

Introduction to R

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<https://aweimann.github.io/floto-lab-learning-bioinformatics/docs/>



The goal

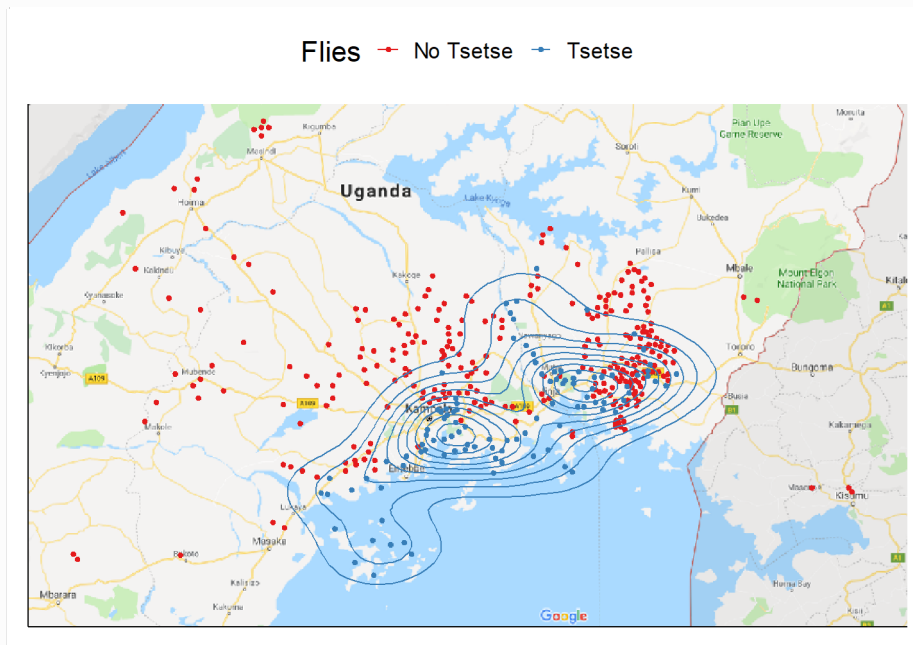
Learning to program is learning to think

Learning R in a workshop is an impossible task. Hence, our goal is to illustrate the capabilities of this tool. How programming can make your work much easier.

Why learn R?

A way to represent ideas

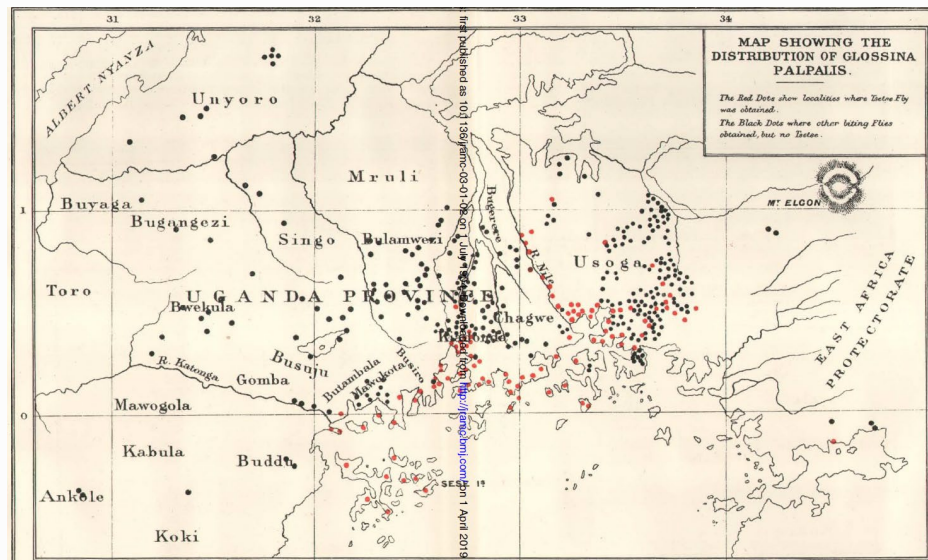
Age	Sex	District	Shamba	Trypanosoma	Filaria
25	M	Sese Island	Sewana	+	+
20	M	Sese Island	Kaganda I	-	+
25	M	Sese Island	Semagala I	+	+
30	M	Sese Island	Kaganda I	-	+
20	M	Sese Island	Semagala I	-	+
25	M	Sese Island	Buvovu I	-	+
25	M	Sese Island	Kaganda I	-	+
20	M	Sese Island	Semagala I	-	+
30	M	Sese Island	Buvu I	-	+
35	M	Sese Island	Semagala I	-	-
20	M	Sese Island	Semagala I	-	-
.....
.....
.....
25	M	Sese Island	Semagala I	+	-
35	M	Sese Island	Bunami I	+	+



Why learn R?

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25	M	Sese Island	Kaganda I	-	+
20	M	Sese Island	Semagala I	-	+
30	M	Sese Island	Buvu I	-	+
35	M	Sese Island	Semagala I	-	-
20	M	Sese Island	Semagala I	-	-
.....
.....
.....
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Dr D. Bruce, 1903

Why learn R?

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30	M	Sese Island	Buvu I	-	+
35	M	Sese Island	Semagala I	-	-
20	M	Sese Island	Semagala I	-	-
.....
.....
.....
25	M	Sese Island	Semagala I	+	-
35	M	Sese Island	Bunami I	+	+



```

1 library(ggplot2)
2 library(magick)
3 library(readxl)
4 library(ggpubr)
5 library(data.table)
6
7 map_u <- image_read("Figures/Mapa_uganda.png")
8
9 sick_coord <- read_xlsx(path = "Datasets/D_Bruce_map_dataset_v1.xlsx",
10 sheet = "Sickness") %>% data.table
11 Flies_coord <- read_xlsx(path = "Datasets/D_Bruce_map_dataset_v1.xlsx",
12 sheet = "Flies") %>% data.table
13
14 # Plot 1
15
16 image_ggplot(map_u) +
17 geom_point(data = Sick_coord, mapping = aes(x = X, y = Y, col = Sickness)) +
18 geom_density_2d(data = Sick_coord[Sickness == "Present"],
19 mapping = aes(x = X, y = Y, col = Sickness)) +
20 scale_color_brewer(palette = "Set1") +
21 labs(x = "", y = "") +
22 theme_pubr(legend = "top", base_size = 20) +
23 theme(axis.title.x = element_blank(),
24 axis.text.x = element_blank(),
25 axis.ticks.x = element_blank(),
26 axis.title.y = element_blank(),
27 axis.text.y = element_blank(),
28 axis.ticks.y = element_blank())
29
30 # Plot 2
31
32 image_ggplot(map_u) +
33 geom_point(data = Flies_coord, mapping = aes(x = X, y = Y, col = Flies)) +
34 geom_density_2d(data = Flies_coord[Flies == "Tsetse"],
35 mapping = aes(x = X, y = Y, col = Flies)) +
36 scale_color_brewer(palette = "Set1") +
37 labs(x = "", y = "") +
38 theme_pubr(legend = "top", base_size = 20) +
39 theme(axis.title.x = element_blank(),
40 axis.text.x = element_blank(),
41 axis.ticks.x = element_blank(),
42 axis.title.y = element_blank(),
43 axis.text.y = element_blank(),
44 axis.ticks.y = element_blank())
45
46
  
