

EX. NO: 7

INSTALLATION GUIDE FOR R AND R STUDIO

Step 1 – Install R

1. Download the R installer from <https://cran.r-project.org/>



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The Comprehensive R Archive Network

Download and Install R

Precompiled binary distributions of the base system and contributed packages, **Windows and Mac** users most likely want one of these versions of R:

- [Download R for Linux](#)
- [Download R for \(Mac\) OS X](#)
- [Download R for Windows](#)

R is part of many Linux distributions, you should check with your Linux package management system in addition to the link above.

Source Code for all Platforms

Windows and Mac users most likely want to download the precompiled binaries listed in the upper box, not the source code. The sources have to be compiled before you can use them. If you do not know what this means, you probably do not want to do it!

- The latest release (2015-06-18, World-Famous Astronaut) [R-3.2.1.tar.gz](#), read [what's new](#) in the latest version.
- Sources of [R alpha and beta releases](#) (daily snapshots, created only in time periods before a planned release).
- Daily snapshots of current patched and development versions are [available here](#). Please read about [new features and bug fixes](#) before filing corresponding feature requests or bug reports.
- Source code of older versions of R is [available here](#).
- Contributed extension [packages](#)

Questions About R

- If you have questions about R like how to download and install the software, or what the license terms are, please read our [answers to frequently asked questions](#) before you send an email.

What are R and CRAN?

Figure 1. Screenshot of <http://cran.csiro.au/>

2. Run the installer. Default settings are fine. If you do not have admin rights on your laptop, then ask you local IT support. In that case, it is important that you also ask them to give you full permissions to the R directories. Without this, you will not be able to install additional packages later

Step 2 – Install RStudio

1. Download RStudio: <https://www.rstudio.com/products/rstudio/download/>

Do you need support or a commercial license? Check out our commercial offerings

RStudio Desktop 0.99.467 — Release Notes

RStudio requires R 2.11.1 (or higher). If you don't already have R, you can download it here.

Share your R code on the web with Shiny
Click here to learn more

Installers for Supported Platforms

Installers	Size	Date	MD5
RStudio 0.99.467 - Windows Vista/7/8	73.9 MB	2015-07-15	5c0bf6987adcfb6dd441326ecc67f6e8
RStudio 0.99.467 - Mac OS X 10.6+ (64-bit)	56.2 MB	2015-07-15	3116a0f3b9b3779b9531e9b08c394558
RStudio 0.99.467 - Ubuntu 12.04+/Debian 8+ (32-bit)	77.4 MB	2015-07-15	0ca919255495cc87112df12a1cf7e29
RStudio 0.99.467 - Ubuntu 12.04+/Debian 8+ (64-bit)	83.9 MB	2015-07-15	dd64fc165de55a0be229f2362cd776da
RStudio 0.99.467 - Fedora 19+/RedHat 7+/openSUSE 13.1+ (32-bit)	76.8 MB	2015-07-15	1e152bafa8b6c5355a2ec0f6822abddf
RStudio 0.99.467 - Fedora 19+/RedHat 7+/openSUSE 13.1+ (64-bit)	77.6 MB	2015-07-15	a82a27b113184e1790ec5bd3c36e2137

Zip/Tarballs

Zip/tar archives	Size	Date	MD5
RStudio 0.99.467 - Windows Vista/7/8	105.5 MB	2015-07-15	b88bbbc042aaa04b83a1786322981c5f
RStudio 0.99.467 - Ubuntu 12.04+/Debian 8+ (32-bit)	78.1 MB	2015-07-15	171de4b2e08edd33988ab1b75d571ae6
RStudio 0.99.467 - Ubuntu 12.04+/Debian 8+ (64-bit)	84.8 MB	2015-07-15	85ca4bc15c2e5e535594bcf586604324
RStudio 0.99.467 - Fedora 19+/RedHat 7+/openSUSE 13.1+ (32-bit)	77.4 MB	2015-07-15	6abc51b45cdd1fa0c14a991b5336a8ad
RStudio 0.99.467 - Fedora 19+/RedHat 7+/openSUSE 13.1+ (64-bit)	78.4 MB	2015-07-15	9d693e5e5ed1b2003e1b18691f31574b

Source Code

A tarball containing source code for RStudio v0.99.467 can be downloaded from [here](#)

Figure 2. Download RStudio on <https://www.rstudio.com/products/rstudio/download/>

- Once the installation of R has completed successfully (and not before), run the RStudio installer.
- If you do not have administrative rights on your laptop, step 2 may fail. Ask your IT Support or download a pre--built zip archive of RStudio which doesn't need installing. The link for this is towards the bottom of the download page, highlighted in Image 2.
 - Download the appropriate archive for your system (Windows/Linux only – the Mac version can be installed into your personal “Applications” folder without admin rights).
 - Double clicking on the zip archive should automatically unpack it on most Windows machines.

