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CSE AI AND ML  
VIT AP

```
In [ ]: import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

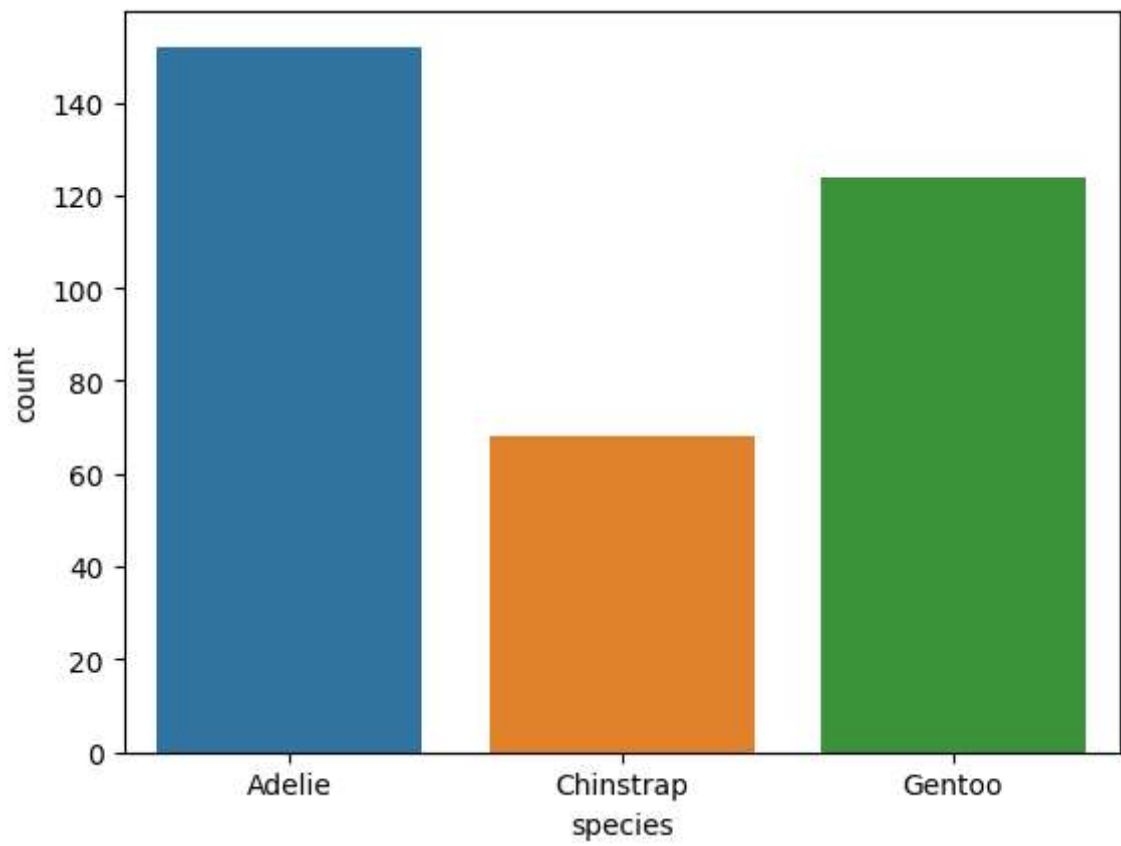
```
In [5]: df = pd.read_csv("/content/penguins_size.csv")
df
```

Out[5]:

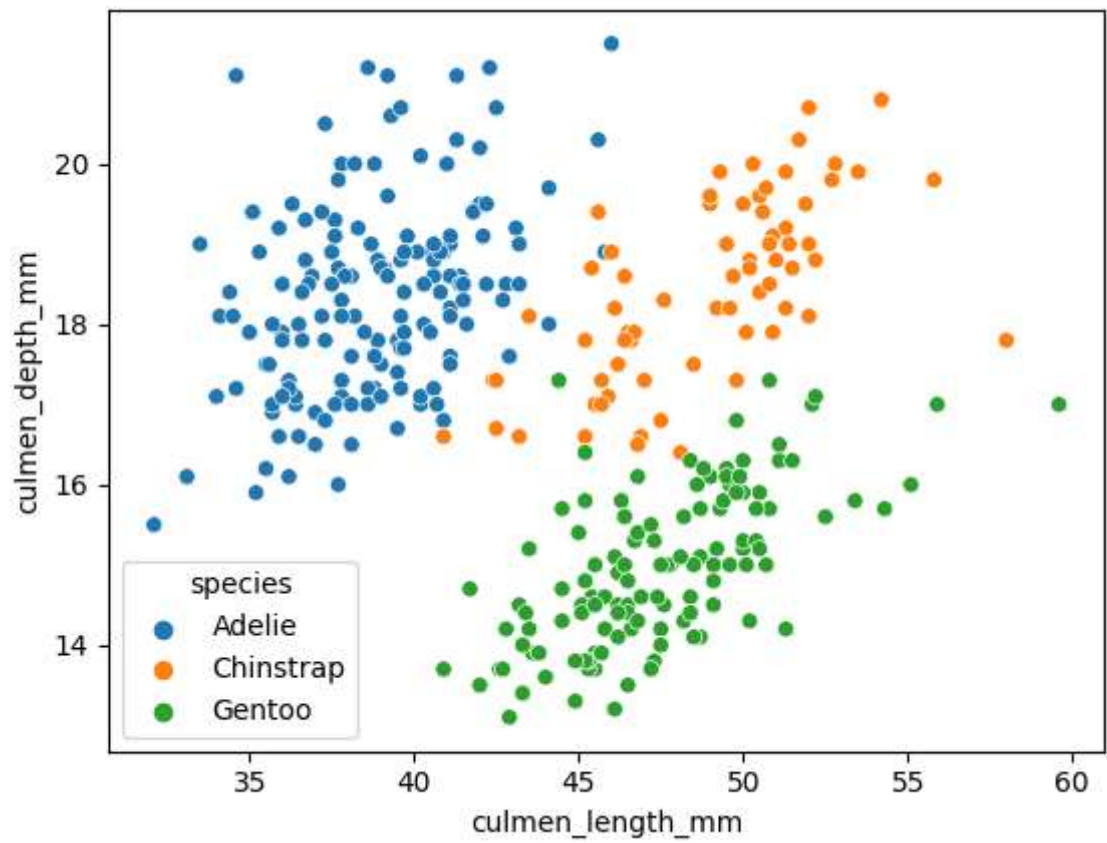