```

The cycle of data analysis

One ring to model them all



What's R

A tool to make life *easier*



- Dynamic programming
- Open source
- Active community
- Endless packages



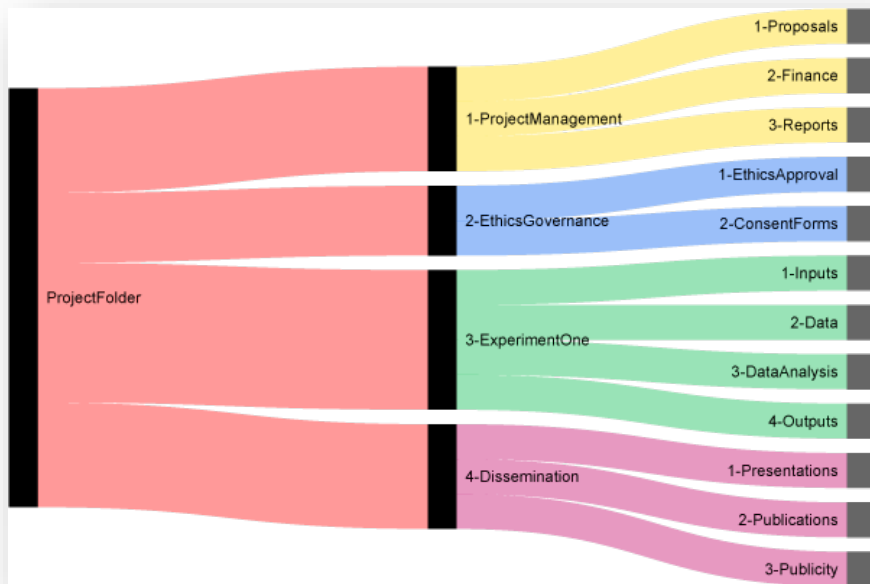
What's R?

How was my first experience

- 1
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- 32

Arrange files and folders

Thinking in hierarchical order



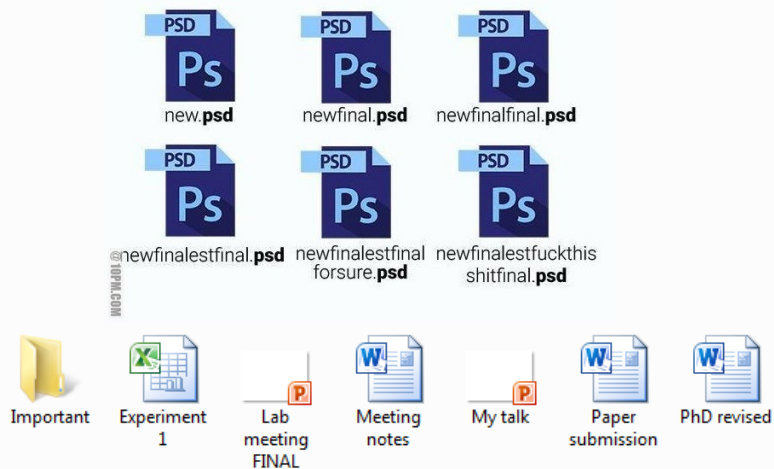
67 items 1 item selected Available when o

© http://nikola.me/folder_structure.html

Arrange files and folders

Naming... really matters?

Copyright: <http://10pm.com/>



What I do

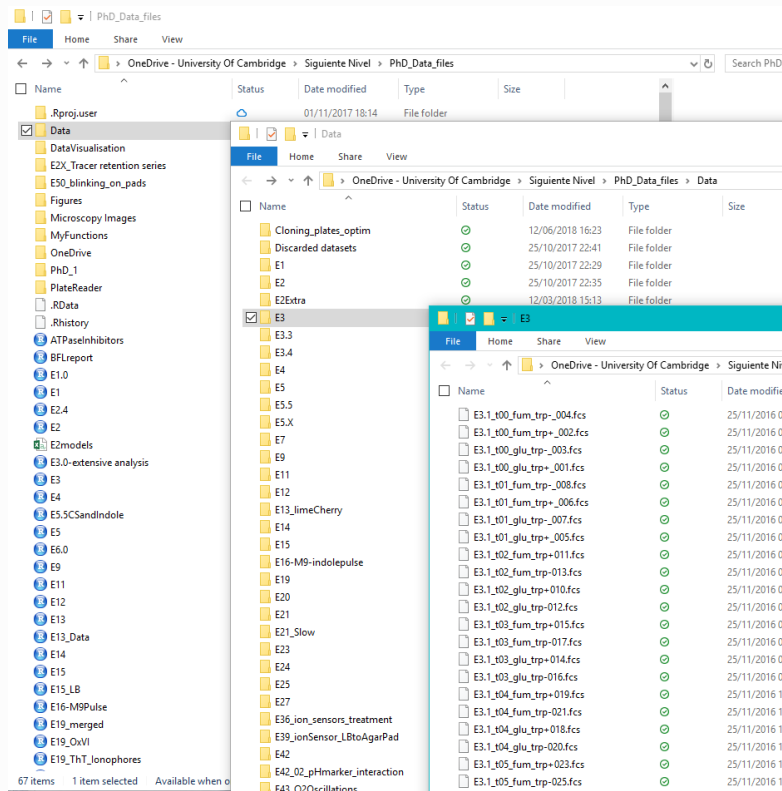
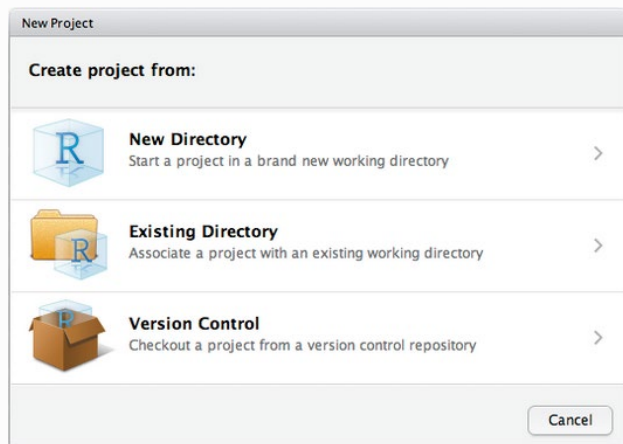
EXP_AAAAMMDD_NOMBRE_VX.X.csv

