Introduction to R

Santiago Caño Muñiz, Dr Aaron Weimann, Dr Chris Ruis

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

The goal

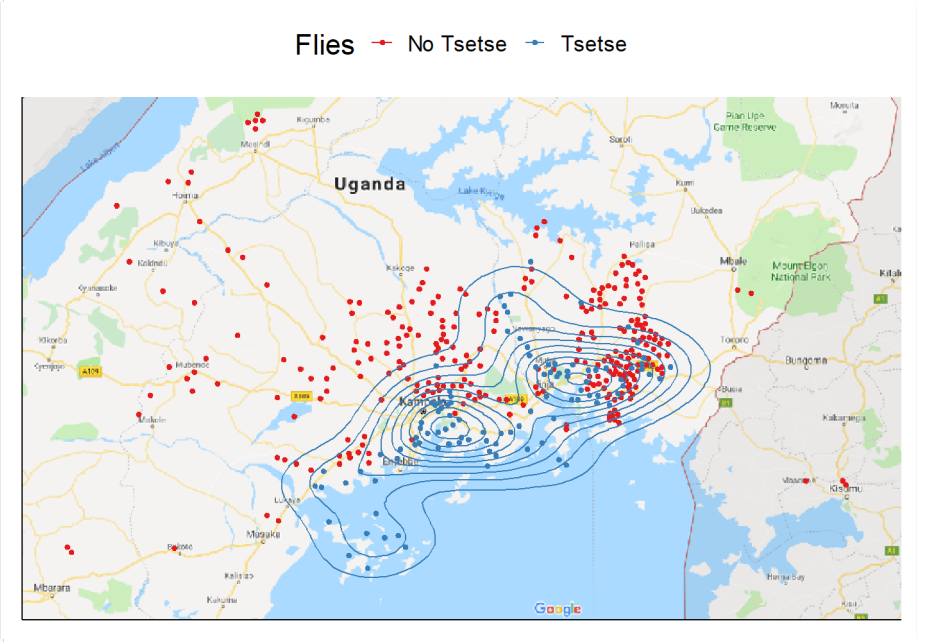
Learning to program is learning to think

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

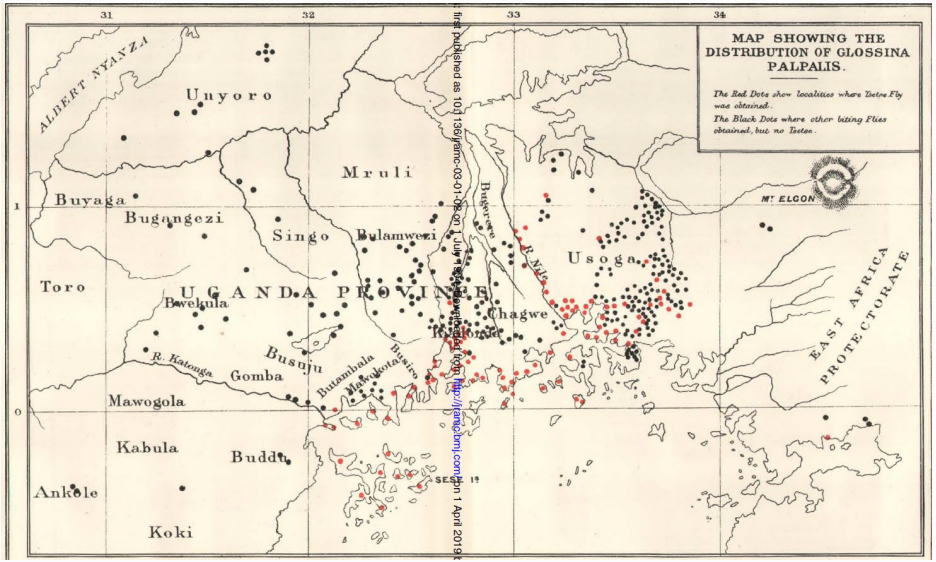
Age	Sex	District	Shamba	Trypanosoma	Filaria
25	M	Sese Island	Sewana	+	+
20	M	Sese Island	Kaganda I	-	+
25	M	Sese Island	Semagala I	+	+
30	M	Sese Island	Kaganda I	-	+
20	M	Sese Island	Semagala I	-	+
25	M	Sese Island	Buvovu I	-	+
25	M	Sese Island	Kaganda I	-	+
20	M	Sese Island	Semagala I	-	+
30	M	Sese 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?

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	+	+



Dr D. Bruce, 1903

Why learn R?

