

# SHARDA SCHOOL OF BASIC SCIENCES AND RESEARCH

# **Department of Mathematics**

# LAB REPORT FILE

**Course Title: Data Cleaning lab** 

Course Code: BDA 253

Program-Bachelor of Science (Hons.)-Data Science

Semester: 4th

Session 2022-2023

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**Submitted to** 

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(Assistant Professor)

## **❖** Aim:

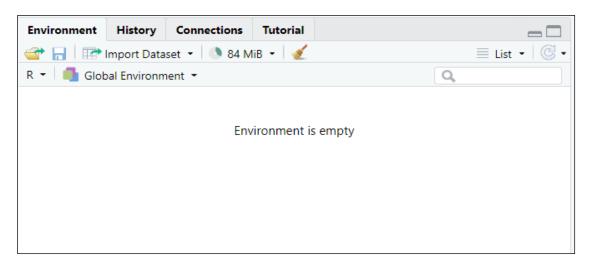
How to import data from different sources in R.

# **\*** Theory:

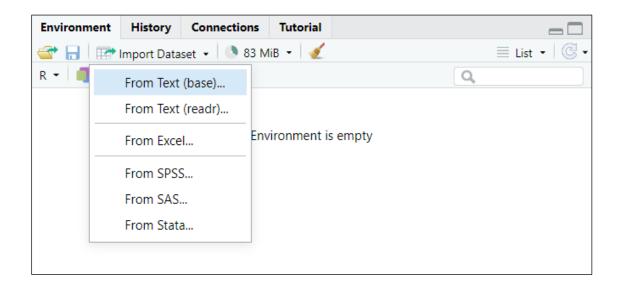
R is capable of reading data from most formats, including files created in other statistical packages. Whether the data was prepared using Excel (in CSV, XLSX, or TXT format), SAS, Stata, SPSS, or others, R can read and load the data into memory.

# **Steps of importing a csv file:**

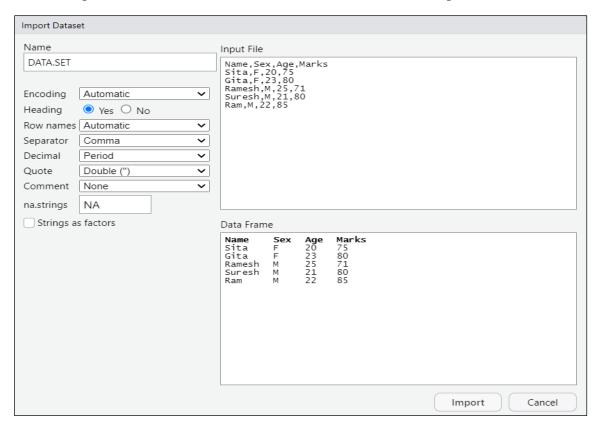
1. Click on the 'import' button in the right window of R studio.



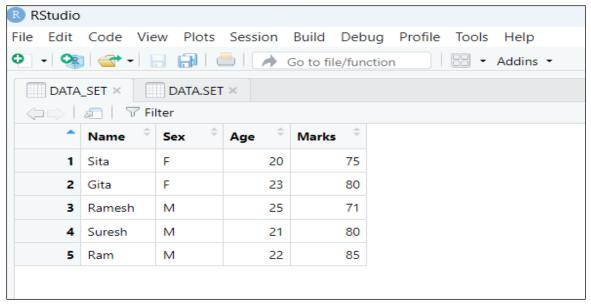
2. A pop-down menu will open. Choose the first option (From text(base))



3. This will open another window enabling you to browse your computer to locate the file you want to import. Double click on the file, and another window will open as show below.



4. Click on the **Import** button at the bottom of this window and the CSV file is successfully imported in the R studio as follows.

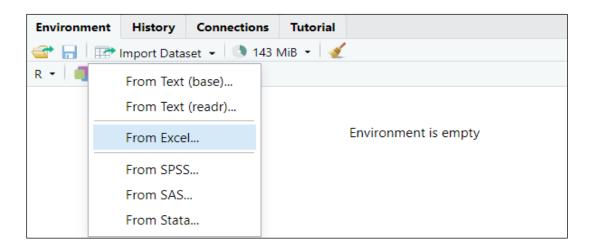


# **Steps of importing a Excel file:**

I. Click on the 'import' button in the right window of R studio



II. A pop-down menu will open. Choose the first option From Excel.

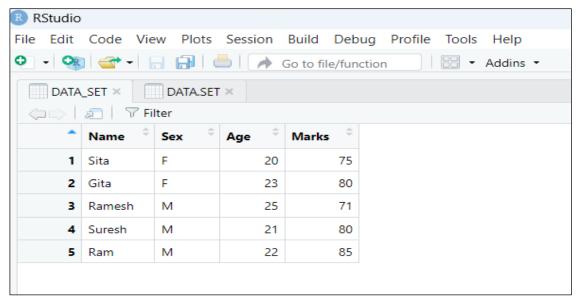


III. This will open another window enabling you to browse your computer to locate the file you want to import. Double click on the file, and another window will open as show below.



