# Coding for Reproducible Research Field Coordinator Training - R Track

Luiza Andrade, Leonardo Viotti & Rob Marty

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#### Introduction

## Objective

Create an R Master Script.

#### Content

- Comments
- Using R Studio to create a code index
- Indenting
- File paths
- Section switches
- Using functions output
- Packages
- O Loops

#### Introduction

- The exercises on this section will help us create a Master R script
- To do them, go to the *DataWork/Code* folder an open the file called *Lab 2 Coding for Reproducible Research.R*

#### Introduction

- Just like in Stata, a Master script must be a map of all data work
- It should be possible to follow all data work in the data folder, from raw data to analysis output, by following the master script
- It should also be possible to run all the code for the project by simply running the Master script
- Finally, in Stata, the Master Script is where you create globals. You can do the same in R by just creating new objects

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# Commenting

- Let's take a look at the script we just opened
- You can see that the first few lines in the script are the header, but they're not commented out
- In R, errors will not always break your code, so you should still be able to run this script
- However, not commenting out comments is still bad practice, as it makes the code harder to read

# Commenting

- To comment a line, write # as its first character
- You can also add # half way through a line to comment whatever comes after it
- In Stata, you can use /\* and \*/ to comment part of a line's code.
   That is not possible in R: whatever comes after # will be a comment
- To comment a selection of lines, press Ctrl + Shift + C

# Commenting

#### Exercise 1

Use the keyboard shortcut to comment the header of the script.

#### Exercise 2

Use the keyboard shortcut to comment the header of the script again. What happened?

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# Creating a document outline in RStudio

- RStudio also allows you to create an interactive index for your scripts
- To add a section to your code, create a commented line with the title
  of your section and add at least 4 trailing dashes, pound signs or equal
  signs after it

#### Exercise 3

Open the script index and make PART 0 a section header. Do the same for PART 1.

#### Exercise 4

Note that once you create a section header, an arrow appears right next to it. Click on the arrows of parts 0 and 1 to see what happens.

# Creating a document outline in RStudio

- The outline can be accessed by clicking on the button on the top right corner of the script window. You can use it to jump from one section to another
- ullet You can also use the keyboard shortcuts Alt + L (Cmd + Option + L on Mac) and Alt + Shift + L to collapse and expand sections

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# **Initial Settings**

• A Stata do-file typically starts with a few settings:

```
clear
set maxvar 120000
set more off
```

## **Initial Settings**

- We don't need to set the memory or the maximum number of variables in R, and more is automatically selected
- However, if you saved the last RStudio session in .Rhistory, the
  objects that were in RStudio's memory last time you closed it will still
  be there whenever you open it again
- Therefore, it's good practice to always clean the memory when starting a new session
- You can see all the objects currently in you memory in the Environment pane

## **Initial Settings**

#### Exercise 5

- Go to Environment window
- Create an object called foo with any content you pick
- Type ls() to print the names of the object in memory
- Type rm(foo) to remove it
- To remove all objects, use rm(list=ls())

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- Since there is a lot of people developing for R, it can have many different functionalities
- To make it simpler, these functionalities are bundled into packages
- A package is the fundamental unit of shareable code
- It may contain new functions, but also more complex functionalities, such as a Graphic User Interface (GUI) or settings for parallel processing (similar to Stata MP)
- They can be shared through R's official repository CRAN (10,000+ packages reviewed and tested) and many other online sources
- There are many other online sources such as Github, but it's important to be careful, as these probably haven't gone through a review process as rigorous as those in CRAN

 To install and use packages you can either do it with the user interface or by the command prompt.

 You only have to install a package once, but you have to load it every new session. To load a package type:

```
library(stargazer)
```

Once a package is loaded, you can use its features and functions. Here's a list of some useful and cool packages:

- Rcmdr Easy to use GUI
- swirl An interactive learning environment for R and statistics.
- ggplot2 beautiful and versatile graphics (the syntax is a pain, though)
- stargazer awesome latex regression and summary statistics tables
- foreign reads dtas and other formats from inferior statistical software
- zoo time series and panel data manipulation useful functions
- data.table some functions to deal with huge data sets
- sp and rgeos spatial analysis
- multiwayvcov and sandwich clustered and robust standard errors
- RODBC, RMySQL, RPostgresSQL, RSQLite For relational databases and using SQL in R.

#### Exercise 6

Install the swirl and stargazer packages, including packages necessary for them to run.

#### Exercise 7

Call the packages you just installed. Note that the library function only accepts one argument, so you will need to load each of them separately.

```
library(stargazer)
##
## Please cite as:
   Hlavac, Marek (2018). stargazer: Well-Formatted Regression and Summary Statisti
##
   R package version 5.2.2. https://CRAN.R-project.org/package=stargazer
 library(swirl)
```

```
##
## | Hi! Type swirl() when you are ready to begin.
```

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## Indenting

- Indenting can be pretty different from what it looks like in Stata
- To indent a whole line, you can select that line and press Tab
- To unindent a whole line, you can select that line and press Shift +
   Tab
- However, this will not always work for different parts of a code in the same line
- In R, we typically don't introduce white space manually
- It's rather introduced by RStudio for us

# Indenting

#### Exercise 8

To see an example of how indenting works in RStudio, add a line between the two arguments of the install.packages function (the vector of package names and the dependents option). Then add a line between the two package names. Note that RStudio formats the different arguments of the function differently.

