

**Batch: A1      Roll No.: 16010123012**

**Experiment No. 1**

**Title: Exploring R for Data Science**

**Course Outcome:**

CO1, CO3

**Books/ Journals/ Websites referred:**

1. [The Comprehensive R Archive Network](#)
2. [Posit](#)

**Resources used:**

<https://www.rdocumentation.org/>

<https://www.w3schools.com/r/>

<https://www.geeksforgeeks.org/r-programming-language-introduction/>

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In this lab, we will introduce some simple R commands. R is a programming language for statistical computing and data visualization. It has been adopted in the fields of data mining, bioinformatics and data analysis

The best way to learn a new language is to try out the commands. R can be downloaded from <http://cran.r-project.org/> . We recommend that you run R within an integrated development environment (IDE) such as RStudio, which can be freely downloaded from <http://rstudio.com>

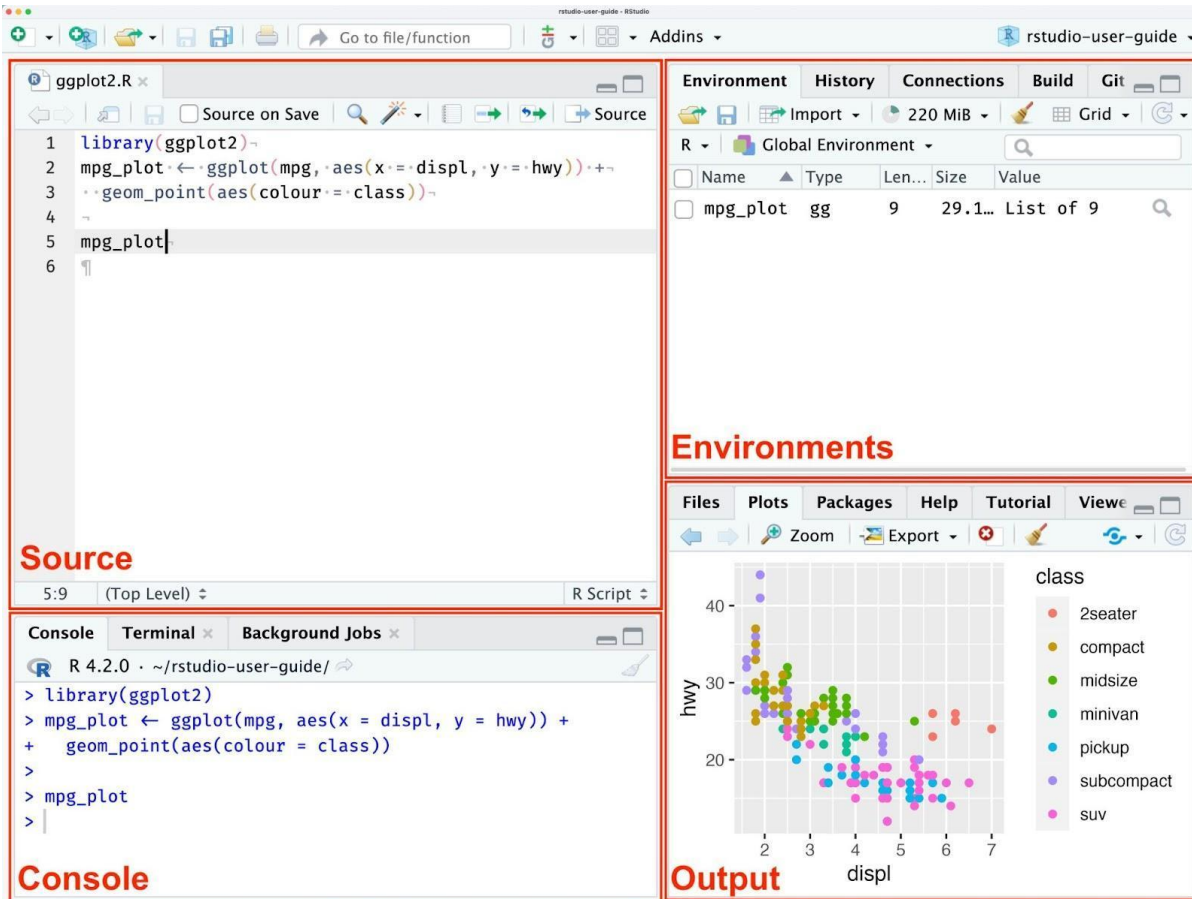
Pane Layout

**The RStudio user interface has 4 primary panes:**

- Source pane
- Console pane
- Environment pane, containing the Environment, History, Connections, Build, VCS , and Tutorial tabs

- Output pane, containing the Files, Plots, Packages, Help, Viewer, and Presentation tabs

Each pane can be minimized or maximized within the column by clicking the minimize/maximize buttons.