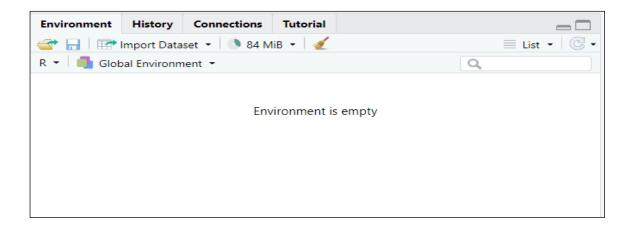
IV. Click on the **Import** button at the bottom of this window and the EXCEL file is successfully imported in the R studio as follows.

## **\*** OUTPUT:

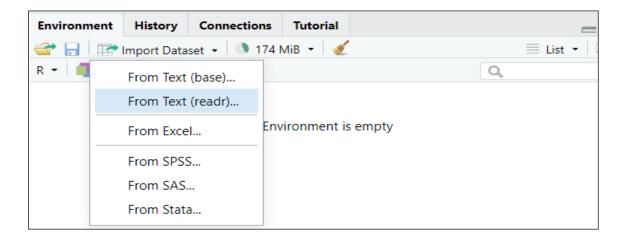


# **Steps of importing a TEXT file:**

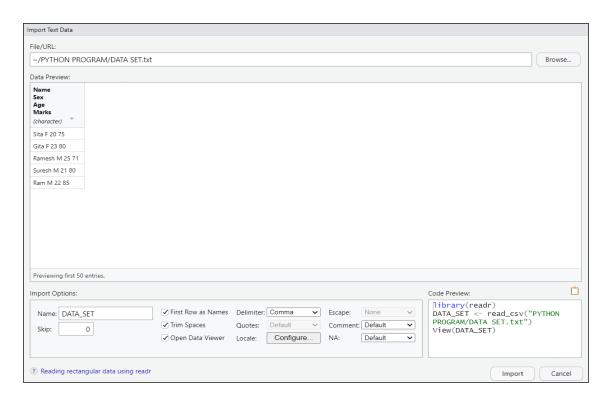
1) Click on the 'import' button in the right window of R studio.



2) A pop-down menu will open. Choose the first option (From text(readr)).

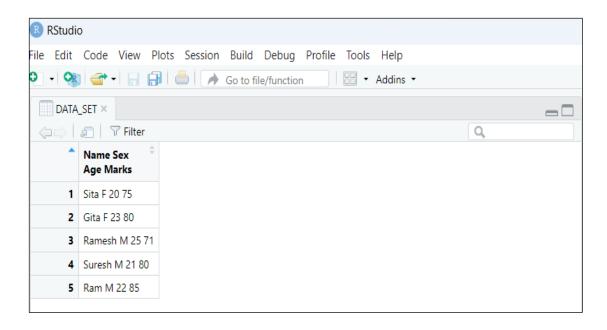


3) This will open another window enabling you to browse your computer to locate the file you want to import. Double click on the file, and another window will open as show below.



4) Click on the **Import** button at the bottom of this window and the CSV file is successfully imported in the R studio as follows.

# **❖** Output:



## **❖** Aim:

How to enter matrix through different command in R with their use.

# **\*** Theory:

A matrix is a 2-D data set with columns and rows. A column is a vertical representation of data, while a row is a horizontal representation of data.

# **Syntax**

The basic syntax for creating a matrix in R is –

Matrix (data, nrow, ncol, byrow, dimnames)

# **\*** Creating a Matrix in R:

To create a matrix in R you need to use the function called **matrix** ().

# **Output:**

```
Background Jobs ×
Console Terminal ×
> Data=matrix(c(25,26,22,24,25,21,32,15,31,42,32,35,16,32,31,36,26,18,33,26,85,39,17,36,41),
            nrow=5,byrow=TRUE)
> Data
    [,1] [,2] [,3] [,4] [,5]
[1,] 25 26 22 24
     21
          32 15
                    31
                        42
[2,]
[3,]
              16
18
           35
                    32
      32
                        31
[4,]
      36
           26
                    33
                        26
          39 17
      85
                    36
[5,]
                        41
```

# **Concatenation of a Row wise:**

> The concatenation of a row to a matrix is done using **rbind** ().