Will we remember what does it means at the end of the thesis or Project?

Avoid: ?,\$,%,^,&,*,(.)-#?,<,>./|\[\],{and};

RStudio

A helper in our quest



Inside R

RStudio

The screenshot displays the RStudio interface with four main panels:

- Editor:** A text editor window showing a single line of code: `1 |`.
- Environment:** A panel showing the current environment, which is empty: "Environment is empty".
- Terminal:** A console window showing the R version and copyright information: "R version 3.5.3 (2019-03-11) -- 'Great Truth' Copyright (C) 2019 The R Foundation for Statistical Computing Platform: x86_64-w64-mingw32/x64 (64-bit)". It also includes instructions on how to use R, such as `license()`, `contributors()`, `demo()`, `help.start()`, and `q()`.
- Files:** A file explorer showing the contents of the current project directory, including files like `.gitignore`, `.Rhistory`, `_config.yml`, `_site.yml`, `docs`, `foto-lab-learning-bioinformatics.Rproj`, `index.Rmd`, `intro_to_stats.Rmd`, `introduction.to.R.Rproj`, `introduction.to.R.Rproj`, `presentations`, and `README.md`.

Red text labels are overlaid on the image to identify the panels:

- Editor** (over the Editor panel)
- Objects and variables** (over the Environment panel)
- Terminal** (over the Console panel)
- Plots, files, packages....** (over the Files panel)

Inside R

Importing data

The screenshot displays the RStudio environment with the 'Import Dataset' dialog box open. The 'Import Dataset' menu item in the top toolbar is circled in red. The dialog box contains the following elements:

- File/URL:** A text input field with a 'Browse...' button.
- Data Preview:** A large empty rectangular area for previewing the data.
- Import Options:**
 - Name:
 - Skip:
 - First Row as Names
 - Trim Spaces
 - Open Data Viewer
 - Delimiter:
 - Quotes:
 - Escape:
 - Comments:
 - Locale:
- Code Preview:**

```
library(readr)
dataset <- read_csv(NULL)
View(dataset)
```

The console at the bottom left shows the R startup message:

```
R version 3.5.3 (2019-03-
Copyright (C) 2019 The R
Platform: x86_64-w64-mt
R is free software and co
You are welcome to redistrib
Type 'license()' or 'lic
R is a collaborative pro
Type 'contributors()' for
'citation()' on how to c
Type 'demo()' for some de
'help.start()' for an HTML
Type 'q()' to quit R.
> |
```

Inside R

Exploring datasets

The screenshot displays the RStudio interface with the following components:

- Environment:** Shows the loaded data object 'd' with 32 observations and 11 variables. The variable 'x' is shown with the value 'nature'.
- Console:** Contains the following R code and output:


```

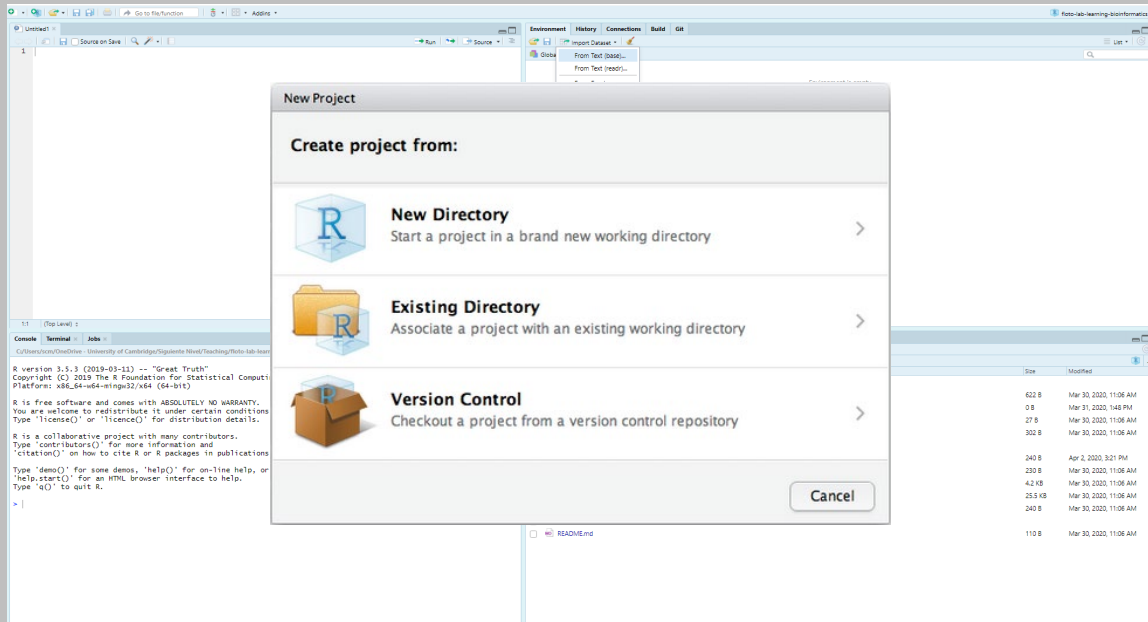
      > library(data.table)
      data.table 1.12.2 using 2 threads (see ?getDTthreads). Latest news: r-datatable.com
      > d <- data.table(mtcars)
      > view(d)
      > print("I'm sure you are not reading this!")
      
```
- Files Panel:** Lists files in the current directory, including 'gignore', 'RData', '.Rhistory', '_config.yml', '_site.yml', 'docs', 'Foto-lab-learning-bioinformatics.Rproj', 'index.Rmd', 'intro_to_stats.Rmd', 'introduction.to.R.Rmd', 'introduction.to.R.Rproj', 'presentations', and 'README.md'.