The screenshot displays the RStudio IDE interface. The **Source** pane on the left shows an R script with the following code:

```
1 library(ggplot2)
2 mpg_plot <- ggplot(mpg, aes(x = displ, y = hwy)) +
3   geom_point(aes(colour = class))
4
5 mpg_plot
6
```

The **Console** pane at the bottom left shows the execution of the code:

```
> library(ggplot2)
> mpg_plot <- ggplot(mpg, aes(x = displ, y = hwy)) +
+   geom_point(aes(colour = class))
>
> mpg_plot
>
```

The **Output** pane on the right contains the **Environments** and **Plots** tabs. The **Environments** tab shows the Global Environment with a table listing the objects:

Name	Type	Len...	Size	Value
mpg_plot	gg	9	29.1...	List of 9

The **Plots** tab shows a scatter plot of highway mileage (hwy) versus engine displacement (displ), colored by vehicle class. The legend indicates the following classes: 2seater, compact, midsize, minivan, pickup, subcompact, and suv.

```
> print("Hello, World!")
```

```
[1] "Hello, World!"
```

### Basic Arithmetic Operators

- Arithmetic operations: +, -, \*, /, ^, %%, %/%
- Assigning values: <- and =

```
> x <- 10
> y <- 5
> quotient <- x %/% y
> remainder <- x %% y
> quotient
[1] 2
> remainder
[1] 0
```

### Basic Data Types

```
> # numeric
> x <- 10.5
> class(x)
[1] "numeric"
>
> # integer
> x <- 1000L
> class(x)
[1] "integer"
>
> # complex
> x <- 9i + 3
> class(x)
[1] "complex"
>
> # character/string
> x <- "R is exciting"
> class(x)
[1] "character"
>
> # logical/boolean
> x <- TRUE
> class(x)
[1] "logical"

> num <- as.numeric("123")
> char <- as.character(123)
> num
[1] 123
> typeof(num)
[1] "double"
> char
[1] "123"
> typeof(char)
[1] "character"
```

### Vectors and Basic Vector Operations

**R uses functions to perform operations.**

**To run a function called *funcname*, we type *funcname(input1, input2)*, where the inputs (or arguments) *input1* and *input2* tell R how to run the function.**

**A function can have any number of inputs. For example, to create a vector of numbers, we use the function *c()* (for concatenate).**

**Any numbers inside the parentheses are joined together.**

**The following command instructs R to join together the numbers 1, 3, 2, and 5, and to save them as a vector named *x*.**

```
> x <- c(1, 3, 2, 5)
> x
[1] 1 3 2 5
```

**When we type *x*, it gives us back the vector.**

**Other ways to create vectors are using *seq()* and *rep()***

```
> seq_vec <- seq(1, 10, by = 2)
> seq_vec
[1] 1 3 5 7 9
> rep_vec <- rep(5, times = 3)
> rep_vec
[1] 5 5 5
```

**We can tell R to add two sets of numbers together. It will then add the first number from *x* to the first number from *y*, and so on. However, *x* and *y* should be the same length. We can check their length using the *length()* function.**

```
> y = c(1, 4, 3)
```

```
> length(x)
[1] 3
> length(y)
[1] 3
> x + y
[1] 2 10 5
```

The `ls()` function allows us to look at a list of all of the objects, such as data and functions, that we have saved so far. The `rm()` function can be used to delete any that we don't want.

```
> ls()
[1] "x" "y"
> rm(x, y)
```

```
> ls()
character(0)
```

It's also possible to remove all objects at once:

```
> rm(list = ls())
```

The `matrix()` function can be used to create a matrix of numbers. Before we use the `matrix()` function, we can learn more about it:

```
> ?matrix
```

The help file reveals that the `matrix()` function takes a number of inputs, but for now we focus on the first three: the data (the entries in the matrix), the number of rows, and the number of columns. First, we create a simple matrix.

```
> x <- matrix(data = c(1, 2, 3, 4), nrow = 2, ncol = 2)
> x
      [,1] [,2]
[1,]    1    3
[2,]    2    4
```