# \* Output:

```
Background Jobs
Console Terminal ×
R 4.2.2 · ~/ ≈
> r1=c(25,26,22,24,25)
> r2=c(21,32,15,31,42)
> r3=c(32,35,16,32,31)
> r4=c(36,26,18,33,26)
> r5=c(85,39,17,36,41)
> rbind= rbind(r1,r2,r3,r4,r5)
> rbind
  [,1] [,2] [,3] [,4] [,5]
r1 25
         26
              22
                    24
r2 21
         32
              15
                    31
                        42
   32
         35
              16
                    32
                         31
r4
    36
         26
              18
                    33
                         26
              17
```

# **Concatenation of a column wise:**

> The concatenation of a column to a matrix is done using **cbind()** 

# **Output:**

## **Create a inverse of Matrix in R:**

Using the solve () function:

```
Background Jobs X
Console Terminal ×
R 4.2.2 · ~/ ≈
> Data=matrix(c(25,26,22,24,25,21,32,15,31,42,32,35,16,32,31,36,26,18,33,26,85,39,17,36,41),
             nrow=5,byrow=TRUE)
> Data
    [,1] [,2] [,3] [,4] [,5]
[1,]
      25
           26
               22
                     24
                          25
           32 15
[2,]
      21
                     31
                          42
[3,]
           35 16
                     32
      32
                          31
[4,]
      36
           26 18
                     33
                          26
[5,]
      85
           39
               17
                     36
                          41
> solve(Data)
            [,1]
                         [,2]
                                       [,3]
[1,] 0.001238593 -0.009125311 -0.0138934602 1.006898e-03 0.018458931
[2,] 0.010981035 -0.044888596 0.1331179799 -9.072346e-02 -0.003830543
[3,] 0.098940180 -0.012156832 -0.0573775958 1.255916e-05 -0.004500942
[4,] -0.069420096 -0.001835369 -0.0003265381 1.113980e-01 -0.026186277
[5,] 0.006917065 0.068269417 -0.0737434779 -1.360763e-02 0.014624490
```

# **\*** Create a Multiplication of Matrix in R:

The multiplication operator\* is used for multiplying a matrix by scalar or element-wise multiplication of two matrices.

```
Terminal × Background Jobs ×
                                                                             -\Box
R 4.2.2 · ~/ ≈
> r1=c(25,26,22,24,25)
> r2=c(21,32,15,31,42)
> r3=c(32,35,16,32,31)
> r4=c(36,26,18,33,26)
> r5=c(85,39,17,36,41)
> rbind= rbind(r1,r2,r3,r4,r5)
> c1=c(25,21,32,34,85)
> c2=c(26,32,35,26,39)
> c3=c(22,15,16,18,17)
> c4=c(24,31,32,33,36)
> c5=c(25,42,31,26,41)
> cbind= cbind(c1,c2,c3,c4,c5)
> Multi=rbind*cbind
  Multi
   [,1] [,2] [,3] [,4] [,5]
r1 625 676 484 576 625
   441 1024
              225
                   961 1764
r2
r3 1024 1225
              256 1024
                         961
r4 1224 676 324 1089 676
r5 7225 1521 289 1296 1681
```

## **❖** Aim:

Create a dataframe and rearrange the columns.

- 1) A Column is Moved to the First Position
- 2) Move a Column to the Last Position
- 3) Rearrange Several Columns
- 4) Alphabetically reorder the columns
- 5) Reverse Column Order

# **\*** Theory:

Each vector will represent a DataFrame column, and the length of any vector will correspond to the number of rows in the new DataFrame. It's also possible to pass only one vector to the data. frame() function, in which case a DataFrame with a single column will be created.

# **A Rearranging Columns in R:**

Tips for Rearranging Columns in R, you might frequently want to reorder the columns in a data frame.

The **select()** function from the **dplyr package,** fortunately, makes this simple to accomplish. **library(dplyr)** 

#### Create a data frame:

```
Console
        Terminal ×
                   Background Jobs X
                                                                                 =\Box
R 4.2.2 · ~/ ≈
> df <- data.frame(player = c('P1', 'P2', 'P3', 'P4'),</pre>
                     points = c(124, 229, 313, 415),
                     result = c('Win', 'Loss', 'Win', 'Loss'))
> df
  player points result
                     Win
1
      Ρ1
             124
2
      Ρ2
             229
                    Loss
3
      Р3
             313
                     Win
      Ρ4
             415
                    Loss
```

#### 1. A Column is Moved to the First Position:

How to shift a particular column in a data frame to the front position is demonstrated by the code below.

#### Output:

```
Console Terminal ×
                Background Jobs X
> #A Column is Moved to the First Position
> select(df,player,points,result)
 player points result
1
           124
     Ρ1
2
     P2
           229
                 Loss
3
     Р3
           313
                  Win
4
     Ρ4
           415
                 Loss
```

#### 2. Move a Column to the Last Position:

How to relocate a particular column in a data frame to the last location is demonstrated by the code below:

#### **❖** OUTPUT:

```
Console Terminal × Background Jobs ×
R 4.2.2 · ~/ ≈
> #Move a Column to the Last Position
> select(df,points,result,player)
 points result player
1
    124
           Win
                    Р1
2
     229
                    P2
          Loss
3
     313
           Win
                    Р3
     415 Loss
                    Р4
```

## 3. Rearrange Several Columns:

The code that follows demonstrates how to sequentially reorder multiple columns.

