

```

# Import necessary libraries
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns

# Load the Iris dataset
iris_df = pd.read_csv('C:/Users/KSK/Downloads/iris/iris.csv')

# Display the first few rows of the dataset
iris_df.head()

```

| | Id | SepalLengthCm | SepalWidthCm | PetalLengthCm | PetalWidthCm | Species |
|---|----|---------------|--------------|---------------|--------------|-------------|
| 0 | 1 | 5.1 | 3.5 | 1.4 | 0.2 | Iris-setosa |
| 1 | 2 | 4.9 | 3.0 | 1.4 | 0.2 | Iris-setosa |
| 2 | 3 | 4.7 | 3.2 | 1.3 | 0.2 | Iris-setosa |
| 3 | 4 | 4.6 | 3.1 | 1.5 | 0.2 | Iris-setosa |
| 4 | 5 | 5.0 | 3.6 | 1.4 | 0.2 | Iris-setosa |

```
iris_df.isnull().sum()
```

```

Id          0
SepalLengthCm  0
SepalWidthCm  0
PetalLengthCm  0
PetalWidthCm  0
Species      0
dtype: int64

```

```

# Summary statistics
iris_df.describe()

```

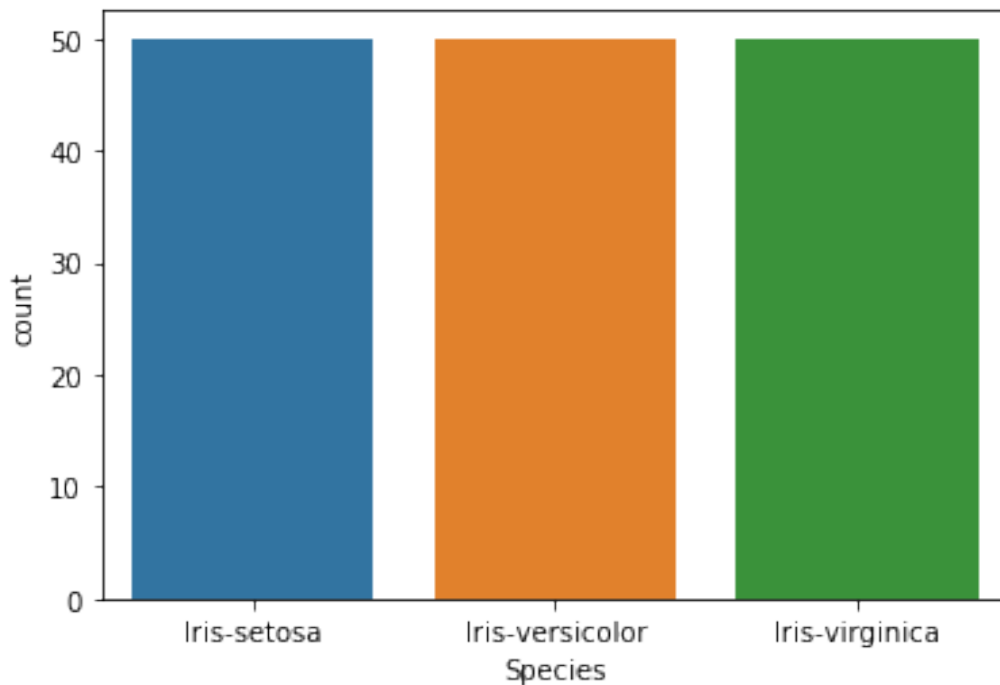
| | Id | SepalLengthCm | SepalWidthCm | PetalLengthCm | PetalWidthCm |
|-------|------------|---------------|--------------|---------------|--------------|
| count | 150.000000 | 150.000000 | 150.000000 | 150.000000 | 150.000000 |
| mean | 75.500000 | 5.843333 | 3.054000 | 3.758667 | 1.198667 |
| std | 43.445368 | 0.828066 | 0.433594 | 1.764420 | 0.763161 |
| min | 1.000000 | 4.300000 | 2.000000 | 1.000000 | 0.100000 |
| 25% | 38.250000 | 5.100000 | 2.800000 | 1.600000 | 0.300000 |
| 50% | 75.500000 | 5.800000 | 3.000000 | 4.350000 | 1.300000 |

```
75%    112.750000    6.400000    3.300000    5.100000
1.800000
max    150.000000    7.900000    4.400000    6.900000
2.500000
```

```
iris_df['Species'].value_counts()
```