Note that we could just as well omit typing `data=`, `nrow=`, and `ncol=` in the `matrix()` command above: that is, we could just type

```
> x <- matrix(c(1, 2, 3, 4), 2, 2)
```

and this would have the same effect. However, it can sometimes be useful to specify the names of the arguments passed in, since otherwise R will assume that the function arguments are passed into the function in the same order that is given in the function's help file. As this example illustrates, by default R creates matrices by successively filling in columns. Alternatively, the `byrow = TRUE` option can be used to populate the matrix in order of the rows.

```
> matrix(c(1, 2, 3, 4), 2, 2, byrow = TRUE)
      [,1] [,2]
[1,]    1    2
[2,]    3    4
```

Notice that in the above command we did not assign the matrix to a value such as `x`. In this case the matrix is printed to the screen but is not saved for future calculations.

```
> mat <- matrix(1:9, nrow = 3, byrow = TRUE)
>
> mat
      [,1] [,2] [,3]
[1,]    1    2    3
[2,]    4    5    6
[3,]    7    8    9
> mat[1,2]
[1] 2
```

The `sqrt()` function returns the square root of each element of a vector or matrix. The command `x^2` raises each element of `x` to the power 2; any powers are possible, including fractional or negative powers.

```
> sqrt(x)
      [,1] [,2]
[1,]  1.00  1.73
[2,]  1.41  2.00
> x^2
      [,1] [,2]
[1,]     1     9
[2,]     4    16
```

### Creating a list

```
> my_list <- list(name = "John", age = 25)
> my_list
$name
[1] "John"

$age
[1] 25

> my_list$name
[1] "John"
```

## Creating dataframes

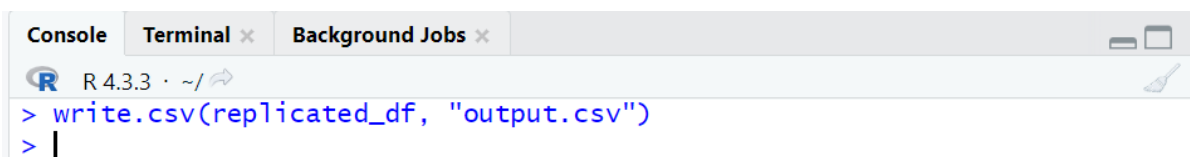
```
> # Create a sample data frame
> df <- data.frame(
+   Name = c("Alice", "Bob", "Charlie", "David"),
+   Age = c(25, 30, 35, 40),
+   City = c("New York", "London", "Paris", "Tokyo")
+ )
> df
```

	Name	Age	City
1	Alice	25	New York
2	Bob	30	London
3	Charlie	35	Paris
4	David	40	Tokyo

```
> # Replicate each row twice
> replicated_df <- cbind(df, rep(row.names(df), each = 2))
> replicated_df
```

	Name	Age	City	rep(row.names(df), each = 2)
1	Alice	25	New York	1
2	Bob	30	London	1
3	Charlie	35	Paris	2
4	David	40	Tokyo	2
5	Alice	25	New York	3
6	Bob	30	London	3
7	Charlie	35	Paris	4
8	David	40	Tokyo	4

## Writing dataframe to csv file



The screenshot shows the R Studio interface with the Console tab selected. The R version is 4.3.3. The command `write.csv(replicated_df, "output.csv")` has been entered, and the cursor is at the next line.

```
R 4.3.3 · ~/
> write.csv(replicated_df, "output.csv")
> |
```



