# standardized markdown

Jordy van Langen December 10, 2019

The following contains a standardized workflow following best practices in the open science community (2019).

## 1. First steps

- Give a meaningful title to your markdown document.
- Change the default date for automaticity to: "r Sys.Date()" or "r format(Sys.time(), "%B %d, %Y")"
- Add 'urlcolor' if you want to highlight hyperlinks in a color of your choice in the rendered document. An example of a standardized markdown header:

```
title: "standardized_markdown"
author: "Jordy van Langen"
date: '`r format(Sys.time(), "%B %d, %Y")`'
output:
  pdf_document: default
  html_document: default
urlcolor: blue
```

## 2. R-version check

• Create an R chunk that prints the current R version you are using.

## R.version\$version.string

```
## [1] "R version 3.6.1 (2019-07-05)"
```

## 3. Loading libraries

• Create an R chunk that loads all the libraries you are using.

```
library("ggplot2") #Version 3.2.1
library("here") #Version 0.1

packageVersion()
sessionInfo()
```

- Always make sure to install the latest version of a package when performing your analysis. You can manually check the version of each package with packageVersion().
- sessionInfo() also gives you a lot of information about packages installed and your R version used, however, this might not be desired since you will receive a lot of information that is not always of interest.

• An alternative might be the packrat package, developed by Rstudio. This is a solid way to manage the R packages your project depends on in an isolated, portable, and reproducible way.

## 4. Importing data

- When loading data from your computer (i.e., not directly from an url), use the here package Müller, 2017. This offers a robust solution to make sure that file access does not break accross different computing platforms. A detailed description of the use of this package is described in: A Reproducible Data Analysis Workflow with R Markdown, Git, Make, and Docker, 2019.
- Use the following folder structure:

```
projectname (your project folder)

data (folder containing your data)
R (folder containing your code)
```

• An example on how to import data files, taken and adjusted from A Reproducible Data Analysis Workflow with R Markdown, Git, Make, and Docker, 2019.

Not robust

```
NHANES <-read.csv("/Users/jordyvanlangen/standardized_markdown/data/NHANES.csv")
```

Not robust

```
NHANES <-read.csv("data/NHANES.csv")
```

Robust

```
NHANES <-read.csv(here("data", "NHANES.csv"))</pre>
```

Robust

```
NHANES_df <- read.csv(here("data", "NHANES.csv"))</pre>
```

• Personal remark to this example: I would add \_df to NHANES to indicate that you are working with a dataframe. This is also advised by the Statistical thinking for the 21st century course by Russel Poldrack (2019).

#### 5. Data pre-processing and analysis

• At this stage you can start writing your analysis code.

## 6. Knitting

- You can knit your markdown to an HTML file, however that does not look really 'fancy' and not 'publication-ready'. Therefore, knitting to PDF might be a better option.
- When you want to knit your .Rmd to pdf, you'll need to have LaTeX installed. For R Markdown users
  who have not installed LaTeX before, it is recommended that you install the TinyTex package by YiHui
  Xie.

```
install.packages("tinytex")
```

## 7. Appendix:

The appendix includes an amount of useful sources that might improve your R (markdown) workflow.

- Check the R Markdown cheatsheet for the extensive possibilities that R markdown offers.
- There is the possibility of writing your whole paper (reported stats, tables, figures) in markdown complied APA style with the papaja package by Frederik Aust, 2019. If you do not have to (or want to) write in APA, you can check out the code below for an example on how to make a basic table.
- The redoc package by Noam Ross, 2019: enables a two-way R Markdown-Microsoft Word workflow. It generates Word documents that can be de-rendered back into R Markdown, retaining edits on the Word document, including tracked changes.
- 4 key-points that researchers should take into account when sharing data. See Analysis of Open Data and Computational Reproducibility in Registered Reports in Psychology, 2019.
- Power analysis: although G\*Power is still being used predominantly, it does not suit reproducible research. Therefore, a good alternative is to perform your power-analysis in R. For simpler power-analysis (e.g. t-tests) you can find a good tutorial from Dan Quintana (2019) on YouTube in which he explains the use of the pwr package Helios de Rosario, 2019. For more extensive, simulation-based, power-analysis have a look at the ANOVApower package by Daniël Lakens, 2019.
- If you want to unload a package and use a clean version of it again, you can use the following code:

```
if("package_name" %in% (.packages())){
  detach('package:package_name', unload=TRUE)}
library("package_name")
```

This workflow was made by Jordy van Langen, December 10, 2019.

If you have suggestions on how to improve, feel welcome to send a message to jordy.vanlangen@sydney.edu.au or contact me on Twitter jordyvanlangen