Two red arrows point to the 'mpg' and 'carb' columns in the data table and the 'd' variable in the Environment pane.

	mpg	cy	disp	hp	drat	wt	qsec	vs	am	gear	carb
1	21.0	6	160.0	110	3.90	2.620	16.46	0	1	4	4
2	21.0	6	160.0	110	3.90	2.875	17.02	0	1	4	4
3	22.8	4	101.0	93	3.85	2.320	18.61	1	1	4	1
4	21.4	6	258.0	110	3.08	3.215	19.44	1	0	3	1
5	18.7	8	360.0	175	3.15	3.440	17.02	0	0	3	2
6	18.1	6	225.0	105	2.76	3.460	20.22	1	0	3	1
7	14.3	8	360.0	245	3.21	3.570	15.84	0	0	3	4
8	24.4	4	146.7	62	3.69	3.190	20.00	1	0	4	2
9	22.8	4	140.8	95	3.92	3.150	22.90	1	0	4	2
10	19.2	6	187.6	123	3.92	3.440	18.30	1	0	4	4
11	17.8	6	187.6	123	3.92	3.440	18.90	1	0	4	4
12	16.4	8	275.8	180	3.07	4.070	17.40	0	0	3	3
13	17.3	8	275.8	180	3.07	3.730	17.60	0	0	3	3
14	15.2	8	275.8	180	3.07	3.780	18.00	0	0	3	3
15	10.4	8	472.0	205	2.90	5.250	17.96	0	0	3	4
16	10.4	8	460.0	215	3.00	5.424	17.82	0	0	3	4
17	14.7	8	440.0	230	3.23	5.345	17.42	0	0	3	4
18	32.4	4	78.7	66	4.08	2.200	19.47	1	1	4	1
19	30.4	4	75.7	52	4.93	1.615	18.52	1	1	4	2
20	33.9	4	71.5	65	4.22	1.835	19.90	1	1	4	1

Time to program

For example



The foundation of R

The heir of mathematical notation

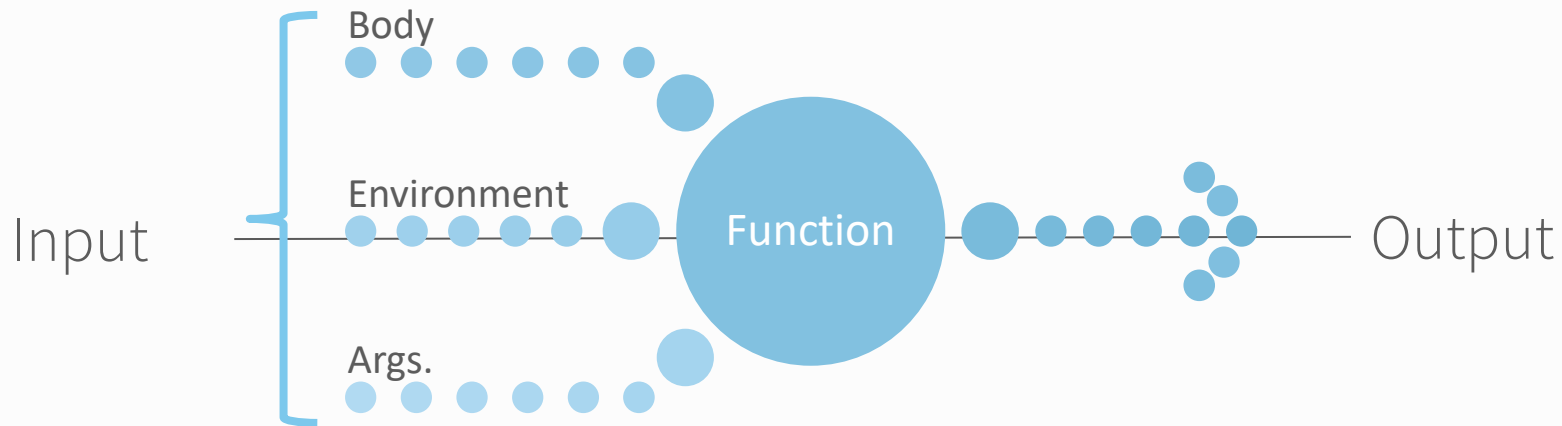
Input $\xrightarrow{f(x)}$ Output

$$A = f(r) = \pi r^2$$

$$S = \frac{dA}{dr} = \frac{df(r)}{dr} = 2\pi r$$

The foundation of R

Functional programming



R like a super-calc on steroids

But... what's a **function**?

```

Action <- function(x, y = 0) {
  z <- x + y
  return(z)
}

```

specify "x" and "y"

Default "y" when we don't specify it

Add "x" e "y"

Return the result

The 3 elements

- **Args:** List of items, specified by **order** or **name**.
- **Body:** Transformations upon the arguments.
- **Env:** Variables accessible within the body.

R like a super-calc on steroids

But... what's a **function**?

```
# Intro to functions

runif(n = 10, min = 3, max = 10) # This function generates 10 random numb between 3 and 10
runif(10, 3, 10)                 # Same but using argument position as reference

runif(n = 10)                    # Get 10 random numbers within the default range
runif(10)                        # 10 random numbers within the default range setting n by position

runif(min = -1, n = 3)          # Using the name allows the order to be alter

# Basic math functions
1 + 8
2 - 9
2 * 3
5/3
```

R like a super-calc on steroids

Logic operations

```
x == y           # Is x equal to y
identical(x, y)  # Same as == but using a function
!x              # Negate (or logic inversión) of x
x & y           # Is x AND y True?
x | y           # Is x OR y True?
x < y           # Is X Lower than y
x > y           # Is X Bigger than y
x <= y          # Is X Lower or equal than y
x >= y          # Is X Bigger or equal than y
x != y          # Is x different from y
xor()           # Is only X or Y True?
isTrue ()       # Is true?
```

