## 

Mattia Guerini

2020/2021 - fall semester

Introduction

Data structures

#### Schedule

- ▶ 08th of September 13-16
- ▶ 15th of September 9-12
- ▶ 22th of September 9-12

#### Rules of the game

- arrive on time
- ▶ 20 minutes break
- ▶ homeworks
- ▶ no book (plenty of open source resources on-line)
- ▶ slides https://github.com/mattiaguerini/slides-intro-to-R
- ▶ take home exam (short project)



#### What is R

R is both a programming language and software environment for statistical computing, which is free and open-source.

The *R Project* was initiated by Robert Gentleman and Ross Ihaka (University of Auckland) in the early 1990s as a different implementation of the S language.

Since 1997, R has been developed by the R Development Core Team.

R is platform independent and can run on Microsoft Windows, Mac OS and Unix/Linux systems.

#### Getting Started

To get started, you'll need to install two pieces of software:

- R, the actual programming language. https://cran.r-project.org
- ► RStudio, an excellent IDE for working with R. https://www.rstudio.com

#### Why RStudio?<sup>1</sup>

- ► Easier to use (everything is in one space)
- ► Many useful integrations (e.g. R-projects, R-markdown, shiny . . . )
- $\triangleright$  Plenty of shortcuts (alt + shift + k)
- ▶ Plenty of cheatsheets (see top panel)

<sup>&</sup>lt;sup>1</sup>You must have installed R before using RStudio.

#### Screenshot of RConsole

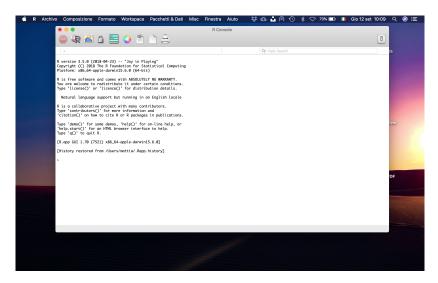


Figure 1: RConsole

#### Screenshot of RStudio

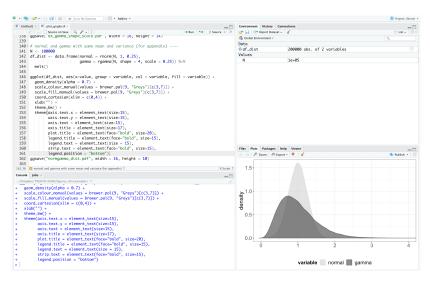


Figure 2: RStudio

### Glossary

- ▶ command: user input (text or numbers) that R understands
- ▶ *script*: a sequence of commands collected in a text file, each separated by a new line
- environment: a list of named variables that we have generated/imported by means of a series of commands
- ▶ history: the list of past commands thaty we have used
- ▶ help: a documentation of all the functions available in R (the user manual)
- ▶ package: a collection of additional functions and dataset

### R as a calculator (I)

```
2+2
## [1] 4
2-2
## [1] 0
2*2
## [1] 4
2/2
## [1] 1
```

# R as a calculator (II)

```
log(1)
## [1] 0
exp(1)
## [1] 2.718282
log(exp(1))
## [1] 1
sqrt(25)
## [1] 5
```

## The help

```
?log
help(log)
```

#### Otherwise:

- ► Google your error message
- ► Ask for help in Stack Overflow

#### Packages

R comes with a number of built-in functions and datasets, but one of the main strengths of R as an open-source project is its package system.

Packages add additional functions and datasets.

Frequently if you want to do something in R, and it is not available by default, there is a good chance that there is a package that will fulfill your needs.

You can install packages using the command install.packages()

You can load packages using the command library()

## Data structures

#### Data types

- ► Numeric/Double (e.g. 2.5, 1/5, 1.0, ...)
- ► Integer (e.g. 1, 2, 3, ...)
- ightharpoonup Complex (e.g. 1 + 2i, ...)
- ► Logical (e.g. TRUE, FALSE or NA)
- ► Character (e.g. "a", "b", "paper", ...)
- ► Factor/Categorical ("male", "female", ...)

