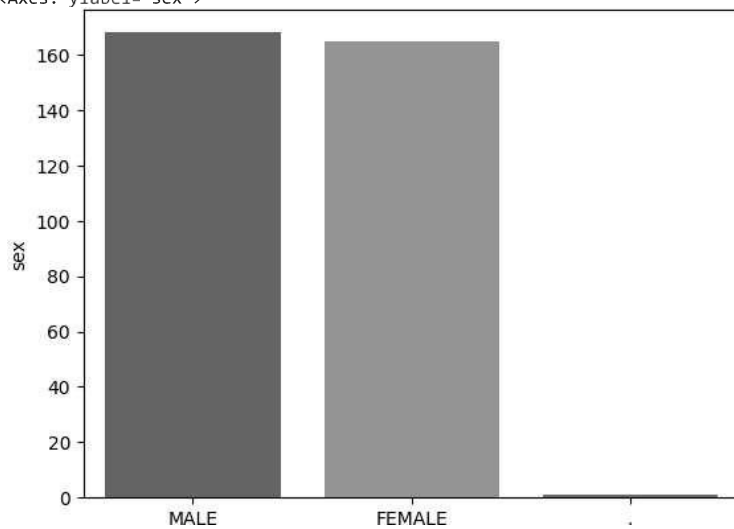
```
sbn.barplot(x=df["island"].value_counts().index,y=df["island"].value_counts())
```

<Axes: ylabel='island'>



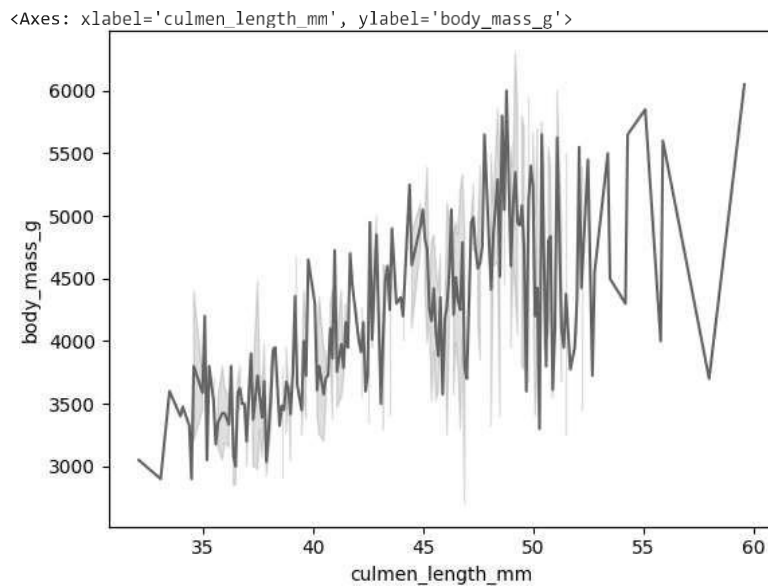
```
sbn.barplot(x=df["sex"].value_counts().index,y=df["sex"].value_counts())
```