R like a super-calc on steroids

In R you have a function for everything

```
A <- ...
c(object, object,...)
length(object)
str(object)
class(object)
names(object)
rm(object)
mean(vector)
median(vector)
sd(vector)
sqrt(vector)
log(vector)
exp(vector)

summary()
getwd()
read.delim()

# Assigning a variable
# Concatenate SAME type objects within a vector
# Number of elements in an object
# Show the internal structure of an object
# Show the class of an object
# Names of the elements
# Remove the object
# Media
# Median
# Standard deviation
# Square root
# Natural logarithm
# E to the power of..

# Summary
# Where we are working
# Import files (also read.csv, read.csv2, read.txt, read.table)
```

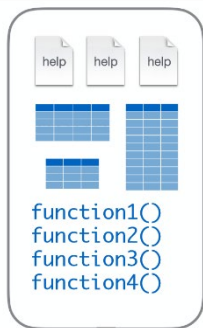
Time to program

For example

```
mean(1:5, 0.1)  
mean(x = 1:5, trim = 0.1)  
mean(1:5, trim = 0.1)  
mean(x = 1:5, 0.1)
```

The R packages

Families of functions comes in packages



Base R

[image taken from Ryan Wesslen presentation]

What they are

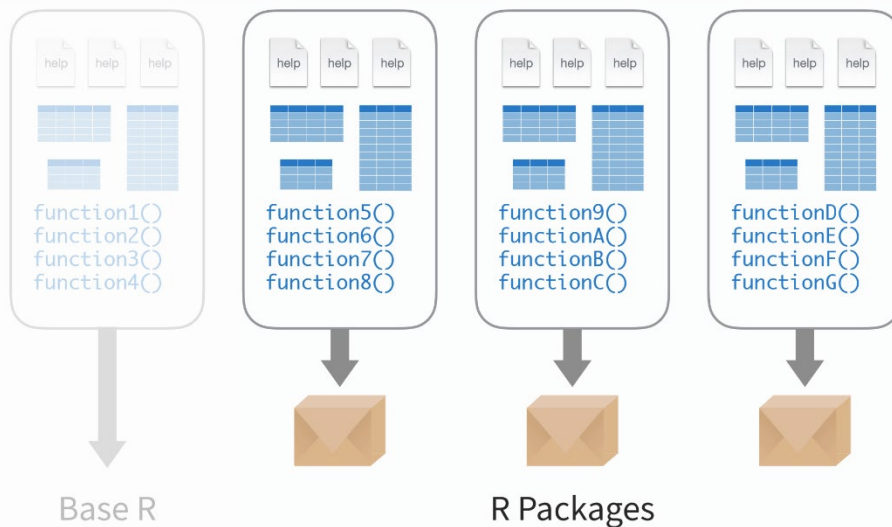
- The R community that develops related functions adds them in a library that other users can download and install.
- This allows us to start our work where others finished and focus our efforts on a single problem.
- They are usually accompanied by instruction books called " **vignette** " that explain how to use the functions of the library.
-

```
install.packages(data.table) # Install library from CRAN
devtools::install_github("/paul-buerkner/brms") # Install from GitHub
```

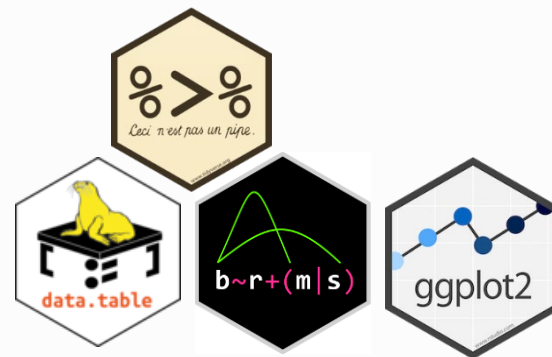
```
library(tidyverse) # Load library
data.table::melt.data.table() # Calling a function from a
# library
```

The R packages

Families of functions comes in packages



[image taken from Ryan Wesslen presentation]



- In this workshop we will work with the following bookstores: data.table, magrittr, ggplot, glmTMB, mgcv
- **Beware of using too many libraries:**
- Overlapping functions with the same name
- Reproducibility on other computers

First contact

Representing mathematical objects

Vectors

$$X = [x_1, x_2, x_3, x_i \dots \dots x_n]$$

Numeric: \mathbb{R}

•0, -8, 3.14...

Logic: 1|0

•TRUE, FALSE, T, F

Character

•"Que", "es", "eso", "Eso es Queso"

```
1 + 1 # Hashtag to write down comments
Nombre_x <- valor # Define objects with the operator <-
y <- 1
```

Principles for naming a variable

- Start with a character, no irregular caraters (&, %, ^, ...)
- Brief and descriptive.
- For compound names connect with `_`

```
`<-` (x, 2) # Assign
x <- y <- 1 # Assign multiple elements
y <- c(1, -2, 8, 5, 5e5) # Manually define a vector
z <- c("A", "B", "C") # Character vectors
x <- c(FALSE, TRUE, F, T) # Logic
x <- c(1L, 2L, 43L) # Integer
```

First contact

Representing mathematical objects

Vectors

```
x + y
x * 3
x / y
x^3
1:5           # Dos puntos ":" para indicar secuencias
5:1
y <- c(1, -2, 8, 5, 5e5)
y[3]         # Extraer elementos
[1] 8
y[-2]       # Numeros negativos para excluir elementos
[1] 1 8 5 5e5
y[2:3]      # Extraer secuencia de elementos

# Nombrar elementos para crear diccionarios
y <- c("a" = 1, "b" = -2, "c" = 8, "d" = 5, "e" = 5e5)
y["d"]
[1] 5
y[y > 5]    # Extraer elementos usando test lógicos ( también ==, >=, <= )
  c      e
8e+00 5e+05
```

First contact

Representing mathematical objects

Recycling

```
# If two vectors have different length, the shorter one is recycled it will recycle to equal the length of the longest with the longest vector.
```

```
x = c(10, 20, 30)
```

```
y = c(1, 2, 3, 4, 5, 6, 7, 8, 9)
```

```
> y + x
```

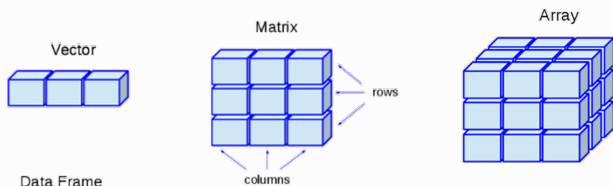
```
# R uses "y" as a 9-element vector and "x" will repeat it 3 times
```

```
[1] 11 22 33 14 25 36 17 28 39
```