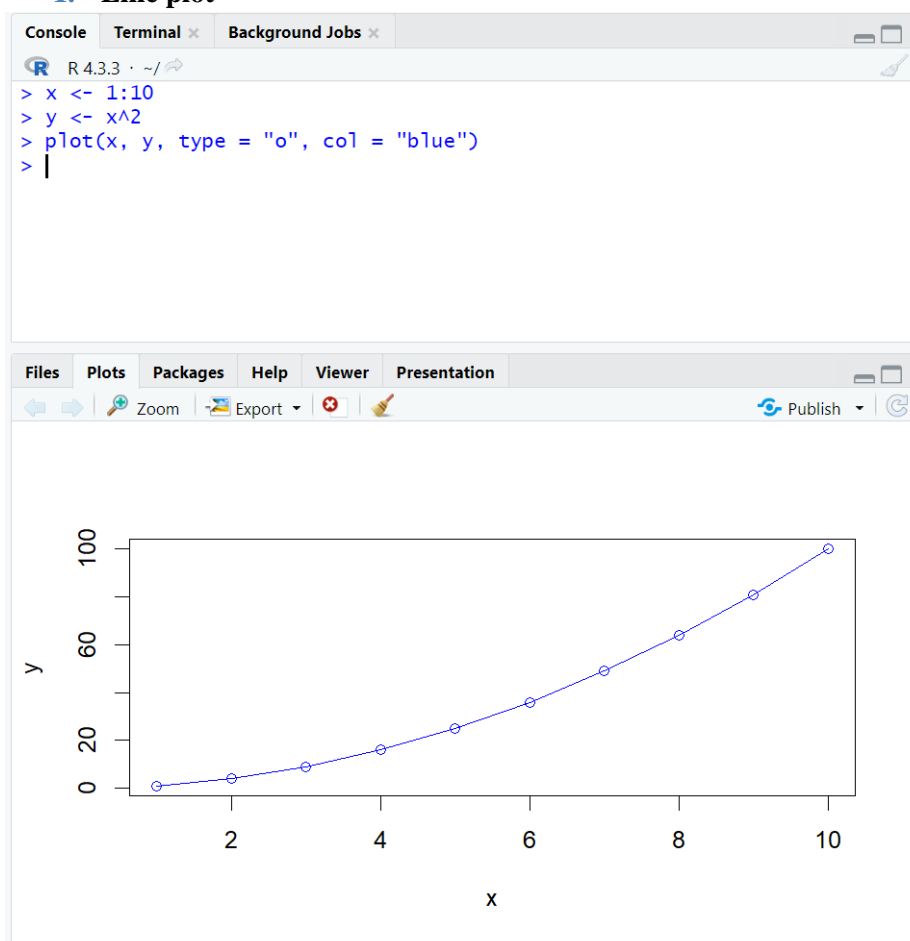
## Reading dataframe from a csv file

```

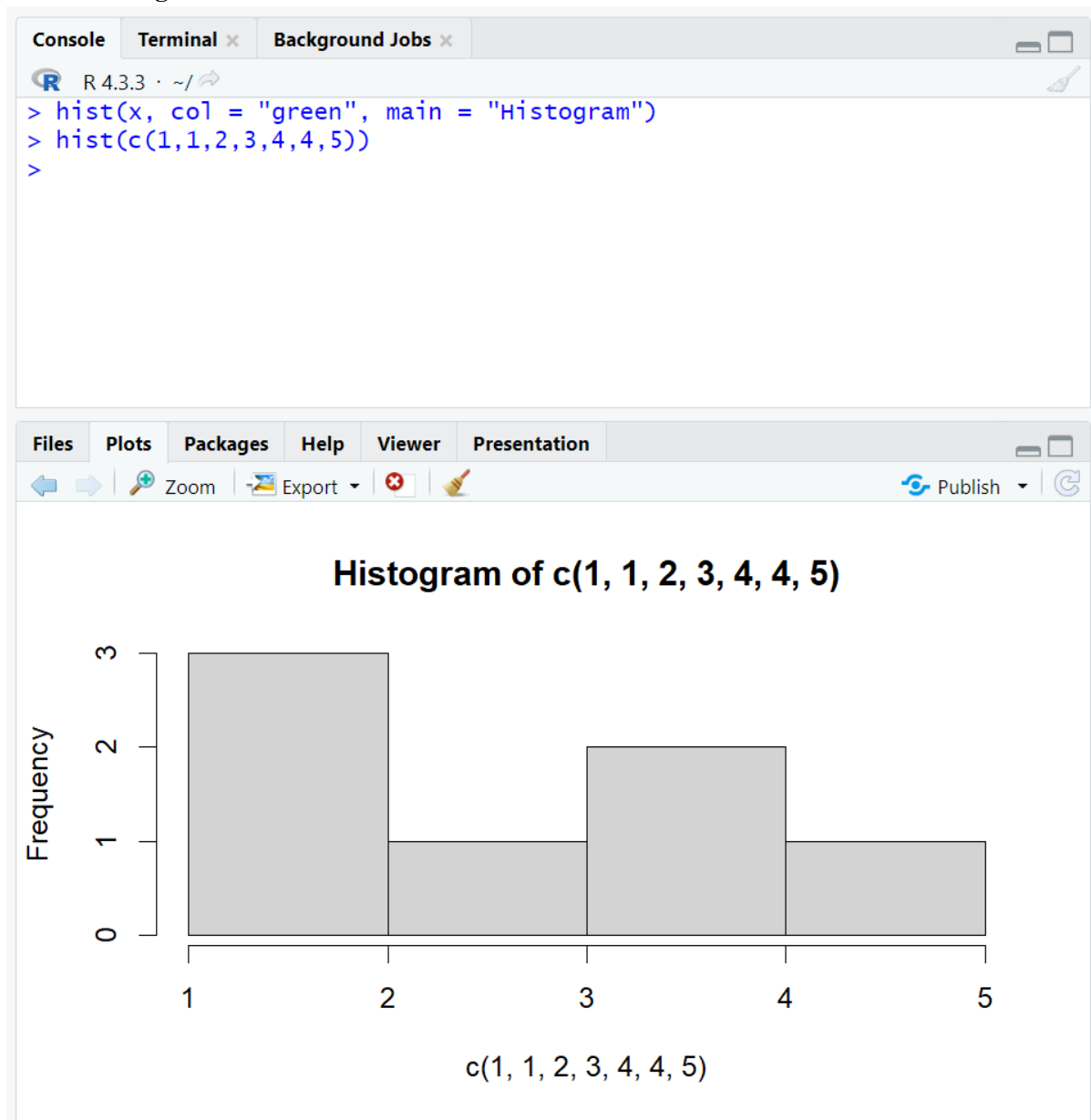
Console Terminal x Background Jobs x
R 4.3.3 · ~/
> write.csv(replicated_df, "output.csv")
> new_df <- read.csv("output.csv")
> new_df
  X   Name Age   City rep.row.names.df...each...2.
1 1  Alice  25 New York                        1
2 2   Bob  30  London                        1
3 3 Charlie 35   Paris                        2
4 4  David 40   Tokyo                        2
5 5  Alice 25 New York                        3
6 6   Bob 30  London                        3
7 7 Charlie 35   Paris                        4
8 8  David 40   Tokyo                        4
  