# Indenting

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## **Functions inception**

- In R, you can use the output of one function as the input of another, as long as they have the same format
- In fact, that's exactly what we just did when installing the packages
- To see that, select just the first argument of the install.packages function and press Ctrl + Enter
- The c() function, as we know, creates a vector with its arguments

```
c("stargazer", "swirl")
```

```
## [1] "stargazer" "swirl"
```

## **Functions inception**

- The resulting is vector is used as an input to the install.packages function
- We could also have stored this vector in the memory as an object and used that object as the input
- In fact, that's exactly what we are going to do next, so the code doesn't get too polluted as we add new packages

## **Functions** inception

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#### Section switches

- Now, installing packages can be time-consuming, especially as the number of packages grow, and each package only needs to be installed once
- What can we bring from our Stata Master do-files to avoid installing packages twice?

#### Section switches

- In Stata, section switches would be saved as locals
- In R, the equivalent to that would be to create a new object

#### Exercise 10

Create a dummy scalar object called PACKAGES.

TIP: Section switches can also be Boolean objects.

#### Section switches

- Now we need to create an if statement using this switch
- If statements in R look like this:

#### Section switches

Possible variations would include

```
# Turn switch on
PACKAGES <- TRUE
# Using a Boolean object
if (PACKAGES == TRUE) {
  install.packages(packages, dep = T)
}
# Which is the same as
if (PACKAGES) {
  install.packages(packages, dep = T)
```

#### Outline

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- The next important part of a Master script are file paths
- We recommend always using **explicit** and **dynamic** file paths

• Implicit and static file path:

• Explicit and static file path:

• Explicit and dynamic file path:

- File paths in R, as in Stata, are basically just strings
- Note, however, that in R we can only use forward slashes (/) to separate folder names

#### Exercise 11

Let's start by adding the folder path to the training's folder in your computer to the beginning of PART 3.

- You can set file paths in your master using the file.path() function
- This function concatenates strings using / as a separator to create file paths

```
# Project folder
projectFolder <-
    "C:/Users/luiza/Documents/GitHub/R-Training"

# Data work folder
dataWorkFolder <- file.path(projectFolder, "DataWork")

# Print data work folder
dataWorkFolder</pre>
```

## [1] "C:/Users/luiza/Documents/GitHub/R-Training/DataWork"

Let's check if that worked, as we will need your Master script to be running smoothly for the other sessions.

#### Exercise 12

- 1 Turn off the switch that installs packages in your Master script
- Q Run the whole script
- Type the following code to open a data set using the file paths you just set:

#### Outline

- Looping

- We're almost at the end of this section
- But the DRY rule can still be applied to part of this code

```
# The DRY rule:
DONT REPEAT YOURSELF
```

- In Stata, we'd usually use a foreach loop to go through a list of objects
- The equivalent to that in R would be to write a for loop like this

```
# A for loop in R
for (number in c(1.2,2.5)) {
    print(round(number))
}
```

```
## [1] 1
## [1] 2
```

- R, however, has a whole function family that allows users to loop through an object in a more efficient way
- They're called apply and there are many of them, for different use cases.
- For the purpose of this training, we will only use two of them, sapply and apply
- If you look for the apply help file, you can see all of them

- sapply(X, FUN, ...): applies a function to all elements of a vector or list and returns the result in a vector. Its arguments are
  - X: a matrix (or data frame) the function will be applied to
  - FUN: the function you want to apply
  - ...: possible function options

```
# A much more elegant for loop in R sapply(c(1.2,2.5), round)
```

```
## [1] 1 2
```

#### Exercise 13

Use the sapply() function to apply the library() function to all packages you have selected.

A more generic version is the apply function.

- apply(X, MARGIN, FUN, ...): applies a function to all columns or rows of matrix. Its arguments are
  - X: a matrix (or data frame) the function will be applied to
  - MARGIN: 1 to apply the function to all rows or 2 to apply the function to all columns
  - FUN: the function you want to apply
  - ...: possible function options

```
## [,1] [,2] [,3]
## [1,] 1 6 2
## [2,] 24 9 74
## [3,] 9 4 2
```

```
# Row means
apply(matrix, 1, mean)

## [1] 3.00000 35.66667 5.00000

# Column means
apply(matrix, 2, mean)
```

## [1] 11.333333 6.333333 26.000000

#### That's all, folks

- Now you have a template master script to use across this training's sessions
- Save the script that you created during this session in the DataWork folder. Call it MASTER.R
- You can run scripts from the Master script by using the source()
  function as you write scripts for future sessions, but we will do that on
  the next session

#### That's all, folks

#### Homework

In the next slide here's a list of all the packages we'll need to the next sessions. Installing them might take a while, so paste them to your Master script and run it again before the next session:

# Thank you!

#### Bonus exercise

- You can customize your loops in R by defining your own function
- This is done using a function conveniently called function()
- For example, if instead of just printing a number we want to print it's square, we could create a function that does both:

```
# A much more elegant for loop in R sapply(c(1,2), function(x) x^2)
```

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