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Im	Implement SVM/Decision tree classification techniques		
AIM: To Implement SVM/Decision tree classification techniques using R.			

PROCEDURE:

- Collect and load the dataset from sources like CSV files or databases.
- Clean and preprocess the data, including handling missing values and encoding categorical variables.
- Split the dataset into training and testing sets to evaluate model performance.
- Normalize or standardize the features, especially for SVM, to ensure consistent scaling.
- Choose the appropriate model: SVM for margin-based classification, Decision Tree for rule-based classification.
- Train the model on the training data using the 'fit' method.
- Make predictions on the testing data using the 'predict' method.
- Evaluate the model using metrics like accuracy, confusion matrix, precision, and recall.
- Visualize the results with plots, such as decision boundaries for SVM or tree structures for Decision Trees.
- Fine-tune the model by adjusting hyperparameters like 'C' for SVM or 'max depth' for Decision Trees.

CODE:

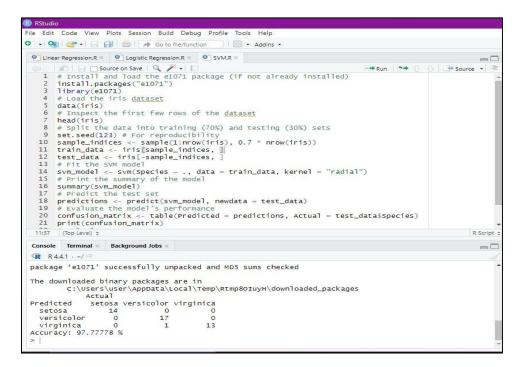
SVM.R:

```
# 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))
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)
print(confusion matrix)
# Calculate accuracy
accuracy <- sum(diag(confusion matrix)) / sum(confusion matrix)
cat("Accuracy:", accuracy * 100, "%\n")
Decision Tree.R:
```

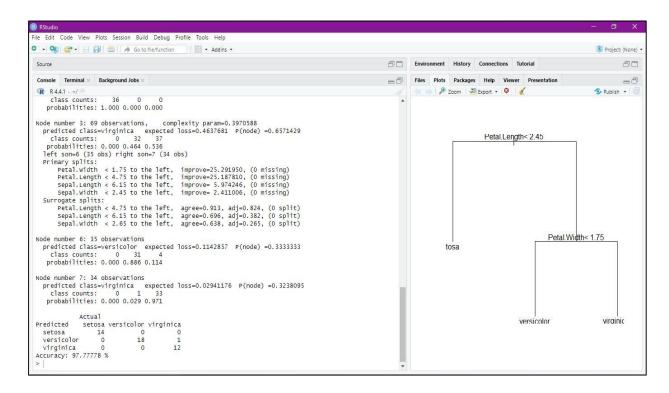
```
# Install and load the rpart package (if not already installed)
install.packages("rpart")
library(rpart)
# Load the iris dataset
data(iris)
# Split the data into training (70%) and testing (30%) sets
set.seed(123) # For reproducibility
sample indices <- sample(1:nrow(iris), 0.7 * nrow(iris))
train data <- iris[sample indices, ]
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")
```

OUTPUT:

SVM in R:



Decision tree:



RESULT:

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