	species	island	culmen_length_mm	culmen_depth_mm	flipper_length_mm	body_mass_g
0	Adelie	Torgersen	39.1	18.7	181.0	3750
1	Adelie	Torgersen	39.5	17.4	186.0	3800
2	Adelie	Torgersen	40.3	18.0	195.0	3800
3	Adelie	Torgersen	NaN	NaN	NaN	NaN
4	Adelie	Torgersen	36.7	19.3	193.0	3400
...	...	...	...	...	...	...
339	Gentoo	Biscoe	NaN	NaN	NaN	NaN
340	Gentoo	Biscoe	46.8	14.3	215.0	4800
341	Gentoo	Biscoe	50.4	15.7	222.0	5700
342	Gentoo	Biscoe	45.2	14.8	212.0	5200
343	Gentoo	Biscoe	49.9	16.1	213.0	5400

344 rows × 7 columns

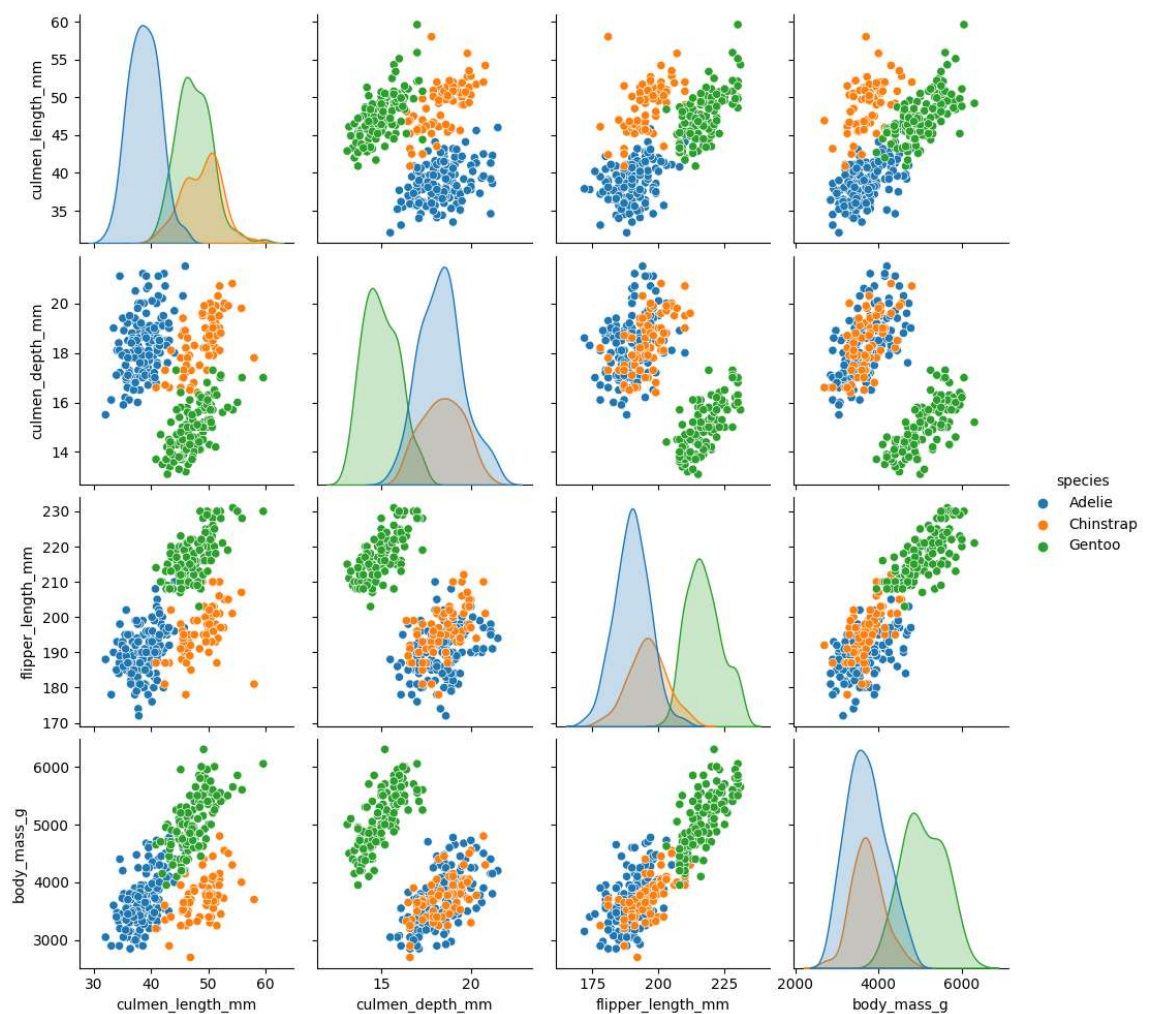
```
In [9]: # Univariate Analysis  
sns.countplot(x='species', data=df)  
plt.show()
```