<Axes: ylabel='sex'>

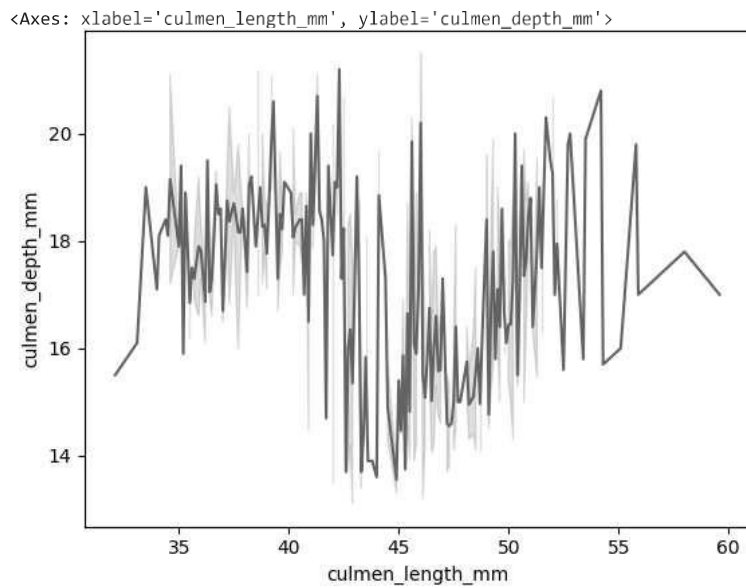


## 2. Bivariate Analysis

```
sbn.lineplot(x=df["culmen_length_mm"],y=df["body_mass_g"])
```

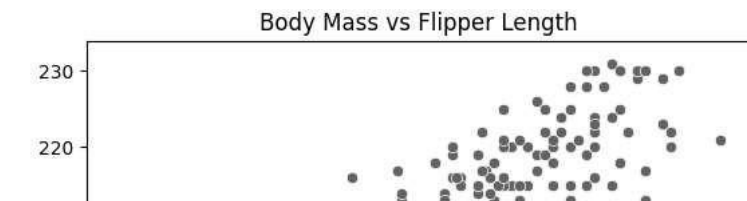


```
sbn.lineplot(x=df['culmen_length_mm'],y=df['culmen_depth_mm'])
```



```
plt.title("Body Mass vs Flipper Length")  
sbn.scatterplot(x=df["body_mass_g"],y=df['flipper_length_mm'])
```

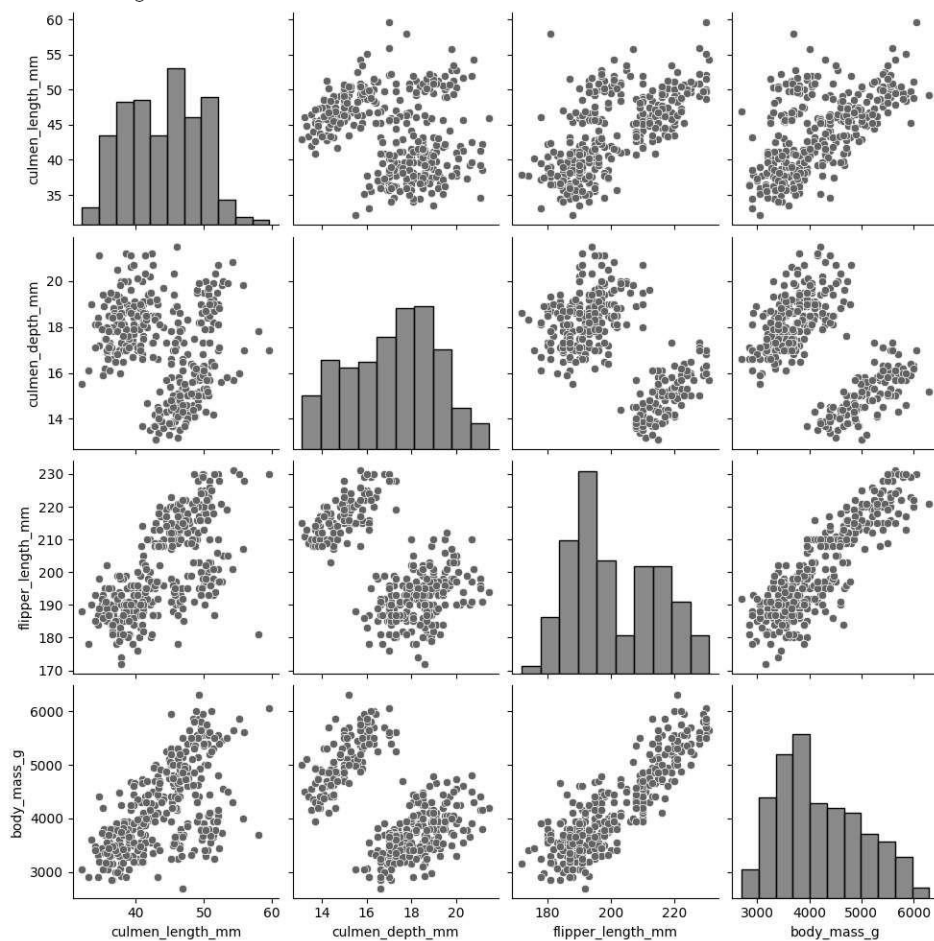
```
<Axes: title={'center': 'Body Mass vs Flipper Length'}, xlabel='body_mass_g',
ylabel='flipper_length_mm'>
```



### 3.Multivariate Analysis

```
gt | |
sbn.pairplot(df)
```

```
<seaborn.axisgrid.PairGrid at 0x7aac8d553880>
```



### 4.Descriptive Statistics

```
dstats = df.describe()
print(dstats)
```

|       | 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

```
#THE MISSING VALUES
missing_values = df.isnull().sum()
print(missing_values)
```

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

### ▼ Dealing with missing values

```
#Input values using mean for numerical value based
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)
df['sex'].fillna(df['sex'].mode()[0],inplace=True)
```