Step 3 – Check that R and RStudio are working

- Open RStudio. It should open a window that looks similar to image 3 below.
- In the left hand window, by the '>' sign, type '4+5'(without the quotes) and hit enter. An output line reading '[1] 9' should appear. This means that R and RStudio are working.
- If this is not successful, contact us or your local IT support for further advice

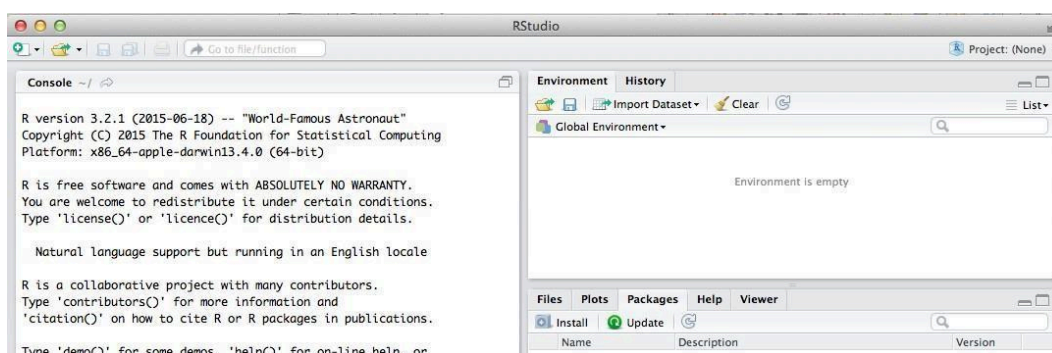


Figure 3. Running R with RStudio

Step 4 – Install R packages required for the workshop

1. Click on the tab 'Packages' then 'Install' as shown in Image 4. Or Tools --> Install packages.
2. Install the following packages: mixOmics **version 6.1.0**, mvtnorm, RColorBrewer, corrplot, igraph (see Image 4). For apple mac users, if you are unable to install the mixOmics imported library rgl, you will need to install the XQuartz software first <https://www.xquartz.org/>
3. Check that the packages are installed by typing 'library(mixOmics)' (without the quotes) in the prompt and press enter (see Image 5).
4. Then type 'sessionInfo()' and check that mixOmics version 6.1.0 has been installed (image 6).

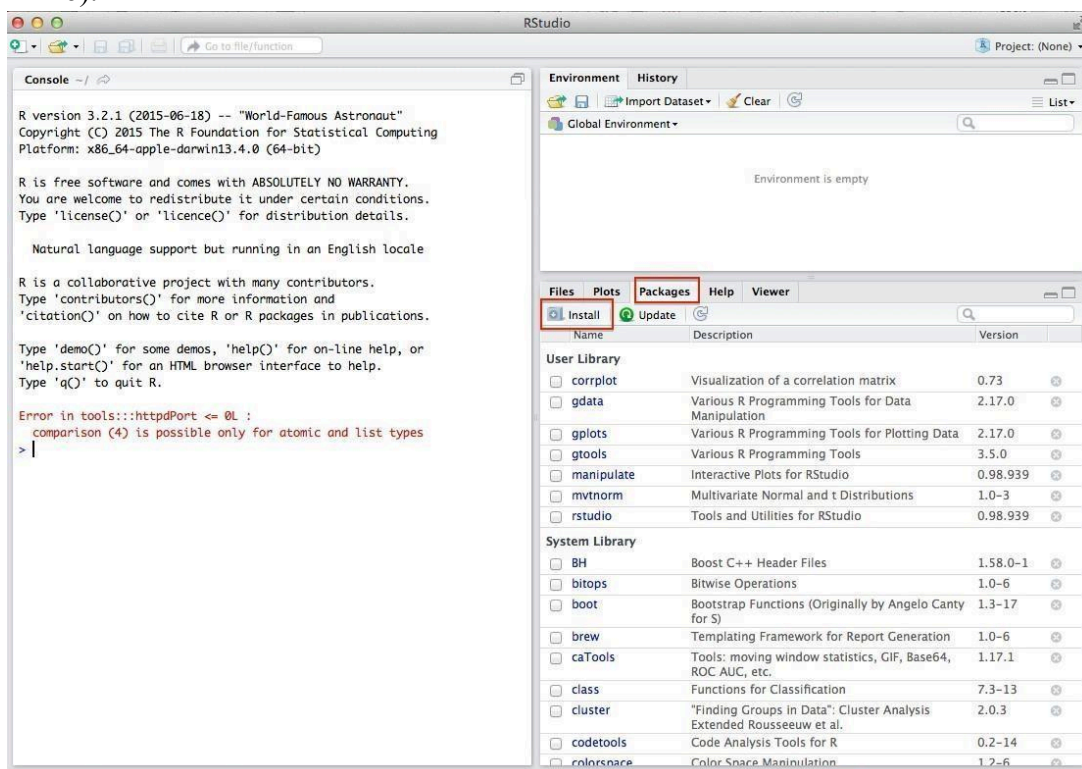


Figure 4. Click on Install to install R packages.

