```
import numpy as nmpy
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sbn

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"])
```

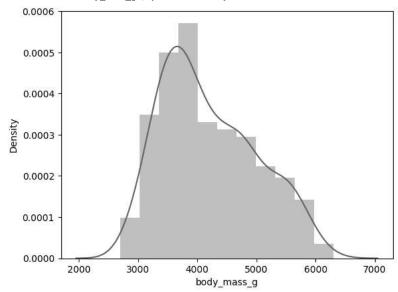
<ipython-input-9-9a0d592a49b0>:1: UserWarning:

`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 $\underline{\text{https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751}}$

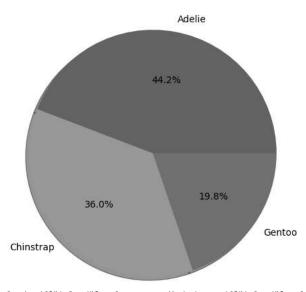
```
sbn.distplot(df["body_mass_g"])
<Axes: xlabel='body_mass_g', ylabel='Density'>
```



```
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
```

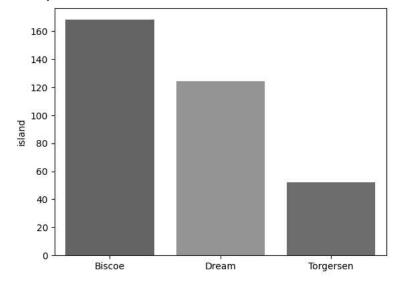
<function matplotlib.pyplot.show(close=None, block=None)>

Species



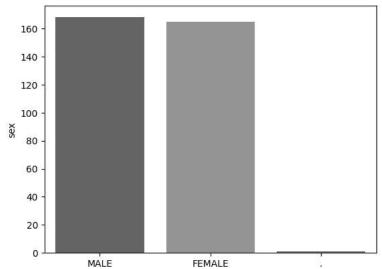
sbn.barplot(x=df["island"].value_counts().index,y=df["island"].value_counts())

<Axes: ylabel='island'>



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

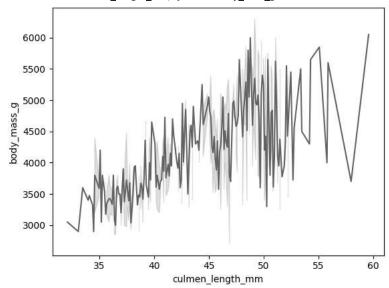
<Axes: ylabel='sex'>



2. Bivariate Analysis

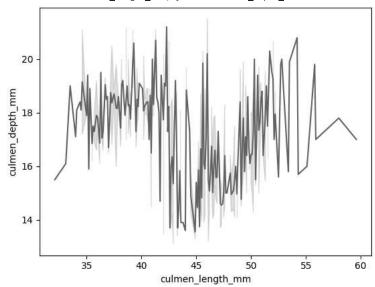
sbn.lineplot(x=df["culmen_length_mm"],y=df["body_mass_g"])

<Axes: xlabel='culmen_length_mm', ylabel='body_mass_g'>



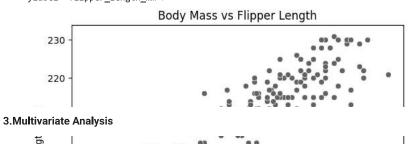
sbn.lineplot(x=df['culmen_length_mm'],y=df['culmen_depth_mm'])

<Axes: xlabel='culmen_length_mm', ylabel='culmen_depth_mm'>

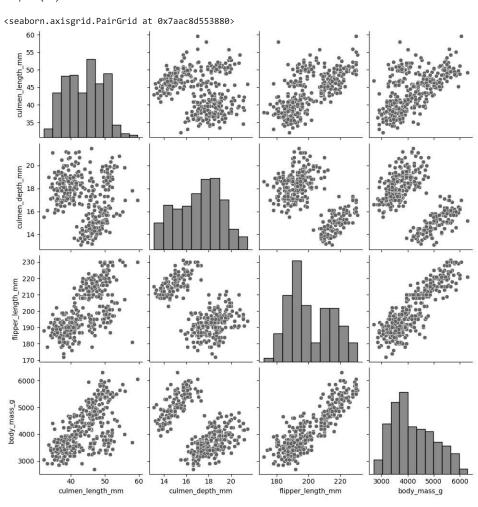


plt.title("Body Mass vs Flipper Length")
sbn.scatterplot(x=df["body_mass_g"],y=df['flipper_length_mm'])

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



sbn.pairplot(df)



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

▼ 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:

6.Outlier Removal

dtype: int64

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

```
<Axes: >
6000 -
5000 -
4000 -
There are no outliers in this data!
```

7. Correlation of independent variables with a target

8. Check for categorical columns and encode them

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

```
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,)
```

✓ 0s completed at 8:45 PM