#### **❖** OUTPUT:

```
Console Terminal × Background Jobs ×
> #Rearrange Several Columns
> select(df,points,result,player)
 points result player
    124
          Win
                  P1
    229
                  P2
         Loss
3
    313
          Win
                  Р3
    415 Loss
                  Р4
```

# 4. Alphabetically reorder the columns:

The code below demonstrates how to alphabetically arrange the columns. alphabetize the columns

## **❖** Output:

```
Console Terminal × Background Jobs ×
R 4.2.2 · ~/ ≈
> #alphabetize the columns
> df %>% select(order(colnames(.)))
 player points result
1
     P1
          124
     P2
           229 Loss
2
        313
3
     Р3
                Win
           415 Loss
     Ρ4
```

# 5. Reverse Column Order:

Reversing the column order in a data frame is demonstrated by the code below.

# **❖** Output:

```
Console Terminal × Background Jobs ×
R 4.2.2 · ~/ ≈
> #column order in reverse
> df %>% select(result:player, everything())
 result points player
   Win 124
                  Ρ1
2 Loss
           229
                   P2
3
           313
                   Р3
   Win
4 Loss
           415
                   Ρ4
>
```

# **PRACTICAL-4**

#### **\*** AIM:

Create a dataframe and remove the rows.

- 1) Remove any row with NA's.
- 2) Delete any rows that contain NA's in specific columns.
- 3) Rows that are duplicated should be removed.
- 4) Rows are removed based on their index position.
- 5) Rows are removed based on their condition.

# **Steps:**

# \* Remove any row with NA's

#### Code:

```
Console
         Terminal ×
                     Jobs ×
> source("~/.active-rstudio-document", echo=TRUE)
> df <- data.frame(player = c('P1', 'P2', 'P3', 'P4', 'P5', 'P6', 'P7'),
+ points = c(122, 144, 154, 155, 120, 218, 229),</pre>
            assists = .... [TRUNCATED]
+
> df
  player points assists
1
       Ρ1
               122
                          43
2
               144
                          55
       P2
3
       Р3
               154
                          77
4
       Р4
               155
                          18
5
       Р5
               120
                         114
6
       Р6
               218
                          NA
7
       Р7
               229
                          29
> df %>% na.omit()
  player points assists
1
                          43
       Р1
               122
2
               144
       P2
                          55
3
       Р3
               154
                          77
4
5
       Р4
               155
                          18
       Р5
               120
                         114
7
       Р7
               229
                          29
```

# ❖ Delete any rows that contain NA's in specific columns

## Code:

```
Console Terminal × Jobs ×
R 4.1.2 · ~/ ≈
> source("~/.active-rstudio-document", echo=TRUE)
> df <- data.frame(player = c('P1', 'P2', 'P3', 'P4', 'P5', 'P6', 'P7'),
+ points = c(122, 144, 154, 155, 120, 218, 229),</pre>
+
           assists = .... [TRUNCATED]
> df
  player points assists
1
      Ρ1
              122
                        43
2
       P2
              144
                        55
3
       Р3
              154
                        77
4
       Ρ4
              155
                        18
5
      Р5
              120
                       114
6
       Р6
              218
                        NΑ
7
      Ρ7
              229
                        29
> df %>%filter(!is.na(assists))
  player points assists
1
      Р1
                        43
              122
2
       P2
              144
                         55
3
                        77
       Р3
              154
4
5
       Р4
              155
                        18
       P5
              120
                       114
6
       Р7
              229
                        29
```

# \* Rows that are duplicated should be removed

## Code:

```
      Untitled4* x
      Doubled5 x

      Image: Control of the control of the
```

```
Console Terminal ×
                  Jobs ×
R 4.1.2 · ~/ ≈
> source("~/.active-rstudio-document", echo=TRUE)
> df <- data.frame(player = c('P1', 'P2', 'P3', 'P4', 'P5', 'P2', 'P6', 'P7'),</pre>
          points = c(122, 144, 154, 155, 120, 144, 218, 229),
           .... [TRUNCATED]
> df
  player points assists
1
      Ρ1
            122
                      43
2
      P2
             144
                      55
3
      Р3
             154
                      77
4
             155
      Ρ4
                      18
5
      Р5
            120
                     114
6
      P2
            144
                      55
7
             218
      Р6
                      NA
8
      Ρ7
             229
                      29
> df %>% distinct()
  player points assists
1
      Р1
            122
                      43
2
3
      P2
             144
                      55
      Р3
             154
                      77
4
      Ρ4
             155
                      18
5
      Р5
             120
                     114
6
      Р6
             218
                      NA
      Р7
                      29
             229
```

# \* Rows are removed based on their index position

## Code:

```
Console Terminal ×
                    Jobs ×
R 4.1.2 · ~/ ≈
> source("~/.active-rstudio-document", echo=TRUE)
> df <- data.frame(player = c('P1', 'P2', 'P3', 'P4', 'P5','P2', 'P6', 'P7'),
+ points = c(122, 144, 154, 155, 120, 144, 218, 229),</pre>
+
            .... [TRUNCATED]
> df
  player points assists
1
                      43
       Р1
              122
              144
                         55
2
       P2
3
              154
                        77
       Р3
4
              155
       Р4
                        18
5
       P5
              120
                       114
6
              144
       P2
                        55
7
       Р6
              218
                         NΑ
8
       Р7
              229
                         29
> df %>% filter(!row_number() %in% c(1, 2, 4))
  player points assists
1
       Р3
              154
                        77
2
                       114
       Р5
              120
3
       P2
              144
                        55
4
       Р6
              218
                         NA
5
       Р7
              229
                         29
```

#### **Rows** are removed based on their condition

The code below demonstrates how to eliminate rows based on certain criteria.

## **Code:**

```
Untitled4** Untitled5**

| Source on Save | Source on Sav
```

```
Console Terminal ×
                 Jobs ×
R 4.1.2 · ~/ ≈
> source("~/.active-rstudio-document", echo=TRUE)
> df <- data.frame(player = c('P1', 'P2', 'P3', 'P4', 'P5','P2', 'P6', 'P7'),</pre>
          points = c(122, 144, 154, 155, 120, 144, 218, 229),
           .... [TRUNCATED]
> df
  player points assists
      Ρ1
            122
                     43
                     55
2
      P2
            144
3
      Р3
                     77
            154
4
      Р4
            155
                     18
5
      P5
                    114
            120
6
      P2
            144
                     55
7
      P6
            218
                     NA
8
      Р7
                     29
            229
> df %>% filter(player=='P1' | assists >100)
  player points assists
     Ρ1
            122
                     43
      Ρ5
                     114
            120
```

#### **❖** Aim:

Create a random sample whit the help of distribution of 100 observations where minimum value is 2 and maximum value is 5 than categorize the data into class interval and calculate the frequency.

# **\*** Theory:

The continuous uniform distribution is also referred to as the probability distribution of any random number selection from the continuous interval defined between intervals a and b. A uniform distribution holds the same probability for the entire interval. Thus, its plot is a rectangle, and therefore it is often referred to as Rectangular distribution. Here we will discuss various functions and cases in which these functions should be used to get a required probability.

For uniform distribution, we first need a randomly created sequence ranging between two numbers. The **runif()** function in R programming language is used to generate a sequence of random following the uniform distribution.

# **Steps:**

#### **Code:**

```
\mathbf{X} = \text{runif}(100, \text{min} = 2, \text{max} = 5)
```

 $\mathbf{X}$ 

#### **Output:**

#### **Code:**

```
Z = cut(x,breaks=c(1,2,3,4,5))
```

 $\mathbf{Z}$ 

#### **Code:**

calculate the frequency of sample

table(z)

#### **Output:**

#### **\*** Result:

we generate a random sample of 100 observations from a uniform distribution using the runif() function with a minimum value of 2 and a maximum value of 5. Next, we create class intervals using the seq() function with a step size of 0.5 to create intervals ranging from 2 to 5. We then categorize the random sample into these class intervals using the **cut()** function, which returns a factor variable representing the category of each observation. Finally, we use the **table()** function to calculate the frequency of each class interval and print the result using **cat()**.

#### \* AIM:

The following data is the number of pages is 20 books on Statistic Mathematics and Physics construct a data frame and categorize the continuous variable statistic & Mathematics in the different class interval

#### **Code:**

```
② Untitled6* × ② Untitled7* × ② Untitled9* × ② Untitled8* ×
                                                                                        O Untitled10* × >> ☐ ☐
ntitled3* × 👂 Untitled4* × 👂 Untitled12* ×
      maths<-c(502,539,550,681,251,491,750,548,546,539,345,407,593,331,832,608,666,502,278,437)
physics<-c(915,623,405,900,611,744,563,417,401,294,669,477,251,804,710,730,872,336,919,645)
  5
      df <- data.frame(statistic,maths,physics)</pre>
  6
  8
      dfstatistic1 <- cut(dfstatistic, breaks = c(200,300,400,500,600,700,800,900))
  9
      df$statistic1
  10
  11
      df\$maths1 <- \ cut(df\$maths,breaks = c(200,300,400,500,600,700,800,900))
 12
      df$maths1
 14
      df
  15
```

```
Console Terminal × Background Jobs ×

R 8.4.2.2 · √ ∞

> statistic <-c(448,831,842,779,556,833,315,665,462,281,532,432,405,277,843,761,410,554,348,473)

> maths<-c(502,539,550,681,251,491,750,548,546,539,345,407,593,331,832,608,666,502,278,437)

> physics<-c(915,623,405,900,611,744,563,417,401,294,669,477,251,804,710,730,872,336,919,645)

> df <- data.frame(statistic,maths,physics)

> df$statistic1 <- cut(df$statistic,breaks = c(200,300,400,500,600,700,800,900))

> df$statistic1

[1] (400,500] (800,900] (800,900] (700,800] (500,600] (800,900] (300,400] (600,700] (400,500]

[19] (300,400] (400,500]

Levels: (200,300] (300,400] (400,500] (500,600] (600,700] (700,800] (800,900]

> df$maths1 <- cut(df$maths,breaks = c(200,300,400,500,600,700,800,900))

> df$maths1

[1] (500,600] (500,600] (500,600] (600,700] (200,300] (400,500] (700,800] (500,600] (500,600]

[19] (200,300] (300,400] (400,500] (500,600] (300,400] (800,900] (600,700] (500,600]

[19] (200,300] (300,400] (400,500] (500,600] (300,400] (800,900] (600,700] (500,600]

[19] (200,300] (300,400] (400,500] (500,600] (500,600] (600,700] (700,800] (800,900]

Levels: (200,300] (300,400] (400,500] (500,600] (600,700] (700,800] (800,900]
```

```
> df
  statistic maths physics statistic1
                                           maths1
         448
               502
                        915
                             (400,500]
                                        (500,600]
         831
                539
                        623
                             (800,900] (500,600]
3
         842
               550
                        405
                             (800,900] (500,600]
         779
               681
                        900
                             (700,800]
                                        (600,700]
5
         556
                             (500,600] (200,300]
               251
                        611
6
         833
               491
                        744
                             (800,900] (400,500]
         315
               750
                        563
                             (300,400] (700,800]
8
         665
                548
                        417
                             (600,700] (500,600]
9
         462
               546
                        401
                             (400,500] (500,600]
10
         281
                        294
                                        (500,600]
                539
                             (200,300]
11
         532
               345
                        669
                             (500,600] (300,400]
                        477
12
         432
               407
                             (400,500] (400,500]
13
         405
                593
                        251
                             (400,500]
                                        (500,600]
14
         277
               331
                        804
                             (200,300] (300,400]
15
         843
                        710
                             (800,900] (800,900]
               832
16
         761
               608
                        730
                             (700,800]
                                        (600,700]
17
         410
               666
                        872
                             (400,500] (600,700]
18
         554
               502
                        336
                             (500,600] (500,600]
19
         348
               278
                        919
                             (300,400]
                                        (200,300]
20
         473
               437
                        645 (400,500] (400,500]
```

*	Result:  Construct a data frame with variables Statistic Mathematics and Physics using data.frame ()  Function and then categorize the continuous variables statistic and mathematics in different class interval using cut () function.	

#### **❖** Aim:

To work with dates and times in R

# **\*** Theory:

R provides several options for dealing with date and date/time data. Three date/time classes commonly used in R are Date, POSIXct and POSIXlt.

# **\*** formats and their descriptions used in date and time in r:

In R, you can use format codes to specify the desired format for representing dates and times when using functions like **format()**, **strftime()**, **and as.Date()**. Here are some commonly used format codes and their descriptions:

- %Y: Year with century as a decimal number (e.g., 2023).
- %y: Year without century as a decimal number (00-99).
- %m: Month as a decimal number (01-12).
- %d: Day of the month as a decimal number (01-31).
- %H: Hour (00-23) as a decimal number.
- %M: Minute (00-59) as a decimal number.
- %S: Second (00-59) as a decimal number.
- %b: Abbreviated month name.
- %B: Full month name.
- %a: Abbreviated weekday name.
- %A: Full weekday name.
- %p: Either AM or PM.
- %z: Time zone offset from UTC as +HHMM or -HHMM.
- %Z: Time zone name.

## **❖** Date class:

The Date class represents dates without any time information. Dates in R are typically represented in the format "YYYY-MM-DD". Here's an example:

```
Console Terminal × Background Jobs ×

R 84.2.2 · ~/ **

> # Create a Date object

> my_date <- as.Date("2023-04-15")

> # Print the Date object

> print(my_date)
[1] "2023-04-15"

> # Extract components of the Date object

> year <- format(my_date, format = "%Y")

> month <- format(my_date, format = "%m")

> day <- format(my_date, format = "%d")

> # Print the components

> cat("Year:", year, "\n")

Year: 2023

> cat("Month:", month, "\n")

Month: 04

> cat("Day:", day, "\n")

Day: 15
```

## **❖ POSIXct class:**

The POSIXct class represents date and time information as a continuous numeric value representing the number of seconds since the UNIX epoch (January 1, 1970, 00:00:00 UTC). Here's an example:

```
> # Create a POSIXct object
> my_datetime <- as.POSIXct("2023-04-15 12:34:56", tz = "UTC")</pre>
> # Print the POSIXct object
> print(my_datetime)
[1] "2023-04-15 12:34:56 UTC"
> # Extract components of the POSIXct object
> year <- format(my_datetime, format = "%Y")
> month <- format(my_datetime, format = "%m")</pre>
> day <- format(my_datetime, format = "%d")</pre>
> hour <- format(my_datetime, format = "%H")</pre>
> minute <- format(my_datetime, format = "%M")</pre>
> second <- format(my_datetime, format = "%S")</pre>
> # Print the components
> cat("Year:", year, "\n")
Year: 2023
> cat("Month:", month, "\n")
Month: 04
> cat("Day:", day, "\n")
Day: 15
> cat("Hour:", hour, "\n")
Hour: 12
> cat("Minute:", minute, "\n")
Minute: 34
> cat("Second:", second, "\n")
Second: 56
```

#### **POSIXIt class:**

The POSIXIt class represents date and time information as a list of components (e.g., year, month, day, hour, minute, second) which allows for more fine-grained manipulation. Here's an example:

```
Console Terminal × Background Jobs ×

    R 4.2.2 · ~/ ∅

> # Create a POSIXlt object
> my_datetime <- as.POSIX1t("2023-04-15 12:34:56", tz = "UTC")
> # Print the POSIXlt object
> print(my_datetime)
[1] "2023-04-15 12:34:56 UTC"
> # Extract components of the POSIXlt object
> year <- my_datetime$year + 1900
> month <- my_datetime$mon + 1
> day <- my_datetime$mday
> hour <- my_datetime$hour
> minute <- my_datetime$min
> second <- my_datetime$sec
> # Print the components
> cat("Year:", year, "\n")
Year: 2023
> cat("Month:", month, "\n")
Month: 4 > cat("Day:", day, "\n")
Day: 15
> cat("Hour:", hour, "\n")
Hour: 12
> cat("Minute:", minute, "\n")
Minute: 34
> cat("Second:", second, "\n")
Second: 56
```

❖ Here's an example of how you can use these format codes in R to get the current date and time in various formats:

```
Console Terminal × Background Jobs
 R 4.2.2 · ~/ @
 > # Get current date and time
> now <- Sys.time()</pre>
> # Year with century (e.g., 2023)
> year_with_century <- format(now, "%Y")
> cat("Year with century: ", year_with_century, "\n")
Year with century: 2023
> # Year without century (00-99)
> year_without_century <- format(now, "%y")
> cat("Year without century: ", year_without_century, "\n")
Year without century: 23
> # Month as decimal number (01-12)
> month <- format(now, "%m")
> cat("Month: ", month, "\n")
Month: 04
> # Day of the month as decimal number (01-31)
> day <- format(now, "%d")
> cat("Day of the month: ", day, "\n")
Day of the month: 15
> # Hour (00-23)
> hour <- format(now, "%H")
> cat("Hour: ", hour, "\n")
Hour: 00
> # Minute (00-59)
> minute (00 33)
> minute <- format(now, "%M")
> cat("Minute: ", minute, "\n")
Minute: 49
```

```
> # Second (00-59)
> second <- format(now, "%S")</pre>
> cat("Second: ", second, "\n")
Second: 53
> # Abbreviated month name
> abbreviated_month <- format(now, "%b")</pre>
> cat("Abbreviated month name: ", abbreviated_month, "\n")
Abbreviated month name: Apr
> # Full month name
> full_month <- format(now, "%B")
> cat("Full month name: ", full_month, "\n")
Full month name: April
> # Abbreviated weekday name
> abbreviated_weekday <- format(now, "%a")</pre>
> cat("Abbreviated weekday name: ", abbreviated_weekday, "\n")
Abbreviated weekday name: Sat
> # Full weekday name
> full_weekday <- format(now, "%A")</pre>
> cat("Full weekday name: ", full_weekday, "\n")
Full weekday name: Saturday
>
```

#### \* Result:

The actual result of dates and times in R will depend on the format and class of the object being used, as well as any formatting or conversion operations applied to the object.

#### ❖ Aim:

To select subset of data in R.

## **\*** Theory:

In R, a subset of data refers to a portion of a larger data set that is selected based on specific conditions or criteria. Subsetting data is a common operation in data analysis and allows you to extract and work with a smaller, more focused portion of the data for further analysis or visualization.

There are several ways to subset data in R, including using indexing, logical conditions, or functions from various packages. Here are some common methods for creating subsets of data in R:

#### 1. Indexing:

You can use square brackets [] to subset data based on row and column indices. For example, data[row\_indices, col\_indices] can be used to select specific rows and columns from a data frame or a matrix.