#### Data structures

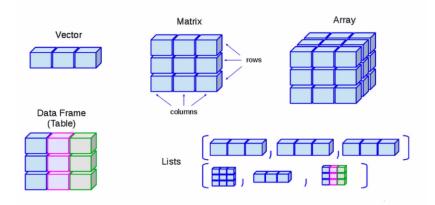


Figure 3: Visualization of data structures

### Vectors (I)

You can create a vector using the command c()

```
x <- c(1, 3, 5, 10)
x
```

## [1] 1 3 5 10

Vectors must contain elements of the same data type.

```
c(1, "intro", TRUE)
```

```
## [1] "1" "intro" "TRUE"
```

You can measure the length of a vector using the command length()

```
length(x)
```

```
## [1] 4
```

## Vectors (II)

```
It is also possible to easily create sequences
1:10
## [1] 1 2 3 4 5 6 7 8 9 10
seq(from = 1, to = 2, by = 0.1)
## [1] 1.0 1.1 1.2 1.3 1.4 1.5 1.6 1.7 1.8 1.9 2.0
rep("A", times = 5)
## [1] "A" "A" "A" "A" "A"
```

#### Vectors (III)

You can combine different vectors

```
x <- 1:3 # from 1 to 3
y <- c(10, 15) # 10 and 11
z \leftarrow c(x,y) \# x \text{ first and then } y
z
## [1] 1 2 3 10 15
And you can repeat vectors (or its elements)
z <- rep(y, each=3) # repeat each element 3 times
z
## [1] 10 10 10 15 15 15
z <- rep(y, times=3) # repeat the whole vector 3 times
z
## [1] 10 15 10 15 10 15
```

## Subsetting Vectors

```
x \leftarrow c(1,5,10,7)
x < 6 # elements lower than 6?
## [1] TRUE TRUE FALSE FALSE
x == 10 \# elements equal to 10?
## [1] FALSE FALSE TRUE FALSE
x[2] # element in the second position?
## [1] 5
x[1:2] # elements in the first 2 positions?
## [1] 1 5
x[c(1,3,4)] # elements in position 1, 3, 4?
## [1] 1 10 7
```

# Vectors' Operations

## [1] 1 25 100 49

```
x <- c(1,5,10,7)
x+2 # adds a scalar to all elements
## [1] 3 7 12 9
x^2 # squares all elements</pre>
```

## Matrices (I)

```
You can create a matrix using the command matrix()
```

```
X <- matrix(1:9, nrow = 3, ncol = 3)
X
## [,1] [,2] [,3]</pre>
```

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

### Matrices (II)

R automatically inserts elements by columns, but we can ask to include by rows

```
X <- matrix(1:9, nrow = 3, ncol = 3, byrow = TRUE)
X

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

You don't even have to specify the options names

```
X <- matrix(1:8, 2, 4, T)
X
```

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

### Matrices (III)

Matrices can also be created by combining vectors

```
X <- cbind(1:4, 6:9) # binds them as columns
X
## [,1] [,2]
## [1,]
## [2,] 2 7
## [3,] 3 8
## [4,] 4 9
X <- rbind(1:4, 6:9) # binds them as rows
X
## [,1] [,2] [,3] [,4]
## [1,] 1 2 3
## [2,] 6 7
```

### Subsetting Matrices

```
X>5 # elements larger than 5
## [,1] [,2] [,3] [,4]
## [1,] FALSE FALSE FALSE FALSE
## [2,] TRUE TRUE TRUE TRUE
X[1,4] # element of first row, fourth column?
## [1] 4
X[1,] # element in the first row?
## [1] 1 2 3 4
X[,2] # elements in the second columns?
## [1] 2 7
```

## Matrices' Operations (I)

 $x \leftarrow c(1,5,4,9)$ 

```
Let's create two matrices X and Y:
```

```
y \leftarrow c(2,4,1,3)
X <- matrix(x, 2, 2)</pre>
Y \leftarrow matrix(y, 2, 2)
X
## [,1] [,2]
## [1,] 1 4
## [2,] 5 9
## [,1] [,2]
## [1,] 2 1
## [2,] 4 3
```

## Matrices' Operations (II)