First contact

Representing mathematical objects

Homogeneous structures



[source: University of Cantabria]

Principles for naming a variable

Within R, a matrix is a vector with two extra attributes:

- Rows
- Columns

```
mat <- matrix(c(1, -2, 8, 5, 7, 0, 3, 6, 9),
              nrow = 3, ncol = 3)

mat[4]                # Use array as vector
[1] 5

mat[1, 2]            # mat[fila, columna]
[1] 5

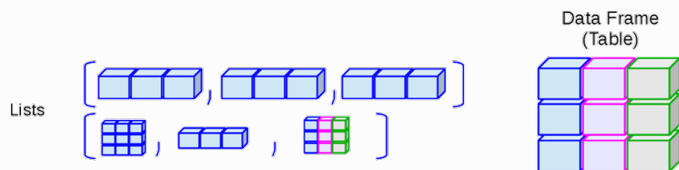
mat[, c(1, 3)]
  [,1] [,2]
[1,]  1   3
[2,] -2   6
[3,]  8   9

A <- array(data = 1:27, dim = c(3, 3, 3))
A[ 2, 2, 3]
[1] 23
```

First contact

Representing mathematical objects

Heterogeneous structures



[source: University of Cantabria]

```
mat <- matrix(c(1, -2, 8, 5), nrow = 2, ncol = 2)
mat[1, 2] # mat[filas, columnas]
[1] 8
Animals <- list("a" = c(1, 4, 5, 2),
               "b" = c("Cow", "Pig", "Chicken"))

Animals[["a"]][1] # [[ to access each item
Animals$a[1]      # $ to access every element
```

Lists and tables

- It contains structures of different types, or even contain a list
- They can also be of different length

Time to program

For example

```
x + y
x * 3
x / y
x^3
1:5 # Use colon ":" to indicate sequences
5:1
y <- c(1, -2, 8, 5, 5e5)
y[3] # Extract items
[1] 8
y[-2] # Negative numbers to delete elements
[1] 1 8 5 5e5
y[2:3] # Extract sequence of elements
# Name items to create "dictionaries"
y <- c("a" = 1, "b" = -2, "c" = 8, "d" = 5, "e" = 5e5)
y["d"]
[1] 5
y[y > 5] # Extract elements using logical tests ( también ==, >=, <= )
      c      e
8e+00 5e+05
```

Tables

Data.frame

- Intuitively, `data.frame` is the classic way to represent information in our minds.
- Like excel sheet
- Each column represents a vector
- Each row a case

Field Name	Area	Slope	Vegetation
Nash's Field	3.6	11	Grassland
Silwood Bottom	5.1	2	Arable
Nursery Field	2.8	3	Grassland
Rush Meadow	2.4	5	Meadow
Gunness' Thicket	3.8	0	Scrub
Oak Mead	3.1	2	Grassland
Church Field	3.5	3	Grassland

```

table <- data.frame(Field_N = c("Nash`s", "Silwood", "Nursery",
                                "Rush", "Gunness", "Oak Mead", "Church Field"),
                    Area = c(3.6, 5.1, 2.8, 2.4, 3.8, 3.1, 3.5),
                    Slope = c(11, 2, 3, 5, 0, 2, 3),
                    Vegetation = c("Grass", "Arabl", "Grass",
                                    "Meadow", "Scrub", "Grass", "Grass"))

# Create by adding pre-existing vectors

table <- data.frame(Field_N, Area, Slope, Vegetation)

# Join tables

table <- cbind(Field_N, Area, Slope, Vegetation)
table <- rbind(table_1, table_2)

# Also via read.*

table <- read.csv("path/to/my/file")

```

Tables

Data.frame

- Intuitively, `data.frame` is the classic way to represent information in our minds.
- Like excel sheet
- Each column represents a vector
- Each row a case

Field Name	Area	Slope	Vegetation	Soil pH
Nash's Field	3.6	11	Grassland	4.1
Silwood Bottom	5.1	2	Arable	5.2
Nursery Field	2.8	3	Grassland	4.3
Rush Meadow	2.4	5	Meadow	4.9
Gunness' Thicket	3.8	0	Scrub	4.2
Oak Mead	3.1	2	Grassland	3.9
Church Field	3.5	3	Grassland	4.2

```

head(table)           # Exploring head/tail values
tail(table)
str(table)

table$Area           # Expose values
mean(table$Slope)

summary(table)

table[1, 2]          # Access values as matrix
table[, 2:3]
table[, "nombre"]    # Access values via naming

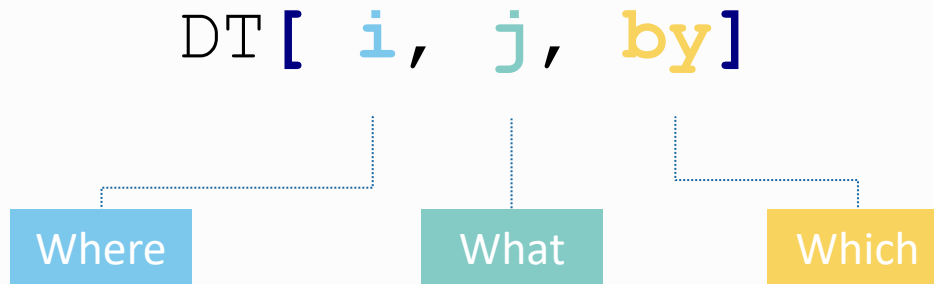
```


Data.table

An expansion to data.frame

Properties

- Syntax and usage is similar to `data.frame`
- It's extremely fast
- It offers tools for:
 - Data aggregate
 - Update cells
 - Join tables
- Allows elegant notation
- **No dependencies**