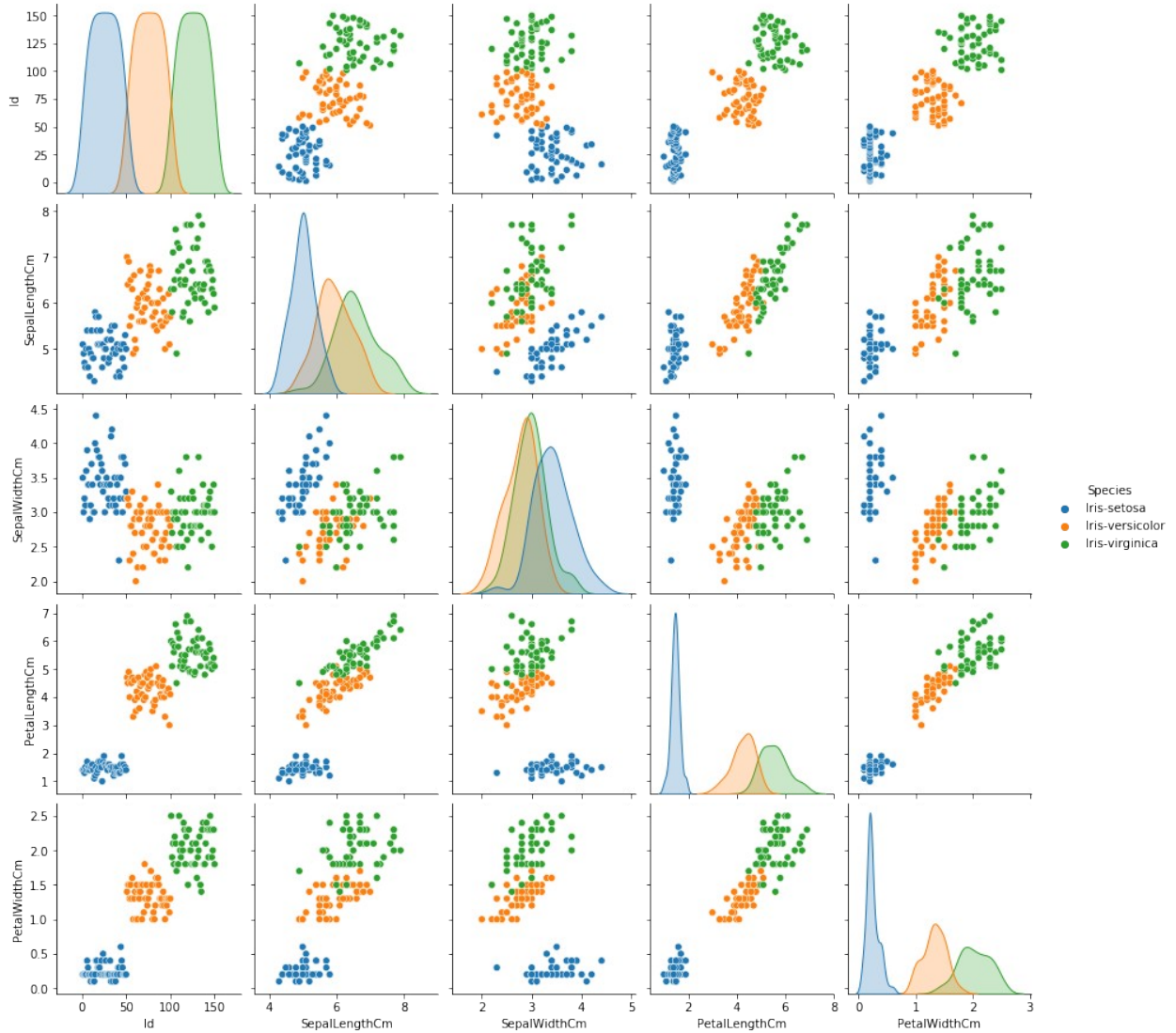
```
Iris-setosa      50
Iris-versicolor  50
Iris-virginica   50
Name: Species, dtype: int64
```

```
# visualizing target variable
sns.countplot(iris_df['Species'])
plt.show()
```