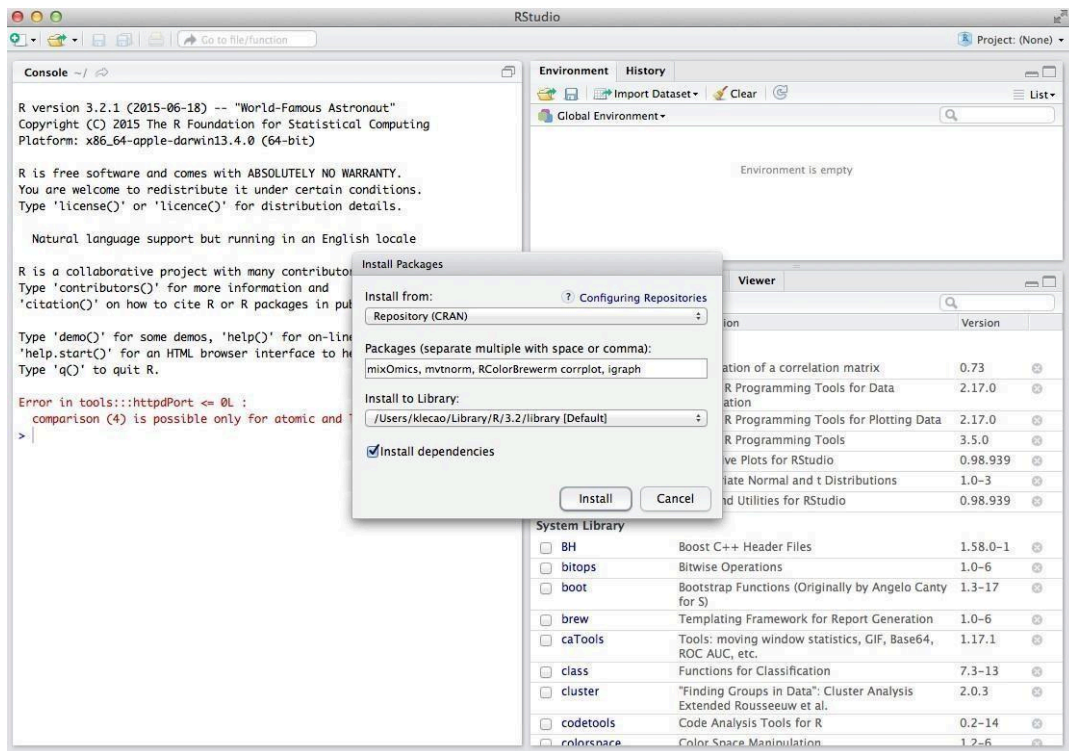


Figure 5. Specify the list of packages to be installed

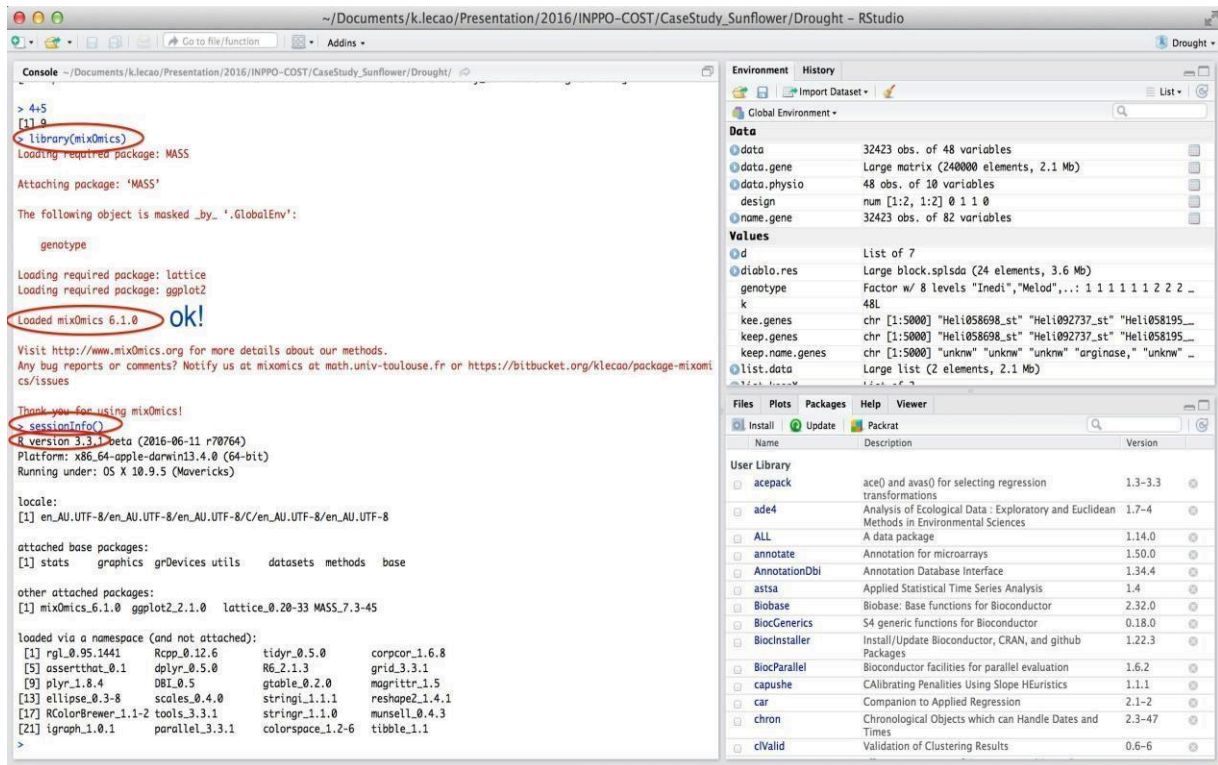


Figure 6. Check that the package mixOmics is installed and has the version 6.1.0.

RESULT: Thus to install R and R Studio is successfully completed.

EX .NO:7a**IMPLEMENT LINEAR AND LOGISTIC REGRESSION****AIM :**

To implement Linear and Logistic regression using R language.

a)Linear regression

```
# Sample data
heights <- c(150, 160, 165, 170, 175, 180, 185)
weights <- c(55, 60, 62, 68, 70, 75, 80)
```

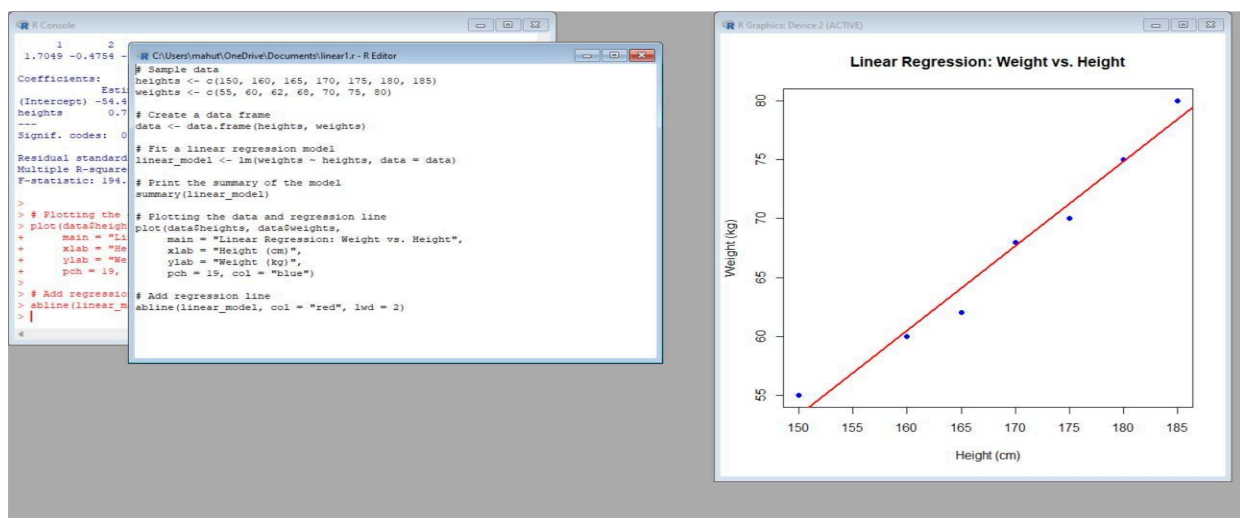
```
# Create a data frame data <-
data.frame(heights, weights)
```

```
# Fit a linear regression model linear_model <-
lm(weights ~ heights, data = data)
```

```
# Print the summary of the model
print(summary(linear_model))
```

```
# Plotting the data and regression line
plot(data$heights, data$weights,
     main = "Linear Regression: Weight vs.
Height", xlab = "Height (cm)", ylab =
"Weight (kg)",
     pch = 19, col = "blue")
```

```
# Add regression line
abline(linear_model, col = "red", lwd = 2)
```



b) Logistic regression

```
# Load the dataset
```

```
data(mtcars)
```

```
# Convert 'am' to a factor (categorical variable) mtcars$am <- factor(mtcars$am,
levels = c(0, 1), labels = c("Automatic", "Manual"))
```

```
# Fit a logistic regression model logistic_model <- glm(am ~
mpg, data = mtcars, family = binomial)
```

```
# Print the summary of the model
```

```
print(summary(logistic_model))
```

```
# Predict probabilities for the logistic model
```

```
predicted_probs <- predict(logistic_model, type =
"response")
```

```
# Display the predicted probabilities
```

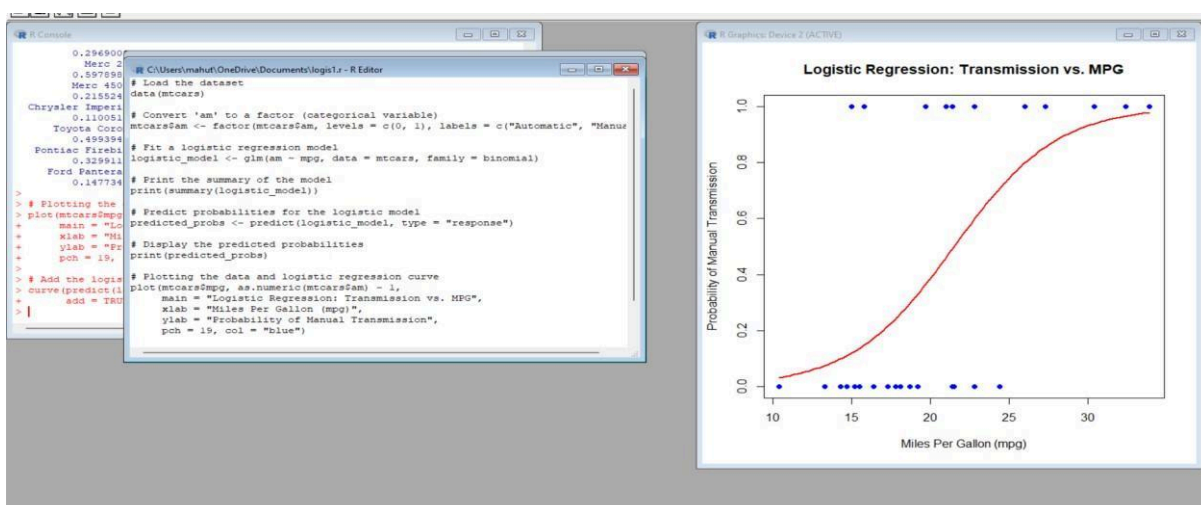
```
print(predicted_probs)
```

```
# Plotting the data and logistic regression curve
```

```
plot(mtcars$mpg, as.numeric(mtcars$am) - 1, main =
"Logistic Regression: Transmission vs. MPG", xlab =
"Miles Per Gallon (mpg)", ylab = "Probability of
Manual Transmission", pch = 19, col = "blue")
```

```
# Add the logistic regression curve
```

```
curve(predict(logistic_model, data.frame(mpg = x), type = "response"),
add = TRUE, col = "red", lwd = 2)
```



RESULT: Thus to implement linear and logistic regression using R language is successfully done.

Ex.No:8

Implement SVM/Decision tree classification techniques

AIM :

to implement SVM and Decision tree classification techniques using R language.

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, ]
```

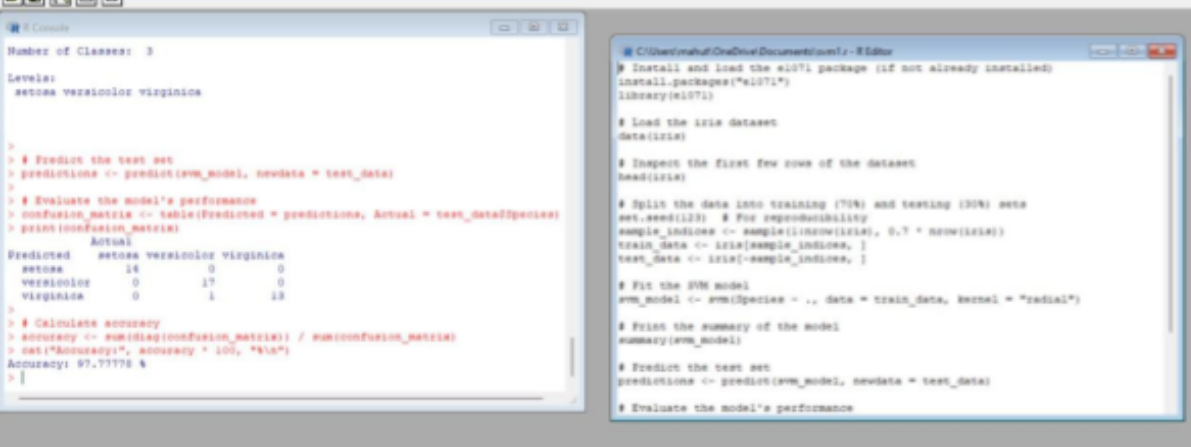
```
# Fit the SVM model svm_model <- svm(Species ~ ., data =
train_data, kernel = "radial")
```

```
# 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")
```

The screenshot shows two windows from the R Studio interface. The left window is the R Console, displaying the output of several R commands. It shows the number of classes (3), the levels (setosa, versicolor, virginica), a confusion matrix, and the calculated accuracy (97.7778%). The right window is the R Editor, showing a script of R code that performs the same steps: installing the e1071 package, loading the iris dataset, splitting it into training and testing sets, fitting an SVM model, and evaluating its performance.

```

R Console
Number of Classes: 3
Levels:
setosa versicolor virginica

>
> # 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)
      Actual
Predicted setosa versicolor virginica
setosa     16         0          0
versicolor  0         17         0
virginica   0          1         13

> # Calculate accuracy
> accuracy <- sum(diag(confusion_matrix)) / sum(confusion_matrix)
> cat("Accuracy:", accuracy * 100, "%\n")
Accuracy: 97.7778 %
>

R Editor
C:\Users\mahul\OneDrive\Documents\unifr - R Editor
# 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(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

```

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)

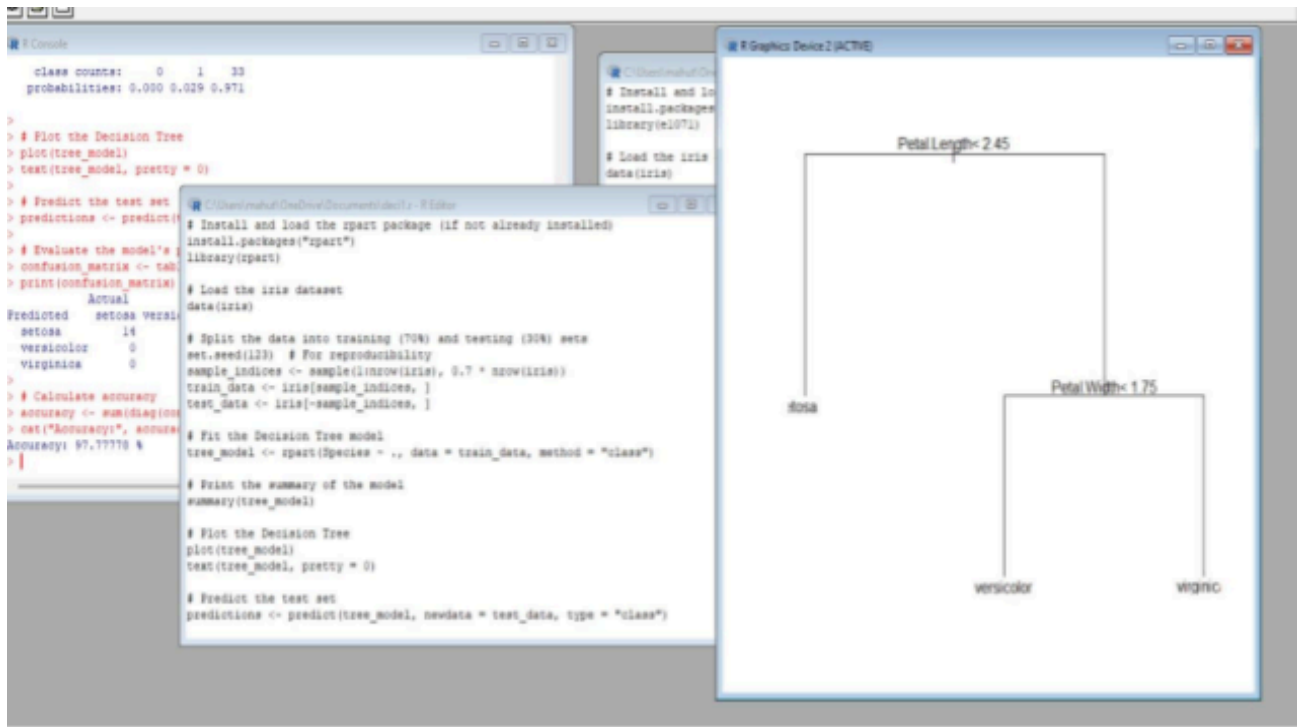
Predict the test set predictions <- predict(tree_model, newdata =
test_data, type = "class")

Evaluate the model's performance

```
confusion_matrix <- table(Predicted = predictions, Actual = test_data$Species)
print(confusion_matrix)
```

```
# Calculate accuracy
```

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



RESULT :

Thus to implement SVM and Decision tree using R language is successfully completed.

Ex.No:9**Implement clustering techniques – Hierarchical and K-Means****AIM:**

To implement clustering techniques (hierarchical and K Means) using R language.

a) HIERARCHIAL CLUSTERING

```
# Load the iris dataset
data(iris)

# Use only the numeric columns for clustering (exclude the Species column)
iris_data <- iris[, -5]

# Standardize the data
iris_scaled <- scale(iris_data)

# Compute the distance matrix distance_matrix <-
dist(iris_scaled, method = "euclidean")

# Perform hierarchical clustering using the "complete" linkage method
hc_complete <- hclust(distance_matrix, method = "complete")

# Plot the dendrogram plot(hc_complete, main = "Hierarchical Clustering Dendrogram",
xlab = "", sub = "", cex =
0.6)

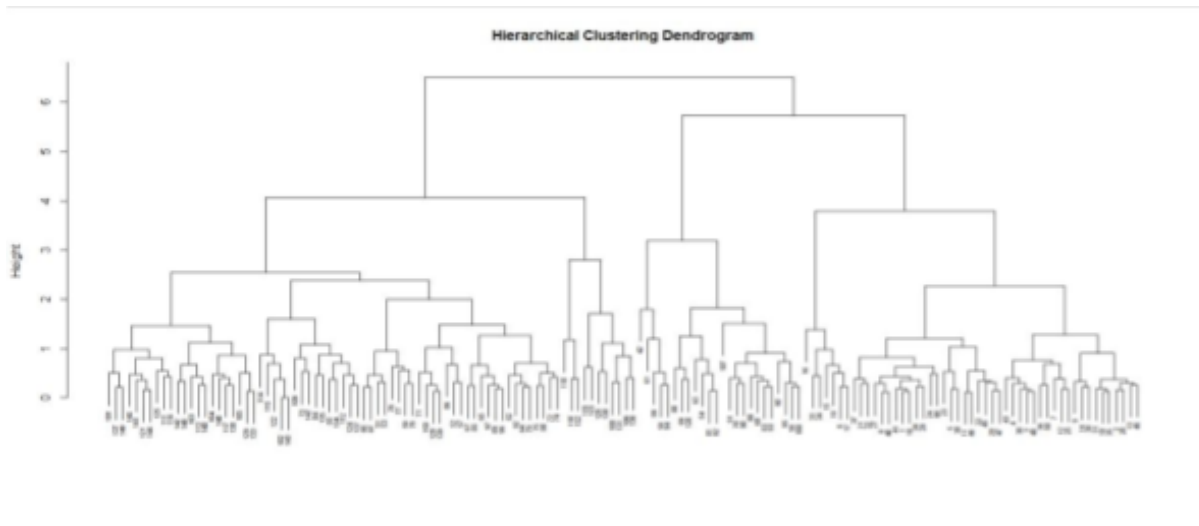
# Cut the tree to form 3 clusters
clusters <- cutree(hc_complete, k = 3)

# Print the cluster memberships
print(clusters)

# Add the clusters to the original dataset
iris$Cluster <- as.factor(clusters)
```

```
# Display the first few rows of the updated dataset
head(iris)
```

```
R Console
>
> # Cut the tree to form 3 clusters
> clusters <- cutree(hc_complete, k = 3)
>
> # Print the cluster memberships
> print(clusters)
[1] 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1
[38] 1 1 1 1 2 1 1 1 1 1 1 1 1 1 3 3 3 2 3 2 3 2 3 2 3 2 3 3 3 3 2 2 2 3 3 3
[75] 3 3 3 3 3 2 2 2 2 3 3 3 3 2 3 2 2 2 3 2 2 2 3 3 3 2 2 3 3 3 3 3 2 3 3 3
[112] 3 3 3 3 3 3 3 3 2 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3
[149] 3 3
>
> # Add the clusters to the original dataset
> iris$Cluster <- as.factor(clusters)
>
> # Display the first few rows of the updated dataset
> head(iris)
  Sepal.Length Sepal.Width Petal.Length Petal.Width Species Cluster
1           5.1         3.5          1.4          0.2  setosa      1
2           4.9         3.0          1.4          0.2  setosa      1
3           4.7         3.2          1.3          0.2  setosa      1
4           4.6         3.1          1.5          0.2  setosa      1
5           5.0         3.6          1.4          0.2  setosa      1
6           5.4         3.9          1.7          0.4  setosa      1
> |
```



b) K-MEANS CLUSTERING

```
# Load the iris dataset
data(iris)
```

```
# Use only the numeric columns for clustering (exclude the Species column)
iris_data <- iris[, -5]
```

```
# Standardize the data
iris_scaled <- scale(iris_data)
```

```
# Set the number of clusters
set.seed(123) # For reproducibility
k <- 3 # Number of clusters
```

```
# Perform K-Means clustering
kmeans_result <- kmeans(iris_scaled, centers = k, nstart = 25)
```

```
# Print the K-Means result
print(kmeans_result)
```

```
# Print the cluster centers
print(kmeans_result$centers)
```

```
# Add the cluster assignments to the original dataset
iris$Cluster <- as.factor(kmeans_result$cluster)
```

```
# Display the first few rows of the updated dataset
head(iris)
```

```
# Plot the clusters
library(ggplot2)
ggplot(iris, aes(x = Sepal.Length, y = Sepal.Width, color = Cluster)) +
  geom_point(size = 3) +
  labs(title = "K-Means Clustering of Iris Dataset", x = "Sepal Length", y = "Sepal Width")
```



RESULT:

Thus to implement clustering techniques (hierarchical and K Means) using R language is successfully completed.

Ex.No:10

VISUALIZE DATA USING ANY PLOTTING FRAMEWORK

AIM:

To visualize data using any plotting framework using R language .

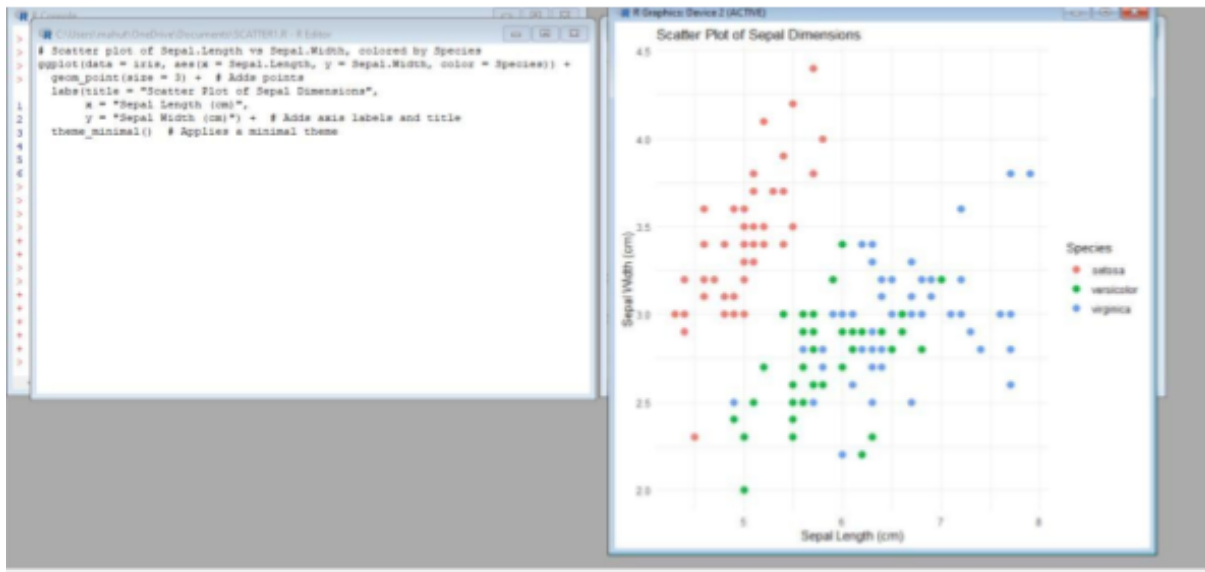
1) SCATTER PLOT

```
# Install ggplot2 (if not already installed)
install.packages("ggplot2")

# Load the ggplot2 package
library(ggplot2)

# Scatter plot of Sepal.Length vs Sepal.Width, colored by Species
ggplot(data = iris, aes(x = Sepal.Length, y = Sepal.Width, color = Species))
+ geom_point(size = 3) + # Adds points labs(title = "Scatter Plot of Sepal
```


Dimensions", x = "Sepal Length (cm)", y = "Sepal Width (cm)) +



Adds axis labels and title theme_minimal() # Applies a minimal theme

2) **BAR CHART**

Install ggplot2 (if not already installed)

install.packages("ggplot2")

Load the ggplot2 package

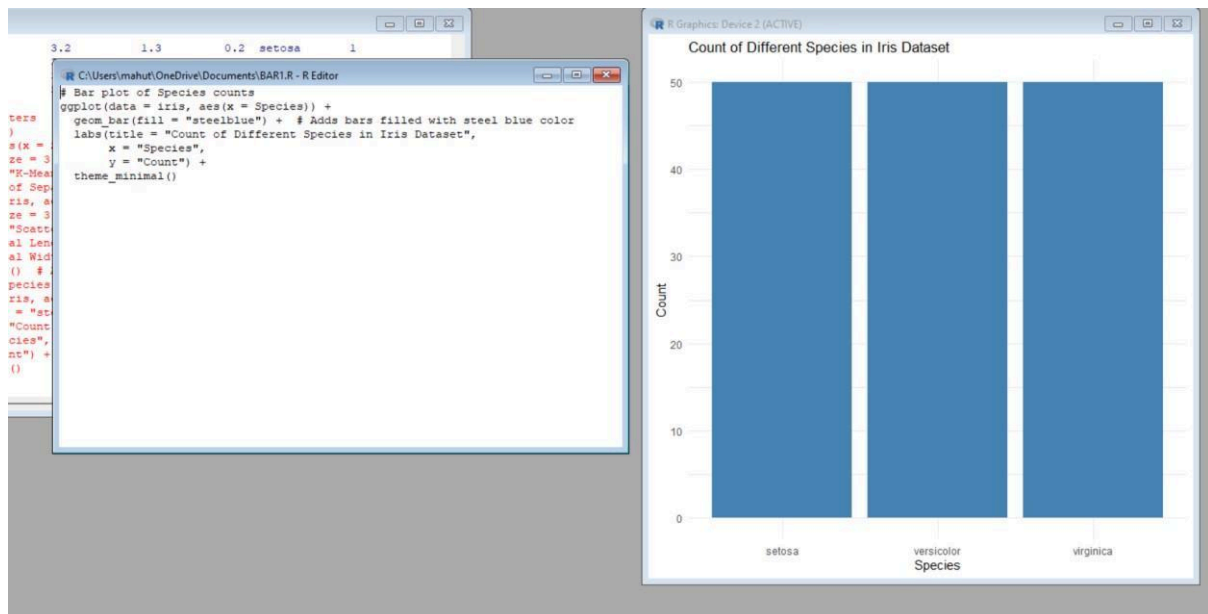
library(ggplot2)

Bar plot of Species counts

ggplot(data = iris, aes(x = Species)) +

geom_bar(fill = "steelblue") + # Adds bars filled with steel blue color

labs(title = "Count of Different Species in Iris
Dataset", x = "Species", y = "Count") +
theme_minimal()

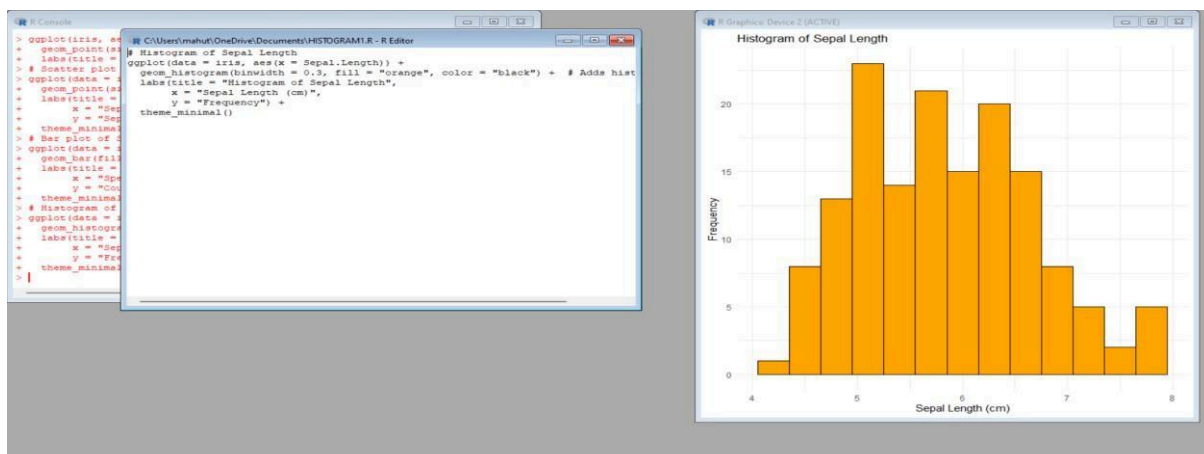


3) HISTOGRAM

Install ggplot2 (if not already installed)
install.packages("ggplot2")

Load the ggplot2 package
library(ggplot2)

Histogram of Sepal Length
ggplot(data = iris, aes(x = Sepal.Length)) +
 geom_histogram(binwidth = 0.3, fill = "orange", color = "black") + # Adds
 histogram bars
 labs(title = "Histogram of Sepal
Length", x = "Sepal Length (cm)",
y = "Frequency") +
 theme_minimal()

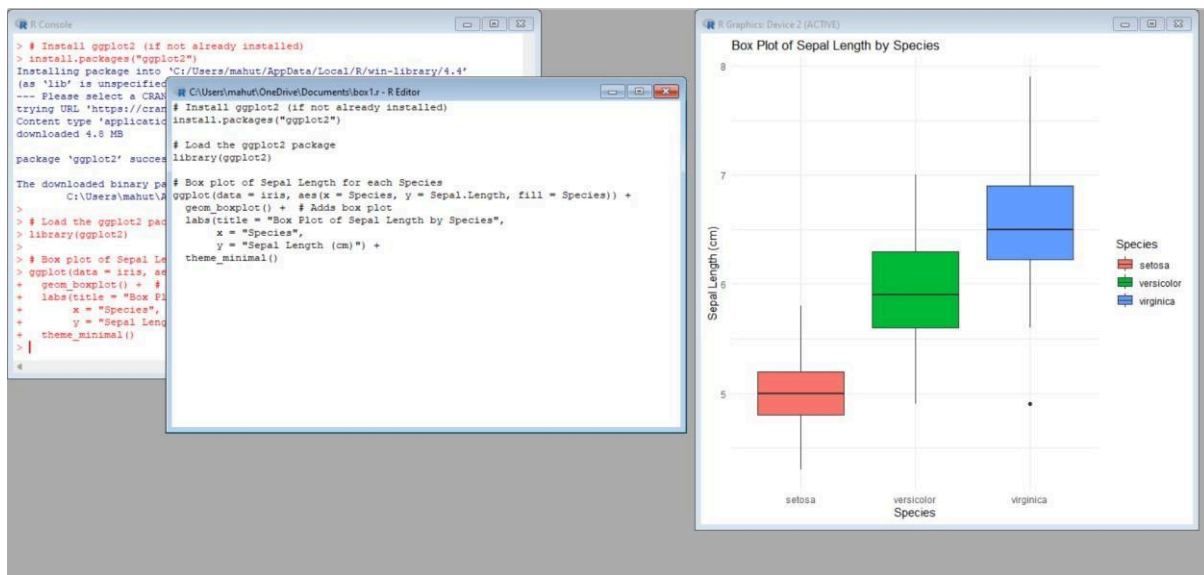


4)BOX PLOT

```
# Install ggplot2 (if not already installed)
install.packages("ggplot2")
```

```
# Load the ggplot2 package
library(ggplot2)
```

```
# Box plot of Sepal Length for each Species
ggplot(data = iris, aes(x = Species, y = Sepal.Length, fill =
Species)) + geom_boxplot() + # Adds box plot labs(title = "Box
Plot of Sepal Length by Species", x = "Species", y = "Sepal
Length (cm)") + theme_minimal()
```



RESULT :

Thus to visualize data using plotting framework using R language is successfully completed.