

R Programming Lab Assignment List

BE Sem VIII Jan-May 2020

1. Installation of R and R Studio, Basics of R [LO1].
2. Datatypes and Data structures in R [LO1, LO2,].
3. Loops and Conditions in R [LO1].
4. EDA on inbuilt R dataset [LO3, LO4].
5. Graphs Plotting using ggplot2 library [LO2, LO5].
6. Regression and Correlation [LO4].
7. Mini-Project Session 1 [LO6].:
Choose Dataset from Kaggle, extracting data from large dataset
8. Mini-Project Session 2:
Cleaning of the Dataset
9. Mini-Project Session 3:
EDA on the dataset
10. Mini-Project Session 4:
Regression analysis
11. Mini-Project Session 5:
Data Visualization using ggplot2

Subject In charges: Dr. Shachi Natu

Dr. Darshan Ingle.

Lab Assignment: 01

Aim:

Installation of R and RStudio

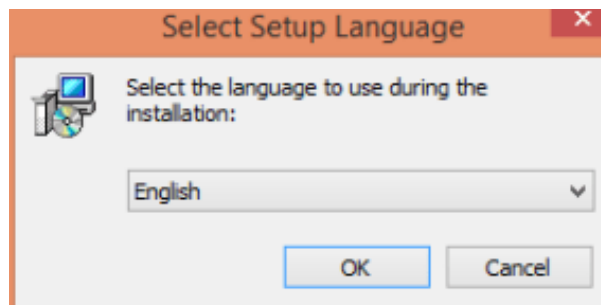
Theory:

R is an open source programming language and software environment for statistical computing. The R language is widely used among statisticians and data miners for developing statistical software and data analysis.

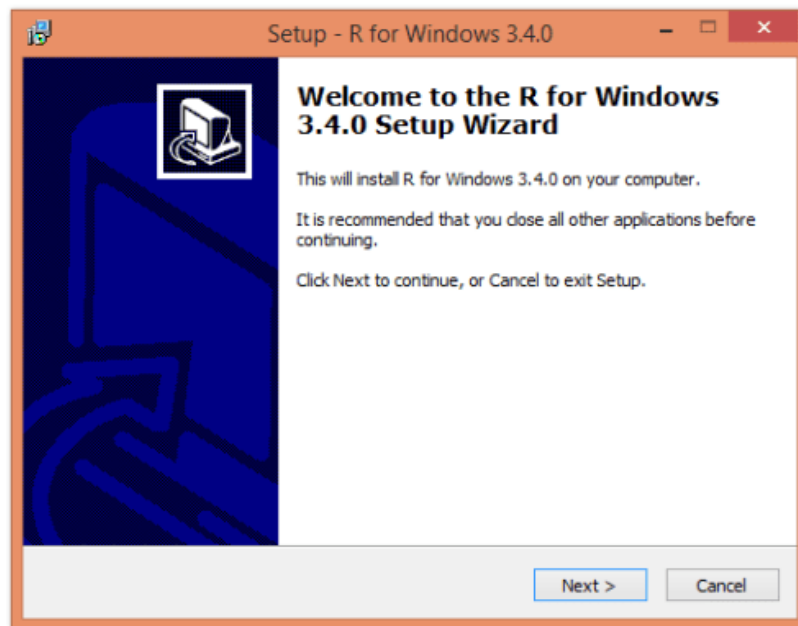
A major advantage that R has over many other statistical packages and is that it's free in the sense of free software. R is both flexible and powerful. It has an amazing ecosystem for developers and It has wide range of packages for data access, data cleaning or munging, performing Analysis, creating Reports etc.

Steps to install R on Windows:

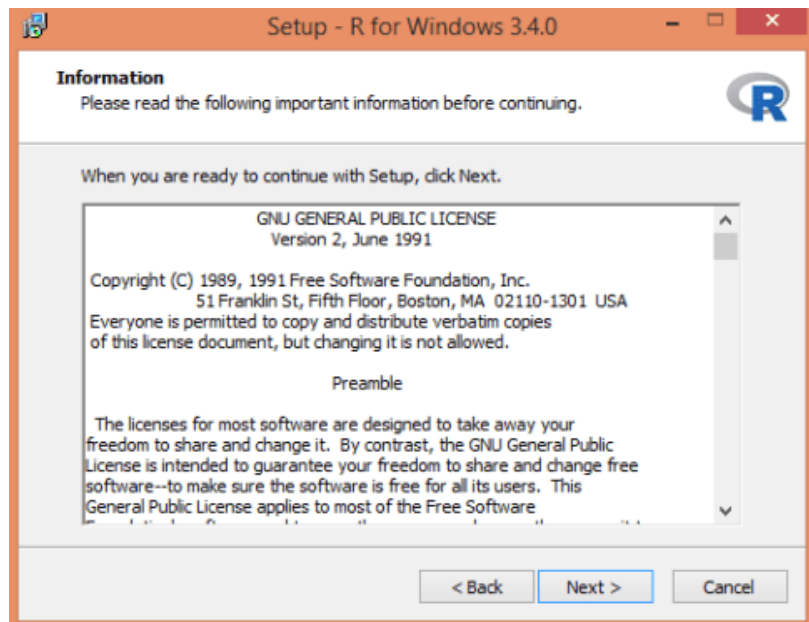
1. Go to <http://ftp.heanet.ie/mirrors/cran.r-project.org/>.
or <https://cran.rstudio.com/bin/windows/base/>
2. Under “Download and Install R”, click on the “Windows” link.
3. Download the required the .exe file. You should see a link saying something like “Download R 3.4.0 for Windows” (or R X.X.X, where X.X.X gives the version of R, eg. R 3.4.0). Click on that link.
4. After downloading double-click on the R-3.4.0-win.exe to run it.
5. You will be asked what language to install it in – choose English.



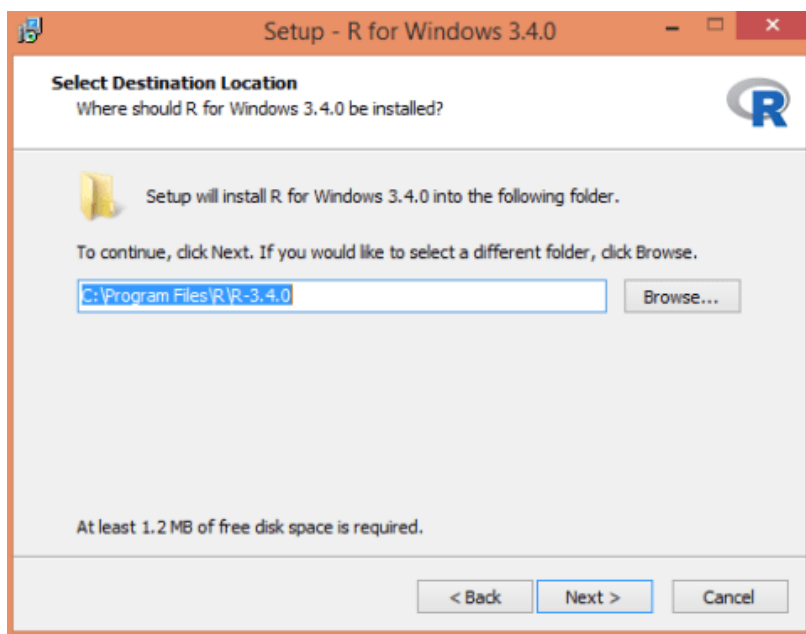
6. The R Setup Wizard will appear in a window. Click “Next” at the bottom of the R setup wizard window.



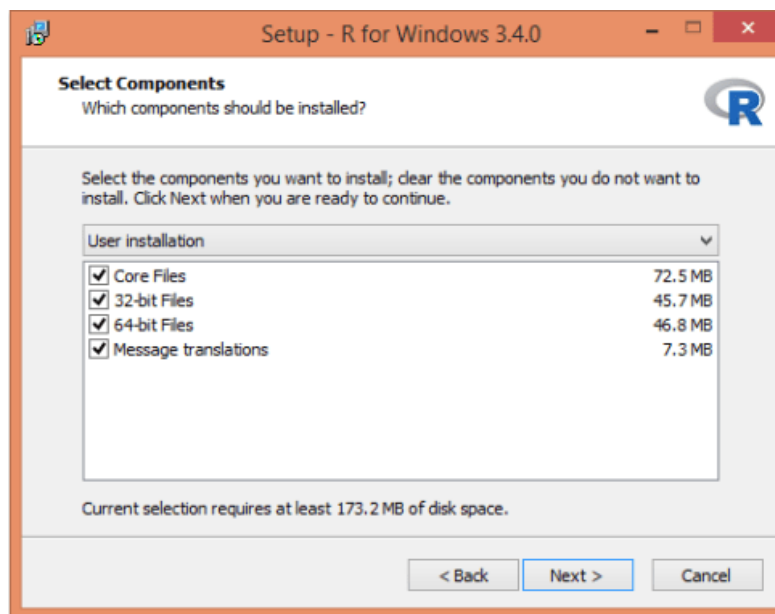
7. Click “Next” at the bottom of the R Setup wizard window.



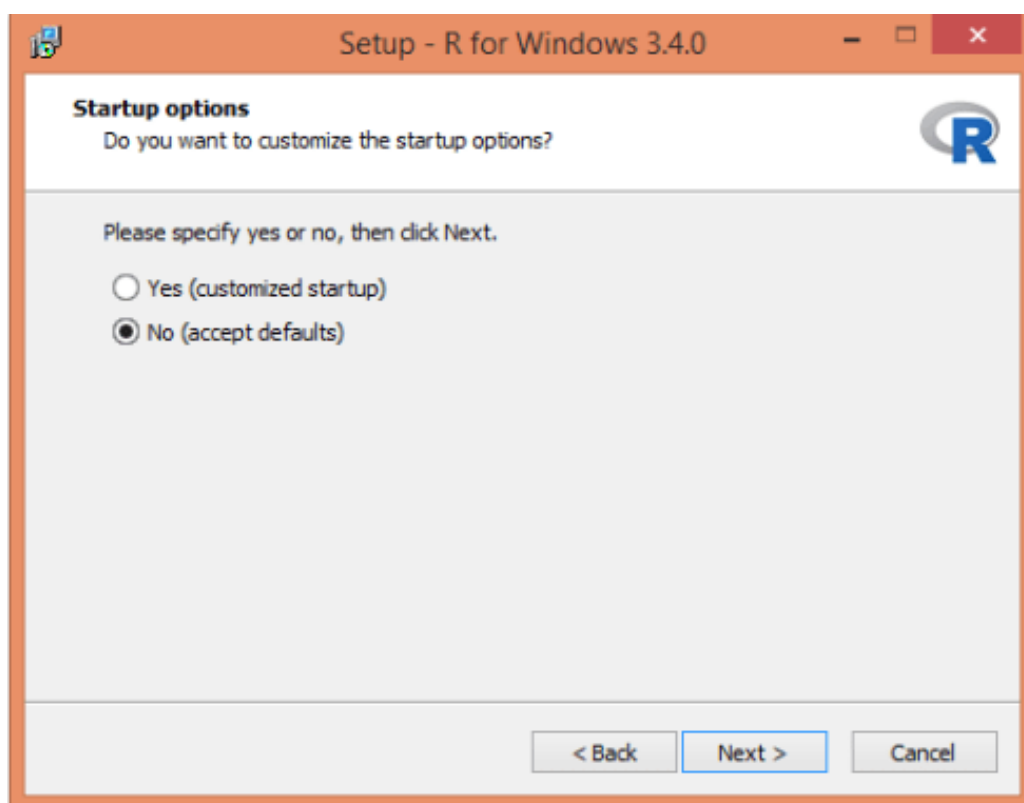
8. The next page says “Select Destination Location” at the top. By default, it will suggest to install R in “C:\Program Files” on your computer.



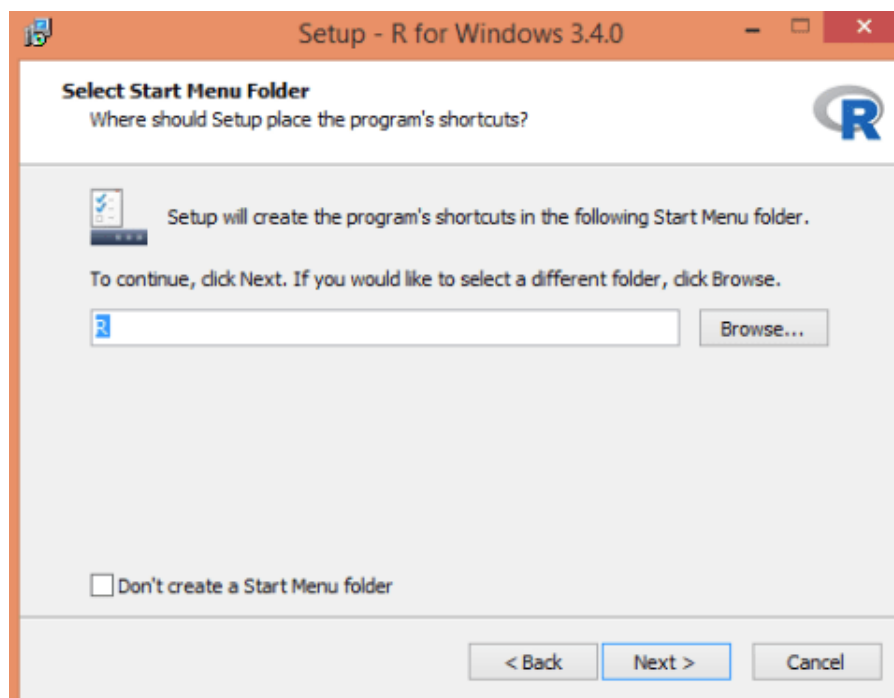
9. The next page says “Select components” at the top. Click “Next” again.



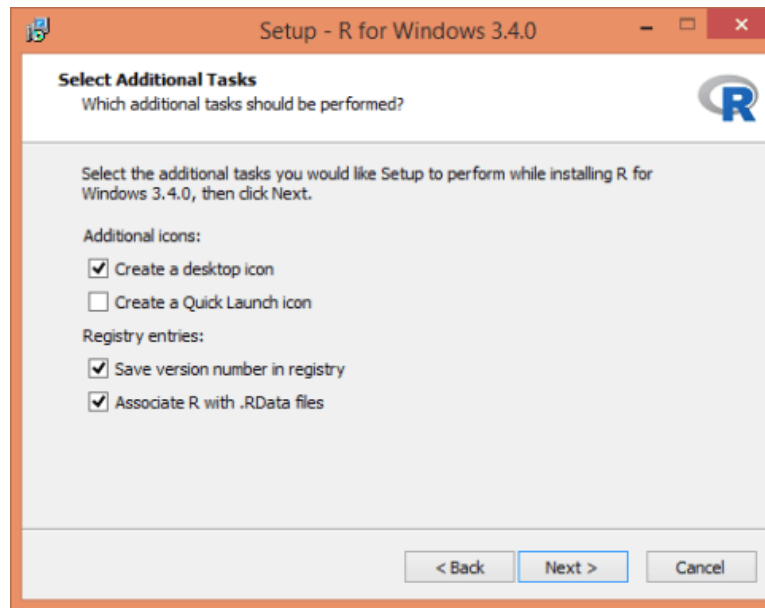
10. The next page says “Startup options” at the top. Click “Next” again.



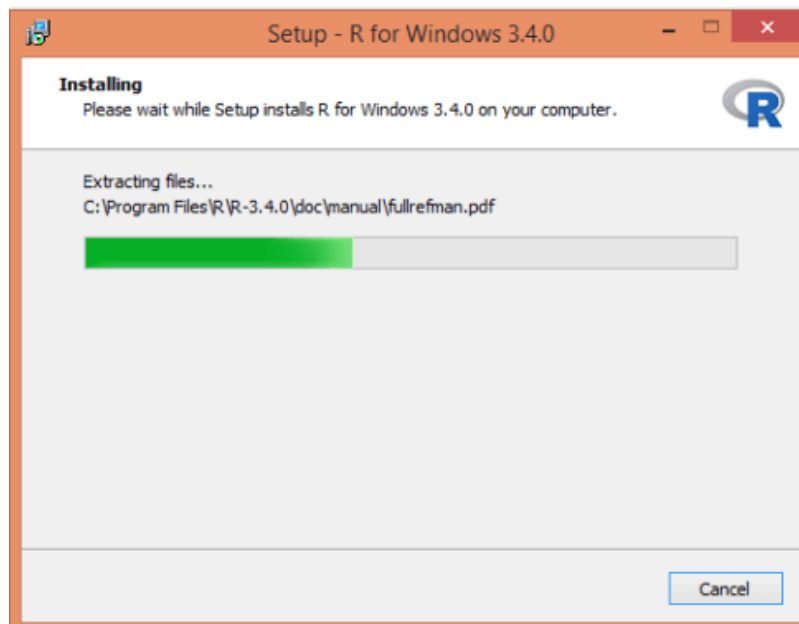
11. The next page says “Select start menu folder” at the top. Click “Next” again.



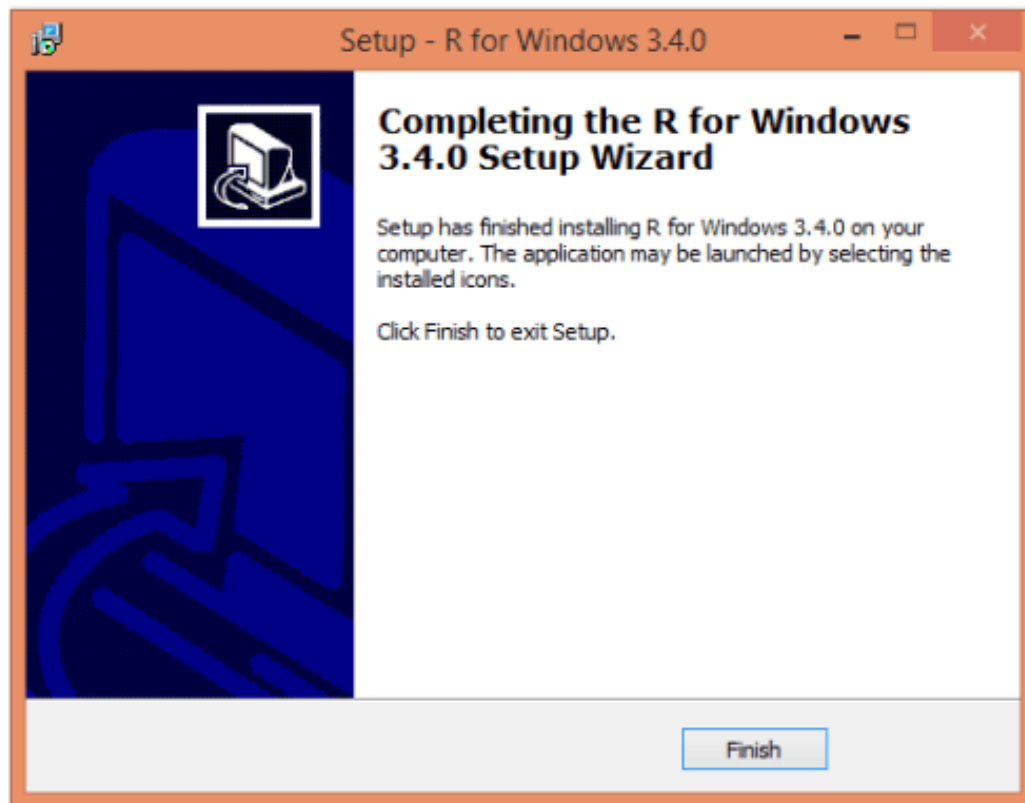
12. The next page says “Select additional tasks” at the top. Click “Next” again.



13. Now it will be installing in your machine. You will see something like this.



14. Now you will see “Completing the R for Windows Setup Wizard” appear. Click “Finish”.



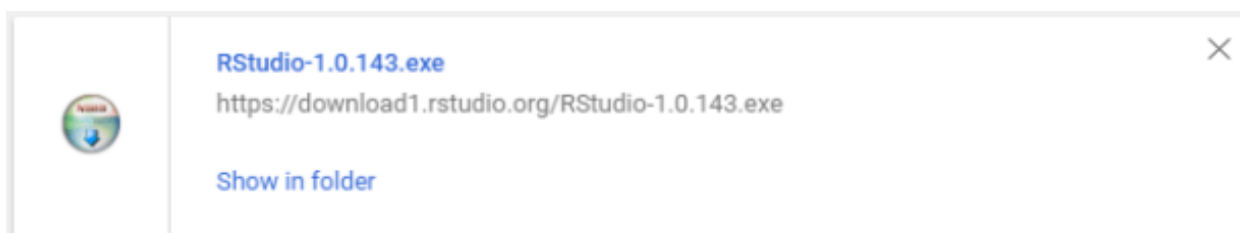
Steps to install R Studio:

1. First, go to <https://www.rstudio.com/products/rstudio/> and click on **DOWNLOAD RSTUDIO DESKTOP**.
2. Then find out the installers for supported for Windows platforms. You will be getting something like this RStudio 1.0.143 – Windows Vista/7/8/10 . Click on that to download RStudio.

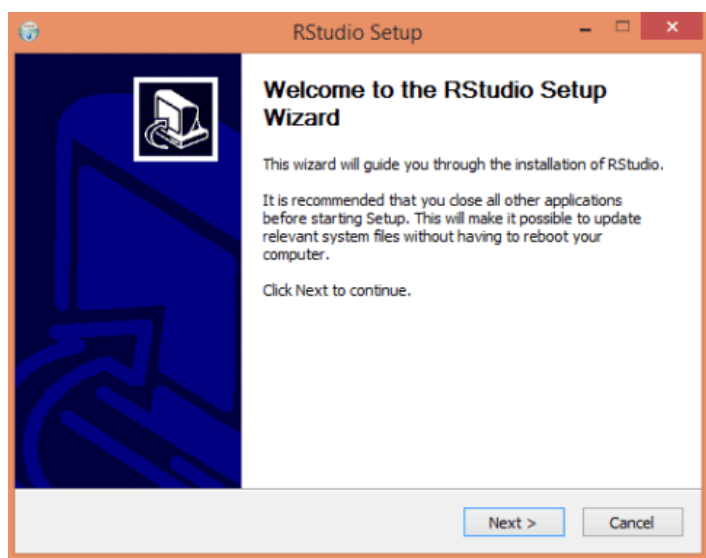
Installers for Supported Platforms

Installers	Size	Date	MD5
RStudio 1.0.143 - Windows Vista/7/8/10	81.9 MB	2017-04-19	76bb84296b9202759b3eb1de555a2231
RStudio 1.0.143 - Mac OS X 10.6+ (64-bit)	71.2 MB	2017-04-19	c7f1ed865428b225b202fd1b431954b4
RStudio 1.0.143 - Ubuntu 12.04+/Debian 8+ (32-bit)	85.5 MB	2017-04-19	21ca14bffcdc1a2361ead2d763d0313d
RStudio 1.0.143 - Ubuntu 12.04+/Debian 8+ (64-bit)	92.1 MB	2017-04-19	75761eae209158d8415d562b3771fbec
RStudio 1.0.143 - Fedora 19+/RedHat 7+/openSUSE 13.1+ (32-bit)	84.7 MB	2017-04-19	2c356d4ee50667ad4042ee196afb3c53
RStudio 1.0.143 - Fedora 19+/RedHat 7+/openSUSE 13.1+ (64-bit)	85.7 MB	2017-04-19	7ab5fc240351debe491c6c5a7acb6068

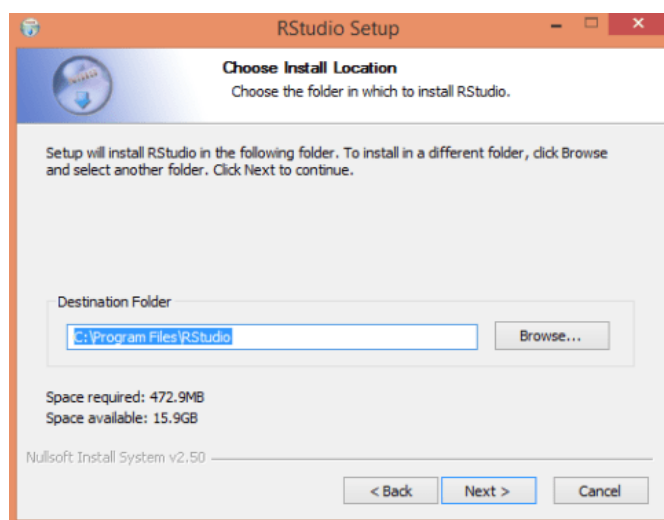
3. Next, find the file that was downloaded in your system and double click it. It will be named something like RStudio-1.0.143.exe. This will start the install process.



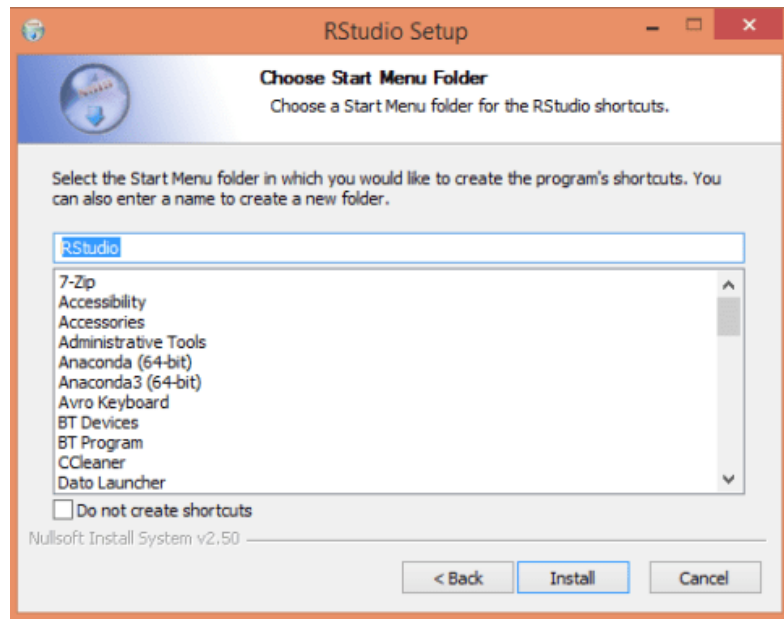
4. Click next to continue when the install wizard opens.



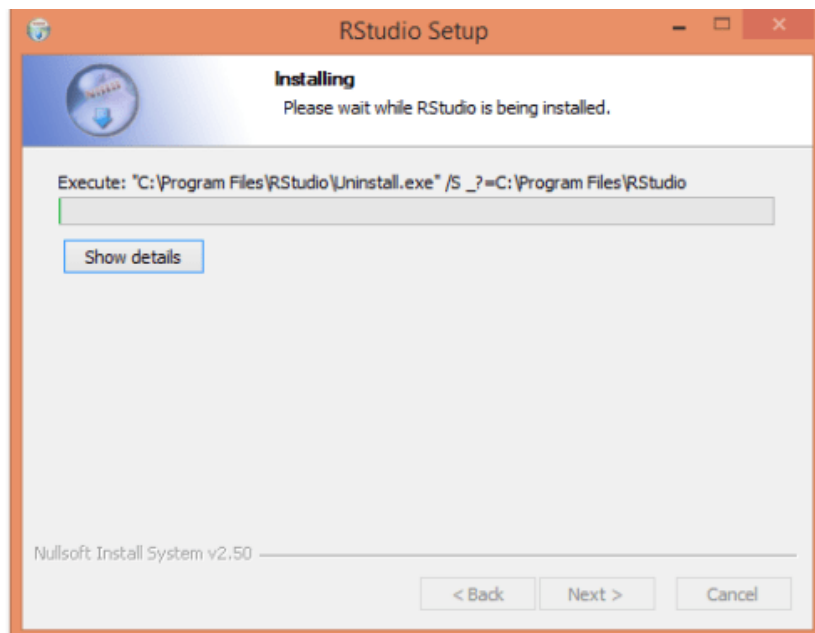
5. Click next to accept the default install location.



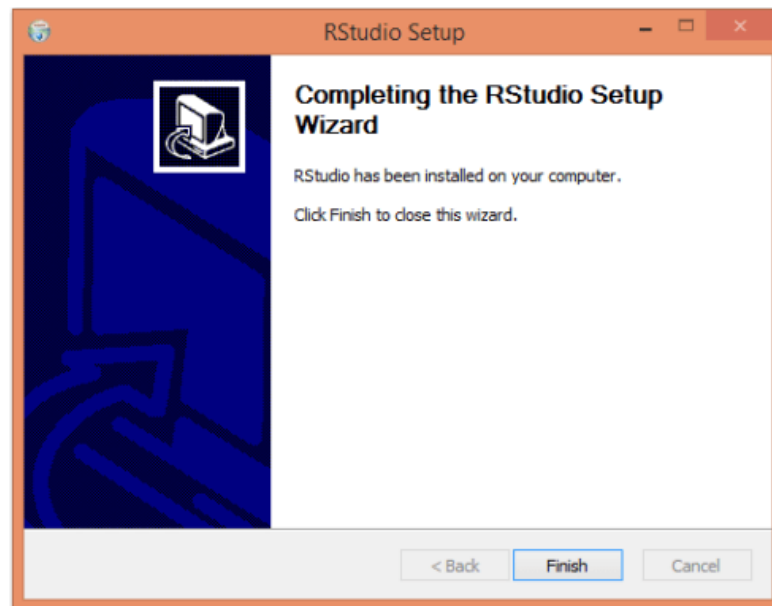
6. Click Install to accept the default start menu folder and install RStudio.



7. Now it will be installing in your system.



8. Next, click Finish to close the wizard.



9. Check if there is an “RStudio” icon on the desktop of the computer. If so, double-click on the “RStudio” icon to start RStudio.

Conclusion:

Thus, R and R Studio has been installed in windows 10.

Lab Outcome: LO1

Lab Assignment: 02

Aim:

Datatypes and Data structures in R.