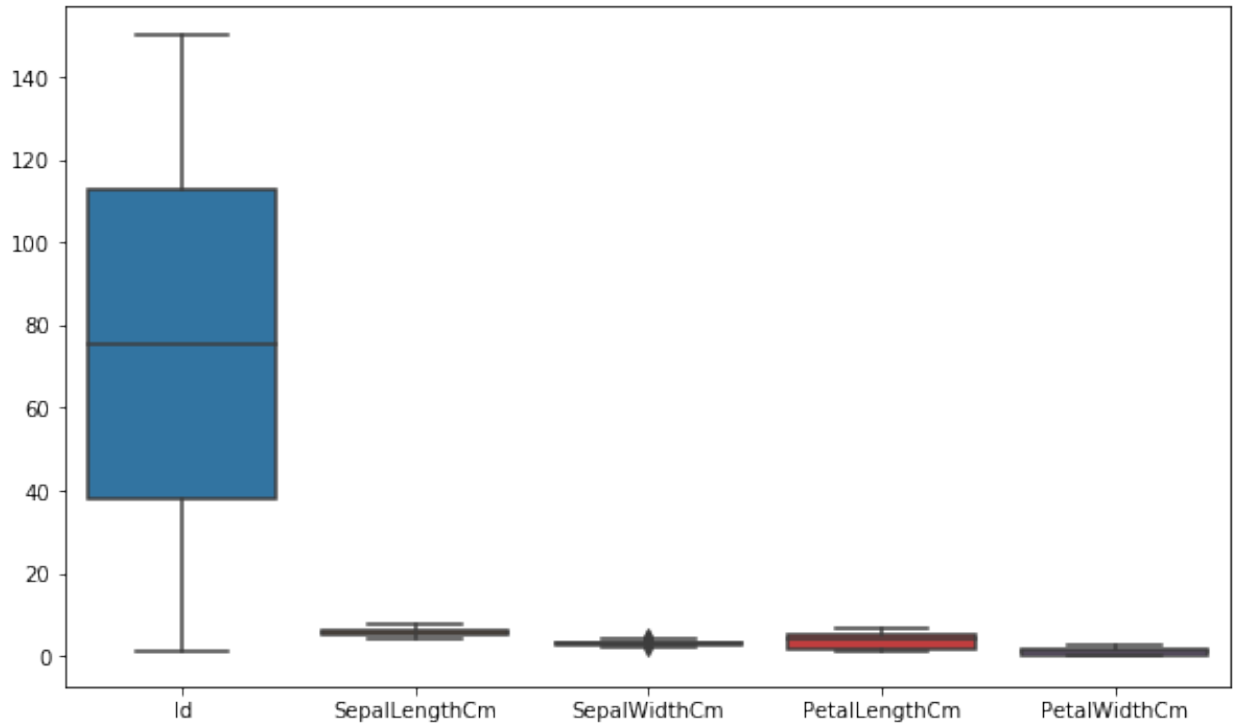
All the class labels are equal in number

```
# Pairplot to visualize relationships between variables
sns.pairplot(iris_df, hue='Species')
plt.show()
```

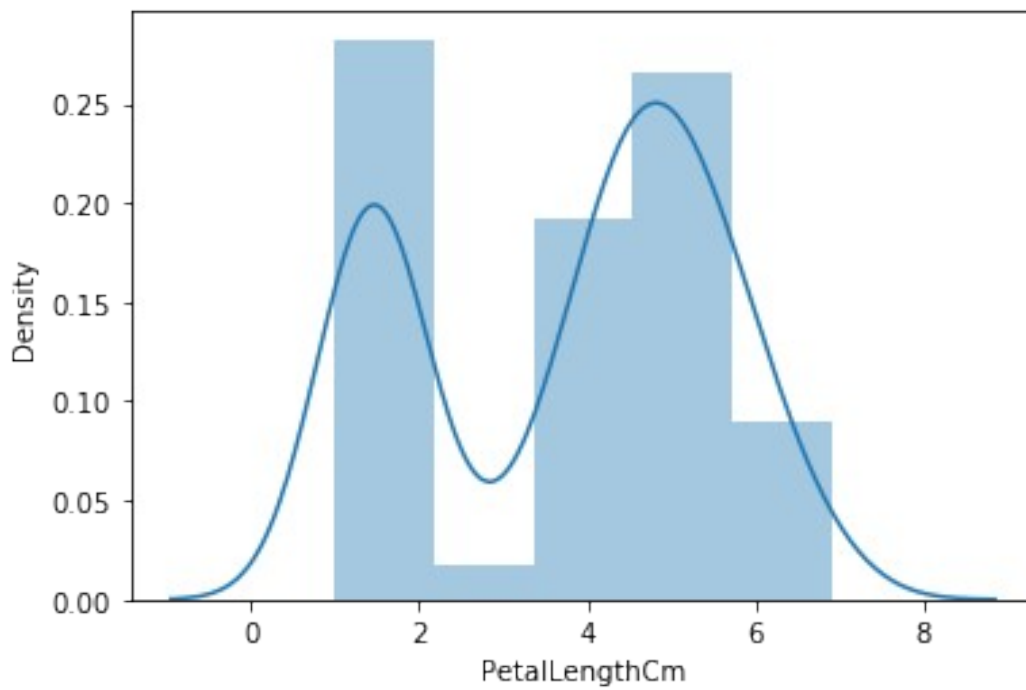


```
# Boxplot to visualize distributions of numerical variables
```