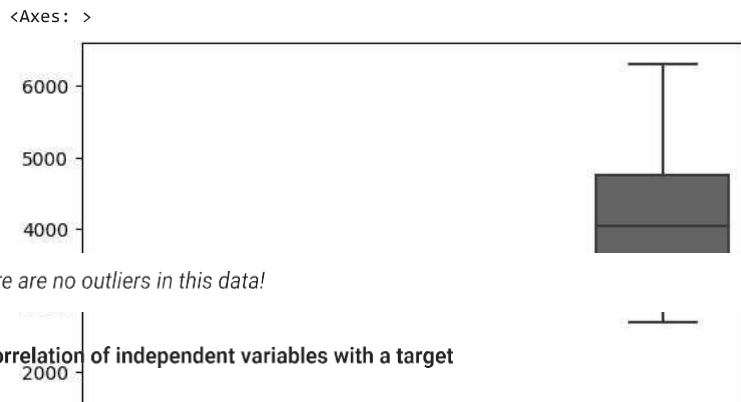
Checking for missing values again:

```
#THE MISSING VALUES
missing_values = df.isnull().sum()
print(missing_values)
```

```
species          0
island           0
culmen_length_mm  0
culmen_depth_mm  0
flipper_length_mm 0
body_mass_g      0
sex              0
dtype: int64
```

## 6.Outlier Removal

```
#viewing the outliers
sbn.boxplot(data=df[['culmen_length_mm', 'culmen_depth_mm', 'flipper_length_mm', 'body_mass_g']])
```



There are no outliers in this data!

### 7. Correlation of independent variables with a target

```
cmatrix = df[['culmen_length_mm', 'culmen_depth_mm', 'flipper_length_mm', 'body_mass_g']].corr()

# correlation with the target variable 'body_mass_g'
target = cmatrix['body_mass_g']
print(target)

culmen_length_mm    0.595110
culmen_depth_mm     -0.471916
flipper_length_mm   0.871202
body_mass_g         1.000000
Name: body_mass_g, dtype: float64
```

### 8. Check for categorical columns and encode them

```
#checking for categorical columns
ccolumn = df.select_dtypes(include=['object']).columns
print(ccolumn)

Index(['species', 'island', 'sex'], dtype='object')

#encoding the columns
df_encode = pd.get_dummies(df, columns=ccolumn, drop_first=True)
df_encode.head()
```

|   | culmen_length_mm | culmen_depth_mm | flipper_length_mm | body_mass_g | species_Chinstrap | species_Gentoo | island_Dream | island_Torgersen |
|---|------------------|-----------------|-------------------|-------------|-------------------|----------------|--------------|------------------|
| 0 | 39.10000         | 18.70000        | 181.000000        | 3750.000000 | 0                 | 0              | 0            | 1                |
| 1 | 39.50000         | 17.40000        | 186.000000        | 3800.000000 | 0                 | 0              | 0            | 1                |
| 2 | 40.30000         | 18.00000        | 195.000000        | 3250.000000 | 0                 | 0              | 0            | 1                |
| 3 | 43.92193         | 17.15117        | 200.915205        | 4201.754386 | 0                 | 0              | 0            | 1                |
| 4 | 36.70000         | 19.30000        | 193.000000        | 3450.000000 | 0                 | 0              | 0            | 1                |

### 9. Splitting of data to Independent and Dependent

```
X = df_encode.iloc[:, :-1].values #taking the independent variable species
Y = df_encode.iloc[:, 9] # our dependent variables
print(X)
print(Y)
```

```
[[ 39.1  18.7 181. ...  0.  1.  0. ]
 [ 39.5  17.4 186. ...  0.  1.  1. ]
 [ 40.3  18.  195. ...  0.  1.  1. ]
 ...
 [ 50.4  15.7 222. ...  0.  0.  0. ]
 [ 45.2  14.8 212. ...  0.  0.  1. ]
 [ 49.9  16.1 213. ...  0.  0.  0. ]]
0      1
1      0
2      0
3      1
4      0
..
339    1
340    0
341    1
342    0
```

```
343    1
Name: sex_MALE, Length: 344, dtype: uint8
```

## 10. Scaling the Data

```
from sklearn.preprocessing import StandardScaler
scaler = StandardScaler()
X_scaled = scaler.fit_transform(X)
```

## 11. Split the data into training and testing

```
from sklearn.model_selection import train_test_split

# Split the data into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(X_scaled, Y, test_size=0.2, random_state=42)
```

## 12. Check Shape

```
print("Training data shape:", X_train.shape, y_train.shape)
print("Testing data shape:", X_test.shape, y_test.shape)
```

```
Training data shape: (275, 9) (275,)
Testing data shape: (69, 9) (69,)
```