Code:

```
> #Experiment 2: Datatypes in R
> #Anirudh Poroorkara
> #Roll: 44
>
> #To check your current working directory, you can run the command
getwd() in the RStudio console.
> #Present working directory
> getwd()
[1] "/home/lab1003"
>
> #To change your working directory, use setwd and specify the path to
the desired folder.
> setwd("~/Desktop")
> getwd()
[1] "/home/lab1003/Desktop"
>
> #The dir R function returns a character vector of file and/or folder
names within a directory.
> dir()
[1] "abc" "Assignment No. 7TCPDUMP.docx"
"c.c"
[4] "dsa softfin.odt" "ds.py"
"EDU1664_Launcher"
[7] "KitcheNette-master" "RSA.class"
"RSA.java"
> #ls and objects return a vector of character strings giving the
names of the objects in the specified environment.
> ls()
[1] "A" "B" "C" "c_emp.data"
"colnames" "column.names"
[7] "factor_traffic" "list2" "matrix.names" "merged.list"
"R" "RandomList"
[13] "result" "resultAdd" "resultDiv" "resultMul"
"resultSub" "rownames"
```

```
[19] "row.names"      "traffic"      "v"            "v1"
"v2"              "vector1"
[25] "vector2"        "w"
> objects()
[1] "A"              "B"            "C"            "c_emp.data"
"colnames"        "column.names"
[7] "factor_traffic" "list2"        "matrix.names" "merged.list"
"R"               "RandomList"
[13] "result"          "resultAdd"    "resultDiv"    "resultMul"
"resultSub"        "rownames"
[19] "row.names"      "traffic"      "v"            "v1"
"v2"              "vector1"
[25] "vector2"        "w"
>
> #The help() function and ? help operator in R provide access to the
documentation pages for R functions, data sets, and other objects,
both for packages in the standard R distribution and for contributed
packages.
> help.start()
> help("sum")
> ?summary
> ?mean.Date
>
> #declared variable x with value 1
> x<-44
>
> #The function class prints the vector of names of classes an object
inherits from.
> class(x)
[1] "numeric"
> #The R specific function typeof returns the type of an R object
> typeof(x)
[1] "double"
> #print x
> x
[1] 44
>
> #declared variable y with value Anirudh
> y<-'Anirudh'
> #typeof y
> typeof(y)
[1] "character"
> #class of y
> class(y)
```

```
[1] "character"
> #Check if character
> is.character(y)
[1] TRUE
> #as.integer attempts to coerce its argument to be of integer type.
> as.integer(y)
[1] NA
Warning message:
NAs introduced by coercion
>
> #Creating Sequence from 4 to 14
> z <- 4:14
> z
[1] 4 5 6 7 8 9 10 11 12 13 14
> length(z)
[1] 11
>
> #Vector
> #A vector is a collection of elements that are most commonly of mode
character, logical, integer or numeric.
>
> v <- c(1, 2, 3)
> v
[1] 1 2 3
> w <- c("Ani", "Nivi", "Dhi")
> w
[1] "Ani" "Nivi" "Dhi"
> length(v)
[1] 3
> class(w)
[1] "character"
>
> #Adding Elements in vectors
> w <- c(w, "Muni")
>
> # Accessing vector elements using position.
> w[c(2,3)]
[1] "Nivi" "Dhi"
> # Accessing vector elements using 0/1 indexing.
> w[c(0,0,0,1)]
[1] "Ani"
>
> #Lists
```

```
> # Create a list containing strings, numbers, vectors and a logical values.
```

```
> RandomList <- list("Black", c(21,32,11), TRUE, 44)
```

```
> RandomList
```

```
[[1]]
```

```
[1] "Black"
```

```
[[2]]
```

```
[1] 21 32 11
```

```
[[3]]
```

```
[1] TRUE
```

```
[[4]]
```

```
[1] 44
```

```
>
```

```
> list2 <- list(45,65.5, "Red")
```

```
>
```

```
> # Merge the two lists.
```

```
> merged.list <- c(RandomList,list2)
```

```
>
```

```
> # Print the merged list.
```

```
> print(merged.list)
```

```
[[1]]
```

```
[1] "Black"
```

```
[[2]]
```

```
[1] 21 32 11
```

```
[[3]]
```

```
[1] TRUE
```

```
[[4]]
```

```
[1] 44
```

```
[[5]]
```

```
[1] 45
```

```
[[6]]
```

```
[1] 65.5
```

```
[[7]]
```

```
[1] "Red"
```

```
>
> # Convert the lists to vectors.
> v1 <- unlist(RandomList)
> v2 <- unlist(list2)
> v1
[1] "Black" "21"      "32"      "11"      "TRUE"  "44"
> v2
[1] "45"      "65.5" "Red"
>
> #Matrix
> #Matrices are the R objects in which the elements are arranged in a
two-dimensional rectangular layout. They contain elements of the same
atomic types.
>
> # Elements are arranged sequentially by row.
> A <- matrix(c(4:15), nrow = 4, byrow = TRUE)
> A
      [,1] [,2] [,3]
[1,]     4     5     6
[2,]     7     8     9
[3,]    10    11    12
[4,]    13    14    15
>
> # Elements are arranged sequentially by column.
> B <- matrix(c(4:15), nrow = 4, byrow = FALSE)
> B
      [,1] [,2] [,3]
[1,]     4     8    12
[2,]     5     9    13
[3,]     6    10    14
[4,]     7    11    15
>
> # Define the column and row names.
> rownames = c("R1", "R2", "R3", "R4")
> colnames = c("C1", "C2", "C3")
> C <- matrix(c(10:21), nrow = 4, byrow = TRUE, dimnames =
list(rownames, colnames))
> C
      C1 C2 C3
R1 10 11 12
R2 13 14 15
R3 16 17 18
R4 19 20 21
```

```
>
> #Add matrix A and B
> resultAdd <- A + B
> resultAdd
      [,1] [,2] [,3]
[1,]    8   13   18
[2,]   12   17   22
[3,]   16   21   26
[4,]   20   25   30
>
> #Subtract matrix A and B
> resultSub <- A - B
> resultSub
      [,1] [,2] [,3]
[1,]    0   -3   -6
[2,]    2   -1   -4
[3,]    4    1   -2
[4,]    6    3    0
>
> #Multiply matrix A and B
> resultMul <- A * B
> resultMul
      [,1] [,2] [,3]
[1,]   16   40   72
[2,]   35   72  117
[3,]   60  110  168
[4,]   91  154  225
>
> #Divide matrix A and B
> resultDiv <- A / B
> resultDiv
      [,1]      [,2]      [,3]
[1,] 1.000000 0.6250000 0.5000000
[2,] 1.400000 0.8888889 0.6923077
[3,] 1.666667 1.1000000 0.8571429
[4,] 1.857143 1.2727273 1.0000000
>
> #Arrays
> #Arrays are the R data objects which can store data in more than two
dimensions
> # Create two vectors of different lengths.
> v1 <- c(4,90,23)
> v2 <- c(20,45,27,44,87,39)
> column.names <- c("C1","C2","C3")
```



```
> row.names <- c("R1","R2","R3")
> matrix.names <- c("Matrix_1","Matrix_2")
>
> # Take these vectors as input to the array.
> R <- array(c(vector1,vector2),dim = c(3,3,2),dimnames =
list(row.names,column.names,matrix.names))
> R
, , Matrix_1

  C1 C2 C3
R1  4 14 17
R2 90 15 18
R3 23 16 19

, , Matrix_2

  C1 C2 C3
R1  4 14 17
R2 90 15 18
R3 23 16 19

>
>
> # Print the third row of the second matrix of the array.
> R[3,,2]
C1 C2 C3
23 16 19
>
> # Print the element in the 1st row and 3rd column of the 1st matrix.
> R[1,3,1]
[1] 17
>
> #Calculate sum of all each row in Array
> rowSums(R)
R1 R2 R3
70 246 116
> #Calculate sum of columns in Array
> colSums(R)
  Matrix_1 Matrix_2
C1      117      117
C2       45       45
C3       54       54
>
>
```

```
> #Factor
> #Factors are the data objects which are used to categorize the data
and store it as levels. They can store both strings and integers. They
are useful in the columns which have a limited number of unique
values.
>
> # Create a vector as input.
> traffic <-
c("Yello", "Green", "Yello", "Red", "Red", "Yello", "Green", "Green", "Green",
"Yello", "Red")
> traffic
[1] "Yello" "Green" "Yello" "Red"   "Red"   "Yello" "Green" "Green"
"Green" "Yello" "Red"
> is.factor(traffic)
[1] FALSE
>
> # Apply the factor function.
> factor_traffic <- factor(traffic)
> factor_traffic
[1] Yello Green Yello Red   Red   Yello Green Green Green Yello Red
Levels: Green Red Yello
> is.factor(factor_traffic)
[1] TRUE
>
> #DataFrame
> #A data frame is a table or a two-dimensional array-like structure
in which each column contains values of one variable and each row
contains one set of values from each column.
>
> # Create the data frame.
>
> c_emp.data <- data.frame(
+   c_id = c (1:5),
+   c_name = c("Vedant", "Amol", "Muneeb", "Maitreyee", "Shah"),
+   c_roll = c(43, 132, 32, 28, 39) ,
+   stringsAsFactors = FALSE
+ )
>
> # Print the data frame.
> c_emp.data
  c_id  c_name c_roll
1    1  Vedant    43
2    2   Amol   132
3    3  Muneeb    32
```

```
4    4 Maitreyee    28
5    5      Shah    39
>
> # Get the structure of the data frame.
> str(c_emp.data)
'data.frame':   5 obs. of  3 variables:
 $ c_id  : int   1 2 3 4 5
 $ c_name: chr   "Vedant" "Amol" "Muneeb" "Maitreyee" ...
 $ c_roll: num   43 132 32 28 39
>
> # Print the summary.
> print(summary(c_emp.data))
      c_id      c_name      c_roll
Min.   :1   Length:5   Min.    : 28.0
1st Qu.:2   Class :character 1st Qu.: 32.0
Median :3   Mode  :character Median : 39.0
Mean    :3                                Mean  : 54.8
3rd Qu.:4                                3rd Qu.: 43.0
Max.    :5                                Max.   :132.0
>
> # Extract first two rows.
> result <- c_emp.data[1:2,]
> result
  c_id c_name c_roll
1    1 Vedant    43
2    2  Amol   132
>
> # Extract Specific columns.
> result <- data.frame(c_emp.data$c_name,c_emp.data$c_roll)
> result
  c_emp.data.c_name c_emp.data.c_roll
1          Vedant          43
2           Amol         132
3          Muneeb          32
4        Maitreyee          28
5           Shah          39
>
> # Add the batch coulmn.
> c_emp.data$batch <- c("B1","B3","B6","B5","B4")
> v <- c_emp.data
> v
  c_id      c_name c_roll batch
1    1    Vedant    43    B1
2    2     Amol   132    B3
```

3	3	Muneeb	32	B6
4	4	Maitreyee	28	B5
5	5	Shah	39	B4

Conclusion:

Thus, datatypes and structures were executed in R.

Lab Outcome: LO1, LO2

Lab Assignment: 03

Aim:

Loops and Conditions in R

Code:

```
> #Lab Assignment 3
> #Loops and conditions
>
> #get the working directory
> getwd()
[1] "/cloud/project"
>
> #Control Structures
> #Control structures in R allow you to control the flow of execution
of a series of R expressions. Basically, control structures allow you
to put some "logic" into your R code, rather than just always executin
g the same R code every time. Control structures allow you to respond
to inputs or to features of the data and execute different R expressio
ns accordingly.

> #1. IF-ELSE
> #The if-else combination is probably the most commonly used control
structure not only in R but also for other programming languages. This
structure allows you to test a condition and act on it depending on wh
ether it's true or false.
> x<-1000
> if(x > 99){
+   print("X is greater than 100")
+ }
[1] "X is greater than 100"
>
> x <- 9
> if(x > 10){
+   print("Value is greater than 10")
+ } else {
+   print("Value is less than or equal to 10")
+ }
[1] "Value is less than or equal to 10"
```

```
> #2. FOR loop
> #For loops are most commonly used for iterating over the elements of
an object (list, vector, etc.)
>
> for(i in 4:12) {
+   print(i)
+ }
[1] 4
[1] 5
[1] 6
[1] 7
[1] 8
[1] 9
[1] 10
[1] 11
[1] 12
>
> x <- c("ani", "nivi", "dhi", "parth")
> for(letter in x) {
+   print(letter)
+ }
[1] "ani"
[1] "nivi"
[1] "dhi"
[1] "parth"

> #3. WHILE loop
> #While loops begin by testing a condition. If it is true, then they
execute the loop body. Once the loop body is executed, the condition i
s tested again, and so forth, until the condition is false, after whic
h the loop exits.
>
> c <- 0
> while(c < 5) {
+   print(c)
+   c <- c + 1
+ }
[1] 0
[1] 1
[1] 2
[1] 3
[1] 4
>
> x <- c("ani", "nivi", "dhi", "parth")
```

```
> i <- 0
> while(i < 5) {
+   print(x[i])
+   i <- i+1
+ }
character(0)
[1] "ani"
[1] "nivi"
[1] "dhi"
[1] "parth"
```

> #4. REPEAT loop

> #A repeat loop is used to iterate over a block of code multiple number of times. There is no condition check in repeat loop to exit the loop. The only way to exit a repeat loop is to call break. These are not commonly used in statistical or data analysis applications but they do have their uses.

> #When the break statement is encountered inside a loop, the loop is immediately terminated and program control resumes at the next statement following the loop.

```
>
> x <- 5
>
> repeat {
+   print(x)
+   x = x+1
+   if (x == 10){
+     break
+   }
+ }
[1] 5
[1] 6
[1] 7
[1] 8
[1] 9
```

> #The next statement in R programming language is useful when we want to skip the current iteration of a loop without terminating it. On encountering next, the R parser skips further evaluation and starts next iteration of the loop.

```
> for(i in 1:10) {
+
+   if(i >= 5) {
+     # Skip the first 5 iterations
```

```
+     next
+   }
+   print(i)
+ }
[1] 1
[1] 2
[1] 3
[1] 4
```

```
> #Functions
> #Functions are defined using the function() directive and are stored
as R objects
>
> #Basic Simple function
> #define the function
> f <- function() {
+   print("Hi Anirudh!!!")
+ }
> #Call the function
> f()
[1] "Hi Anirudh!!!"
>
> #Build-in Functions
> # Create a sequence of numbers 4 to 14
> print(seq(4,14))
[1] 4 5 6 7 8 9 10 11 12 13 14
>
> # Find mean of numbers 4 to 14
> print(mean(4:14))
[1] 9
>
> # Find sum of numbers 4 to 14
> print(sum(4:14))
[1] 99
>
> #User defined function
> pow <- function(x, y = 2) {
+   result <- x^y
+   print(paste(x,"raised to the power", y, "is", result))
+ }
>
> #Give both arguments
> pow(2,3)
```



```
[1] "2 raised to the power 3 is 8"
>
> #Give one argument, the default value of other argument is taken
> pow(2)
[1] "2 raised to the power 2 is 4"
>
> #Return statement
> #The return statement terminates the execution of a function and returns control to the calling function.
>
> calculate <- function(x) {
+   if (x > 0) {
+     result <- "Positive"
+   } else if (x < 0) {
+     result <- "Negative"
+   } else {
+     result <- "Zero"
+   }
+   return(result)
+ }
>
> calculate(9)
[1] "Positive"
> calculate(-5)
[1] "Negative"
> calculate(0)
[1] "Zero"

> #Different math functions
> x = 3.456
> n = 2
> #absolute value
> abs(x)
[1] 3.456
> #ceiling(3.475) is 4
> ceiling(x)
[1] 4
> #square root
> sqrt(x)
[1] 1.859032
> #floor(3.475) is 3
> floor(x)
[1] 3
> #natural logarithm
```

```
> log(x)
[1] 1.240112
> #trunc(5.99) is 5
> trunc(x)
[1] 3
> #round(3.475, digit=2) is 3.48
> round(x, digits=n)
[1] 3.46
> #common logarithm
> log10(x)
[1] 0.5385737
> #signif(3.475, digit=2) is 3.5
> signif(x, digits=n)
[1] 3.5
> #e^x
> exp(x)
[1] 31.68996
> #Trigonometric functions
> cos(x)
[1] -0.9509798
> sin(x)
[1] -0.3092529
> tan(x)
[1] 0.325194
```

Conclusion:

Thus functions, loops and conditions was studied and implemented in R.

Lab Outcome: LO1

Lab Assignment: 04

Aim:

EDA on inbuilt R dataset

Code:

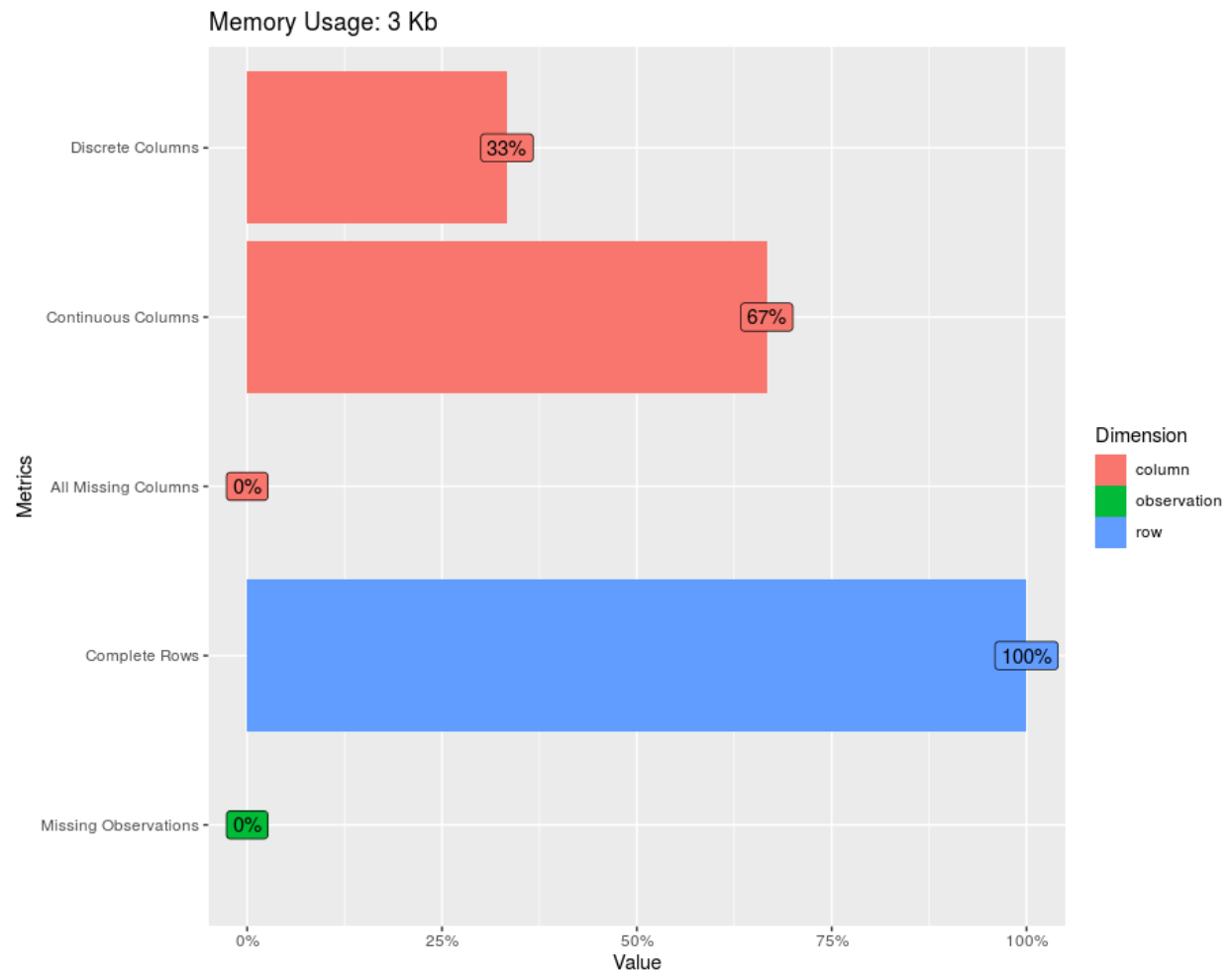
```
> #Lab Assignment 4
> #Exploratory Data Analysis
>
> #Dataset name - UFC Fight Data
> #Kaggle link
>
>
> #get the working directory
> getwd()
[1] "/cloud/project"
>
> #Choosing dataset into R
> #data() function is used to list all the internal datasets present
> data()
>
> #we load the chicken weights dataset
> data("Orange")
>
> # ? is used to get information about any entity in R
> # here we use ?(Orange) to get information about the Orange dataset
> ?(Orange)
>
> #load the dataset in a variable
> df <- Orange
>
> #Check the class before doing any cleaning
> class(df)
[1] "nfnGroupedData" "nfGroupedData"
[3] "groupedData"    "data.frame"
>
> #Check the number of rows and columns the data frame has
> dim(df)
[1] 35 3
> #We can see that the data frame has 35 rows and 3 columns
>
```

```
> #Finding summary of entire dataset
> summary(df)
  Tree      age      circumference
3:7  Min.    : 118.0    Min.      : 30.0
1:7  1st Qu.: 484.0    1st Qu.: 65.5
5:7  Median :1004.0    Median :115.0
2:7  Mean   : 922.1    Mean     :115.9
4:7  3rd Qu.:1372.0    3rd Qu.:161.5
      Max.    :1582.0    Max.      :214.0

>
> #nrow() function is used to print the number of rows in dataset
> nrow(df)
[1] 35
>
> #ncol() function is used to print the number of columns in the dataset
> ncol(df)
[1] 3
>
> #check for null values in dataset
> is.null(df)
[1] FALSE
>
> #duplicated function is used to display duplicate values
> #it prints 'false' for non duplicate values and 'true' for duplicate values
> duplicated(df)
[1] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[9] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[17] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[25] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[33] FALSE FALSE FALSE
>
> #to remove the duplicate value we use Unique function
> #it returns only the unique values in the dataset
> dfd<-unique(df)
>
> #check again if all duplicate values are removed
> duplicated(dfd)
[1] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[9] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[17] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[25] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[33] FALSE FALSE FALSE
```

```
> #Exploratory data analysis is the process to get to know your data,
so that you can generate and test your hypothesis. Visualization techn
iques are usually applied.
> #EDA consists of univariate (1-variable) and bivariate (2-variables)
analysis.
> #DataExplorer can help you with different tasks throughout your data
exploration process.
> #Install the package if not installed
> #install.packages('DataExplorer')
> #install.packages('GGally')
>
> #import the library
> library(DataExplorer)
> library(GGally)
>
> #To get introduced to your newly created dataset:
> introduce(df)
  rows columns discrete_columns continuous_columns all_missing_columns
total_missing_values complete_rows total_observations
1    35         3                1                2                0
0         35         105
  memory_usage
1         3040
>
> #plot_intro gives a brief Introduction to the dataset.
> #It covers basic information and gives us an idea about the content.
> plot_intro(df)
```

>



> #str gives the entire list of features in a network graph. We have the choice of viewing it radially or diagonally

> plot_str(df)

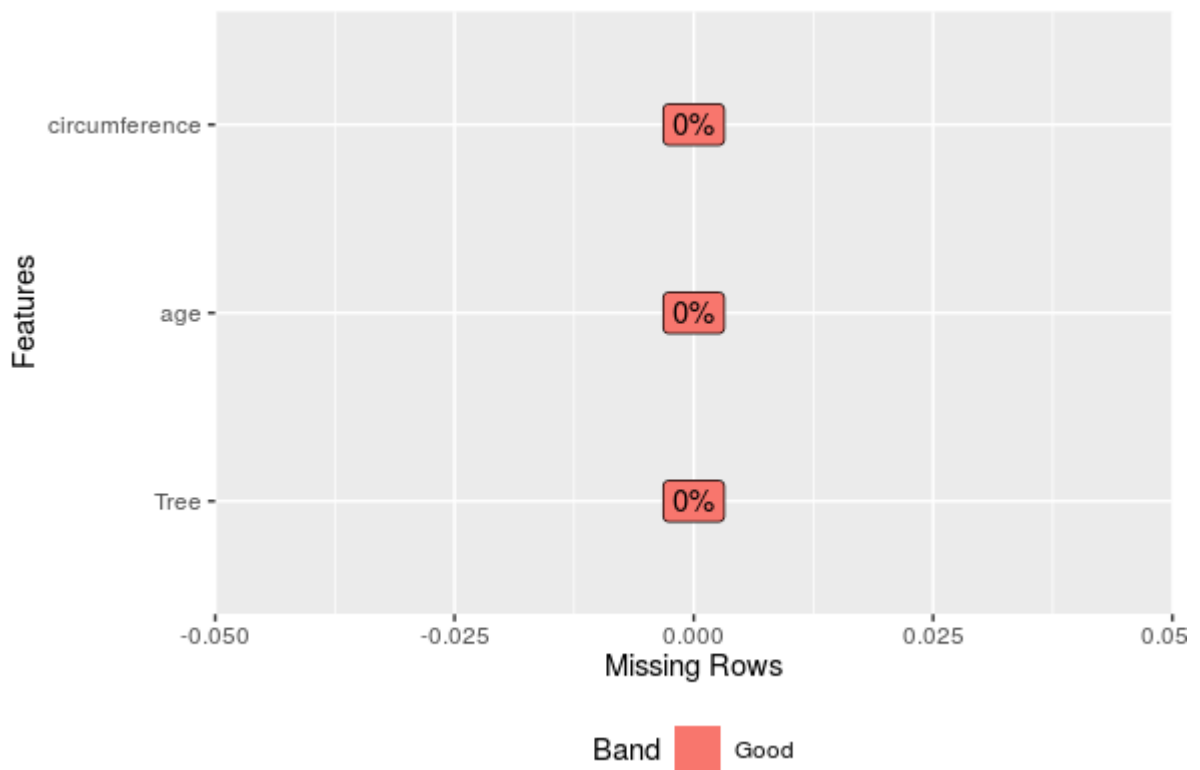
>

> #missing function returns and plots frequency of missing values for each feature.

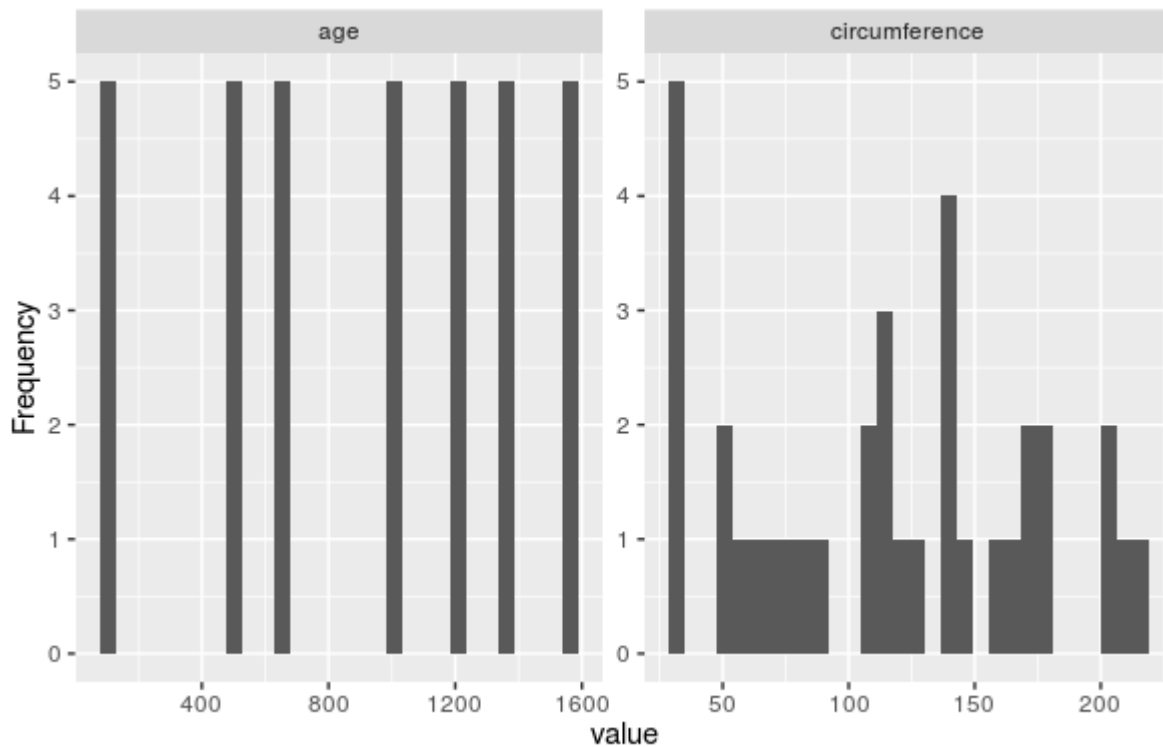
> #It also advises us whether to remove certain columns before carrying out our analysis

> #All our features are in acceptable form

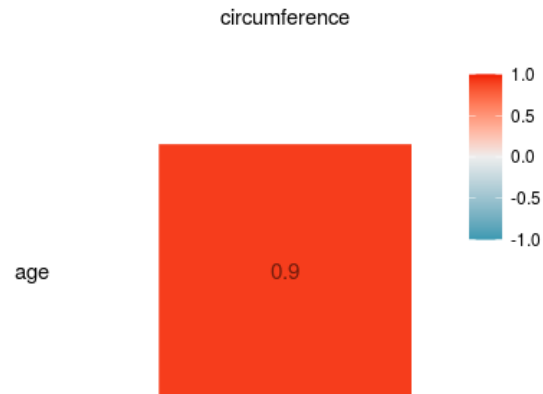
> plot_missing(df)



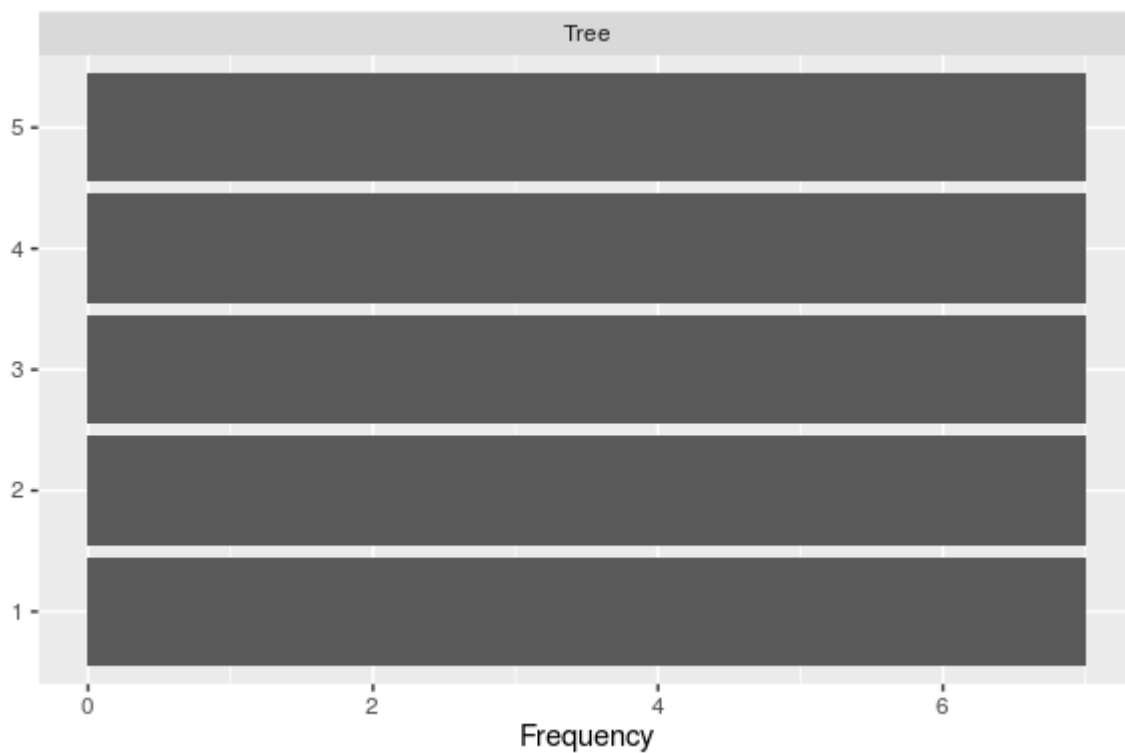
```
>  
> #Let us first analyse and represent the continuous variables  
> #Histograms can be used to analyse continuous variables  
> # you can use a Histogram to organize and display the data in a more  
userfriendly format.  
> #A Histogram will make it easy  
> #to see where the majority of values falls in a measurement scale, a  
nd how much variation there is.  
> plot_histogram(df)
```



```
>  
> #For multivariate analysis, let us do correlation analysis  
> #Correlation is used to test relationships between quantitative variables or categorical variables.  
> #It's a measure of how things are related and how well they are related.  
> ggcorr(df, label =TRUE, label_alpha =TRUE)  
Warning message:  
In ggcorr(df, label = TRUE, label_alpha = TRUE) :  
  data in column(s) 'Tree' are not numeric and were ignored
```

```
>  
> #For categorical analysis, we use a bar graph  
> #Each bar represents one value. When the bars are stacked next to one another,  
> #the viewer can compare the different bars, or values, at a glance.  
> plot_bar(df)
```



Conclusion:

EDA was conducted on the inbuilt dataset.

Lab Outcome: LO3, LO4

Lab Assignment: 05

Aim:

Plot graphs using GGPLOT2 on the internal dataset.

Code:

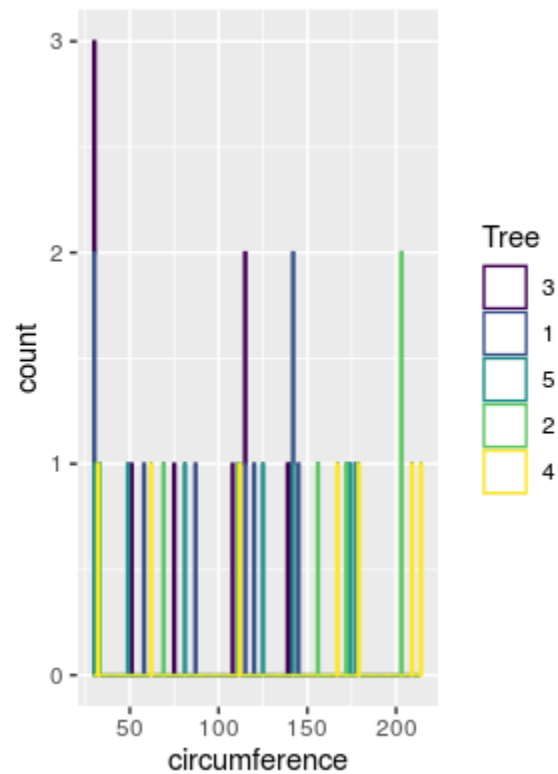
```
> #Lab Assignment 5
> #Data Visualization using ggplot2
>
> #get the working directory
> getwd()
[1] "/cloud/project"
>
> #load the dataset in a variable
> df <- Orange
>
> #ggplot2 is a robust and a versatile R package for generating aesthetic plots and charts.
> #Plot = data + Aesthetics + Geometry
> #1. Data refers to a data frame
> #2. Aesthetics indicates x and y variables. It is also used to tell R how data are displayed in a plot, e.g. color, size and shape of points etc.
> #3. Geometry refers to the type of graphics
>
> #install the package if not installed
> #install.packages("ggplot2")
> library(ggplot2)
> df
```

	Tree	age	circumference
1	1	118	30
2	1	484	58
3	1	664	87
4	1	1004	115
5	1	1231	120
6	1	1372	142
7	1	1582	145
8	2	118	33
9	2	484	69
10	2	664	111
11	2	1004	156
12	2	1231	172
13	2	1372	203
14	2	1582	203
15	3	118	30
16	3	484	51
17	3	664	75
18	3	1004	108

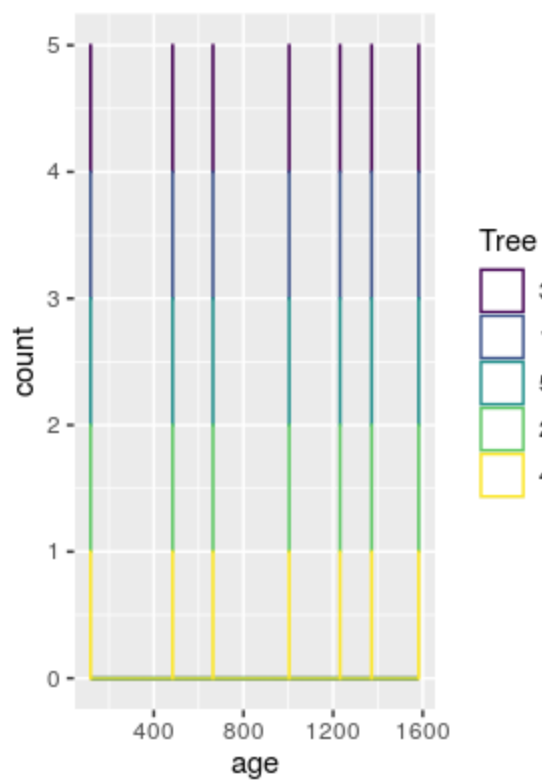
19	3	1231	115
20	3	1372	139
21	3	1582	140
22	4	118	32
23	4	484	62
24	4	664	112
25	4	1004	167
26	4	1231	179
27	4	1372	209
28	4	1582	214
29	5	118	30
30	5	484	49
31	5	664	81
32	5	1004	125
33	5	1231	142
34	5	1372	174
35	5	1582	177

```
> #Plotting with R_Reach and winner using a group histogram
```

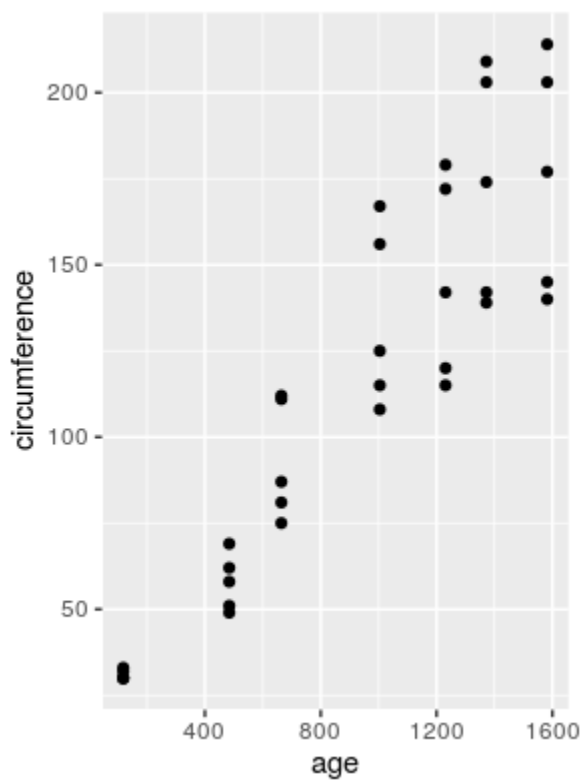
```
> ggplot(data = df, aes( x = circumference, color=Tree)) + geom_histogram(fill="white", binwidth = 1)
```



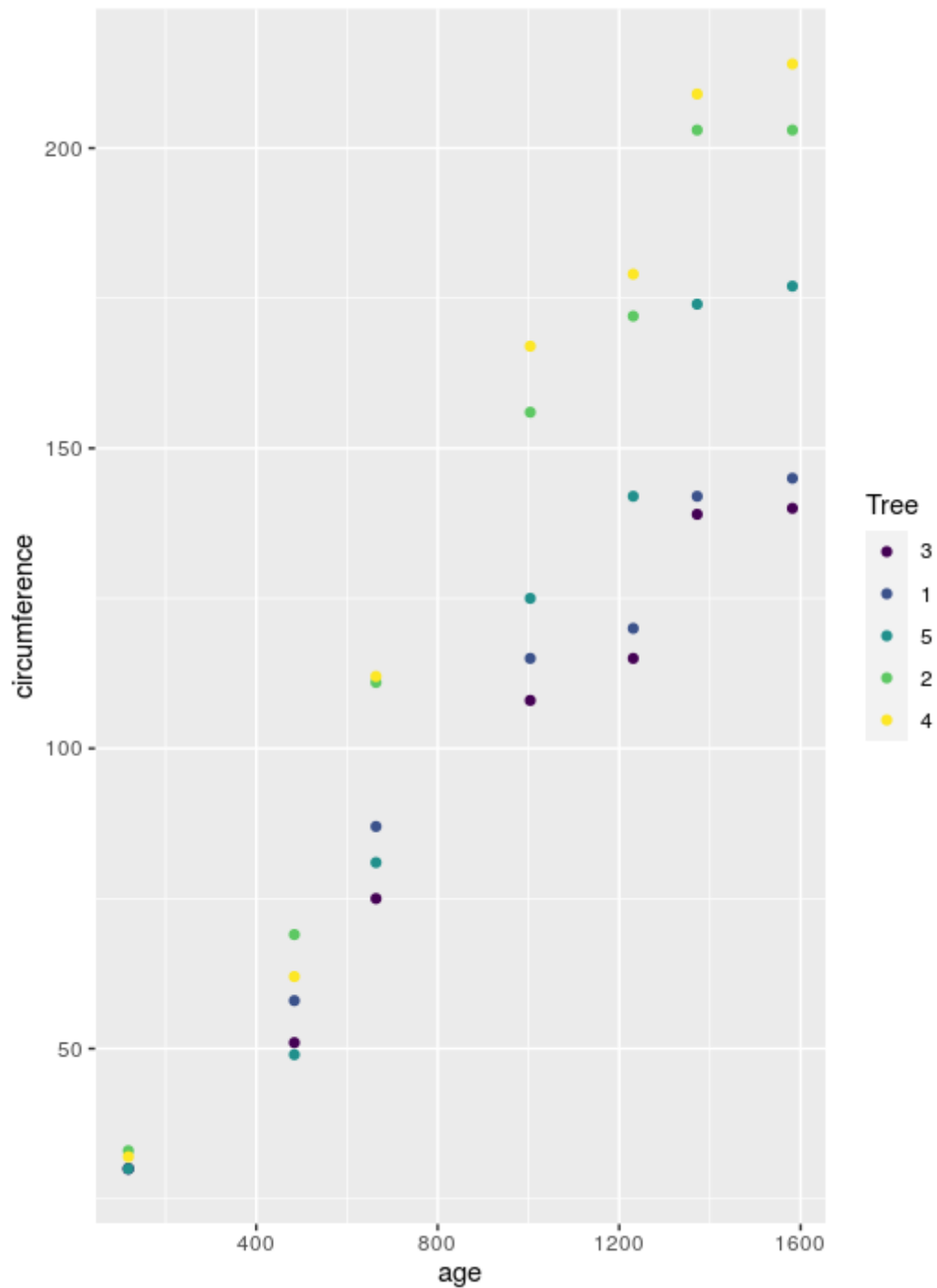
```
> ggplot(data = df, aes( x = age, color=Tree)) + geom_histogram(fill="white", binwidth = 1)
```



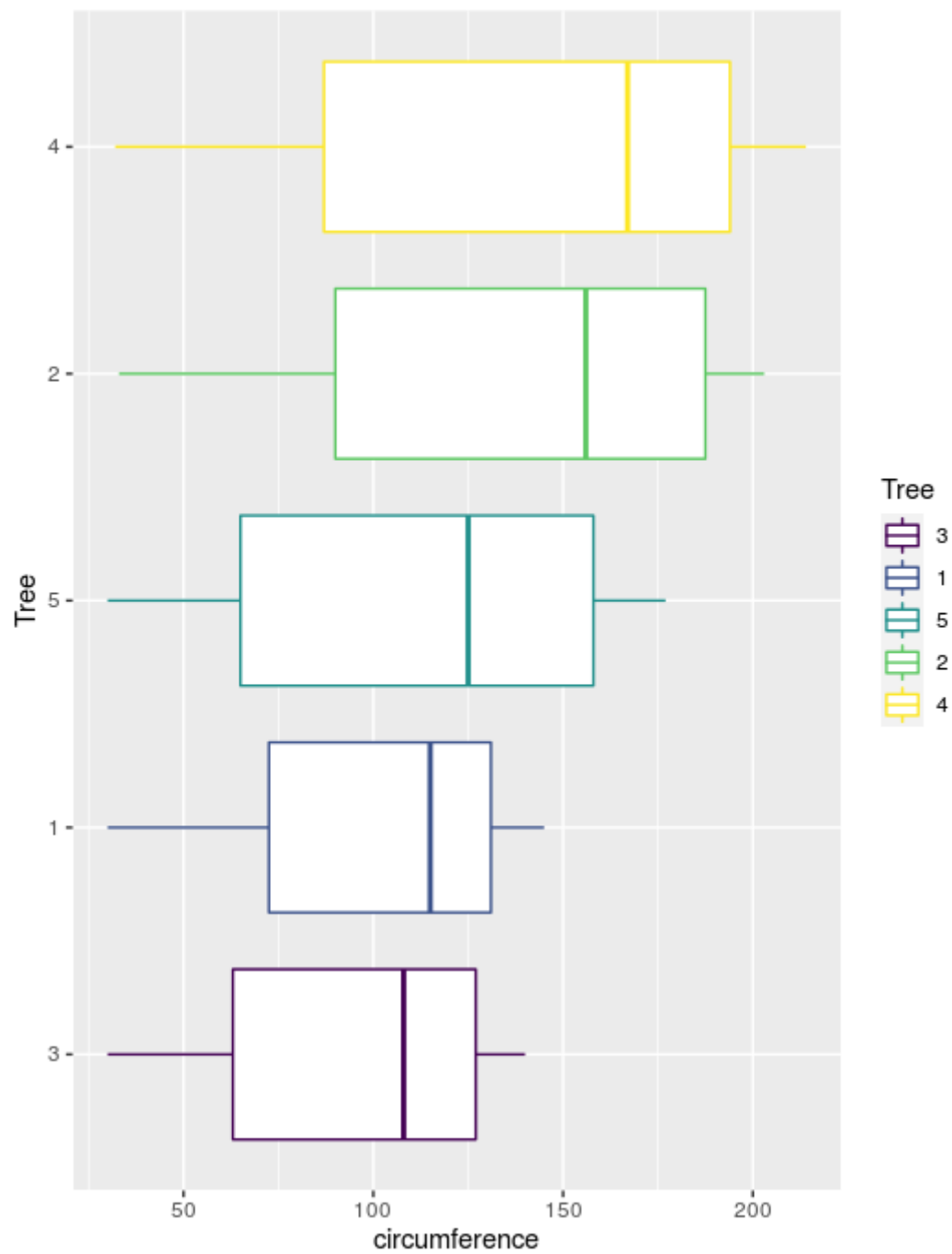
```
> ggplot(data = df, mapping = aes(x = age, y = circumference)) +geom_point()
```



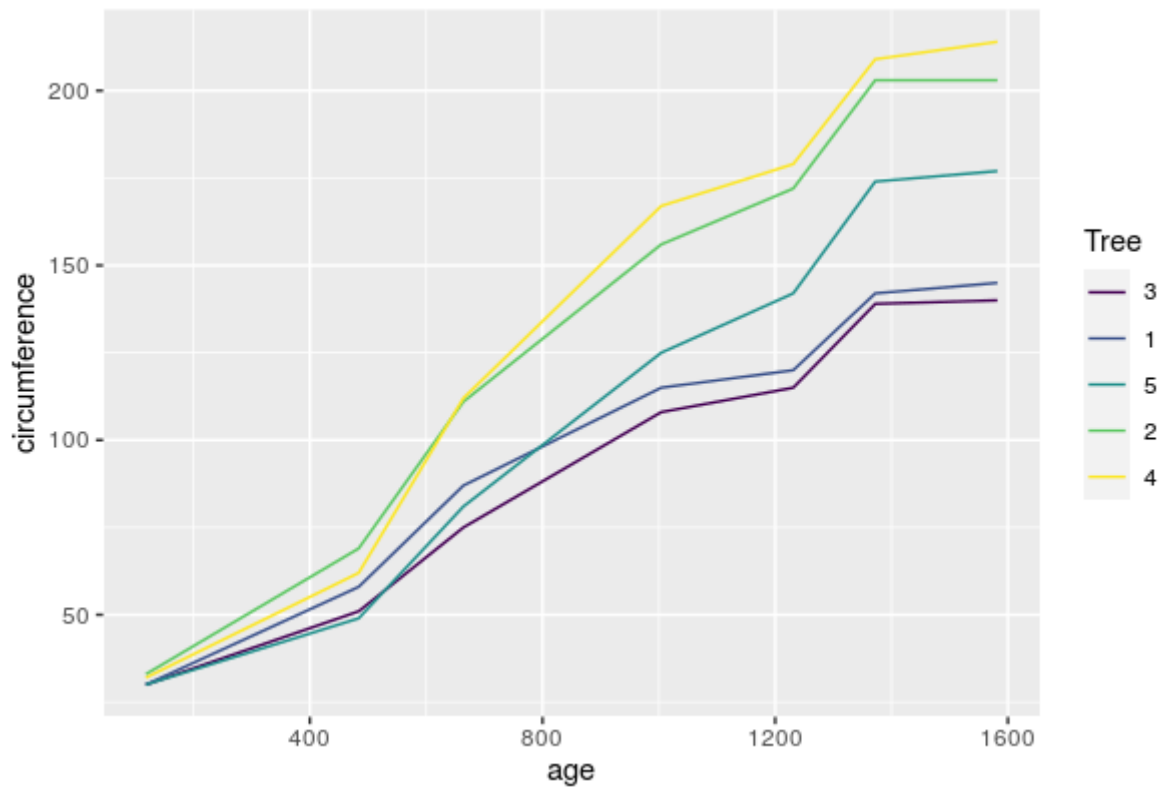
```
> ggplot(data = df, mapping = aes(x = age, y = circumference)) +geom_point(aes(color = Tree))
```



```
> ggplot(data = df, mapping = aes(x = circumference, y = Tree)) +geom_boxplot(aes(color = Tree))
```



```
> ggplot(data = df, mapping = aes(x = age, y = circumference, color = Tree)) +  
  geom_line()
```



Conclusion:

Thus, graphs were plotted using ggplot2.

Lab Outcome: LO2, LO5

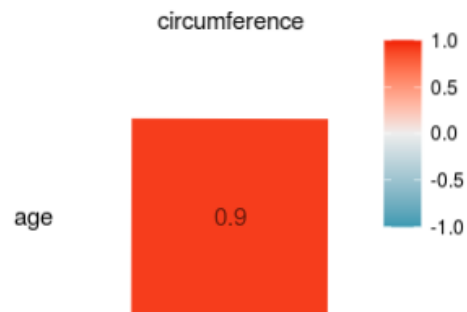
Lab Assignment: 06

Aim:

Regression and Correlation

Code:

```
> #Lab Assignment 6
> #Regression and Correlation
>
>
> #get the working directory
> getwd()
[1] "/cloud/project"
>
> #Choosing dataset into R
> #data() function is used to list all the internal datasets present
> data()
>
> library(ggplot2)
> #install.packages("GGally")
> library(GGally)
> #we load the chicken weights dataset
> data("Orange")
>
> #For multivariate analysis, let us do correlation analysis
> #Correlation is used to test relationships between quantitative variables or categorical variables.
> #It's a measure of how things are related and how well they are related.
> ggcorr(df, label =TRUE, label_alpha =TRUE)
Warning message:
In ggcorr(df, label = TRUE, label_alpha = TRUE) :
  data in column(s) 'Tree' are not numeric and were ignored
```



```
>
> #Regression analysis is used in stats to find trends in data
> #It will provide you with an equation for a graph so that you can make predictions about your data
> #Since we have a lot of values, we use multiple regression analysis
> #Multiple linear regression is an extension of simple linear regression used to
> #predict an outcome variable (y) on the basis of multiple distinct predictor variables (x).
>
> #Here we will predict the circumference based on age of the tree.
>
> model <- lm(circumference ~ age ,data = df)
> summary(model)
```

Call:

```
lm(formula = circumference ~ age, data = df)
```

Residuals:

Min	1Q	Median	3Q	Max
-46.310	-14.946	-0.076	19.697	45.111

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	17.399650	8.622660	2.018	0.0518 .
age	0.106770	0.008277	12.900	1.93e-14 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 23.74 on 33 degrees of freedom

Multiple R-squared: 0.8345, Adjusted R-squared: 0.8295

F-statistic: 166.4 on 1 and 33 DF, p-value: 1.931e-14

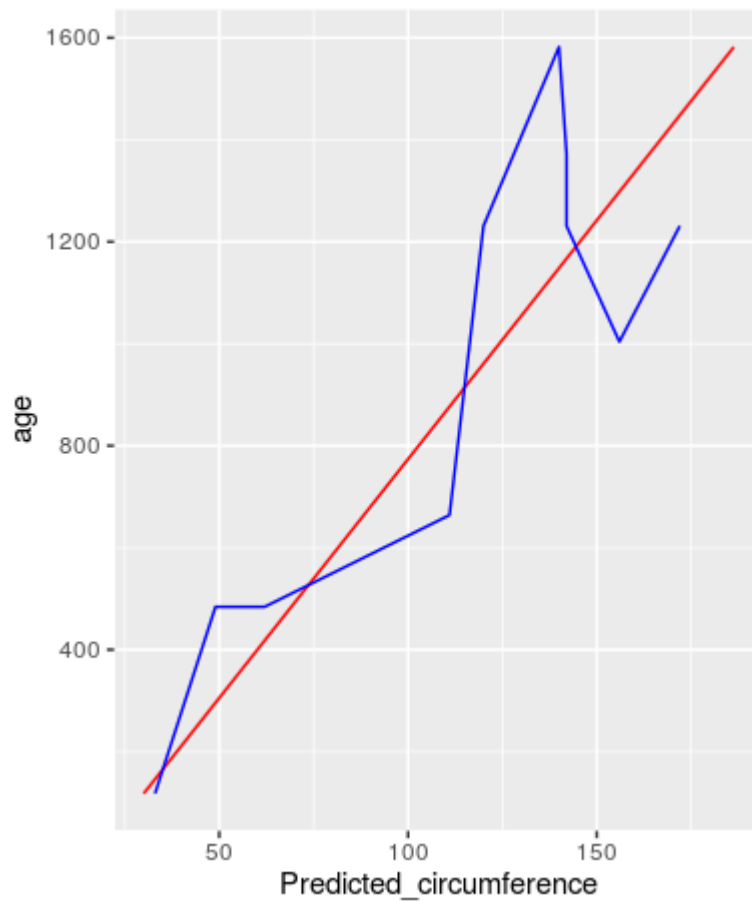
```
>
>
> #The first step in interpreting the regression analysis is
> #to examine the F-statistic and the associated p-value, at the bottom of model summary.
> summary(model)$coefficient
              Estimate Std. Error  t value    Pr(>|t|)
(Intercept) 17.3996502  8.622659801  2.017898 5.179267e-02
age          0.1067703  0.008276623 12.900228 1.930596e-14
>
> #For a given the predictor, the t-statistic evaluates whether or not
there is significant association between the predictor and the outcome
variable,
> #that is whether the beta coefficient of the predictor is significantly
different from zero.
>
> #The RSE estimate gives a measure of error of prediction. The lower
the RSE, the more accurate the model (on the data in hand).
> #The error rate can be estimated by dividing the RSE by the mean outcome
variable:
> sigma(model)/mean(df$circumference)
[1] 0.2048874
>
> #We get a 0.204 which means it has a 20% error rate
>
> #Let us now test the created prediction model
> #install.packages("caTools")
> library(caTools)
> #Splitting the data set into training and testing data
> split = sample.split(df$circumference, SplitRatio = 2/3)
> #Training is 2 parts
> training_set = subset(df, split == TRUE)
> #testing is 1 part
> test_set = subset(df, split == FALSE)
>
> test_set
  Tree age circumference
2    1  484             58
5    1 1231            120
6    1 1372            142
8    2  118             33
10   2  664            111
11   2 1004            156
```

12	2	1231	172
16	3	484	51
21	3	1582	140
23	4	484	62
30	5	484	49
33	5	1231	142

```
>
> #Do a prediction on the model
> pred = predict(model, newdata = test_set)
>
> #Store and compare the result in the test cases
> result <- data.frame(age = test_set$age, circumference = test_set$ci
rcumference, Predicted_circumference = pred)
> result
```

	age	circumference	Predicted_circumference
2	484	58	69.07649
5	1231	120	148.83392
6	1372	142	163.88854
8	118	33	29.99855
10	664	111	88.29515
11	1004	156	124.59706
12	1231	172	148.83392
16	484	51	69.07649
21	1582	140	186.31030
23	484	62	69.07649
30	484	49	69.07649
33	1231	142	148.83392

```
> ggplot(data = result) + geom_line(aes(x=Predicted_circumference, y=a
ge), color='red') + geom_line(aes(x=circumference, y=age), color='blue
')
```



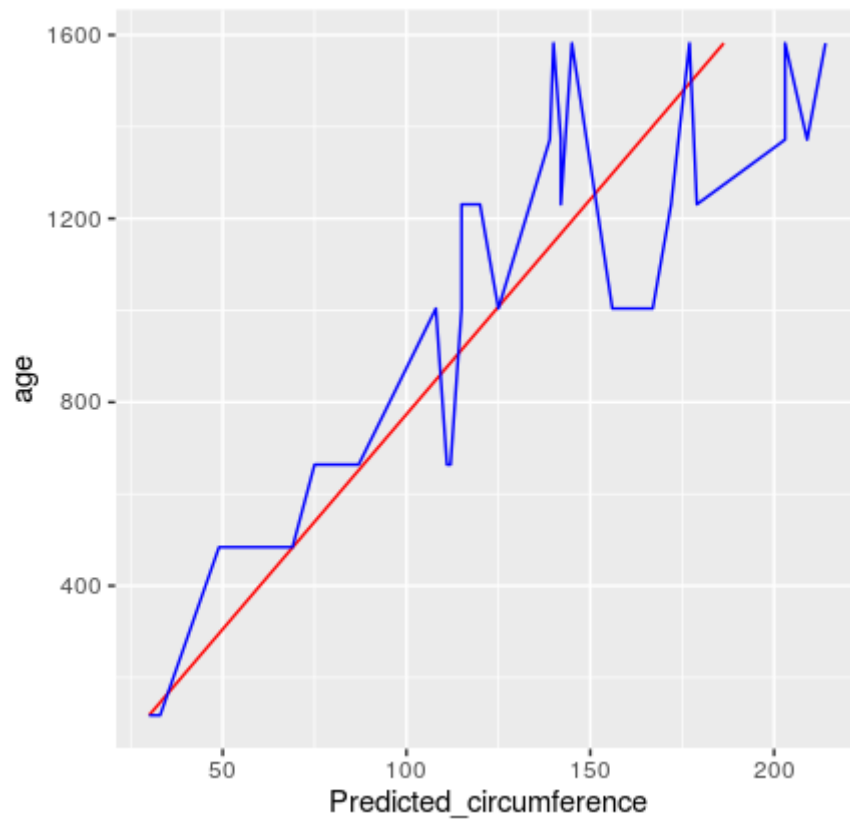
```
> #Do entire df prediction on the model
> pred = predict(model, newdata = df)
> result <- data.frame(age = df$age, circumference = df$circumference,
Predicted_circumference = pred)
> result
```

	age	circumference	Predicted_circumference
1	118	30	29.99855
2	484	58	69.07649
3	664	87	88.29515
4	1004	115	124.59706
5	1231	120	148.83392
6	1372	142	163.88854
7	1582	145	186.31030
8	118	33	29.99855
9	484	69	69.07649
10	664	111	88.29515
11	1004	156	124.59706
12	1231	172	148.83392
13	1372	203	163.88854

14	1582	203	186.31030
15	118	30	29.99855
16	484	51	69.07649
17	664	75	88.29515
18	1004	108	124.59706
19	1231	115	148.83392
20	1372	139	163.88854
21	1582	140	186.31030
22	118	32	29.99855
23	484	62	69.07649
24	664	112	88.29515
25	1004	167	124.59706
26	1231	179	148.83392
27	1372	209	163.88854
28	1582	214	186.31030
29	118	30	29.99855
30	484	49	69.07649
31	664	81	88.29515
32	1004	125	124.59706
33	1231	142	148.83392
34	1372	174	163.88854
35	1582	177	186.31030

```
> #Plotting model with results
```

```
> ggplot(data = result) + geom_line(aes(x=Predicted_circumference, y=age), color='red') + geom_line(aes(x=circumference, y=age), color='blue')
'
```



Conclusion:

Thus, regression and correlation analysis were done on the internal dataset and graphs were plotted.

