# Exp.No: 8

# Implement SVM/Decision tree classification techniques

# AIM:

To Implement SVM/Decision tree classification techniques using R.

## **PROCEDURE:**

- Load the dataset from sources such as CSV files or databases using suitable libraries.
- Perform data cleaning and preprocessing, including addressing missing values and encoding categorical variables.
- Divide the dataset into training and testing sets to effectively assess model performance.
- Normalize or standardize the features, particularly for SVM, to ensure uniform scaling across inputs.
- Select the appropriate model: use SVM for margin-based classification or Decision Tree for rule-based classification tasks.
- Train the model on the training data using the fit method.
- Generate predictions on the testing data with the predict method.
- Evaluate the model's performance using metrics such as accuracy, confusion matrix, precision, and recall.
- Visualize the outcomes with relevant plots, like decision boundaries for SVM or tree structures for Decision Trees.
- Tune the model by adjusting hyperparameters, such as C for SVM or max\_depth for Decision Trees.

### **PROGRAM:**

# SVM.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))</pre>
train_data <- iris[sample_indices, ]</pre>
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)
print(confusion matrix)
# Calculate accuracy
accuracy <- sum(diag(confusion matrix)) / sum(confusion matrix)
cat("Accuracy:", accuracy * 100, "%\n")
```

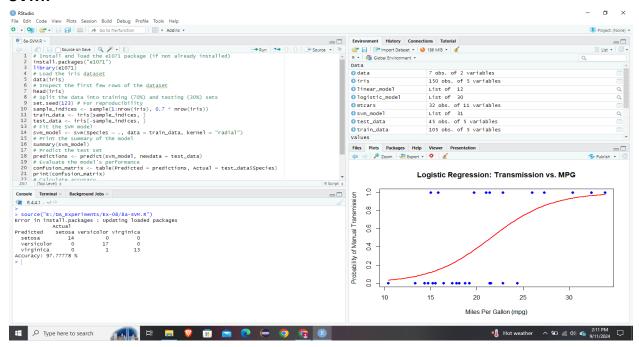
#### **Decision Tree.R:**

```
# 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, ]</pre>
```

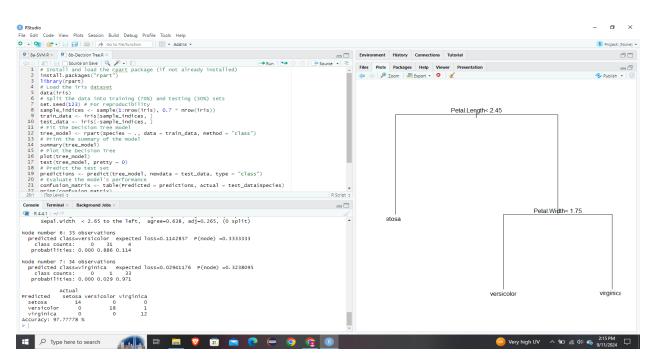
```
test data <- iris[-sample indices, ]
# Fit the Decision Tree model
tree_model <- rpart(Species ~ ., data = train_data, method = "class")
# Print the summary of the model
summary(tree_model)
# Plot the Decision Tree
plot(tree model)
text(tree_model, pretty = 0)
# Predict the test set
predictions 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

# **OUTPUT:**

## SVM:



## **Decision tree:**



## **RESULT:**

Thus SVM and Decision tree classification techniques has been successfully executed.