## **Code:**

**Output:** 

```
# Create example data frame
data <- data.frame(
Name = c("Alice", "Bob", "Charlie", "Dave", "Eve"),
Age = c(25, 30, 35, 40, 45),
City = c("New York", "Los Angeles", "Chicago", "San Francisco", "Miami")
)
```

```
Console Terminal × Background Jobs ×
R 4.2.2 · ~/ ≈
> # Create example data frame
/ "Create Champe attent"
> data <- data.frame(
+ Name = c("Alice", "Bob", "Charlie", "Dave", "Eve"),
+ Age = c(25, 30, 35, 40, 45),
+ City = c("New York", "Los Angeles", "Chicago", "San Francisco", "Miami")</pre>
       Name Age
                                 City
                     New York
1 Alice 25 New York
2 Bob 30 Los Angeles
3 Charlie 35 Chicago
      Dave 40 San Francisco
        Eve 45
                               Miami
> # Select rows 2 and 3, and columns 1 and 3
> subset_data <- data[c(2, 3), c(1, 3)]</pre>
> # Print the subset of data
> print(subset_data)
       Name
                       City
        Bob Los Angeles
3 Charlie
                    Chicago
```

# 2. Logical conditions:

You can use logical conditions to subset data based on values in one or more columns. For example, data[data\$Age > 30, ] selects rows where the value in the Age column is greater than 30.

```
Console Terminal x Background Jobs x

R 84.22 · -/ **

> data <- data.frame(
+ Name = c("Alice", "Bob", "Charlie", "Dave", "Eve"),
+ Age = c(25, 30, 35, 40, 45),
+ City = c("New York", "Los Angeles", "Chicago", "San Francisco", "Miami")
+ )
>

## Select rows where Age is greater than 30
> subset_data <- data[data$Age > 30, ]
>
## Print the subset of data
> print(subset_data)
Name Age City
3 Charlie 35 Chicago
4 Dave 40 San Francisco
5 Eve 45 Miami
>
```

#### 3. Functions:

R provides various functions for subsetting data, such as subset(), filter() from the dplyr package, and slice() from the dplyr or data.table package. These functions allow you to specify conditions or criteria to filter and extract specific rows or columns from a data frame.

#### **OUTPUT:**

```
Console Terminal × Background Jobs ×
                                                                                                                                        R 4.2.2 · ~/ ≈
> data <- data.frame(</pre>
   nata <- data.Trame(
Name = c("Alice", "Bob", "Charlie", "Dave", "Eve"),
Age = c(25, 30, 35, 40, 45),
City = c("New York", "Los Angeles", "Chicago", "San Francisco", "Miami")
> # Load the dplyr package
> library(dplyr)
> # Select rows where Age is greater than 30 using filter()
> subset_data <- filter(data, Age > 30)
> # Print the subset of data
> print(subset_data)
      Name Age
                              City
1 Charlie 35 Chicago
2 Dave 40 San Francisco
3
       Eve 45
                             Miami
```

# 4. Regular expressions:

You can use regular expressions with functions like grep() or grepl() to subset data based on patterns or text matching in character vectors.

#### **OUTPUT:**

## **\*** Aim:

To select a random sample from a dataset in R.

# **\*** Theory:

A random sample from a dataset in R refers to a subset of data that is randomly selected without any specific pattern or order. This random selection is useful for various purposes, such as statistical inference, data exploration, and model validation.

In R, you can use the sample() function to generate a random sample from a dataset. The sample() function randomly selects elements from a vector or a data frame. Here's an example:

#### **Code:**

```
data <- data.frame(
    Name = c("Alice", "Bob", "Charlie", "Dave", "Eve"),
    Age = c(25, 30, 35, 40, 45),
    City = c("New York", "Los Angeles", "Chicago", "San Francisco", "Miami")
)
```

#### **Output:**

```
Console Terminal × Background Jobs ×

R R4.22 · ~/ →

> # Create example data frame

> data <- data.frame(

+ Name = c("Alice", "Bob", "Charlie", "Dave", "Eve"),

+ Age = c(25, 30, 35, 40, 45),

+ City = c("New York", "Los Angeles", "Chicago", "San Francisco", "Miami")

+ )

> # Set seed for reproducibility (optional)

> set.seed(123)

> # Select a random sample of 3 rows from the data frame

> random_sample <- data[sample(nrow(data), 3), ]

> # Print the random sample

print(random_sample)

Name Age City

3 Charlie 35 Chicago

2 Bob 30 Los Angeles

5 Eve 45 Miami

> |
```

#### **\*** Result:

we use the resulting row indices to select the corresponding rows from the data frame and store them in random\_sample, which is then printed using cat() and print() functions.

	Practical-10
❖ Aim:	
* Theory:	