Lab Outcome: LO4

Lab Assignment: 07

Aim:

Choose Dataset from Kaggle, extracting data from large dataset

Code:

```
> #Lab Assignment 7
> #Mini-Project Session 1
>
> #Dataset name - UFC Fight Data
> #Kaggle link
>
>
> #get the working directory
> getwd()
[1] "/cloud/project"
>
> #Loading dataset into R
> #Header is set to true if file contains Header information
> #Sep stands for separator which is , in our csv file
>
> df <- read.csv("data.csv", header = TRUE, sep = ",")
>
> #Display data file
> head(df)
```

	R_fighter	B_fighter	Referee	date
location Winner				
1	Henry Cejudo	Marlon Moraes	Marc Goddard	2019-06-08 Chicago, Illinois, USA
2	Valentina Shevchenko	Jessica Eye	Robert Madrigal	2019-06-08 Chicago, Illinois, USA
3	Tony Ferguson	Donald Cerrone	Dan Miragliotta	2019-06-08 Chicago, Illinois, USA
4	Jimmie Rivera	Petr Yan	Kevin MacDonald	2019-06-08 Chicago, Illinois, USA
5	Tai Tuivasa	Blagoy Ivanov	Dan Miragliotta	2019-06-08 Chicago, Illinois, USA
6	Tatiana Suarez	Nina Ansaroff	Robert Madrigal	2019-06-08 Chicago, Illinois, USA
title_bout		weight_class	no_of_rounds	B_current_lose_streak
current_win_streak	B_draw			
1	True	Bantamweight	5	0
4	0			

2	True	Women's Flyweight	5	0
3	0			
3	False	Lightweight	3	0
3	0			
4	False	Bantamweight	3	0
4	0			
5	False	Heavyweight	3	0
1	0			
6	False	Women's Strawweight	3	0
4	0			

	B_avg_BODY_att	B_avg_BODY_landed	B_avg_CLINCH_att	B_avg_CLINCH_landed
1	9.20000	6.00000	0.200000	0.00000
0	62.60000			
2	14.60000	9.10000	11.800000	7.30000
0	124.70000			
3	15.35484	11.32258	6.741935	4.38709
7	84.74194			
4	17.00000	14.00000	13.750000	11.00000
0	109.50000			
5	17.00000	14.50000	2.500000	2.00000
0	201.00000			
6	19.50000	12.33333	11.833333	7.16666
7	142.33333			

	B_avg_DISTANCE_landed	B_avg_GROUND_att	B_avg_GROUND_landed	B_avg_HEA
1	20.60000	2.600000	2.000000	48.
60000	11.20000			
2	42.10000	2.400000	1.900000	112.
00000	32.00000			
3	38.58065	5.516129	3.806452	67.
64516	23.25806			
4	48.75000	13.000000	10.500000	116.
25000	53.75000			
5	59.50000	0.000000	0.000000	184.
50000	45.00000			
6	63.83333	6.000000	4.166667	117.
83333	42.66667			

	B_avg_KD	B_avg_LEG_att	B_avg_LEG_landed	B_avg_PASS	B_avg_REV	B_avg
1	0.8000000	7.60000	5.40000	0.4000000	0.00000000	
65.4000						
2	0.0000000	12.30000	10.20000	0.8000000	0.00000000	
138.9000						

3	0.6451613	14.00000	12.19355	0.9354839	0.09677419
	97.0000				
4	0.5000000	3.00000	2.50000	0.5000000	0.25000000
	136.2500				
5	0.0000000	2.00000	2.00000	0.0000000	0.00000000
	203.5000				
6	0.0000000	22.83333	20.16667	1.3333333	0.16666667
	160.1667				

B_avg_SIG_STR_landed B_avg_SIG_STR_pct B_avg_SUB_ATT B_avg_TD_att B_avg_TD_landed B_avg_TD_pct

1	22.60000	0.466000	0.4000000	0.8000000
	0.2000000	0.1000000		
2	51.30000	0.399000	0.7000000	1.0000000
	0.5000000	0.2250000		
3	46.77419	0.496129	0.3548387	2.1612903
	0.6774194	0.2954839		
4	70.25000	0.550000	0.2500000	2.5000000
	1.2500000	0.2875000		
5	61.50000	0.310000	0.0000000	0.0000000
	0.0000000	0.0000000		
6	75.16667	0.470000	0.6666667	0.8333333
	0.3333333	0.2500000		

B_avg_TOTAL_STR_att B_avg_TOTAL_STR_landed B_longest_win_streak B_lo
sses B_avg_opp_BODY_att

1	66.4000	23.60000	4
1	6.40000		
2	158.7000	69.60000	3
6	13.00000		
3	103.7097	52.54839	8
8	17.90323		
4	154.7500	86.75000	4
0	12.25000		
5	204.0000	62.00000	1
1	42.50000		
6	183.5000	95.66667	4
2	12.00000		

B_avg_opp_BODY_landed B_avg_opp_CLINCH_att B_avg_opp_CLINCH_landed B_avg_opp_DISTANCE_att

1	4.000000	1.000000	0.60000
	51.20000		
2	9.300000	12.800000	9.60000
	101.70000		
3	11.870968	8.419355	5.83871
	84.54839		

4	6.000000	6.000000	3.75000
94.25000			

5	23.500000	0.500000	0.50000
205.00000			

6	7.333333	9.666667	7.00000
95.16667			

B_avg_opp_DISTANCE_landed	B_avg_opp_GROUND_att	B_avg_opp_GROUND_landed	B_avg_opp_HEAD_att
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1	17.40000	0.600000	0.20000
00	39.60000		

2	32.00000	8.100000	6.90000
00	97.70000		

3	38.06452	1.741935	0.93548
39	67.64516		

4	26.75000	1.750000	1.25000
00	82.50000		

5	89.50000	0.000000	0.00000
00	152.50000		

6	38.33333	5.166667	3.50000
00	86.66667		

B_avg_opp_HEAD_landed	B_avg_opp_KD	B_avg_opp_LEG_att	B_avg_opp_LEG_landed	B_avg_opp_PASS
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1	9.40000	0.2000000	6.80000	4.8
00000	0.00000000			

2	30.80000	0.1000000	11.90000	8.4
00000	1.40000000			

3	25.48387	0.2258065	9.16129	7.4
83871	0.03225806			

4	21.50000	0.2500000	7.25000	4.2
50000	0.00000000			

5	56.50000	0.0000000	10.50000	10.0
00000	0.00000000			

6	33.16667	0.0000000	11.33333	8.3
33333	1.50000000			

B_avg_opp_REV	B_avg_opp_SIG_STR_att	B_avg_opp_SIG_STR_landed	B_avg_opp_SIG_STR_pct
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1	0.00000000	52.80000	18.20000
0.2360000			

2	0.00000000	122.60000	48.50000
0.4080000			

3	0.03225806	94.70968	44.83871
0.4532258			

4	0.00000000	102.00000	31.75000
0.3375000			

5	0.00000000	205.50000	90.00000
	0.4300000		
6	0.16666667	110.00000	48.83333
	0.4266667		
	B_avg_opp_SUB_ATT	B_avg_opp_TD_att	B_avg_opp_TD_landed
	pct B_avg_opp_TOTAL_STR_att		
1	0.00000000	1.000000	0.4000000
	000	53.8000	0.10000
2	0.70000000	2.300000	0.9000000
	000	151.5000	0.23100
3	0.09677419	2.096774	0.2258065
	839	100.3871	0.06354
4	0.00000000	4.500000	0.7500000
	000	104.7500	0.09750
5	0.00000000	0.500000	0.0000000
	000	205.5000	0.00000
6	0.00000000	6.000000	1.1666667
	000	131.5000	0.14000
	B_avg_opp_TOTAL_STR_landed	B_total_rounds_fought	B_total_time_fought
	.seconds. B_total_title_bouts		
1	19.20000	9	
	419.400	0	
2	75.40000	29	
	849.000	0	
3	49.77419	68	
	581.871	1	
4	34.25000	9	
	652.000	0	
5	90.00000	8	
	1200.000	0	
6	68.66667	18	
	886.500	0	
	B_win_by_Decision_Majority	B_win_by_Decision_Split	B_win_by_Decision
	_Unanimous B_win_by_KO.TKO		
1	0	1	
	0	2	
2	0	2	
	1	0	
3	0	0	
	7	10	
4	0	0	
	2	2	
5	0	0	
	1	0	

6	0	0	
3	0		
B_win_by_Submission	B_win_by_TKO_Doctor_Stoppage	B_wins	B_Stance B_H
eight_cms B_Reach_cms			
1	1	0	4 Orthodox
167.64	170.18		
2	0	1	4 Orthodox
167.64	167.64		
3	6	0	23 Orthodox
185.42	185.42		
4	0	0	4 Switch
170.18	170.18		
5	0	0	1 Southpaw
180.34	185.42		
6	1	0	4 Orthodox
165.10	162.56		
B_Weight_lbs	R_current_lose_streak	R_current_win_streak	R_draw R_avg
_BODY_att R_avg_BODY_landed			
1	135	0	4 0
21.90000	16.400000		
2	125	0	2 0
12.00000	7.714286		
3	155	0	11 0
13.86667	8.666667		
4	135	1	0 0
18.25000	10.250000		
5	250	1	0 0
7.75000	6.750000		
6	115	0	4 0
8.75000	7.500000		
R_avg_CLINCH_att	R_avg_CLINCH_landed	R_avg_DISTANCE_att	R_avg_DISTAN
CE_landed R_avg_GROUND_att			
1	17.000000	11.000000	75.00000
26.50000	9.400000		
2	9.285714	6.857143	88.14286
36.14286	18.428571		
3	2.866667	1.733333	116.13333
49.46667	5.333333		
4	5.875000	4.125000	104.87500
41.00000	1.000000		
5	11.000000	7.250000	50.75000
24.75000	0.500000		
6	3.000000	2.250000	12.75000
4.75000	42.250000		

R_avg_GROUND_landed	R_avg_HEAD_att	R_avg_HEAD_landed	R_avg_KD	R_avg_LEG_att	R_avg_LEG_landed
1	6.500000	74.20000	23.90	0.400	
5.30000	3.70000				
2	16.428571	84.57143	37.00	0.000	1
9.28571	14.71429				
3	4.266667	96.73333	35.60	0.200	1
3.73333	11.20000				
4	0.625000	80.50000	24.00	0.375	1
3.00000	11.50000				
5	0.500000	50.75000	22.75	0.500	
3.75000	3.00000				
6	35.750000	44.75000	31.25	0.000	
4.50000	4.00000				

R_avg_PASS	R_avg_REV	R_avg_SIG_STR_att	R_avg_SIG_STR_landed	R_avg_SIG_STR_pct	R_avg_SUB_ATT
1	1.2000000	0.0000000	101.4000	44.00000	
0.4660000	0.1000000				
2	1.7142857	0.1428571	115.8571	59.42857	
0.5757143	0.4285714				
3	0.3333333	0.1333333	124.3333	55.46667	
0.4300000	1.0000000				
4	0.1250000	0.0000000	111.7500	45.75000	
0.3662500	0.0000000				
5	0.2500000	0.0000000	62.2500	32.50000	
0.5450000	0.0000000				
6	7.7500000	0.0000000	58.0000	42.75000	
0.6375000	0.5000000				

R_avg_TD_att	R_avg_TD_landed	R_avg_TD_pct	R_avg_TOTAL_STR_att	R_avg_TOTAL_STR_landed
1	5.3000000	1.900000	0.4580000	129.9000
69.1000				
2	5.1428571	2.428571	0.6014286	161.5714
102.8571				
3	0.9333333	0.400000	0.2773333	133.0000
63.4000				
4	2.2500000	0.625000	0.1037500	117.3750
50.7500				
5	0.5000000	0.000000	0.0000000	63.5000
32.7500				
6	5.5000000	4.500000	0.8175000	101.5000
80.5000				

R_longest_win_streak	R_losses	R_avg_opp_BODY_att	R_avg_opp_BODY_landed	R_avg_opp_CLINCH_att
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1	4	2	13.30000	8.8000
00	7.50000			
2	2	2	24.57143	14.1428
57	10.57143			
3	11	1	14.46667	8.1333
33	2.80000			
4	5	2	20.25000	13.3750
00	6.87500			
5	3	1	6.25000	4.7500
00	4.50000			
6	4	0	3.00000	2.2500
00	3.50000			

R_avg_opp_CLINCH_landed R_avg_opp_DISTANCE_att R_avg_opp_DISTANCE_la
nded R_avg_opp_GROUND_att

1	5.100000	90.50000	26.8
0000	0.800000		
2	7.8571429	98.57143	32.5
7143	6.428571		
3	0.7333333	91.06667	32.2
0000	4.866667		
4	5.6250000	103.12500	38.5
0000	0.875000		
5	3.5000000	42.75000	16.2
5000	7.750000		
6	3.0000000	5.75000	2.0
0000	2.000000		

R_avg_opp_GROUND_landed R_avg_opp_HEAD_att R_avg_opp_HEAD_landed R_a
vg_opp_KD R_avg_opp_LEG_att

1	0.300000	76.10000	17.30000
0.1000000	9.40000		
2	4.285714	61.85714	12.42857
0.0000000	29.14286		
3	2.800000	78.26667	23.20000
0.2666667	6.00000		
4	0.750000	77.37500	20.37500
0.1250000	13.25000		
5	2.750000	43.25000	14.00000
0.2500000	5.50000		
6	1.500000	8.00000	4.00000
0.0000000	0.25000		

R_avg_opp_LEG_landed R_avg_opp_PASS R_avg_opp_REV R_avg_opp_SIG_STR_
att R_avg_opp_SIG_STR_landed

1	6.10000	0.0000000	0.0000000	98.80
000	32.20000			

2	18.14286	1.1428571	0.0000000	115.57
143	44.71429			
3	4.40000	0.3333333	0.1333333	98.73
333	35.73333			
4	11.12500	0.0000000	0.0000000	110.87
500	44.87500			
5	3.75000	0.7500000	0.0000000	55.00
000	22.50000			
6	0.25000	0.0000000	0.5000000	11.25
000	6.50000			

	R_avg_opp_SIG_STR_pct	R_avg_opp_SUB_ATT	R_avg_opp_TD_att	R_avg_opp_T D_landed R_avg_opp_TD_pct
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1	0.3360000	0.00000000	0.900000	0
.1000000	0.0500000			
2	0.4371429	0.28571429	3.285714	0
.8571429	0.1471429			
3	0.3400000	0.06666667	2.866667	0
.6666667	0.1313333			
4	0.4462500	0.00000000	2.375000	0
.0000000	0.0000000			
5	0.3975000	0.00000000	1.000000	0
.0000000	0.0000000			
6	0.5400000	0.75000000	0.500000	0
.0000000	0.0000000			

	R_avg_opp_TOTAL_STR_att	R_avg_opp_TOTAL_STR_landed	R_total_rounds_fo ught
--	-------------------------	----------------------------	---------------------------

1	110.5000	43.30000
27		
2	158.1429	82.28571
25		
3	102.1333	38.60000
33		
4	115.1250	48.87500
20		
5	60.5000	27.75000
7		
6	38.0000	26.50000
8		

	R_total_time_fought.seconds.	R_total_title_bouts	R_win_by_Decision_M ajority
--	------------------------------	---------------------	--------------------------------

1	742.60	3
0		
2	1062.00	2
0		


```

3          604.40          2
0
4          690.25          0
0
5          440.75          0
0
6          540.00          1
0
  R_win_by_Decision_Split R_win_by_Decision_Unanimous R_win_by_KO.TKO
R_win_by_Submission
1          2          4          2
0
2          1          2          0
2
3          1          3          3
6
4          1          4          1
0
5          0          1          2
0
6          0          1          1
2
  R_win_by_TKO_Doctor_Stoppage R_wins R_Stance R_Height_cms R_Reach_cm
s R_Weight_lbs B_age R_age
1          0          8 Orthodox          162.56          162.5
6          135          31          32          0          5 Southpaw          165.10          167.6
2
4          125          32          31          0          14 Orthodox          180.34          193.0
3
4          155          36          35          0          6 Orthodox          162.56          172.7
4
2          135          26          29          0          3 Southpaw          187.96          190.5
5
0          264          32          26          0          4          165.10          167.6
6
4          115          33          28
>
> #Extracting relevant information of data needed for analysis
> #For this dataset we will use the following columns extracted.
> #To extract certain columns, we will use the subset function.
> #Within the subset function, we need to specify the name of our data
matrix (i.e. df)
> #and the columns we want to select (i.e. 6 to 100)
> #We can also make use of column names to extract needed columns

```

```
> #Notice that the data is extracted in our order of preference
> final_df <- subset(df, select = c(4, 6:12, 38, 63, 64, 65, 72, 73:76
, 144, 77:79 , 105, 130:132, 139:143, 145 ))
```

```
>
```

```
> #Display dataframe
```

```
> final_df
```

	date	Winner	title_bout	weight_class	no_of_rounds	B_c
urrent_lose_streak						
1	2019-06-08	Red	True	Bantamweight	5	
0						
2	2019-06-08	Red	True	Women's Flyweight	5	
0						
3	2019-06-08	Red	False	Lightweight	3	
0						
4	2019-06-08	Blue	False	Bantamweight	3	
0						
5	2019-06-08	Blue	False	Heavyweight	3	
0						
6	2019-06-08	Red	False	Women's Strawweight	3	
0						
7	2019-06-08	Red	False	Bantamweight	3	
0						
8	2019-06-08	Blue	False	Women's Strawweight	3	
1						
9	2019-06-08	Blue	False	Featherweight	3	
0						
10	2019-06-08	Red	False	Women's Strawweight	3	
0						
11	2019-06-08	Blue	False	Middleweight	3	
1						
12	2019-06-08	Red	False	Bantamweight	3	
0						
13	2019-06-08	Red	False	Women's Flyweight	3	
0						
14	2019-06-01	Blue	False	Light Heavyweight	5	
1						
15	2019-06-01	Blue	False	Light Heavyweight	3	
0						
16	2019-06-01	Red	False	Featherweight	3	
0						
17	2019-06-01	Blue	False	Lightweight	3	
0						
18	2019-06-01	Red	False	Featherweight	3	
0						

19 0	2019-06-01	Blue	False	Welterweight	3
20 1	2019-06-01	Blue	False	Women's Bantamweight	3
21 0	2019-06-01	Blue	False	Lightweight	3
22 2	2019-06-01	Blue	False	Lightweight	3
23 0	2019-06-01	Red	False	Women's Bantamweight	3
24 1	2019-06-01	Blue	False	Light Heavyweight	3
25 0	2019-06-01	Red	False	Lightweight	3
26 1	2019-05-18	Red	False	Welterweight	5
27 0	2019-05-18	Blue	False	Middleweight	3
28 0	2019-05-18	Blue	False	Women's Featherweight	3
29 0	2019-05-18	Red	False	Welterweight	3
30 0	2019-05-18	Red	False	Lightweight	3
31 0	2019-05-18	Red	False	Lightweight	3
32 0	2019-05-18	Red	False	Women's Bantamweight	3

B_current_win_streak B_draw B_losses B_total_rounds_fought B_total_time_fought.seconds.

1 419.4000	4	0	1	9
2 849.0000	3	0	6	29
3 581.8710	3	0	8	68
4 652.0000	4	0	0	9
5 1200.0000	1	0	1	8
6 886.5000	4	0	2	18
7 495.2500	3	0	4	23

8	0	0	2	10
716.0000				
9	1	0	1	10
670.7500				
10	1	0	6	26
763.1000				
11	0	0	5	14
488.7143				
12	0	0	0	0
NA				
13	2	0	3	18
628.6250				
14	0	0	4	27
625.0000				
15	3	0	0	7
681.6667				
16	1	0	1	3
310.5000				
17	1	0	3	10
558.2000				
18	0	0	0	0
NA				
19	0	0	0	0
NA				
20	0	0	3	13
720.4000				
21	4	0	1	14
656.3333				
22	0	0	2	8
727.6667				
23	0	0	0	0
NA				
24	0	0	3	13
620.6667				
25	0	0	0	0
NA				
26	0	0	4	36
724.8571				
27	1	0	0	3
900.0000				
28	0	0	0	0
NA				
29	0	0	0	0
NA				

30		2	0	8		59
754.9091						
31		0	0	0		0
NA						
32		2	0	0		6
900.0000						
B_total_title_bouts		B_wins	B_Stance	B_Height_cms	B_Reach_cms	B_Weig
ht_lbs	B_age					
1		0	4 Orthodox	167.64		170.18
135	31					
2		0	4 Orthodox	167.64		167.64
125	32					
3		1	23 Orthodox	185.42		185.42
155	36					
4		0	4 Switch	170.18		170.18
135	26					
5		0	1 Southpaw	180.34		185.42
250	32					
6		0	4 Orthodox	165.10		162.56
115	33					
7		0	8 Orthodox	167.64		165.10
135	32					
8		0	2 Orthodox	165.10		167.64
115	25					
9		0	3 Orthodox	180.34		182.88
145	31					
10		0	4 Orthodox	160.02		162.56
115	34					
11		0	2 Orthodox	182.88		187.96
185	28					
12		0	0 Switch	170.18		172.72
135	35					
13		0	5 Orthodox	167.64		165.10
125	33					
14		1	7 Orthodox	193.04		193.04
205	30					
15		0	3 Orthodox	193.04		198.12
205	27					
16		0	1 Orthodox	172.72		172.72
145	26					
17		0	2 Orthodox	177.80		180.34
155	29					
18		0	0 Orthodox	182.88		182.88
145	26					

19		0	0 Orthodox	185.42	187.96
170	27				
20		0	2 Orthodox	170.18	165.10
135	37				
21		1	5 Orthodox	182.88	190.50
155	39				
22		0	1 Orthodox	177.80	185.42
170	30				
23		0	0 Orthodox	170.18	185.42
145	23				
24		0	3 Orthodox	182.88	190.50
205	29				
25		0	0 Southpaw	182.88	187.96
155	24				
26		1	10 Orthodox	175.26	195.58
170	26				
27		0	1 Orthodox	180.34	182.88
185	30				
28		0	0 Orthodox	167.64	172.72
145	28				
29		0	0 Orthodox	177.80	182.88
170	31				
30		0	14 Orthodox	172.72	172.72
155	34				
31		0	0 Orthodox	177.80	180.34
155	27				
32		0	2	162.56	170.18
135	34				

	R_current_lose_streak	R_current_win_streak	R_draw	R_losses	R_total_
rounds_fought					
1	0		4	0	2
27					
2	0		2	0	2
25					
3	0		11	0	1
33					
4	1		0	0	2
20					
5	1		0	0	1
7					
6	0		4	0	0
8					
7	0		3	0	3
32					

8	2	0	0	4
25				
9	0	1	0	5
34				
10	0	3	0	0
9				
11	1	0	0	1
3				
12	2	0	0	7
29				
13	1	0	0	2
18				
14	1	0	0	5
38				
15	3	0	0	5
21				
16	0	1	0	1
11				
17	0	2	0	2
12				
18	3	0	0	3
8				
19	0	0	0	0
0				
20	2	0	0	2
4				
21	0	1	0	3
22				
22	2	0	0	3
19				
23	0	0	0	0
0				
24	0	1	0	0
1				
25	1	0	0	1
3				
26	2	0	0	9
74				
27	0	4	0	3
21				
28	0	1	0	1
4				
29	0	4	0	2
18				

30	0	4	0	9
42				
31	0	3	0	1
8				
32	0	2	0	0
3				
R_total_time_fought.seconds. R_total_title_bouts R_wins R_Stance R_				
Height_cms R_Reach_cms				
1	742.6000	3	8	Orthodox
162.56	162.56			
2	1062.0000	2	5	Southpaw
165.10	167.64			
3	604.4000	2	14	Orthodox
180.34	193.04			
4	690.2500	0	6	Orthodox
162.56	172.72			
5	440.7500	0	3	Southpaw
187.96	190.50			
6	540.0000	1	4	
165.10	167.64			
7	750.6667	0	9	Orthodox
170.18	180.34			
8	800.1111	1	5	Orthodox
160.02	162.56			
9	624.0667	1	10	Orthodox
172.72	180.34			
10	900.0000	0	3	Orthodox
165.10	160.02			
11	692.0000	0	0	Orthodox
190.50	200.66			
12	668.0000	1	5	Orthodox
170.18	175.26			
13	900.0000	0	4	Orthodox
175.26	172.72			
14	660.8667	3	10	Orthodox
195.58	200.66			
15	455.7273	0	6	Orthodox
185.42	200.66			
16	561.8000	0	4	Southpaw
177.80	182.88			
17	610.8000	0	3	Orthodox
175.26	177.80			
18	681.6667	0	0	Orthodox
165.10	175.26			

19		NA	0	0	Switch
177.80	182.88				
20		461.0000	1	0	Orthodox
170.18	177.80				
21		686.8889	0	6	Southpaw
177.80	177.80				
22		807.8571	0	4	Southpaw
167.64	167.64				
23		NA	0	0	Orthodox
175.26	187.96				
24		193.0000	0	1	Orthodox
182.88	193.04				
25		900.0000	0	0	Orthodox
190.50	195.58				
26		794.7692	4	17	Southpaw
172.72	177.80				
27		634.2222	1	6	Orthodox
187.96	200.66				
28		480.5000	0	1	Orthodox
182.88	182.88				
29		453.4000	0	8	Orthodox
180.34	190.50				
30		400.5652	0	14	Orthodox
177.80	187.96				
31		480.5000	0	3	Orthodox
167.64	177.80				
32		329.5000	0	2	Orthodox
167.64	167.64				

	R_Weight_lbs	R_age
1	135	32
2	125	31
3	155	35
4	135	29
5	264	26
6	115	28
7	135	29
8	115	33
9	145	37
10	115	29
11	185	28
12	135	34
13	125	30
14	205	32
15	205	39

```

16      145      30
17      155      32
18      145      31
19      170      27
20      135      37
21      155      29
22      155      35
23      135      27
24      205      27
25      155      25
26      170      34
27      185      29
28      145      29
29      170      27
30      155      29
31      155      32
32      135      24

```

```

[ reached 'max' / getOption("max.print") -- omitted 5112 rows ]
>
> #Save this data frame for future lab assignments using write.csv com
mands
> #To save a dataframe as CSV is easy.
> #Simply need to use the write.csv function with the name of the data
frame and the name of the file you want to write to
> write.csv(final_df, file = "df.csv")
>
> #Structure of dataframe
> str(final_df)
'data.frame': 5144 obs. of  31 variables:
 $ date          : chr  "2019-06-08" "2019-06-08" "2019-
06-08" "2019-06-08" ...
 $ Winner        : chr  "Red" "Red" "Red" "Blue" ...
 $ title_bout    : chr  "True" "True" "False" "False" ..
 .
 $ weight_class  : chr  "Bantamweight" "Women's Flyweigh
t" "Lightweight" "Bantamweight" ...
 $ no_of_rounds  : int   5 5 3 3 3 3 3 3 3 3 ...
 $ B_current_lose_streak : num  0 0 0 0 0 0 0 1 0 0 ...
 $ B_current_win_streak  : num  4 3 3 4 1 4 3 0 1 1 ...
 $ B_draw        : num  0 0 0 0 0 0 0 0 0 0 ...
 $ B_losses      : num  1 6 8 0 1 2 4 2 1 6 ...
 $ B_total_rounds_fought : num  9 29 68 9 8 18 23 10 10 26 ...
 $ B_total_time_fought.seconds.: num  419 849 582 652 1200 ...
 $ B_total_title_bouts   : num  0 0 1 0 0 0 0 0 0 0 ...

```

```

$ B_wins          : num  4 4 23 4 1 4 8 2 3 4 ...
$ B_Stance        : chr   "Orthodox" "Orthodox" "Orthodox"
"Switch" ...
$ B_Height_cms    : num  168 168 185 170 180 ...
$ B_Reach_cms     : num  170 168 185 170 185 ...
$ B_Weight_lbs    : num  135 125 155 135 250 115 135 115
145 115 ...
$ B_age           : num  31 32 36 26 32 33 32 25 31 34 ..
.
$ R_current_lose_streak : num  0 0 0 1 1 0 0 2 0 0 ...
$ R_current_win_streak  : num  4 2 11 0 0 4 3 0 1 3 ...
$ R_draw                : num  0 0 0 0 0 0 0 0 0 0 ...
$ R_losses              : num  2 2 1 2 1 0 3 4 5 0 ...
$ R_total_rounds_fought : num  27 25 33 20 7 8 32 25 34 9 ...
$ R_total_time_fought.seconds.: num  743 1062 604 690 441 ...
$ R_total_title_bouts   : num  3 2 2 0 0 1 0 1 1 0 ...
$ R_wins                : num  8 5 14 6 3 4 9 5 10 3 ...
$ R_Stance              : chr   "Orthodox" "Southpaw" "Orthodox"
"Orthodox" ...
$ R_Height_cms         : num  163 165 180 163 188 ...
$ R_Reach_cms          : num  163 168 193 173 190 ...
$ R_Weight_lbs         : num  135 125 155 135 264 115 135 115
145 115 ...
$ R_age                : num  32 31 35 29 26 28 29 33 37 29 ..
.

```

```

>
> #Summary of entire dataframe
> print(summary(final_df))
      date      Winner      title_bout      weight_class
no_of_rounds
Length:5144      Length:5144      Length:5144      Length:5144
Min.   :1.000
Class  :character Class :character Class :character Class :chara
cter   1st Qu.:3.000
Mode   :character Mode  :character Mode  :character Mode  :chara
cter   Median :3.000

Mean    :3.119

3rd Qu.:3.000

Max.    :5.000

```

B_current_lose_streak	B_current_win_streak	B_draw	B_losses
B_total_rounds_fought			
Min. :0.0000	Min. : 0.0000	Min. :0	Min. : 0.000
Min. : 0.000			
1st Qu.:0.0000	1st Qu.: 0.0000	1st Qu.:0	1st Qu.: 0.000
1st Qu.: 1.000			
Median :0.0000	Median : 0.0000	Median :0	Median : 1.000
Median : 5.000			
Mean :0.4298	Mean : 0.8373	Mean :0	Mean : 1.464
Mean : 8.921			
3rd Qu.:1.0000	3rd Qu.: 1.0000	3rd Qu.:0	3rd Qu.: 2.000
3rd Qu.:13.000			
Max. :6.0000	Max. :13.0000	Max. :0	Max. :13.000
Max. :75.000			

B_total_time_fought.seconds.	B_total_title_bouts	B_wins	B
_Stance			
Min. : 7.0	Min. : 0.0000	Min. : 0.000	Len
gth:5144			
1st Qu.: 445.9	1st Qu.: 0.0000	1st Qu.: 0.000	Clas
ss :character			
Median : 610.4	Median : 0.0000	Median : 1.000	Mod
e :character			
Mean : 592.4	Mean : 0.2799	Mean : 2.484	
3rd Qu.: 767.2	3rd Qu.: 0.0000	3rd Qu.: 4.000	
Max. :1500.0	Max. :16.0000	Max. :23.000	
NA's :1265			

B_Height_cms	B_Reach_cms	B_Weight_lbs	B_age	R_cur
rent_lose_streak				
Min. :152.4	Min. :152.4	Min. :115.0	Min. :18.00	Min.
:0.0000				
1st Qu.:172.7	1st Qu.:177.8	1st Qu.:145.0	1st Qu.:26.00	1st Q
u.:0.0000				
Median :180.3	Median :182.9	Median :170.0	Median :29.00	Media
n :0.0000				
Mean :179.2	Mean :183.3	Mean :172.1	Mean :29.17	Mean
:0.5509				
3rd Qu.:185.4	3rd Qu.:190.5	3rd Qu.:185.0	3rd Qu.:32.00	3rd Q
u.:1.0000				
Max. :210.8	Max. :213.4	Max. :770.0	Max. :51.00	Max.
:7.0000				
NA's :8	NA's :666	NA's :6	NA's :172	
R_current_win_streak	R_draw	R_losses	R_total_rounds_foug	ht

Min. : 0	Min. :0	Min. : 0.000	Min. : 0.00
1st Qu.: 0	1st Qu.:0	1st Qu.: 0.000	1st Qu.: 3.00
Median : 0	Median :0	Median : 1.000	Median : 9.00
Mean : 1	Mean :0	Mean : 1.951	Mean :12.85
3rd Qu.: 1	3rd Qu.:0	3rd Qu.: 3.000	3rd Qu.:19.00
Max. :16	Max. :0	Max. :14.000	Max. :80.00

R_total_time_fought.seconds.	R_total_title_bouts	R_wins	R
_Stance			
Min. : 7.0	Min. : 0.000	Min. : 0.000	Len
1st Qu.: 470.6	1st Qu.: 0.000	1st Qu.: 1.000	Cla
Median : 620.3	Median : 0.000	Median : 2.000	Mod
Mean : 603.8	Mean : 0.597	Mean : 3.598	
3rd Qu.: 762.3	3rd Qu.: 1.000	3rd Qu.: 5.000	
Max. :1500.0	Max. :16.000	Max. :20.000	
NA's :650			
R_Height_cms	R_Reach_cms	R_Weight_lbs	R_age
Min. :152.4	Min. :152.4	Min. :115.0	Min. :19.00
1st Qu.:172.7	1st Qu.:177.8	1st Qu.:145.0	1st Qu.:26.00
Median :180.3	Median :182.9	Median :170.0	Median :29.00
Mean :179.3	Mean :183.7	Mean :172.1	Mean :29.44
3rd Qu.:185.4	3rd Qu.:190.5	3rd Qu.:185.0	3rd Qu.:32.00
Max. :210.8	Max. :213.4	Max. :345.0	Max. :47.00
NA's :4	NA's :316	NA's :3	NA's :64

Conclusion:

Thus dataset has been selected from kaggle loaded into Rstudio.

Lab Outcome: LO6

Lab Assignment: 08

Aim:

Cleaning of the Dataset

Code:

```
> #Lab Assignment 8
> #Mini-Project Session 2
>
> #Cleaning of Dataset
>
> #get the working directory
> getwd()
[1] "/cloud/project"
>
> #Loading dataset into R
> #Header is set to to true if file contains Header information
> #Sep stands for seperator which is , in our csv file
> df <- read.csv("data.csv", header = TRUE, sep = ",")
>
> #Check the class before doing any cleaning
> class(df)
[1] "data.frame"
>
> #Check the number of rows and columns the data frame has
> dim(df)
[1] 5144  145
> #We can see that the data frame has 5144 rows and 32 columns
>
> #Finding summary of entire dataset
> summary(df)
  R_fighter      B_fighter      Referee      date
location
Length:5144    Length:5144    Length:5144    Length:5144
Length:5144
Class :character Class :character Class :character Class :chara
cter  Class :character
Mode  :character  Mode  :character  Mode  :character  Mode  :chara
cter  Mode  :character
```

Winner	title_bout	weight_class	no_of_round
s B_current_lose_streak			
Length:5144	Length:5144	Length:5144	Min. :1.00
0 Min. :0.0000			
Class :character	Class :character	Class :character	1st Qu.:3.00
0 1st Qu.:0.0000			
Mode :character	Mode :character	Mode :character	Median :3.00
0 Median :0.0000			
			Mean :3.11
9 Mean :0.4298			
			3rd Qu.:3.00
0 3rd Qu.:1.0000			
			Max. :5.00
0 Max. :6.0000			
B_current_win_streak	B_draw	B_avg_BODY_att	B_avg_BODY_landed B
_avg_CLINCH_att			
Min. : 0.0000	Min. :0	Min. : 0.000	Min. : 0.000 M
in. : 0.000			
1st Qu.: 0.0000	1st Qu.:0	1st Qu.: 3.500	1st Qu.: 2.333 1
st Qu.: 3.000			
Median : 0.0000	Median :0	Median : 7.000	Median : 5.000 M
edian : 6.333			
Mean : 0.8373	Mean :0	Mean : 8.689	Mean : 6.083 M
ean : 8.241			
3rd Qu.: 1.0000	3rd Qu.:0	3rd Qu.:12.225	3rd Qu.: 8.500 3
rd Qu.:11.422			
Max. :13.0000	Max. :0	Max. :49.000	Max. :39.000 M
ax. :87.000			
B_avg_CLINCH_landed	B_avg_DISTANCE_att	B_avg_DISTANCE_landed	B_avg_GR
OUND_att B_avg_GROUND_landed			
Min. : 0.000	Min. : 0.00	Min. : 0.000	Min. :
0.000 Min. : 0.000			
1st Qu.: 2.000	1st Qu.: 22.00	1st Qu.: 7.667	1st Qu.:
2.500 1st Qu.: 1.667			
Median : 4.200	Median : 44.67	Median : 15.200	Median :
6.500 Median : 4.333			
Mean : 5.556	Mean : 53.16	Mean : 19.329	Mean :
8.754 Mean : 5.773			
3rd Qu.: 7.739	3rd Qu.: 74.33	3rd Qu.: 27.143	3rd Qu.:
12.167 3rd Qu.: 8.000			
Max. :68.000	Max. :271.00	Max. :130.000	Max. :
88.000 Max. :47.000			

B_avg_HEAD_att	B_avg_HEAD_landed	B_avg_KD	B_avg_LEG_att	
B_avg_LEG_landed				
Min. : 0.00	Min. : 0.00	Min. :0.0000	Min. : 0.000	
Min. : 0.000				
1st Qu.: 29.41	1st Qu.: 11.00	1st Qu.:0.0000	1st Qu.: 2.000	
1st Qu.: 1.500				
Median : 49.00	Median : 17.75	Median :0.1111	Median : 4.500	
Median : 3.600				
Mean : 55.45	Mean : 19.82	Mean :0.2464	Mean : 6.009	
Mean : 4.759				
3rd Qu.: 74.69	3rd Qu.: 26.00	3rd Qu.:0.4000	3rd Qu.: 8.500	
3rd Qu.: 6.750				
Max. :277.00	Max. :137.00	Max. :5.0000	Max. :61.000	
Max. :47.000				
B_avg_PASS	B_avg_REV	B_avg_SIG_STR_att	B_avg_SIG_STR_lan	
ded B_avg_SIG_STR_pct				
Min. : 0.000	Min. :0.0000	Min. : 0.00	Min. : 0.00	
Min. :0.0000				
1st Qu.: 0.200	1st Qu.:0.0000	1st Qu.: 38.50	1st Qu.: 17.25	
1st Qu.:0.3875				
Median : 1.000	Median :0.0000	Median : 63.00	Median : 28.26	
Median :0.4550				
Mean : 1.278	Mean :0.1692	Mean : 70.15	Mean : 30.66	
Mean :0.4563				
3rd Qu.: 1.879	3rd Qu.:0.2500	3rd Qu.: 94.53	3rd Qu.: 40.83	
3rd Qu.:0.5240				
Max. :15.000	Max. :3.0000	Max. :299.00	Max. :154.00	
Max. :1.0000				
B_avg_SUB_ATT	B_avg_TD_att	B_avg_TD_landed	B_avg_TD_pct	B
_avg_TOTAL_STR_att				
Min. :0.0000	Min. : 0.0000	Min. : 0.00	Min. :0.0000	M
in. : 0.00				
1st Qu.:0.0000	1st Qu.: 0.8819	1st Qu.: 0.25	1st Qu.:0.1000	1
st Qu.: 57.78				
Median :0.3333	Median : 2.0000	Median : 1.00	Median :0.3000	M
edian : 87.50				
Mean :0.5481	Mean : 2.8049	Mean : 1.19	Mean :0.3155	M
ean : 92.64				
3rd Qu.:0.9000	3rd Qu.: 4.0000	3rd Qu.: 1.75	3rd Qu.:0.4861	3
rd Qu.:122.65				
Max. :8.0000	Max. :19.0000	Max. :10.00	Max. :1.0000	M
ax. :360.00				
B_avg_TOTAL_STR_landed	B_longest_win_streak	B_losses	B_avg_op	
p_BODY_att				

Min. : 0.00 0.000	Min. : 0.000	Min. : 0.000	Min. :
1st Qu.: 31.00 3.358	1st Qu.: 0.000	1st Qu.: 0.000	1st Qu.:
Median : 47.67 7.000	Median : 1.000	Median : 1.000	Median :
Mean : 50.71 8.297	Mean : 1.588	Mean : 1.464	Mean :
3rd Qu.: 66.50 11.639	3rd Qu.: 2.000	3rd Qu.: 2.000	3rd Qu.:
Max. :230.00 61.000	Max. :16.000	Max. :13.000	Max. :
B_avg_opp_BODY_landed	B_avg_opp_CLINCH_att	B_avg_opp_CLINCH_landed	B_avg_opp_DISTANCE_att
Min. : 0.000 n. : 0.00	Min. : 0.000	Min. : 0.000	Min. : 0.00
1st Qu.: 2.333	1st Qu.: 2.857	1st Qu.: 1.667	1st Qu.: 2.333
Median : 4.600	Median : 5.909	Median : 3.667	Median : 4.600
Mean : 5.639	Mean : 7.455	Mean : 4.922	Mean : 5.639
3rd Qu.: 7.714	3rd Qu.: 10.000	3rd Qu.: 6.667	3rd Qu.: 7.714
Max. :48.000 x. :361.00	Max. :105.000	Max. :84.000	Max. :48.000
B_avg_opp_DISTANCE_landed	B_avg_opp_GROUND_att	B_avg_opp_GROUND_landed	B_avg_opp_HEAD_att
Min. : 0.00 Min. : 0.00	Min. : 0.000	Min. : 0.000	Min. : 0.00
1st Qu.: 7.00 1st Qu.: 27.00	1st Qu.: 1.667	1st Qu.: 1.000	1st Qu.: 7.00
Median : 14.67 Median : 46.57	Median : 4.714	Median : 3.000	Median : 14.67
Mean : 18.23 Mean : 52.09	Mean : 7.100	Mean : 4.633	Mean : 18.23
3rd Qu.: 25.32 3rd Qu.: 69.20	3rd Qu.: 9.454	3rd Qu.: 6.000	3rd Qu.: 25.32
Max. :150.00 Max. :335.00	Max. :94.000	Max. :84.000	Max. :150.00
B_avg_opp_HEAD_landed	B_avg_opp_KD	B_avg_opp_LEG_att	B_avg_opp_LEG_landed
Min. : 0.000 00	Min. :0.0000	Min. : 0.000	Min. : 0.000
Min. : 0.0000			

1st Qu.: 8.857	1st Qu.:0.0000	1st Qu.: 2.143	1st Qu.: 1.8
00 1st Qu.: 0.0000			
Median : 15.000	Median :0.0000	Median : 4.800	Median : 3.7
50 Median : 0.6667			
Mean : 17.368	Mean :0.1609	Mean : 6.048	Mean : 4.7
74 Mean : 1.0892			
3rd Qu.: 23.000	3rd Qu.:0.2500	3rd Qu.: 8.114	3rd Qu.: 6.5
64 3rd Qu.: 1.5000			
Max. :126.000	Max. :3.0000	Max. :57.000	Max. :50.0
00 Max. :19.0000			
B_avg_opp_REV	B_avg_opp_SIG_STR_att	B_avg_opp_SIG_STR_landed	B_avg
_opp_SIG_STR_pct			
Min. :0.0000	Min. : 0.00	Min. : 0.00	Min.
:0.0000			
1st Qu.:0.0000	1st Qu.: 35.35	1st Qu.: 15.00	1st Q
u.:0.3533			
Median :0.0000	Median : 60.25	Median : 25.07	Media
n :0.4157			
Mean :0.1536	Mean : 66.43	Mean : 27.78	Mean
:0.4272			
3rd Qu.:0.2000	3rd Qu.: 87.45	3rd Qu.: 36.55	3rd Q
u.:0.4946			
Max. :3.0000	Max. :401.00	Max. :202.00	Max.
:1.0000			
B_avg_opp_SUB_ATT	B_avg_opp_TD_att	B_avg_opp_TD_landed	B_avg_opp_TD_p
ct B_avg_opp_TOTAL_STR_att			
Min. :0.0000	Min. : 0.000	Min. : 0.0000	Min. :0.0000
Min. : 0.00			
1st Qu.:0.0000	1st Qu.: 1.000	1st Qu.: 0.2817	1st Qu.:0.0827
1st Qu.: 54.00			
Median :0.2500	Median : 2.364	Median : 0.8571	Median :0.2394
Median : 83.00			
Mean :0.4625	Mean : 2.901	Mean : 1.0941	Mean :0.2765
Mean : 86.98			
3rd Qu.:0.6667	3rd Qu.: 4.000	3rd Qu.: 1.5000	3rd Qu.:0.3928
3rd Qu.:113.63			
Max. :7.0000	Max. :20.000	Max. :11.5000	Max. :1.0000
Max. :404.00			
B_avg_opp_TOTAL_STR_landed	B_total_rounds_fought	B_total_time_fought.	
seconds. B_total_title_bouts			
Min. : 0.00	Min. : 0.000	Min. : 7.0	
Min. : 0.0000			
1st Qu.: 28.68	1st Qu.: 1.000	1st Qu.: 445.9	
1st Qu.: 0.0000			

Median : 43.33	Median : 5.000	Median : 610.4
Median : 0.0000		
Mean : 46.16	Mean : 8.921	Mean : 592.4
Mean : 0.2799		
3rd Qu.: 59.00	3rd Qu.:13.000	3rd Qu.: 767.2
3rd Qu.: 0.0000		
Max. :232.00	Max. :75.000	Max. :1500.0
Max. :16.0000		
B_win_by_Decision_Majority	B_win_by_Decision_Split	B_win_by_Decision_
Unanimous	B_win_by_KO.TKO	
Min. :0.00000	Min. :0.0000	Min. : 0.0000
Min. : 0.0000		
1st Qu.:0.00000	1st Qu.:0.0000	1st Qu.: 0.0000
1st Qu.: 0.0000		
Median :0.00000	Median :0.0000	Median : 0.0000
Median : 0.0000		
Mean :0.01691	Mean :0.2123	Mean : 0.7801
Mean : 0.8727		
3rd Qu.:0.00000	3rd Qu.:0.0000	3rd Qu.: 1.0000
3rd Qu.: 1.0000		
Max. :2.00000	Max. :5.0000	Max. :10.0000
Max. :11.0000		
B_win_by_Submission	B_win_by_TKO_Doctor_Stoppage	B_wins
_Stance		B
Min. : 0.0000	Min. :0.0000	Min. : 0.000
Min. : 0.0000		Len
1st Qu.: 0.0000	1st Qu.:0.0000	1st Qu.: 0.000
1st Qu.: 0.0000		Clas
Median : 0.0000	Median :0.0000	Median : 1.000
Median : 0.0000		Mod
Mean : 0.5515	Mean :0.0453	Mean : 2.484
Mean : 0.5515		
3rd Qu.: 1.0000	3rd Qu.:0.0000	3rd Qu.: 4.000
3rd Qu.: 1.0000		
Max. :11.0000	Max. :2.0000	Max. :23.000
Max. :11.0000		
B_Height_cms	B_Reach_cms	B_Weight_lbs
R_current_win_streak		R_current_lose_streak
Min. :152.4	Min. :152.4	Min. :115.0
Min. :152.4		Min. :0.0000
1st Qu.:172.7	1st Qu.:177.8	1st Qu.:145.0
1st Qu.:172.7		1st Qu.:0.0000
Median :180.3	Median :182.9	Median :170.0
Median :180.3		Median :0.0000
Mean :179.2	Mean :183.3	Mean :172.1
Mean :179.2		Mean :0.5509
Mean : 1		

3rd Qu.:185.4 3rd Qu.:190.5 3rd Qu.:185.0 3rd Qu.:1.0000
 3rd Qu.: 1
 Max. :210.8 Max. :213.4 Max. :770.0 Max. :7.0000
 Max. :16
 R_draw R_avg_BODY_att R_avg_BODY_landed R_avg_CLINCH_att R_avg
 _CLINCH_landed
 Min. :0 Min. : 0.000 Min. : 0.000 Min. : 0.000 Min.
 : 0.000
 1st Qu.:0 1st Qu.: 4.000 1st Qu.: 2.800 1st Qu.: 3.400 1st Q
 u.: 2.000
 Median :0 Median : 7.297 Median : 5.000 Median : 6.545 Media
 n : 4.333
 Mean :0 Mean : 8.811 Mean : 6.138 Mean : 8.174 Mean
 : 5.525
 3rd Qu.:0 3rd Qu.:12.000 3rd Qu.: 8.500 3rd Qu.:11.000 3rd Q
 u.: 7.444
 Max. :0 Max. :51.000 Max. :39.000 Max. :82.000 Max.
 :52.000
 R_avg_DISTANCE_att R_avg_DISTANCE_landed R_avg_GROUND_att R_avg_GROUN
 D_landed R_avg_HEAD_att
 Min. : 0.00 Min. : 0.000 Min. : 0.000 Min. : 0.
 000 Min. : 0.00
 1st Qu.: 24.14 1st Qu.: 8.437 1st Qu.: 3.333 1st Qu.: 2.
 146 1st Qu.: 32.67
 Median : 46.00 Median : 16.000 Median : 7.439 Median : 5.
 000 Median : 50.03
 Mean : 53.58 Mean : 19.551 Mean : 9.557 Mean : 6.
 325 Mean : 56.17
 3rd Qu.: 74.47 3rd Qu.: 27.851 3rd Qu.:13.000 3rd Qu.: 8.
 587 3rd Qu.: 75.00
 Max. :287.50 Max. :131.000 Max. :96.000 Max. :62.
 000 Max. :264.00
 R_avg_HEAD_landed R_avg_KD R_avg_LEG_att R_avg_LEG_landed
 R_avg_PASS
 Min. : 0.00 Min. :0.0000 Min. : 0.000 Min. : 0.000
 Min. : 0.0000
 1st Qu.: 12.17 1st Qu.:0.0000 1st Qu.: 2.000 1st Qu.: 1.667
 1st Qu.: 0.3333
 Median : 18.45 Median :0.1667 Median : 4.786 Median : 3.905
 Median : 1.0000
 Mean : 20.27 Mean :0.2577 Mean : 6.329 Mean : 4.997
 Mean : 1.4136
 3rd Qu.: 26.33 3rd Qu.:0.4000 3rd Qu.: 8.990 3rd Qu.: 7.000
 3rd Qu.: 2.0000

Max. :119.00	Max. :4.0000	Max. :63.000	Max. :44.000
Max. :14.0000			
R_avg_REV	R_avg_SIG_STR_att	R_avg_SIG_STR_landed	R_avg_SIG_STR
_pct R_avg_SUB_ATT			
Min. :0.0000	Min. : 0.00	Min. : 0.00	Min. :0.000
0 Min. :0.0000			
1st Qu.:0.0000	1st Qu.: 42.52	1st Qu.: 19.50	1st Qu.:0.393
3 1st Qu.:0.0000			
Median :0.0000	Median : 64.18	Median : 28.71	Median :0.457
0 Median :0.3750			
Mean :0.1533	Mean : 71.31	Mean : 31.40	Mean :0.463
8 Mean :0.5431			
3rd Qu.:0.2133	3rd Qu.: 95.25	3rd Qu.: 41.50	3rd Qu.:0.530
0 3rd Qu.:0.8451			
Max. :3.0000	Max. :298.50	Max. :141.00	Max. :1.000
0 Max. :9.0000			
R_avg_TD_att	R_avg_TD_landed	R_avg_TD_pct	R_avg_TOTAL_STR_at
t R_avg_TOTAL_STR_landed			
Min. : 0.000	Min. : 0.000	Min. :0.0000	Min. : 0.00
Min. : 0.00			
1st Qu.: 1.000	1st Qu.: 0.375	1st Qu.:0.1650	1st Qu.: 61.16
1st Qu.: 33.61			
Median : 2.333	Median : 1.000	Median :0.3221	Median : 89.79
Median : 48.52			
Mean : 3.027	Mean : 1.305	Mean :0.3336	Mean : 93.84
Mean : 51.52			
3rd Qu.: 4.333	3rd Qu.: 2.000	3rd Qu.:0.4900	3rd Qu.:121.00
3rd Qu.: 65.40			
Max. :30.000	Max. :11.000	Max. :1.0000	Max. :325.50
Max. :202.50			
R_longest_win_streak	R_losses	R_avg_opp_BODY_att	R_avg_opp_BO
DY_landed R_avg_opp_CLINCH_att			
Min. : 0.000	Min. : 0.000	Min. : 0.000	Min. : 0.0
00 Min. : 0.000			
1st Qu.: 1.000	1st Qu.: 0.000	1st Qu.: 3.600	1st Qu.: 2.5
00 1st Qu.: 3.000			
Median : 2.000	Median : 1.000	Median : 7.000	Median : 4.8
00 Median : 6.000			
Mean : 2.254	Mean : 1.951	Mean : 8.157	Mean : 5.4
99 Mean : 7.345			
3rd Qu.: 3.000	3rd Qu.: 3.000	3rd Qu.:11.495	3rd Qu.: 7.5
00 3rd Qu.: 9.744			
Max. :16.000	Max. :14.000	Max. :75.000	Max. :41.0
00 Max. :82.000			

R_avg_opp_CLINCH_landed R_avg_opp_DISTANCE_att R_avg_opp_DISTANCE_landed R_avg_opp_GROUND_att

Min. : 0.000	Min. : 0.00	Min. : 0.00
Min. : 0.000		
1st Qu.: 2.000	1st Qu.: 23.50	1st Qu.: 7.50
1st Qu.: 1.750		
Median : 3.833	Median : 45.00	Median : 14.81
Median : 4.750		
Mean : 4.813	Mean : 52.16	Mean : 17.89
Mean : 6.611		
3rd Qu.: 6.352	3rd Qu.: 73.00	3rd Qu.: 25.00
3rd Qu.: 9.000		
Max. : 51.000	Max. : 440.00	Max. : 144.00
Max. : 104.000		

R_avg_opp_GROUND_landed R_avg_opp_HEAD_att R_avg_opp_HEAD_landed R_avg_opp_KD R_avg_opp_LEG_att

Min. : 0.000	Min. : 0.00	Min. : 0.00	Min. : 0.000
: 0.000			
Min. : 0.000			
1st Qu.: 1.000	1st Qu.: 28.25	1st Qu.: 9.00	1st Qu.: 0.000
1st Qu.: 2.431			
Median : 3.000	Median : 46.67	Median : 14.83	Median : 0.000
Median : 4.931			
Mean : 4.273	Mean : 52.00	Mean : 16.86	Mean : 0.154
Mean : 5.958			
3rd Qu.: 5.893	3rd Qu.: 69.33	3rd Qu.: 22.48	3rd Qu.: 0.250
3rd Qu.: 8.000			
Max. : 53.000	Max. : 400.00	Max. : 132.00	Max. : 3.000
Max. : 63.000			

R_avg_opp_LEG_landed R_avg_opp_PASS R_avg_opp_REV R_avg_opp_SIG_STR_att

Min. : 0.000	Min. : 0.0000	Min. : 0.0000	Min. : 0.0000
1st Qu.: 2.000	1st Qu.: 0.1667	1st Qu.: 0.0000	1st Qu.: 36.79
Median : 3.800	Median : 0.6667	Median : 0.0000	Median : 60.53
Mean : 4.615	Mean : 1.0138	Mean : 0.1603	Mean : 66.12
3rd Qu.: 6.333	3rd Qu.: 1.4000	3rd Qu.: 0.2222	3rd Qu.: 88.26
Max. : 41.000	Max. : 17.0000	Max. : 3.0000	Max. : 454.00

R_avg_opp_SIG_STR_landed R_avg_opp_SIG_STR_pct R_avg_opp_SUB_ATT R_avg_opp_TD_att

Min. : 0.00	Min. :0.0000	Min. :0.0000	Min.
: 0.000			
1st Qu.: 15.62	1st Qu.:0.3440	1st Qu.:0.0000	1st
Qu.: 1.000			
Median : 24.67	Median :0.4091	Median :0.2727	Medi
an : 2.400			
Mean : 26.97	Mean :0.4133	Mean :0.4491	Mean
: 2.831			
3rd Qu.: 35.58	3rd Qu.:0.4743	3rd Qu.:0.6667	3rd
Qu.: 4.000			
Max. :151.00	Max. :1.0000	Max. :8.0000	Max.
:22.000			
R_avg_opp_TD_landed	R_avg_opp_TD_pct	R_avg_opp_TOTAL_STR_att	R_avg_op
p_TOTAL_STR_landed			
Min. : 0.0000	Min. :0.0000	Min. : 0.00	Min. :
0.00			
1st Qu.: 0.3333	1st Qu.:0.0900	1st Qu.: 54.50	1st Qu.:
29.33			
Median : 0.8000	Median :0.2246	Median : 82.79	Median :
42.75			
Mean : 1.0230	Mean :0.2579	Mean : 86.28	Mean :
45.03			
3rd Qu.: 1.5000	3rd Qu.:0.3697	3rd Qu.:111.00	3rd Qu.:
57.31			
Max. :11.0000	Max. :1.0000	Max. :461.00	Max. :
202.00			
R_total_rounds_fought	R_total_time_fought.seconds.	R_total_title_bouts	
R_win_by_Decision_Majority			
Min. : 0.00	Min. : 7.0	Min. : 0.000	
Min. :0.00000			
1st Qu.: 3.00	1st Qu.: 470.6	1st Qu.: 0.000	
1st Qu.:0.00000			
Median : 9.00	Median : 620.3	Median : 0.000	
Median :0.00000			
Mean :12.85	Mean : 603.8	Mean : 0.597	
Mean :0.02761			
3rd Qu.:19.00	3rd Qu.: 762.3	3rd Qu.: 1.000	
3rd Qu.:0.00000			
Max. :80.00	Max. :1500.0	Max. :16.000	
Max. :2.00000			
R_win_by_Decision_Split	R_win_by_Decision_Unanimous	R_win_by_KO.TKO	
R_win_by_Submission			
Min. :0.0000	Min. : 0.000	Min. : 0.000	
Min. : 0.0000			

```

1st Qu.:0.0000      1st Qu.: 0.000      1st Qu.: 0.000
1st Qu.: 0.0000
Median :0.0000      Median : 1.000      Median : 1.000
Median : 0.0000
Mean :0.2809        Mean : 1.177        Mean : 1.255
Mean : 0.7776
3rd Qu.:0.0000      3rd Qu.: 2.000      3rd Qu.: 2.000
3rd Qu.: 1.0000
Max. :5.0000        Max. :10.000       Max. :11.000
Max. :13.0000

R_win_by_TKO_Doctor_Stoppage      R_wins      R_Stance      R_H
eight_cms      R_Reach_cms
Min. :0.00000      Min. : 0.000      Length:5144      Min.
:152.4      Min. :152.4
1st Qu.:0.00000      1st Qu.: 1.000      Class :character      1st
Qu.:172.7      1st Qu.:177.8
Median :0.00000      Median : 2.000      Mode :character      Medi
an :180.3      Median :182.9
Mean :0.07135      Mean : 3.598      Mean
:179.3      Mean :183.7
3rd Qu.:0.00000      3rd Qu.: 5.000      3rd
Qu.:185.4      3rd Qu.:190.5
Max. :2.00000      Max. :20.000      Max.
:210.8      Max. :213.4

R_Weight_lbs      B_age      R_age
Min. :115.0      Min. :18.00      Min. :19.00
1st Qu.:145.0      1st Qu.:26.00      1st Qu.:26.00
Median :170.0      Median :29.00      Median :29.00
Mean :172.1      Mean :29.17      Mean :29.44
3rd Qu.:185.0      3rd Qu.:32.00      3rd Qu.:32.00
Max. :345.0      Max. :51.00      Max. :47.00
[ reached getOption("max.print") -- omitted 1 row ]

```

```

>
> #Let us now check for missing values in the entire dataset column by
column
> #This function returns the number of all missing values in each of t
he columns
> colSums(is.na(df))

```

```

Referee      R_fighter      B_fighter
0
0
date      location
Winner

```


0	0		
	title_bout	weight_class	
no_of_rounds			
0	0		
B_current_lose_streak	B_current_win_streak		
B_draw			
0	0		
B_avg_BODY_att	B_avg_BODY_landed		
B_avg_CLINCH_att			
1265	1265		
1265			
B_avg_CLINCH_landed	B_avg_DISTANCE_att	B_avg	
_DISTANCE_landed			
1265	1265		
1265			
B_avg_GROUND_att	B_avg_GROUND_landed		
B_avg_HEAD_att			
1265	1265		
1265			
B_avg_HEAD_landed	B_avg_KD		
B_avg_LEG_att			
1265	1265		
1265			
B_avg_LEG_landed	B_avg_PASS		
B_avg_REV			
1265	1265		
1265			
B_avg_SIG_STR_att	B_avg_SIG_STR_landed	B	
_avg_SIG_STR_pct			
1265	1265		
1265			
B_avg_SUB_ATT	B_avg_TD_att		
B_avg_TD_landed			
1265	1265		
1265			
B_avg_TD_pct	B_avg_TOTAL_STR_att	B_avg_	
TOTAL_STR_landed			
1265	1265		
1265			
B_longest_win_streak	B_losses	B_	
avg_opp_BODY_att			

0	0	
1265		
B_avg_opp_BODY_landed	B_avg_opp_CLINCH_att	B_avg_o
pp_CLINCH_landed		
1265	1265	
1265		
B_avg_opp_DISTANCE_att	B_avg_opp_DISTANCE_landed	B_av
g_opp_GROUND_att		
1265	1265	
1265		
B_avg_opp_GROUND_landed	B_avg_opp_HEAD_att	B_avg
_opp_HEAD_landed		
1265	1265	
1265		
B_avg_opp_KD	B_avg_opp_LEG_att	B_av
g_opp_LEG_landed		
1265	1265	
1265		
B_avg_opp_PASS	B_avg_opp_REV	B_avg
_opp_SIG_STR_att		
1265	1265	
1265		
B_avg_opp_SIG_STR_landed	B_avg_opp_SIG_STR_pct	B
_avg_opp_SUB_ATT		
1265	1265	
1265		
B_avg_opp_TD_att	B_avg_opp_TD_landed	
B_avg_opp_TD_pct		
1265	1265	
1265		
B_avg_opp_TOTAL_STR_att	B_avg_opp_TOTAL_STR_landed	B_tot
al_rounds_fought		
1265	1265	
0		
B_total_time_fought.seconds.	B_total_title_bouts	B_win_by_D
ecision_Majority		
1265	0	
0		
B_win_by_Decision_Split	B_win_by_Decision_Unanimous	
B_win_by_KO.TKO		
0	0	
0		
B_win_by_Submission	B_win_by_TKO_Doctor_Stoppage	
B_wins		

0	0	0	
B_Reach_cms	B_Stance	B_Height_cms	
666	0	8	
urrent_win_streak	B_Weight_lbs	R_current_lose_streak	R_cu
0	6	0	
_avg_BODY_landed	R_draw	R_avg_BODY_att	R
650	0	650	
avg_DISTANCE_att	R_avg_CLINCH_att	R_avg_CLINCH_landed	R_
650	650	650	
vg_GROUND_landed	R_avg_DISTANCE_landed	R_avg_GROUND_att	R_a
650	650	650	
R_avg_KD	R_avg_HEAD_att	R_avg_HEAD_landed	
650	650	650	
R_avg_PASS	R_avg_LEG_att	R_avg_LEG_landed	
650	650	650	
g_SIG_STR_landed	R_avg_REV	R_avg_SIG_STR_att	R_av
650	650	650	
R_avg_TD_att	R_avg_SIG_STR_pct	R_avg_SUB_ATT	
650	650	650	
vg_TOTAL_STR_att	R_avg_TD_landed	R_avg_TD_pct	R_a
650	650	650	
R_avg_TOTAL_STR_landed		R_longest_win_streak	
R_losses			

0	650	0	
R_avg_opp_BODY_att		R_avg_opp_BODY_landed	R_av
g_opp_CLINCH_att	650	650	
650			
R_avg_opp_CLINCH_landed		R_avg_opp_DISTANCE_att	R_avg_opp
_DISTANCE_landed	650	650	
650			
R_avg_opp_GROUND_att		R_avg_opp_GROUND_landed	R_
avg_opp_HEAD_att	650	650	
650			
R_avg_opp_HEAD_landed		R_avg_opp_KD	R
_avg_opp_LEG_att	650	650	
650			
R_avg_opp_LEG_landed		R_avg_opp_PASS	
R_avg_opp_REV	650	650	
650			
R_avg_opp_SIG_STR_att		R_avg_opp_SIG_STR_landed	R_avg
_opp_SIG_STR_pct	650	650	
650			
R_avg_opp_SUB_ATT		R_avg_opp_TD_att	R_a
vg_opp_TD_landed	650	650	
650			
R_avg_opp_TD_pct		R_avg_opp_TOTAL_STR_att	R_avg_opp_
TOTAL_STR_landed	650	650	
650			
R_total_rounds_fought	R_total_time_fought.seconds.		R_t
otal_title_bouts	0	650	
0			
R_win_by_Decision_Majority	R_win_by_Decision_Split	R_win_by_De	
cision_Unanimous	0	0	
0			
R_win_by_KO.TKO	R_win_by_Submission	R_win_by_TKO	
_Doctor_Stoppage			

```

0
0
R_Height_cms      0
4
R_Reach_cms      0
B_age            316
172              64
R_age            64
>
> #Store and identify all columns that contain missing values.
> list_na <- colnames(df)[ apply(df, 2, anyNA) ]
> #Display list
> list_na
[1] "B_avg_BODY_att"      "B_avg_BODY_landed"      "B
_avg_CLINCH_att"
[4] "B_avg_CLINCH_landed" "B_avg_DISTANCE_att"      "B
_avg_DISTANCE_landed"
[7] "B_avg_GROUND_att"    "B_avg_GROUND_landed"    "B
_avg_HEAD_att"
[10] "B_avg_HEAD_landed"   "B_avg_KD"                "B
_avg_LEG_att"
[13] "B_avg_LEG_landed"    "B_avg_PASS"              "B
_avg_REV"
[16] "B_avg_SIG_STR_att"   "B_avg_SIG_STR_landed"    "B
_avg_SIG_STR_pct"
[19] "B_avg_SUB_ATT"       "B_avg_TD_att"            "B
_avg_TD_landed"
[22] "B_avg_TD_pct"        "B_avg_TOTAL_STR_att"     "B
_avg_TOTAL_STR_landed"
[25] "B_avg_opp_BODY_att"  "B_avg_opp_BODY_landed"   "B
_avg_opp_CLINCH_att"
[28] "B_avg_opp_CLINCH_landed" "B_avg_opp_DISTANCE_att" "B
_avg_opp_DISTANCE_landed"
[31] "B_avg_opp_GROUND_att" "B_avg_opp_GROUND_landed" "B
_avg_opp_HEAD_att"
[34] "B_avg_opp_HEAD_landed" "B_avg_opp_KD"            "B
_avg_opp_LEG_att"
[37] "B_avg_opp_LEG_landed" "B_avg_opp_PASS"          "B
_avg_opp_REV"

```

[40] "B_avg_opp_SIG_STR_att"	"B_avg_opp_SIG_STR_landed"	"B
_avg_opp_SIG_STR_pct"		
[43] "B_avg_opp_SUB_ATT"	"B_avg_opp_TD_att"	"B
_avg_opp_TD_landed"		
[46] "B_avg_opp_TD_pct"	"B_avg_opp_TOTAL_STR_att"	"B
_avg_opp_TOTAL_STR_landed"		
[49] "B_total_time_fought.seconds."	"B_Height_cms"	"B
_Reach_cms"		
[52] "B_Weight_lbs"	"R_avg_BODY_att"	"R
_avg_BODY_landed"		
[55] "R_avg_CLINCH_att"	"R_avg_CLINCH_landed"	"R
_avg_DISTANCE_att"		
[58] "R_avg_DISTANCE_landed"	"R_avg_GROUND_att"	"R
_avg_GROUND_landed"		
[61] "R_avg_HEAD_att"	"R_avg_HEAD_landed"	"R
_avg_KD"		
[64] "R_avg_LEG_att"	"R_avg_LEG_landed"	"R
_avg_PASS"		
[67] "R_avg_REV"	"R_avg_SIG_STR_att"	"R
_avg_SIG_STR_landed"		
[70] "R_avg_SIG_STR_pct"	"R_avg_SUB_ATT"	"R
_avg_TD_att"		
[73] "R_avg_TD_landed"	"R_avg_TD_pct"	"R
_avg_TOTAL_STR_att"		
[76] "R_avg_TOTAL_STR_landed"	"R_avg_opp_BODY_att"	"R
_avg_opp_BODY_landed"		
[79] "R_avg_opp_CLINCH_att"	"R_avg_opp_CLINCH_landed"	"R
_avg_opp_DISTANCE_att"		
[82] "R_avg_opp_DISTANCE_landed"	"R_avg_opp_GROUND_att"	"R
_avg_opp_GROUND_landed"		
[85] "R_avg_opp_HEAD_att"	"R_avg_opp_HEAD_landed"	"R
_avg_opp_KD"		
[88] "R_avg_opp_LEG_att"	"R_avg_opp_LEG_landed"	"R
_avg_opp_PASS"		
[91] "R_avg_opp_REV"	"R_avg_opp_SIG_STR_att"	"R
_avg_opp_SIG_STR_landed"		
[94] "R_avg_opp_SIG_STR_pct"	"R_avg_opp_SUB_ATT"	"R
_avg_opp_TD_att"		
[97] "R_avg_opp_TD_landed"	"R_avg_opp_TD_pct"	"R
_avg_opp_TOTAL_STR_att"		
[100] "R_avg_opp_TOTAL_STR_landed"	"R_total_time_fought.seconds."	"R
_Height_cms"		
[103] "R_Reach_cms"	"R_Weight_lbs"	"B
_age"		

```
[106] "R_age"
```

```
> #Import dplyr library
```

```
> #The dplyr package consists of many functions specifically used for  
data manipulation.
```

```
> #These functions process data faster than Base R functions and are k  
nown the best for data exploration and transformation, as well.
```

```
> library(dplyr)
```

```
> #Dropping all rows that contain missing values using na.omit()
```

```
> #This function removes all incomplete cases of a data object typical  
ly of a data frame, matrix or vector
```

```
> df_drop <-df %>% na.omit()
```

```
> #Storing in another file
```

```
> dim(df_drop)
```

```
[1] 3355 145
```

```
> #Finding missing values in the new dataframe
```

```
> colSums(is.na(df_drop))
```

	R_fighter	B_fighter	
Referee	0	0	
0			
	date	location	
Winner	0	0	
0			
	title_bout	weight_class	
no_of_rounds	0	0	
0			
	B_current_lose_streak	B_current_win_streak	
B_draw	0	0	
0			
	B_avg_BODY_att	B_avg_BODY_landed	
B_avg_CLINCH_att	0	0	
0			
	B_avg_CLINCH_landed	B_avg_DISTANCE_att	B_avg
_DISTANCE_landed	0	0	
0			
	B_avg_GROUND_att	B_avg_GROUND_landed	
B_avg_HEAD_att	0	0	
0			

B_avg_HEAD_landed	B_avg_KD	
B_avg_LEG_att		
0	0	
B_avg_LEG_landed	B_avg_PASS	
B_avg_REV		
0	0	
B_avg_SIG_STR_att	B_avg_SIG_STR_landed	B
_avg_SIG_STR_pct		
0	0	
B_avg_SUB_ATT	B_avg_TD_att	
B_avg_TD_landed		
0	0	
B_avg_TD_pct	B_avg_TOTAL_STR_att	B_avg_
TOTAL_STR_landed		
0	0	
B_longest_win_streak	B_losses	B_
avg_opp_BODY_att		
0	0	
B_avg_opp_BODY_landed	B_avg_opp_CLINCH_att	B_avg_o
pp_CLINCH_landed		
0	0	
B_avg_opp_DISTANCE_att	B_avg_opp_DISTANCE_landed	B_av
g_opp_GROUND_att		
0	0	
B_avg_opp_GROUND_landed	B_avg_opp_HEAD_att	B_avg
_opp_HEAD_landed		
0	0	
B_avg_opp_KD	B_avg_opp_LEG_att	B_av
g_opp_LEG_landed		
0	0	
B_avg_opp_PASS	B_avg_opp_REV	B_avg
_opp_SIG_STR_att		
0	0	

B_avg_opp_SIG_STR_landed	B_avg_opp_SIG_STR_pct	B
_avg_opp_SUB_ATT		
0	0	
B_avg_opp_TD_att	B_avg_opp_TD_landed	
B_avg_opp_TD_pct		
0	0	
B_avg_opp_TOTAL_STR_att	B_avg_opp_TOTAL_STR_landed	B_tot
al_rounds_fought		
0	0	
B_total_time_fought.seconds.	B_total_title_bouts	B_win_by_D
ecision_Majority		
0	0	
B_win_by_Decision_Split	B_win_by_Decision_Unanimous	
B_win_by_KO.TKO		
0	0	
B_win_by_Submission	B_win_by_TKO_Doctor_Stoppage	
B_wins		
0	0	
B_Stance	B_Height_cms	
B_Reach_cms		
0	0	
B_Weight_lbs	R_current_lose_streak	R_cu
rrent_win_streak		
0	0	
R_draw	R_avg_BODY_att	R
_avg_BODY_landed		
0	0	
R_avg_CLINCH_att	R_avg_CLINCH_landed	R_
avg_DISTANCE_att		
0	0	
R_avg_DISTANCE_landed	R_avg_GROUND_att	R_a
vg_GROUND_landed		
0	0	

R_avg_KD	R_avg_HEAD_att	R_avg_HEAD_landed	
0	0	0	
R_avg_PASS	R_avg_LEG_att	R_avg_LEG_landed	
0	0	0	
g_SIG_STR_landed	R_avg_REV	R_avg_SIG_STR_att	R_av
0	0	0	
R_avg_TD_att	R_avg_SIG_STR_pct	R_avg_SUB_ATT	
0	0	0	
vg_TOTAL_STR_att	R_avg_TD_landed	R_avg_TD_pct	R_a
0	0	0	
R_avg_TOTAL_STR_landed	R_longest_win_streak		
R_losses	0	0	
0			
g_opp_CLINCH_att	R_avg_opp_BODY_att	R_avg_opp_BODY_landed	R_av
0	0	0	
_DISTANCE_landed	R_avg_opp_CLINCH_landed	R_avg_opp_DISTANCE_att	R_avg_opp
0	0	0	
avg_opp_HEAD_att	R_avg_opp_GROUND_att	R_avg_opp_GROUND_landed	R_
0	0	0	
_avg_opp_LEG_att	R_avg_opp_HEAD_landed	R_avg_opp_KD	R
0	0	0	
R_avg_opp_LEG_landed	R_avg_opp_PASS		
R_avg_opp_REV	0	0	
0			

R_avg_opp_SIG_STR_att	R_avg_opp_SIG_STR_landed	R_avg
_opp_SIG_STR_pct		
0	0	
R_avg_opp_SUB_ATT	R_avg_opp_TD_att	R_a
vg_opp_TD_landed		
0	0	
R_avg_opp_TD_pct	R_avg_opp_TOTAL_STR_att	R_avg_opp_
TOTAL_STR_landed		
0	0	
R_total_rounds_fought	R_total_time_fought.seconds.	R_t
otal_title_bouts		
0	0	
R_win_by_Decision_Majority	R_win_by_Decision_Split	R_win_by_De
cision_Unanimous		
0	0	
R_win_by_KO.TKO	R_win_by_Submission	R_win_by_TKO
_Doctor_Stoppage		
0	0	
R_wins	R_Stance	
R_Height_cms		
0	0	
R_Reach_cms	R_Weight_lbs	
B_age		
0	0	
R_age		
0		

>

> #Another method of cleaning data.

> #Let us now use complete.cases to find the number of complete rows

> #We can also create a complete subset of our data by using the complete.cases function.

> comp_df<-df[complete.cases(df),]

> #Finding the columns and rows in this dataset

> dim(comp_df)

[1] 3355 145

> #Finding missing values in the new dataframe

```
> colSums(is.na(comp_df))
```

Referee	R_fighter	B_fighter	
0	0	0	
Winner	date	location	
0	0	0	
no_of_rounds	title_bout	weight_class	
0	0	0	
B_current_lose_streak	B_current_win_streak		
B_draw	0	0	
0			
B_avg_BODY_att	B_avg_BODY_landed		
B_avg_CLINCH_att	0	0	
0			
B_avg_CLINCH_landed	B_avg_DISTANCE_att	B_avg	
_DISTANCE_landed	0	0	
0			
B_avg_GROUND_att	B_avg_GROUND_landed		
B_avg_HEAD_att	0	0	
0			
B_avg_HEAD_landed	B_avg_KD		
B_avg_LEG_att	0	0	
0			
B_avg_LEG_landed	B_avg_PASS		
B_avg_REV	0	0	
0			
B_avg_SIG_STR_att	B_avg_SIG_STR_landed	B	
_avg_SIG_STR_pct	0	0	
0			
B_avg_SUB_ATT	B_avg_TD_att		
B_avg_TD_landed			

0	0		
0	B_avg_TD_pct	B_avg_TOTAL_STR_att	B_avg_
TOTAL_STR_landed	0	0	
0	B_longest_win_streak	B_losses	B_
avg_opp_BODY_att	0	0	
0	B_avg_opp_BODY_landed	B_avg_opp_CLINCH_att	B_avg_o
pp_CLINCH_landed	0	0	
0	B_avg_opp_DISTANCE_att	B_avg_opp_DISTANCE_landed	B_av
g_opp_GROUND_att	0	0	
0	B_avg_opp_GROUND_landed	B_avg_opp_HEAD_att	B_avg
_opp_HEAD_landed	0	0	
0	B_avg_opp_KD	B_avg_opp_LEG_att	B_av
g_opp_LEG_landed	0	0	
0	B_avg_opp_PASS	B_avg_opp_REV	B_avg
_opp_SIG_STR_att	0	0	
0	B_avg_opp_SIG_STR_landed	B_avg_opp_SIG_STR_pct	B
_avg_opp_SUB_ATT	0	0	
0	B_avg_opp_TD_att	B_avg_opp_TD_landed	
B_avg_opp_TD_pct	0	0	
0	B_avg_opp_TOTAL_STR_att	B_avg_opp_TOTAL_STR_landed	B_tot
al_rounds_fought	0	0	
0	B_total_time_fought.seconds.	B_total_title_bouts	B_win_by_D
ecision_Majority			

0	0	0	
B_win_by_Decision_Split	B_win_by_Decision_Unanimous		
B_win_by_KO.TKO			
0	0		
B_win_by_Submission	B_win_by_TKO_Doctor_Stoppage		
B_wins			
0	0		
B_Reach_cms	B_Stance	B_Height_cms	
0	0		
rrrent_win_streak	B_Weight_lbs	R_current_lose_streak	R_cu
0	0		
_avg_BODY_landed	R_draw	R_avg_BODY_att	R
0	0		
avg_DISTANCE_att	R_avg_CLINCH_att	R_avg_CLINCH_landed	R_
0	0		
vg_GROUND_landed	R_avg_DISTANCE_landed	R_avg_GROUND_att	R_a
0	0		
R_avg_KD	R_avg_HEAD_att	R_avg_HEAD_landed	
0	0		
R_avg_PASS	R_avg_LEG_att	R_avg_LEG_landed	
0	0		
g_SIG_STR_landed	R_avg_REV	R_avg_SIG_STR_att	R_av
0	0		
R_avg_TD_att	R_avg_SIG_STR_pct	R_avg_SUB_ATT	

0	0	0	
R_avg_TD_landed	R_avg_TD_pct	R_a	
vg_TOTAL_STR_att			
0	0		
R_avg_TOTAL_STR_landed	R_longest_win_streak		
R_losses			
0	0		
R_avg_opp_BODY_att	R_avg_opp_BODY_landed	R_av	
g_opp_CLINCH_att			
0	0		
R_avg_opp_CLINCH_landed	R_avg_opp_DISTANCE_att	R_avg_opp	
_DISTANCE_landed			
0	0		
R_avg_opp_GROUND_att	R_avg_opp_GROUND_landed	R_	
avg_opp_HEAD_att			
0	0		
R_avg_opp_HEAD_landed	R_avg_opp_KD	R	
_avg_opp_LEG_att			
0	0		
R_avg_opp_LEG_landed	R_avg_opp_PASS		
R_avg_opp_REV			
0	0		
R_avg_opp_SIG_STR_att	R_avg_opp_SIG_STR_landed	R_avg	
_opp_SIG_STR_pct			
0	0		
R_avg_opp_SUB_ATT	R_avg_opp_TD_att	R_a	
vg_opp_TD_landed			
0	0		
R_avg_opp_TD_pct	R_avg_opp_TOTAL_STR_att	R_avg_opp_	
TOTAL_STR_landed			
0	0		
R_total_rounds_fought	R_total_time_fought.seconds.	R_t	
otal_title_bouts			

```

0
0
R_win_by_Decision_Majority      R_win_by_Decision_Split  R_win_by_De
cision_Unanimous
0
0
R_win_by_KO.TKO      R_win_by_Submission  R_win_by_TKO
_Doctor_Stoppage
0
0
R_wins      R_Stance
R_Height_cms
0
0
R_Reach_cms      R_Weight_lbs
B_age
0
0
R_age
0
>
>
> #We can also replace missing values in our dataset using the mean or
median
> #Find and identify the columns like done previously
> list_na<- colnames(df)[ apply(df, 2, anyNA) ]
> #Display list
> list_na
[1] "B_avg_BODY_att"      "B_avg_BODY_landed"      "B
_avg_CLINCH_att"
[4] "B_avg_CLINCH_landed"  "B_avg_DISTANCE_att"      "B
_avg_DISTANCE_landed"
[7] "B_avg_GROUND_att"     "B_avg_GROUND_landed"     "B
_avg_HEAD_att"
[10] "B_avg_HEAD_landed"    "B_avg_KD"                "B
_avg_LEG_att"
[13] "B_avg_LEG_landed"     "B_avg_PASS"              "B
_avg_REV"
[16] "B_avg_SIG_STR_att"    "B_avg_SIG_STR_landed"    "B
_avg_SIG_STR_pct"
[19] "B_avg_SUB_ATT"        "B_avg_TD_att"            "B
_avg_TD_landed"
[22] "B_avg_TD_pct"         "B_avg_TOTAL_STR_att"     "B
_avg_TOTAL_STR_landed"

```


[25] "B_avg_opp_BODY_att"	"B_avg_opp_BODY_landed"	"B
_avg_opp_CLINCH_att"		
[28] "B_avg_opp_CLINCH_landed"	"B_avg_opp_DISTANCE_att"	"B
_avg_opp_DISTANCE_landed"		
[31] "B_avg_opp_GROUND_att"	"B_avg_opp_GROUND_landed"	"B
_avg_opp_HEAD_att"		
[34] "B_avg_opp_HEAD_landed"	"B_avg_opp_KD"	"B
_avg_opp_LEG_att"		
[37] "B_avg_opp_LEG_landed"	"B_avg_opp_PASS"	"B
_avg_opp_REV"		
[40] "B_avg_opp_SIG_STR_att"	"B_avg_opp_SIG_STR_landed"	"B
_avg_opp_SIG_STR_pct"		
[43] "B_avg_opp_SUB_ATT"	"B_avg_opp_TD_att"	"B
_avg_opp_TD_landed"		
[46] "B_avg_opp_TD_pct"	"B_avg_opp_TOTAL_STR_att"	"B
_avg_opp_TOTAL_STR_landed"		
[49] "B_total_time_fought.seconds."	"B_Height_cms"	"B
_Reach_cms"		
[52] "B_Weight_lbs"	"R_avg_BODY_att"	"R
_avg_BODY_landed"		
[55] "R_avg_CLINCH_att"	"R_avg_CLINCH_landed"	"R
_avg_DISTANCE_att"		
[58] "R_avg_DISTANCE_landed"	"R_avg_GROUND_att"	"R
_avg_GROUND_landed"		
[61] "R_avg_HEAD_att"	"R_avg_HEAD_landed"	"R
_avg_KD"		
[64] "R_avg_LEG_att"	"R_avg_LEG_landed"	"R
_avg_PASS"		
[67] "R_avg_REV"	"R_avg_SIG_STR_att"	"R
_avg_SIG_STR_landed"		
[70] "R_avg_SIG_STR_pct"	"R_avg_SUB_ATT"	"R
_avg_TD_att"		
[73] "R_avg_TD_landed"	"R_avg_TD_pct"	"R
_avg_TOTAL_STR_att"		
[76] "R_avg_TOTAL_STR_landed"	"R_avg_opp_BODY_att"	"R
_avg_opp_BODY_landed"		
[79] "R_avg_opp_CLINCH_att"	"R_avg_opp_CLINCH_landed"	"R
_avg_opp_DISTANCE_att"		
[82] "R_avg_opp_DISTANCE_landed"	"R_avg_opp_GROUND_att"	"R
_avg_opp_GROUND_landed"		
[85] "R_avg_opp_HEAD_att"	"R_avg_opp_HEAD_landed"	"R
_avg_opp_KD"		
[88] "R_avg_opp_LEG_att"	"R_avg_opp_LEG_landed"	"R
_avg_opp_PASS"		

```

[91] "R_avg_opp_REV"          "R_avg_opp_SIG_STR_att"      "R
_avg_opp_SIG_STR_landed"
[94] "R_avg_opp_SIG_STR_pct"  "R_avg_opp_SUB_ATT"         "R
_avg_opp_TD_att"
[97] "R_avg_opp_TD_landed"    "R_avg_opp_TD_pct"          "R
_avg_opp_TOTAL_STR_att"
[100] "R_avg_opp_TOTAL_STR_landed" "R_total_time_fought.seconds." "R
_Height_cms"
[103] "R_Reach_cms"           "R_Weight_lbs"              "B
_age"
[106] "R_age"

```

```

>
> #Compute the mean with the argument na.rm = TRUE.
> #This argument is compulsory because the columns have missing data,
and this tells R to ignore them.
> # apply() function takes list, vector or data frame as input and gi
ves output in vector or matrix.
> #It is useful for operations on list objects and returns a list obje
ct of same length of original set.
> df_new <-data.frame(
+   apply(
+     df, #Your dataset
+     function(x) ifelse(is.na(x), #Function to check for missing valu
es and replace it with mean
+                               mean(x, na.rm = TRUE),
+                               x)))
>
>
> #Check the missing values now
> colSums(is.na(df_new))

```

	R_fighter	B_fighter
Referee	0	0
0		
	date	location
Winner	0	0
0		
	title_bout	weight_class
no_of_rounds	0	0
0		
	B_current_lose_streak	B_current_win_streak
B_draw		

0	0		
B_avg_BODY_att	B_avg_BODY_landed		
B_avg_CLINCH_att			
0	0		
B_avg_CLINCH_landed	B_avg_DISTANCE_att	B_avg	
_DISTANCE_landed			
0	0		
B_avg_GROUND_att	B_avg_GROUND_landed		
B_avg_HEAD_att			
0	0		
B_avg_HEAD_landed	B_avg_KD		
B_avg_LEG_att			
0	0		
B_avg_LEG_landed	B_avg_PASS		
B_avg_REV			
0	0		
B_avg_SIG_STR_att	B_avg_SIG_STR_landed	B	
_avg_SIG_STR_pct			
0	0		
B_avg_SUB_ATT	B_avg_TD_att		
B_avg_TD_landed			
0	0		
B_avg_TD_pct	B_avg_TOTAL_STR_att	B_avg_	
TOTAL_STR_landed			
0	0		
B_longest_win_streak	B_losses	B_	
avg_opp_BODY_att			
0	0		
B_avg_opp_BODY_landed	B_avg_opp_CLINCH_att	B_avg_o	
pp_CLINCH_landed			
0	0		
B_avg_opp_DISTANCE_att	B_avg_opp_DISTANCE_landed	B_av	
g_opp_GROUND_att			

0	0	0	
B_avg_opp_GROUND_landed	B_avg_opp_HEAD_att	B_avg	
_opp_HEAD_landed			
0	0		
B_avg_opp_KD	B_avg_opp_LEG_att	B_av	
g_opp_LEG_landed			
0	0		
B_avg_opp_PASS	B_avg_opp_REV	B_avg	
_opp_SIG_STR_att			
0	0		
B_avg_opp_SIG_STR_landed	B_avg_opp_SIG_STR_pct	B	
_avg_opp_SUB_ATT			
0	0		
B_avg_opp_TD_att	B_avg_opp_TD_landed		
B_avg_opp_TD_pct			
0	0		
B_avg_opp_TOTAL_STR_att	B_avg_opp_TOTAL_STR_landed	B_tot	
al_rounds_fought			
0	0		
B_total_time_fought.seconds.	B_total_title_bouts	B_win_by_D	
ecision_Majority			
0	0		
B_win_by_Decision_Split	B_win_by_Decision_Unanimous		
B_win_by_KO.TKO			
0	0		
B_win_by_Submission	B_win_by_TKO_Doctor_Stoppage		
B_wins			
0	0		
B_Stance	B_Height_cms		
B_Reach_cms			
0	0		
B_Weight_lbs	R_current_lose_streak	R_cu	
rrent_win_streak			

0	0	0	
	R_draw	R_avg_BODY_att	R
_avg_BODY_landed	0	0	
0	R_avg_CLINCH_att	R_avg_CLINCH_landed	R_
avg_DISTANCE_att	0	0	
0	R_avg_DISTANCE_landed	R_avg_GROUND_att	R_a
vg_GROUND_landed	0	0	
0	R_avg_HEAD_att	R_avg_HEAD_landed	
R_avg_KD	0	0	
0	R_avg_LEG_att	R_avg_LEG_landed	
R_avg_PASS	0	0	
0	R_avg_REV	R_avg_SIG_STR_att	R_av
g_SIG_STR_landed	0	0	
0	R_avg_SIG_STR_pct	R_avg_SUB_ATT	
R_avg_TD_att	0	0	
0	R_avg_TD_landed	R_avg_TD_pct	R_a
vg_TOTAL_STR_att	0	0	
0	R_avg_TOTAL_STR_landed	R_longest_win_streak	
R_losses	0	0	
0	R_avg_opp_BODY_att	R_avg_opp_BODY_landed	R_av
g_opp_CLINCH_att	0	0	
0	R_avg_opp_CLINCH_landed	R_avg_opp_DISTANCE_att	R_avg_opp
_DISTANCE_landed			

0	0	0	
R_avg_opp_GROUND_att	R_avg_opp_GROUND_landed	R_avg_opp_HEAD_att	R_avg_opp_HEAD_landed
0	0	0	0
R_avg_opp_HEAD_landed	R_avg_opp_KD	R_avg_opp_LEG_att	R_avg_opp_LEG_landed
0	0	0	0
R_avg_opp_LEG_landed	R_avg_opp_PASS	R_avg_opp_REV	
0	0	0	0
R_avg_opp_SIG_STR_att	R_avg_opp_SIG_STR_landed	R_avg_opp_SIG_STR_pct	
0	0	0	0
R_avg_opp_SUB_ATT	R_avg_opp_TD_att	R_avg_opp_TD_landed	
0	0	0	0
R_avg_opp_TD_pct	R_avg_opp_TOTAL_STR_att	R_avg_opp_TOTAL_STR_landed	
0	0	0	0
R_total_rounds_fought	R_total_time_fought.seconds.	R_total_title_bouts	
0	0	0	0
R_win_by_Decision_Majority	R_win_by_Decision_Split	R_win_by_Decision_Unanimous	
0	0	0	0
R_win_by_KO.TKO	R_win_by_Submission	R_win_by_TKO	
_Doctor_Stoppage			
0	0	0	0
R_wins	R_Stance		
R_Height_cms			
0	0	0	0
R_Reach_cms	R_Weight_lbs		
B_age			

```
0
0
0
R_age
0
>
> #Save cleaned file for future lab session
> write.csv(comp_df, file = "df_clean.csv")
```

Conclusion:

Thus, dataset has been cleaned and made ready for analysis.

Lab Outcome:LO6

Lab Assignment: 09

Aim:

EDA on the dataset

Code:

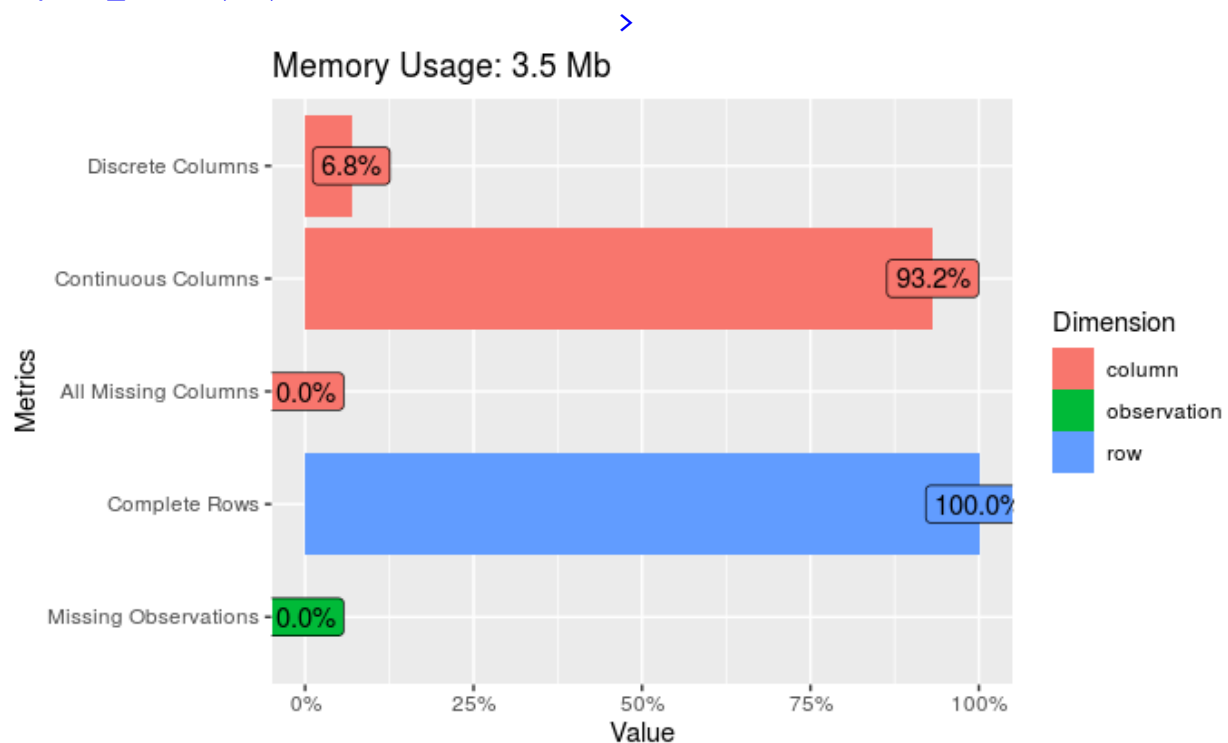
```
> #Lab Assignment 9
> #Mini-Project Session 3
>
> #Exploratory Data Analysis
>
> #get the working directory
> getwd()
[1] "/cloud/project"
>
> #Loading dataset into R
> #Header is set to to true if file contains Header information
> #Sep stands for seperator which is , in our csv file
> df <- read.csv("df_clean.csv", header = TRUE, sep = ",")
>
>
> #Exploratory data analysis is the process to get to know your data,
so that you can generate and test your hypothesis. Visualization techn
iques are usually applied.
> #EDA consists of univariate (1-variable) and bivariate (2-variables)
analysis.
> #DataExplorer can help you with different tasks throughout your data
exploration process.
> #Install the package if not installed
> #install.packages('DataExplorer')
>
> #import the library
> library(DataExplorer)
> library(GGally)
Loading required package: ggplot2
Registered S3 method overwritten by 'GGally':
  method from
+.gg      ggplot2
>
> #To get introduced to your newly created dataset:
> introduce(df)
```



```

rows columns discrete_columns
1 3355      146          10
continuous_columns all_missing_columns
1              136          0
total_missing_values complete_rows
1              0          3355
total_observations memory_usage
1          489830      3683888
>
> #plot_intro gives a brief Introduction to the dataset.
> #It covers basic information and gives us an idea about the content.
> plot_intro(df)

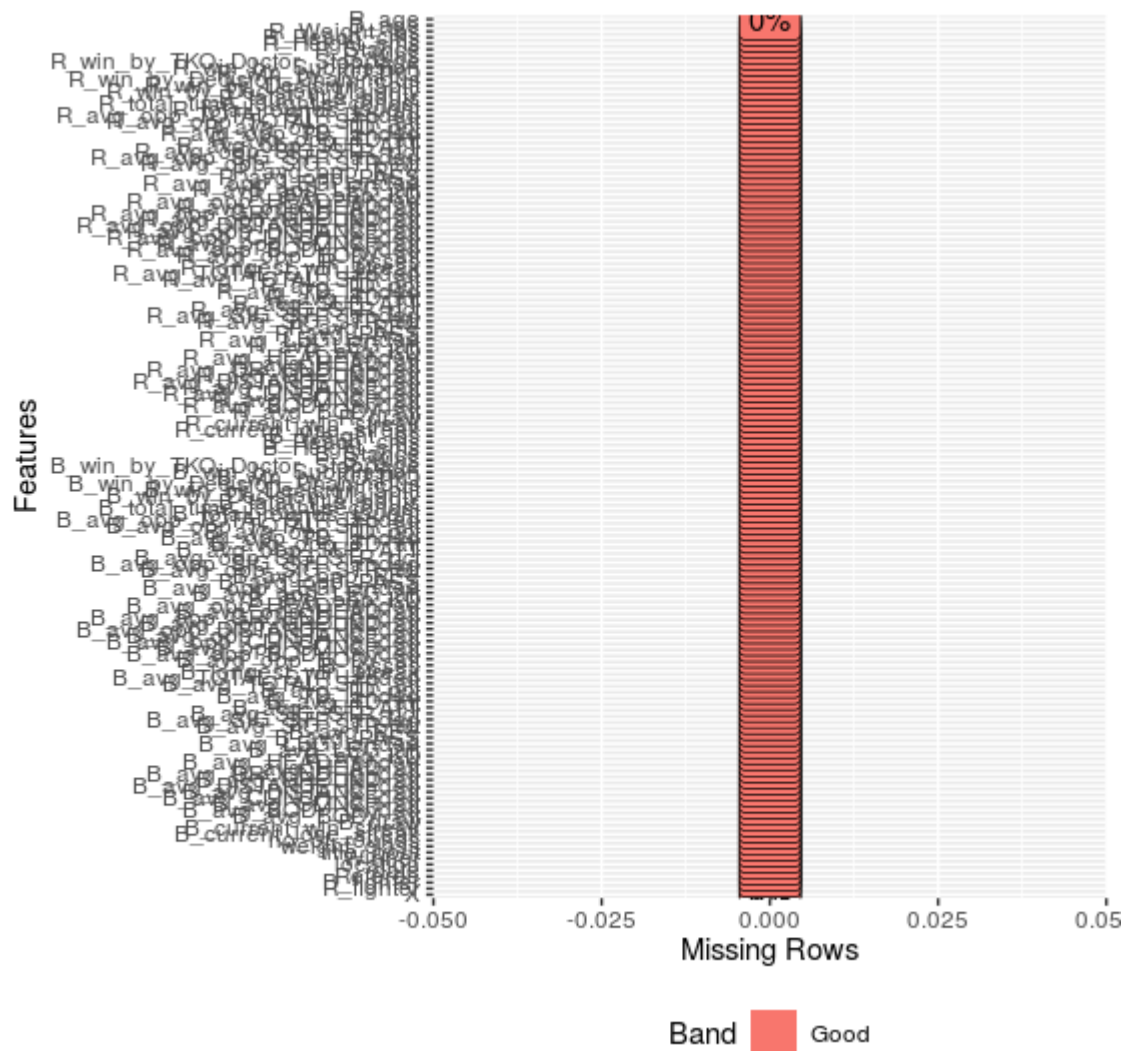
```



```

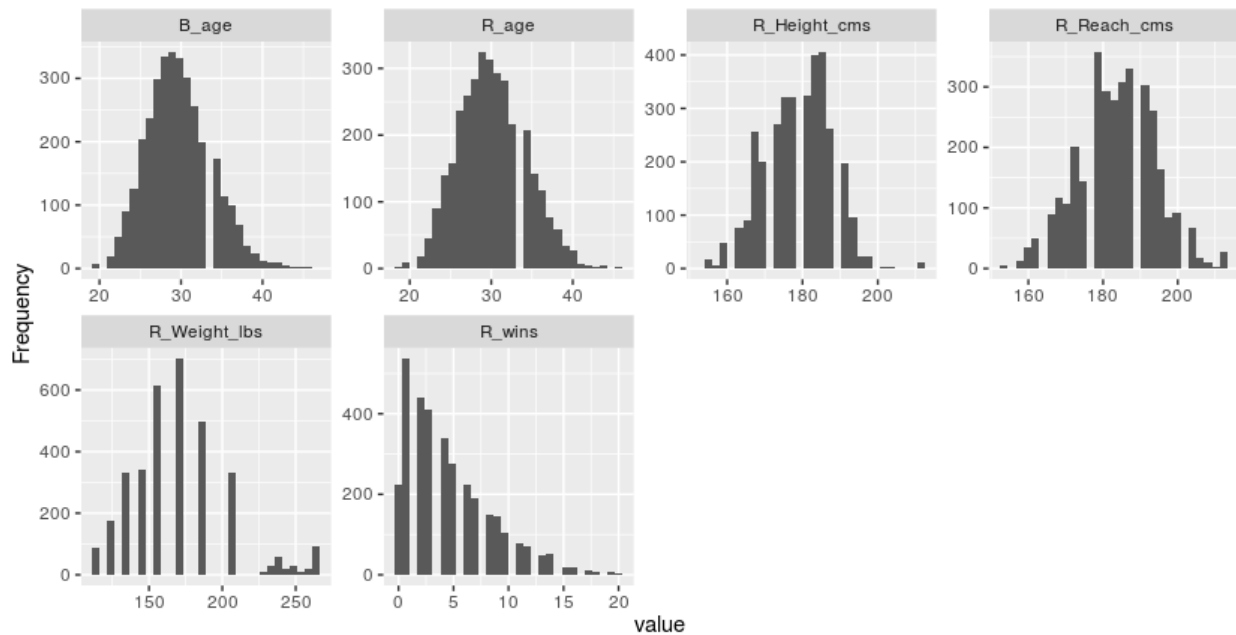
> #str gives the entire list of features in a network graph. We have the
  choice of viewing it radially or diagonally
> plot_str(df)
>
> #missing function returns and plots frequency of missing values for
  each feature.
> #It also advises us whether to remove certain columns before carrying
  out our analysis
> #All our features are in acceptable form
> plot_missing(df)

```



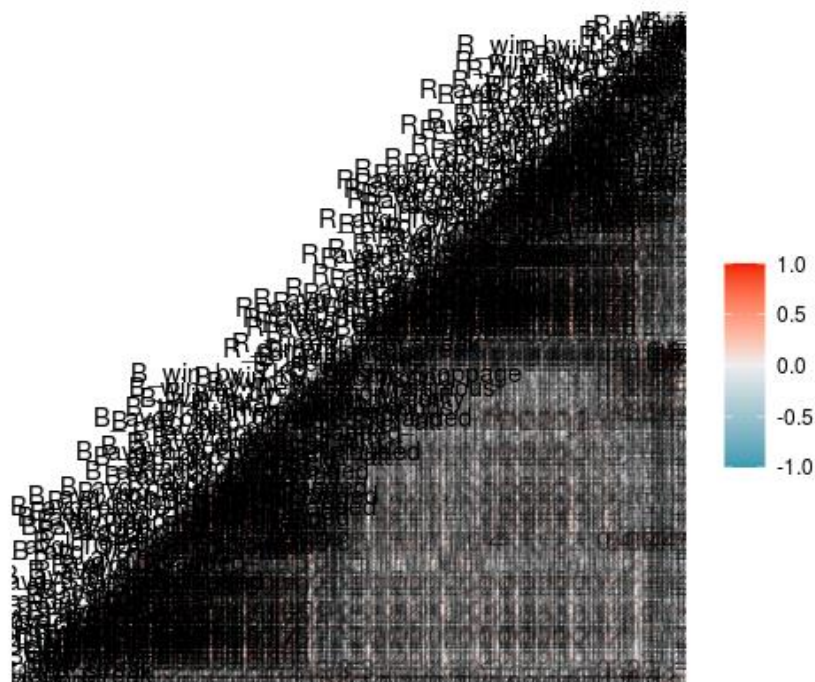
```
>
> #Let us first analyse and represent the continuous variables
> #Histograms can be used to analyse continuous variables
> # you can use a Histogram to organize and display the data in a more
userfriendly format.
> #A Histogram will make it easy
> #to see where the majority of values falls in a measurement scale, a
nd how much variation there is.
> plot_histogram(df)
```

>

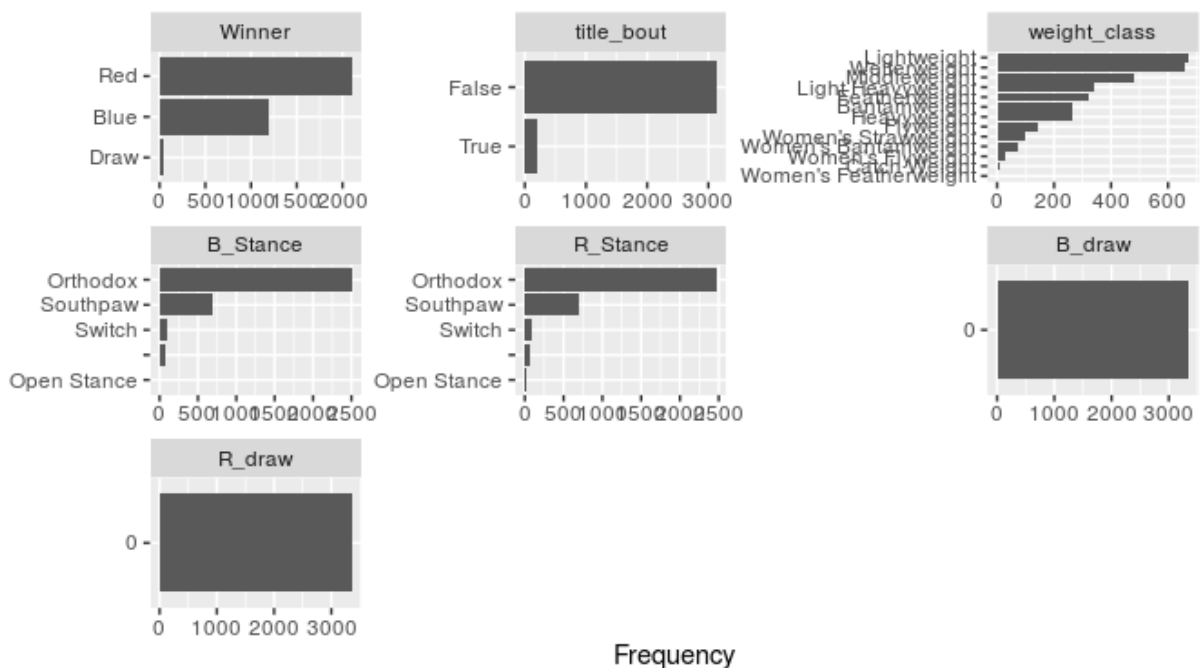


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```
> #For multivariate analysis, let us do correlation analysis
> #Correlation is used to test relationships between quantitative variables or categorical variables.
> #It's a measure of how things are related and how well they are related.
> ggcorr(df, label = TRUE, label_alpha = TRUE)
>
```



```
> #For categorical analysis, we use a bar graph
> #Each bar represents one value. When the bars are stacked next to one another,
> #the viewer can compare the different bars, or values, at a glance.
> plot_bar(df)
5 columns ignored with more than 50 categories.
R_fighter: 923 categories
B_fighter: 1114 categories
Referee: 171 categories
date: 445 categories
location: 146 categories
```



```
>
> #For additional complete reports, we can use the create_report function from dataexplorer
> #It creates an automatic EDA report and generates the report in an html file.
> #report_EDA.pdf is attached in the folder
> #create_report(df)
```

Conclusion:

Thus, EDA is performed on the dataset.

Lab Outcome:LO6

Lab Assignment: 10

Aim:

Regression analysis

Code:

```
> #Lab Assignment 10
> #Mini-Project Session 4
>
> #Regression analysis
>
> #get the working directory
> getwd()
[1] "/cloud/project"
>
> #Loading dataset into R
> #Header is set to true if file contains Header information
> #Sep stands for separator which is , in our csv file
> df <- read.csv("df_clean.csv", header = TRUE, sep = ",")
>
>
> #Regression analysis is used in stats to find trends in data
> #It will provide you with an equation for a graph so that you can make predictions about your data
> #Since we have a lot of values, we use multiple regression analysis
> #Multiple linear regression is an extension of simple linear regression used to
> #predict an outcome variable (y) on the basis of multiple distinct predictor variables (x).
> #Here we will predict the Wins of R and B based on variables
> #1. Height
> #2. Weight
> #3. Age
> #4. Reach
> #5. Total rounds fought
> #6. Total time fought
>
>
> modelR <- lm(R_wins ~ R_Height_cms + R_Reach_cms + R_Weight_lbs + R_age + R_total_rounds_fought + R_total_time_fought.seconds, data = df)
> summary(modelR)
```

Call:

```
lm(formula = R_wins ~ R_Height_cms + R_Reach_cms + R_Weight_lbs +
    R_age + R_total_rounds_fought + R_total_time_fought.seconds.,
    data = df)
```

Residuals:

Min	1Q	Median	3Q	Max
-7.1217	-0.7885	-0.0042	0.7619	6.0647

Coefficients:

	Estimate	
(Intercept)	-2.6123404	
R_Height_cms	0.0020585	
R_Reach_cms	0.0267762	
R_Weight_lbs	0.0002106	
R_age	-0.0270183	
R_total_rounds_fought	0.2679589	
R_total_time_fought.seconds.	-0.0027533	
	Std. Error	
(Intercept)	0.7106651	
R_Height_cms	0.0065516	
R_Reach_cms	0.0051411	
R_Weight_lbs	0.0012112	
R_age	0.0064898	
R_total_rounds_fought	0.0019300	
R_total_time_fought.seconds.	0.0001199	
	t value	Pr(> t)
(Intercept)	-3.676	0.000241
R_Height_cms	0.314	0.753386
R_Reach_cms	5.208	2.02e-07
R_Weight_lbs	0.174	0.861988
R_age	-4.163	3.22e-05
R_total_rounds_fought	138.836	< 2e-16
R_total_time_fought.seconds.	-22.966	< 2e-16

(Intercept)	***
R_Height_cms	
R_Reach_cms	***
R_Weight_lbs	
R_age	***
R_total_rounds_fought	***
R_total_time_fought.seconds.	***

Signif. codes:

```
0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Residual standard error: 1.338 on 3348 degrees of freedom

Multiple R-squared: 0.8795, Adjusted R-squared: 0.8793

F-statistic: 4074 on 6 and 3348 DF, p-value: < 2.2e-16

```
>
> #The first step in interpreting the multiple regression analysis is
> #to examine the F-statistic and the associated p-value, at the bottom of model summary.
```

```
> summary(modelR)$coefficient
```

	Estimate
(Intercept)	-2.6123403679
R_Height_cms	0.0020585422
R_Reach_cms	0.0267761544
R_Weight_lbs	0.0002105744
R_age	-0.0270182895
R_total_rounds_fought	0.2679588625
R_total_time_fought.seconds.	-0.0027532817
	Std. Error
(Intercept)	0.7106650709
R_Height_cms	0.0065516249
R_Reach_cms	0.0051410955
R_Weight_lbs	0.0012111899
R_age	0.0064897934
R_total_rounds_fought	0.0019300441
R_total_time_fought.seconds.	0.0001198842
	t value
(Intercept)	-3.6759093
R_Height_cms	0.3142033
R_Reach_cms	5.2082585
R_Weight_lbs	0.1738575
R_age	-4.1631972
R_total_rounds_fought	138.8356197
R_total_time_fought.seconds.	-22.9661718
	Pr(> t)
(Intercept)	2.407215e-04
R_Height_cms	7.533862e-01
R_Reach_cms	2.021309e-07
R_Weight_lbs	8.619880e-01
R_age	3.217298e-05
R_total_rounds_fought	0.000000e+00
R_total_time_fought.seconds.	1.630348e-108

```
>
```



```
> #For a given the predictor, the t-statistic evaluates whether or not
there is significant association between the predictor and the outcome
variable,
> #that is whether the beta coefficient of the predictor is significant
ly different from zero.
> # so for our given set of predictors we find that height and weight
is not significant in analysing the wins because its value is close to
0
> #Therefore we remove height and weight
> modelR <- lm(R_wins ~ R_Reach_cms + R_age + R_total_rounds_fought +
R_total_time_fought.seconds. ,data = df)
> summary(modelR)
```

Call:

```
lm(formula = R_wins ~ R_Reach_cms + R_age + R_total_rounds_fought +
    R_total_time_fought.seconds., data = df)
```

Residuals:

Min	1Q	Median	3Q	Max
-7.1208	-0.7902	-0.0062	0.7637	6.0593

Coefficients:

	Estimate		
(Intercept)	-2.5845949		
R_Reach_cms	0.0287740		
R_age	-0.0265275		
R_total_rounds_fought	0.2678970		
R_total_time_fought.seconds.	-0.0027602		
	Std. Error		
(Intercept)	0.4772775		
R_Reach_cms	0.0023339		
R_age	0.0062450		
R_total_rounds_fought	0.0019240		
R_total_time_fought.seconds.	0.0001187		
	t value	Pr(> t)	
(Intercept)	-5.415	6.55e-08	
R_Reach_cms	12.329	< 2e-16	
R_age	-4.248	2.22e-05	
R_total_rounds_fought	139.242	< 2e-16	
R_total_time_fought.seconds.	-23.247	< 2e-16	
(Intercept)	***		
R_Reach_cms	***		
R_age	***		

```
R_total_rounds_fought      ***
R_total_time_fought.seconds. ***
---
```

```
Signif. codes:
```

```
0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
Residual standard error: 1.338 on 3350 degrees of freedom
```

```
Multiple R-squared:  0.8795,      Adjusted R-squared:  0.8794
```

```
F-statistic:  6114 on 4 and 3350 DF,  p-value: < 2.2e-16
```

```
>
> #The RSE estimate gives a measure of error of prediction. The lower
the RSE, the more accurate the model (on the data in hand).
> #The error rate can be estimated by dividing the RSE by the mean out
come variable:
> #Our outcome variable is R_wins
> sigma(modelR)/mean(df$R_wins)
[1] 0.2850224
>
> #We get a 0.28 which means it has a 28% error rate
> #install.packages('caTools')
> library(caTools)
> #Let us now test the created prediction model
> #Splitting the data set into training and testing data
> split = sample.split(df$R_wins,SplitRatio = 2/3)
> #Training is 2 parts
> training_set = subset(df, split == TRUE)
> #testing is 1 part
> test_set = subset(df, split == FALSE)
>
> #Do a prediction on the model
> R_pred = predict(modelR, newdata = test_set)
>
> #Store and compare the result in the test cases
> resultR <- data.frame(Reach = test_set$R_Reach_cms, Age = test_set$R
_age, Total_rounds= test_set$R_total_rounds_fought, Total_seconds=test
_set$R_total_time_fought.seconds., Actual_Value = test_set$R_wins, Pre
dicted_Value = R_pred)
> resultR
```

	Reach	Age	Total_rounds	Total_seconds
3	193.04	35	33	604.4000
5	190.50	26	7	440.7500
6	167.64	28	8	540.0000
8	162.56	33	25	800.1111

10	160.02	29	9	900.0000
15	182.88	30	11	561.8000
17	177.80	37	4	461.0000
19	167.64	35	19	807.8571
20	193.04	27	1	193.0000
26	193.04	38	26	565.8333
28	177.80	29	9	858.3333
29	165.10	26	23	768.5000
33	172.72	31	12	900.0000
34	190.50	42	24	576.1818
36	182.88	28	22	656.6667
37	177.80	40	74	787.5769
39	195.58	35	22	355.6429
40	177.80	35	46	807.9375
41	162.56	27	6	900.0000
44	180.34	28	2	383.0000
46	182.88	39	23	509.1667
49	193.04	39	36	566.3750
56	213.36	26	1	261.0000
57	175.26	36	13	953.0000
58	187.96	34	6	554.0000
63	190.50	31	18	876.0000
65	203.20	36	41	565.7222
69	190.50	26	4	500.0000
72	198.12	37	19	611.6250
73	177.80	34	15	711.3333
78	177.80	29	18	900.0000
83	170.18	33	29	685.1667
93	187.96	31	14	510.8571
94	185.42	29	11	606.2000
99	200.66	34	36	568.5000
101	190.50	35	41	608.5556
111	187.96	36	32	723.8333
116	182.88	37	73	738.7500
119	198.12	35	26	736.1000
124	193.04	34	3	900.0000
135	185.42	24	3	900.0000
137	175.26	35	21	868.8571
140	203.20	29	11	802.0000
143	187.96	22	3	891.0000
144	193.04	25	1	46.0000
149	177.80	32	43	1027.9167
154	182.88	23	9	839.3333
156	180.34	32	9	436.8000

158	165.10	24	20	728.2857
159	193.04	39	35	590.7333
161	167.64	32	39	681.6875
162	182.88	30	19	776.8571
163	213.36	31	55	844.0556
165	190.50	34	43	801.4000
168	195.58	39	56	507.1481
177	180.34	31	15	497.7500
180	172.72	30	6	527.3333
190	177.80	29	26	697.5000
194	180.34	32	19	561.3333
195	190.50	28	12	695.8000
197	193.04	37	45	654.7778
198	182.88	44	43	656.1176
200	190.50	37	52	711.5000
204	177.80	34	69	766.5600
205	165.10	32	21	549.2000
206	165.10	34	39	737.6000
212	182.88	23	2	453.0000
213	160.02	29	6	900.0000
223	170.18	30	26	850.3333
226	177.80	32	7	601.6667
230	198.12	34	32	621.4615
233	195.58	34	21	360.6154
234	180.34	26	21	746.7500
235	170.18	33	3	900.0000
236	177.80	29	6	837.5000
240	205.74	26	5	659.5000
249	190.50	32	5	659.5000
254	200.66	33	31	555.2143
256	175.26	25	34	906.6364
266	177.80	40	42	693.4706
268	198.12	37	54	746.6500
271	195.58	39	53	492.0385
273	185.42	30	13	413.8571
274	190.50	26	3	899.0000
277	157.48	26	32	705.8462
278	185.42	27	8	777.3333
286	180.34	35	69	653.5517
290	185.42	32	44	660.5000
291	170.18	31	20	728.7500
293	167.64	34	35	739.3077
296	177.80	26	2	528.0000
297	187.96	30	12	822.5000

300	170.18	33	35	685.0000
302	167.64	31	64	1091.2941
305	165.10	31	18	510.2222
308	177.80	25	21	737.8750
313	185.42	28	18	581.6250
316	177.80	31	2	104.0000
317	172.72	33	33	654.5000
318	172.72	33	20	684.3750
319	165.10	29	33	778.1667
323	193.04	38	32	568.6429
324	198.12	32	19	874.5000
326	167.64	34	16	792.5000
328	203.20	29	7	645.3333
329	175.26	32	8	702.0000
330	180.34	22	16	569.1429
332	198.12	28	16	692.3333
337	160.02	36	6	900.0000
345	180.34	26	9	585.5000
350	185.42	30	48	744.8333
352	187.96	28	9	592.0000
356	185.42	35	55	527.0000
360	185.42	30	28	733.8182
363	160.02	24	9	900.0000
364	177.80	28	5	463.6667
365	172.72	29	9	900.0000
370	203.20	38	25	472.2308
374	190.50	38	61	784.6818
381	180.34	33	39	519.7368
385	185.42	28	9	900.0000
388	180.34	25	18	724.8571
392	165.10	36	7	600.6667
394	182.88	40	73	745.7037
397	175.26	26	8	707.6667
399	190.50	33	16	549.1250
402	177.80	29	15	900.0000
403	175.26	37	9	544.5000
404	182.88	22	2	415.0000
407	160.02	25	3	900.0000
411	203.20	40	10	389.8333
413	185.42	31	15	876.6000
416	195.58	36	6	511.3333
420	198.12	30	4	526.5000
422	180.34	28	24	742.8889
426	200.66	28	25	674.5000

432	187.96	37	47	538.0000
436	190.50	31	3	900.0000
441	182.88	28	11	772.2500
442	185.42	27	5	716.0000
444	162.56	32	18	880.5000
449	172.72	24	5	614.0000
450	187.96	26	22	773.6250
458	193.04	34	36	499.4444
462	177.80	32	13	774.6000
463	157.48	26	29	689.6667
468	170.18	26	10	697.5000
470	170.18	33	32	679.8462
477	167.64	31	1	236.0000
481	182.88	43	39	665.8667
482	190.50	24	1	275.0000
484	200.66	26	5	465.6667
485	172.72	33	11	612.8000
487	170.18	32	23	661.9000
489	175.26	26	15	681.0000
490	182.88	29	6	900.0000
492	170.18	36	25	787.5556
498	180.34	37	37	626.6875
522	180.34	33	43	702.2353
524	172.72	30	3	894.0000
525	185.42	30	12	883.2500
526	152.40	28	9	900.0000
527	175.26	23	3	686.0000
532	190.50	33	38	817.2308
534	187.96	24	7	690.0000
542	187.96	36	19	777.8571

	Actual_Value	Predicted_Value
3	14	9.21381664
5	3	2.86585879
6	4	2.14898046
8	5	5.70646758
10	3	1.17742806
15	4	3.27795971
17	0	1.34904229
19	4	4.17082164
20	1	1.98886925
26	6	7.36540580
28	2	1.80403643
29	6	5.51670423
33	2	2.29349337

34	6	6.62185220
36	6	6.01603310
37	12	19.12084126
39	9	7.02664934
40	10	11.69616308
41	2	0.49987782
44	0	1.34037557
46	9	6.39925395
49	11	10.01635351
56	0	2.41239136
57	1	2.35554924
58	1	2.00006562
63	3	4.47872077
65	11	11.72956635
69	1	1.89862710
72	4	5.53643259
73	4	3.68452738
78	5	4.10010226
83	8	7.31458047
93	5	4.34190639
94	2	3.25502112
99	12	10.36238320
101	10	11.27243697
111	9	8.44356305
116	17	19.21346948
119	6	7.12119356
124	0	0.38752433
135	1	0.43354185
137	5	4.75750233
140	4	3.22617902
143	1	0.58452432
144	1	2.44767056
149	9	10.36487265
154	3	2.16181655
156	5	2.96104715
158	4	4.87706676
159	10	9.68122313
161	9	9.95659557
162	5	4.82753979
163	16	15.13685592
165	7	11.30247338
168	16	15.61085669
177	5	4.42672394
180	2	1.74126534

190	7	6.80221473
194	6	5.29628319
195	3	3.44830532
197	9	12.23647433
198	8	11.21894607
200	14	13.88210423
204	17	17.99853161
205	6	5.42705235
206	12	9.67612621
212	0	1.35288643
213	2	0.37373696
223	6	6.13458235
226	2	1.89710575
230	10	9.07152591
233	9	6.77155475
234	4	5.47945913
235	1	-0.24372077
236	2	1.05784906
240	2	2.16479068
249	1	1.56711057
254	11	9.08609649
256	8	8.40116139
266	12	10.80788657
268	12	14.54013554
271	16	14.84887104
273	6	4.29518837
274	0	0.52941871
277	9	7.88145580
278	3	2.03202643
286	17	18.35701300
290	9	11.86616293
291	3	4.83626440
293	8	8.67291040
296	1	0.92011880
297	2	2.97244981
300	10	8.92242269
302	15	15.54996134
305	5	4.75747440
308	4	5.45739737
313	4	5.22465924
316	2	1.95779711
317	8	8.54389993
318	6	4.97877817
319	7	8.08941030

323	10	8.96503320
324	3	4.94348810
326	4	3.43604651
328	3	2.58701891
329	1	1.81497899
330	5	4.73631147
332	4	4.74871961
337	0	0.18804442
345	1	2.70977359
350	8	12.75803099
352	2	2.85803495
356	20	15.10193164
360	5	7.43049415
363	2	1.31006559
364	1	1.84832639
365	1	1.54285729
370	8	7.64821197
374	14	16.06465510
381	10	10.74251007
385	1	1.93481402
388	4	4.76272374
392	1	1.42832668
394	19	19.11469350
397	2	1.95850301
399	6	4.79210516
402	4	3.29641117
403	1	2.38496667
404	0	1.48430073
407	1	-0.32384411
411	4	3.80413314
413	4	3.52720188
416	3	2.28403577
420	2	1.93862988
422	6	6.24075257
426	6	7.28210193
432	12	12.94842426
436	1	0.39402100
441	3	2.75013508
442	2	1.39762631
444	4	3.63582816
449	1	1.39331784
450	6	5.89243389
458	10	10.33373135
462	4	3.02716101

```

463      8      7.12242298
468      3      2.37618721
470      9      8.13295713
477      0      1.03321309
481      8     10.14697641
482      1      1.76903126
484      2      2.55363366
485      2      2.76526469
487      6      5.79794595
489      3      3.90738700
490      1      1.03150957
492      7      5.88079818
498      9      9.80540304
522     13     11.31036983
524      0     -0.07449134
525      2      2.73168310
526      2      0.98469803
527      0      0.75840425
532      7      9.94581998
534      2      2.15785338
542      5      4.81178627

```

```

[ reached 'max' / getOption("max.print") -- omitted 952 rows ]
>
>
> #We now repeat the same with B
> modelB <- lm(B_wins ~ B_Height_cms + B_Reach_cms + B_Weight_lbs + B
_age + B_total_rounds_fought + B_total_time_fought.seconds. ,data = d
f)
> summary(modelB)

```

Call:

```
lm(formula = B_wins ~ B_Height_cms + B_Reach_cms + B_Weight_lbs +
    B_age + B_total_rounds_fought + B_total_time_fought.seconds.,
    data = df)
```

Residuals:

```

      Min       1Q   Median       3Q      Max
-5.7887 -0.7017 -0.0318  0.7241  5.7063

```

Coefficients:

```

              Estimate
(Intercept)    0.0427642
B_Height_cms    0.0006284
B_Reach_cms     0.0085128

```

```

B_Weight_lbs          0.0041488
B_age                 -0.0317970
B_total_rounds_fought 0.2743816
B_total_time_fought.seconds. -0.0023744
Std. Error
(Intercept)          0.6512501
B_Height_cms         0.0057748
B_Reach_cms          0.0045366
B_Weight_lbs         0.0010759
B_age                0.0057644
B_total_rounds_fought 0.0019104
B_total_time_fought.seconds. 0.0001005
t value Pr(>|t|)
(Intercept)          0.066 0.947649
B_Height_cms         0.109 0.913358
B_Reach_cms          1.876 0.060680
B_Weight_lbs         3.856 0.000117
B_age                -5.516 3.73e-08
B_total_rounds_fought 143.623 < 2e-16
B_total_time_fought.seconds. -23.616 < 2e-16

```

```

(Intercept)
B_Height_cms
B_Reach_cms      .
B_Weight_lbs     ***
B_age            ***
B_total_rounds_fought ***
B_total_time_fought.seconds. ***
---

```

Signif. codes:

0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 1.189 on 3348 degrees of freedom

Multiple R-squared: 0.8785, Adjusted R-squared: 0.8783

F-statistic: 4035 on 6 and 3348 DF, p-value: < 2.2e-16

> summary(modelB)\$coefficient

```

Estimate
(Intercept) 0.0427642265
B_Height_cms 0.0006283718
B_Reach_cms 0.0085127503
B_Weight_lbs 0.0041488348
B_age -0.0317970175
B_total_rounds_fought 0.2743815706

