

Lab Program 10

Aim:

Implement and demonstrate classification algorithm using Support vector machine Algorithm.

Program:

Implement and demonstrate the working of SVM algorithm for classification.

Import necessary libraries:

In [2]:

```
import numpy as np
```

```
import pandas as pd
```

```
import matplotlib.pyplot as plt
```

```
from sklearn import datasets
```

```
from sklearn.model_selection import train_test_split
```

```
from sklearn.svm import SVC
```

```
from sklearn.metrics import classification_report, confusion_matrix,  
accuracy_score
```

Load and Visualize the dataset

In [10]:

```
iris.target_names
```

Out[10]:

```
array(['setosa', 'versicolor', 'virginica'], dtype='<U10')
```

In [13]:

```
# Load the Iris dataset
```

```
iris = datasets.load_iris()
```

```
X = iris.data
```

```
y = iris.target
```

```
print("Features:\n", iris.feature_names)
```

```
print("Classes:\n", iris.target_names)
```

Convert to DataFrame for better visualization

```
df = pd.DataFrame(data=np.c_[iris['data'], iris['target']],
                  columns=iris['feature_names'] + ['target'])
```

```
print("\nFirst 5 rows of the dataset:\n", df.head())
```

Plotting (optional)

```
plt.figure(figsize=(10, 6))
```

```
plt.scatter(X[:, 0], X[:, 1], c=y, cmap='viridis', edgecolor='k', s=50)
```

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

```
plt.ylabel('Sepal Width')
```

```
plt.title('Iris Data Sepal Length vs Sepal Width')
```

```
plt.show()
```

Features:

['sepal length (cm)', 'sepal width (cm)', 'petal length (cm)', 'petal width (cm)']

Classes:

['setosa' 'versicolor' 'virginica']

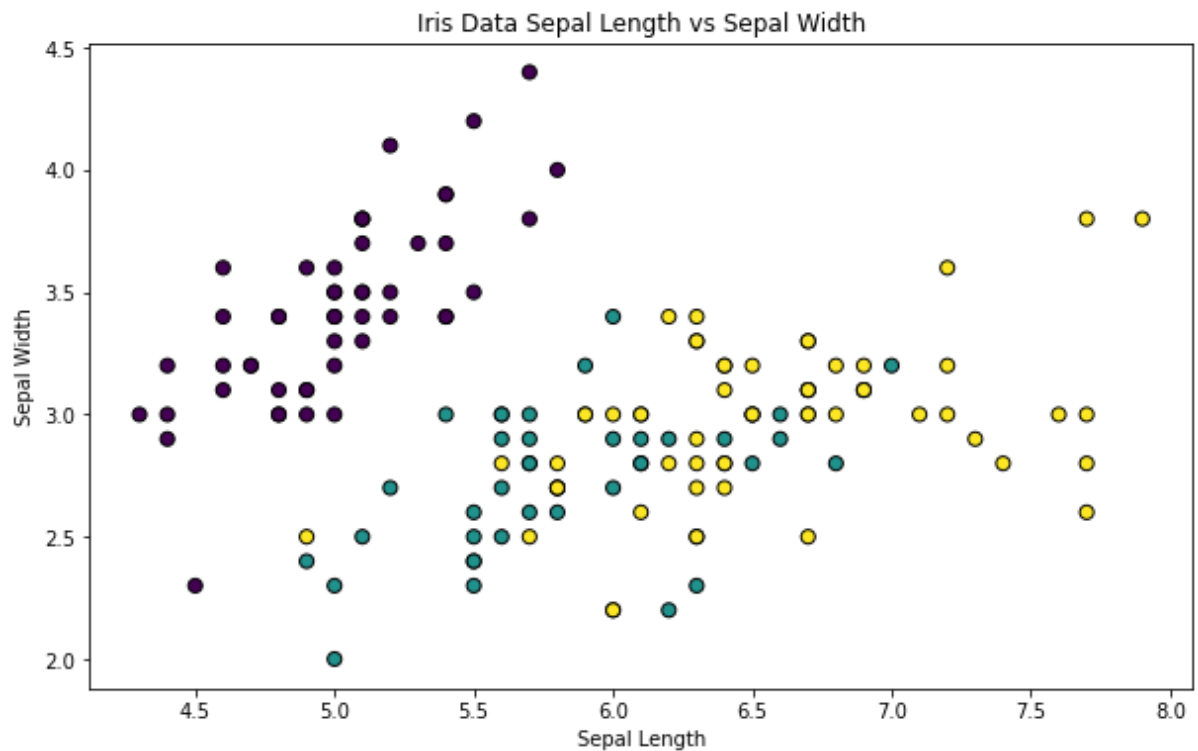
First 5 rows of the dataset:

	sepal length (cm)	sepal width (cm)	petal length (cm)	petal width (cm) \
0	5.1	3.5	1.4	0.2
1	4.9	3.0	1.4	0.2
2	4.7	3.2	1.3	0.2
3	4.6	3.1	1.5	0.2
4	5.0	3.6	1.4	0.2

target

0 0.0

```
1 0.0
2 0.0
3 0.0
4 0.0
```



Split the dataset

In [4]:

```
# Split the dataset into training and testing sets
```

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3,  
random_state=42)
```

Train the SVM classifier

In [5]:

```
# Create an SVM classifier
```

```
svm_classifier = SVC(kernel='linear', C=1.0, random_state=42)
```

```
# Train the classifier
```

```
svm_classifier.fit(X_train, y_train)
```

Out[5]:

```
SVC(kernel='linear', random_state=42)
```

Make predictions and evaluate the model:

In [6]:

```
# Make predictions
```

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

```
# Evaluate the model
```

```
print("\nConfusion Matrix:\n", confusion_matrix(y_test, y_pred))
```

```
print("\nClassification Report:\n", classification_report(y_test, y_pred))
```

```
print("\nAccuracy Score:\n", accuracy_score(y_test, y_pred))
```

Confusion Matrix:

```
[[19  0  0]
```

```
 [ 0 13  0]
```

```
 [ 0  0 13]]
```

Classification Report:

	precision	recall	f1-score	support
0	1.00	1.00	1.00	19
1	1.00	1.00	1.00	13
2	1.00	1.00	1.00	13
accuracy		1.00		45
macro avg	1.00	1.00	1.00	45
weighted avg	1.00	1.00	1.00	45

Accuracy Score:

1.0

In [15]:

df.info()

<class 'pandas.core.frame.DataFrame'>

RangeIndex: 150 entries, 0 to 149

Data columns (total 5 columns):

#	Column	Non-Null Count	Dtype
0	sepal length (cm)	150 non-null	float64
1	sepal width (cm)	150 non-null	float64
2	petal length (cm)	150 non-null	float64
3	petal width (cm)	150 non-null	float64
4	target	150 non-null	float64

dtypes: float64(5)

memory usage: 6.0 KB

In [22]:

df['target'].value_counts()

Out[22]:

2.0 50

1.0 50

0.0 50

Name: target, dtype: int64

In [25]:

svm_classifier_rbf = SVC(random_state = 42)

svm_classifier_rbf.fit(X_train, y_train)

Out[25]:

```
SVC(random_state=42)
```

In [26]:

```
# Make predictions
```

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

```
# Evaluate the model
```

```
print("\nConfusion Matrix:\n", confusion_matrix(y_test, y_pred))
```

```
print("\nClassification Report:\n", classification_report(y_test, y_pred))
```

```
print("\nAccuracy Score:\n", accuracy_score(y_test, y_pred))
```

Confusion Matrix:

```
[[19 0 0]
```

```
 [ 0 13 0]
```

```
 [ 0 0 13]]
```

Classification Report:

	precision	recall	f1-score	support
0	1.00	1.00	1.00	19
1	1.00	1.00	1.00	13
2	1.00	1.00	1.00	13
accuracy		1.00		45
macro avg	1.00	1.00	1.00	45
weighted avg	1.00	1.00	1.00	45

Accuracy Score:

1.0

In [27]:

```
svm_classifier_poly = SVC(kernel = 'poly', random_state = 42)
svm_classifier_poly.fit(X_train, y_train)
```

Out[27]:

```
SVC(kernel='poly', random_state=42)
```

In [28]:

```
# Make predictions
```

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

```
# Evaluate the model
```

```
print("\nConfusion Matrix:\n", confusion_matrix(y_test, y_pred))
```

```
print("\nClassification Report:\n", classification_report(y_test, y_pred))
```

```
print("\nAccuracy Score:\n", accuracy_score(y_test, y_pred))
```

Confusion Matrix:

```
[[19  0  0]
```

```
 [ 0 12  1]
```

```
 [ 0  0 13]]
```

Classification Report:

```
precision  recall  f1-score  support
```

```
0    1.00    1.00    1.00     19
```

```
1    1.00    0.92    0.96     13
```

```
2    0.93    1.00    0.96     13
```

accuracy		0.98		45
macro avg	0.98	0.97	0.97	45
weighted avg	0.98	0.98	0.98	45

Accuracy Score:

0.9777777777777777

In []: