



SAVEETHA INSTITUTE OF MEDICAL AND TECHNICAL SCIENCES

SAVEETHA SCHOOL OF ENGINEERING



LABORATORY MANUAL

ITA0447 Statistics with R Programming for NLP

Submitted By :

Register Number :

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A. BASIC OPERATIONS IN R

1. Write the Commands to Perform Basic Arithmetic in R.

Aim:

Procedure:

Code & Output:

1. Addition (+) 2 + 3 # Output: [1] 5 # Adding a sequence of numbers 1:5 + 3 # Output: [1] 4 5 6 7 8	2. Subtraction (-) # Subtracting two numbers 10 - 3 # Output: [1] 7 # Subtracting a sequence of numbers 1:5 - 3 # Output: [1] -2 -1 0 1 2	3. Multiplication (*) # Multiplying two numbers 2 * 3 # Output: [1] 6 # Multiplying a sequence of numbers 1:5 * 3 # Output: [1] 3 6 9 12 15
4. Division (/) # Dividing two numbers 10 / 5 # Output: [1] 2 # Dividing a sequence of numbers 1:5 / 2 # Output: [1] 0.5 1.0 1.5 2.0 2.5	5. Exponentiation (^) # Raising a number to a power 2^3 # Output: [1] 8 # Raising a sequence of numbers to a power 1:5^2 # Output: [1] 1 4 9 16 25	6. Modulus (%%) # Finding the remainder of a division 10 %% 3 # Output: [1] 1 # Finding the remainder of a sequence of divisions 1:5 %% 2 # Output: [1] 1 0 1 0 1

Result:

2. Declare Variables In R And Also Write The Commands For Retrieving The Value of The Stored Variables In R Console.

Aim:

Procedure:

Code:	Output:
# declare variables x <- 5 y <- "Hello, World!" z <- TRUE # retrieve values and display output x	# Output: [1] 5 y # Output: [1] "Hello, World!" z # Output: [1] TRUE

Result:

3. Write R script to calculate the area of Rectangle.

Aim:

Procedure:

Code:	Output:
<pre># declare variables length <- 5 width <- 3 # calculate area area <- length * width # display result cat("The area of the rectangle is", area)</pre>	The area of the rectangle is 15

Result:

4. Enumerate The Process To Check Whether A Given Input Is Numeric, Integer, Double, Complex in R.

Aim:

Procedure:

Code:	Output:
<pre># declare variables x <- 5 y <- 3.14 z <- 2+3i a <- "hello" # check if variables are numeric is.numeric(x) # TRUE is.numeric(y) # TRUE is.numeric(z) # TRUE is.numeric(a) # FALSE # check if variables are integer is.integer(x) # TRUE is.integer(y) # FALSE is.integer(z) # FALSE is.integer(a) # FALSE # check if variables are double is.double(x) # TRUE is.double(y) # TRUE is.double(z) # FALSE is.double(a) # FALSE # check if variables are complex is.complex(x) # FALSE is.complex(y) # FALSE is.complex(z) # TRUE is.complex(a) # FALSE</pre>	<pre>> is.numeric(x) # TRUE [1] TRUE > is.numeric(y) # TRUE [1] TRUE > is.numeric(z) # TRUE [1] TRUE > is.numeric(a) # FALSE [1] FALSE > is.integer(x) # TRUE [1] TRUE > is.integer(y) # FALSE [1] FALSE > is.integer(z) # FALSE [1] FALSE > is.integer(a) # FALSE [1] FALSE > is.double(x) # TRUE [1] TRUE > is.double(y) # TRUE [1] TRUE > is.double(z) # FALSE [1] FALSE > is.double(a) # FALSE [1] FALSE > is.complex(x) # FALSE [1] FALSE > is.complex(y) # FALSE [1] FALSE > is.complex(z) # TRUE [1] TRUE > is.complex(a) # FALSE [1] FALSE</pre>

Result:

5. Illustration of Vector Arithmetic.

Aim:

Procedure:

Code & Output:

# create two vectors a <- c(1, 2, 3) b <- c(4, 5, 6)	# add the two vectors c <- a + b print(c) # output: 5 7 9	# subtract the two vectors d <- b - a print(d) # output: 3 3 3
# multiply the two vectors element-wise e <- a * b print(e) # output: 4 10 18	# divide the two vectors element-wise f <- b / a print(f) # output: 4 2.5 2	# calculate the dot product of the two vectors g <- sum(a * b) print(g) # output: 32

Result:

6. Write an R Program to Take Input From User.

Input name as "Jack" and age as 17.

The program should display the output as

"Hai , Jack next year you will be 18 years old"

Aim:

Procedure:

Code:

```
# take input from user
name <- readline(prompt = "Enter your name: ")
age <- readline(prompt = "Enter your age: ")

# convert age to integer and add 1
age_next_year <- as.integer(age) + 1

# display output
cat("Hai, ", name, " next year you will be ",
age_next_year, " years old.", "\n")
```

Output:

```
Enter your name: Jack
Enter your age: 17
Hai, Jack next year you will be 18 years old.
```

Result:

B.DATA STRUCTURES IN R

1) Perform Addition, Subtraction, Multiplication and Transpose of a Matrix in R

Aim:

Procedure:

Program:	Output:
<pre>matrix1 <- matrix(c(1, 2, 3, 4, 5, 6), nrow = 2, ncol = 3) matrix2 <- matrix(c(7, 8, 9, 10, 11, 12), nrow = 2, ncol = 3) addition_result <- matrix1 + matrix2 subtraction_result <- matrix1 - matrix2 multiplication_result <- matrix1 %*% t(matrix2) # Use %*% for matrix multiplication transpose_result <- t(matrix1) print("Addition Result:") print(addition_result) print("Subtraction Result:") print(subtraction_result) print("Multiplication Result:") print(multiplication_result) print("Transpose Result:") print(transpose_result)</pre>	<pre>[1] "Addition Result:" [1] [2] [3] [1,] 8 10 12 [2,] 14 16 18 [1] "Subtraction Result:" [1] [2] [3] [1,] -6 -6 -6 [2,] -6 -6 -6 [1] "Multiplication Result:" [1] [2] [1,] 50 68 [2,] 122 167 [1] "Transpose Result:" [1] [2] [1,] 1 4 [2,] 2 5 [3,] 3 6</pre>

Result:

2) Find Transpose of a matrix in R and deconstruct a matrix.

Aim:

Procedure:

Program:	Output:
<pre># Transpose of Matrix mat <- matrix(1:6, nrow = 2) transpose <- t(mat) print(mat) print(transpose) # Deconstruct of matrix mat <- matrix(1:6, nrow = 2) element_1 <- mat[1, 1] # Extracts the element at row 1, column 1 element_2 <- mat[2, 3] # Extracts the element at row 2, column 3 row_1 <- mat[1,] column_2 <- mat[, 2] print(element_1)</pre>	<pre>Output: [1] [2] [3] [1,] 1 3 5 [2,] 2 4 6 [1] [2] [1,] 1 2 [2,] 3 4 [3,] 5 6 Output: [1] 1</pre>

<pre>print(element_2) print(row_1) print(column_2)</pre>	<pre>[1] 6 [1] 1 3 5 [1] 3 4</pre>
--	------------------------------------

Result:

3) Perform the operation of combining matrices in R using cbind() and rbind() functions.

Aim :

Procedure:

Program: <pre>mat1 <- matrix(1:4, nrow = 2) mat2 <- matrix(5:8, nrow = 2) combined <- cbind(mat1, mat2) print(combined)</pre>	Output: <pre>[,1] [,2] [,3] [,4] [1,] 1 3 5 7 [2,] 2 4 6 8</pre> Output: <pre>[,1] [,2] [1,] 1 2 [2,] 3 4 [3,] 5 6 [4,] 7 8</pre>
--	--

Result:

4) Perform array manipulation in R .

Aim:

Procedure:

Program: #Reshaping Arrays in R program <pre>mat <- matrix(1:12, nrow = 3, ncol = 4) reshaped_array <- array(mat, dim = c(2, 6)) print(reshaped_array)</pre> #Subsetting Arrays in R <pre>mat <- matrix(1:12, nrow = 3, ncol = 4) subset_array <- mat[2,] print(subset_array)</pre>	Output: <pre>[,1] [,2] [,3] [,4] [,5] [,6] [1,] 1 4 7 10 1 4 [2,] 2 5 8 11 2 5</pre> Output: <pre>[1] 2 5 8 11</pre> Output: <pre>[,1] [,2] [,3] [,4] [,5] [,6]</pre>
--	---

```
#Merging Arrays in R
mat1 <- matrix(1:6, nrow = 2)
mat2 <- matrix(7:12, nrow = 2)
merged_array <- cbind(mat1, mat2)
print(merged_array)
```

```
#Transforming Arrays
mat <- matrix(1:12, nrow = 3, ncol = 4)
transformed_array <- apply(mat, 2, sum)
print(transformed_array)
```

```
[1,] 1 3 5 7 9 11
[2,] 2 4 6 8 10 12
```

Output:

```
[1] 6 15 24 33
```

Result:

5) Perform calculations across array elements in an array using the apply() function.

Aim:

Procedure:

Program:

```
my_array <- array(1:12, dim = c(3, 4))
print(my_array)
product <- function(x) {
  prod(x)}
result <- apply(my_array, 1, product)
print(result)
```

Output:

```
> print(my_array)
 [,1] [,2] [,3] [,4]
 [1,] 1 4 7 10
 [2,] 2 5 8 11
 [3,] 3 6 9 12
> print(result)
[1] 280 1760 11880
```

Result:

6) Create a data frame and print the structure of the data frame in R.

Aim:

Procedure:

Programm:

```
df <- data.frame(
  Name = c("John", "Jane", "Mike"),
  Age = c(25, 30, 35),
  Height = c(170, 165, 180),
  StringsAsFactors = FALSE
)
```

Output:

```
'data.frame': 3 obs. of 3 variables:
 $ Name : chr "John" "Jane" "Mike"
 $ Age  : num 25 30 35
 $ Height: num 170 165 180
```

str(df)	
---------	--

Result :

7) Demonstrate the creation of S3 class and S4 class in R.

Aim:

Procedure:

S3 Class	S4 Class:
----------	-----------

Program:

S3 Class

```
setClass("Person", representation(name =  
"character", age = "numeric"))  
person <- structure(list(name = "John Doe", age =  
30), class = "Person")  
print.Person <- function(obj)  
{  
  cat("Name:", obj$name, "\n")  
  cat("Age:", obj$age, "\n")  
}  
print(person)
```

S4 Class

```
setClass("Rectangle",  
  slots = c(length = "numeric", width =  
"numeric"),  
  prototype = list(length = 0, width = 0)  
)  
setMethod("area", "Rectangle", function(object) {  
  object@length * object@width  
})  
rectangle <- new("Rectangle", length = 5, width = 3)  
area(rectangle)
```

Output:

Name: John Doe
Age: 30

Output:

[1] 15

Result:

8) Demonstrate the creation of S3 class and S4 class in R. Also illustrate how the fields of the object can be accessed using the \$ operator. Modify the Name field by reassigning the name to Paul.

Aim :

Procedure:

Program:

```
person <- list(name = "John Doe", age = 30)
cat("Original Name:", person$name, "\n")
cat("Original Age:", person$age, "\n")
person$name <- "Paul"
cat("Modified Name:", person$name, "\n")
```

Output:

```
Original Name: John Doe
Original Age: 30
Modified Name: Paul
```

Result :

C. WORKING WITH LOOPING AND FUNCTIONS IN R

1. Write a program to check whether an integer (entered by the user) is a prime number or not using control statements.

AIM:

PROCEDURE:

CODE:

```
num <- as.integer(readline("Enter a number: "))

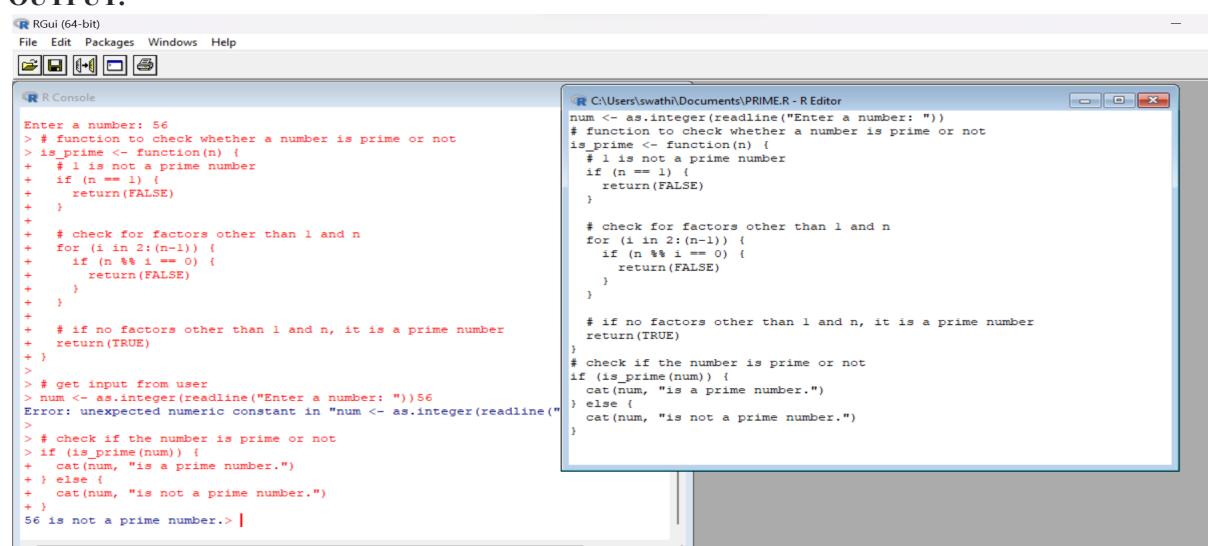
# function to check whether a number is prime or not
is_prime <- function(n) {
  # 1 is not a prime number
  if (n == 1) {
    return(FALSE)
  }

  # check for factors other than 1 and n
  for (i in 2:(n-1)) {
    if (n %% i == 0) {
      return(FALSE)
    }
  }

  # if no factors other than 1 and n, it is a prime number
  return(TRUE)
}

# check if the number is prime or not
if (is_prime(num)) {
  cat(num, "is a prime number.")
} else {
  cat(num, "is not a prime number.")
}
```

OUTPUT:



The screenshot shows the RGui interface with two windows open. On the left is the R Console window, which displays the R code for checking prime numbers and its execution. On the right is the R Editor window, which shows the same R code. Both windows have a light gray background and black text.

R Gui (64-bit)

File Edit Packages Windows Help

R Console

```
Enter a number: 56
> # function to check whether a number is prime or not
> is_prime <- function(n) {
+   # 1 is not a prime number
+   if (n == 1) {
+     return(FALSE)
+   }

+   # check for factors other than 1 and n
+   for (i in 2:(n-1)) {
+     if (n %% i == 0) {
+       return(FALSE)
+     }
+   }

+   # if no factors other than 1 and n, it is a prime number
+   return(TRUE)
+ }

> # get input from user
> num <- as.integer(readline("Enter a number: "))56
Error: unexpected numeric constant in "num<- as.integer(readline("))

> # check if the number is prime or not
> if (is_prime(num)) {
+   cat(num, "is a prime number.")
+ } else {
+   cat(num, "is not a prime number.")
+ }
56 is not a prime number.> |
```

R C:\Users\swathi\Documents\PRIMER.R - R Editor

```
num <- as.integer(readline("Enter a number: "))
# function to check whether a number is prime or not
is_prime <- function(n) {
  # 1 is not a prime number
  if (n == 1) {
    return(FALSE)
  }

  # check for factors other than 1 and n
  for (i in 2:(n-1)) {
    if (n %% i == 0) {
      return(FALSE)
    }
  }

  # if no factors other than 1 and n, it is a prime number
  return(TRUE)
}

# check if the number is prime or not
if (is_prime(num)) {
  cat(num, "is a prime number.")
} else {
  cat(num, "is not a prime number.")
}
```

RESULT:

2. Write a program to check whether a number entered by the user is positive number or a negative number or zero.
AIM:

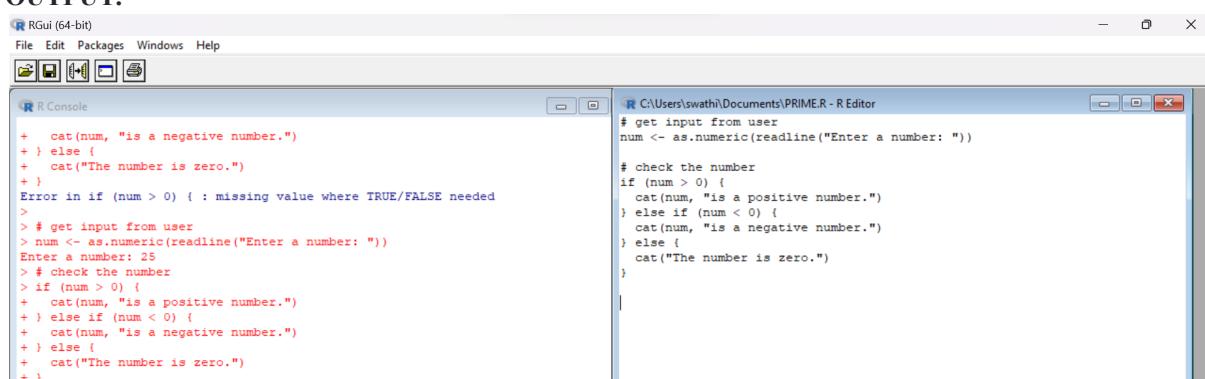
PROCEDURE:

CODE:

```
# get input from user
num <- as.numeric(readline("Enter a number: "))

# check the number
if(num > 0) {
  cat(num, "is a positive number.")
} else if (num < 0) {
  cat(num, "is a negative number.")
} else {
  cat("The number is zero.")
}
```

OUTPUT:



The screenshot shows the RGui interface with two windows open. On the left is the 'R Console' window, which contains the R code for checking a number's sign. On the right is the 'C:\Users\swathi\Documents\PRIMER.R - R Editor' window, which shows the same code with some parts highlighted in red. The R Editor window has a vertical scrollbar on its right side.

```
RGui (64-bit)
File Edit Packages Windows Help
R Console
C:\Users\swathi\Documents\PRIMER.R - R Editor
```

```
+   cat(num, "is a negative number.")
+ } else {
+   cat("The number is zero.")
+
Error in if (num > 0) { : missing value where TRUE/FALSE needed
>
> # get input from user
> num <- as.numeric(readline("Enter a number: "))
Enter a number: 25
> # check the number
> if (num > 0) {
+   cat(num, "is a positive number.")
+ } else if (num < 0) {
+   cat(num, "is a negative number.")
+ } else {
+   cat("The number is zero.")
+ }
```

RESULT:

3. Write a program to check whether a number is an Armstrong number or not using a while loop.
AIM:

PROCEDURE:

CODE:

```
num = as.integer(readline(prompt="Enter a number: "))
sum = 0
temp = num
while(temp > 0) {
  digit = temp %% 10
  sumsum = sum + (digit ^ 3)
  temp = floor(temp / 10)
}
if(num == sum)
{
```

```

    print(paste(num, "is an Armstrong number"))
} else
{
    print(paste(num, "is not an Armstrong number"))
}

```

OUTPUT:

The screenshot shows the RGui interface with two windows open. The left window is the R Console, which displays the R code for checking if a number is an Armstrong number, followed by an error message and the result "[1] "1221 is not an Armstrong number"" when the number 1221 is entered. The right window is the R Editor, showing the same code as the console.

```

RGui (64-bit)
File Edit Packages Windows Help
[R] [New] [Open] [Save] [Save As] [Print]
[R Console] C:\Users\swathi\Documents\ARMSTRONG.R - R Editor
> num = as.integer(readline(prompt="Enter a number: "))
Enter a number: 1221
> sum = 0
> temp = num
> while(temp > 0) {
+   digit = temp %% 10
+   sumsum = sum + (digit ^ 3)
+   temp = floor(temp / 10)
+ }
> if(num ==sum)
Error: unexpected '=' in "if(num =="
> {
+   print(paste(num, "is an Armstrong number"))
+ } else
Error: unexpected 'else' in:
" " print(paste(num, "is an Armstrong number"))
} else"
> {
+   print(paste(num, "is not an Armstrong number"))
+ }
[1] "1221 is not an Armstrong number"
>

```

RESULT:

4. Write a program to demonstrate Repeat Loop in R

AIM:

PROCEDURE:

CODE:

```

x = 1
# Repeat loop
repeat {
  print(x)

  # Break statement to terminate if x > 4
  if(x > 4) {
    break
  }

  # Increment x by 1
  x = x + 1
}

```

OUTPUT:

```

> x = 1
>
> # Repeat loop
> repeat {
+   print(x)
+   # Break statement to terminate if x > 4
+   if (x > 4) {
+     break
+   }
+   # Increment x by 1
+   x = x + 1
+
+ }
[1] 1
[1] 2
[1] 3
[1] 4
[1] 5
> 

```

RESULT:

5.Using functions develop a simple calculator in R.

AIM:

PROCEDURE:

CODE:

```
num1=2
num2=12
```

```

# Menu for the user to choose an operation
cat("Select an operation:\n")
cat("1. Addition\n")
cat("2. Subtraction\n")
cat("3. Multiplication\n")
cat("4. Division\n")

option <- as.integer(readline("Enter your choice (1-4): "))

# Function to add two numbers
add <- function(a, b) {
  return(a + b)
}

# Function to subtract two numbers
subtract <- function(a, b) {
  return(a - b)
}

# Function to multiply two numbers
multiply <- function(a, b) {
  return(a * b)
}

# Function to divide two numbers
divide <- function(a, b) {
  return(a / b)
}

# Get input from the user

```

```

# Function to add two numbers
add <- function(a, b) {
  return(a + b)
}

# Function to subtract two numbers
subtract <- function(a, b) {
  return(a - b)
}

# Function to multiply two numbers
multiply <- function(a, b) {
  return(a * b)
}

# Function to divide two numbers
divide <- function(a, b) {
  return(a / b)
}

# Get input from the user
num1=12
num2=2

# Menu for the user to choose an operation
cat("Select an operation:\n")
cat("1. Addition\n")
cat("2. Subtraction\n")
cat("3. Multiplication\n")
cat("4. Division\n")

option <- as.integer(readline("Enter your choice (1-4): "))

# Perform the selected operation
result <- switch(option,
  add(num1, num2),
  subtract(num1, num2),
  multiply(num1, num2),
  divide(num1, num2))

# Display the result
cat("The result is:", result)

# Perform the selected operation
result <- switch(option,
  add(num1, num2),
  subtract(num1, num2),
  multiply(num1, num2),
  divide(num1, num2))

# Display the result
cat("The result is:", result)
OUTPUT:

```

```

RGui (64-bit)
File Edit Packages Windows Help
R Console
>
>
> num1=2
> num2=12
>
>
> num1=2
> num2=12
> # Menu for the user to choose an operation
> cat("Select an operation:\n")
Select an operation:
> cat("1. Addition\n")
1. Addition
> cat("2. Subtraction\n")
2. Subtraction
> cat("3. Multiplication\n")
3. Multiplication
> cat("4. Division\n")
4. Division
>
> option <- as.integer(readline("Enter your choice (1-4): "))
Enter your choice (1-4): 1
> # Perform the selected operation
> result <- switch(option,
+                   add(num1, num2),
+                   subtract(num1, num2),
+                   multiply(num1, num2),
+                   divide(num1, num2))
>
> # Display the result
> cat("The result is:", result)
The result is: 14> |

```

```

C:\Users\swathi\Documents\CALCULATOR.R - R Editor
num1=2
num2=12

# Menu for the user to choose an operation
cat("Select an operation:\n")
cat("1. Addition\n")
cat("2. Subtraction\n")
cat("3. Multiplication\n")
cat("4. Division\n")

option <- as.integer(readline("Enter your choice (1-4): "))

# Function to add two numbers
add <- function(a, b) {
  return(a + b)
}

# Function to subtract two numbers
subtract <- function(a, b) {
  return(a - b)
}

# Function to multiply two numbers
multiply <- function(a, b) {
  return(a * b)
}

# Function to divide two numbers
divide <- function(a, b) {
  return(a / b)
}

```

RESULT:

6. Demonstrate the creation of a complex number in R.

AIM:

PROCEDURE:

CODE:

```
# Create a complex number using complex() function
complex_num_1 <- complex(real = 2, imaginary = 3)
cat("Complex number 1:", complex_num_1, "\n")
```

OUTPUT:

```

RGui (64-bit)
File Edit Packages Windows Help
R Console
>
>
> # Create a complex number using complex() function
> complex_num_1 <- complex(real = 2, imaginary = 3)
> cat("Complex number 1:", complex_num_1, "\n")
Complex number 1: 2+3i
>
> |

```

```

C:\Users\swathi\Documents\COMPLEX.R - R Editor
# Create a complex number using complex() function
complex_num_1 <- complex(real = 2, imaginary = 3)
cat("Complex number 1:", complex_num_1, "\n")

```

RESULT:

7. Write a program to multiply two numbers using a function with a default value. Assume default value as NULL.

AIM:

PROCEDURE:

CODE:

```

multiply_numbers <- function(a, b = NULL) {
  if (is.null(b)) {
    print("Error: Second number is missing.")
  } else {
    product <- a * b
    return(product)
  }
}
# Example usage
result1 <- multiply_numbers(5, 3) # Multiply 5 and 3
print(result1) # Output: 15

```

result2 <- multiply_numbers(2) # Error: Second number is missing

OUTPUT:

The screenshot shows the RGui interface. On the left is the R Console window, which contains the R code for the `multiply_numbers` function and its execution. On the right is the R Editor window, which shows the same code with color-coded syntax highlighting. The R Console output shows the function definition, an example usage where `b` is provided, and an attempt to run the function with only `a`, which results in an error message.

```

RGui (64-bit)
File Edit Packages Windows Help
R Console
C:\Users\swathi\Documents\MULTIPLY.R - R Editor
> multiply_numbers <- function(a, b = NULL) {
+   if (is.null(b)) {
+     print("Error: Second number is missing.")
+   } else {
+     product <- a * b
+     return(product)
+   }
>
> # Example usage
> result1 <- multiply_numbers(5, 3) # Multiply 5 and 3
> print(result1) # Output: 15
[1] 15
>
> result2 <- multiply_numbers(2) # Error: Second number is missing
[1] "Error: Second number is missing."
> |
```

RESULT:**8.Find sum, mean and product of vector elements using built-in functions.****AIM:****PROCEDURE:****CODE:**

```

# Create a sample vector
v <- c(2, 4, 6, 8, 10)

# Sum of vector elements
sum_v <- sum(v)
cat("Sum of vector elements: ", sum_v, "\n")

# Mean of vector elements
mean_v <- mean(v)
cat("Mean of vector elements: ", mean_v, "\n")

# Product of vector elements
prod_v <- prod(v)
cat("Product of vector elements: ", prod_v, "\n")

```

OUTPUT:

The screenshot shows the RGui interface. On the left is the R Console window with the following R code and output:

```
> v <- c(2, 4, 6, 8, 10)
>
> # Sum of vector elements
> sum_v <- sum(v)
> cat("Sum of vector elements: ", sum_v, "\r")
Sum of vector elements: 30
>
> # Mean of vector elements
> mean_v <- mean(v)
> cat("Mean of vector elements: ", mean_v, "\r")
Mean of vector elements: 6
>
> # Product of vector elements
> prod_v <- prod(v)
> cat("Product of vector elements: ", prod_v, "\r")
Product of vector elements: 3840
> |
```

On the right is the R Editor window titled "C:\Users\swathi\Documents\VECTOR.R - R Editor" with the same R code:

```
# Create a sample vector
v <- c(2, 4, 6, 8, 10)

# Sum of vector elements
sum_v <- sum(v)
cat("Sum of vector elements: ", sum_v, "\n")

# Mean of vector elements
mean_v <- mean(v)
cat("Mean of vector elements: ", mean_v, "\n")

# Product of vector elements
prod_v <- prod(v)
cat("Product of vector elements: ", prod_v, "\n")|
```

RESULT:

9.Sort a vector in R using sort() function. Also find the index of the sorted vector.

AIM:

PROCEDURE:

CODE:

```
# Create a vector
my_vector <- c(8, 3, 1, 6, 2)

# Sort the vector
sorted_vector <- sort(my_vector)

# Find the index of the sorted vector
index <- order(my_vector)

# Print the sorted vector and its index
print(sorted_vector)
print(index)
```

OUTPUT:

```

RGui (64-bit)
File Edit Packages Windows Help
R Console
> cat("Product of vector elements: ", prod_v,
Product of vector elements: 3840
> # Create a vector
> my_vector <- c(8, 3, 1, 6, 2)
>
> # Sort the vector
> sorted_vector <- sort(my_vector)
>
> # Find the index of the sorted vector
> index <- order(my_vector)
>
> # Print the sorted vector and its index
> print(sorted_vector)
[1] 1 2 3 6 8
> print(index)
[1] 3 5 2 4 1
>

```

```

C:\Users\swathi\Documents\INDEX.R - R ...
# Create a vector
my_vector <- c(8, 3, 1, 6, 2)

# Sort the vector
sorted_vector <- sort(my_vector)

# Find the index of the sorted vector
index <- order(my_vector)

# Print the sorted vector and its index
print(sorted_vector)
print(index)

```

RESULT:

10. Find the L.C.M of two numbers entered by the user by creating a user-defined function.

AIM:

PROCEDURE:

CODE:

```

# Function to calculate the LCM of two numbers
calculate_lcm <- function(a, b) {
  # Find the maximum of the two numbers
  max_num <- max(a, b)

  # Initialize lcm variable
  lcm <- max_num
  while (TRUE) {
    # Check if lcm is divisible by both numbers
    if (lcm %% a == 0 && lcm %% b == 0) {
      break # Found the LCM, exit the loop
    }

    # Increment lcm by the maximum number
    lcm <- lcm + max_num
  }
  return(lcm)
}

# Get input from the user
num1 <- as.integer(readline("Enter the first number: "))
num2 <- as.integer(readline("Enter the second number: "))

# Call the function to calculate the LCM
result <- calculate_lcm(num1, num2)

# Print the result
cat("The LCM of", num1, "and", num2, "is", result, "\n")

```

OUTPUT:

The screenshot shows the RGui interface with two windows open. On the left is the R Console window, which contains R code for calculating the LCM of two numbers. On the right is the R Editor window, which shows the same code in a more structured, highlighted format.

```

RGui (64-bit)
File Edit Packages Windows Help
R Console
+ # Initialize lcm variable
+ lcm <- max_num
+
+ while (TRUE) {
+   # Check if lcm is divisible by both numbers
+   if (lcm %% a == 0 && lcm %% b == 0) {
+     break # Found the LCM, exit the loop
+   }
+
+   # Increment lcm by the maximum number
+   lcm <- lcm + max_num
+ }
+
+ return(lcm)
+ }
> # Get input from the user
> num1 <- as.integer(readline("Enter the first number: "))
Enter the first number: num2 <- as.integer(readline("Enter the second number: "))
Warning message:
NAs introduced by coercion
>
> # Call the function to calculate the LCM
> result <- calculate_lcm(num1, num2)
Error in if (lcm%%a == 0 && lcm%%b == 0) { :
  missing value where TRUE/FALSE needed
>
> # Print the result
> cat("The LCM of", num1, "and", num2, "is", result, "\n")
The LCM of NA and 2 is 14
> |
```

```

C:\Users\swathi\Documents\LCM.R - R Editor
# Function to calculate the LCM of two numbers
calculate_lcm <- function(a, b) {
  # Find the maximum of the two numbers
  max_num <- max(a, b)

  # Initialize lcm variable
  lcm <- max_num

  while (TRUE) {
    # Check if lcm is divisible by both numbers
    if (lcm %% a == 0 && lcm %% b == 0) {
      break # Found the LCM, exit the loop
    }

    # Increment lcm by the maximum number
    lcm <- lcm + max_num
  }

  return(lcm)
}

# Get input from the user
num1 <- as.integer(readline("Enter the first number: "))
num2 <- as.integer(readline("Enter the second number: "))

# Call the function to calculate the LCM
result <- calculate_lcm(num1, num2)

# Print the result
cat("The LCM of", num1, "and", num2, "is", result, "\n")
```

RESULT:

D. IMPLEMENTATION OF VECTOR RECYCLING, APPLY FAMILY & RECURSION

1. Demonstrate Vector Recycling in R.

AIM:

PROCEDURE:

CODE:	OUTPUT:
a<-c(23,45,67,89) b<-c(4,5) print(a+b)	27 50 71 94

RESULT:

2.Demonstrate the usage of apply function in R

AIM:

PROCEDURE:

CODE:	OUTPUT:
m<- matrix((1:12), nrow=3) print(m) print(apply(m,1,max)) print(apply(m,2,sum))	[,1] [,2] [,3] [,4] [1,] 1 4 7 10 [2,] 2 5 8 11 [3,] 3 6 9 12 > print(apply(m,1,max)) [1] 10 11 12 > print(apply(m,2,sum)) [1] 6 15 24 33

RESULT:

3.Demonstrate the usage of lapply function in R

AIM:

PROCEDURE:

CODE:	OUTPUT:
<pre> names <- c("nikitha", "abhi","pawan", "hitler","deva") print("original data:") names # apply lapply() function print("data after lapply():") lapply(names, toupper) </pre>	<pre> [1] "nikitha" "abhi" "pawan" "hitler" "deva" [[1]] [1] "NIKITHA" [[2]] [1] "ABHI" [[3]] [1] "PAWAN" [[4]] [1] "HITLER" [[5]] [1] "DEVA" </pre>

4. Demonstrate the usage of sapply function in R

AIM:

PROCEDURE:

CODE:	OUTPUT:
<pre> sample_data<- data.frame(x=c(1,2,3,4,5,6),y=c(3,2,4,2,34,5)) print("original data:") sample_data # apply sapply() function print("data after sapply():") sapply(sample_data, max) </pre>	<pre> x y 1 1 3 2 2 2 3 3 4 4 4 2 5 5 34 6 6 5 [1] "data after sapply():" > sapply(sample_data, max) x y 6 34 </pre>

5. Demonstrate the usage of tapply function in R

AIM:

PROCEDURE:

CODE:

```
# load library tidyverse
library(tidyverse)

# print head of diamonds dataset
print(" Head of data:")
head(diamonds)

# apply tapply function to get average price by cut
print("Average price for each cut of diamond:")
tapply(diamonds$price, diamonds$cut, mean)
```

OUTPUT:

```
>
> # print head of diamonds dataset
> print(" Head of data:")
[1] " Head of data:"
> head(diamonds)
# A tibble: 6 × 10
   carat    cut     color clarity depth table price     x     y     z
   <dbl> <ord>    <ord> <ord> <dbl> <dbl> <int> <dbl> <dbl> <dbl>
1  0.23 Ideal      E     SI2     61.5    55   326  3.95  3.98  2.43
2  0.21 Premium    E     SI1     59.8    61   326  3.89  3.84  2.31
3  0.23 Good       E     VS1     56.9    65   327  4.05  4.07  2.31
4  0.29 Premium    I     VS2     62.4    58   334  4.2    4.23  2.63
5  0.31 Good       J     SI2     63.3    58   335  4.34  4.35  2.75
6  0.24 Very Good  J     VVS2    62.8    57   336  3.94  3.96  2.48
>
> # apply tapply function to get average price by cut
> print("Average price for each cut of diamond:")
[1] "Average price for each cut of diamond:"
> tapply(diamonds$price, diamonds$cut, mean)
  Fair        Good      Very Good      Premium      Ideal
4358.758 3928.864 3981.760 4584.258 3457.542
> |
```

The screenshot shows the RStudio interface with the R console tab active. The code is identical to the one above, but the output is displayed in a larger, monospaced font. It shows the head of the diamonds dataset and the resulting data frame, followed by the average price for each cut of diamond.

```
R C:\Users\NIKITHA\Desktop\R programs\Day 2 lab\qn 5.R - R Editor
# load library tidyverse
library(tidyverse)

# print head of diamonds dataset
print(" Head of data:")
head(diamonds)

# apply tapply function to get average price by cut
print("Average price for each cut of diamond:")
tapply(diamonds$price, diamonds$cut, mean)
```

RESULT:**6. Demonstrate the usage of mapply function in R****AIM:****PROCEDURE:****CODE:**

```
mapply(rep, 1:3, times=5)
```

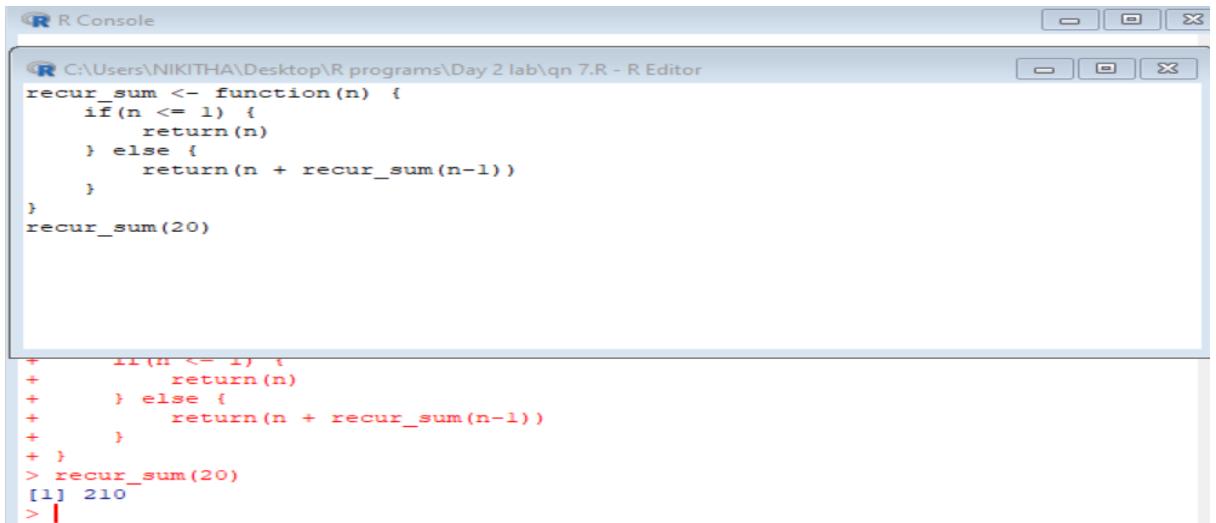
OUTPUT:

```
[,1] [,2] [,3]
[1,] 1 2 3
[2,] 1 2 3
[3,] 1 2 3
[4,] 1 2 3
[5,] 1 2 3
```

RESULT:**7. Sum of Natural Numbers using Recursion****AIM:**

PROCEDURE:

CODE & OUTPUT:



The screenshot shows the R Console window. The code in the editor pane is:

```
recur_sum <- function(n) {
  if(n <= 1) {
    return(n)
  } else {
    return(n + recur_sum(n-1))
  }
}
recur_sum(20)
```

The output pane shows the results of running the function:

```
+   if(n <= 1) {
+     return(n)
+   } else {
+     return(n + recur_sum(n-1))
+
+ }
> recur_sum(20)
[1] 210
> |
```

RESULT:

8. Write a program to generate Fibonacci sequence using Recursion in R

AIM:

PROCEDURE:

CODE:

```
fibonacci <- function(n) {
  if(n <= 1) {
    return(n)
  } else {
    return(fibonacci(n-1) + fibonacci(n-2))
  }
}

for(i in 0:10) {
  result <- fibonacci(i)
  cat(result, " ")
}
```

OUTPUT:

```
0 1 1 2 3 5 8 13 21 34 55
```

RESULT:

9. Write a program to find factorial of a number in R using recursion.

AIM:

PROCEDURE:

CODE:

```
factorial <- function(n) {  
  if(n <= 1) {  
    return(1)  
  } else {  
    return(n * factorial(n-1))  
  }  
}  
result <- factorial(9)  
cat("9! =", result)
```

OUTPUT:

9!=362880

RESULT:

E. CREATION AND MANIPULATION OF DATAFRAMES IN R

1. Consider two vectors: `x=seq(1,43,along.with=Id)` `y=seq(-20,0,along.with=Id)`

Create a data frame ‘df’ as shown below.

```
>df  
Id Letter x y  
1 1 a 1.000000 -20.000000  
2 1 b 4.818182 -18.181818  
3 1 c 8.636364 -16.363636  
4 2 a 12.454545 -14.545455  
5 2 b 16.272727 -12.727273  
6 2 c 20.090909 -10.909091  
7 3 a 23.909091 -9.090909  
8 3 b 27.727273 -7.272727  
9 3 c 31.545455 -5.454545  
10 4 a 35.363636 -3.636364  
11 4 b 39.181818 -1.818182  
12 4 c 43.000000 0.000000
```

AIM:

PROCEDURE:

CODE:

```
x <- seq(1, 43, along.with = Id)  
y <- seq(-20, 0, along.with = Id)  
letters <- c("a", "b", "c")  
  
df <- data.frame(Id = rep(1:4, each = 3),  
                  Letter = rep(letters, 4),  
                  x = x,  
                  y = y)
```

df

OUTPUT:

```

> x <- seq(1, 43, along.with = Id)
> y <- seq(-20, 0, along.with = Id)
> letters <- c("a", "b", "c")
>
> df <- data.frame(Id = rep(1:4, each = 3),
+                     Letter = rep(letters, 4),
+                     x = x,
+                     y = y)
>
> df
   Id Letter      x      y
1  1     a 1.000000 -20.000000
2  1     b 4.818182 -18.181818
3  1     c 8.636364 -16.363636
4  2     a 12.454545 -14.545455
5  2     b 16.272727 -12.727273
6  2     c 20.090909 -10.909091
7  3     a 23.909091 -9.090909
8  3     b 27.727273 -7.272727
9  3     c 31.545455 -5.454545
10 4     a 35.363636 -3.636364
11 4     b 39.181818 -1.818182
12 4     c 43.000000  0.000000
> |

```

RESULT:

2. Create two data frame df1 and df2:

```

> df1
Id Age
1 1 14
2 2 12
3 3 15
4 4 10
> df2
Id Sex Code
1 1 F a
2 2 M b
3 3 M c
4 4 F d

```

From df1 and df2 create M:

```

>M
Id Age Sex Code
1 1 14 F a
2 2 12 M b
3 3 15 M c 4 4 10 F d

```

AIM:

PROCEDURE:

CODE:

```

# Create df1
df1 <- data.frame(Id = 1:4,

```

```

Age = c(14, 12, 15, 10))
df1
# Create df2
df2 <- data.frame(Id = 1:4,
Sex = c("F", "M", "M", "F"),
Code = c("a", "b", "c", "d"))
df2
# Combine df1 and df2 into M
M <- merge(df1, df2)
M

```

OUTPUT:

The screenshot shows the R Console window with two panes. The left pane displays the R code and its execution results. The right pane shows the code being typed into the editor.

```

R Console
C:\Users\NIKITHA\Desktop\R programs\Day 2 lab\Day 4 lab\EX E\EX 2.R - R Editor

# Create df1
df1 <- data.frame(Id = 1:4,
Age = c(14, 12, 15, 10))

# Create df2
df2 <- data.frame(Id = 1:4,
Sex = c("F", "M", "M", "F"),
Code = c("a", "b", "c", "d"))

# Combine df1 and df2 into M
M <- merge(df1, df2)

M

```

RESULT:

3. Create a data frame df3:

```

> df3 id2 score 1 4 100
2 3 98
3 2 94
4 1 99

```

From M (used in Exercise-3) and df3 create N:

	Id	Age	Sex	Code	score
1	1	14	F	a	99
2	2	12	M	b	94
3	3	15	M	c	98
4	4	10	F	d	100

AIM:

PROCEDURE:

CODE:

```

# Create df1
df1 <- data.frame(Id = 1:4,
                    Age = c(14, 12, 15, 10))
# Create df2
df2 <- data.frame(Id = 1:4,
                    Sex = c("F", "M", "M", "F"),
                    Code = c("a", "b", "c", "d"))
# Combine df1 and df2 into M
M <- merge(df1, df2)
M
# Create df3
df3 <- data.frame(id2 = c(4, 3, 2, 1),
                    score = c(100, 98, 94, 99))
df3
# Combine M and df3 into N
N <- merge(M, df3, by.x = "Id", by.y = "id2")
N

```

OUTPUT:

The screenshot shows the R Console window with the following output:

```

R Console
C:\Users\NIKITHA\Desktop\R programs\Day 2 lab\Day 4 lab\EX E\ex3.r - R Editor

# Create df1
df1 <- data.frame(Id = 1:4,
                    Age = c(14, 12, 15, 10))
# Create df2
df2 <- data.frame(Id = 1:4,
                    Sex = c("F", "M", "M", "F"),
                    Code = c("a", "b", "c", "d"))
# Combine df1 and df2 into M
M <- merge(df1, df2)
M

# Create df3
df3 <- data.frame(id2 = c(4, 3, 2, 1),
                    score = c(100, 98, 94, 99))
df3
# Combine M and df3 into N
N <- merge(M, df3, by.x = "Id", by.y = "id2")
N

N
  Id Age Sex Code score
1  1   14   F    a    99
2  2   12   M    b    94
3  3   15   M    c    98
4  4   10   F    d   100
> |

```

RESULT:

4. Consider the previous one data frame N:

- 1) Remove the variables Sex and Code
- 2) From N, create a data frame:

values ind

1	1	Id
2	2	Id
3	3	Id
4	4	Id
5	14	Age
6	12	Age
7	15	Age
8	10	Age
9	99	score
10	94	score
11	98	score
12	100	score

AIM:

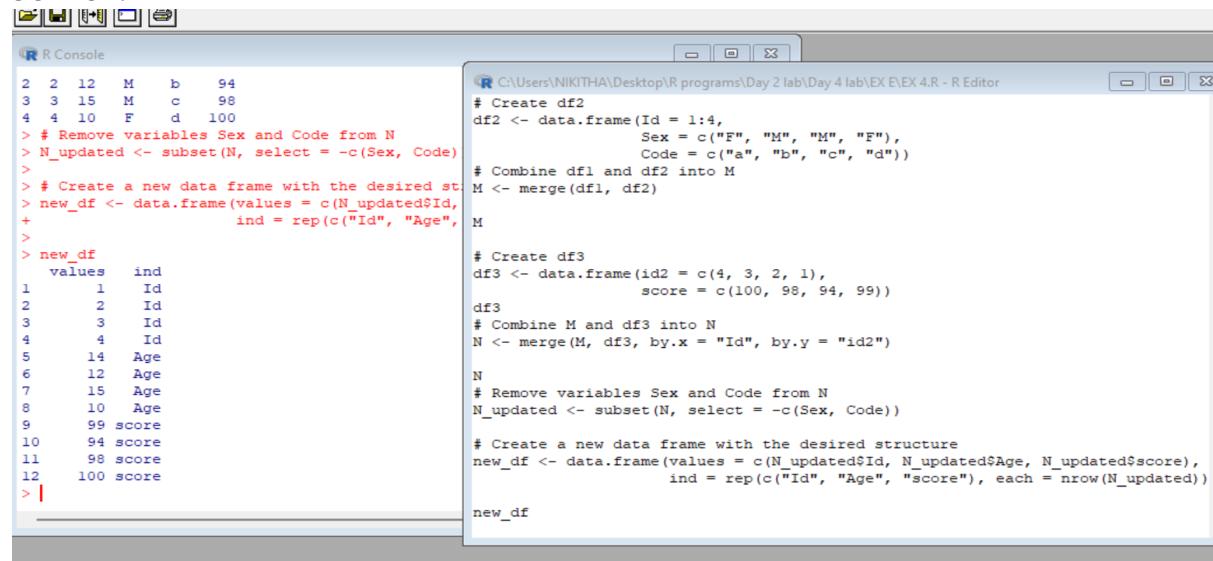
PROCEDURE:

CODE:

```
# Create df1
df1 <- data.frame(Id = 1:4,
                    Age = c(14, 12, 15, 10))
# Create df2
df2 <- data.frame(Id = 1:4,
                    Sex = c("F", "M", "M", "F"),
                    Code = c("a", "b", "c", "d"))
# Combine df1 and df2 into M
M <- merge(df1, df2)
M
# Create df3
df3 <- data.frame(id2 = c(4, 3, 2, 1),
                    score = c(100, 98, 94, 99))
df3
# Combine M and df3 into N
N <- merge(M, df3, by.x = "Id", by.y = "id2")
N
# Remove variables Sex and Code from N
N_updated <- subset(N, select = -c(Sex, Code))

# Create a new data frame with the desired structure
new_df <- data.frame(values = c(N_updated$Id, N_updated$Age, N_updated$score),
                      ind = rep(c("Id", "Age", "score"), each = nrow(N_updated)))
```

new_df

OUTPUT:

The screenshot shows the RStudio interface with two main windows. The left window is the R Console, displaying the command history and the resulting data frame 'new_df'. The right window is the R Editor, showing the R script with the code for creating data frames df1, df2, M, df3, and N, and for creating the final new_df. The data frame 'new_df' is shown in the R Console window.

```
R Console
2 2 12 M b 94
3 3 15 M c 98
4 4 10 F d 100
> # Remove variables Sex and Code from N
> N_updated <- subset(N, select = -c(Sex, Code))
>
> # Create a new data frame with the desired structure
> new_df <- data.frame(values = c(N_updated$Id,
+                                     ind = rep(c("Id", "Age",
+
> new_df
values ind
1 1 Id
2 2 Id
3 3 Id
4 4 Id
5 14 Age
6 12 Age
7 15 Age
8 10 Age
9 99 score
10 94 score
11 98 score
12 100 score
> |
```

```
R C:\Users\NIKITHA\Desktop\R programs\Day 2 lab\Day 4 lab\EX E\EX 4.R - R Editor
# Create df2
df2 <- data.frame(Id = 1:4,
                    Sex = c("F", "M", "M", "F"),
                    Code = c("a", "b", "c", "d"))
# Combine df1 and df2 into M
M <- merge(df1, df2)

# Create df3
df3 <- data.frame(id2 = c(4, 3, 2, 1),
                    score = c(100, 98, 94, 99))
df3
# Combine M and df3 into N
N <- merge(M, df3, by.x = "Id", by.y = "id2")

N
# Remove variables Sex and Code from N
N_updated <- subset(N, select = -c(Sex, Code))

# Create a new data frame with the desired structure
new_df <- data.frame(values = c(N_updated$Id, N_updated$Age, N_updated$score),
                      ind = rep(c("Id", "Age", "score"), each = nrow(N_updated)))
```

RESULT:**6. For this exercise, we'll use the (built-in) dataset trees.**

- Make sure the object is a data frame, if not change it to a data frame.
- Create a new data frame A:

>A

Girth Height Volume

```

mean_tree 13.24839 76 30.17097
min_tree 8.30000 63 10.20000
max_tree 20.60000 87 77.00000
sum_tree 410.70000 2356 935.30000

```

AIM:

PROCEDURE:

CODE:

```

# Check if the object is a data frame and convert it if necessary
if (!is.data.frame(trees)) {
  trees <- as.data.frame(trees)
}

# Create the new data frame A
A <- data.frame(Girth = c(mean(trees$Girth), min(trees$Girth), max(trees$Girth), sum(trees$Girth)),
                 Height = c(mean(trees$Height), min(trees$Height), max(trees$Height), sum(trees$Height)),
                 Volume = c(mean(trees$Volume), min(trees$Volume), max(trees$Volume), sum(trees$Volume)))

# Assign row names to A
row.names(A) <- c("mean_tree", "min_tree", "max_tree", "sum_tree")

```

A

OUTPUT:

```

> # Check if the object is a data frame and convert it if necessary
> if (!is.data.frame(trees)) {
+   trees <- as.data.frame(trees)
+ }
>
> # Create the new data frame A
> A <- data.frame(Girth = c(mean(trees$Girth), min(trees$Girth),
+                           Height = c(mean(trees$Height), min(trees$Height),
+                           Volume = c(mean(trees$Volume), min(trees$Volume),
+ >
> # Assign row names to A
> row.names(A) <- c("mean_tree", "min_tree", "max_tree", "sum_tree")
>
> A
      Girth Height     Volume
mean_tree 13.24839    76 30.17097
min_tree   8.30000    63 10.20000
max_tree  20.60000    87 77.00000
sum_tree  410.70000   2356 935.30000
> |

```

RESULT:

7. Consider the data frame A:

- Order the entire data frame by the first column.
- Rename the row names as follows: mean, min, max, tree

AIM:

PROCEDURE:

CODE:

```

# Check if the object is a data frame and convert it if necessary
if (!is.data.frame(trees)) {
  trees <- as.data.frame(trees)
}
# Create the new data frame A
A <- data.frame(Girth = c(mean(trees$Girth), min(trees$Girth), max(trees$Girth), sum(trees$Girth)),
                 Height = c(mean(trees$Height), min(trees$Height), max(trees$Height), sum(trees$Height)),
                 Volume = c(mean(trees$Volume), min(trees$Volume), max(trees$Volume), sum(trees$Volume)))
# Assign row names to A
row.names(A) <- c("mean_tree", "min_tree", "max_tree", "sum_tree")
# Order the data frame by the first column
A_ordered <- A[order(A[, 1]), ]
# Rename the row names
row.names(A_ordered) <- c("mean", "min", "max", "tree")
A_ordered

```

OUTPUT:

The screenshot shows the RStudio interface with two panes. The left pane contains the R code, and the right pane shows the console output.

```

> # Check if the object is a data frame and convert it if necessary
> if (!is.data.frame(trees)) {
+   trees <- as.data.frame(trees)
+ }
>
> # Create the new data frame A
> A <- data.frame(Girth = c(mean(trees$Girth),
+                           Height = c(mean(trees$Height),
+                           Volume = c(mean(trees$Volume),
+
> # Assign row names to A
> row.names(A) <- c("mean_tree", "min_tree",
> # Order the data frame by the first column
> A_ordered <- A[order(A[, 1]), ]
>
> # Rename the row names
> row.names(A_ordered) <- c("mean", "min", "max", "tree")
>
> A_ordered
      Girth Height     Volume
mean    8.30000   63 10.20000
min    13.24839    76 30.17097
max    20.60000    87 77.00000
tree  410.70000  2356 935.30000
> |

```

RESULT:**8. Create a data frame XY**

```

X=c(1,2,3,1,4,5,2)
Y=c(0,3,2,0,5,9,3)
> XY
X Y
1 0
2 3
3 2
1 0
4 5
5 9
2 3

```

look at duplicated elements using a provided R function. keep only the unique lines on XY using a provided R function.

AIM:**PROCEDURE:**

CODE:

```
# Create the vectors X and Y
X <- c(1, 2, 3, 1, 4, 5, 2)
Y <- c(0, 3, 2, 0, 5, 9, 3)
# Create the data frame XY
XY <- data.frame(X, Y)
# Check for duplicated elements using the duplicated() function
duplicated_rows <- duplicated(XY)
# Keep only the unique lines in XY using the unique() function
XY_unique <- unique(XY)
XY_unique
```

```
> # Create the vectors X and Y
> X <- c(1, 2, 3, 1, 4, 5, 2)
> Y <- c(0, 3, 2, 0, 5, 9, 3)
>
> # Create the data frame XY
> XY <- data.frame(X, Y)
>
> # Check for duplicated
> duplicated_rows <- dupl
>
> # Keep only the unique
> XY_unique <- unique(XY)
>
> XY_unique
   X   Y
1 1  0
2 2  3
3 3  2
4 4  5
5 5  9
6 5  9
> |
```

9. Use the (built-in) dataset Titanic.

Make sure the object is a data frame, if not change it to a data frame.

Define a data frame with value 1st in Class variable, and value NO in Survived variable and variables Sex, Age and Freq.

Sex Age Freq

1	Male	Child	0
5	Female	Child	0
9	Male	Adult	118
13	Female	Adult	4

AIM:

PROCEDURE:**CODE:**

```
# Check if the object is a data frame and convert it if necessary
if (!is.data.frame(Titanic)) {
  Titanic <- as.data.frame(Titanic)
}
```

```
# Define the new data frame with the specified values
new_df <- data.frame(Sex = c("Male", "Female", "Male", "Female"),
                      Age = c("Child", "Child", "Adult", "Adult"),
                      Freq = c(0, 0, 118, 4),
                      stringsAsFactors = FALSE)
```

```
# Set the Class variable to "1st"
new_df$Class <- "1st"
```

```
# Set the Survived variable to "NO"
new_df$Survived <- "NO"
```

new_df

OUTPUT:

```
> # Check if the object is a data frame and convert it if necessary
> if (!is.data.frame(Titanic)) {
+   Titanic <- as.data.frame(Titanic)
+ }
>
> # Define the new data frame with the specified values
> new_df <- data.frame(Sex = c("Male", "Female", "Male", "Female"),
+                       Age = c("Child", "Child", "Adult", "Adult"),
+                       Freq = c(0, 0, 118, 4),
+                       stringsAsFactors = FALSE)
>
> # Set the Class variable to "1st"
> new_df$Class <- "1st"
>
> # Set the Survived variable to "NO"
> new_df$Survived <- "NO"
>
> new_df
  Sex   Age Freq Class Survived
1  Male Child    0  1st      NO
2 Female Child    0  1st      NO
3  Male Adult   118  1st      NO
4 Female Adult     4  1st      NO
> |
```

R C:\Users\NIKITHA\Desktop\R programs\Day 2 lab\Day 4 lab\EX E\EX 9.R - R Editor

```
# Check if the object is a data frame and convert it if necessary
if (!is.data.frame(Titanic)) {
  Titanic <- as.data.frame(Titanic)
}

# Define the new data frame with the specified values
new_df <- data.frame(Sex = c("Male", "Female", "Male", "Female"),
                      Age = c("Child", "Child", "Adult", "Adult"),
                      Freq = c(0, 0, 118, 4),
                      stringsAsFactors = FALSE)

# Set the Class variable to "1st"
new_df$Class <- "1st"

# Set the Survived variable to "NO"
new_df$Survived <- "NO"

new_df
```

RESULT:

MERGING DATAFRAMES

10 a) Create the following dataframes to merge:

```
buildings<- data.frame(location=c(1, 2, 3), name=c("building1", "building2", "building3"))
```

```
data <- data.frame(survey=c(1,1,1,2,2,2),location=c(1,2,3,2,3,1),efficiency=c(51,64,70,7,80,58))
```

The dataframes, buildings and data have a common key variable called, “location”.

Use the merge() function to merge the two dataframes by “location”, into a new dataframe, “buildingStats”.

AIM:

PROCEDURE:

CODE:

```
# Create the buildings data frame
```

```
buildings <- data.frame(location = c(1, 2, 3), name = c("building1", "building2", "building3"))
```

```
# Create the data data frame
```

```
data <- data.frame(survey = c(1, 1, 1, 2, 2, 2),
                   location = c(1, 2, 3, 2, 3, 1),
```

```

efficiency = c(51, 64, 70, 7, 80, 58))
# Merge the buildings and data data frames by "location"
buildingStats <- merge(buildings, data, by = "location")
buildingStats

```

OUTPUT:

```

> # Create the buildings data frame
> buildings <- data.frame(location = c(1, 2, 3), name = c("building1", "building2", "building3"))
>
> # Create the data data frame
> data <- data.frame(survey = c(1, 1, 1, 2, 2, 2),
+                      location = c(1, 2, 3, 2, 3, 1),
+                      efficiency = c(51, 64, 70, 7, 80, 58))
>
> # Merge the buildings and data data frames by "location"
> buildingStats <- merge(buildings, data, by = "location")
>
> buildingStats
   location      name survey efficiency
1       1 building1      1      51
2       1 building1      2      58
3       2 building2      1      64
4       2 building2      2       7
5       3 building3      1      70
6       3 building3      2      80

```

RESULT:

b) Give the dataframes different key variable names:

```

buildings<- data.frame(location=c(1, 2, 3), name=c("building1", "building2", "building3"))
data <- data.frame(survey=c(1,1,1,2,2,2), LocationID=c(1,2,3,2,3,1), efficiency=c(51,64,70,71,80,58))
The dataframes, buildings and data have corresponding variables called, location, and LocationID. Use the merge() function to merge the columns of the two dataframes by the corresponding variables.

```

AIM:

PROCEDURE:

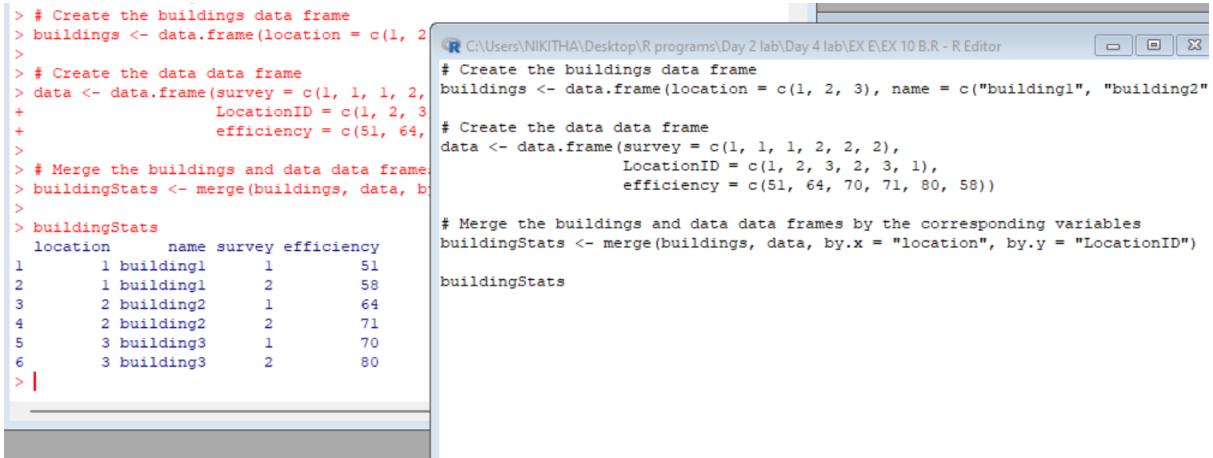
CODE:

```

# Create the buildings data frame
buildings <- data.frame(location = c(1, 2, 3), name = c("building1", "building2", "building3"))
# Create the data data frame
data <- data.frame(survey = c(1, 1, 1, 2, 2, 2),
                   LocationID = c(1, 2, 3, 2, 3, 1),
                   efficiency = c(51, 64, 70, 71, 80, 58))
# Merge the buildings and data data frames by the corresponding variables
buildingStats <- merge(buildings, data, by.x = "location", by.y = "LocationID")
buildingStats

```

OUTPUT:



```

> # Create the buildings data frame
> buildings <- data.frame(location = c(1, 2,
>
> # Create the data data frame
> data <- data.frame(survey = c(1, 1, 1, 2,
+                     LocationID = c(1, 2, 3
+                     efficiency = c(51, 64,
>
> # Merge the buildings and data data frame
> buildingStats <- merge(buildings, data, by =
>
> buildingStats
location      name survey efficiency
1      1 building1     1      51
2      1 building1     2      58
3      2 building2     1      64
4      2 building2     2      71
5      3 building3     1      70
6      3 building3     2      80
> |

```

RESULT:

DIFFERENT TYPES OF MERGE IN R

11 a) InnerJoin:

The R merge() function automatically joins the frames by common variable names. In that case, demonstrate how you would perform the merge in Exercise 10 a without specifying the key variable.

AIM:

PROCEDURE:

CODE:

```

# Create the buildings data frame
buildings <- data.frame(location = c(1, 2, 3), name = c("building1", "building2", "building3"))
# Create the data data frame
data <- data.frame(survey = c(1, 1, 1, 2, 2, 2),
                   location = c(1, 2, 3, 2, 3, 1),
                   efficiency = c(51, 64, 70, 7, 80, 58))
# Merge the data frames by "location" using explicit key variable specification
buildingStats <- merge(buildings, data, by = "location")
buildingStats
# Merge the data frames without specifying the key variable
buildingStats_auto <- merge(buildings, data)
buildingStats_auto

```

OUTPUT:

```

>
> # Merge the data frames by "location" using explicit key variable specification
> buildingStats <- merge(buildings, data, by = "location")
>
> buildingStats
  location name survey efficiency
1       1 building1    1      51
2       1 building1    2      58
3       2 building2    1      64
4       2 building2    2       7
5       3 building3    1      70
6       3 building3    2      80
>
> # Merge the data frames without specifying the key variable
> buildingStats_auto <- merge(buildings, data)
>
> buildingStats_auto
  location name survey efficiency
1       1 building1    1      51
2       1 building1    2      58
3       2 building2    1      64
4       2 building2    2       7
5       3 building3    1      70
6       3 building3    2      80
> |

```

RESULT:

b)OuterJoin:

Merge the two dataframes from Exercise 10 a. Use the “all=” parameter in the merge() function to return all records from both tables. Also, merge with the key variable, “location”.

AIM:

PROCEDURE:

CODE:

```

# Create the buildings data frame
buildings <- data.frame(location = c(1, 2, 3), name = c("building1", "building2", "building3"))
# Create the data data frame
data <- data.frame(survey = c(1, 1, 1, 2, 2, 2),
                   location = c(1, 2, 3, 2, 3, 1),
                   efficiency = c(51, 64, 70, 7, 80, 58))
# Merge the data frames by "location" using an outer join
buildingStats <- merge(buildings, data, by = "location", all = TRUE)
buildingStats

```

OUTPUT:

```

> 
> # Create the buildings data frame
> buildings <- data.frame(location = c(1, 2, 3), name = c("building1", "building2", "building3"))
>
> # Create the data data frame
> data <- data.frame(survey = c(1, 1, 1, 2, 2, 2),
+                     location = c(1, 2, 3, 2, 3, 1),
+                     efficiency = c(51, 64, 70, 7, 80, 58))
>
> # Merge the data frames by "location" using an outer join
> buildingStats <- merge(buildings, data, by = "location", all = TRUE)
>
> buildingStats
  location name survey efficiency
1       1 building1    1      51
2       1 building1    2      58
3       2 building2    1      64
4       2 building2    2       7
5       3 building3    1      70
6       3 building3    2      80
> |

```

c)Left Join:

Merge the two dataframes from Exercise 10 a, and return all rows from the left table. Specify the matching key from Exercise 10 a.

AIM:

PROCEDURE:

CODE:

```
# Create the buildings dataframe
buildings <- data.frame(location = c(1, 2, 3), name = c("building1", "building2", "building3"))
# Create the data dataframe
data <- data.frame(survey = c(1, 1, 1, 2, 2, 2), location = c(1, 2, 3, 2, 3, 1), efficiency = c(51, 64, 70, 7, 80, 58))
# Merge the dataframes using left join
buildingStats <- merge(buildings, data, by = "location", all.x = TRUE)
# Print the merged dataframe
BuildingStats
```

OUTPUT:

[Previously saved workspace restored]

location	name	survey	efficiency
1	building1	1	51
2	building1	2	58
3	building2	1	64
4	building2	2	7
5	building3	1	70
6	building3	2	80

RESULT:**d)Right Join:**

Merge the two dataframes from Exercise 10a, and return all rows from the right table. Use the matching key from Exercise 10a to return matching rows from the left table.

AIM:**PROCEDURE:****CODE:**

```
# Create the buildings dataframe
buildings <- data.frame(location = c(1, 2, 3), name = c("building1", "building2", "building3"))
# Create the data dataframe
data <- data.frame(survey = c(1, 1, 1, 2, 2, 2), location = c(1, 2, 3, 2, 3, 1), efficiency = c(51, 64, 70, 7, 80, 58))
# Merge the dataframes using right join
buildingStats <- merge(buildings, data, by = "location", all.y = TRUE)
# Print the merged dataframe
BuildingStats
```

OUTPUT:

```

> # Create the buildings dataframe
> buildings <- data.frame(location = c(1, 2, 3),
>
> # Create the data dataframe
> data <- data.frame(survey = c(1, 1, 1, 2, 2),
>
> # Merge the dataframes using right join
> buildingStats <- merge(buildings, data, by =
>
> # Print the merged dataframe
> buildingStats
  location     name survey efficiency
1      1 building1     1       51
2      1 building1     2       58
3      2 building2     1       64
4      2 building2     2        7
5      3 building3     1       70
6      3 building3     2       80
> |

```

e) Cross Join:

Merge the two dataframes from Exercise 10a, into a “Cross Join” with each row of “buildings” matched to each row of “data”. What new column names are created in “buildingStats”?

AIM:

PROCEDURE:

CODE:

```

# Create the buildings dataframe
buildings <- data.frame(location = c(1, 2, 3), name = c("building1", "building2", "building3"))
# Create the data dataframe
data <- data.frame(survey = c(1, 1, 1, 2, 2, 2), location = c(1, 2, 3, 2, 3, 1), efficiency = c(51, 64, 70, 7, 80, 58))
# Merge the dataframes using cross join
buildingStats <- merge(buildings, data, by=NULL)
# Print the merged dataframe
BuildingStats

```

OUTPUT:

	location.x	name	survey	location.y	efficiency
1	1	building1	1	1	51
2	2	building2	1	1	51
3	3	building3	1	1	51
4	1	building1	1	2	64
5	2	building2	1	2	64
6	3	building3	1	2	64
7	1	building1	1	3	70
8	2	building2	1	3	70
9	3	building3	1	3	70
10	1	building1	2	2	7
11	2	building2	2	2	7
12	3	building3	2	2	7
13	1	building1	2	3	80
14	2	building2	2	3	80
15	3	building3	2	3	80
16	1	building1	2	1	58
17	2	building2	2	1	58
18	3	building3	2	1	58

RESULT:

12 Merging Dataframe rows:

To join two data frames (datasets) vertically, use the rbind function. The two data frames must have the same variables, but they do not have to be in the same order.

Merge the rows of the following two dataframes:

```
buildings<- data.frame(location=c(1, 2, 3), name=c("building1",
"building2", "building3"))
buildings2 <- data.frame(location=c(5, 4, 6), name=c("building5", "building4", "building6"))
Also, specify the new dataframe as, "allBuildings".
```

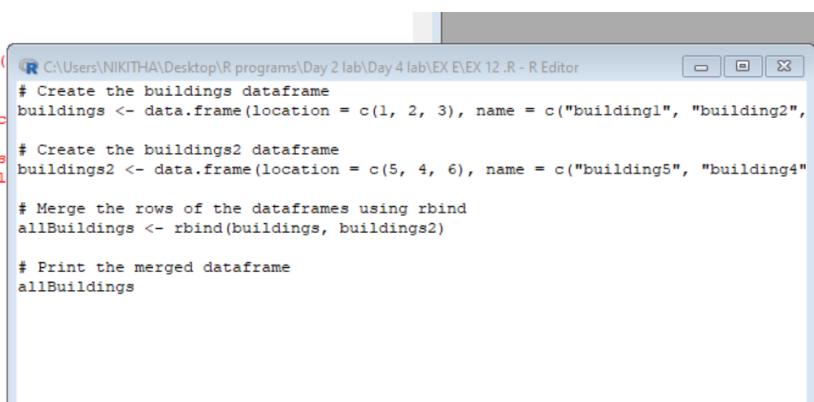
AIM:

PROCEDURE:

CODE:

```
# Create the buildings dataframe
buildings <- data.frame(location = c(1, 2, 3), name = c("building1", "building2", "building3"))
# Create the buildings2 dataframe
buildings2 <- data.frame(location = c(5, 4, 6), name = c("building5", "building4", "building6"))
# Merge the rows of the dataframes using rbind
allBuildings <- rbind(buildings, buildings2)
# Print the merged dataframe
AllBuildings
```

OUTPUT:



The screenshot shows an RStudio session window titled 'C:\Users\NIKITHA\Desktop\R programs\Day 2 lab\Day 4 lab\EX E\EX 12.R - R Editor'. The code is identical to the one in the 'CODE' section. The output pane displays the resulting data frame:

location	name
1	building1
2	building2
3	building3
4	building5
5	building4
6	building6

RESULT:

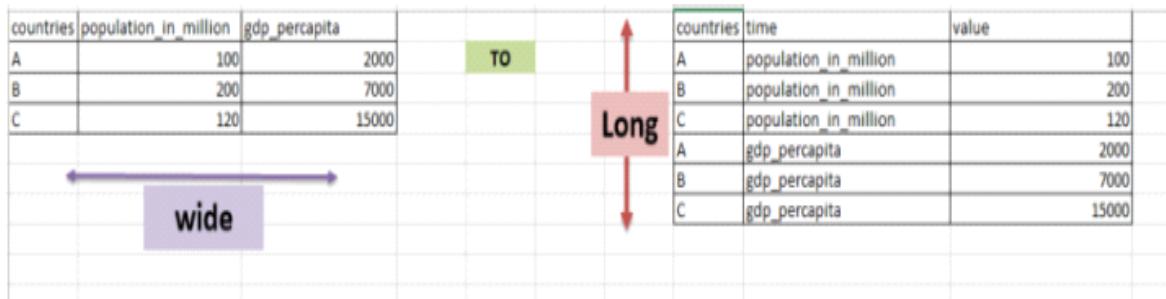
F. RESHAPE FUNCTION IN R

- Construct the following data frame ‘country’.

	countries	value.population_in_million	value.gdp_per capita
1	A	100	2000
2	B	200	7000
3	C	120	15000

- Reshape in R from wide to long:

Reshape the above data frame from wide to long format in R.



- data frame “country” is passed to reshape function
- idvar is the variable which need to be left unaltered which is “countries”
- varying are the ones that needs to converted from wide to long
- v.names are the values that should be against the times in the resultant [data frame](#).
- new.row.names is used to assign row names to the resultant dataset
- direction is, to which format the data needs to be transformed

Aim:

Procedure:

Code:

```
> country<-data.frame(c("A","B","C"),c(100,200,120),c(2000,7000,15000))
> colnames(country)<- c("countries","population_in_million","gdp_per capita")
> country
```

OUTPUT:

```
countries population_in_million gdp_per capita
1      A           100        2000
2      B           200        7000
3      C           120       15000
```

Code: Reshape in R from wide to long

```
> country_w_to_L<- reshape(data=country, idvar="countries",
+                             varying = c("population_in_million","gdp_per capita"),
+                             v.name=c("value"),
+                             times=c("population_in_million","gdp_per capita"),
+                             new.row.names = 1:1000,
+                             direction="long")
```

> country_w_to_L

OUTPUT:

```
countries          time value
1      A population_in_million 100
2      B population_in_million 200
```

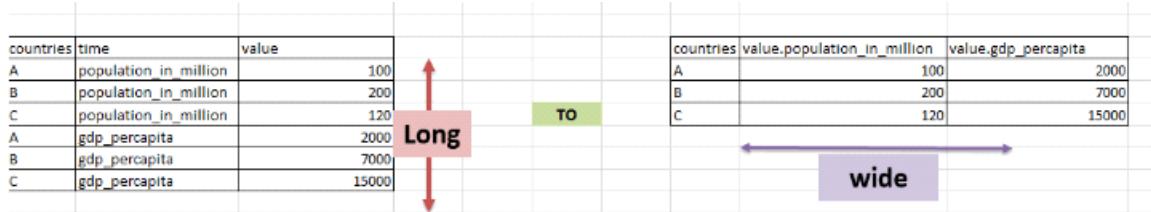
```

3   C population_in_million 120
4   A     gdp_percapita 2000
5   B     gdp_percapita 7000
6   C     gdp_percapita 15000

```

Result:

- **Reshape in R from long to wide:**



- data (country_w_to_L) which is in long format, is passed to reshape function
- idvar is the variable which need to be left unaltered, which is “countries”
- timevar are the variables that needs to converted to wide format
- v.names are the value variable
- direction is, to which format the data needs to be transformed

Aim:

Procedure:

Code:

```

library(tidyr)
data_long = gather(country, detail, value, population_in_million:gdp_percapita, factor_key=TRUE)
data_long

```

OUTPUT:

	countries	detail	value
1	A	population_in_million	100
2	B	population_in_million	200
3	C	population_in_million	120
4	A	gdp_percapita	2000
5	B	gdp_percapita	7000
6	C	gdp_percapita	15000

Result:

G. MELTING AND CASTING IN R

1. Melt airquality data set and display as a long – format data ?

Aim :

Procedure :

Code & Output :

```
> library(reshape2)
>
> # Load the airquality dataset
> data(airquality)
>
> # Melt the dataset
> airquality_long <- melt(airquality, id.vars = c("Month", "Day"))
>
> # View the first few rows of the long-format dataset
> head(airquality_long)
  Month Day variable value
1      5    1   Ozone    41
2      5    2   Ozone    36
3      5    3   Ozone    12
4      5    4   Ozone    18
5      5    5   Ozone     NA
6      5    6   Ozone    28
> |
```

RESULT :

2.Melt airquality data and specify month and day to be “ID variables” ?

AIM :

Procedure :

Code & Output :

```

> library(reshape2)
>
> # Load the airquality dataset
> data(airquality)
>
> # Melt the dataset and specify Month and Day as ID variables
> airquality_long <- melt(airquality, id.vars = c("Month", "Day"), variable.name = "Measurements", value.name = "Values")
>
> # View the first few rows of the long-format dataset
> head(airquality_long)
  Month Day Measurements Values
1      5    1        Ozone     41
2      5    2        Ozone     36
3      5    3        Ozone     12
4      5    4        Ozone     18
5      5    5        Ozone      NA
6      5    6        Ozone     28
> |

```

Result :

3.Cast the molten airquality data set .

Aim :

Procedure :

Code & Output :

```

> library(reshape2)
>
> # Load the airquality dataset
> data(airquality)
>
> # Melt the dataset
> airquality_long <- melt(airquality, id.vars = c("Month", "Day"))
>
> # Cast the dataset back to its original wide-format
> airquality_wide <- dcast(airquality_long, Month + Day ~ variable)
>
> # View the first few rows of the wide-format dataset
> head(airquality_wide)
  Month Day Ozone Solar.R Wind Temp
1      5    1     41     190   7.4   67
2      5    2     36     118   8.0   72
3      5    3     12     149  12.6   74
4      5    4     18     313  11.5   62
5      5    5     NA     NA  14.3   56
6      5    6     28     NA  14.9   66
> |

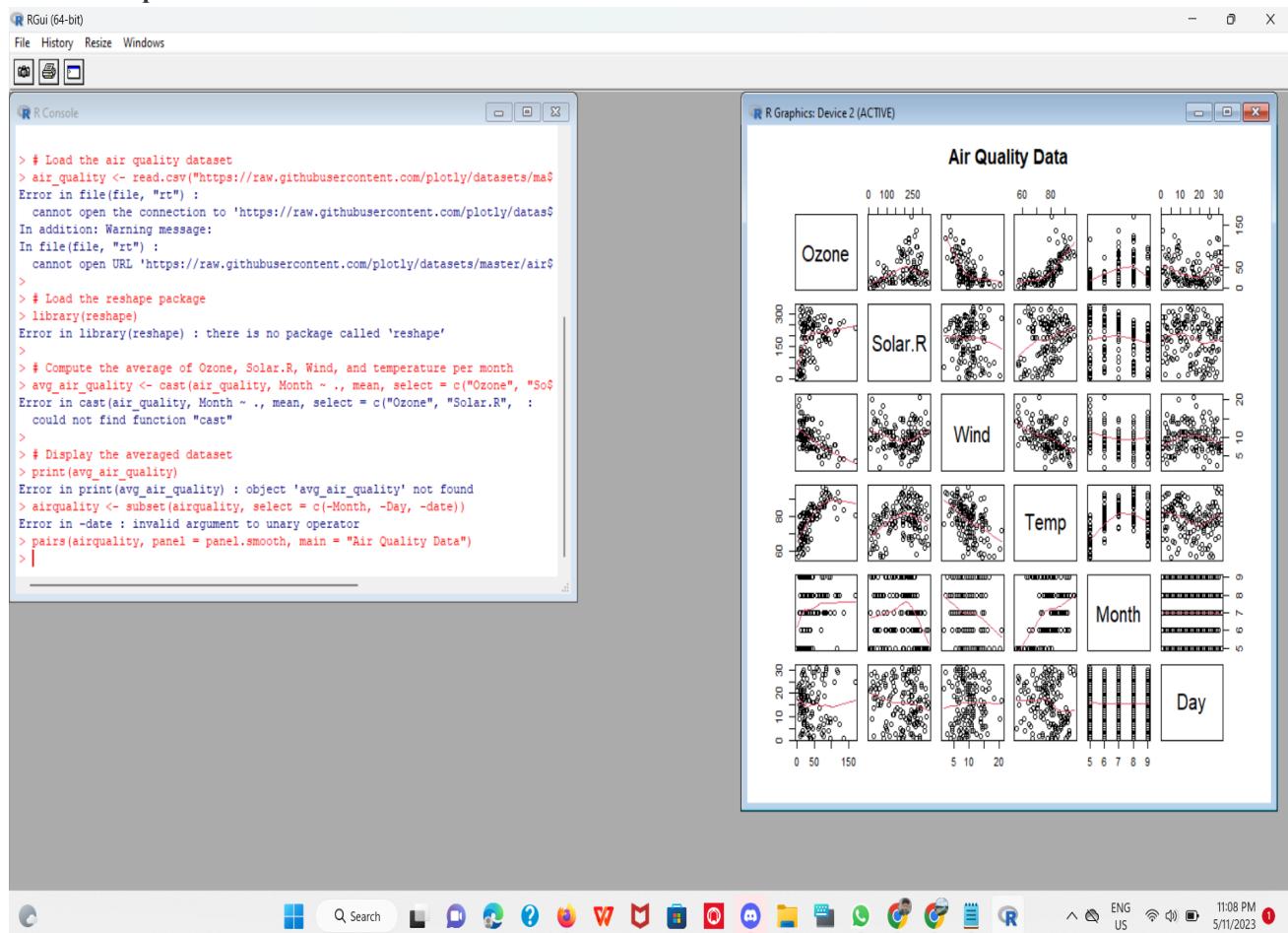
```

Result :

4.Use cast function appropriately and compute the average of Ozone, Solar.R , Wind and temperature per month ?
Aim :

Procedure :

Code & Output :



Result :

H. FILE MANUPULATION IN R

1. Consider the following data present. Create this file using windows notepad . Save the file as **input.csv** using the save As All files(*.*) option in notepad.

```
id,name,salary,start_date,dept
1,Rick,623.3,2012-01-01,IT
2,Dan,515.2,2013-09-23,Operations
3,Michelle,611,2014-11-15,IT
4,Ryan,729,2014-05-11,HR
5,Gary,843.25,2015-03-27,Finance
6,Nina,578,2013-05-21,IT
7,Simon,632.8,2013-07-30,Operations
8,Guru,722.5,2014-06-17,Finance
```

Aim:

Procedure:

Program :

```
data <- read.csv("input.csv")
print(data)
```

Output:

The screenshot shows two windows from the R environment. On the left, the R Console window displays the R startup message and the command history for reading the CSV file and printing its contents. The output shows the data frame 'data' with 8 rows and 5 columns. On the right, the R Editor window shows the same R code used to generate the output.

```
R Console
Natural language support but running in an English locale
R is a collaborative project with many contributors.
Type 'contributors()' for more information and
'citation()' on how to cite R or R packages in publications.

Type 'demo()' for some demos, 'help()' for on-line help, or
'help.start()' for an HTML browser interface to help.
Type 'q()' to quit R.

[Previously saved workspace restored]
> data <- read.csv("input.csv")
>
> print(data)
   id   name salary start_date     dept
1  1    Rick  623.30 2012-01-01       IT
2  2     Dan  515.20 2013-09-23 Operations
3  3 Michelle  611.00 2014-11-15       IT
4  4    Ryan  729.00 2014-05-11      HR
5  5    Gary 843.25 2015-03-27 Finance
6  6    Nina  578.00 2013-05-21       IT
7  7   Simon  632.80 2013-07-30 Operations
8  8    Guru  722.50 2014-06-17 Finance
```

```
R Editor
data <- read.csv("input.csv")
print(data)
```

Result:

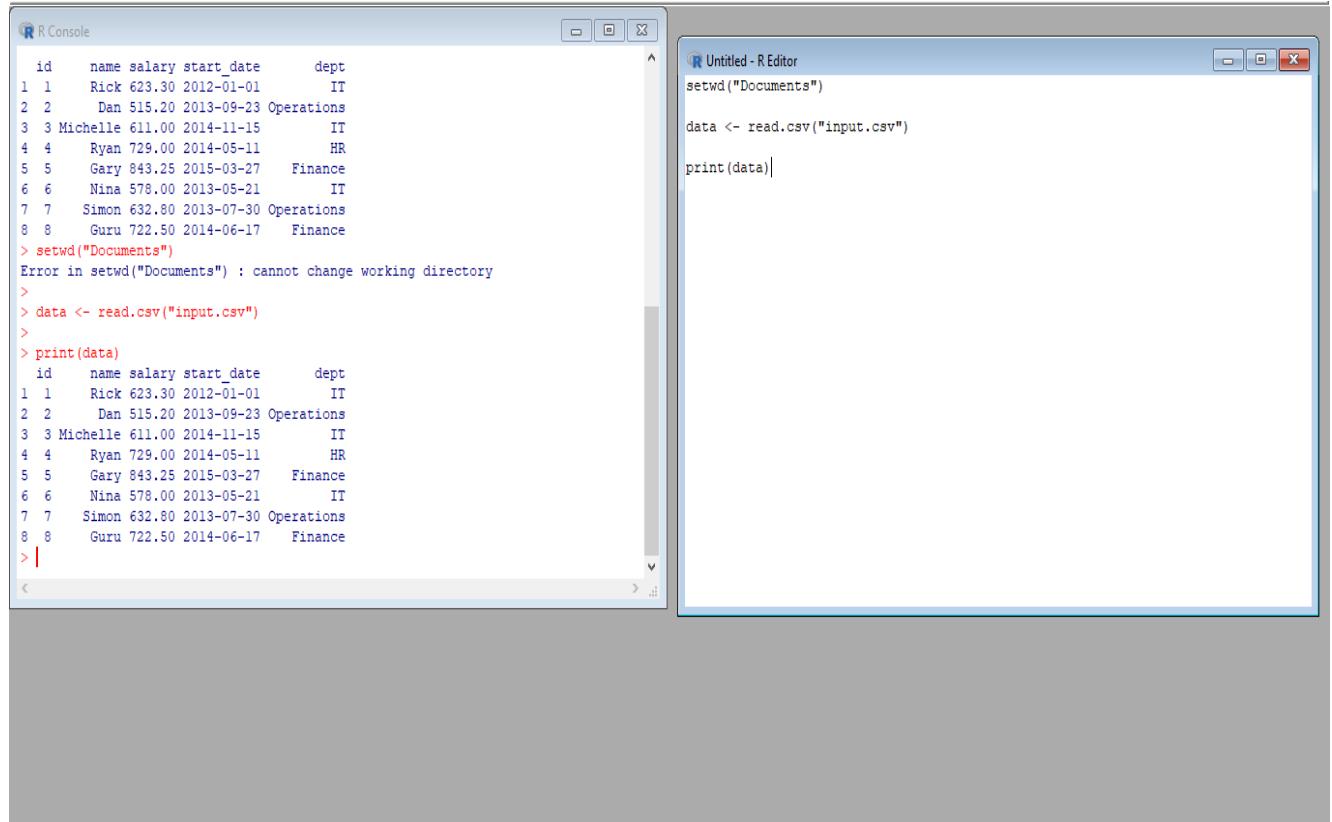
2. Use appropriate R commands to read **input.csv** file.

Aim:

Procedure:

Program:

```
setwd("Documents")
data <- read.csv("input.csv")
print(data)
```

Output:

The screenshot shows the RStudio interface with two main windows. On the left is the 'R Console' window, which displays the R code and its output. The output shows a data frame with columns: id, name, salary, start_date, and dept. The data consists of 8 rows of employee information. An error message is also present. On the right is the 'Untitled - R Editor' window, which contains the same R code as the console.

```
R Console
id    name salary start_date      dept
1 1   Rick 623.30 2012-01-01      IT
2 2   Dan  515.20 2013-09-23 Operations
3 3 Michelle 611.00 2014-11-15      IT
4 4   Ryan 729.00 2014-05-11      HR
5 5   Gary 843.25 2015-03-27 Finance
6 6   Nina 578.00 2013-05-21      IT
7 7 Simon 632.80 2013-07-30 Operations
8 8   Guru 722.50 2014-06-17 Finance
> setwd("Documents")
Error in setwd("Documents") : cannot change working directory
>
> data <- read.csv("input.csv")
>
> print(data)
id    name salary start_date      dept
1 1   Rick 623.30 2012-01-01      IT
2 2   Dan  515.20 2013-09-23 Operations
3 3 Michelle 611.00 2014-11-15      IT
4 4   Ryan 729.00 2014-05-11      HR
5 5   Gary 843.25 2015-03-27 Finance
6 6   Nina 578.00 2013-05-21      IT
7 7 Simon 632.80 2013-07-30 Operations
8 8   Guru 722.50 2014-06-17 Finance
> |
```

```
Untitled - R Editor
setwd("Documents")
data <- read.csv("input.csv")
print(data)|
```

Result:

3. Analyze the CSV File and compute the following.
 - a. Get the maximum salary
 - b. Get the details of the person with max salary
 - c. Get all the people working in IT department
 - d. Get the persons in IT department whose salary is greater than 600
 - e. Get the people who joined on or after 2014

Aim:**Procedure:**

Program:

```
setwd("Documents")
data <- read.csv("input.csv")
a) max_salary <- max(data$salary)
print(paste("Maximum Salary:", max_salary))
b) person_with_max_salary <- data[data$salary == max_salary,]
print("Details of person with max salary:")
print(person_with_max_salary)
c) it_dept_employees <- data[data$dept == "IT",]
print("Employees working in IT department:")
print(it_dept_employees)
d) it_dept_high_salary_employees <- it_dept_employees[it_dept_employees$salary > 600,]
print("Employees working in IT department with salary greater than 600:")
print(it_dept_high_salary_employees)
e) employees_joined_after_2014 <- data[as.Date(data$start_date) >= as.Date("2014-01-01"),]
print("Employees who joined on or after 2014:")
print(employees_joined_after_2014)
```

Output:

```
a) > max_salary <- max(data$salary)
> print(paste("Maximum Salary:", max_salary))
[1] "Maximum Salary: 843.25"

b) [1] "Details of person with max salary:"
> print(person_with_max_salary)
  id name salary start_date dept
5 5 Gary 843.25 2015-03-27 Finance

[1] "Employees working in IT department:"
> print(it_dept_employees)
  id      name salary start_date dept
1 1      Rick 623.3 2012-01-01  IT
3 3 Michelle 611.0 2014-11-15  IT
6 6      Nina 578.0 2013-05-21  IT
c) ~

[1] "Employees working in IT department with salary greater than 600:"
> print(it_dept_high_salary_employees)
  id      name salary start_date dept
1 1      Rick 623.3 2012-01-01  IT
3 3 Michelle 611.0 2014-11-15  IT
d) >

[1] "Employees who joined on or after 2014:"
> print(employees_joined_after_2014)
  id      name salary start_date dept
3 3 Michelle 611.00 2014-11-15  IT
4 4      Ryan 729.00 2014-05-11  HR
5 5      Gary 843.25 2015-03-27 Finance
8 8      Guru 722.50 2014-06-17 Finance
e) > |
```

Result:

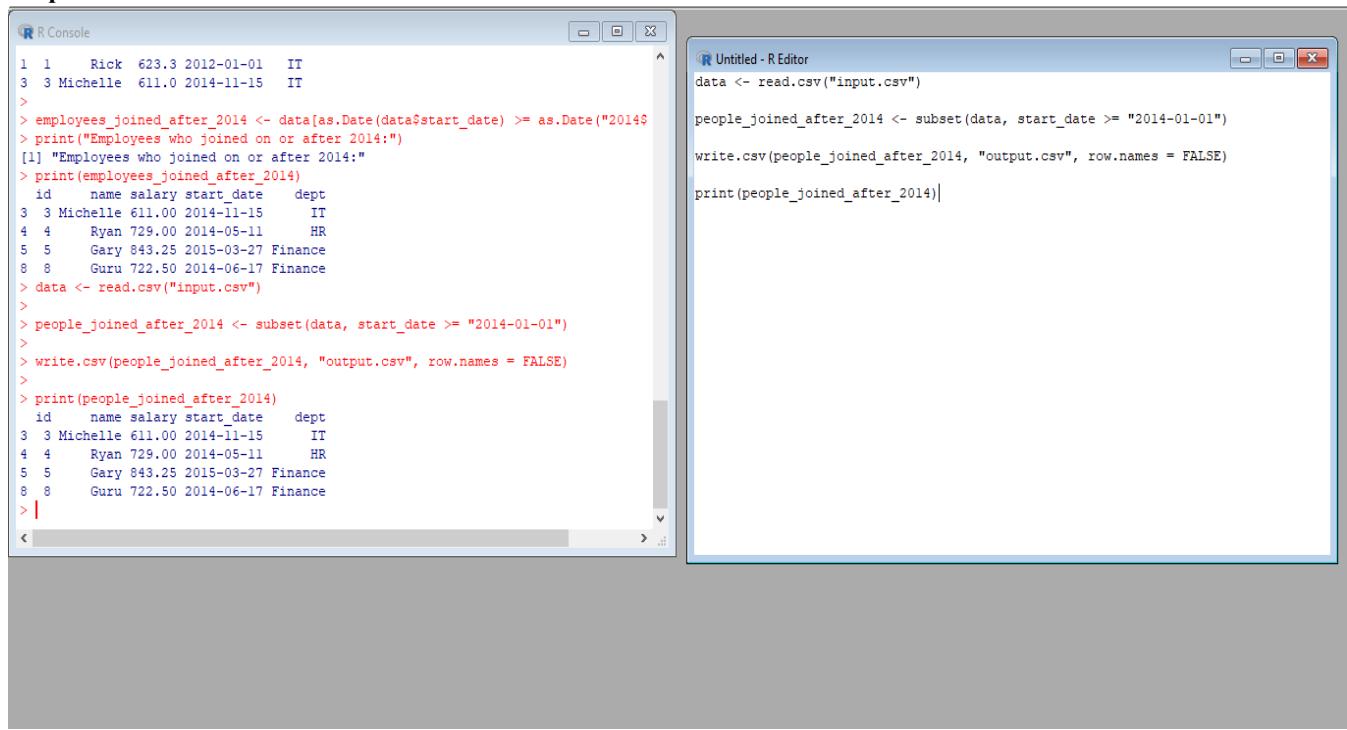
4. Get the people who joined on or after 2014 and write the output onto a file called output.csv
- Aim:**

Procedure:

Program:

```
data <- read.csv("input.csv")
people_joined_after_2014 <- subset(data, start_date >= "2014-01-01")
write.csv(people_joined_after_2014, "output.csv", row.names = FALSE)
print(people_joined_after_2014)
```

Output:



The screenshot shows the RStudio interface with two windows open. On the left is the 'R Console' window, which displays the R code and its output. The output shows a subset of data from 'input.csv' where 'start_date' is on or after '2014-01-01'. On the right is the 'Untitled - R Editor' window, which contains the same R code used to generate the output.

```
R Console
data <- read.csv("input.csv")
people_joined_after_2014 <- subset(data, start_date >= "2014-01-01")
write.csv(people_joined_after_2014, "output.csv", row.names = FALSE)
print(people_joined_after_2014)

[1] "Employees who joined on or after 2014:"
[1] "Rick 623.3 2012-01-01 IT
[1] 3 Michelle 611.0 2014-11-15 IT
>
> employees_joined_after_2014 <- data[as.Date(data$start_date) >= as.Date("2014-01-01")]
> print("Employees who joined on or after 2014:")
[1] "Employees who joined on or after 2014:"
> print(employees_joined_after_2014)
   id   name salary start_date   dept
3 3 Michelle 611.00 2014-11-15     IT
4 4   Ryan 729.00 2014-05-11     HR
5 5   Gary 843.25 2015-03-27 Finance
8 8   Guru 722.50 2014-06-17 Finance
> data <- read.csv("input.csv")
>
> people_joined_after_2014 <- subset(data, start_date >= "2014-01-01")
>
> write.csv(people_joined_after_2014, "output.csv", row.names = FALSE)
>
> print(people_joined_after_2014)
   id   name salary start_date   dept
3 3 Michelle 611.00 2014-11-15     IT
4 4   Ryan 729.00 2014-05-11     HR
5 5   Gary 843.25 2015-03-27 Finance
8 8   Guru 722.50 2014-06-17 Finance
> |
```

```
Untitled - R Editor
data <- read.csv("input.csv")
people_joined_after_2014 <- subset(data, start_date >= "2014-01-01")
write.csv(people_joined_after_2014, "output.csv", row.names = FALSE)
print(people_joined_after_2014)
```

Result:

I. UNIVARIATE ANALYSIS IN R - MEASURES OF CENTRAL TENDENCY

1. ARITHMETIC MEAN

a) Write suitable R code to compute the average of the following values 12,7,3,4.2,18,2,54, -21,8, -5

Aim:

Procedure:

CODE:

```
values<- c(12, 7, 3, 4.2, 18, 2, 54, -21, 8, -5)
mean_value <- mean(values)
print(mean_value)
```

OUTPUT:

```
[1] 8.22
```

b) Compute the mean after applying the trim option and removing 3 values from each end.

Aim:

Procedure:

CODE:

```
values <- c(12, 7, 3, 4.2, 18, 2, 54, -21, 8, -5)
trimmed_mean <- mean(values, trim = 0.3)
print(trimmed_mean)
```

OUTPUT:

```
[1] 5.55
```

c) Compute the mean of the following vector. (12,7,3,4.2,18,2,54,-21,8,-5,NA)

#If there are missing values, then the mean function returns NA.

Find mean dropping NA values.

#To drop the missing values from the calculation use na.rm = TRUE

Aim:

Procedure:

CODE:

```
values <- c(12, 7, 3, 4.2, 18, 2, 54, -21, 8, -5, NA)
mean_without_na <- mean(values, na.rm = TRUE)
print(mean_without_na)
```

OUTPUT:

```
[1] 8.22
```

RESULT:**2.MEDIAN**

Write suitable R code to compute the median of the following values. 12,7,3,4.2,18,2,54,-21,8,-5

AIM:**PROCEDURE:****CODE:**

```
a<- c(12, 7, 3, 4.2, 18, 2, 54, -21, 8, -5)
median_of_a <- median(a)
print(median_of_a)
```

OUTPUT:

```
[1] 5.6
```

RESULT:**3. MODE**

Calculate the mode for the following numeric as well as character data set in R.
 $(2,1,2,3,1,2,3,4,1,5,5,3,2,3)$, ("o","it","the","it","it")

AIM:**PROCEDURE:****CODE:**

```
get_mode <- function(data) {
  unique_values <- unique(data)
  frequencies <- tabulate(match(data, unique_values))
  mode <- unique_values[which.max(frequencies)]
  return(mode)
}
```

```
numeric_data <- c(2, 1, 2, 3, 1, 2, 3, 4, 1, 5, 5, 3, 2, 3)
```

```
mode_numeric <- get_mode(numeric_data)
print(mode_numeric)
character_data <- c("o", "it", "the", "it", "it")
mode_character <- get_mode(character_data)
print(mode_character)
```

OUTPUT:

```
> print(mode_numeric)
[1] 2
> print(mode_character)
[1] "it"
```

RESULT:

J. UNIVARIATE ANALYSIS IN R - MEASURES OF DISPERSION

1. Download mpg dataset which contains Fuel economy data from 1999 and 2008 for 38 popular models of car from the URL given below.

<https://vincentarelbundock.github.io/Rdatasets/datasets.html>

Aim:

Procedure:

Answer the following queries

- **Find the car which gives maximum city miles per gallon**

Code:

```
data("mtcars")
```

```
# Find the car with the maximum city miles per gallon  
max_city_mpg <- mtcars[which.max(mtcars$mpg), ]
```

```
# Print the car name and its city miles per gallon
```

```
cat("The car with the maximum city miles per gallon is", rownames(max_city_mpg), "with", max_city_mpg$mpg, "mpg.")
```

Output:

The car with the maximum city miles per gallon is Toyota Corolla with 33.9 mpg.

Result:

- **Find the cars which gives minimum disp in compact and subcompact class**

Code:

```
# Load the MTcars dataset
```

```
data(mtcars)
```

```
# Create a subset of the data containing only the compact and subcompact cars  
compact_cars <- subset(mtcars, cyl == 4 & (am == 0 | gear == 4))
```

```
# Find the car with the minimum displacement in the compact and subcompact classes  
min_disp_compact <- compact_cars[which.min(compact_cars$disp), ]
```

```
# Print the car names and their displacements
```

```
cat("The car(s) with the minimum displacement in the compact and subcompact classes is/are:")  
print(data.frame(disp = min_disp_compact$disp, car = rownames(min_disp_compact)))
```

Output:

The car(s) with the minimum displacement in the compact and subcompact classes is/are:

disp car

1 71.1 Toyota Corolla

Result:

2. Use the same dataset as used in Exercise 1 and perform the following queries

Aim:

Procedure:

- Find the standard deviation of city milles per gallon

Code:

```
# Load the MTcars dataset  
data(mtcars)  
# Calculate the standard deviation of city miles per gallon (mpg)  
sd_mpg_city <- sd(mtcars$mpg)  
  
# Print the standard deviation of city miles per gallon  
cat("The standard deviation of city miles per gallon is:", sd_mpg_city)
```

Output:

The standard deviation of city miles per gallon is: 6.026948

Result:

- **Find the variance of highway milles per gallon**

Code:

```
data("mtcars")  
# Calculate the variance of highway miles per gallon (mpg)  
var_mpg_highway <- var(mtcars$hp)  
  
# Print the variance of highway miles per gallon  
cat("The variance of highway miles per gallon is:", var_mpg_highway)
```

Output:

The variance of highway miles per gallon is: 4700.867

Result:

3. Use the same dataset and perform the following queries

Aim:

Procedure:

- **Find the range of the disp in the data set mpg**

Code:

```
# Load the MTcars dataset  
data(mtcars)  
  
# Calculate the range of displacement (disp) values in the dataset  
range_disp <- range(mtcars$disp)  
  
# Print the range of displacement values  
cat("The range of displacement values in the dataset is:", range_disp)
```

Output:

The range of displacement values in the dataset is: 71.1 472

Result:

- **Find the Quartile of the disp in the data set mpg**

Code:

```
# Load the MTcars dataset
```

```

data(mtcars)

# Calculate the quartiles of displacement (disp) values in the dataset
quartiles_disp <- quantile(mtcars$disp, probs = c(0.25, 0.5, 0.75))

# Print the quartiles of displacement values
cat("The quartiles of displacement values in the dataset are:")
print(quartiles_disp)

```

Output:

The variance of highway miles per gallon is: 4700.867

Result:

- **Find the IQR of the disp column in the data set mpg**

Code:

```

data("mtcars")
# Calculate the variance of highway miles per gallon (mpg)
var_mpg_highway <- var(mtcars$hp)
# Print the variance of highway miles per gallon
cat("The variance of highway miles per gallon is:", var_mpg_highway)

```

Output:

The quartiles of displacement values in the dataset are:

```

25%   50%   75%
120.825 196.300 326.000

```

Result:

4.

#Install Library

```
library(e1071)
```

- Find the skewness of city miles per mileage in the data set mpg ?
Use qplot function and display the graph for the city miles per mileage column

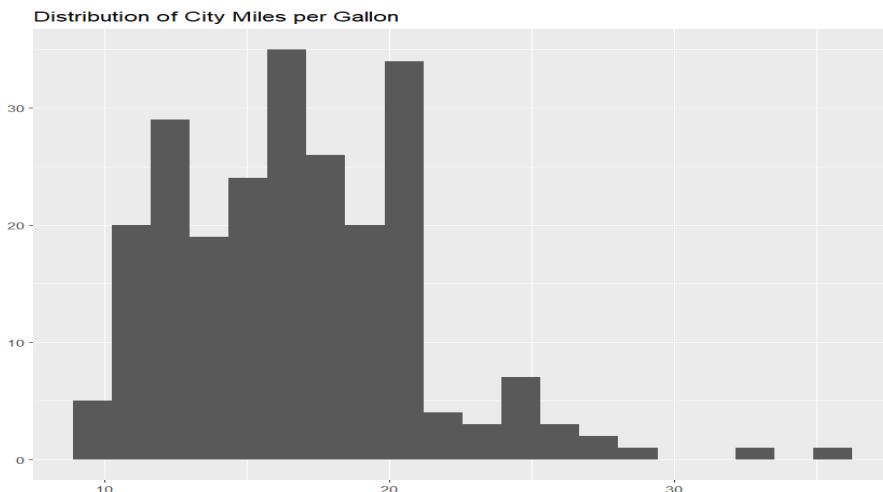
Code:

```

# Load the mpg dataset
library(ggplot2)
data(mpg)
# Calculate the skewness of city miles per gallon
library(moments)
skew_mpg_city <- skewness(mpg$cty)
# Print the skewness value
cat("The skewness of city miles per gallon is:", skew_mpg_city)
# Create a histogram of city miles per gallon
qplot(x = cty, data = mpg, geom = "histogram", bins = 20,
      main = "Distribution of City Miles per Gallon", xlab = "City MPG")

```

Output:



Result:

- Find the kurtosis of city miles per mileage in the data set mpg
Use qplot function and display the graph for the city miles per mileage column

Code:

```
# Load the mpg dataset
```

```
library(ggplot2)
```

```
data(mpg)
```

```
# Calculate the kurtosis of city miles per gallon
```

```
library(moments)
```

```
kurt_mpg_city <- kurtosis(mpg$cty)
```

```
# Print the kurtosis value
```

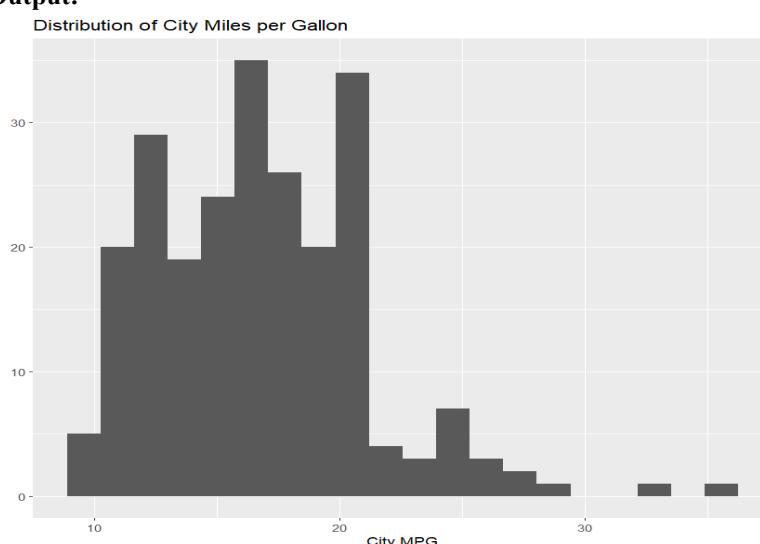
```
cat("The kurtosis of city miles per gallon is:", kurt_mpg_city)
```

```
# Create a histogram of city miles per gallon
```

```
qplot(x = cty, data = mpg, geom = "histogram", bins = 20,
```

```
main = "Distribution of City Miles per Gallon", xlab = "City MPG")
```

Output:



Result:

K. BIVARIATEANALYSIS IN R -COVARIANCE, CORRELATION, CROSSTAB

1 Reference Status Gender TestNewOrFollowUp
 1 KRXH Accepted Female Test1 New
 2 KRPT Accepted Male Test1 New
 3 FHRA Rejected Male Test2 New
 4 CZKK Accepted Female Test3 New
 5 CQTN Rejected Female Test1 New
 6 PZXW Accepted Female Test4 Follow-up
 7 SZRZ Rejected Male Test4 New
 8 RMZE Rejected Female Test2 New
 9 STNX Accepted Female Test3 New
 10 TMDW Accepted Female Test1 New
 i) Load the dataset and Create a data frame and name it as dataframe1
 ii) Load the function for crosstab
 xtabs(~colname , data=Data frame name)

AIM:

PROCEDURE:

CODE:

```
# load the data from dataframe1
dataframe1 <- data.frame(
  Reference = c("KRXH", "KRPT", "FHRA", "CZKK", "CQTN", "PZXW", "SZRZ", "RMZE", "STNX",
  "TMDW"),
  Status = c("Accepted", "Accepted", "Rejected", "Accepted", "Rejected", "Accepted", "Rejected", "Rejected",
  "Accepted", "Accepted"),
  Gender = c("Female", "Male", "Male", "Female", "Female", "Female", "Male", "Female", "Female",
  "Female"),
  TestNewOrFollowUp = c("Test1 New", "Test1 New", "Test2 New", "Test3 New", "Test1 New",
  "Follow-up", "Test4 New", "Test2 New", "Test3 New", "Test1 New")
)
```

```

# i) explore the relationship between two categorical variables
# contingency table between Gender and Status
gender_status <- table(dataframe1$Gender, dataframe1>Status)
print(gender_status)

# ii) create a table between Reference and Status
reference_status <- xtabs(~ Reference + Status, data = dataframe1)
print(reference_status)

# iii) save the file in the name of dataframe2
dataframe2 <- reference_status

```

OUTPUT:

The screenshot shows the R GUI interface. On the left, the R Console window displays R code and its output. The output includes two tables: one for gender status and one for reference status. On the right, the R Editor window shows the same R code. The taskbar at the bottom indicates it's running on a Windows 10 system.

```

RGui (64-bit)
File Edit Packages Windows Help
R R Console
> status_gender <- table(data>Status, data$Gender)
> reference_status <- table(data$Reference, data>Status)
>
> # print the results
> print(status_gender)
>
> Female Male
Accepted      5   1
Rejected      2   2
> print(reference_status)
>
> Accepted Rejected
CQTN          0     1
CZKK          1     0
FHRA          0     1
KRPT          1     0
KRXH          1     0
PZXW          1     0
RMZE          0     1
STNX          1     0
SZRZ          0     1
TMDW          1     0
> |
```

```

C:\Users\Admin\OneDrive\Documents\day 3 ex 8.R - R Editor
# create the data frame
data <- data.frame(
  Reference = c("KRXH", "KRPT", "FHRA", "CZKK", "CQTN", "PZXW", "SZRZ", "RMZE", "STNX"),
  Status = c("Accepted", "Accepted", "Rejected", "Accepted", "Rejected", "Accepted", "Rejected", "Accepted", "Accepted"),
  Gender = c("Female", "Male", "Male", "Female", "Female", "Male", "Female", "Male", "Female"),
  TestNewOrFollowUp = c("Test1 New", "Test1 New", "Test2 New", "Test3 New", "Test4 New", "Test5 New", "Test6 New", "Test7 New", "Test8 New")
)

# perform the crosstabulations
status_gender <- table(data>Status, data$Gender)
reference_status <- table(data$Reference, data>Status)

# print the results
print(status_gender)
print(reference_status)
```

RESULT:

2

- Use Two Categorical Variables and Discover the relationships within a dataset
- Next, using the xtabs() function, apply two variables from “dataframe1”, to create a table delineating the relationship between the “Reference” category, and the “Status” category.
- Save the file in the name of dataframe2

AIM:

PROCEDURE:

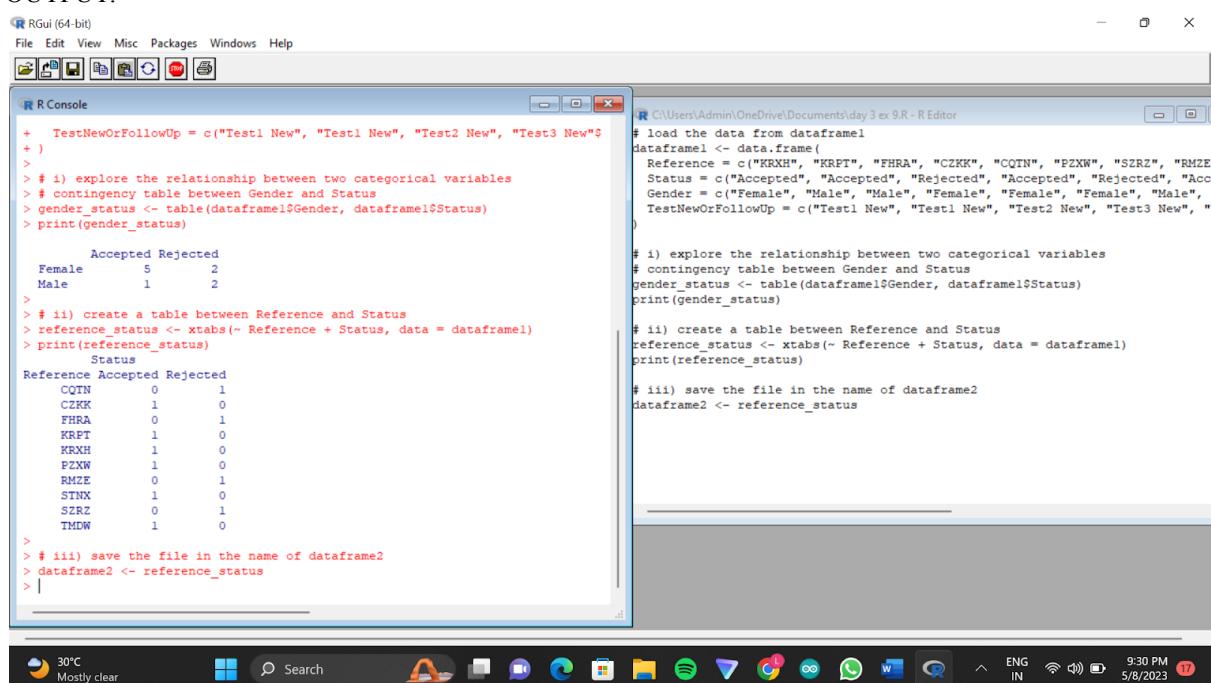
CODE:

```
# load the data from dataframe1
dataframe1 <- data.frame(
  Reference = c("KRXH", "KRPT", "FHRA", "CZKK", "CQTN", "PZXW", "SZRZ", "RMZE", "STNX", "TMDW"),
  Status = c("Accepted", "Accepted", "Rejected", "Accepted", "Rejected", "Accepted", "Rejected", "Accepted", "Accepted"),
  Gender = c("Female", "Male", "Male", "Female", "Female", "Male", "Female", "Female", "Female"),
  TestNewOrFollowUp = c("Test1 New", "Test1 New", "Test2 New", "Test3 New", "Test1 New", "Follow-up", "Test4 New", "Test2 New", "Test3 New", "Test1 New")
)

# i) explore the relationship between two categorical variables
# contingency table between Gender and Status
gender_status <- table(dataframe1$Gender, dataframe1>Status)
print(gender_status)

# ii) create a table between Reference and Status
reference_status <- xtabs(~ Reference + Status, data = dataframe1)
print(reference_status)

# iii) save the file in the name of dataframe2
dataframe2 <- reference_status
```

OUTPUT:

The screenshot shows the RGui interface with two windows open. The R Console window on the left displays the R code and its execution results. The R Editor window on the right shows the same code with syntax highlighting. The desktop taskbar at the bottom includes icons for weather, search, and various applications.

```
R Gui (64-bit)
File Edit View Misc Packages Windows Help
R Console
+  TestNewOrFollowUp = c("Test1 New", "Test1 New", "Test2 New", "Test3 New"
+ )
> # i) explore the relationship between two categorical variables
> # contingency table between Gender and Status
> gender_status <- table(dataframe1$Gender, dataframe1>Status)
> print(gender_status)
>
Accepted Rejected
Female      5      2
Male        1      2
>
> # ii) create a table between Reference and Status
> reference_status <- xtabs(~ Reference + Status, data = dataframe1)
> print(reference_status)
>
Status
Reference Accepted Rejected
CQTN      0      1
CZKK      1      0
FHRA      0      1
KRPT      1      0
KRXH      1      0
PZXW      1      0
RMZE      0      1
STNX      1      0
SZRZ      0      1
TMDW      1      0
>
> # iii) save the file in the name of dataframe2
> dataframe2 <- reference_status
>
```

```
C:\Users\Admin\OneDrive\Documents\day 3 ex 9.R - R Editor
# load the data from datafram1
datafram1 <- data.frame(
  Reference = c("KRXH", "KRPT", "FHRA", "CZKK", "CQTN", "PZXW", "SZRZ", "RMZE",
  Status = c("Accepted", "Accepted", "Rejected", "Accepted", "Rejected", "Accepted", "Rejected", "Accepted", "Accepted"),
  Gender = c("Female", "Male", "Male", "Female", "Female", "Male", "Female", "Female", "Female"),
  TestNewOrFollowUp = c("Test1 New", "Test1 New", "Test2 New", "Test3 New", "Test1 New", "Follow-up", "Test4 New", "Test2 New", "Test3 New", "Test1 New")
)

# i) explore the relationship between two categorical variables
# contingency table between Gender and Status
gender_status <- table(dataframe1$Gender, dataframe1>Status)
print(gender_status)

# ii) create a table between Reference and Status
reference_status <- xtabs(~ Reference + Status, data = datafram1)
print(reference_status)

# iii) save the file in the name of datafram2
dataframe2 <- reference_status
```

RESULT:**3**

Use the same data frame using three Categorical Variables create a Multi-Dimensional Table
Apply three variables from “dataframe1” to create a Multi-Dimensional Cross-Tabulation of “Status”, “Gender”, and “Test”.

AIM:**PROCEDURE:**

CODE:

```

dataframe2 <- dataframe1 # Copy the original data frame

# Create the multi-dimensional table
multi_table <- xtabs(~ Status + Gender + TestNewOrFollowUp, dataframe2)

```

View the table

multi_table

OUTPUT:

```

RGui (64-bit)
File Edit View Misc Packages Windows Help
R C:\Users\Admin\OneDrive\Documents\day 3 ex 10.R - R Editor
dataframe2 <- dataframe1 # Copy the original data frame
# Create the multi-dimensional table
multi_table <- xtabs(~ Status + Gender + TestNewOrFollowUp, dataframe2)

# View the table
multi_table

```

		Gender	
Status	Female	Male	
Accepted	1	0	
Rejected	0	0	

		Gender	
Status	Female	Male	
Accepted	2	1	
Rejected	1	0	

		Gender	
Status	Female	Male	
Accepted	0	0	
Rejected	1	1	

		Gender	
Status	Female	Male	
Accepted	2	0	
Rejected	0	0	

		Gender	
Status	Female	Male	
Accepted	2	0	
Rejected	0	0	

RESULT:**4 Covariance**

- For the Dataframe1 created from exercise 2 calculate the covariance between Reference column and Status column
- Display the covariance matrix

AIM:**PROCEDURE:****CODE:**

```

# Create the Dataframe1
Dataframe1 <- data.frame(
  Reference = c("KRXH", "KRPT", "FHRA", "CZKK", "CQTN", "PZXW", "SZRZ", "RMZE", "STNX", "TMDW"),
  Status = c("Accepted", "Accepted", "Rejected", "Accepted", "Rejected", "Accepted", "Rejected", "Rejected",
            "Accepted", "Accepted"),
  Gender = c("Female", "Male", "Male", "Female", "Female", "Female", "Male", "Female", "Female", "Female"))

```

```

TestNewOrFollowUp = c("Test1", "Test1", "Test2", "Test3", "Test1", "Test4", "Test4", "Test2", "Test3", "Test1")
)

# Convert categorical variables to numerical representations
Dataframe1$Reference <- as.numeric(factor(Dataframe1$Reference))
Dataframe1>Status <- as.numeric(factor(Dataframe1>Status))

# Calculate the covariance between Reference and Status columns
covariance <- cov(Dataframe1$Reference, Dataframe1>Status)

# Print the covariance
print(covariance)

```

OUTPUT:

RESULT:

5 Correlation

Find the Correlation between gender and status. what kind of correlation does exist between the two?

AIM:

PROCEDURE:

CODE:

```

# Create the dataset
Dataframe1 <- data.frame(
  Reference = c("KRXH", "KRPT", "FHRA", "CZKK", "CQTN", "PZXW", "SZRZ", "RMZE", "STNX", "TMDW"),
  Status = c("Accepted", "Accepted", "Rejected", "Accepted", "Rejected", "Accepted", "Rejected", "Accepted", "Accepted"),
  Gender = c("Female", "Male", "Male", "Female", "Female", "Male", "Female", "Female", "Female"),
  TestNewOrFollowUp = c("Test1", "Test1", "Test2", "Test3", "Test1", "Test4", "Test4", "Test2", "Test3", "Test1")
)

# Convert categorical variables to numerical representations
Dataframe1$Gender <- as.numeric(factor(Dataframe1$Gender))
Dataframe1>Status <- as.numeric(factor(Dataframe1>Status))

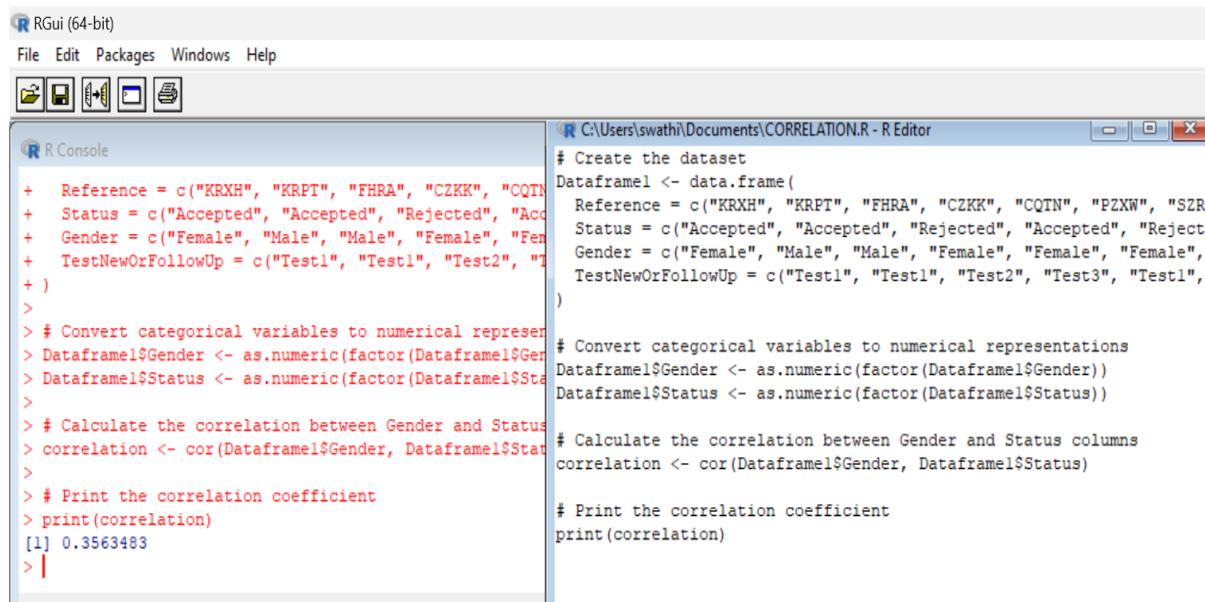
# Calculate the correlation between Gender and Status columns
correlation <- cor(Dataframe1$Gender, Dataframe1>Status)

```

```
# Print the correlation coefficient
```

```
print(correlation)
```

OUTPUT:



The screenshot shows the RGui interface with two windows open. The left window is the R Console, displaying R code and its output. The right window is the R Editor, showing the same R code. Both windows have a blue header bar with the R logo and the text 'RGui (64-bit)'.

R Console:

```
+ Reference = c("KRXH", "KRPT", "FHRA", "CZKK", "CQTN")
+ Status = c("Accepted", "Accepted", "Rejected", "Accepted", "Accepted")
+ Gender = c("Female", "Male", "Male", "Female", "Female")
+ TestNewOrFollowUp = c("Test1", "Test1", "Test2", "Test3", "Test1")
+
>
> # Convert categorical variables to numerical representations
> Dataframe1$Gender <- as.numeric(factor(Dataframe1$Gender))
> Dataframe1$Status <- as.numeric(factor(Dataframe1$Status))
>
> # Calculate the correlation between Gender and Status
> correlation <- cor(Dataframe1$Gender, Dataframe1$Status)
>
> # Print the correlation coefficient
> print(correlation)
[1] 0.3563483
>
```

R Editor:

```
# Create the dataset
Dataframe1 <- data.frame(
  Reference = c("KRXH", "KRPT", "FHRA", "CZKK", "CQTN", "PZXW", "SZRZ"),
  Status = c("Accepted", "Accepted", "Rejected", "Accepted", "Accepted", "Rejected"),
  Gender = c("Female", "Male", "Male", "Female", "Female", "Female"),
  TestNewOrFollowUp = c("Test1", "Test1", "Test2", "Test3", "Test1"),
)

# Convert categorical variables to numerical representations
Dataframe1$Gender <- as.numeric(factor(Dataframe1$Gender))
Dataframe1$Status <- as.numeric(factor(Dataframe1$Status))

# Calculate the correlation between Gender and Status columns
correlation <- cor(Dataframe1$Gender, Dataframe1$Status)

# Print the correlation coefficient
print(correlation)
```

RESULT:

L. VISUALIZATION IN R

1. Write a program for creating a pie-chart in R using the input vector(21,62,10,53). Provide labels for the chart as ‘London’, ‘New York’, ‘Singapore’, ‘Mumbai’. Add a title to the chart as ‘city pie-chart’ and add a legend at the top right corner of the chart.

Aim:

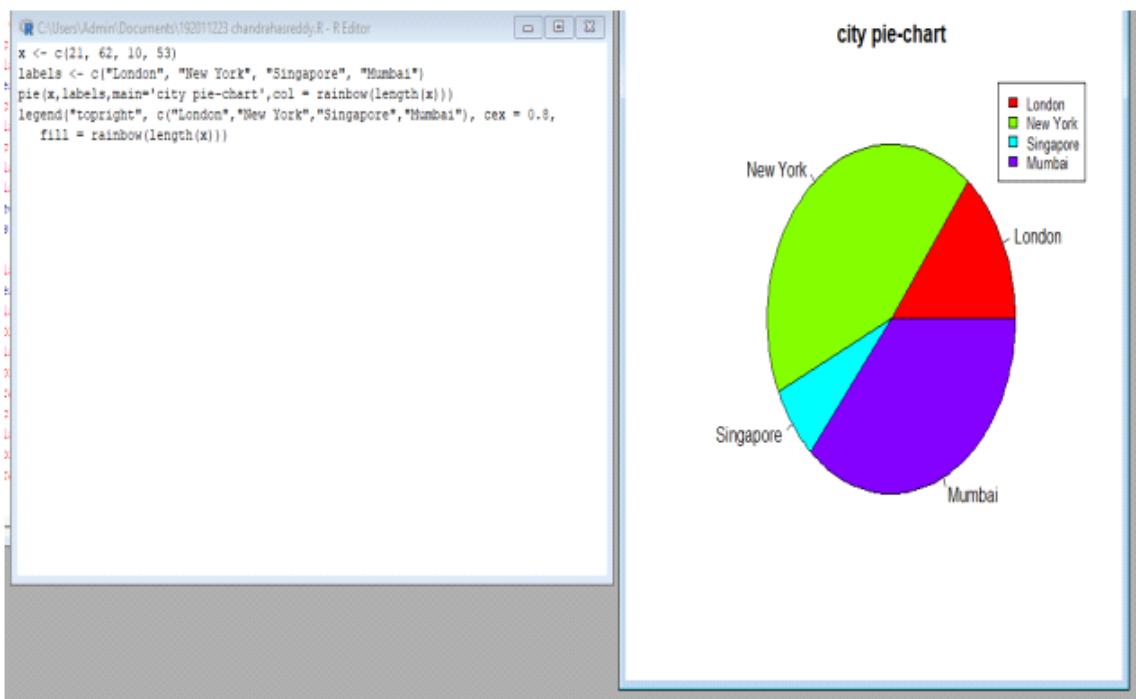
Procedure:

Program:

A)

```
x <- c(21, 62, 10, 53)
labels <- c("London", "New York", "Singapore", "Mumbai")
pie(x, labels, main='city pie-chart', col = rainbow(length(x)))
legend("topright", c("London", "New York", "Singapore", "Mumbai"), cex = 0.8,
      fill = rainbow(length(x)))
```

Output:



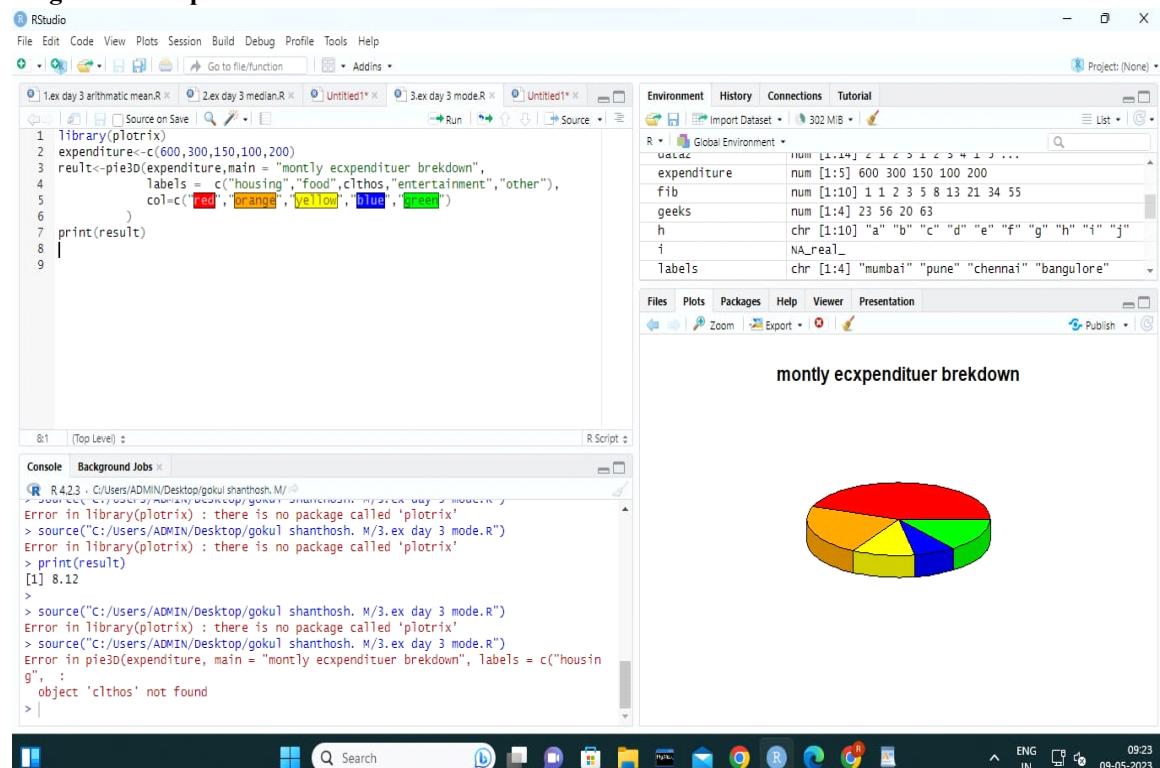
Result:

2. Create a 3D Pie Chart for the dataset “political Knowledge” with suitable labels, colours and a legend at the top right corner of the chart

Aim:

Procedure:

Program & Output:



Result:

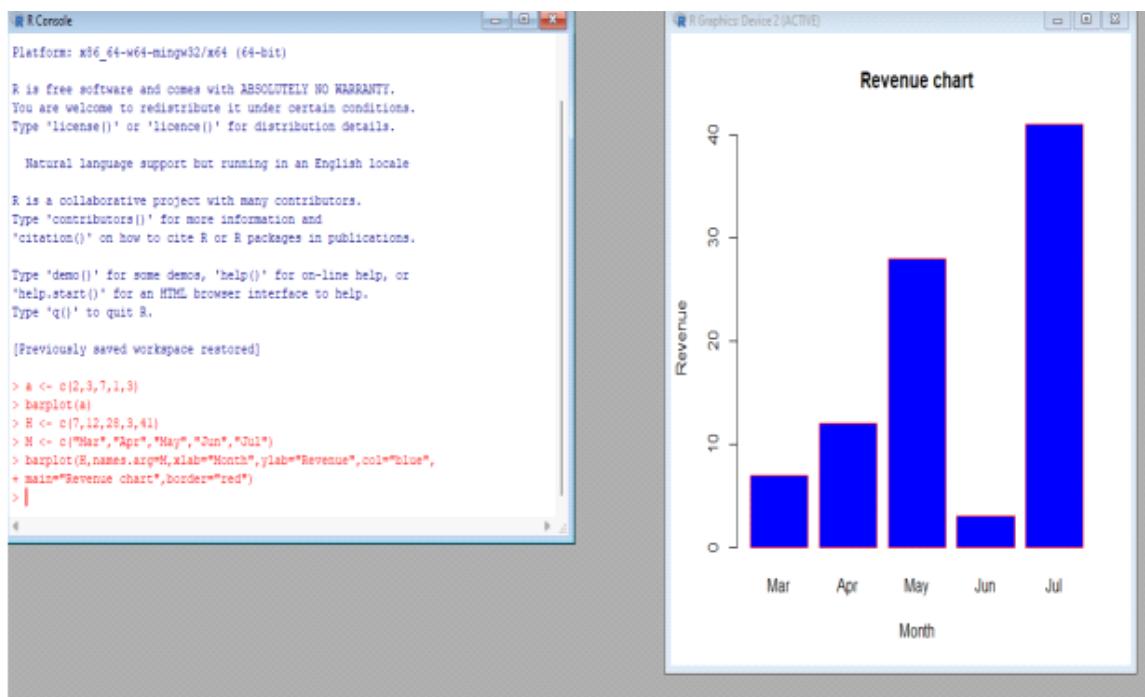
3. Write a program for creating a bar chart using the vectors $H=c(7,12,28,3,41)$ and $M=c("mar", "apr", "may", "jun", "jul")$. Add a title to the chart as “Revenue chart”

Aim:

Procedure:

Program:

```
A ] a <- c(2,3,7,1,3)
> barplot(a)
> H <- c(7,12,28,3,41)
> M <- c("Mar","Apr","May","Jun","Jul")
> barplot(H,names.arg=M,xlab="Month",ylab="Revenue",col="blue",
+ main="Revenue chart",border="red")
```

Output:**Result:**

4. Make a histogram for the “Air Passengers” dataset, start at 100 on the x-axis, and from values 200 to 700, make the bins 200 wide

Aim:**Procedure:****Program:**

```
A)hist(AirPassengers,
main="Histogram for Air Passengers",
```

```

xlab="Passengers",
border="blue",
col="green",
xlim=c(100,700),
las=1,
breaks=5)

```

Output:

```

pe 'contributors()' for more information and
itation() on how to cite R or R packages in publications.

```

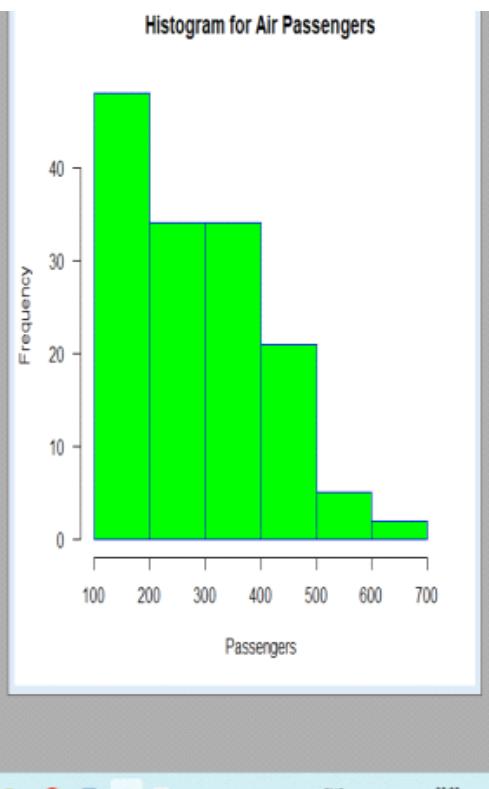
```

pe 'de
elp.sta
pe 'q()
hist(AirPassengers,
main="Histogram for Air Passengers",
xlab="Passengers",
border="blue",
col="green",
xlim=c(100,700),
las=1,
breaks=5)

```

```

|
```



Result:

5. Create a Boxplot graph for the relation between "mpg"(miles per gallon) and "cyl"(number of Cylinders) for the dataset "mtcars" available in R Environment.

Aim:

Procedure:

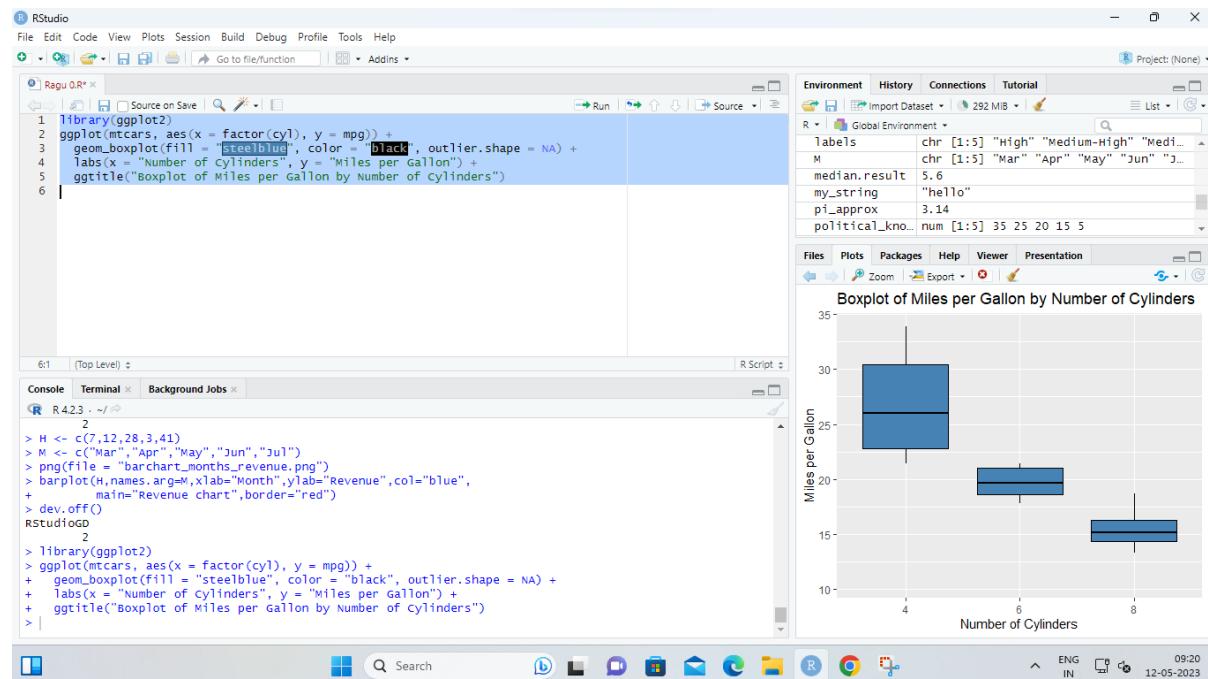
Program:

```

library(ggplot2)
ggplot(mtcars, aes(x = factor(cyl), y = mpg)) +
  geom_boxplot(fill = "steelblue", color = "black", outlier.shape = NA) +
  labs(x = "Number of Cylinders", y = "Miles per Gallon") +
  ggtitle("Boxplot of Miles per Gallon by Number of Cylinders")

```

Output:



Result:

M. LINEAR REGRESSION ANALYSIS IN R

- Using linear regression analysis establish a relationship between height and weight of a person using the input vector given below.

```
# Values of height  
151, 174, 138, 186, 128, 136, 179, 163, 152, 131  
# Values of weight.  
63, 81, 56, 91, 47, 57, 76, 72, 62, 48
```

Predict the weight of a person with height 170. Visualize the regression graphically.

Aim:

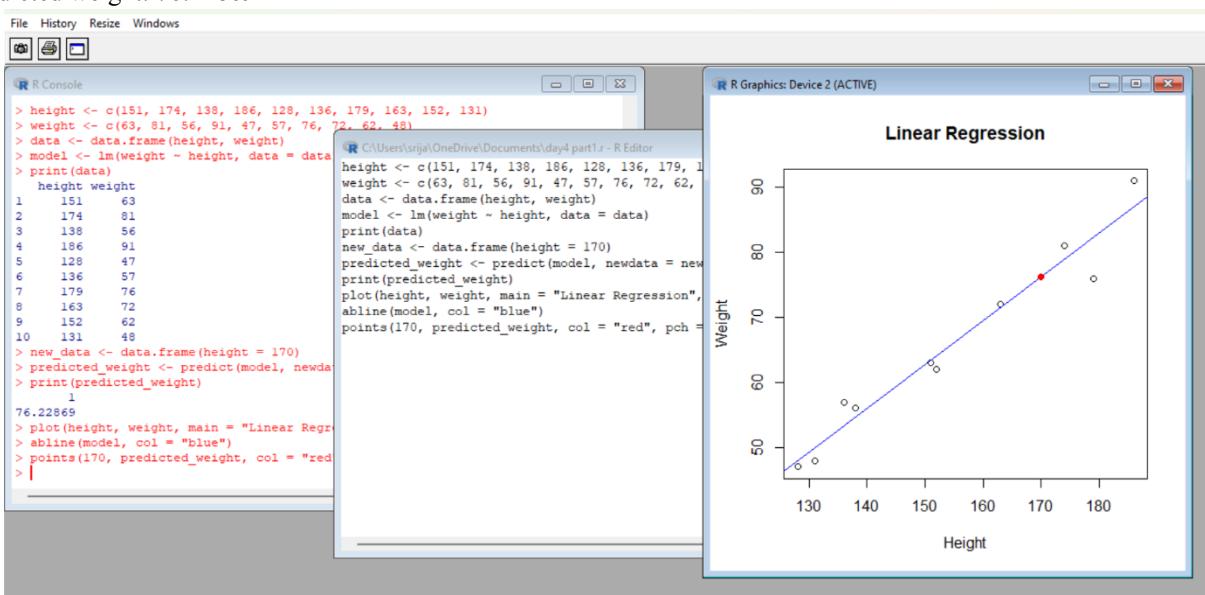
Procedure:

Program:

```
height <- c(151, 174, 138, 186, 128, 136, 179, 163, 152, 131)  
weight <- c(63, 81, 56, 91, 47, 57, 76, 72, 62, 48)  
data <- data.frame(height, weight)  
model <- lm(weight ~ height, data = data)  
print(data)  
new_data <- data.frame(height = 170)  
predicted_weight <- predict(model, newdata = new_data)  
print(predicted_weight)  
plot(height, weight, main = "Linear Regression", xlab = "Height", ylab = "Weight")  
abline(model, col = "blue")  
points(170, predicted_weight, col = "red", pch = 16)
```

Output:

Predicted weight: 76.22869



Result:

- Download the Dataset "water". Find out whether there is a linear relation between attributes "mortality" and "hardness" by plot function. Fit the Data into the Linear Regression model. Predict the mortality for the hardness=88

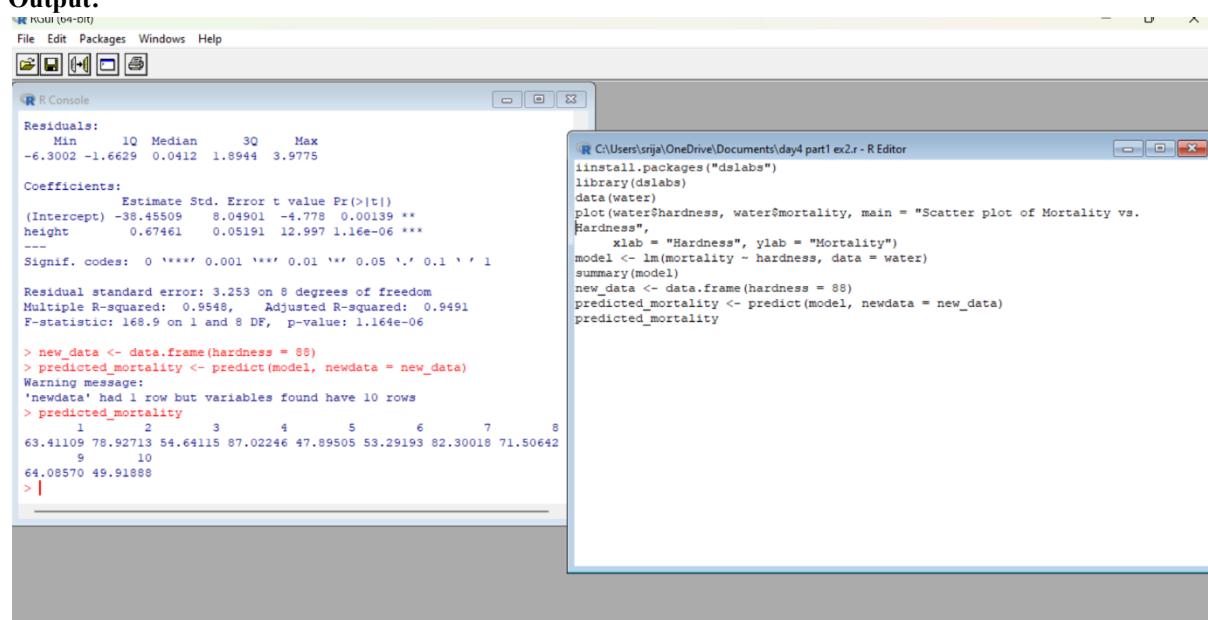
AIM:

PROCEDURE:

PROGRAM:

```
iinstall.packages("dslabs")
library(dslabs)
data(water)
plot(water$hardness, water$mortality, main = "Scatter plot of Mortality vs.Hardness",
xlab = "Hardness", ylab = "Mortality")
model <- lm(mortality ~ hardness, data = water)
summary(model)
new_data <- data.frame(hardness = 88)
predicted_mortality <- predict(model, newdata = new_data)
predicted_mortality
```

Output:



The screenshot shows two windows side-by-side. The left window is the R Console, displaying the output of the R code. The right window is the R Editor, showing the source code.

R Console Output:

```
Residuals:
    Min      1Q  Median      3Q     Max 
-6.3002 -1.6629  0.0412  1.8944  3.9775 

Coefficients:
            Estimate Std. Error t value Pr(>|t|)    
(Intercept) -38.45509   8.04901  -4.778  0.00139 **  
height       0.67461   0.05191  12.997 1.16e-06 ***  
---
Signif. codes:  0 '****' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1 

Residual standard error: 3.253 on 8 degrees of freedom
Multiple R-squared:  0.9548, Adjusted R-squared:  0.9491 
F-statistic: 168.9 on 1 and 8 DF,  p-value: 1.164e-06

> new_data <- data.frame(hardness = 88)
> predicted_mortality <- predict(model, newdata = new_data)
Warning message:
'newdata' had 1 row but variables found have 10 rows
> predicted_mortality
   1     2     3     4     5     6     7     8 
63.41109 78.82713 54.64115 87.02246 47.89505 53.29193 82.30018 71.50642 
  9     10    
64.08570 49.91888 
> |
```

R Editor Content:

```
iinstall.packages("dslabs")
library(dslabs)
data(water)
plot(water$hardness, water$mortality, main = "Scatter plot of Mortality vs.
Hardness",
xlab = "Hardness", ylab = "Mortality")
model <- lm(mortality ~ hardness, data = water)
summary(model)
new_data <- data.frame(hardness = 88)
predicted_mortality <- predict(model, newdata = new_data)
predicted_mortality
```

Result:

N. MULTIPLE REGRESSION ANALYSIS IN R

1. Generate a multiple regression model using the built in dataset mtcars. It gives a comparison between different car models in terms of mileage per gallon (mpg), cylinder displacement("disp"), horse power("hp"), weight of the car("wt") and some more parameters.

Establish the relationship between "mpg" as a response variable with "disp", "hp" and "wt" as predictor variables. Predict the mileage of the car with disp=221, hp=102 and wt=2.91.

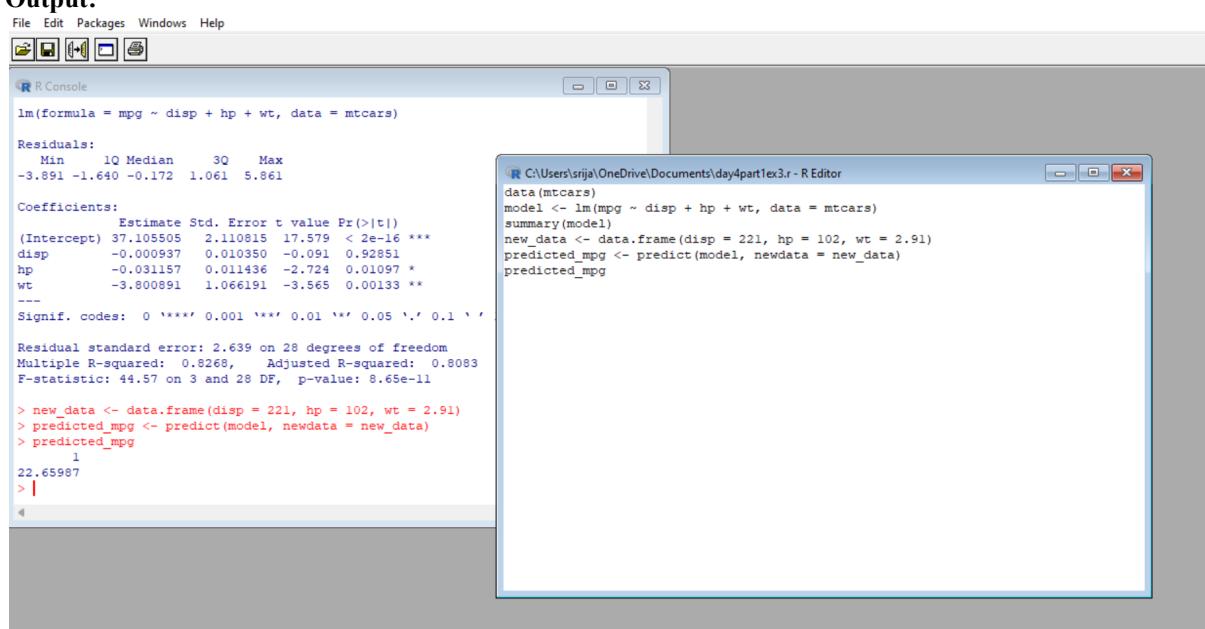
AIM:

PROCEDURE:

PROGRAM:

```
data(mtcars)
model <- lm(mpg ~ disp + hp + wt, data = mtcars)
summary(model)
new_data <- data.frame(disp = 221, hp = 102, wt = 2.91)
predicted_mpg <- predict(model, newdata = new_data)
predicted_mpg
```

Output:



The screenshot shows two windows. The left window is the R Console with the following text:

```
File Edit Packages Windows Help
R Console
lm(formula = mpg ~ disp + hp + wt, data = mtcars)

Residuals:
    Min      1Q Median      3Q     Max 
-3.891 -1.640 -0.172  1.061  5.861 

Coefficients:
            Estimate Std. Error t value Pr(>|t|)    
(Intercept) 37.10505   2.110815 17.579 < 2e-16 ***
disp        -0.000937   0.010350 -0.091  0.92851    
hp         -0.031157   0.011436 -2.724  0.01097 *  
wt         -3.800891   1.066191 -3.565  0.00133 ** 
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' '
Residual standard error: 2.639 on 28 degrees of freedom
Multiple R-squared:  0.8268, Adjusted R-squared:  0.8083 
F-statistic: 44.57 on 3 and 28 DF, p-value: 8.65e-11

> new_data <- data.frame(disp = 221, hp = 102, wt = 2.91)
> predicted_mpg <- predict(model, newdata = new_data)
> predicted_mpg
1
22.65987
> |
```

The right window is titled "C:\Users\srija\OneDrive\Documents\day4part1ex3.r - R Editor" and contains the R code used to generate the results.

Result:

2. Consider the data set "delivery" available in the R environment. It gives a deliverytime ("delTime") of production materials (number of productions "n.prod") with the given distance ("distance") to reach the destination place.

a) Create the model to establish the relationship between "delTime" as a response variable with "n.prod" and "distance" as predictor variables.

b) Predict the delTime for the given number of production("n.prod")=9 and distance("distance")=450

Aim:

Procedure:

Program & Output:

```
data(delivery)
names(delivery)
model <- lm(delTime ~ n.prod + distance, data = delivery)
summary(model)
new_data <- data.frame(n.prod = 9, distance = 450)
predicted_delTime <- predict(model, newdata = new_data)
predicted_delTime
```

The screenshot shows two windows from an R environment. The left window is the R Console, displaying the output of the R code provided above. The right window is the R Editor, showing the same R code. The R Console output includes statistical results like residuals, coefficients, and a summary of the linear model fit.

```
R Console
Residuals:
    Min      1Q  Median      3Q     Max 
-6.3002 -1.6629  0.0412  1.8944  3.9775 

Coefficients:
            Estimate Std. Error t value Pr(>|t|)    
(Intercept) -38.45509   8.04901  -4.778 0.00139 **  
height       0.67461   0.05191  12.997 1.16e-06 *** 
---
Signif. codes:  0 '****' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1 

Residual standard error: 3.253 on 8 degrees of freedom
Multiple R-squared:  0.9548, Adjusted R-squared:  0.9491 
F-statistic: 168.9 on 1 and 8 DF, p-value: 1.164e-06

> new_data <- data.frame(n.prod = 9, distance = 450)
> predicted_delTime <- predict(model, newdata = new_data)
Warning message:
'newdata' had 1 row but variables found have 10 rows
> predicted_delTime
  1     2     3     4     5     6     7     8 
63.41109 78.92713 54.64115 87.02246 47.89505 53.29193 82.30018 71.50642 
  9     10    
64.08570 49.91888 
> |
```

```
R C:\Users\srija\OneDrive\Documents\day4.1ex4.r - R Editor
data(delivery)
names(delivery)
model <- lm(delTime ~ n.prod + distance, data = delivery)
summary(model)
new_data <- data.frame(n.prod = 9, distance = 450)
predicted_delTime <- predict(model, newdata = new_data)
predicted_delTime
```

Result:

O. LOGISTIC REGRESSION ANALYSIS IN R

1. Create a logistic regression model using the "mtcars" data set with the information given below. The in-built data set "mtcars" describes different models of a car with their various engine specifications. In "mtcars" data set, the transmission mode (automatic or manual) is described by the column am which is a binary value (0 or 1). Create a logistic regression model between the columns "am" and 3 other columns - hp, wt and cyl.

AIM:

PROCEDURE:

CODE:

```
# Load the mtcars dataset
data(mtcars)
# Create a logistic regression model
model <- glm(am ~ hp + wt + cyl, data = mtcars, family = binomial)
# Print the summary of the model
summary(model)
```

OUTPUT:

```
R Console
Call:
glm(formula = am ~ hp + wt + cyl, family = binomial, data = mtcars)

Deviance Residuals:
    Min      1Q  Median      3Q     Max 
-2.17272 -0.14907 -0.01464  0.14116  1.27641 

Coefficients:
            Estimate Std. Error z value Pr(>|z|)    
(Intercept) 19.70288   8.11637   2.428   0.0152 *  
hp          0.03259   0.01886   1.728   0.0840 .  
wt         -9.14947   4.15332  -2.203   0.0276 *  
cyl         0.48760   1.07162   0.455   0.6491    
---
Signif. codes:  0 '****' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for binomial family taken to be 1)

Null deviance: 43.2297  on 31  degrees of freedom
Residual deviance: 9.8415  on 28  degrees of freedom
AIC: 17.841

Number of Fisher Scoring iterations: 8
> |
```

RESULT:

P. POISSON REGRESSION ANALYSIS IN R

1. Create a Poisson regression model using the in-built data set “warpbreaks” with information given below.

Aim :

Procedure:

Program & Output:

```
> input <- warpbreaks  
> print(head(input))
```

OUTPUT:

breaks wool tension

1	26	A	L
2	30	A	L
3	54	A	L
4	25	A	L
5	70	A	L
6	52	A	L

```
> # Load the "warpbreaks" dataset  
> data(warpbreaks)  
>  
> # Check the structure of the dataset  
> str(warpbreaks)
```

OUTPUT:

```
'data.frame': 54 obs. of 3 variables:
 $ breaks : num 26 30 54 25 70 52 51 26 67 18 ...
 $ wool   : Factor w/ 2 levels "A","B": 1 1 1 1 1 1 1 1 1 1 ...
 $ tension: Factor w/ 3 levels "L","M","H": 1 1 1 1 1 1 1 1 1 2 ...
```

```
>
> # Fit a Poisson regression model
> model <- glm(breaks ~ wool + tension, data = warpbreaks, family = poisson)
>
> # Display the summary of the model
> summary(model)
```

OUTPUT:

Call:

```
glm(formula = breaks ~ wool + tension, family = poisson, data = warpbreaks)
```

Deviance Residuals:

Min	1Q	Median	3Q	Max
-3.6871	-1.6503	-0.4269	1.1902	4.2616

Coefficients:

	Estimate	Std. Error	z value	Pr(> z)
(Intercept)	3.69196	0.04541	81.302	< 2e-16 ***
woolB	-0.20599	0.05157	-3.994	6.49e-05 ***
tensionM	-0.32132	0.06027	-5.332	9.73e-08 ***
tensionH	-0.51849	0.06396	-8.107	5.21e-16 ***

Signif. codes: 0 ‘***’ 0.001 ‘**’ 0.01 ‘*’ 0.05 ‘.’ 0.1 ‘ ’ 1

(Dispersion parameter for poisson family taken to be 1)

Null deviance: 297.37 on 53 degrees of freedom
 Residual deviance: 210.39 on 50 degrees of freedom
 AIC: 493.06

Number of Fisher Scoring iterations: 4

```
>
> # Make predictions on new data
> new_data <- data.frame(wool = "A", tension = "L")
> predict(model, newdata = new_data, type = "response")
```

OUTPUT:

```
1
40.12354
```

Result: