

# Jornada de Ciência de Dados 11 anos!





# R-BASICS

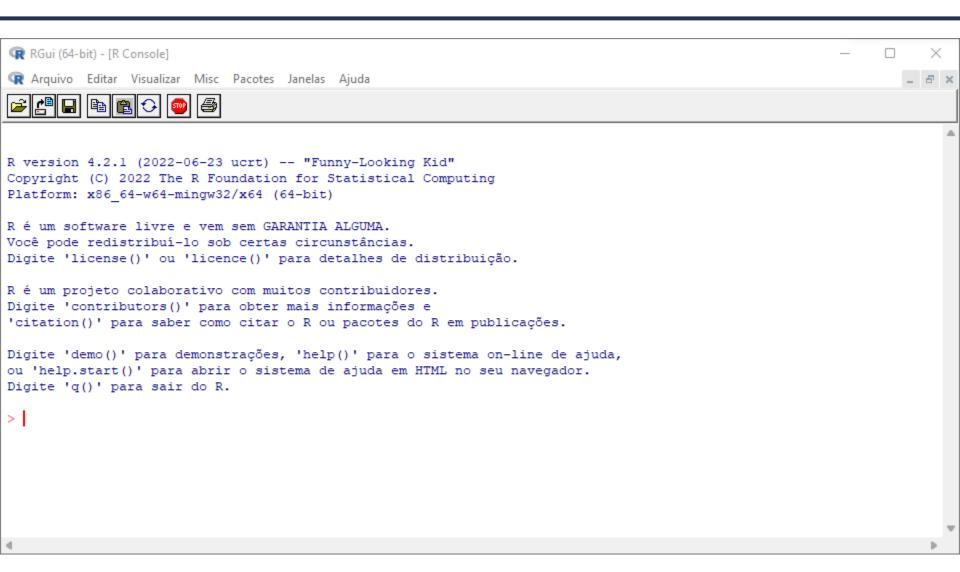
Eduardo Ogasawara eduardo.ogasawara@cefet-rj.br https://eic.cefet-rj.br/~eogasawara



## Introduction to R

- R is a programming language and free software environment for statistical computing
  - Supported by the R Foundation for Statistical Computing
- Created by Ross Ihaka and Robert Gentleman at Auckland University, New Zealand
- R was derived by S (Bell Laboratories AT&T)
- R is a language broadly used by statisticians, data miners, and data scientists
- Current version 4.4

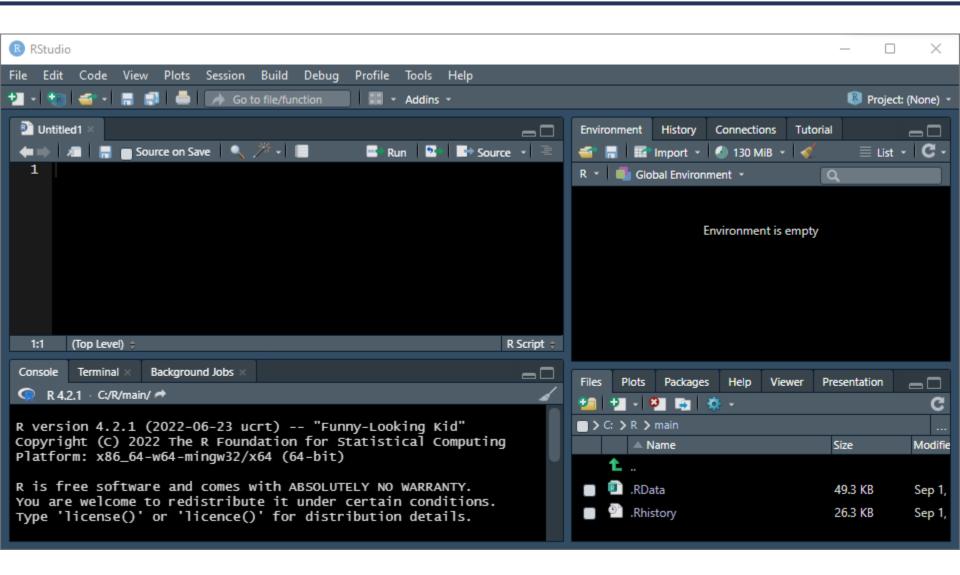
#### R Console



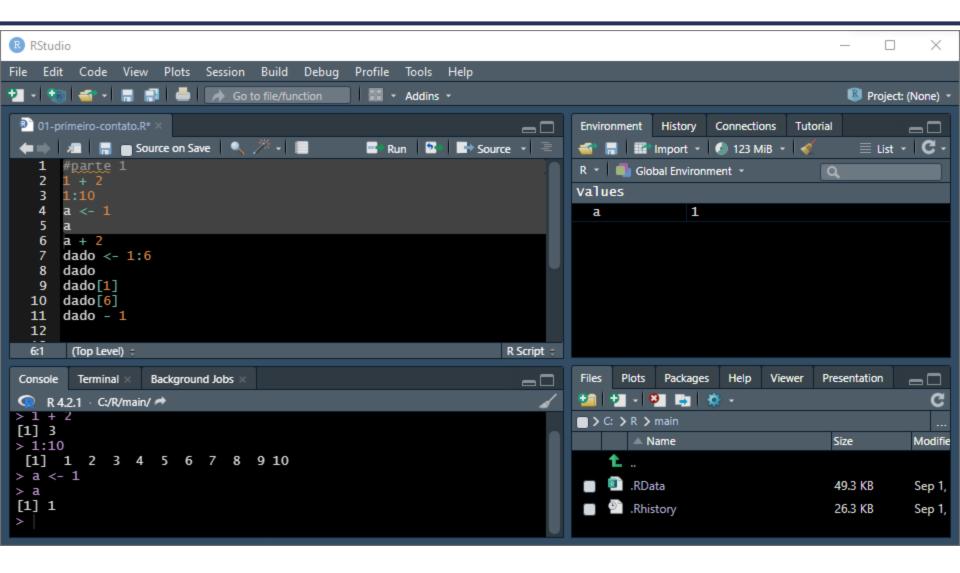
# R Console (Basic operations)

```
RGui (64-bit) - [R Console]
R Arquivo Editar Visualizar Misc Pacotes Janelas Ajuda
> 1 + 2
[1] 3
> 1:10
[1] 1 2 3 4 5 6 7 8 9 10
> a <- 1
> a
[1] 1
> a + 2
[1] 3
> dado <- 1:6
> dado
[1] 1 2 3 4 5 6
> dado[1]
[1] 1
> dado[6]
[1] 6
> dado - 1
[1] 0 1 2 3 4 5
```

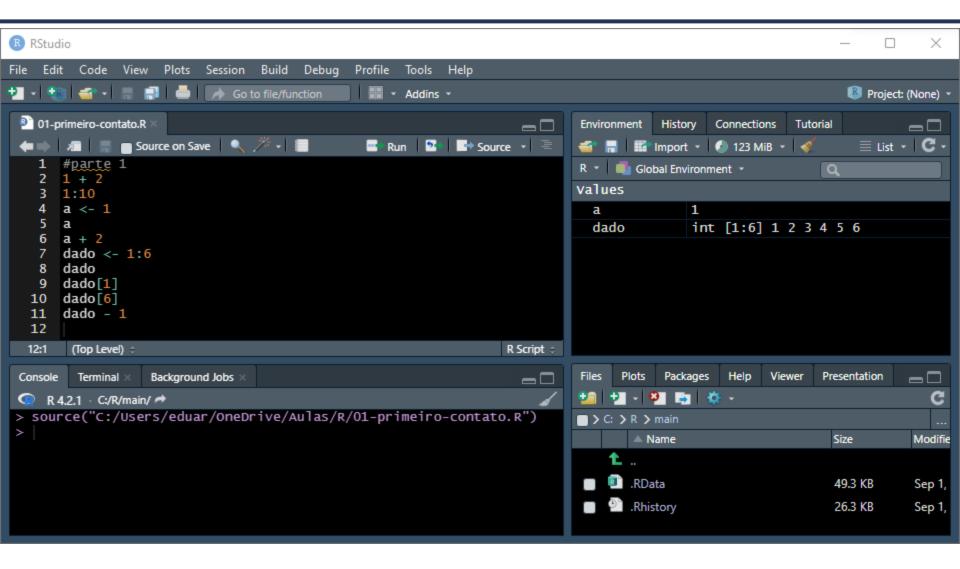
#### R Studio



## Run a block of code



## Run entire source file



## R Packages

- Packages are collections of functions made available as libraries
  - Published in the CRAN Repository
    - Quality control
    - good documentation
  - Uploaded from a GitHub repository
    - Versions under development
- Currently have more than 22000 packages
  - https://cran.r-project.org/
- R has a very active community
  - Several researchers, professors, programmers, and statisticians
- DAL Packages: 7 pacotes
  - https://cran.r-project.org/web/packages/daltoolbox/index.html
  - https://cran.r-project.org/web/packages/harbinger/index.html
  - https://cran.r-project.org/web/packages/tspredit/index.html