```

## Basic Visualization

### 1. Line plot



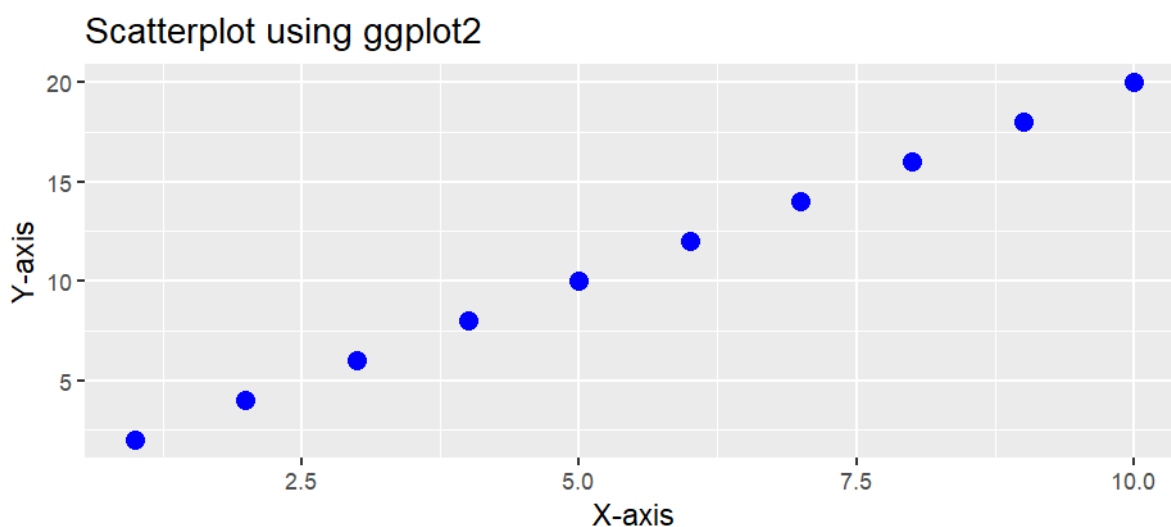
## 2. Histogram



### 3. Scatterplot, using ggplot2

```

Console Terminal x Background Jobs x
R 4.3.3 ~ /
> library(ggplot2)
Learn more about the underlying theory at
https://ggplot2-book.org/
> df <- data.frame(
+   x = 1:10,
+   y = c(2, 4, 6, 8, 10, 12, 14, 16, 18, 20)
+ )
> ggplot(df, aes(x = x, y = y)) +
+   geom_point(color = "blue", size = 3) +
+   ggtitle("Scatterplot using ggplot2") +
+   xlab("X-axis") +
+   ylab("Y-axis")
  
