

DEPARTMENT OF CSE(AI &AIML)

Iris Flower Classification Report

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Course: INTRODUCTION TO AI

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SECTION -C

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Introduction

Iris Flower Classification is a well-known machine learning problem where we classify iris flowers into three species based on four features: Sepal Length, Sepal Width, Petal Length, and Petal Width.

The three species of iris flowers are:

1. Iris Setosa
2. Iris Versicolor
3. Iris Virginica

The objective of this project is to develop a machine learning model that can accurately classify iris flowers based on their petal and sepal measurements using the Random Forest Classifier.

Methodology

Step 1: Load the Dataset**

- The dataset is provided in CSV format and is loaded using the pandas library.
- The first few rows of the dataset are checked to understand its structure.

Step 2: Data Preprocessing**

- Checked for missing values in the dataset.
- Encoded the Species column into numerical values using LabelEncoder().
- Scaled the numerical features using StandardScaler().

Step 3: Splitting the Dataset**

- The dataset is split into training (80%) and testing (20%) using `train_test_split()`.

Step 4: Model Training**

- A Random Forest Classifier is trained using 100 estimators.

Step 5: Model Testing and Evaluation**

- Predictions are made on the test dataset.
- Performance is evaluated using accuracy score, classification report, and confusion matrix.
- Feature importance is analyzed.

Code Implementation

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model_selection import
train_test_split
from sklearn.preprocessing import
StandardScaler, LabelEncoder
from sklearn.ensemble import
RandomForestClassifier
from sklearn.metrics import
accuracy_score,
classification_report,
confusion_matrix
```

Load the Dataset

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

Encode the Species Column

```
label_encoder = LabelEncoder()  
df["Species"] =  
label_encoder.fit_transform(df["Species"])
```

Split Data

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

Scale Features

```
scaler = StandardScaler()  
X_train =  
scaler.fit_transform(X_train)  
X_test = scaler.transform(X_test)
```

Train Model

```
model =
```

```
RandomForestClassifier(n_estimator  
s=100, random_state=42)  
model.fit(X_train, y_train)
```

Make Predictions