# R Package installation and loading

#### Package checking and instalation

```
if(!require(daltoolbox)) {
  install.packages("daltoolbox")
}
```

#### Package loading

library(daltoolbox)

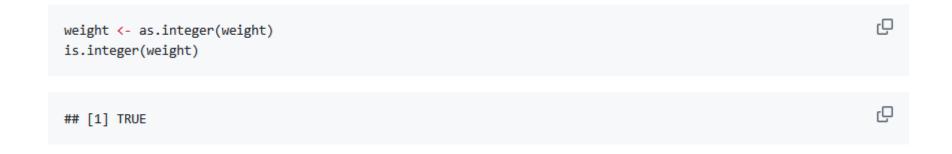
# Variable definitions

# Variable definition and assignment weight <- 60 height = 1.75 subject <- "A" healthy <- TRUE Variable evaluation weight ## [1] 60

# Functions for type checking

<pre>is.numeric(weight) # variable</pre>	C
## [1] TRUE	C)
is.double(weight)	-C
## [1] TRUE	<sub>C</sub>
is.integer(weight)	-C
## [1] FALSE	-C
is.character(subject)	-C
## [1] TRUE	-C

# Functions for variable conversion



# Formulas and String formatting

#### Computing the body mass index (BMI) from the weight and height

```
#Body mass index (BMI)
bmi <- weight/height^2
bmi

## [1] 19.59184
```

#### Functions for string manipulation

```
message <- sprintf("%.1f", bmi)
print(message)

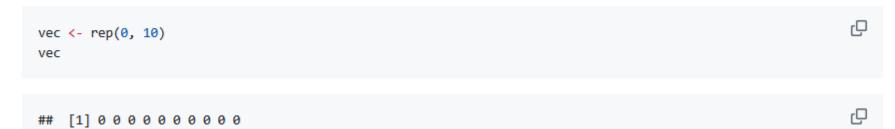
## [1] "19.6"
```

#### **Vectors**

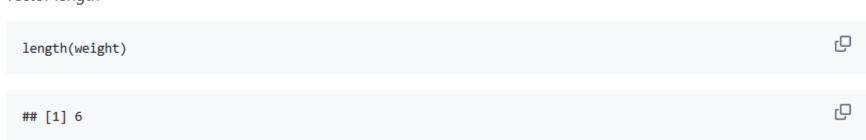
```
definition
  weight <- c(60, 72, 57, 90, 95, 72)
  height <- c(1.75, 1.80, 1.65, 1.90, 1.74, 1.91)
  subject <- c("A", "B", "C", "D", "E", "F")</pre>
evaluation
                                                                                                                 Q
  weight
  ## [1] 60 72 57 90 95 72
  height
                                                                                                                 Q
  ## [1] 1.75 1.80 1.65 1.90 1.74 1.91
  subject
  ## [1] "A" "B" "C" "D" "E" "F"
```

## **Vector indexes**

#### vector of size 10

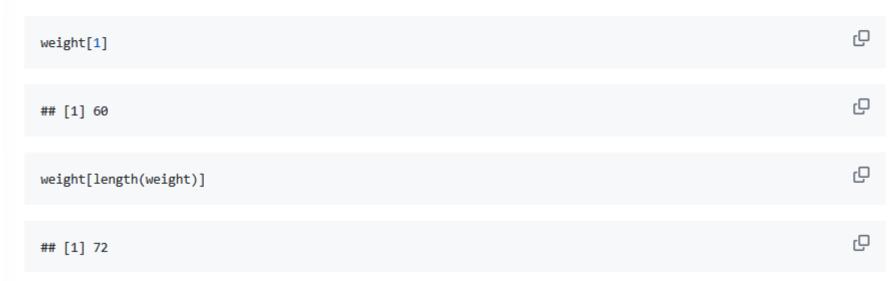


#### vector length



## **Vector limits**

from one to the length of the vector



# **Iterations:** for loop

from one to the length of weight

```
bmi <- 0
for (i in 1:length(weight)) {
   bmi[i] <- weight[i]/height[i]^2
}</pre>
```

evaluation of the bmi vector

```
bmi
```

```
## [1] 19.59184 22.22222 20.93664 24.93075 31.37799 19.73630
```

# **Iterations:** while loop

run while i is below or equal to the length of weight

```
bmi <- 0
i <- 1
while (i <= length(weight)) {
    bmi[i] <- weight[i]/height[i]^2
    i <- i + 1
}</pre>
bmi

## [1] 19.59184 22.22222 20.93664 24.93075 31.37799 19.73630
```

# Removing variables

```
rm(bmi)
exists("bmi")

## [1] TRUE
```

# **Processing vectors at once**



# **Creating functions**

```
name <- function(parameters) { body }
  compute_bmi <- function(weight, height) {</pre>
    bmi <- weight/height^2</pre>
    return(bmi)
calling it
  bmi <- compute_bmi(60, 1.75)</pre>
  bmi
  ## [1] 19.59184
```

Functions are defined with parameters and code assigned to a function name. Functions should return a value.

# **Function processing vectors**



# **Example**

- Knowing that mean is computed as
  - $\mu(X) = \frac{\sum_{i=1}^{n} x_i}{n}$
- Create a function to compute the mean of a vector

# **Example:** function to compute the average

(iterating in all elements of the vector)

```
average <- function(vec) {
    s <- 0
    n <- length(vec)
    for (x in vec) {
        s <- s + x
    }
    return(s/n)
}</pre>
```

#### invoking the function

```
avg_bmi <- average(bmi)
avg_bmi

## [1] 23.13262
```

# Example of a function to compute the average (improved version)

manipulating vectors at once average <- function(vec) {</pre> s <- sum(vec) n <- length(vec) return(s/n) invoking the function avg bmi <- average(bmi) avg bmi ## [1] 23.13262

# Example: function using mean function

Major statistical functions are available in R

```
average <- function(vec) {
    return(mean(vec))
}</pre>
```

#### invoking the function

```
avg_bmi <- average(bmi)
avg_bmi

## [1] 23.13262
```

# Not available (NA)

Operations with NA lead to NA.

```
x <- c(10, NA, 13)
y <- average(x)
y

## [1] NA
```

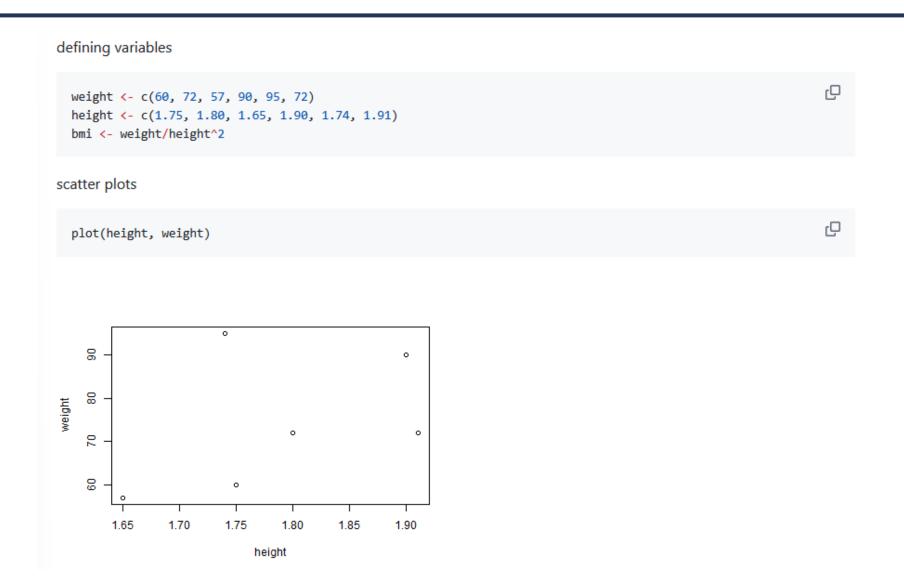
Addressing NA with na.rm=TRUE

```
average <- function(vec) {
    return(mean(vec, na.rm=TRUE))
}

x <- c(10, NA, 13)
y <- average(x)
y</pre>
## [1] 11.5
```

# Part II

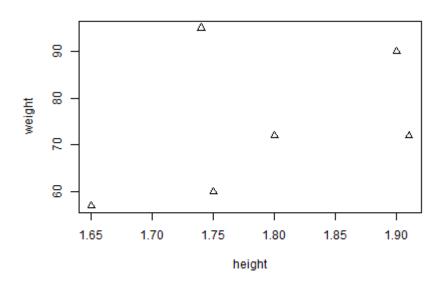
# **Plotting basic graphics**



The plot function is the basic plot function in R. The first parameter is the x-axis observations. In contrast, the second one is the y-axis observations.

# Most functions contain many default parameters

plot(height, weight, pch=2)



## All functions in R CRAN packages have help with examples

?base::plot

plot.default {graphics} R Documentation

## The Default Scatterplot Function

### Description

Draw a scatter plot with decorations such as axes and titles in the active graphics window.

#### Usage

```
## Default S3 method:
plot(x, y = NULL, type = "p", xlim = NULL, ylim = NULL,
    log = "", main = NULL, sub = NULL, xlab = NULL, ylab = NULL,
    ann = par("ann"), axes = TRUE, frame.plot = axes,
    panel.first = NULL, panel.last = NULL, asp = NA,
    xgap.axis = NA, ygap.axis = NA,
    ...)
```

#### Arguments

х, у

the x and y arguments provide the x and y coordinates for the plot. Any reasonable way of defining the coordinates is acceptable. See the function xy.coords for details. If supplied separately, they must be of the same length.

type

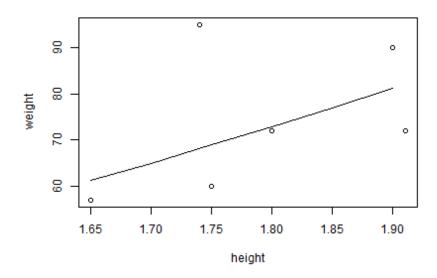
1-character string giving the type of plot desired. The following values are possible, for details, see

# Canvas for plotting is still active until a new plot

```
plot(height, weight)

hh = c(1.65, 1.70, 1.75, 1.80, 1.85, 1.90)

lines(hh, 22.5 * hh^2)
```



#### **Factors**

- Factors are variables in R that refer to categorical data
- Factors in R are stored as a vector of integer values with a corresponding set of character values to use when the factor is displayed
- Both numeric and character variables can be made into factors, but a factor's levels are always character values

#### **Factors**

Factors are used to handle categorical data.

```
pain \leftarrow c(0,3,2,2,1)
  fpain <- factor(pain,levels=0:3, ordered=TRUE)</pre>
  fpain
  ## [1] 0 3 2 2 1
  ## Levels: 0 < 1 < 2 < 3
Levels provide correspondence between numerical values and categorical labels
  levels(fpain) <- c("none", "mild", "medium", "severe")</pre>
  fpain
  ## [1] none severe medium medium mild
  ## Levels: none < mild < medium < severe
```

## **Example**

- Consider the height variable
  - Persons lower than 1.5 m are small
  - Persons greater than 1.9 m are tall
  - Persons in between are medium
  - Convert the height variable into a factor with small, medium, tall

## Convert height to factor with levels: small, medium, high

coding setting element by element

```
Q
lev <- rep("", length(height))</pre>
for (i in 1:length(height)) {
  if (height[i] < 1.5)</pre>
    lev[i] <- "short"</pre>
  else if (height[i] < 1.9)</pre>
    lev[i] <- "medium"</pre>
  else
    lev[i] <- "tall"</pre>
lev <- as.factor(lev)</pre>
lev
## [1] medium medium tall medium tall
## Levels: medium tall
```

#### Using the function cut

#### **Matrix**

Matrices can be filled from vectors or data frames.

```
x <- 1:9
x
## [1] 1 2 3 4 5 6 7 8 9
```

#### Converting a vector to matrix

```
dim(x) <- c(3,3)
x

## [,1] [,2] [,3]
## [1,] 1 4 7
## [2,] 2 5 8
## [3,] 3 6 9
```

## **Scalar multiplication**

```
y <- 3*x
y

## [,1] [,2] [,3]
## [1,] 3 12 21
## [2,] 6 15 24
## [3,] 9 18 27
```

# **Matrix** manipulation

#### Converting a vector to a matrix by row

```
x <- matrix(1:9,nrow=3,byrow=T)
x

## [,1] [,2] [,3]
## [1,] 1 2 3
## [2,] 4 5 6
## [3,] 7 8 9</pre>
```

#### transposing a matrix

```
x <- t(x)
x

## [,1] [,2] [,3]
## [1,] 1 4 7
## [2,] 2 5 8
## [3,] 3 6 9</pre>
```

#### Lists

- Lists are the R objects which contain elements of different types, such as numbers, strings, vectors, matrix, data frame, and another list inside it
- A list can also contain a matrix or a function as its elements
- A list is created using the list() function
- List manipulation
  - Slicing a list []
  - Accessing an element inside a list [[]]

