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))</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)</pre>
print(confusion_matrix)
# Calculate accuracy
accuracy <- sum(diag(confusion_matrix)) / sum(confusion_matrix)</pre>
cat("Accuracy:", accuracy * 100, "%\n")
```

Output:

```
> printe(contraston_macritx)
          Actual
Predicted setosa versicolor virginica
 setosa
                14
                            0
 versicolor
                 0
                            17
                                       0
 virginica
                 0
                            1
                                      13
> # 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))</pre>
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)</pre>
```

```
print(confusion_matrix)
# Calculate accuracy
accuracy <- sum(diag(confusion_matrix)) / sum(confusion_matrix)
cat("Accuracy:", accuracy * 100, "%\n")</pre>
```

Output:

```
- 1 100 che beeteten 1100
> 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)
           Actual
redicted
           setosa versicolor virginica
 setosa
                 14
                            0
                 0
 versicolor
                            18
                                        1
 virginica
                 0
                             0
                                      12
```