```
In [11]: # Bi-Variate Analysis  
sns.scatterplot(x=df['culmen_length_mm'], y=df['culmen_depth_mm'], hue=df['  
plt.show()
```



```
In [13]: # Multi-Variate Analysis
sns.pairplot(df, hue='species')
plt.show()
```



```
In [14]: df.describe()
```

```
Out[14]:
```

	culmen_length_mm	culmen_depth_mm	flipper_length_mm	body_mass_g
<b>count</b>	342.000000	342.000000	342.000000	342.000000
<b>mean</b>	43.921930	17.151170	200.915205	4201.754386
<b>std</b>	5.459584	1.974793	14.061714	801.954536
<b>min</b>	32.100000	13.100000	172.000000	2700.000000
<b>25%</b>	39.225000	15.600000	190.000000	3550.000000
<b>50%</b>	44.450000	17.300000	197.000000	4050.000000
<b>75%</b>	48.500000	18.700000	213.000000	4750.000000
<b>max</b>	59.600000	21.500000	231.000000	6300.000000

```
In [15]: # Check for missing values
print(df.isnull().sum())

# Impute or remove (example: using mean for numeric columns)
df['culmen_length_mm'].fillna(df['culmen_length_mm'].mean(), inplace=True)
```

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

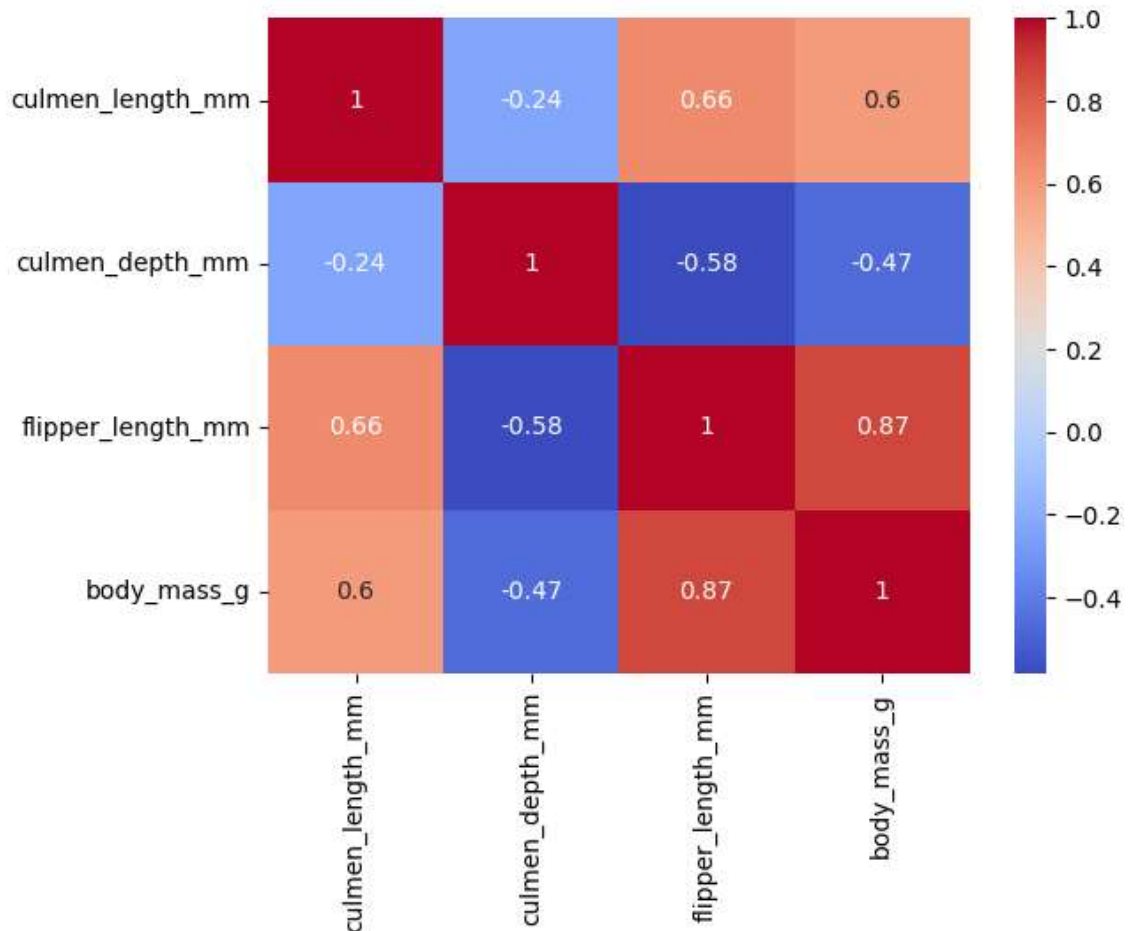
```
In [18]: Q1 = df.quantile(0.25, numeric_only=True)
Q3 = df.quantile(0.75, numeric_only=True)
IQR = Q3 - Q1

numerical_cols = df.select_dtypes(include=['float64', 'int64']).columns
outliers = ((df[numerical_cols] < (Q1 - 1.5 * IQR)) | (df[numerical_cols] >
df = df[~outliers.any(axis=1)]
```

```
In [19]: correlation = df.corr()
sns.heatmap(correlation, annot=True, cmap='coolwarm')
plt.show()
```

<ipython-input-19-b80295645867>:1: FutureWarning: The default value of numeric\_only in DataFrame.corr is deprecated. In a future version, it will default to False. Select only valid columns or specify the value of numeric\_only to silence this warning.

```
correlation = df.corr()
```



```
In [36]: print(df.columns)
```

```
Index(['species', 'culmen_length_mm', 'culmen_depth_mm', 'flipper_length_m
m',
      'body_mass_g', 'island_Dream', 'island_Torgersen', 'sex_FEMALE',
      'sex_MALE'],
      dtype='object')
```

```
In [30]: X = df.drop("species", axis=1)
y = df["species"]
print("Data split into dependent and independent variables!")
```

Data split into dependent and independent variables!

In [31]: `from sklearn.preprocessing import StandardScaler`

```
scaler = StandardScaler()  
X_scaled = scaler.fit_transform(X)  
print("Data scaled!")
```

Data scaled!

In [32]: `from sklearn.model_selection import train_test_split`

```
X_train, X_test, y_train, y_test = train_test_split(X_scaled, y, test_size=  
print("Data split into training and testing sets!")
```

Data split into training and testing sets!

In [26]: `print("X_train shape:", X_train.shape)  
print("y_train shape:", y_train.shape)  
print("X_test shape:", X_test.shape)  
print("y_test shape:", y_test.shape)`

```
X_train shape: (275, 8)  
y_train shape: (275,)  
X_test shape: (69, 8)  
y_test shape: (69,)
```