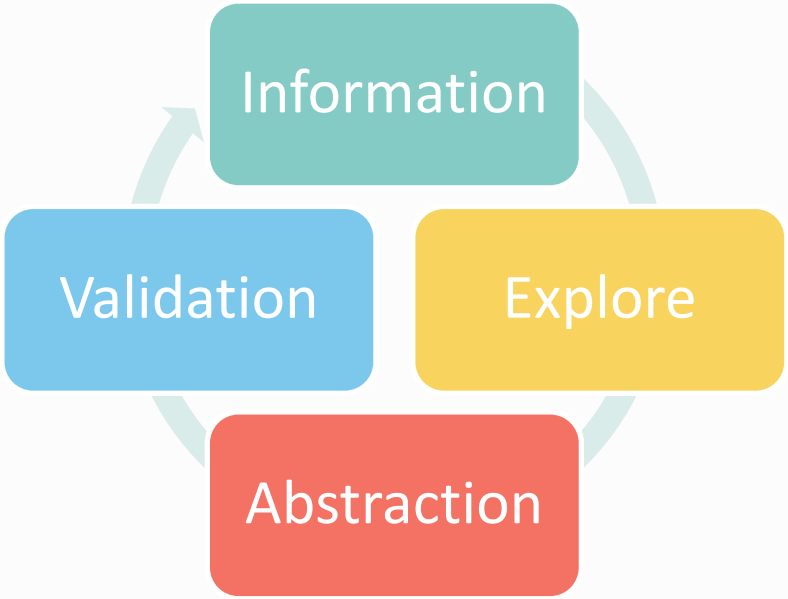
A way to represent ideas

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	+	+

```r
library(ggplot2)
library(magick)
library(readxl)
library(ggpubr)
library(data.table)

map_u <- image_read("Figures/Mapa_uganda.png")

Sick_coord <- read_xlsx(path = "Datasets/D_Bruce_map_dataset_v1.xlsx",
                        sheet = "Sickness") %>% data.table
Flies_coord <- read_xlsx(path = "Datasets/D_Bruce_map_dataset_v1.xlsx",
                         sheet = "Flies") %>% data.table

# Plot 1

image_ggplot(map_u) +
  geom_point(data = Sick_coord, mapping = aes(x = X , y = Y, col = Sickness)) +
  geom_density_2d(data = Sick_coord[Sickness == "Present"],
                  mapping = aes(x = X , y = Y, col = Sickness)) +
  scale_color_brewer(palette = "Set1") +
  labs(x = "", y = "") +
  theme_pubr(legend = "top", base_size = 20) +
  theme(axis.title.x = element_blank(),
        axis.text.x = element_blank(),
        axis.ticks.x = element_blank(),
        axis.title.y = element_blank(),
        axis.text.y = element_blank(),
        axis.ticks.y = element_blank())
# Plot 2

image_ggplot(map_u) +
  geom_point(data = Flies_coord, mapping = aes(x = X , y = Y, col = Flies)) +
  geom_density_2d(data = Flies_coord[Flies == "Tsetse"],
                  mapping = aes(x = X , y = Y, col = Flies)) +
  scale_color_brewer(palette = "Set1") +

  labs(x = "", y = "") +
  theme_pubr(legend = "top", base_size = 20) +
  theme(axis.title.x = element_blank(),
        axis.text.x = element_blank(),
        axis.ticks.x = element_blank(),
        axis.title.y = element_blank(),
        axis.text.y = element_blank(),
        axis.ticks.y = element_blank())
```

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

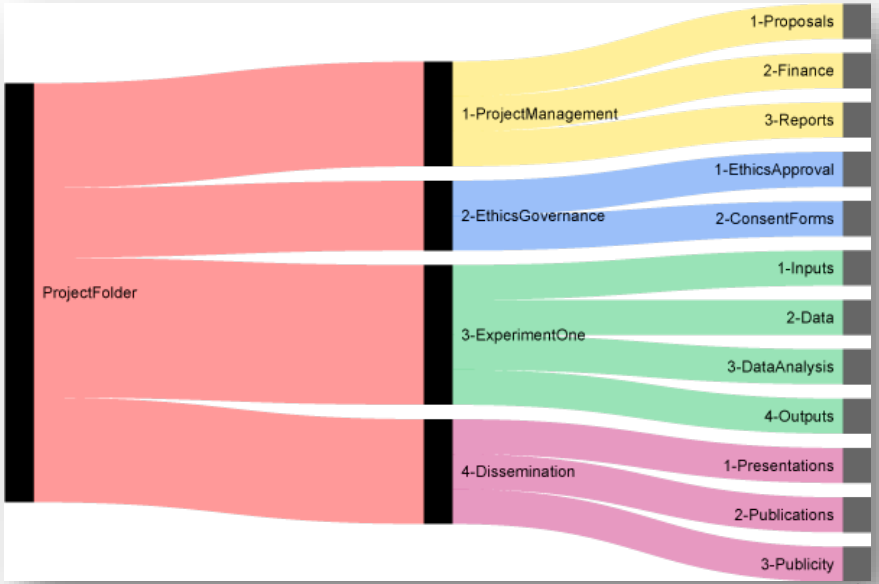
What's R?