#### Creating a list

Lists are used to work with "objects"

```
a <- c(5260,5470,5640,6180,6390,6515,6805,7515,7515,8230,8770)
b <- c(3910,4220,3885,5160,5645,4680,5265,5975,6790,6900,7335)
mybag <- list(a, b, 0, "a")
mybag
## [[1]]
## [1] 5260 5470 5640 6180 6390 6515 6805 7515 7515 8230 8770
##
## [[2]]
## [1] 3910 4220 3885 5160 5645 4680 5265 5975 6790 6900 7335
##
## [[3]]
## [1] 0
##
## [[4]]
## [1] "a"
```

#### Adding elements into a list

```
n <- length(mybag)</pre>
mybag[[n+1]] <- "b"
mybag
## [[1]]
## [1] 5260 5470 5640 6180 6390 6515 6805 7515 7515 8230 8770
##
## [[2]]
## [1] 3910 4220 3885 5160 5645 4680 5265 5975 6790 6900 7335
##
## [[3]]
## [1] 0
## [[4]]
## [1] "a"
##
## [[5]]
## [1] "b"
```

# List slicing

```
slice <- mybag[1]
slice

## [[1]]
## [1] 5260 5470 5640 6180 6390 6515 6805 7515 7515 8230 8770

is.list(slice)

## [1] TRUE
```

#### **Lists: Member reference**

#### Picking the first element

```
h <- mybag[[1]]
  h
     [1] 5260 5470 5640 6180 6390 6515 6805 7515 7515 8230 8770
An element can be evaluated. In this case, it is a vector.
  is.vector(h)
  ## [1] TRUE
  is.list(h)
  ## [1] FALSE
```

#### **Creating lists with attributes**

They are properties on the list

```
mybag <- list(x=a, y=b, const=0, lit="a")

## $x

## [1] 5260 5470 5640 6180 6390 6515 6805 7515 7515 8230 8770

##

## $y

## [1] 3910 4220 3885 5160 5645 4680 5265 5975 6790 6900 7335

##

## $const

## [1] 0

##

## $lit

## [1] "a"
```

#### Adding, accessing, and removing elements

#### Adding, accessing, and removing elements

```
mybag$c <- mybag$x - mybag$y
mybag$const <- NULL
mybag$lit <- NULL
mybag

## $x
## [1] 5260 5470 5640 6180 6390 6515 6805 7515 7515 8230 8770
##
##
##
##
##
$y
##
[1] 3910 4220 3885 5160 5645 4680 5265 5975 6790 6900 7335
##
##
##
##
$c
##
[1] 1350 1250 1755 1020 745 1835 1540 1540 725 1330 1435</pre>
```

# Part III

#### **Data frames**

Data frames (tables) provide support for structured data.

```
a <- c(5260,5470,5640,6180,6390,6515,6805,7515,7515,8230,8770)
b <- c(3910,4220,3885,5160,5645,4680,5265,5975,6790,6900,7335)

data <- data.frame(A=a, B=b)
head(data)

## A B
## 1 5260 3910
## 2 5470 4220
## 3 5640 3885
## 4 6180 5160
## 5 6390 5645
## 6 6515 4680
```

#### Adding a column in a data frame

```
data$c <- data$A + data$B
head(data)

## A B c
## 1 5260 3910 9170
## 2 5470 4220 9690
## 3 5640 3885 9525
## 4 6180 5160 11340
## 5 6390 5645 12035
## 6 6515 4680 11195
```

#### Removing a column of a data frame

```
data$A <- NULL head(data)

## B c ## 1 3910 9170 ## 2 4220 9690 ## 3 3885 9525 ## 4 5160 11340 ## 5 5645 12035 ## 6 4680 11195
```

#### Reading a csv file

There are many functions for reading CSV, Excel, and RData formats.

```
wine = read.table(
    "http://archive.ics.uci.edu/ml/machine-learning-databases/wine/wine.data",
                  header = TRUE, sep = ",")
  colnames(wine) <- c('Type', 'Alcohol', 'Malic', 'Ash',</pre>
                      'Alcalinity', 'Magnesium', 'Phenols',
                      'Flavanoids', 'Nonflavanoids',
                      'Proanthocyanins', 'Color', 'Hue',
                      'Dilution', 'Proline')
head(wine)
     Type Alcohol Malic Ash Alcalinity Magnesium Phenols Flavanoids Nonflavanoids Proanthocyanins Color H 🖵
           13.20 1.78 2.14
                                                                                              1.28 4.38 1.
## 1
                                   11.2
                                              100
                                                     2.65
                                                                2.76
                                                                              0.26
                                   18.6
                                                     2.80
                                                                3.24
                                                                              0.30
                                                                                               2.81 5.68 1.
## 2
           13.16 2.36 2.67
                                              101
       1 14.37 1.95 2.50
                                   16.8
                                                     3.85
                                                                                               2.18 7.80 0.
                                                                3.49
                                                                              0.24
## 3
                                              113
```