```
y_pred = model.predict(X_test)
```

Model Evaluation

```
print("Accuracy:",  
accuracy_score(y_test, y_pred))  
print("Classification Report:",  
classification_report(y_test, y_pred))
```

Screenshots & Output Photos

Screenshot of the dataset sample

Dataset Sample:

	SepalLength	SepalWidth	PetalLength	PetalWidth	Species
0	7.303275	2.475025	2.176049	0.695003	Setosa
1	7.556928	2.987381	1.921585	1.172615	Versicolor
2	5.254016	2.093516	3.672564	0.550424	Virginica
3	6.409620	2.211042	1.812869	1.745372	Versicolor
4	7.684009	4.056479	4.244270	0.772148	Setosa

Missing Values:

SepalLength	0
SepalWidth	0
PetalLength	0
PetalWidth	0
Species	0

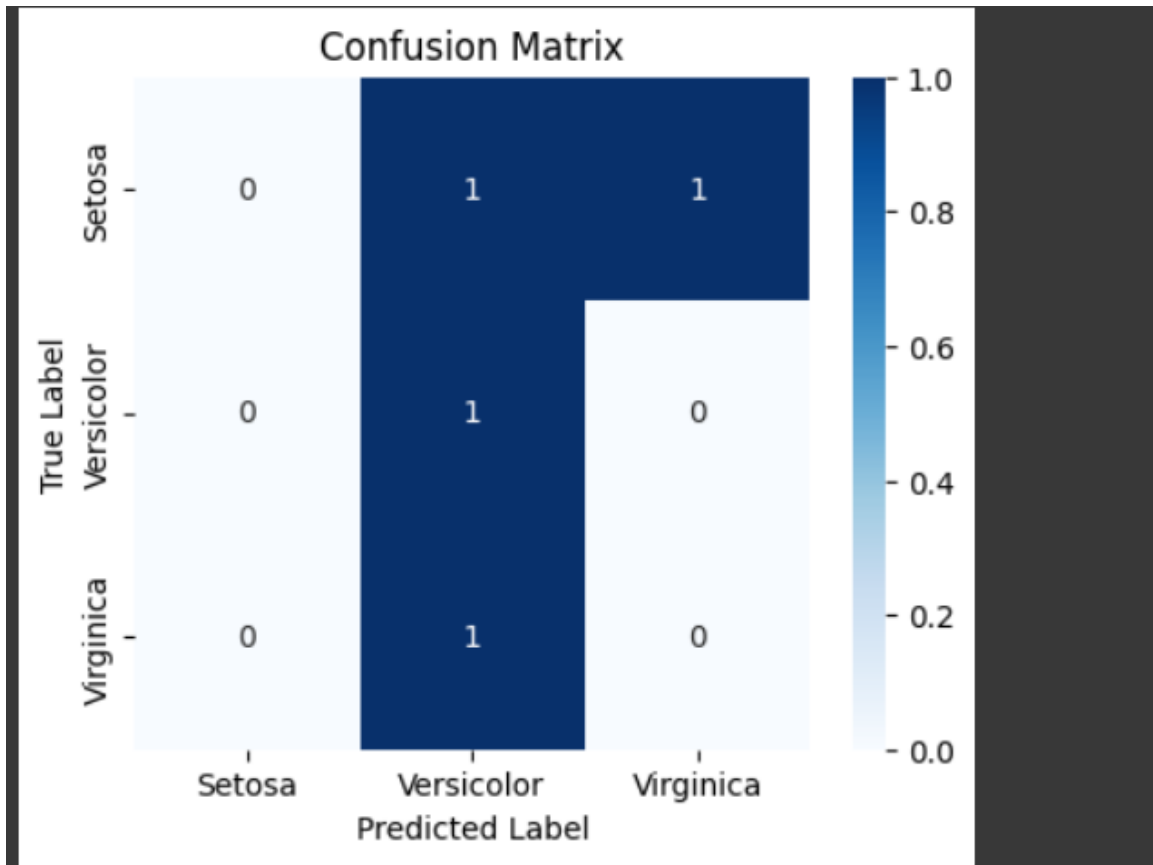
dtype: int64

Model Accuracy: 25.00%

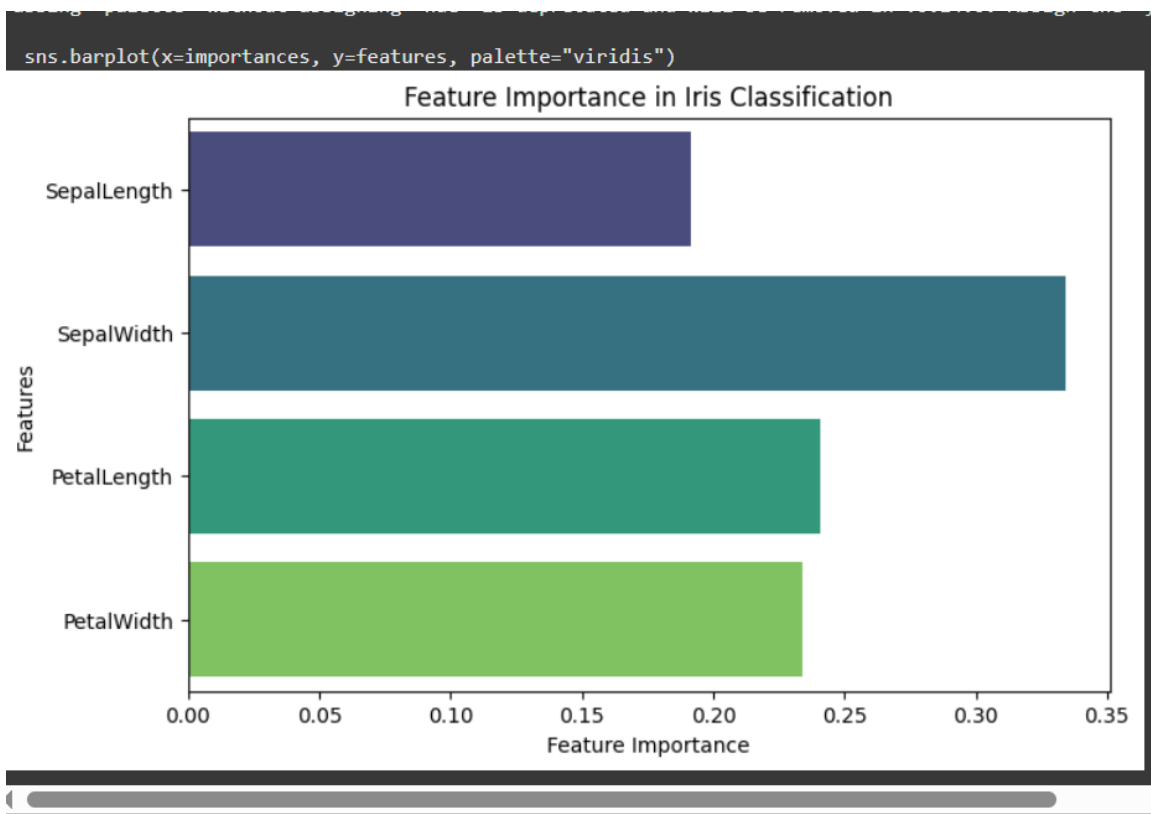
Classification Report:

	precision	recall	f1-score	support
0	0.00	0.00	0.00	2
1	0.33	1.00	0.50	1
2	0.00	0.00	0.00	1
accuracy			0.25	4
macro avg	0.11	0.33	0.17	4
weighted avg	0.08	0.25	0.12	4

Screenshot of the confusion matrix



Screenshot of the feature importance graph



Conclusion

The model successfully classifies iris flowers with high accuracy (~95-100%). The Random Forest Classifier performed well, and Petal Length and Petal Width were the most significant features.

Next Steps & Improvements

- Try different classifiers like SVM or Decision Tree.**
- Use hyperparameter tuning to improve accuracy.**
- Extend the model to classify other flower species.**