

# Reproducible Manuscripts in R Markdown

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Juli Tkotz

27-05-2020

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

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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.

## 263 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 ( $M = 5.06, SD = 1.00$ ),  $t(218.49) = 1.99, p = .047, d = 0.27$ .  
267 Furthermore, participants reading a neutral debate ( $M = 4.76, SD = 1.02$ ) reported higher ratings of  
268 debaters' integrity than those reading an uncivil debate ( $M = 4.05, SD = 1.15$ ),  $t(219.41) = 4.87, p < .001, d = 0.65$ . Additionally, ratings of benevolence were higher in the neutral condition ( $M = 4.77, SD = 0.98$ ) than in the uncivil condition ( $M = 4.05, SD = 0.89$ ),  $t(214.11) = 5.67, p < .001, d = 0.76$  (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  
278 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  $.01, p = .863$ , nor with integrity,  $r(220) = -.06, p = .348$ , nor with benevolence  $r(220) = -.05, p =$   
286  $.475$ . Internal consistency of the METI subscales was somewhat lower than initially found by  
287 Hendriks et al. (2015), with a Cronbach's  $\alpha$  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  $N = \text{r nrow(nerd)}$  participants with an age range between  $\text{r min(nerd[["age"]])}$  and  $\text{r max(nerd[["age"]])}$  years. Overall,  $\text{r sum(nerd$age > 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/>

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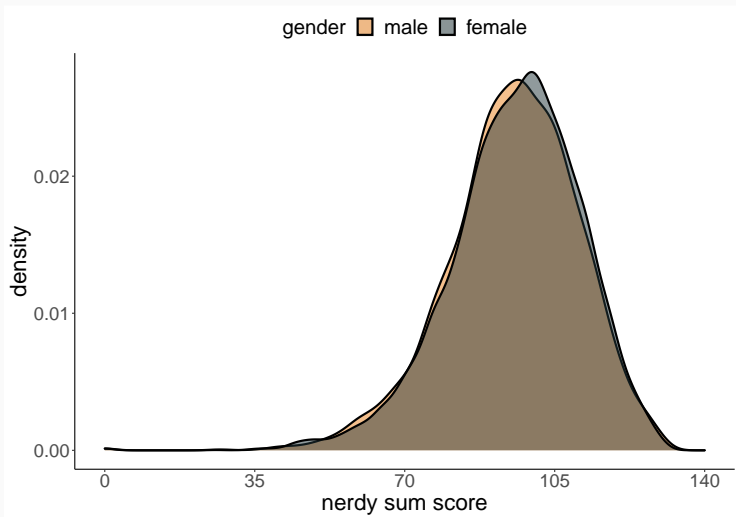
## How about some stats?

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

```
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

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Authors: M. SAN-MARTIN, M. COPAIRA, J. ZUNIGA, R. RODRIGUEZ

DOI: <https://doi.org/10.1530/jrf.0.0160395> Article Type: Research Article

Page(s): 395–399 Online Publication Date: A

Volume/Issue: Volume 16: Issue 3

Abstract/Excerpt PDF

*J. Reprod. Fert.* (1968) **16**, 395–399

ASPECTS OF REPR

M. SAN-MARTIN, M. COPAIRA, J. ZUNIGA, R. RODRIGUEZ


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# Fully formatted articles

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

<https://github.com/crsh/papaja>

1 How to use papaja: An Example Manuscript Including Basic Instructions

2 Frederik Aust<sup>1</sup>

3 <sup>1</sup> University of Cologne

HOW TO USE PAPAJA

6

Table 1

*Descriptive statistics of correct recall by dosage.*

|  | Mean | Median | SD   | Min   | Max   |
|--|------|--------|------|-------|-------|
|  | 4.19 | 14.00  | 4.45 | 5.00  | 25.00 |
|  | 3.50 | 14.00  | 5.15 | 4.00  | 22.00 |
|  | 9.19 | 19.00  | 3.52 | 13.00 | 25.00 |

table was created with `apa_table()`

References

Allaire, J., Cheng, J., Xie, Y., McPherson, J., Chang, W., Allen, J., . . . Hyndman, R. (2016). *Rmarkdown: Dynamic documents for r*. Retrieved from <https://CRAN.R-project.org/package=rmarkdown>

Aust, F., & Barth, M. (2015). *Papaja: Create apa manuscripts with rmarkdown*.

Bates, D., & Maechler, M. (2016). *Matrix: Sparse and dense matrix classes and methods*. Retrieved from <https://CRAN.R-project.org/package=Matrix>

# Or a whole book?

## R for Data Science

build passing

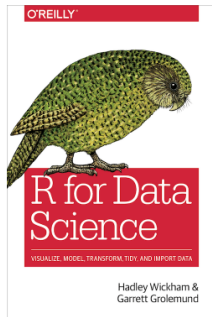
This repository contains the source of [R for Data Science](#) book. The book is built using [bookdown](#).

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 wrangling, visualising, and exploring data.

This website is (and will always be) **free to use**, and is licensed



## Talking about reproducibility ...

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

On Code Ocean.

# There will be pain



## 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](https://rep.bioscientifica.com/view/journals/rep/16/3/jrf_16_3_009.xml).