Data.table

Continuation

```
matrix(data.table)
d <- data.table(mtcars, keep.rownames = TRUE)

>
      rn  mpg  cyl  disp  hp  drat   wt   qsec vs  am gear carb
1:   Mazda RX4 21.0   6 160.0 110 3.90 2.620 16.46 0  1    4    4
2:   Mazda RX4 Wag 21.0   6 160.0 110 3.90 2.875 17.02 0  1    4    4
3:   Datsun 710 22.8   4 108.0  93 3.85 2.320 18.61 1  1    4    1
4:  Hornet 4 Drive 21.4   6 258.0 110 3.08 3.215 19.44 1  0    3    1
5:  Hornet Sportabout 18.7   8 360.0 175 3.15 3.440 17.02 0  0    3    2
6:   Valiant 18.1   6 225.0 105 2.76 3.460 20.22 1  0    3    1
7:   Duster 360 14.3   8 360.0 245 3.21 3.570 15.84 0  0    3    4

# How to perform simple extraction operations?

d[cyl > 4] # Filter with a simple logical test

d[rn %in% c("Mazda RX4", "Hornet Sportabout")] # Elementos within a vector

d[rn %like% "Mazda"] # Extract similar elements
```

Data.table

Continuation

```
# Summarise data

d[, .(g_mean = mean(cyl)),           # By group
    by = gear]

d[, .(g_mean2 = mean(cyl)),         # Multiple groups
    by = .(gear, vs)]

d[, Ref_col_mean := mean(hp),       # Multiple-groups and create a new column
    by = .(gear, vs)]

d[, .("mean_hp", "mean_cyl") :=    # Multiple-groups and multiple columns
    .(mean(hp),
      mean(cyl),
      by = .(gear, vs)]

d[, mean(wt[vs == 0])/mean(wt[vs == 1])] # Vector within a column
```

Time to program

For example

```
# Summarise data

d <- data.table(mtcars)
d[, .(m = mean(cyl)),           # By group
     by = gear]

d[, .(m = mean(cyl)),           # Multiple groups
     by = .(gear, vs)]

d[, m_hp := mean(hp),           # Multiple-groups and create a new column
     by = .(gear, vs)]

d[, c("m_hp", "m_cyl") :=      # Multiple-groups and multiple columns
     .(mean(hp),
       mean(cyl)),
     by = .(gear, vs)]

d[, mean(wt[vs == 0])/mean(wt[vs == 1])] # Vector within a column
```

Pipes

Moving A `%>%` B



The `%>%` operator

- One of the most useful and powerful elements of R.
- The operator `"%>%"` helps structure the code and minimizes the creation of "transitional variables."
- Requires package `magrittr` or `tidyverse`.
- The basic idea:
 - `x %>% f` is `f(x)`
 - `x %>% f %>% g %>% h` is `h(g(f(x)))`

Placeholder

Object `%>%`

```
function1(.) %>%
function2(.) ->
result
```

Pipes

Moving A %>% B



The %>% operator

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- Requires package `magrittr` or `tidyverse`.
- The basic idea:
 - `x %>% f` is `f(x)`
 - `x %>% f %>% g %>% h` is `h(g(f(x)))`

Example: Representing averages by group

```
# Extract values
Subset <- table[table$a > "Value_x"]
# Sumity the information
mus <- aggregate(test ~ condition, Subset, FUN = mean)
stds <- aggregate(test ~ condition, Subset, FUN = sd)
summary_Ss <- merge(mus, stds, by = "condition")

# Create the graph
colnames(summary_Ss) <- c("condition", "mus", "stds")
ggplot(summary_Ss, aes(x = condition, y = mus)) +
  geom_point() +
  geom_errorbar(aes(ymin = mus - stds,
                    ymax = mus + stds)) +
  theme_bw()
```

Stack of lazy variables

Pipes

Moving A `%>%` B



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 - `x %>% f` is `f(x)`
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Example: Representing averages by group **Code Unreadable**

```
summary_Ss <- merge(aggregate(test ~ condition,
                              table[, table$a > "valor_x"],
                              FUN = mean),
                   aggregate(test ~ condition,
                              table[, table$a > "valor_x"],
                              FUN = sd), by = "condition")

# Create the graph
colnames(summary_Ss) <- c("condition", "mus", "stds")
ggplot(summary_Ss, aes(x = condition, y = mus)) +
  geom_point() +
  geom_errorbar(aes(ymin = mus - stds,
                   ymax = mus + stds)) +
  theme_bw()
```

Pipes

Moving A `%>%` B



The `%>%` operator

- One of the most useful and powerful elements of R.
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- Requires package `magrittr` or `tidyverse`.
- The basic idea:
 - `x %>% f` is `f(x)`
 - `x %>% f %>% g %>% h` is `h(g(f(x)))`

Example: Representing averages by group

```
table[a > "value_x" ] %>% # filter
  .[, .(mus = mean(test),
        stds = sd(test),
        by = condition) ] %>% # Calculate parameters

# Graphical representation
ggplot(., aes(x = condition, y = mus) +
  geom_point()
  geom_errorbar(aes(ymin = mus - stds,
                    ymax = mus + stds) +
  theme_bw()
```


Time to program

For example

```
install.packages(magrittr)
install.packages(ggplot2)
library(ggplot2)
library(magrittr)
library(data.table)
data("mtcars") # Load data
d <- data.table(mtcars, keep.rownames = T)

d[ cyl > 3 ] %>%
  ggplot(., aes(x = disp, y = wt)) +
  geom_point() +
  theme_bw()

d %>%
  lm(formula = disp ~ wt, data = .) %>%
  summary
```

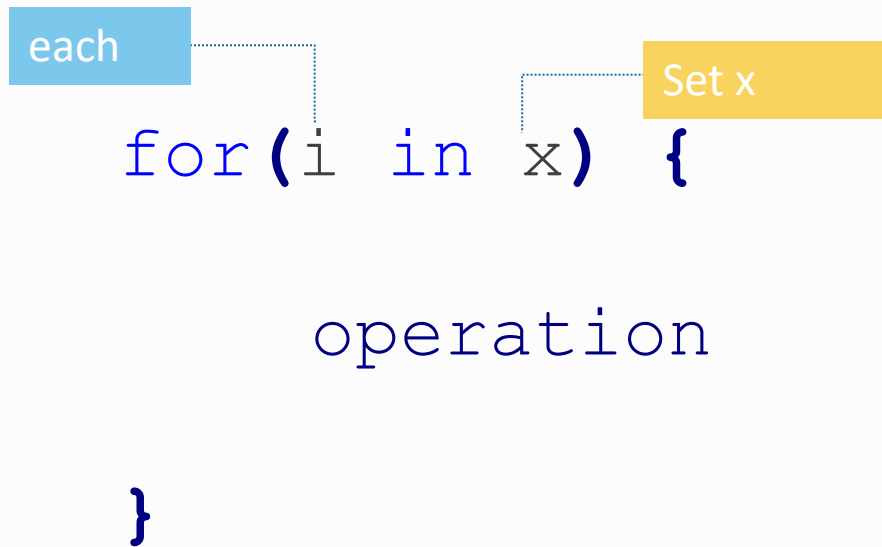
Loops and functionals

The essence of [programming](#)

What are they?