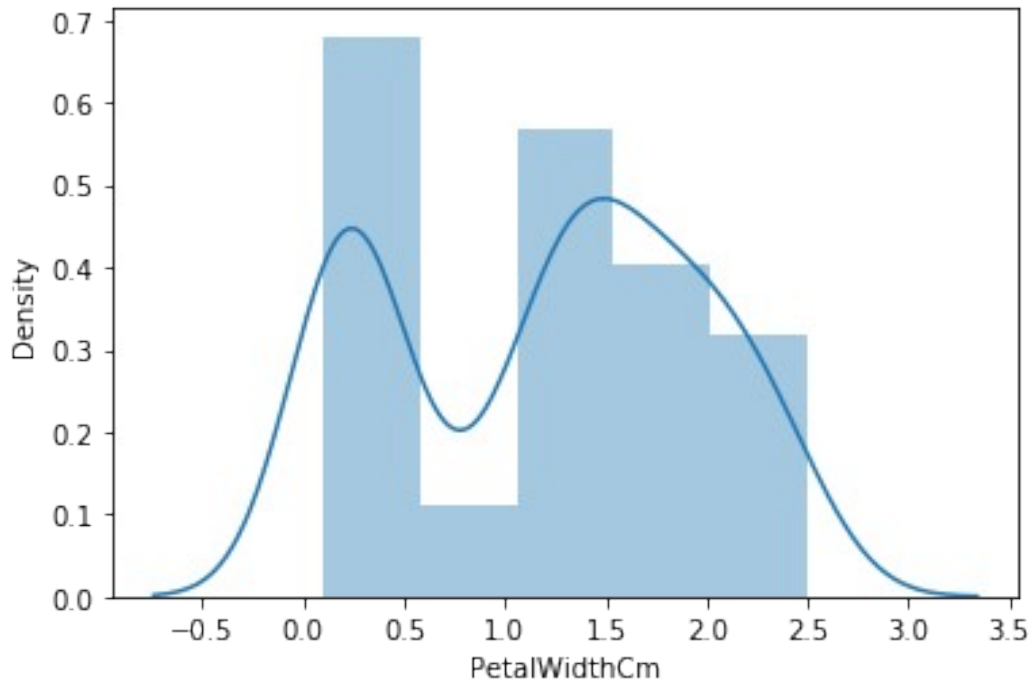
```
plt.figure(figsize=(10, 6))
sns.boxplot(data=iris_df.drop('Species', axis=1))
plt.show()
```



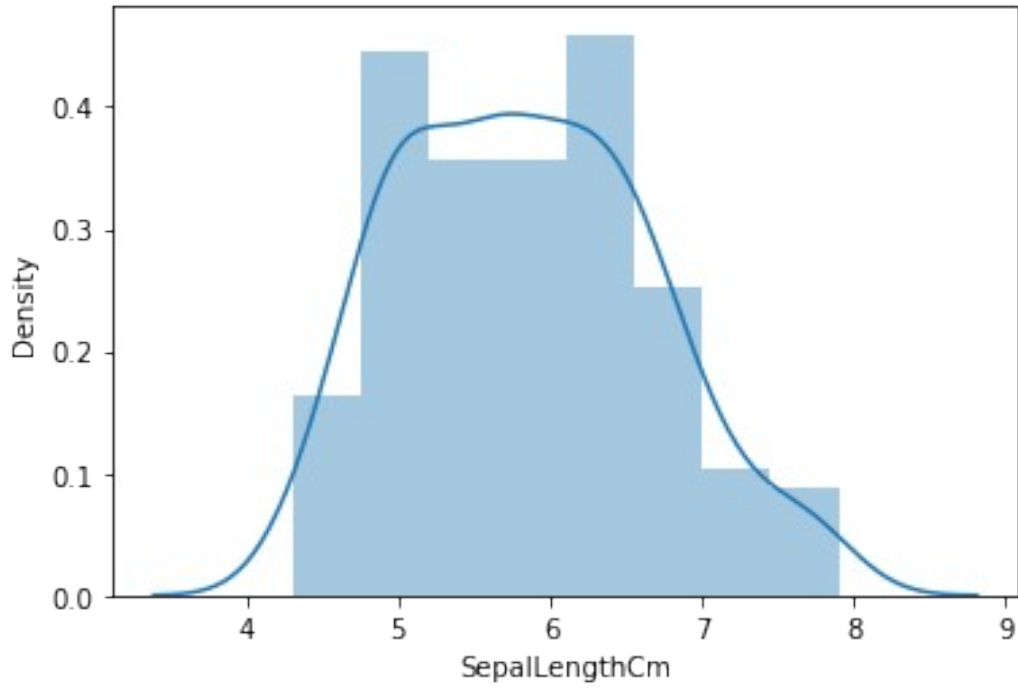
```
# Histograms for each numerical variable
import warnings
warnings.filterwarnings('ignore')
sns.distplot(iris_df['PetalLengthCm'])
plt.show()
```