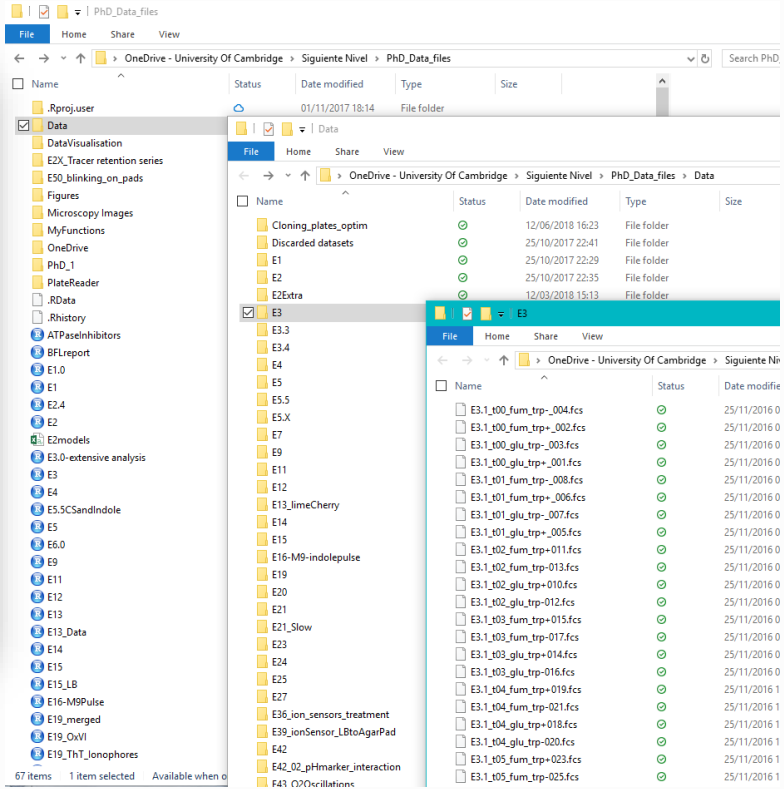
How was my first experience

```
1
2
3
4
5
6
7
8
9
10
11
12
13
14
15
16
17
18
19
20
21
22
23
24
25
26
27
28
29
30
31
32
```

Arrange files and folders

Thinking in hierarchical order

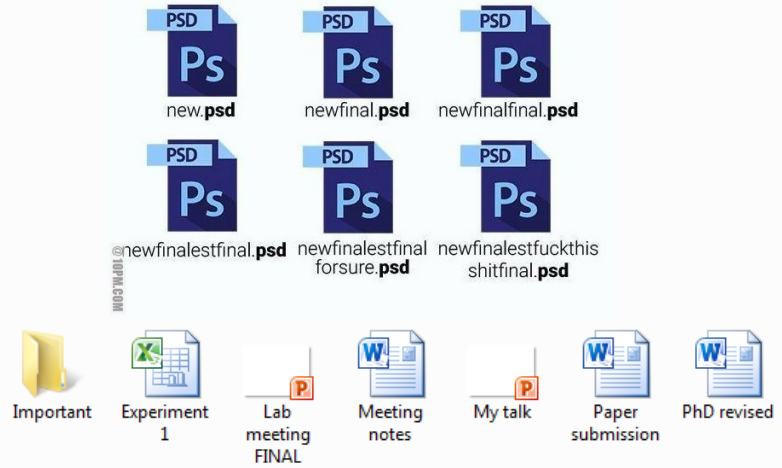




® http://nikola.me/folder_structure.html

Arrange files and folders

Naming… really matters?