#### Saving and loading Rdata files

```
saving a data frame
  save(wine, file="wine.RData")
removing a data frame from memory
                                                                                                            ſĠ
  rm(wine)
loading it
  load("wine.RData")
  head(wine, 3)
       Type Alcohol Malic Ash Alcalinity Magnesium Phenols Flavanoids Nonflavanoids Proanthocyanins Color H 🖵
             13.20 1.78 2.14
                                    11.2
                                                      2.65
                                                                2.76
                                                                              0.26
                                                                                              1.28 4.38 1.
  ## 1
                                               100
             13.16 2.36 2.67
                                    18.6
                                               101
                                                     2.80
                                                                3.24
                                                                              0.30
                                                                                              2.81 5.68 1.
  ## 2
                                    16.8
                                                                                              2.18 7.80 0.
         1 14.37 1.95 2.50
                                               113
                                                     3.85
                                                                3.49
                                                                              0.24
      Dilution Proline
  ## 1
          3.40
                 1050
                 1185
  ## 2
          3.17
  ## 3
          3.45
                 1480
```

# **Exporting data.frame into csv file**

```
write.table(wine, file="wine.csv", row.names=FALSE, quote = FALSE, sep = ",")
```

Try to see the wine.csv

# Filtering vectors

```
a <- c(5260,5470,5640,6180,6390,6515,6805,7515,7515,8230,8770)
b <- c(3910,4220,3885,5160,5645,4680,5265,5975,6790,6900,7335)
data <- data.frame(a = a, b = b)
head(data[data$a > 7000,])

## a b
## 8 7515 5975
## 9 7515 6790
## 10 8230 6900
## 11 8770 7335
```

# Filtering data frames

#### filtering data frames (rows and columns)



#### Benchmarking performance of matrix and data frames

```
rheight <- rnorm(100000, 1.8, sd=0.2)
rweight <- rnorm(100000, 72, sd=15)
```

#### Computing a entire column at once in a data frame

```
start_time <- Sys.time()</pre>
hw <- data.frame(height=rheight, weight=rweight)</pre>
hw$bmi <- hw$weight/hw$height^2
end_time <- Sys.time()</pre>
end_time - start_time
## Time difference of 0.002799988 secs
object.size(hw)
## 2400984 bytes
```

#### Computing a column in a data frame cell by cell

#### processing cell by cell

```
start_time <- Sys.time()
hw <- data.frame(height=rheight, weight=rweight)
for (i in 1:nrow(hw)) {
   hw$bmi[i] <- hw$weight[i]/hw$height[i]^2
}
end_time <- Sys.time()
end_time - start_time</pre>
## Time difference of 12.53602 secs
```

#### Converting a data frame to a matrix, processing it, and going back

```
start_time <- Sys.time()</pre>
hw <- data.frame(height=rheight, weight=rweight)</pre>
hw <- as.matrix(hw)
hw <- cbind(hw, 0)
for (i in 1:nrow(hw)) {
  hw[i,3] \leftarrow hw[i,2]/hw[i,1]^2
hw <- as.data.frame(hw)
end time <- Sys.time()
end time - start time
## Time difference of 0.272609 secs
```

# Part IV

#### **Pipelines**

The operator |> creates a pipeline.

The first parameter of the next invoked function receives the data from the pipeline.

Library dplyr contains a set of functions that support relational algebra operations.

```
flight_data <- read.table(text = "Year Quarter Flights Delays

2016 1 11 6

2016 2 12 5

2016 3 13 3

2016 4 12 5

2017 1 10 4

2017 2 9 3

2017 3 11 4

2017 4 25 15

2018 1 14 3

2018 2 12 5

2018 3 13 3

2018 4 15 4",

header = TRUE, sep = "")
```

## Displaying the data frame

## **Basic Query**

```
library(dplyr)

result <- flight_data |>
    filter(Delays > 5) |>
    select(Year, Quarter, Flights)
head(result)

## Year Quarter Flights
## 1 2016  1  11
## 2 2017  4  25
```