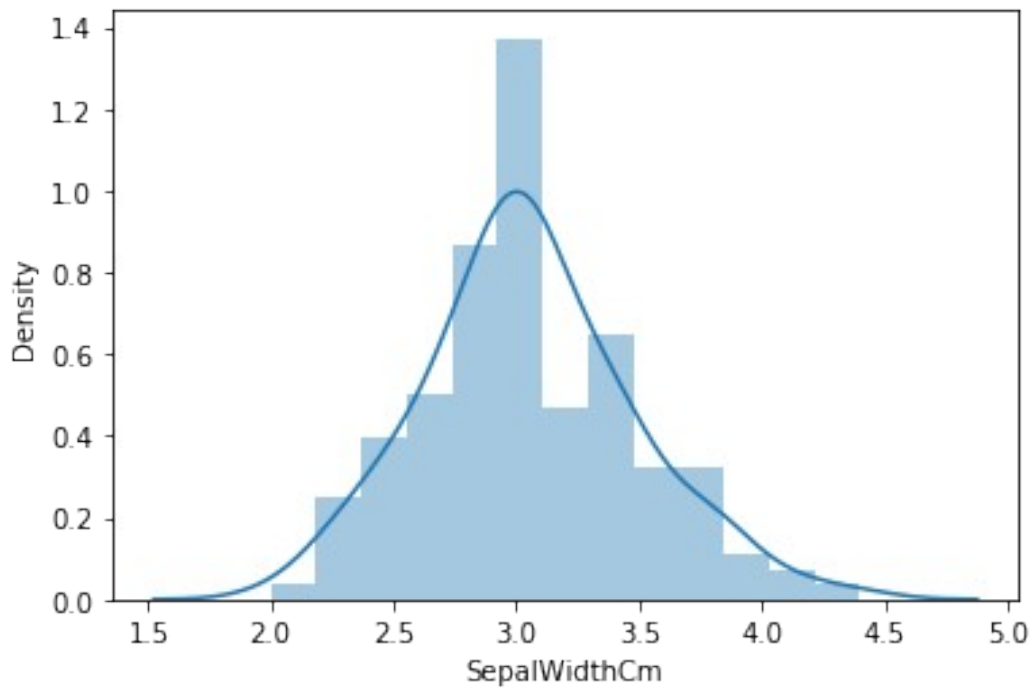
```
# Histograms for each numerical variable
import warnings
warnings.filterwarnings('ignore')
sns.distplot(iris_df['PetalWidthCm'])
plt.show()
```



```
# Histograms for each numerical variable
import warnings
warnings.filterwarnings('ignore')
sns.distplot(iris_df['SepalLengthCm'])
plt.show()
```



```
# Histograms for each numerical variable
import warnings
warnings.filterwarnings('ignore')
sns.distplot(iris_df['SepalWidthCm'])
plt.show()
```



