

5.EDA-Data Visualization

AIM:

To explore and understand the underlying patterns, distributions, and relationships within the dataset through visual representations, which aids in uncovering insights, detecting anomalies, and guiding further analysis.

PROGRAM:

```
import seaborn as sns

import matplotlib.pyplot as plt


# Load iris dataset

df = sns.load_dataset('iris')


# 1. Histogram and KDE plot of sepal_length

plt.figure(figsize=(8, 4))

sns.histplot(df['sepal_length'], kde=True, bins=20, color='skyblue')

plt.title('Distribution of Sepal Length')

plt.xlabel('Sepal Length (cm)')

plt.ylabel('Frequency')

plt.show()


# 2. Boxplot of sepal_width by species (to see spread and outliers)

plt.figure(figsize=(8, 4))

sns.boxplot(x='species', y='sepal_width', data=df, palette='Set2')

plt.title('Sepal Width by Species')

plt.show()
```

```
# 3. Scatter plot of sepal_length vs sepal_width colored by species

plt.figure(figsize=(8, 6))

sns.scatterplot(x='sepal_length', y='sepal_width', hue='species', data=df,
palette='deep')

plt.title('Sepal Length vs Sepal Width by Species')

plt.show()


# 4. Pairplot of all numerical features colored by species

sns.pairplot(df, hue='species', palette='bright')

plt.suptitle('Pairplot of Iris Dataset', y=1.02)

plt.show()


# 5. Correlation heatmap of numeric variables

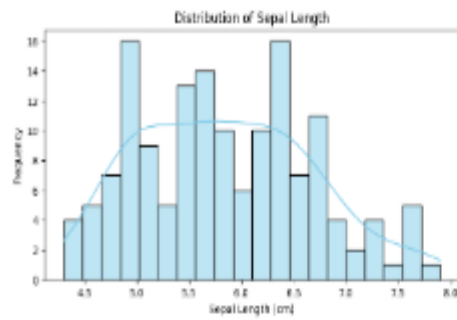
plt.figure(figsize=(6, 5))

sns.heatmap(df.corr(), annot=True, cmap='coolwarm', fmt=".2f")

plt.title('Correlation Heatmap')

plt.show()
```

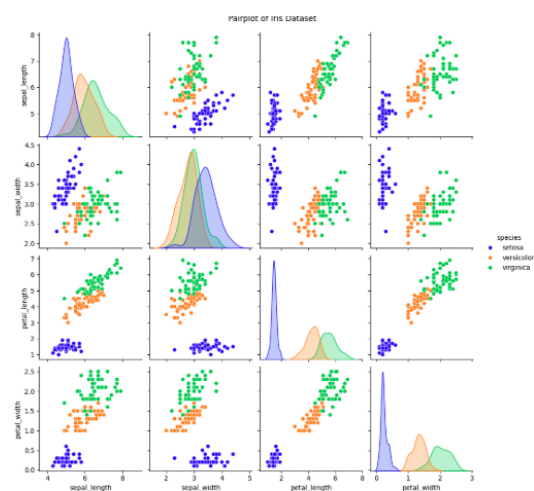
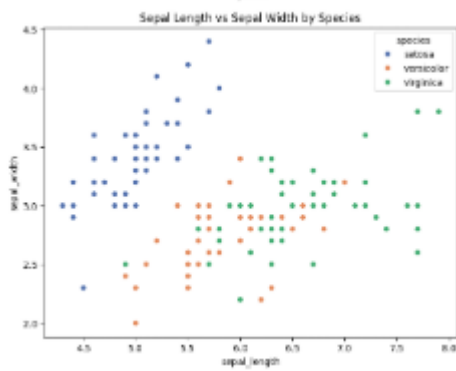
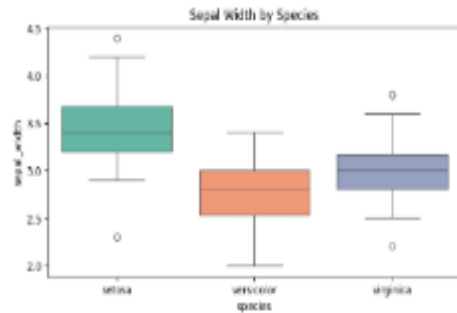
OUTPUT:

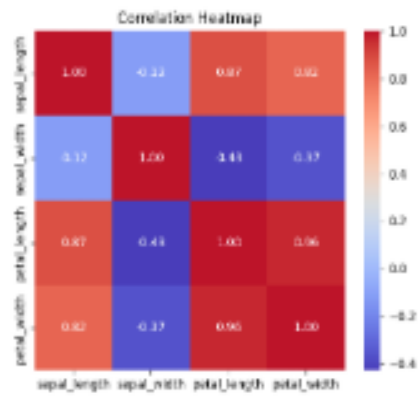


`fig, (ax1, ax2) = plt.subplots(2, 1); plt.tight_layout();`

using 'palette' without assigning 'hue' is deprecated and will be removed in v0.11.0. Assign the 'x' variable to 'hue' and set 'legend=False' for the same effect.

`sns.boxplot(x='species', y='sepal_width', data=df, palette='set2')`





RESULT:

Thus, the given program was written and executed successfully.

