## Exp No: 8

## Implement SVM/Decision tree classification techniques

## a) SVM IN R

```
# Install and load the e1071 package (if not already installed)
install.packages("e1071")
library(e1071)
# Load the iris dataset
data(iris)
# Inspect the first few rows of the dataset
head(iris)
# Split the data into training (70%) and testing (30%) sets
set.seed(123) # For reproducibility
sample indices <- sample(1:nrow(iris), 0.7 * nrow(iris))
train data <- iris[sample indices, ]
test data <- iris[-sample indices, ]
# Fit the SVM model
svm model <- svm(Species ~ ., data = train_data, kernel = "radial")
# Print the summary of the model
summary(svm model)
# Predict the test set
predictions <- predict(svm model, newdata = test data)</pre>
# Evaluate the model's performance
confusion_matrix <- table(Predicted = predictions, Actual = test_data$Species)</pre>
print(confusion matrix)
# Calculate accuracy
accuracy <- sum(diag(confusion matrix)) / sum(confusion matrix)</pre>
cat("Accuracy:", accuracy * 100, "%\n")
```

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 12 test_data <- iris[-sample_indices, ]</pre>
   13
   14 # Fit the SVM model
   15 svm_model <- svm(Species ~ ., data = train_data, kernel = "radial")</pre>
   16 summary(svm_model)
   17
   18 # Predict and evaluate
   19
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 R 4.4.1 · ~/ ≈
 > # Predict and evaluate
 > predictions <- predict(svm_model, newdata = test_data)</pre>
 > confusion_matrix <- table(Predicted = predictions, Actual = test_data$Species)</pre>
 > print(confusion_matrix)
            Actual
 Predicted
             setosa versicolor virginica
   setosa
                 14
                            0
   versicolor
                  0
                            17
                                      0
                  0
                             1
                                     13
   virginica
 > # Calculate accuracy
 > accuracy <- sum(diag(confusion_matrix)) / sum(confusion_matrix)</pre>
 > cat("Accuracy:", accuracy * 100, "%\n")
 Accuracy: 97.77778 %
 >
 >
 >
```

## b) Decision tree in R

```
# Install and load the rpart package (if not already installed)
install.packages("rpart")
library(rpart)
# Load the iris dataset
data(iris)
# Split the data into training (70%) and testing (30%) sets
set.seed(123) # For reproducibility
```

```
sample indices <- sample(1:nrow(iris), 0.7 * nrow(iris))
train data <- iris[sample indices, ]
test data <- iris[-sample indices, ]
# Fit the Decision Tree model
tree model <- rpart(Species ~ ., data = train data, method = "class")
# Print the summary of the model
summary(tree_model)
# Plot the Decision Tree
plot(tree model)
text(tree model, pretty = 0)
# Predict the test set
predictions <- predict(tree model, newdata = test data, type = "class")</pre>
# Evaluate the model's performance
confusion matrix <- table(Predicted = predictions, Actual = test_data$Species)
print(confusion matrix)
# Calculate accuracy
accuracy <- sum(diag(confusion matrix)) / sum(confusion matrix)
cat("Accuracy:", accuracy * 100, "%\n")
```