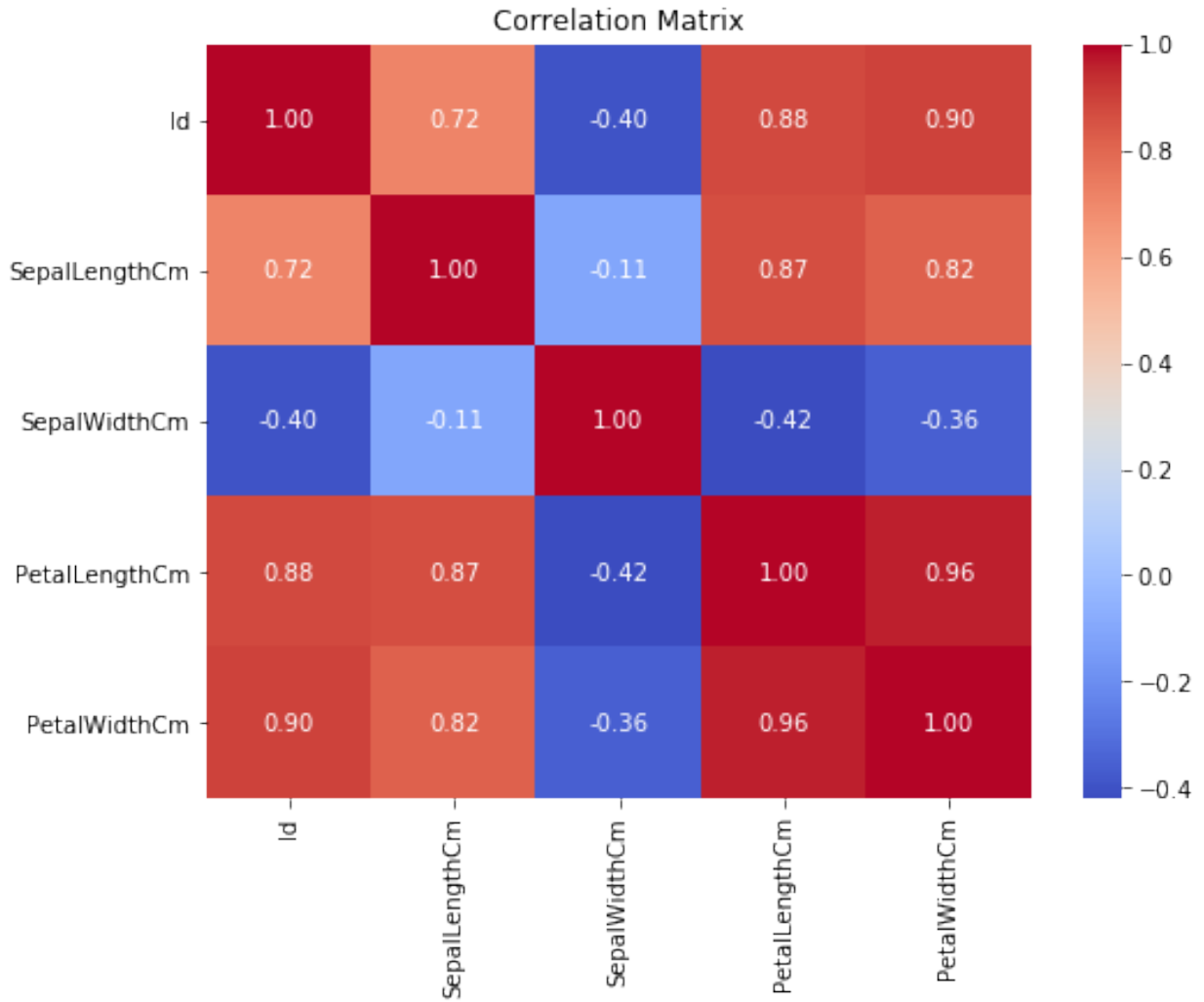
```
iris_df.corr()
```

```
      Id  SepalLengthCm  SepalWidthCm  PetalLengthCm  \
Id      1.000000      0.716676      -0.397729      0.882747
SepalLengthCm  0.716676      1.000000      -0.109369      0.871754
SepalWidthCm  -0.397729     -0.109369      1.000000     -0.420516
PetalLengthCm  0.882747      0.871754     -0.420516      1.000000
PetalWidthCm  0.899759      0.817954     -0.356544      0.962757
```

```
      PetalWidthCm
Id      0.899759
SepalLengthCm  0.817954
SepalWidthCm  -0.356544
PetalLengthCm  0.962757
PetalWidthCm  1.000000
```

```
# Correlation matrix
```

```
plt.figure(figsize=(8, 6))
sns.heatmap(iris_df.corr(), annot=True, cmap='coolwarm', fmt='.2f')
plt.title('Correlation Matrix')
plt.show()
```



There is a strong positive correlation between petal length and petal width