```
X+Y # element by element (also subtraction is equal)
## [,1] [,2]
## [1,] 3 5
## [2,] 9 12
X*Y # element by element multiplication
## [,1] [,2]
## [1,] 2 4
## [2,] 20 27
X%*%Y # matrix multiplication
## [,1] [,2]
## [1,] 18 13
## [2,] 46 32
```

## Matrices' Operations (III)

## [2,] 4 9

```
solve(Y) # inverse

## [,1] [,2]
## [1,] 1.5 -0.5
## [2,] -2.0 1.0

t(X) # transpose

## [,1] [,2]
## [1,] 1 5
```

## Arrays (I)

## [1,] 3 1 3 ## [2,] 4 2 4

```
x < -1:4
X \leftarrow \operatorname{array}(\operatorname{data} = x, \operatorname{dim} = c(2,3,2))
Х
## , , 1
##
## [,1] [,2] [,3]
## [1,] 1 3 1
## [2,] 2 4 2
##
## , , 2
##
## [,1] [,2] [,3]
```

#### Notes about the Arrays

- ▶ Remember that vectors, matrices and arrays can include only data types of the same kind.
- ▶ An 3D array is basically a combination of matrices each laid on top of other (e.g. write N matrix in N different pages in your notebook)
- ▶ A 4D array is basically a combination of arrays each laid on top of other (e.g. take two notebooks of 3D arrays)
- ► A 5D array ...
- ▶ Pay attention to the recycling rule

#### Lists

A list is a one-dimensional heterogeneous data structure.

It is indexed like a vector with a single integer value (or a name), but each element can contain an element of any data type.

```
x <- 1:4
y <- c("a", "b", "c")
L <- list(numbers = x, letters = y)
L</pre>
```

```
## $numbers
## [1] 1 2 3 4
##
## $letters
## [1] "a" "b" "c"
```

### Subsetting Lists

```
L[[1]] # extract the first element
## [1] 1 2 3 4
L$numbers # extract the element called numbers
## [1] 1 2 3 4
I.$letters # extract the element called letters
## [1] "a" "b" "c"
You can even "work" with the subsetted element:
L$numbers[1:3] > 2
## [1] FALSE FALSE TRUE
```

## Data Frames (I)

A data.frame is similar to a typical spreadsheet in excel.

There are rows, and there are columns.

A row is typically thought of as an *observation*.

A column is a certain *variable*, characteristic or feature of that observation.

### Data Frames (II)

A data frame is a list of column vectors where:

- each column has a name
- each column must contain the same data type, but the different columns can store different data types.
- each column must be of same length

## Data Frames (III)

```
set.seed(1)
df <- data.frame(id = 1:5,
    name = c("Ted", "Barney", "Lily", "Marshall", "Robin"),
    surname = c("Mosby", "Stinson", "Aldrin", "Eriksen", "Scherbatsky"),
    wage = rnorm(n=5, mean = 1000, sd = 100), # normal random sample
    origin = c("Cleveland", "New York", "New York", "St. Cloud", "Canada"),
    male = c(T, T, F, T, F)
    )
df</pre>
```

```
##
    id
                              wage origin male
          name
                   surname
## 1 1
           Ted
                    Mosby 937.3546 Cleveland TRUE
## 2 2 Barney
                   Stinson 1018.3643 New York TRUE
## 3 3
          Lilv
                Aldrin 916.4371 New York FALSE
## 4 4 Marshall
                   Eriksen 1159.5281 St. Cloud TRUE
         Robin Scherbatsky 1032.9508 Canada FALSE
## 5 5
```

You can verify the size of the data.frame using the command dim()

You can get the data type info using the command str()

# Subsetting Data Frames (I)

Robin Canada

## 5

```
df$name # subset a column

## [1] Ted Barney Lily Marshall Robin
## Levels: Barney Lily Marshall Robin Ted

df[,c(2,5)] # can also subset like a matrix

## name origin
## 1 Ted Cleveland
## 2 Barney New York
## 3 Lily New York
## 4 Marshall St. Cloud
```

## Subsetting Data Frames (II)

name

surname

## 4 4 Marshall Eriksen 1159.5281 St. Cloud TRUE

##

## 3 3

id

## 5 5

