

## exp6-support-vector-machines

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# Import necessary libraries
import numpy as np
import matplotlib.pyplot as plt
from sklearn.datasets import make_moons
from sklearn.model_selection import train_test_split
from sklearn.svm import SVC
from sklearn.metrics import classification_report, confusion_matrix

# Generate a non-linearly separable dataset (moons dataset)
X, y = make_moons(n_samples=300, noise=0.2, random_state=42)

# 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)

# Initialize the SVM classifier with an RBF kernel
svm_classifier = SVC(kernel='rbf', C=1, gamma='scale')

# Train the SVM classifier
svm_classifier.fit(X_train, y_train)

# Predict the labels for the test set
y_pred = svm_classifier.predict(X_test)

# Evaluate the classifier
print("Confusion Matrix:")
print(confusion_matrix(y_test, y_pred))
print("\nClassification Report:")
print(classification_report(y_test, y_pred))

# Plot the decision boundary
def plot_decision_boundary(clf, X, y):
    # Create a grid to plot the decision boundary
    x_min, x_max = X[:, 0].min() - 1, X[:, 0].max() + 1
    y_min, y_max = X[:, 1].min() - 1, X[:, 1].max() + 1
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xx, yy = np.meshgrid(np.arange(x_min, x_max, 0.01),
                     np.arange(y_min, y_max, 0.01))

# Predict the classification for each grid point
Z = clf.predict(np.c_[xx.ravel(), yy.ravel()])
Z = Z.reshape(xx.shape)

# Plot the decision boundary
plt.contourf(xx, yy, Z, alpha=0.3)
plt.scatter(X[:, 0], X[:, 1], c=y, s=30, edgecolor='k')
plt.title("SVM with RBF Kernel")
plt.xlabel("Feature 1")
plt.ylabel("Feature 2")
plt.show()

# Visualize the decision boundary
plot_decision_boundary(svm_classifier, X, y)

```

Confusion Matrix:

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[[37  1]
 [ 2 50]]

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Classification Report:

|              | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| 0            | 0.95      | 0.97   | 0.96     | 38      |
| 1            | 0.98      | 0.96   | 0.97     | 52      |
| accuracy     |           |        | 0.97     | 90      |
| macro avg    | 0.96      | 0.97   | 0.97     | 90      |
| weighted avg | 0.97      | 0.97   | 0.97     | 90      |