## **Aggregated query**

```
result <- flight_data |>
    group_by(Year) |>
    summarize(mean = mean(Flights), sd = sd(Flights))
head(result)

## # A tibble: 3 x 3
## Year mean sd
## <int> <dbl> <dbl> <dbl> <dbl> ## 1 2016 12 0.816
## 2 2017 13.8 7.54
## 3 2018 13.5 1.29
```

#### Tables join

#### Store table

```
ſĠ
stores <- data.frame(</pre>
 city = c("Rio de Janeiro", "Sao Paulo", "Paris", "New York", "Tokyo"),
 value = c(10, 12, 20, 25, 18))
head(stores)
                                                                                                       Q
              city value
##
## 1 Rio de Janeiro
                    10
## 2
       Sao Paulo
                    12
## 3
             Paris
                    20
## 4
          New York
                    25
## 5
          Tokyo
                    18
```

#### Division table

```
ſĠ
divisions <- data.frame(</pre>
 city = c("Rio de Janeiro", "Sao Paulo", "Paris", "New York", "Tokyo"),
  country = c("Brazil", "Brazil", "France", "US", "Japan"))
head(divisions)
                                                                                                          Q
##
              city country
## 1 Rio de Janeiro Brazil
## 2
         Sao Paulo Brazil
## 3
           Paris France
## 4
        New York
                        US
           Tokyo Japan
## 5
```

#### **Merge function**

The function merge can be used to join data frames. It can be used to produce inner, left, right, and outer joins.

```
stdiv <- merge(stores, divisions, by.x="city", by.y="city")</pre>
head(stdiv)
                                                                                                         Q
              city value country
##
          New York
                      25
## 1
                             US
             Paris
## 2
                     20 France
## 3 Rio de Janeiro
                    10 Brazil
## 4
         Sao Paulo
                    12 Brazil
             Tokyo
                     18 Japan
## 5
```

# Aggregating merged data frame

# Part V

#### Statistical analysis

There are many statistical tests in R. One of the most used is the t-test. It checks if the mean of observations is not different from a theoretical value.

```
weight <- c(60, 72, 57, 90, 95, 72)
height <- c(1.75, 1.80, 1.65, 1.90, 1.74, 1.91)
bmi <- weight/height^2</pre>
                                                                                                              Q
t.test(bmi, mu=22.5)
                                                                                                              ф
##
        One Sample t-test
## data: bmi
## t = 0.34488, df = 5, p-value = 0.7442
## alternative hypothesis: true mean is not equal to 22.5
## 95 percent confidence interval:
## 18.41734 27.84791
## sample estimates:
## mean of x
## 23.13262
```

#### **Python** + R integration

#### Python code at retic.py

```
import pyreadr
import pandas

def add(x, y):
    return x + y

def read_rdata_mem(data):
    x = data["x"]
    print(x)
    y = data["y"]
    data["z"] = x + y
    return(data)
```

Library reticulate enables seamless integration with Python

```
library(reticulate)
source_python('https://raw.githubusercontent.com/eogasawara/analise-dados/refs/heads/main/python/retic.py')
x <- add(5, 10)
x

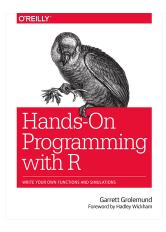
## [1] 15
```

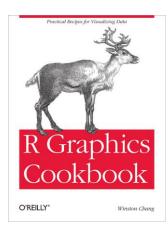
#### **Python** + **R** dataset integration

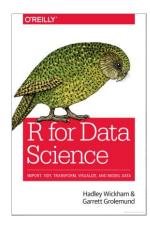
```
data <- data.frame(x = c(1:5), y=c(11:15))
dfm <- read rdata mem(data)
                                                                                                         Q
## 0
       1
## 1
       2
      3
## Name: x, dtype: int32
head(dfm)
## x y z
## 1 1 11 12
## 2 2 12 14
## 3 3 13 16
## 4 4 14 18
## 5 5 15 20
```

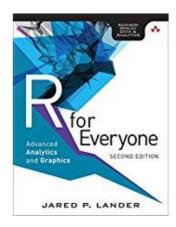
#### **Main References**

Código: <a href="https://github.com/eogasawara/analise-dados/blob/main/examples/2-R-Basics.md">https://github.com/eogasawara/analise-dados/blob/main/examples/2-R-Basics.md</a>









Further reading: Hands-On Programming with R, Chapters 1-12

Slides e vídeos em: https://eic.cefet-rj.br/~eogasawara/analise-de-dados



- https://rstudio-education.github.io/hopr/index.html
- https://r-graphics.org