```



**Students have to perform the above tasks and add their code and screenshots of the output here :**

```
> print("Hello, world!")
[1] "Hello, world!"
> x = 10
> y = 5
> class(x)
[1] "numeric"
> typeof(y)
[1] "double"
> z = 'KJSSE'
Error: unexpected input in "z = '"
> z = 'KJSSE'
> print(z)
[1] "KJSSE"
> class(z)
[1] "character"
> is.complex(z)
[1] FALSE
> print(ls())
[1] "x" "y" "z"
> a = readline()
12
> a = as.integer(a)
> 24
[1] 24
> l = as.integer(readline())
78
> var = readline(prompt = "Enter random number: ")
Enter random number: 546
> g = 0
> g = scan()
1: 13
2: 56
3: 345
4: 789
5: 23
6:
Read 5 items
>
```

Environment	History	Connections	Tutorial
<div> <div>Import Dataset</div> <div>149 MiB</div> <div>List</div> </div>			
<div> <div>R</div> <div>Global Environment</div> </div>			
values			
a	12L		
g	num [1:5] 13 56 345 789 23		
l	78L		
var	"546"		
x	10		
y	5		
z	"KJSSE"		

```
> #numeric data type
> x<-10.5
> class(x)
[1] "numeric"
> x<-1000L
> class(x)
[1] "integer"
> x<-5i+3
> class(x)
[1] "complex"
> x<-"R is exciting"
> class(x)
[1] "character"
> x<-FALSE
> class(x)
[1] "logical"

> num<-as.numeric("124")
> char<-as.character(123)
> class(num)
[1] "numeric"
> typeof(num)
[1] "double"
> class(char)
[1] "character"
> typeof(char)
[1] "character"
>

> vec<-c(1,3,2,4,5)
> vec
[1] 1 3 2 4 5
```

```
> seq_vec=seq(1,10,by=1)
> seq_vec
[1] 1 2 3 4 5 6 7 8 9 10
> rep_vec=rep(1,times=4)
> rep_vec
[1] 1 1 1 1
```

```
> x<-c(1,2,3)
> y<-c(4,5,6)
> length(x)
[1] 3
> length(y)
[1] 3
> x+y
[1] 5 7 9
```

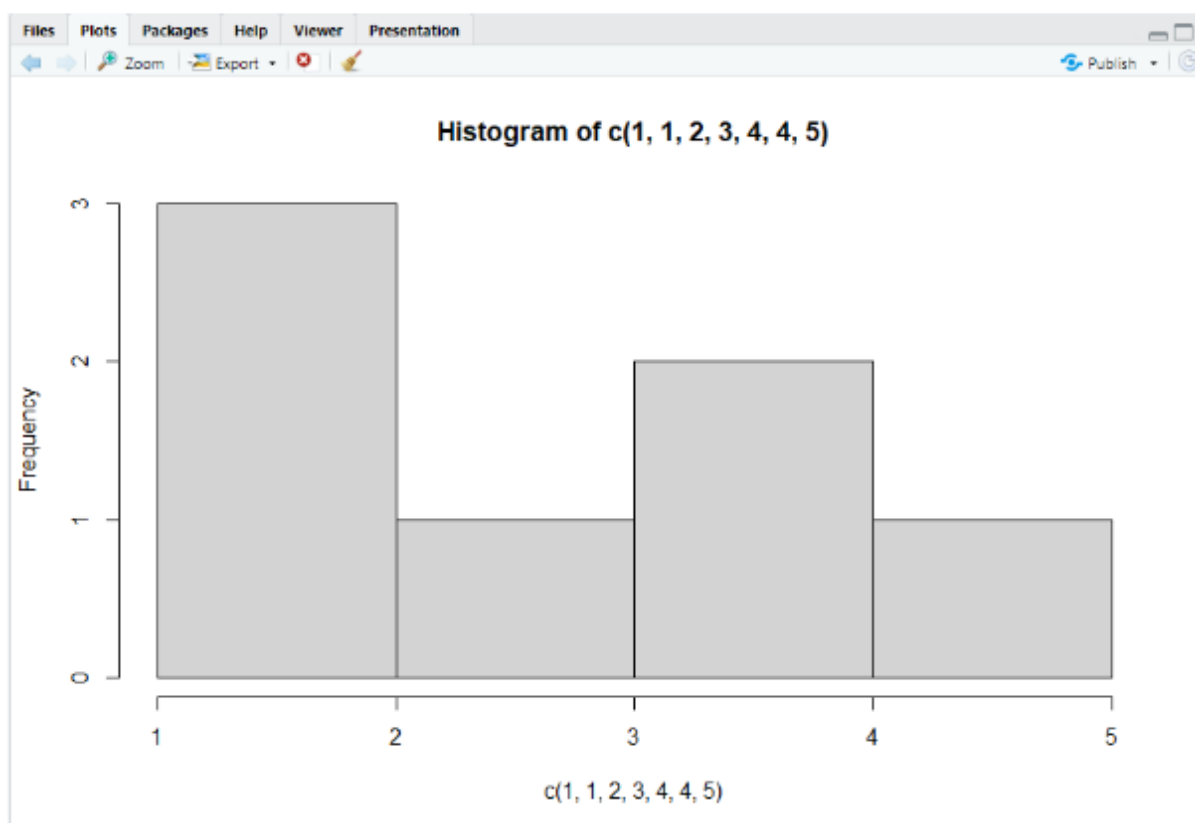
```
> x<-c(1,3,2,5)
> y<-c(1,4,3)
> ls()
[1] "a" "x" "y"
> rm(a)
```

```
> x<-matrix(data=c(1,2,3,4),nrow=2,ncol=2)
> x
      [,1] [,2]
[1,]    1    3
[2,]    2    4
> mat<-matrix(1:9,nrow=3,byrow=TRUE)
> mat
      [,1] [,2] [,3]
[1,]    1    2    3
[2,]    4    5    6
[3,]    7    8    9
> sqrt(x)
      [,1] [,2]
[1,] 1.000000 1.732051
[2,] 1.414214 2.000000
> x^2
      [,1] [,2]
[1,]    1    9
[2,]    4   16
```

```

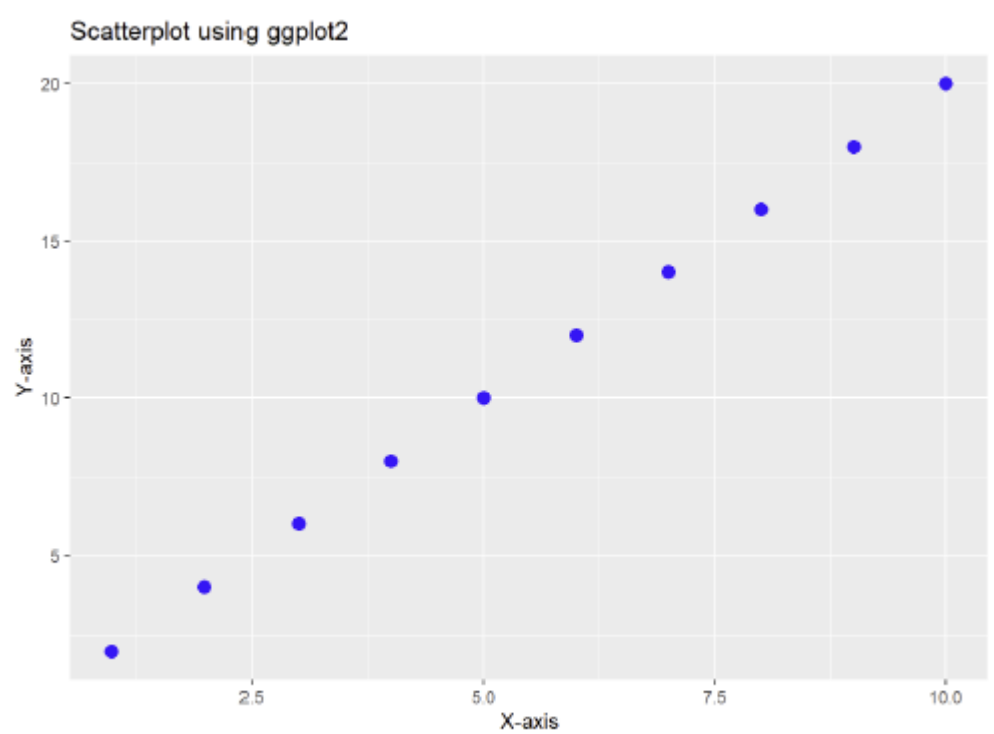
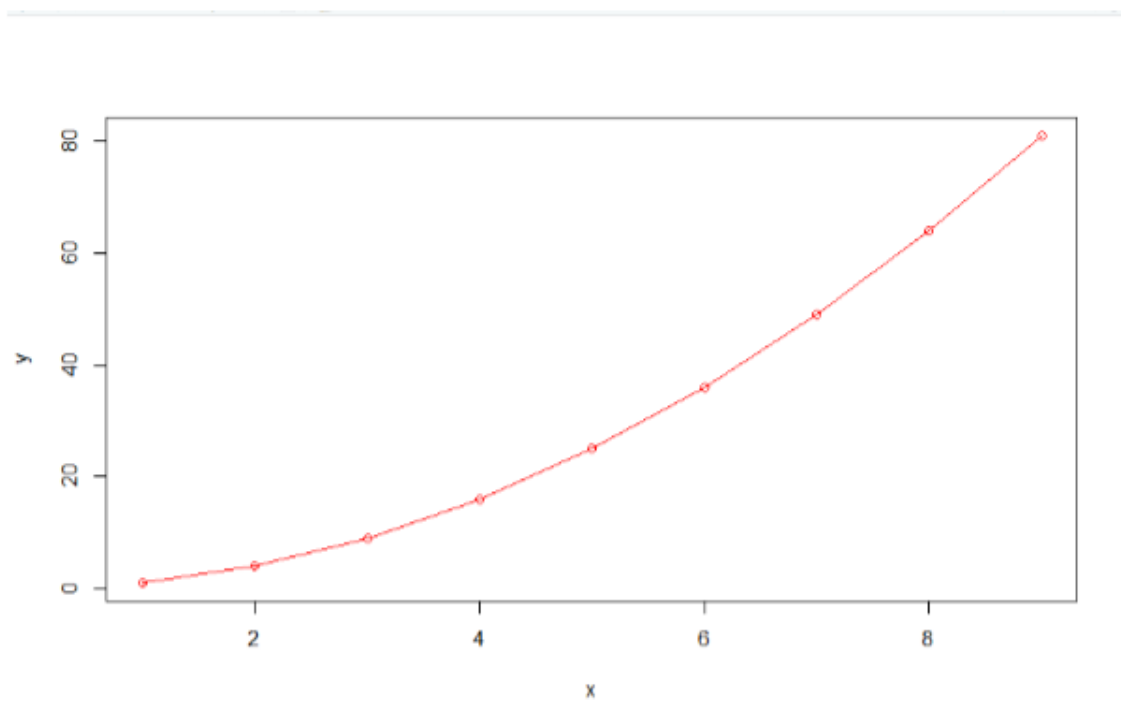
> df=data.frame(
+ Name=c("Alan","Bob","Cam","Dan"),
+ Age=c(20,23,25,27),
+ City=c("Mumbai","Delhi","Tokyo","Berlin")
+ )
> df
  Name Age  City
1 Alan  20 Mumbai
2 Bob   23  Delhi
3 Cam   25  Tokyo
4 Dan   27  Berlin

> hist(x, col="green", main="Histogram")
> hist(c(1,1,2,3,4,4,5))
  