```

```
B_total_time_fought.seconds. -0.0023744479
```

```
Std. Error
```

```
(Intercept) 0.6512501322
```

```
B_Height_cms 0.0057747983
```

```
B_Reach_cms 0.0045366119
```

```
B_Weight_lbs 0.0010759183
```

```
B_age 0.0057644288
```

```
B_total_rounds_fought 0.0019104324
```

```
B_total_time_fought.seconds. 0.0001005447
```

```
t value
```

```
(Intercept) 0.06566483
```

```
B_Height_cms 0.10881276
```

```
B_Reach_cms 1.87645548
```

```
B_Weight_lbs 3.85608725
```

```
B_age -5.51607433
```

```
B_total_rounds_fought 143.62275475
```

```
B_total_time_fought.seconds. -23.61583478
```

```
Pr(>|t|)
```

```
(Intercept) 9.476486e-01
```

```
B_Height_cms 9.133575e-01
```

```
B_Reach_cms 6.067968e-02
```

```
B_Weight_lbs 1.173829e-04
```

```
B_age 3.729752e-08
```

```
B_total_rounds_fought 0.000000e+00
```

```
B_total_time_fought.seconds. 3.522706e-114
```

```
>
```

```
> #In B_wins, Height has a T value close to zero, hence we can remove  
the height
```

```
> modelB <- lm(B_wins ~ B_Reach_cms + B_Weight_lbs + B_age + B_total_  
rounds_fought + B_total_time_fought.seconds. ,data = df)
```

```
> summary(modelB)
```

```
Call:
```

```
lm(formula = B_wins ~ B_Reach_cms + B_Weight_lbs + B_age + B_total_rou  
nds_fought +  
    B_total_time_fought.seconds., data = df)
```

```
Residuals:
```

```
    Min      1Q  Median      3Q     Max  
-5.7880 -0.7020 -0.0320  0.7239  5.7072
```

```
Coefficients:
```

```
                Estimate  
(Intercept) 0.0841889
```

```

B_Reach_cms          0.0088642
B_Weight_lbs         0.0041952
B_age                -0.0318252
B_total_rounds_fought 0.2743800
B_total_time_fought.seconds. -0.0023751
                        Std. Error
(Intercept)          0.5283127
B_Reach_cms           0.0031853
B_Weight_lbs          0.0009879
B_age                 0.0057578
B_total_rounds_fought 0.0019101
B_total_time_fought.seconds. 0.0001004
                        t value Pr(>|t|)
(Intercept)           0.159  0.87340
B_Reach_cms            2.783  0.00542
B_Weight_lbs           4.246 2.23e-05
B_age                 -5.527 3.50e-08
B_total_rounds_fought 143.647 < 2e-16
B_total_time_fought.seconds. -23.664 < 2e-16

```

```

(Intercept)
B_Reach_cms          **
B_Weight_lbs         ***
B_age                ***
B_total_rounds_fought ***
B_total_time_fought.seconds. ***
---
Signif. codes:
0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

```

Residual standard error: 1.188 on 3349 degrees of freedom
Multiple R-squared:  0.8785,    Adjusted R-squared:  0.8783
F-statistic: 4844 on 5 and 3349 DF, p-value: < 2.2e-16

```

```

>
> sigma(modelB)/mean(df$B_wins)
[1] 0.3349714
>
> #We get a 0.33 which means it has a 33% error rate which is not very
efficient to solve models.
>
> #Let us now test the created prediction model
> #Splitting the data set into training and testing data
> split = sample.split(df$B_wins,SplitRatio = 2/3)

```

```
> #Training is 2 parts
> training_set = subset(df, split == TRUE)
> #testing is 1 part
> test_set = subset(df, split == FALSE)
>
> #Do a prediction on the model
> B_pred = predict(modelB, newdata = test_set)
> #Store and compare the result in the test cases
> resultB <- data.frame(Reach = test_set$B_Reach_cms, Weight = test_set$B_Weight_lbs, Age = test_set$B_age, Total_rounds = test_set$B_total_rounds_fought, Total_seconds = test_set$B_total_time_fought.seconds., Actual_Value = test_set$B_wins, Predicted_Value = B_pred)
> resultB
```

	Reach	Weight	Age	Total_rounds
1	170.18	135	31	9
2	167.64	125	32	29
4	170.18	135	26	9
5	185.42	250	32	8
7	165.10	135	32	23
8	167.64	115	25	10
9	182.88	145	31	10
14	198.12	205	27	7
21	195.58	170	26	36
22	182.88	185	30	3
27	187.96	185	26	5
32	180.34	170	26	3
33	162.56	135	35	21
37	177.80	155	37	67
41	172.72	135	28	9
42	170.18	135	31	13
45	203.20	265	27	3
46	195.58	185	30	13
48	180.34	170	27	20
53	162.56	115	32	6
59	193.04	235	37	15
68	182.88	170	30	22
69	177.80	135	30	14
71	177.80	155	30	10
73	185.42	145	32	50
74	157.48	115	33	15
75	182.88	145	28	9
77	185.42	155	29	15
79	195.58	185	31	9
84	177.80	145	28	6

87	162.56	125	30	3
88	162.56	115	34	22
89	187.96	170	34	41
90	182.88	170	30	23
94	190.50	185	33	4
95	182.88	145	31	8
98	165.10	125	30	2
102	198.12	265	37	23
111	193.04	170	31	28
112	165.10	135	32	22
123	170.18	135	26	6
125	172.72	145	26	1
126	160.02	125	23	4
127	180.34	155	34	9
130	177.80	155	34	27
132	182.88	170	29	19
136	175.26	125	30	3
140	195.58	185	43	55
142	167.64	125	23	3
143	190.50	205	32	34
145	182.88	145	27	8
147	177.80	135	32	3
148	170.18	135	30	8
149	182.88	145	29	15
155	170.18	125	32	42
156	190.50	155	31	21
157	165.10	125	30	34
161	182.88	155	28	2
173	185.42	170	34	14
175	177.80	155	31	28
177	175.26	135	25	37
178	187.96	155	29	39
181	195.58	185	30	11
182	190.50	185	34	19
189	162.56	115	33	15
190	180.34	155	32	19
191	167.64	125	32	26
192	190.50	185	31	15
193	175.26	135	31	12
194	187.96	170	29	11
197	200.66	205	27	7
198	198.12	264	31	7
201	165.10	125	30	11
206	165.10	135	26	6

214	190.50	185	33	1
215	185.42	170	32	21
220	180.34	125	29	15
221	180.34	145	26	20
226	182.88	155	32	3
227	172.72	125	27	12
228	172.72	135	27	9
229	200.66	260	33	34
231	195.58	185	34	15
232	185.42	185	28	8
238	180.34	155	26	12
241	193.04	205	30	19
243	193.04	205	37	25
249	190.50	249	26	4
250	185.42	155	26	6
251	187.96	155	30	20
258	187.96	155	35	15
259	172.72	135	28	1
262	190.50	185	31	12
267	195.58	170	35	30
268	180.34	185	40	26
269	203.20	240	41	11
272	190.50	155	34	16
274	187.96	205	30	1
282	162.56	135	28	9
287	170.18	125	24	9
293	172.72	125	30	5
295	182.88	170	27	20
296	180.34	135	38	38
300	170.18	135	32	8
303	182.88	145	29	14
307	165.10	125	26	2
309	182.88	155	29	43
311	152.40	115	28	23
313	182.88	170	27	12
318	172.72	125	29	12
320	162.56	115	32	12
324	213.36	265	30	38
326	177.80	155	31	7
327	190.50	170	32	6
335	170.18	135	28	18
340	180.34	170	26	16
343	182.88	185	27	5
346	193.04	170	32	10

349	203.20	185	28	5
356	187.96	170	26	25
358	167.64	125	31	23
362	185.42	170	23	21
364	180.34	125	21	3
366	167.64	125	30	3
367	185.42	185	41	30
374	193.04	205	29	17
375	175.26	135	24	31
377	180.34	145	26	3
379	182.88	155	35	11
381	195.58	170	35	29
389	187.96	170	36	20
393	190.50	185	37	23
395	167.64	115	27	4
399	170.18	170	36	28
402	182.88	135	28	6
403	167.64	135	37	21
404	195.58	155	34	2
405	170.18	135	29	21
407	162.56	115	26	7
409	187.96	185	39	61
412	167.64	155	34	15
414	187.96	170	33	4
	Total_seconds	Actual_Value	Predicted_Value	
1	419.4000	4	2.64578048	
2	849.0000	4	7.01675887	
4	652.0000	4	2.25246512	
5	1200.0000	1	1.10313008	
7	495.2500	8	6.23009638	
8	716.0000	2	2.30024746	
9	670.7500	3	2.47771361	
14	681.6667	3	2.14274681	
21	724.8571	10	9.85966630	
22	900.0000	1	0.21220050	
27	610.5000	2	1.62087420	
32	900.0000	1	0.25405870	
33	702.3750	4	5.07140933	
37	702.1538	14	17.84875885	
41	900.0000	1	1.62231243	
42	754.2000	2	2.94812710	
45	658.0000	1	1.39817682	
46	378.0000	6	4.30836258	
48	603.1111	5	5.59182576	

53	900.0000	0	0.49790804
59	700.5000	4	4.05562946
68	792.2500	5	5.61840719
69	793.2000	2	3.22924977
71	611.7500	2	2.64658947
73	684.4500	11	13.41106584
74	687.6667	4	3.39477946
75	890.0000	2	1.77807505
77	900.0000	2	3.43324601
79	344.6667	4	3.25818631
84	900.0000	1	0.88615416
87	900.0000	0	-0.21962998
88	808.3750	3	5.04195379
89	797.7333	9	10.73633365
90	548.3636	8	6.47203441
94	541.5000	0	1.31011275
95	551.7500	2	2.21158691
98	565.0000	0	0.32415353
102	558.4000	6	6.75905219
111	894.1111	9	7.08099510
112	513.7273	7	5.91183155
123	569.3333	3	1.62566419
125	251.0000	0	1.07429478
126	284.0000	2	1.71805446
127	410.6000	4	2.74516910
130	688.8182	7	7.00070673
132	655.7500	5	5.15128938
136	900.0000	1	-0.10705467
140	639.9545	17	14.79643559
142	900.0000	1	0.04817647
143	570.0000	10	9.58955238
145	776.0000	1	1.80627820
147	900.0000	0	-0.10623832
148	475.0000	3	2.27117175
149	707.3333	5	3.82637611
155	745.8000	12	10.85132143
156	413.0000	6	6.21756499
157	735.9231	8	8.69835990
161	314.0000	0	1.26740692
173	607.1667	4	3.75816722
175	671.0000	8	7.41288175
177	906.0833	8	9.40849529
178	418.2381	12	11.18509983
181	451.6667	4	3.58463905

182	588.1111	6	5.28328322
189	883.8000	3	2.97397917
190	583.3333	6	5.14236591
191	843.3333	3	6.20707754
192	871.2000	3	3.60888276
193	633.6000	2	3.00521066
194	406.0000	1	3.59445309
197	361.0000	3	2.92686767
198	651.0000	3	2.33579635
201	447.8333	4	3.07185276
206	513.3333	3	1.71363799
214	238.0000	0	1.20780646
215	566.9000	8	5.83811389
220	466.0000	4	4.29314130
221	794.1429	6	5.06505802
226	900.0000	1	0.02269513
227	900.0000	1	2.43532602
228	619.5000	1	2.32034477
229	577.4667	12	9.86078773
231	563.4286	3	4.28941602
232	532.7500	2	2.54251052
238	648.4000	1	3.25811901
241	501.0000	6	5.72389735
243	602.7273	6	6.90579175
249	514.0000	1	1.86669414
250	900.0000	1	1.05930139
251	487.8000	9	5.77483986
258	704.0000	5	3.73032367
259	205.0000	0	1.07794596
262	864.0000	3	2.80284320
267	492.0000	9	8.48001151
268	628.3636	3	6.82732941
269	349.1429	5	3.77634254
272	725.5000	4	4.00797993
274	263.0000	0	1.30529353
282	900.0000	3	1.53225219
287	900.0000	2	1.68514646
293	737.0000	2	0.80632673
295	682.7500	6	5.42519289
296	593.7059	10	10.05609631
300	494.2500	2	2.16180127
303	797.8000	4	3.33713142
307	475.0000	1	0.66521059
309	540.6500	15	11.94685294

311	831.6250	6	5.36200431
313	686.4000	3	3.22148372
318	900.0000	3	2.37167564
320	879.7500	2	2.19228334
324	450.7500	12	11.48829548
326	577.0000	1	1.87415791
327	900.0000	2	0.97630788
335	619.3750	6	4.73572159
340	606.0000	4	4.51926956
343	275.3333	3	2.34006320
346	648.0000	2	2.69486067
349	708.5000	2	1.45955733
356	787.5556	7	6.62502776
358	836.2500	2	5.43258608
362	659.1111	6	5.90553274
364	900.0000	1	0.22440216
366	900.0000	0	-0.17459985
367	758.2000	9	7.62968397
374	623.8571	4	4.91516818
375	907.3000	7	7.79115069
377	900.0000	0	0.14917954
379	610.4000	4	2.81008002
381	518.8571	8	8.14184389
389	688.0000	6	5.17132720
393	708.4444	5	5.99952764
395	561.5000	2	0.95726528
399	798.8000	8	6.94560416
402	900.0000	1	0.88923262
403	873.4286	4	4.64652486
404	78.5000	1	1.74836010
405	787.1250	6	5.12861846
407	474.0000	2	1.97501904
409	729.7826	15	16.28912334
412	900.0000	4	3.11651463
414	531.0000	1	1.24960843

[reached 'max' / getOption("max.print") -- omitted 974 rows]

Conclusion:

Thus, Regression analysis is performed on the dataset.

Anirudh Poroorkara

Roll: 44

IT-1 B12

Lab Outcome: LO6

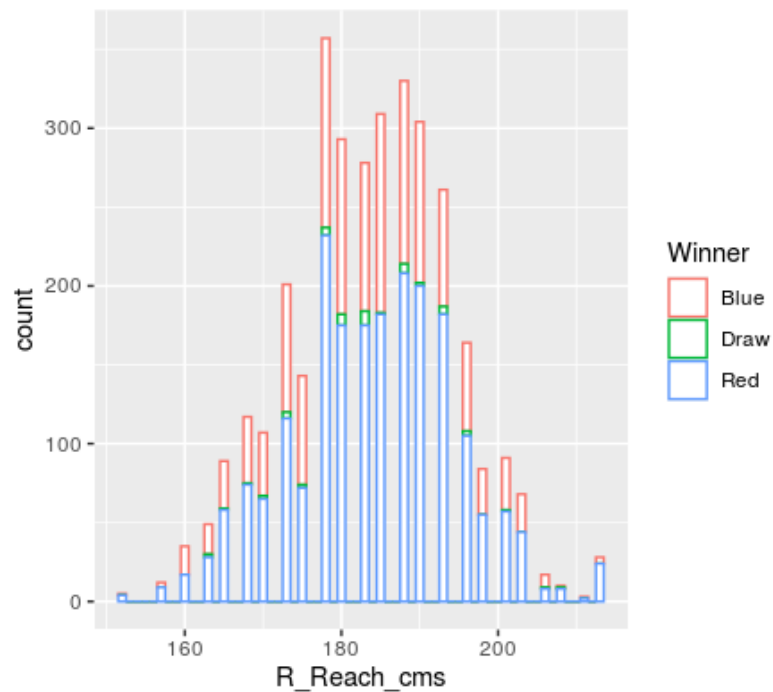
Lab Assignment: 11

Aim:

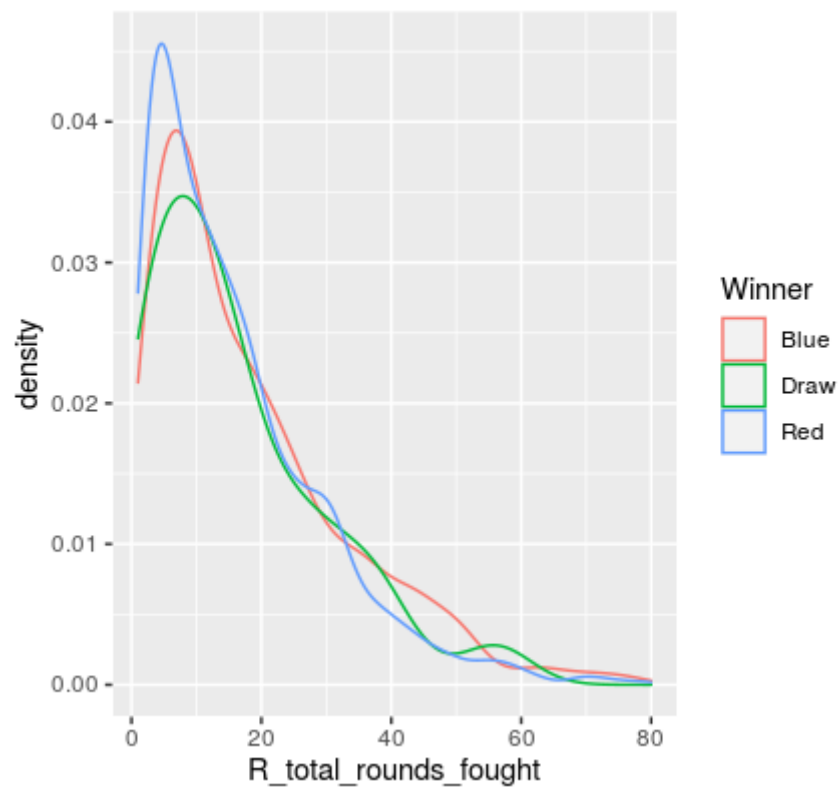
Data Visualization using ggplot2

Code:

```
> #Lab Assignment 11
> #Mini-Project Session 5
>
> #Data Visualization using ggplot2
>
> #get the working directory
> getwd()
[1] "/cloud/project"
>
> #Loading dataset into R
> #Header is set to true if file contains Header information
> #Sep stands for separator which is , in our csv file
> df <- read.csv("df_clean.csv", header = TRUE, sep = ",")
>
> #ggplot2 is a robust and a versatile R package for generating aesthetic plots and charts.
> #Plot = data + Aesthetics + Geometry
> #1. Data refers to a data frame
> #2. Aesthetics indicates x and y variables. It is also used to tell R how data are displayed in a plot, e.g. color, size and shape of points etc.
> #3. Geometry refers to the type of graphics
>
> #install the package if not installed
> #install.packages("ggplot2")
> library(ggplot2)
>
> #Plotting with R_Reach and winner using a group histogram
> ggplot(data = df, aes( x = R_Reach_cms, color=Winner)) + geom_histogram(fill="white", binwidth = 1)
```

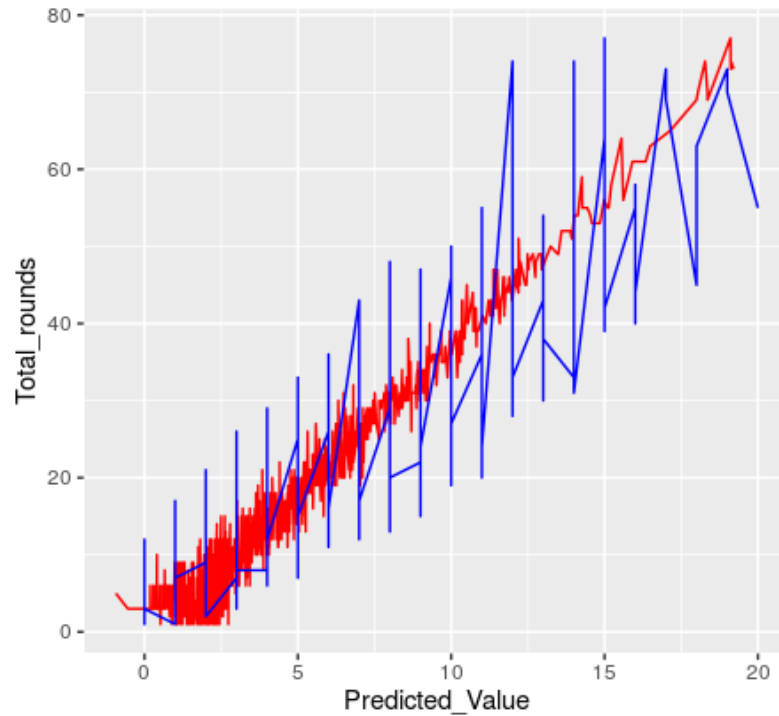


```
> #Plotting R_total_rounds_fought with winner using density graph
> ggplot(df, aes( x = R_total_rounds_fought, color=Winner)) + geom_density( )
```

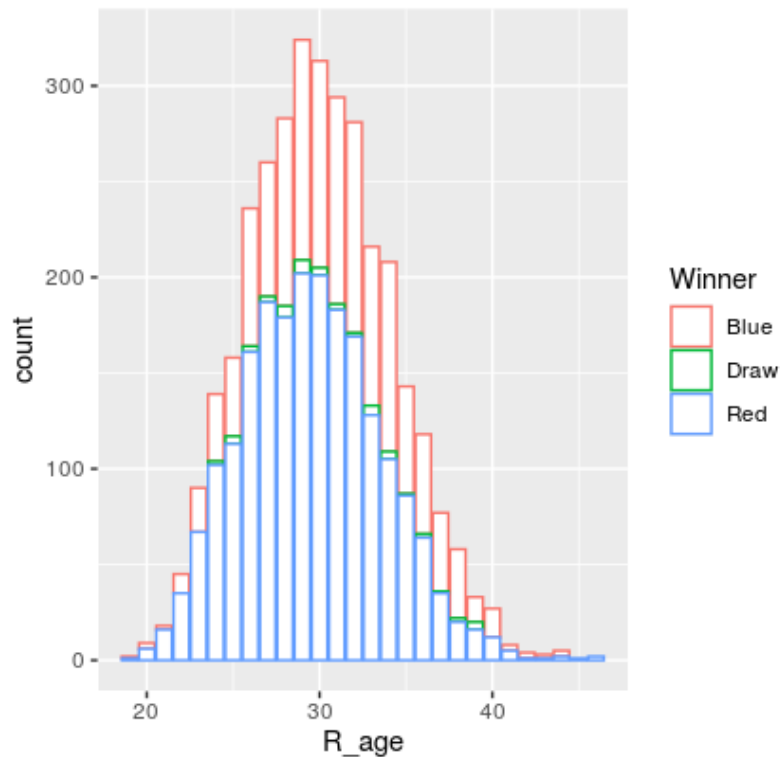


```
> #These two features were chosen as they had a significant hand in determining
the chances of R winning
>
```

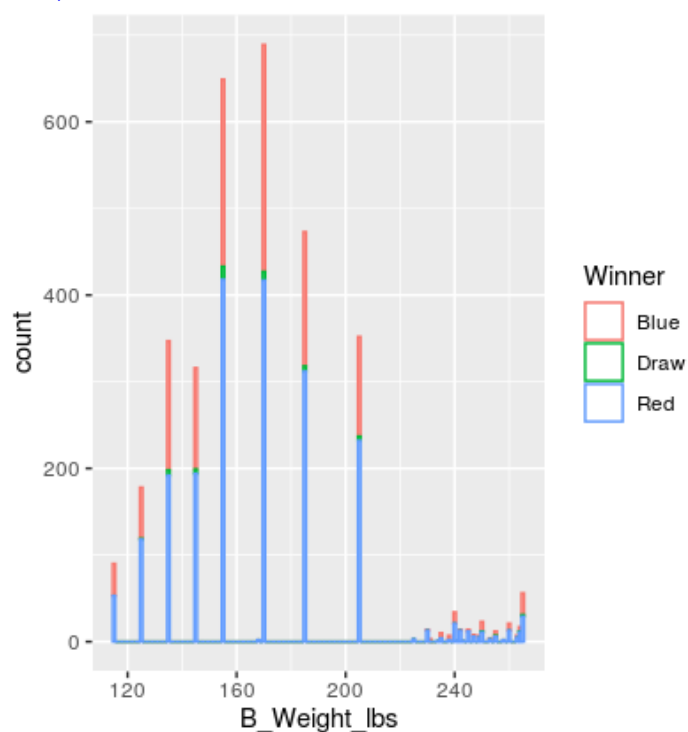
```
> #Plotting regression model results of model R
> ggplot(data = resultR) + geom_line(aes(x=Predicted_Value, y=Total_rounds), color='red') + geom_line(aes(x=Actual_Value, y=Total_rounds), color='blue')
```



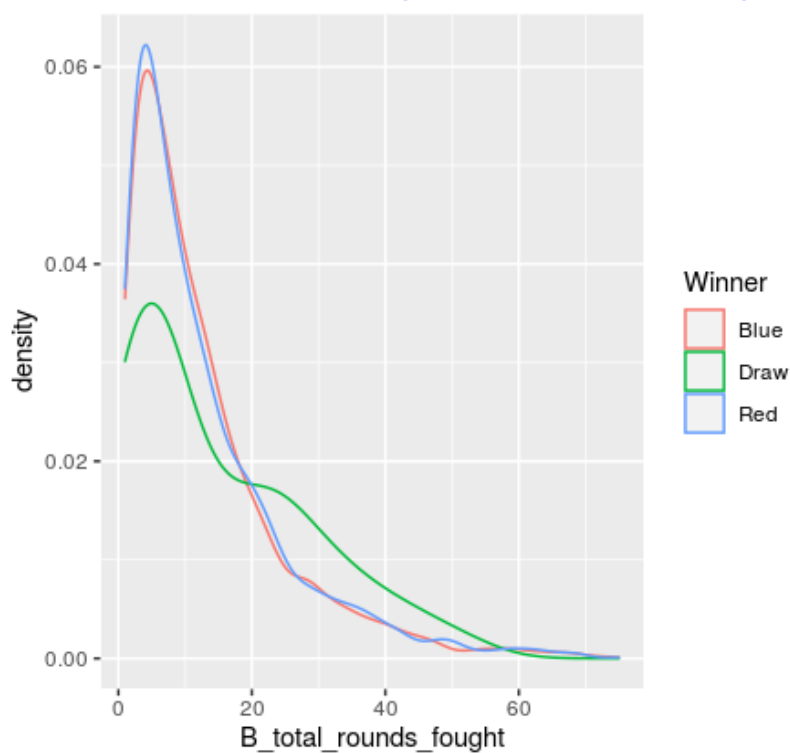
```
> #Plotting age of fighter to get an idea of the general age range of fighters
> ggplot(df, aes(x= R_age, color = Winner)) + geom_bar(fill='white')
```



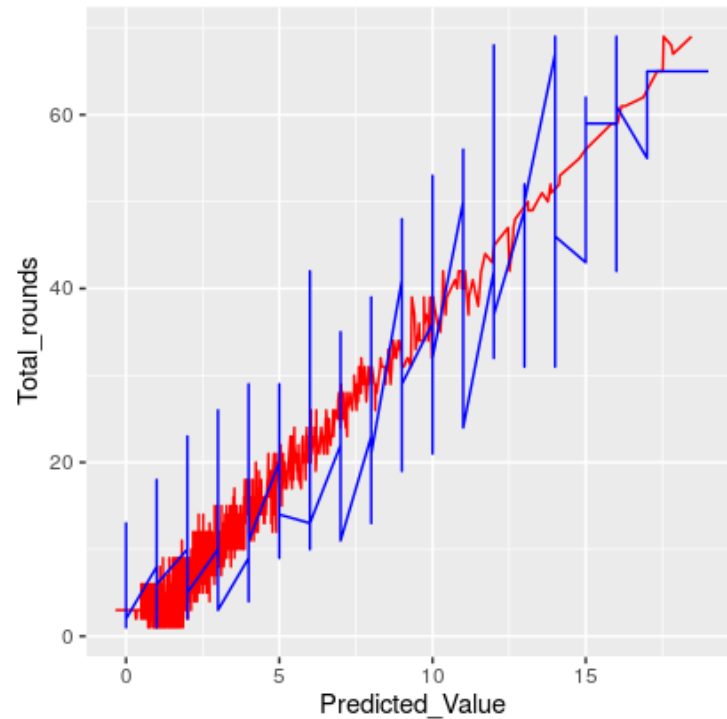

```
> #Plotting with B_weight and winner using a group histogram  
> ggplot(data = df, aes( x = B_Weight_lbs, color=Winner)) + geom_histogram(fill=  
"white", binwidth = 1)
```



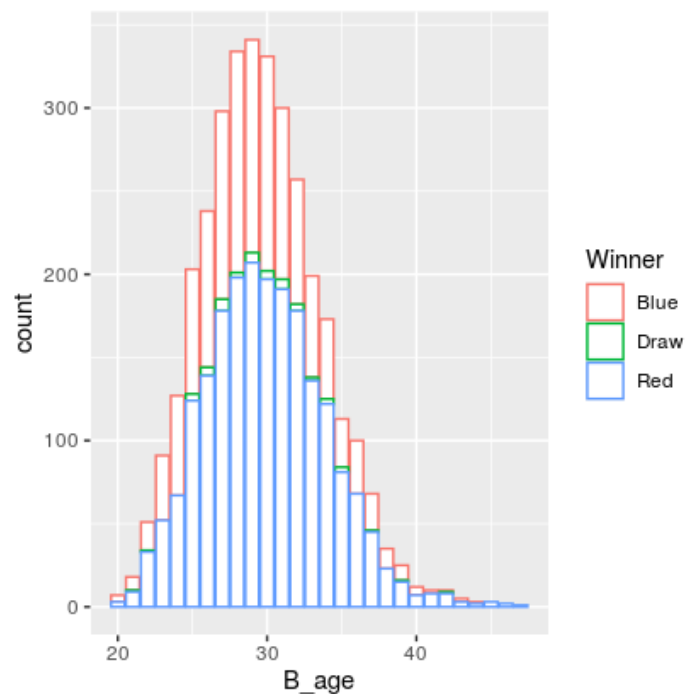
```
> #Plotting B_total_rounds_fought with winner using density graph  
> ggplot(df, aes( x = B_total_rounds_fought, color=Winner)) + geom_density( )
```



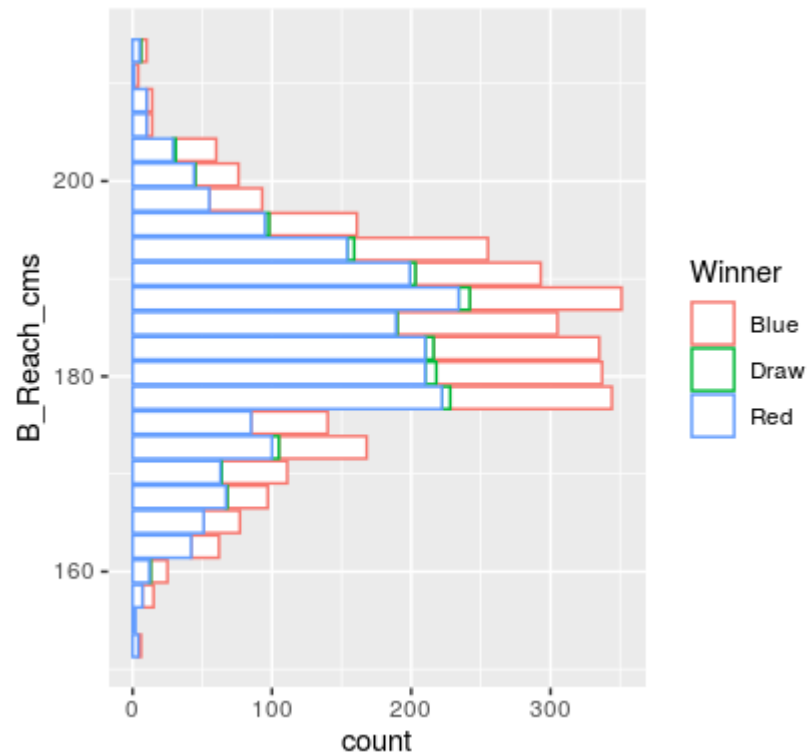
```
> #These two features were choosen as they had a significant hand in determining  
the chances of B winning  
> #Plotting regression model results of model B  
> ggplot(data = resultB) + geom_line(aes(x=Predicted_Value, y=Total_rounds), colo  
r='red') + geom_line(aes(x=Actual_Value, y=Total_rounds), color='blue')
```



```
> #Plotting age of fighter to get an idea of the general age range of fighters  
> ggplot(df, aes(x= B_age, color = Winner)) + geom_bar(fill='white')
```



```
> #Plotting reach of B fighter  
> ggplot(df, aes(x= B_Reach_cms , color = Winner)) + geom_bar(fill='white') + coord_flip()
```



Conclusion:

Thus, data is visualized using ggplot2.

Lab Outcome: LO6