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Exp8:

Implement SVM/Decision tree classification techniques

Aim: To implement SVM/Decision Tree classification techniques in RStudio using R language.

PROCEDURE:

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
# 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")
```

```
7 sample_indices <- sample(1:nrow(iris), 0.7 * nrow(iris))
8 train_data <- iris[sample_indices, ]
9 test_data <- iris[-sample_indices, ]
10 # Fit the SVM model
11 svm_model <- svm(Species ~ ., data = train_data, kernel = "radial")
12 # Print the summary of the model</pre>
                                                                                                                                                            ▲ R • Global Environment •
                                                                                                                                                                  0 data
                                                                                                                                                                                                         7 obs. of 2 variables
                                                                                                                                                                                                        150 obs. of 5 variables
List of 12
                                                                                                                                                                  0 iris
                                                                                                                                                                  ① linear_model
          # First the summary of the model's
summary(sym_model)
# Predict the test set
predictions <- predict(sym_model, newdata = test_data)
# Evaluate the model's performance</pre>
                                                                                                                                                                  0 logistic_model
                                                                                                                                                                                                         List of 30
                                                                                                                                                                                                        32 obs. of 11 variables
List of 31
                                                                                                                                                                  ① mtcars
                                                                                                                                                                  Osvm_model
          # Evaluate the model's performance
confusion_matrix <- table(Predicted = predictions, Actual = test_data$Specie
print(confusion_matrix)
                                                                                                                                                                  ① test_data
                                                                                                                                                                  O train_data 105 obs. of 5 variables
              Calculate accuracy
          \label{eq:accuracy} \begin{split} & accuracy <- \mbox{sum(diag(confusion\_matrix))} \ / \mbox{sum(confusion\_matrix)} \\ & cat("Accuracy:", accuracy <math>\mbox{$^{\circ}$} \mbox{100, "}\mbox{$\%$}\mbox{$n$"}) \end{split}
                                                                                                                                                                  Values
                                                                                                                                                                      confusion_matrix 'table' int [1:3, 1:3] 14 0 0 0 17 1 0 0 13
                                                                                                                                              R Script $
                                                                                                                                                                     heights
predicted_probs
                                                                                                                                                                                                        num [1:7] 150 160 165 170 175 180 185
   22:1
                                                                                                                                                                                                        Named num [1:32] 0.461 0.461 0.598 0.492 0.297...
                                                                                                                                                                                                         Factor w/ 3 levels "setosa", "versicolor", ...: 1....
int [1:105] 14 50 118 43 150 148 90 91 143 92 ...
 Console Terminal × Background Jobs ×
                                                                                                                                                                      predictions
                                                                                                                                                                      sample_indices
 R 4.4.1 · ~/
 > accuracy <- sum(diag(confusion_matrix)) / sum(confusion_matrix)
                                                                                                                                                                                                         num [1:7] 55 60 62 68 70 75 80
> accuracy <- swm(diag(confusion_matrix))
Error: object 'confusion_matrix' not found
> cat("accuracy:", accuracy * 100, "%\n")
Error: object 'accuracy' not found
> source("-/.active-rstudio-document")
Actual
Predicted
setosa versicolor virginica
setosa 14 0 0
versicolor 0 17 0
virginica 0 17 0
virginica 0 13
virginica 0
Accuracy: 97.77778 %
```

b) Decision tree in R

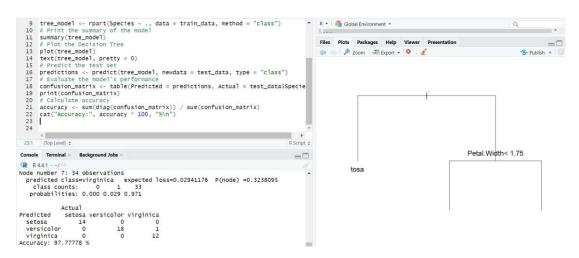
```
# Install and load the rpart package (if not already installed)
install.packages("rpart") library(rpart)
# Load the iris dataset data(iris)
# Split the data into training (70%) and testing (30%) sets set.seed(123)
# For reproducibility sample indices <- sample(1:nrow(iris), 0.7
* nrow(iris)) train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

Evaluate the model's performance

confusion_matrix <- table(Predicted = predictions, Actual = test_data\$Species)
print(confusion_matrix)</pre>

Calculate accuracy

accuracy <- sum(diag(confusion_matrix)) / sum(confusion_matrix) cat("Accuracy:", accuracy * 100, "%\n")



Result: Thus SVM and Decision Tree techniques are implemented in RStudio using R language.