```



```

> x<-1:9
> y<-x^2
> plot(x,y, type="o",col="red")
  
```





**Conclusion:**

I have successfully completed this experiment and learned about the basics of R programming.

**Post Lab Questions:**

1. Explain the difference between vectors, lists, and data frames in R. Provide an example of when each would be used.
2. What are the differences between the plot() and ggplot() functions in R? When would you use each?
3. What parameters can you customize in the hist() function to change the appearance of the histogram?
4. Write an R command to create a vector of numbers from 1 to 100, but only include multiples of 5.

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16010123012

## IDS Exp 1

- 1) Vector: Homogeneous. 1-D object that contain elements of same <sup>type</sup> ~~time~~. eg: `ages = c(18, 13, 20)`. # Numeric Vector  
Used when same type of elements are needed to be stored.  
List: Heterogeneous objects that can contain elements of diff types. Used when related but diverse data types are stored together.  
eg. `list <- c("a", "b", 1, 2)` # List  
DataFrame: Two D. tabular structure, where each column in a vector & columns can have diff types. Used when we need to store & analyze structured data.  
ex. `df <- data.frame(name = c("A", "B"), no. = c(1, 2))`
- 2) plot():  
  - (i) Base R graphics
  - (ii) Offers quick, simple plotting with limited customisation.
  - (iii) Less flexible for complex visualisations
  - (iv) Used for simple exploratory plotting.ggplot():  
  - (i) Part of ggplot2 package
  - (ii) Highly customisable & supports layered, complex visualisations.
  - (iii) Allows for modular addition of components.
  - (iv) Used for creating publication-quality or complex visualisations
- 3) breaks: Number or method to determine number of bins.  
  - (i) `col`: Color of the bars
  - (ii) `main`: Title of the histogram
  - (iii) `xlab/ylab`: Labels for x & y axis
  - (iv) `xlim/ylim`: Limits for x & y axis
  - (v) `border`: Color of the bar borders
- 4) `multi <- seq(5, 100, by = 5)`