Computers are especially useful when the task requires repetition

- R provides us with three basic tools to repeat:
 - `For`
 - The of `*apply` family
- In addition, it is possible to filter with logical tests:
 - `if`
 - `ifelse`



Loops and functionals

An example

```
# Generate a sample dataset

set.seed(2018)

d <- data.frame(replicate(6, sample(c(1:10, -99), 100, rep = TRUE)))
names(d) <- letters[1:6]
head(d)
  a  b  c  d  e  f
1 4  7 -99 9  9  2
2 6  2  8 10  6 10
3 1 -99  9  3  4  1
4 3  7  7  7 -99  6
```

How do I calculate the mean of each column?

```
mean(d$a)
mean(d$b)
mean(d$c)
mean(d$d)
mean(d$e)
```

Loops and functionals

An example

```
# Looping with for

for (i in 1:ncol(d)) {                                # Using ":" to generate sequence of 1 to the number of columns
  x <- mean(d[[i]])
  print(x)
}

# If we want to save the result, first we create an empty vector (or other format)

medias <- rep(NA, ncol(d))

for (i in seq_along(medias)) {
  x <- mean(d[[i]])                                  # We iterate by column
  medias[i] <- x                                     # Save the result in the vector "means", position "i"
}

}
```

Loops and functionals

The essence of R

What are they?

Computers are especially useful when the task requires repetition

- R provides us with three basic tools to repeat:
 - `For`
 - The of `*apply` family
- In addition, it is possible to filter with logical tests:
 - `if`
 - `ifelse`

`*apply(x, Fun = f(x))`



To x apply function f

Loops and functionals

An example

```
# Iteration-*apply

medias <- apply(X = d, MARGIN = 2, FUN = mean)

> medias
      a      b      c      d      e      f
-4.26 -4.75 -1.85 -5.92 -4.43  1.40

# If we also want to specify other arguments, we can indicate them at the end

apply(X = d, MARGIN = 2, FUN = quantile, probs = c(0.1, 0.5, 0.9))

>   a  b c  d e f
10% 1 -9 1 -9 -9 1
50% 4  5 5  5 6 5
90% 9  9 9  9 10 10
```

Loops and functionals

Basic programming

What are they?

Computers are especially useful when the task requires repetition

- R provides us with three basic tools to repeat:
 - `For`
 - The of `*apply` family
- In addition, it is possible to filter with logical tests:
 - `if`
 - `ifelse`

```
ifelse(test = ***,  
       yes = Accion_A,  
       no  = Accion_B)
```

Loops and functionals

An example

```
# Basic if loop
if (paper_accepted == TRUE) {           # Notice, it only accepts 1 element at a time
  print( "We are the best!")
}

# Nested if loop
if (publisher == "science") {
  print ("We are the best")
} else if ( publisher == "arXiv" ){    # else if allow us to make another question
  print ("Andres will kill you" )
} else {                               # Only else resolve all the remaining cases
  print ("If you look for me, I am in the crying room")
}

# Iterate through if-test through a vector

all_papers <- c("accepted", "accepted", "accepted", "rejected")

ifelse(all_papers == "accepted",
       print("Fantastic, it's a science?"),
       print("If you look for me, I am in the crying room")
)
```


Loops and functionals

An example

```
Looping with *apply
d <- apply(d, 2, function(x) {ifelse(x == -99, NA, x)})
medias <- apply(X = d, MARGIN = 2, FUN = mean)

# Looping with *apply and the pipe operator

medias <- apply(d, 2, function(x) {tmp <- ifelse(x == -99, NA, x)}) %>%
  apply(X = ., MARGIN = 2, FUN = mean, na.rm = TRUE)

medias

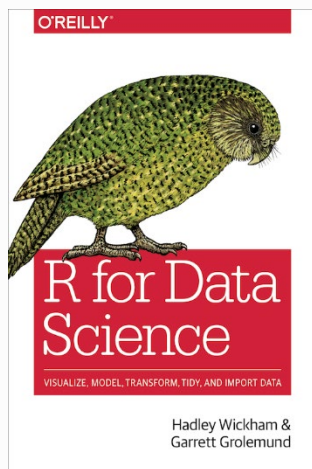
# looping with*apply

medias <- apply(d, 2, function(x) {tmp <- ifelse(x == -99, NA, x)
  mean(tmp, na.rm = TRUE)})

> medias
      X1      X2      X3      X4      X5      X6
5.585106 5.369565 5.263736 5.866667 5.423913 5.217391
```

Support channels

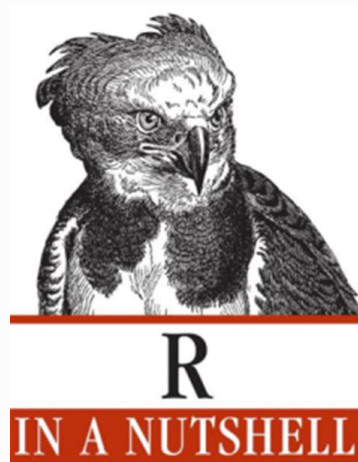
Online resources



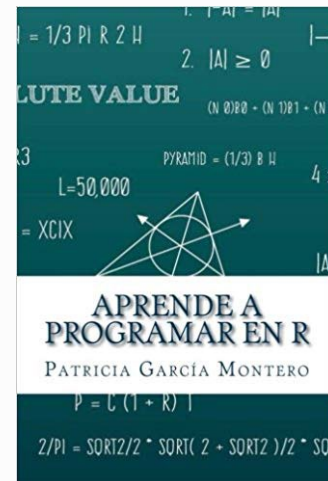
R for Data Science,
H. Wickham &
G. Golemund



Learning R,
R. Cotton



R in a nutshell,
J. Adler



Aprender a
programar en R
P. García Montero



¡Gracias por
¿Preguntas?
vuestro tiempo!