Reproducible Manuscripts in R Markdown

Juli Tkotz

27-05-2020

ZI Mannheim, RG Psychology and Neurobiology of Sleep and Memory

Nice to meet(up) you!



- Juli Tkotz
- PhD student ZI Mannheim
- RG Psychology and Neurobiology of Sleep and Memory

What is a reproducible manuscript?

- A manuscript that directly embeds your research data and analysis code.
- Any person with the raw data can run the code and reproduce your manuscript.
- Interactive stand-alone versions are possible.

Why do we need it?

Analysis of Open Data and Computational Reproducibility in Registered Reports in Psychology

Pepijn Obels¹, Daniel Lakens¹, Nicholas A. Coles², Jaroslav Gottfried³, & Seth Ariel Green⁴

Eindhoven University of Technology, The Netherlands
 University of Tennessee, Knoxville, USA
 Masaryk University, Brno, Czech Republic
 Code Ocean, New York, USA

Ongoing technological developments have made it easier than ever before for scientists to share their data, materials, and analysis code. Sharing data and analysis code makes it easier for other researchers to re-use or check published research. These benefits will only emerge if researchers can reproduce the analysis reported in published articles, and if data is annotated well enough so that it is clear what all variables mean. Because most researchers have not been trained in computational reproducibility, it is important to evaluate current practices to identify practices that can be improved. We examined data and code sharing, as well as computational reproducibility of the main results, without contacting the original authors, for Registered Reports published in the psychological literature between 2014 and 2018. Of the 62 articles that met our inclusion criteria, data was available for 40 articles, and analysis scripts for 37 articles. For the 35 articles that shared both data and code and performed analyses in SPSS, R, Python, MATLAB, or JASP, we could run the scripts for 31 articles, and reproduce the main results for 20 articles. Although the articles that shared both data and code (35 out of 62, or 56%) and articles that could be computationally reproduced (20 out of 35, or 57%) was relatively high compared to other studies, there is clear room for improvement. We provide practical recommendations based on our observations, and link to examples of good research practices in the papers we reproduced.

Obels et al. (2019)

Beyond reproducible analyses

263 4 3.1 Epistemic Trustworthiness 264 Participants placed more epistemic trust in the debaters when reading a neutral debate: Student 265 teachers in the neutral condition (M = 5.06, SD = 1.00) perceived the debaters to have more expertise 266 than those in the uncivil condition (M = 5.06, SD = 1.00), t(218.49) = 1.99, p = .047, d = 0.27. Furthermore, participants reading a neutral debate (M = 4.76, SD = 1.02) reported higher ratings of 267 268 debaters' integrity than those reading an uncivil debate (M = 4.05, SD = 1.15), t(219.41) = 4.87, p < 0.00269 .001, d = 0.65. Additionally, ratings of benevolence were higher in the neutral condition (M = 4.77). 270 SD = 0.98) than in the uncivil condition (M = 4.05, SD = 0.89), t(214.11) = 5.67, p < .001, d = 0.76271 (see Figure 2). 272 We further explored the correlation between the conflict explanation items and the METI subscales. 273 that is, if the perception of various aspects of a conflict was associated with different degrees of 274 epistemic trust. Those who agreed that the debaters in the scenario referred to different research 275 results also thought them to have more expertise, r(220) = .14, p = .039. There was no relation with 276 integrity, r(220) = .07, p = .321, or benevolence, r(220) = .03, p = .679. Assuming personal reasons 277 for the conflict had the strongest relationship with epistemic trust. The more participants perceived 2.78 the conflict to be personal, the less expertise they assigned to the debaters r(220) = -.25, p < .001. In 279 a similar manner, perception of a personal conflict lead to decreased ratings of integrity, r(220) = 280 -36, p < .001, and benevolence, r(220) = -.41, p < .001. How much participants agreed that the 281 debaters referred to different goals of PAVLOV did not correlate with any of the METI subscales. 282 neither with expertise, r(220) = .10, p = .122, nor with integrity, r(220) = -.00, p = .946, nor with 283 benevolence r(220) = -.00, p = .994. Embracement of the statement that debaters referred to different 284 effects of PAVLOV was not associated with epistemic trust either, neither with expertise, r(220) = 285 0.01, p = .863, nor with integrity, r(220) = -.06, p = .348, nor with benevolence r(220) = -.05, p = .06286 .475. Internal consistency of the METI subscales was somewhat lower than initially found by 287 Hendriks et al. (2015), with a Cronbach's α of .87 for expertise, .83 for integrity and .76 for 288 benevolence.