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

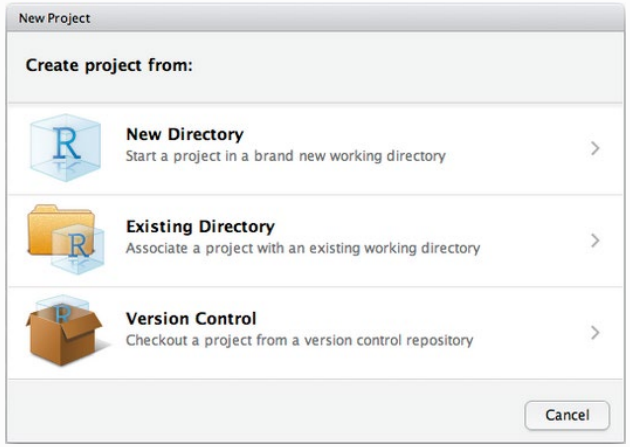
What I do

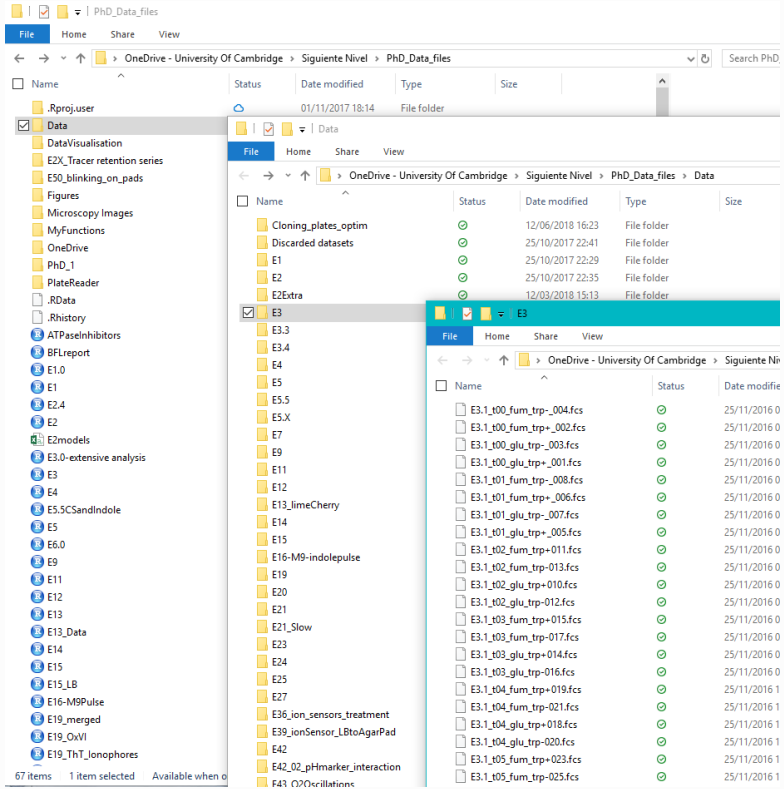
EXP_AAAAMMDD_NOMBRE_VX.X.csv

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

RStudio

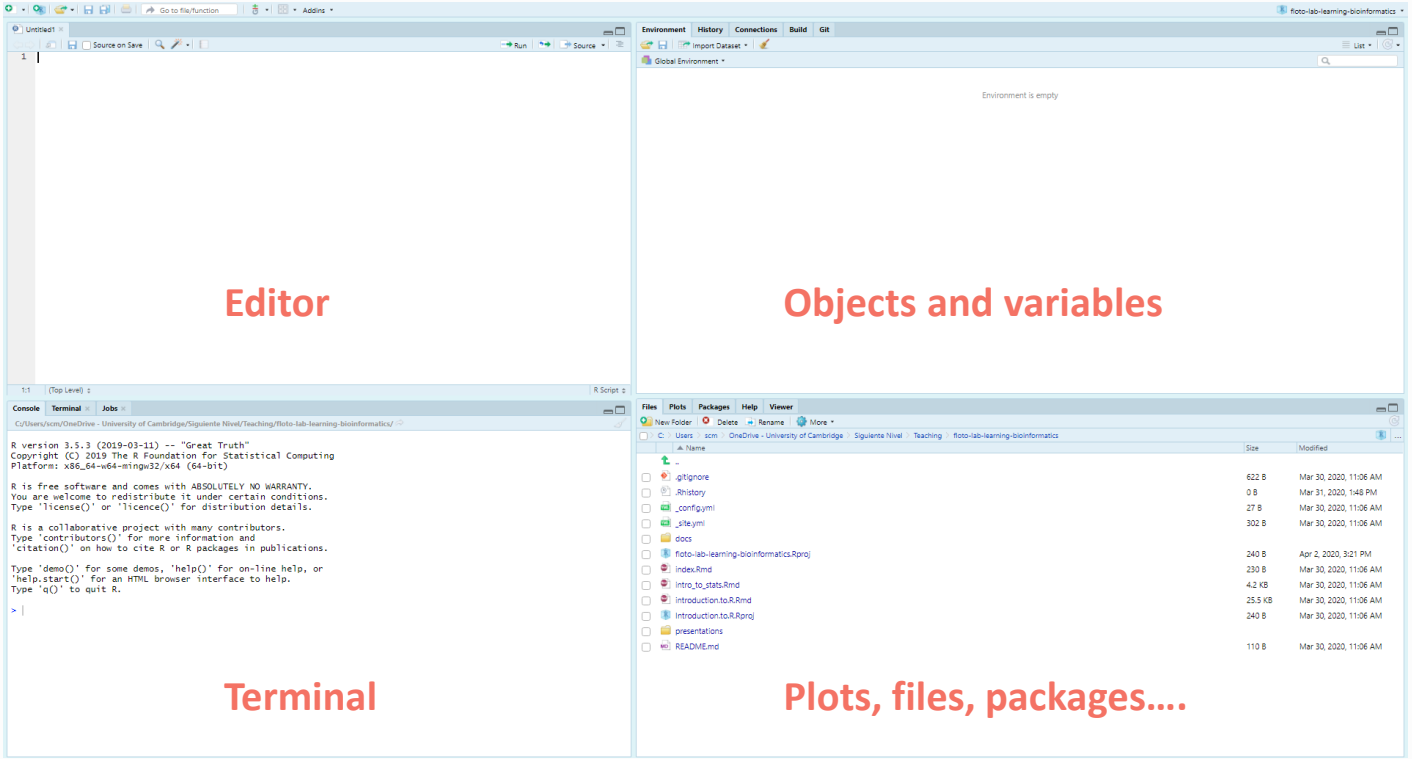
A helper in our quest





