

```
import pandas as pd
import numpy as np
from sklearn.model_selection import train_test_split
from sklearn.tree import DecisionTreeClassifier
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import accuracy_score, precision_score, recall_score
import matplotlib.pyplot as plt
import seaborn as sns
```

Importing Libraries

Read Dataset

```
df= pd.read_csv("Iris.csv")
```

Splitting Data

```
X = df.drop("Species", axis=1)
y = df["Species"]
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
```

Decision Tree

```
tree_model = DecisionTreeClassifier()
tree_model.fit(X_train, y_train)
y_pred_tree = tree_model.predict(X_test)
```

Logistic Regression

```
log_model = LogisticRegression()
log_model.fit(X_train, y_train)
y_pred_log = log_model.predict(X_test)
```

```
/usr/local/lib/python3.10/dist-packages/sklearn/linear_model/_logistic.py:458: ConvergenceWarning: lbfgs failed to converge (status=
STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
```

Increase the number of iterations (max_iter) or scale the data as shown in:

<https://scikit-learn.org/stable/modules/preprocessing.html>

Please also refer to the documentation for alternative solver options:

https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression

```
n_iter_i = _check_optimize_result(
```

Evaluate Model

```
tree_accuracy = accuracy_score(y_test, y_pred_tree)
tree_precision = precision_score(y_test, y_pred_tree, average="weighted")
tree_recall = recall_score(y_test, y_pred_tree, average="weighted")
log_accuracy = accuracy_score(y_test, y_pred_log)
log_precision = precision_score(y_test, y_pred_log, average="weighted")
log_recall = recall_score(y_test, y_pred_log, average="weighted")
print("Decision Tree Results:")
print("Accuracy:", tree_accuracy)
print("Precision:", tree_precision)
print("Recall:", tree_recall)
print("\nLogistic Regression Results:")
print("Accuracy:", log_accuracy)
print("Precision:", log_precision)
print("Recall:", log_recall)
```

```
Decision Tree Results:
```

```
Accuracy: 1.0
```

```
Precision: 1.0
```

```
Recall: 1.0
```

```
Logistic Regression Results:
```

```
Accuracy: 1.0
```

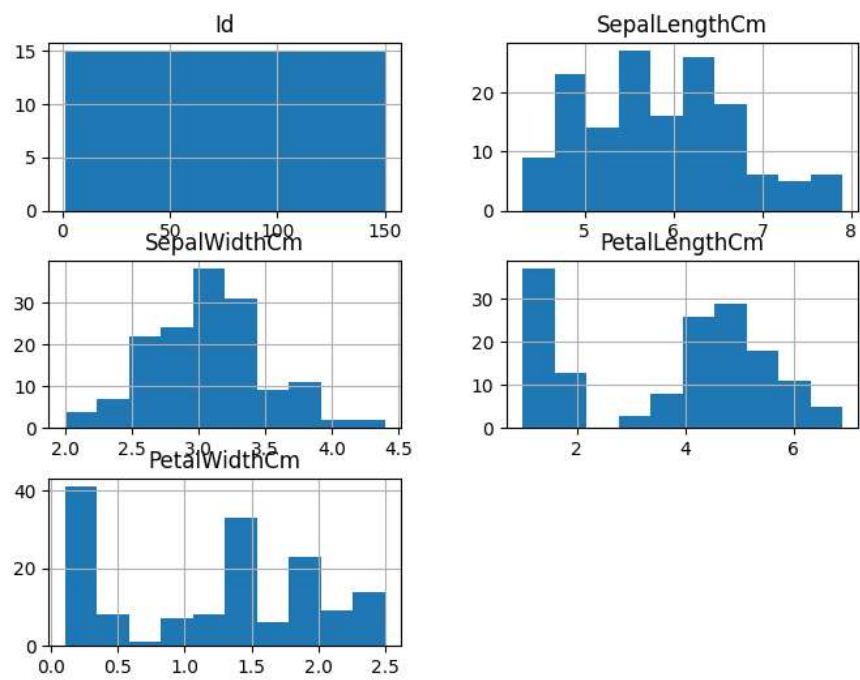
```
Precision: 1.0
```

```
Recall: 1.0
```

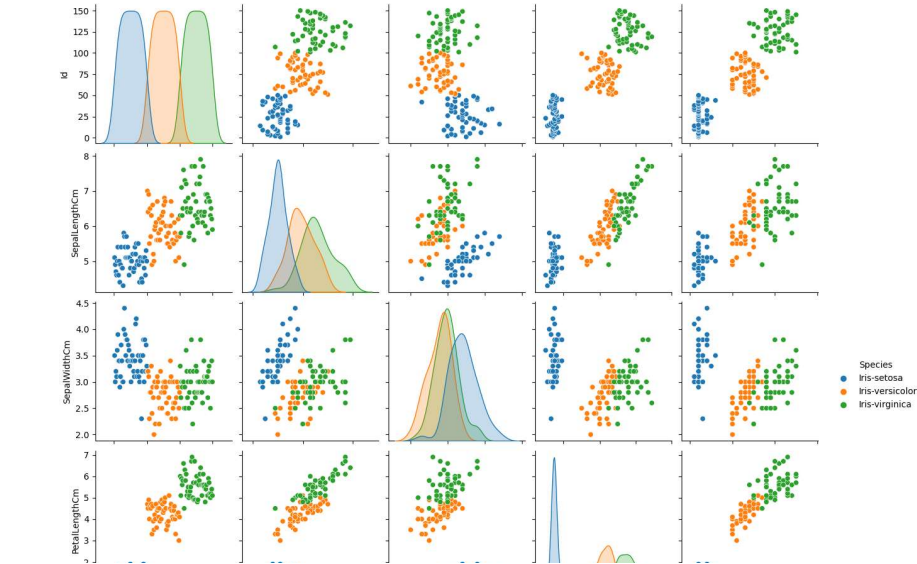
```
df.describe()
```

| | 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 |

```
df.hist(figsize=(8, 6))
plt.show()
```



```
sns.pairplot(df, hue="Species")
plt.show()
```



df.corr()

<ipython-input-11-2f6f6606aa2c>:1: FutureWarning: The default value of numeric_only i
df.corr()

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