```
head(df, n=3) # first n observations

## id name surname wage origin male
## 1 1 Ted Mosby 937.3546 Cleveland TRUE
## 2 2 Barney Stinson 1018.3643 New York TRUE
## 3 3 Lily Aldrin 916.4371 New York FALSE
tail(df, n=3) # last n observations
```

Lily Aldrin 916.4371 New York FALSE

Robin Scherbatsky 1032.9508 Canada FALSE

wage origin male

## Inspecting data frames (I)

R comes with many data bases included. These can be used for learning R.

One of the most famous is the one called mtcars.

```
head(mtcars)
```

```
##
                    mpg cyl disp hp drat wt qsec vs am gear carb
## Mazda RX4
                   21.0
                         6 160 110 3.90 2.620 16.46
                                                               4
## Mazda RX4 Wag
                   21.0
                         6 160 110 3.90 2.875 17.02 0 1
                                                               4
## Datsun 710
                   22.8
                         4 108 93 3.85 2.320 18.61 1 1
## Hornet 4 Drive
                   21.4 6 258 110 3.08 3.215 19.44 1 0
## Hornet Sportabout 18.7 8 360 175 3.15 3.440 17.02 0 0
## Valiant
                   18.1
                         6 225 105 2.76 3.460 20.22 1
dim(mtcars)
```

```
## [1] 32 11
```

### Inspecting data frames (II)

[11] "carb"

```
str(mtcars)
## 'data.frame': 32 obs. of 11 variables:
   $ mpg : num 21 21 22.8 21.4 18.7 18.1 14.3 24.4 22.8 19.2 ...
   $ cyl : num 6 6 4 6 8 6 8 4 4 6 ...
##
##
   $ disp: num 160 160 108 258 360 ...
   $ hp : num 110 110 93 110 175 105 245 62 95 123 ...
##
   $ drat: num 3.9 3.9 3.85 3.08 3.15 2.76 3.21 3.69 3.92 3.92 ...
##
##
   $ wt : num 2.62 2.88 2.32 3.21 3.44 ...
   $ qsec: num 16.5 17 18.6 19.4 17 ...
##
##
   $ vs : num 0 0 1 1 0 1 0 1 1 1 ...
##
   $ am : num 1 1 1 0 0 0 0 0 0 0 ...
   $ gear: num 4 4 4 3 3 3 3 4 4 4 ...
##
##
   $ carb: num 4 4 1 1 2 1 4 2 2 4 ...
names(mtcars)
```

[1] "mpg" "cyl" "disp" "hp" "drat" "wt" "qsec" "vs"

"gear"

## Subsetting data frames (III)

We are interesting in the cylinders and the weights of inefficient cars (lower than 15 miles per gallon).

```
poll_cars <- mtcars[mtcars$mpg<15, c("cyl", "wt")]
poll_cars</pre>
```

```
## cyl wt
## Duster 360 8 3.570
## Cadillac Fleetwood 8 5.250
## Lincoln Continental 8 5.424
## Chrysler Imperial 8 5.345
## Camaro Z28 8 3.840
```

## Subsetting data frames (IV)

#### Alternatively:

```
poll_cars <- subset(mtcars, subset = mpg<15, select = c("cyl", "wt"))
poll_cars</pre>
```

```
## cyl wt
## Duster 360 8 3.570
## Cadillac Fleetwood 8 5.250
## Lincoln Continental 8 5.424
## Chrysler Imperial 8 5.345
## Camaro Z28 8 3.840
```

## Importing data frames (I)

#### You can import csv data using:

```
setwd("~/Google Drive/T_2020a_UCA_introR/data/")
nyc_ab <- read.csv("AB_NYC_2019.csv")</pre>
```

#### where:

- ▶ setwd() sets the working directory to the place where the data is saved;
- read.csv() loads the csv file with the specified name.