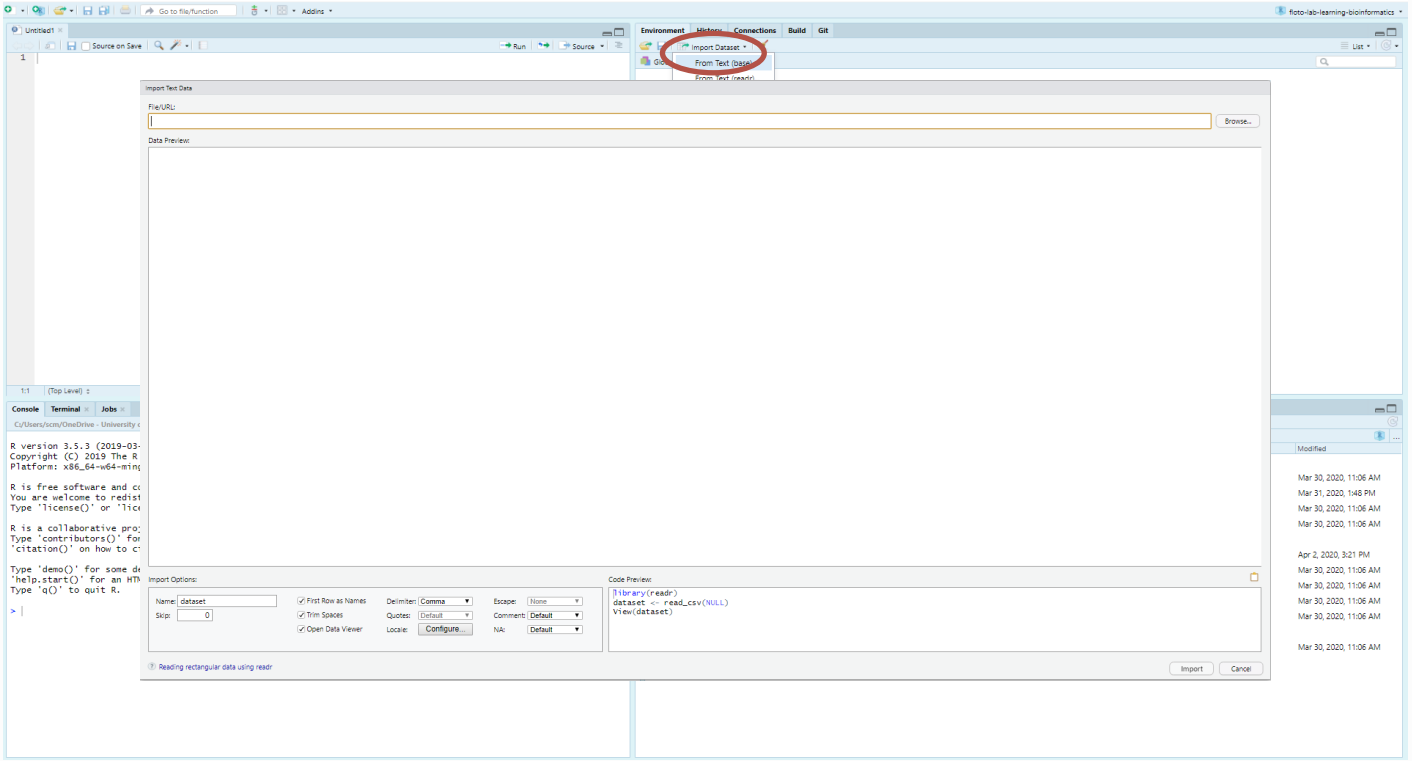
Inside R

RStudio



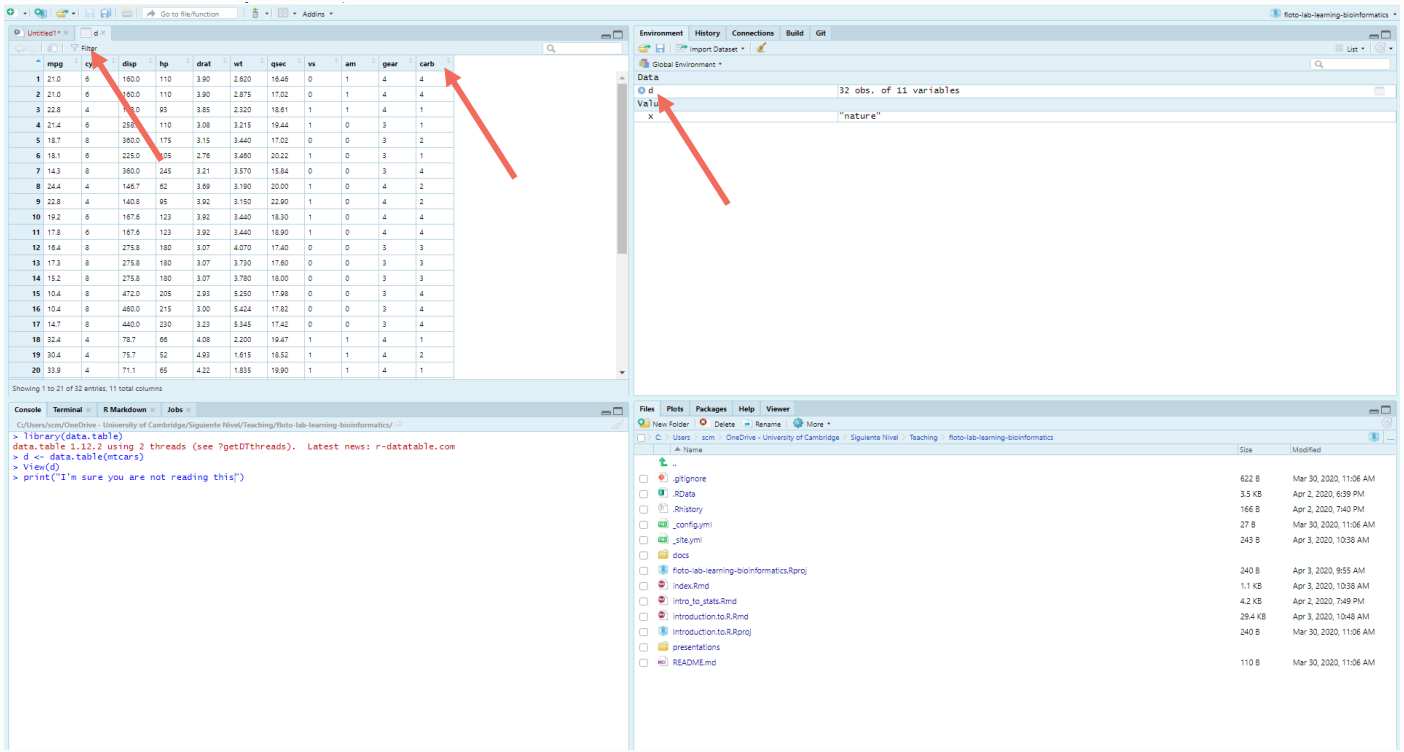
Inside R

Importing data



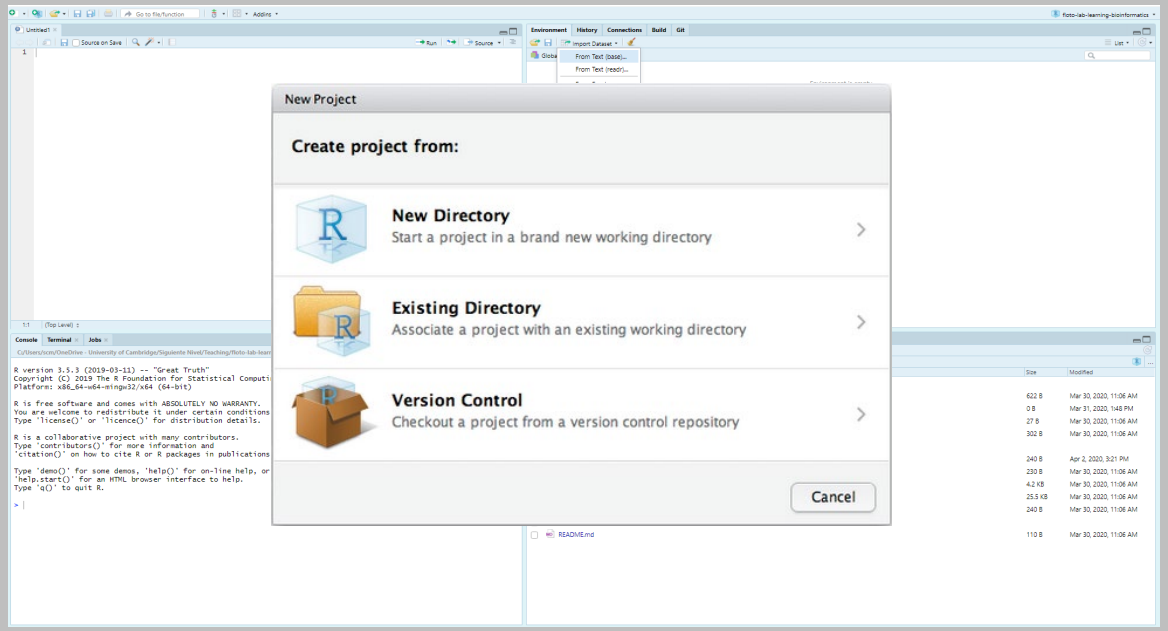
Inside R

Exploring datasets



Time to program

For example



The foundation of R

The heir of mathematical notation

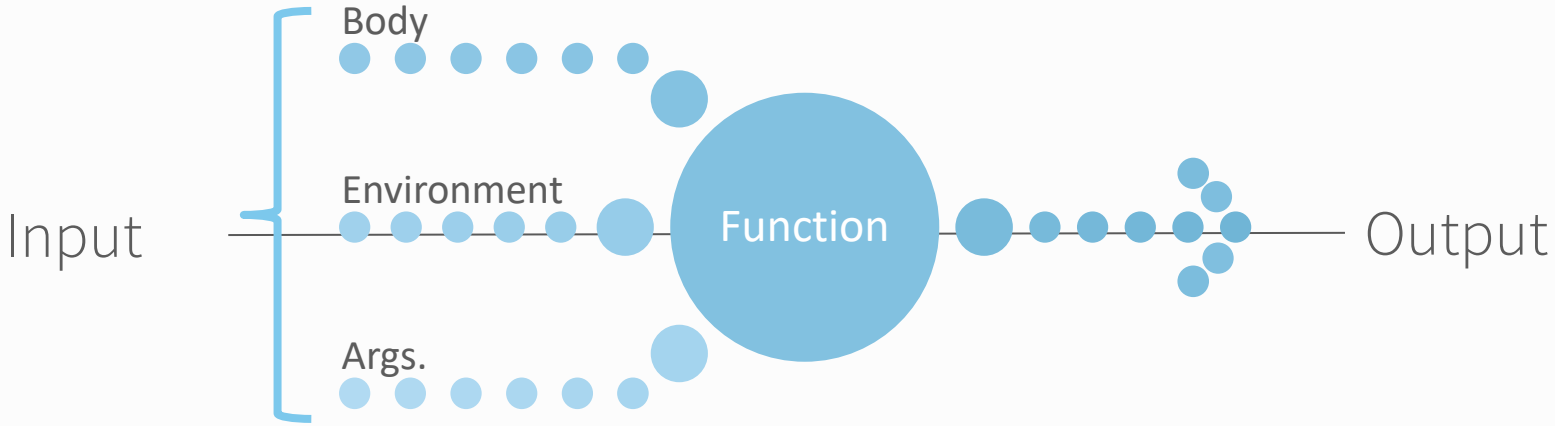
Input $f(x)$ Output

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

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

The foundation of R

Functional programming



R like a super-calc on steroids

But… what's a function?

specify "*x*" and "*y*"

```
Action <- function(x, y = 0) {

          z <- x + y

  return(z)
}
```

