

# iris

February 19, 2025

```
[1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.datasets import load_iris
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn.svm import SVC
from sklearn.metrics import accuracy_score, confusion_matrix, \
    classification_report
```

```
[20]: df = pd.read_csv(r"C:\Users\apvis\Downloads\archive (5)\iris.csv")
print(df.head())
```

	sepal_length	sepal_width	petal_length	petal_width	species
0	5.1	3.5	1.4	0.2	Iris-setosa
1	4.9	3.0	1.4	0.2	Iris-setosa
2	4.7	3.2	1.3	0.2	Iris-setosa
3	4.6	3.1	1.5	0.2	Iris-setosa
4	5.0	3.6	1.4	0.2	Iris-setosa

```
[22]: print(df.isnull().sum())
print(df.describe())
```

sepal_length	0			
sepal_width	0			
petal_length	0			
petal_width	0			
species	0			
dtype: int64				
	sepal_length	sepal_width	petal_length	petal_width
count	150.000000	150.000000	150.000000	150.000000
mean	5.843333	3.054000	3.758667	1.198667
std	0.828066	0.433594	1.764420	0.763161
min	4.300000	2.000000	1.000000	0.100000
25%	5.100000	2.800000	1.600000	0.300000
50%	5.800000	3.000000	4.350000	1.300000
75%	6.400000	3.300000	5.100000	1.800000

max	7.900000	4.400000	6.900000	2.500000
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```
[24]: from sklearn.preprocessing import LabelEncoder
le = LabelEncoder()
df['species'] = le.fit_transform(df['species'])
print(df.head())
```

	sepal_length	sepal_width	petal_length	petal_width	species
0	5.1	3.5	1.4	0.2	0
1	4.9	3.0	1.4	0.2	0
2	4.7	3.2	1.3	0.2	0
3	4.6	3.1	1.5	0.2	0
4	5.0	3.6	1.4	0.2	0

```
[26]: 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)
print(f"Training data shape: {X_train.shape}")
print(f"Testing data shape: {X_test.shape}")
```

Training data shape: (120, 4)

Testing data shape: (30, 4)

```
[28]: scaler = StandardScaler()
X_train_scaled = scaler.fit_transform(X_train)
X_test_scaled = scaler.transform(X_test)
print("Scaled training data:")
print(X_train_scaled[:5])
```

Scaled training data:

```
[[-1.47393679  1.22037928 -1.5639872  -1.30948358]
 [-0.13307079  3.02001693 -1.27728011 -1.04292204]
 [ 1.08589829  0.09560575  0.38562104  0.28988568]
 [-1.23014297  0.77046987 -1.21993869 -1.30948358]
 [-1.7177306   0.32056046 -1.39196294 -1.30948358]]
```

```
[30]: svc = SVC(kernel='linear', random_state=42)
svc.fit(X_train_scaled, y_train)
y_pred = svc.predict(X_test_scaled)
print(f"Predicted labels: {y_pred[:5]}")
```

Predicted labels: [1 0 2 1 1]

```
[34]: species_names = df['species'].unique()
accuracy = accuracy_score(y_test, y_pred)
print(f"Model Accuracy: {accuracy:.2f}")
cm = confusion_matrix(y_test, y_pred)
```

```

print("Confusion Matrix:")
print(cm)
cr = classification_report(y_test, y_pred)
print("\nClassification Report:")
print(cr)
sns.heatmap(cm, annot=True, fmt='d', cmap='Blues', xticklabels=species_names,
            yticklabels=species_names)
plt.ylabel('Actual')
plt.xlabel('Predicted')
plt.title('Confusion Matrix')
plt.show()

```

Model Accuracy: 0.97

Confusion Matrix:

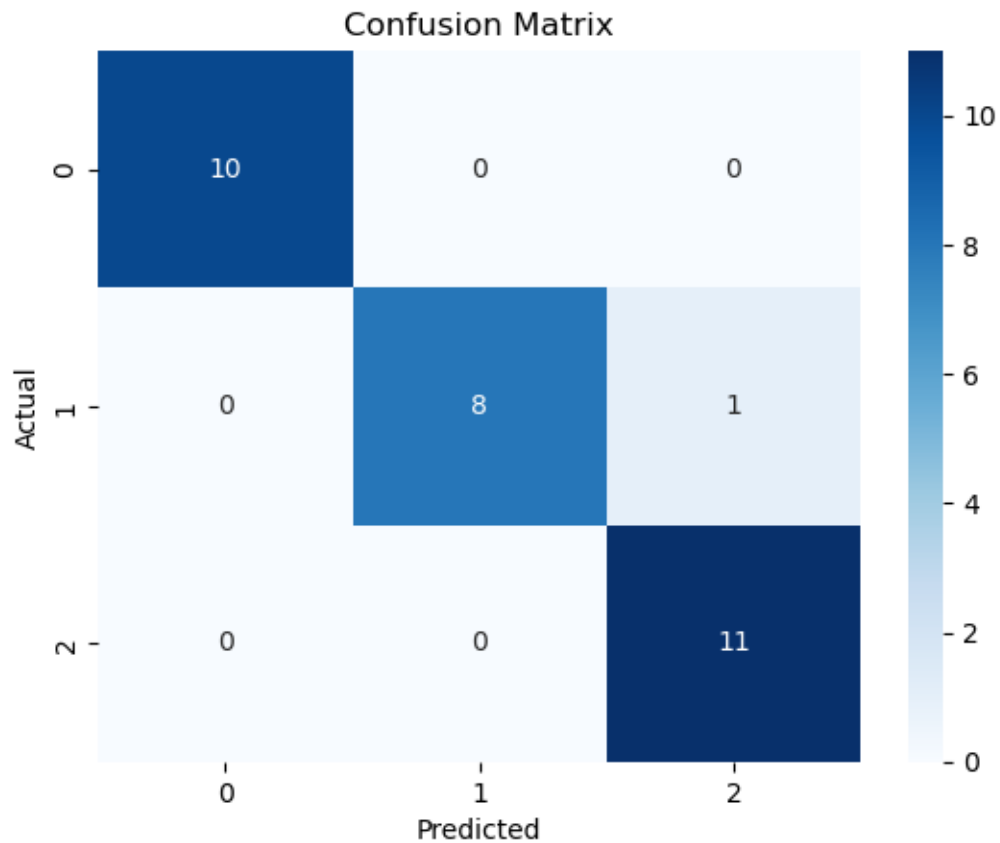
```

[[10  0  0]
 [ 0  8  1]
 [ 0  0 11]]

```

Classification Report:

	precision	recall	f1-score	support
0	1.00	1.00	1.00	10
1	1.00	0.89	0.94	9
2	0.92	1.00	0.96	11
accuracy			0.97	30
macro avg	0.97	0.96	0.97	30
weighted avg	0.97	0.97	0.97	30



```
[36]: svc_improved = SVC(C=10, kernel='rbf', random_state=42)
svc_improved.fit(X_train_scaled, y_train)
y_pred_improved = svc_improved.predict(X_test_scaled)
accuracy_improved = accuracy_score(y_test, y_pred_improved)
print(f"Improved SVC Model Accuracy: {accuracy_improved:.2f}")
```

Improved SVC Model Accuracy: 0.97