R Markdown to the rescue

```
## R Markdown to the rescue

'''{r intext_stats, echo = TRUE}
nerd <- read.csv("./data/nerd.csv", sep = "\t")

'''{r copy_paste_hell}
include_graphics("./pics/slide_inception.png")

This example dataset consists of $N = $ `r nrow(nerd)` participants with an age range between `r min(nerd[["age"]])` and `r max(nerd[["age"]])` years.
overall, 'r sum(nerdsage > 100)` participants reported to be older than 100, so we probably can't trust this data set a lot.
```

This example dataset consists of N=14955 participants with an age range between 13 and 38822 years. Overall, 8 participants reported to be older than 100, so we probably can't trust this data set a lot.

Data retrieved from https://openpsychometrics.org/

R Markdown to the rescue

```
## R Markdown to the rescue

"'[r intext_stats, echo = TRUE]
nerd <- read.Csv("./data/nerd.csv", sep = "\t")

"'[r copy_paste_hell]
include_graphics("./pics/slide_inception.png")

This example dataset consists of $N = $ `r nrow(nerd) `participants with an age range between `r min(nerd[["age"]]) ` and `r max(nerd[["age"]]) `years.
overall, 'r sum(nerdsage > 100) `participants reported to be older than 100, so we probably can't trust this data set a lot.
```

This example dataset consists of N=14955 participants with an age range between 13 and 38822 years. Overall, 8 participants reported to be older than 100, so we probably can't trust this data set a lot.

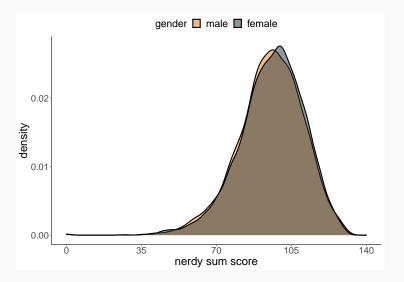
How about some stats?

```
nerd_ttest <- t.test(sum_score ~ gender, data = nerd)
nerd_effsize <- cohen.d(sum_score ~ gender, data = nerd)</pre>
```

```
In this dataset, men `r print_mean_sd(nerd[["sum_score"]][nerd[["gender"]] == "male"])` have a
significantly lower nerd score than women `r print_mean_sd(nerd[["sum_score"]][nerd[["gender"]]
== "female"])`, `r print_ttest(nerd_ttest, nerd_effsize)`.
```

In this dataset, men (M=95.18, SD=15.27) have a significantly lower nerd score than women (M=95.82, SD=15.16), t(9800.06)=-2.39, p=.017, d=-0.04.

Yeah, plots!



Yeah, references!

```
## Yeah, references!

```{r citation}
include_graphics("./pics/citation.png")

```

If I want to cite a paper, I can do this [@san_martin_1968].
This also works if I cite @san_martin_1968 as an in-text citation.
```

If I want to cite a paper, I can do this (San-Martin et al. 1968). This also works if I cite San-Martin et al. (1968) as an in-text citation.

How to get bibtex references



Fully formatted articles

The R-packaga papaja offers you documents that are formatted according to APA (6) style.

https://github.com/crsh/papaja



Or a whole book?

R for Data Science

build passing

This repository contains the source of R for Data Science book. The book is built using bookdown.

The R packages used in this book can be installed via

devtools::install_github("hadley/r4ds")

This is the website for "R for Data Science". This book will teach you how to do data science with R: You'll learn how to get your data into R, get it into the most useful structure, transform it, visualise it and model it. In this book, you will find a practicum of skills for data science. Just as a chemist learns how to clean test tubes and stock a lab, you'll learn how to clean data and draw plots—and many other things besides. These are the skills that allow data science to happen, and here you will find the best practices for doing each of these things with R. You'll learn how to use the grammar of graphics, literate programming, and reproducible research to save time. You'll also learn how to manage cognitive resources to facilitate discoveries when wrandling, visualising, and exploring data.



Talking about reproducibility ...

```
> anticlusters <- anticlust::anticlustering(
+ iris[, -5],
+ K = 3,
+ objective = "variance",
+ method = "exchange"
+ )
Error in loadNamespace(name) : there is no package called 'anticlust'</pre>
```

Code capsules



https://codeocean.com/capsule/3923848/tree/v1

There will be pain



But it's worth it

- Saves time for others and for future you.
- Mistakes are easier to spot and easier to correct.
- Your data and your manuscript will survive longer.
- Others can learn from your analyses.

Thank you!

 $Find \ this \ presentation \ on \ Git Hub.$

Or on CodeOcean.

Or on the OSF.

References

Obels, Pepijn, Daniel Lakens, Nicholas A Coles, Jaroslav Gottfried, and Seth A Green. 2019. "Analysis of Open Data and Computational Reproducibility in Registered Reports in Psychology." PsyArXiv.

https://doi.org/10.31234/osf.io/fk8vh.

San-Martin, M., M. Copaira, J. Zuniga, R. Rodreguez, G. Bustinza, and L. Acosta. 1968. "Aspects of Reproduction in the Alpaca." Reproduction 16 (3): 395-99. https://rep.bioscientifica.com/view/journals/rep/16/3/jrf_16_3_009.xml.