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 <- ...                        # Assigning a variable
c(object, object,...)           # Concatenate SAME type objects within a vector
length(object)                  # Number of elements in an object
str(object)                     # Show the internal structure of an object
class(object)                   # Show the class of an object
names(object)                   # Names of the elements
rm(object)                      # Remove the object
mean(vector)                    # Media
median(vector)                  # Median
sd(vector)                      # Standard deviation
sqrt(vector)                    # Square root
log(vector)                     # Natural logarithm
exp(vector)                     # E to the power of..

summary()                       # Summary
getwd()                         # Where we are working
read.delim()                    # Imporf files (also read.csv, read.csv2, read.txt, read.table)
```

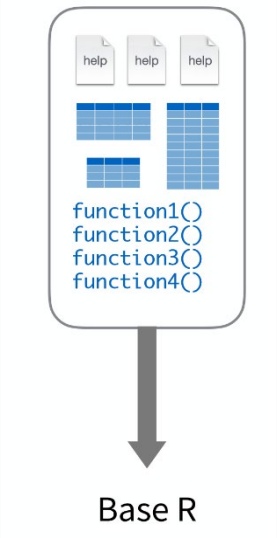
Time to program

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



[image taken from Ryan Wesslen presentation]

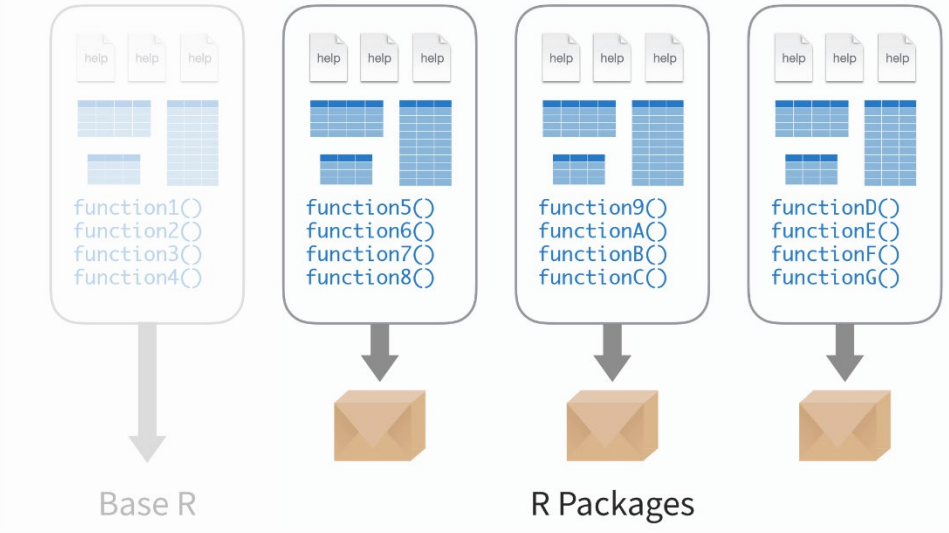
What they are

- 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.
-

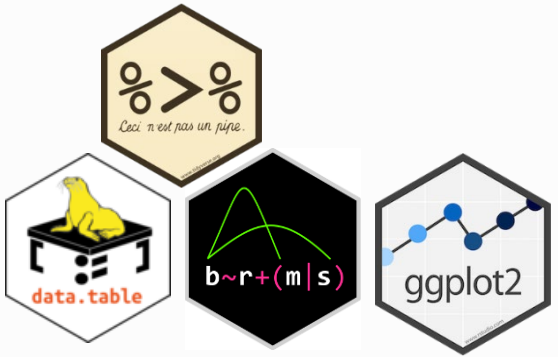
```
install.package(data.table)                          # Install library from CRAN
devtools::install_github("/paul-buerkner/brms")  # Install from GitHub


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, glmmTMB, 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: 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

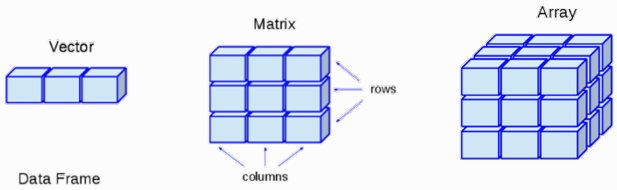
Recycling

```
# 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:

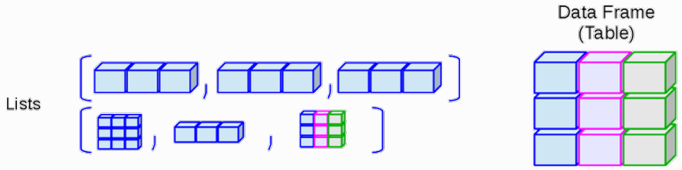
- Rows
- Columns

```
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[fila, columna]
[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.fame

- 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.fame

- 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

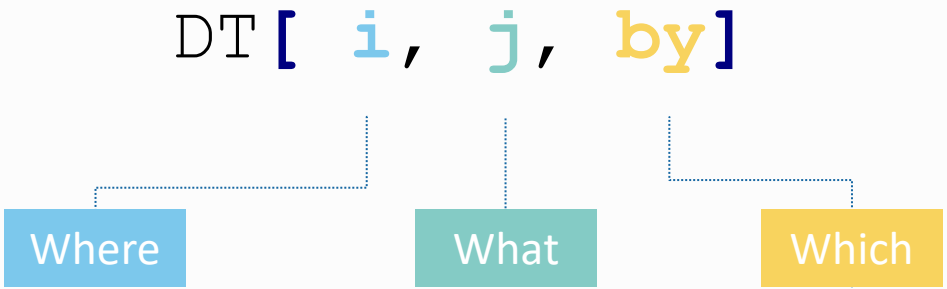
```
head(table)                 # Exploring head/tail values
tail(table)
str(table)

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")]  # Elementros 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)),
            by = .(gear, vs)]        # Multiple groups

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)),
            by = .(gear, vs)]                  # Multiple groups

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` o `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

- 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` o `tidyverse.`
- The basic idea:
  - `x %>% f` is `f(x)`
  - `x %>% f %>% g %>% h` is `h(g(f(x)))`

Example: Representing averages by group

Stack of lazy variables

```
# 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()
```

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` o `tidyverse.`
- The basic idea:
  - `x %>% f` is `f(x)`
  - `x %>% f %>% g %>% h` is `h(g(f(x)))`

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.
- The operator ¨%>% helps structure the code and minimizes the creation of "transitional variables.
- Requires package `magrittr` o `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
```

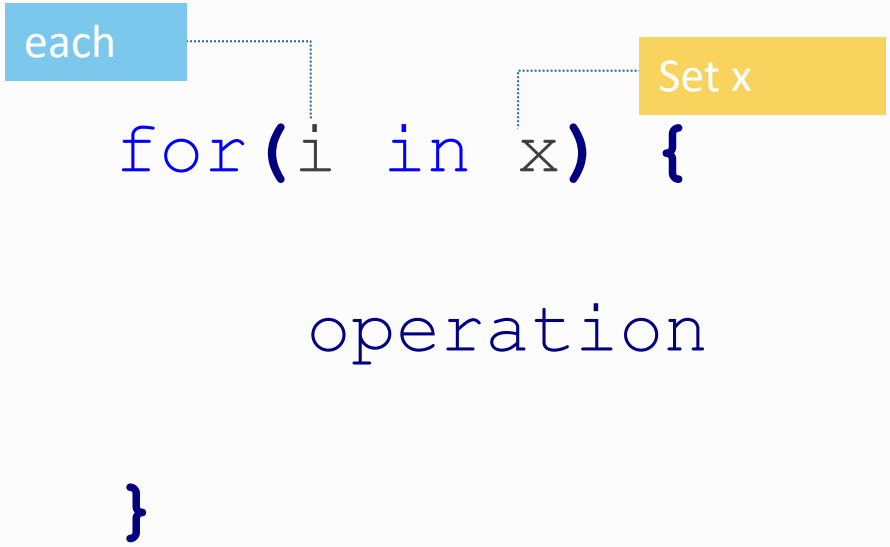
Loops and functionals

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$b)
mean(d$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 -99 -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_acepted == 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 throught if-test throught 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")
            )
```

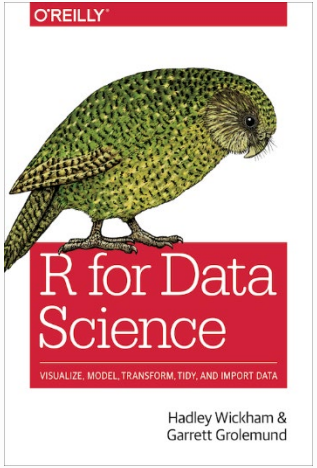
Loops and functionals

An example

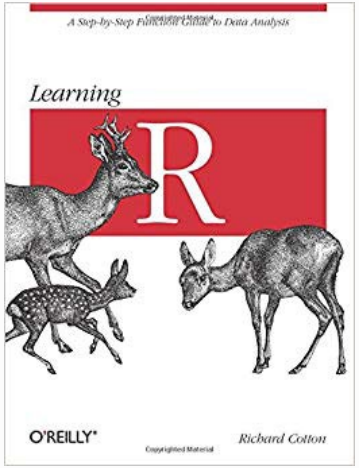
```
Looping with *apply
d <- apply(d, 2, function(x) {ifelse(x == -99, NA, x)})
medias <- apply(X = d, MARGIN = 2, FUN = mean)

# 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

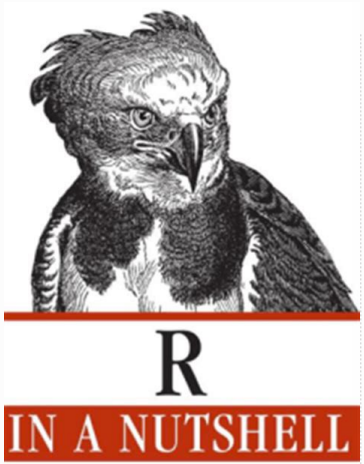
Online resources



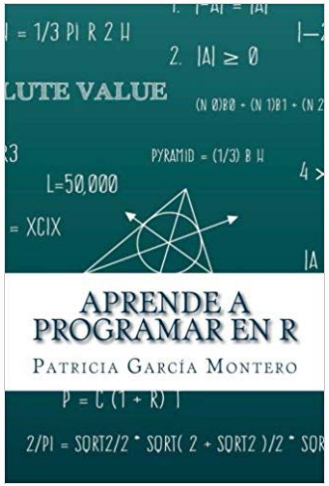
R for Data Science,
H. Wickham &
G. Grolemund



Learning R,
R. Cotton



R in a nutshell,
J. Adler



Aprender a
programar en R
P. García Montero

¡Gracias por vuestro tiempo!

¿Preguntas?