Automating Reproducibility

Challenges and what it takes to meet them

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Why do we trust science?

Scientific claims should not be credible because of their originators' authority but by the transparency and replicability of their supporting evidence.

Reproduction ≠ Replication

World \longrightarrow Data \longrightarrow Results \longrightarrow Conclusions

Reproduction ≠ Replication



Replication can not be automated, but reproducibility can and should be automated.

Reproduction ≠ Replication



"Insanity is doing the same thing over and over again and expecting different results."

- Albert Einstein (disputed)

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As it turns out, doing the same thing is pretty complicated.



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- 2. Copy-and-paste errors
- 3. Ambiguous order of code execution
- 4. Broken dependencies

Lessons from software engeniering

Four solutions:

- 1. Version control
- 2. Dynamic document creation
- 3. Dependency tracking
- 4. Software management

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Methods in Behavioral Sciences, 1, Article e3763. https://doi.org/10.5964/qcmb.3763

Specify Everything

The relations between code, data, results and their environment need to be unambiguously specified.

Why should I care?

Productivity:

- ► reuse
- easier collaboration
- avoid trouble (during review, questions after publication, etc.)

Why should I care?

Good scientific practice:

- reproducibility is a precondition for replication
- ► increases transparency and (longterm) accessibility

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time

Tools for R Users

In the R universe and beyond, the most flexible tools are:

- Dynamic document creation = RMarkdown*
- ► Version control = Git**
- Dependency tracking = Make**
- Software management = Docker**

 * RMarkdown supports more then 40 languages e.g.: Python, Julia, SAS, Scala & Octave
 ** Language agnostic

RMarkdown—Literate Programming

Text and code are mixed in a single source document that can be dynamically compiled into various representations:

- ► (APA conformable) manuscripts
- presentations
- websites
- books
- ► posters
- ► CV

Silly Heading

```
````{r t-test}
data("sleep")
result <- t.test(extra ~ group, data = sleep)
````</pre>
```

This is an example of students' sleep data taken from `help(t.test)`.

```
`r apa_print.htest(result)$full_result`
```

```
I can now assert that what I *believe* to be true
--- that there is a difference in means between the groups ---
is `r ifelse(result$p.value > .025, "**not**", "")` supported by the data.
```

A simple R markdown example

Aaron Peikert & Andreas M. Brandmaier

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library("knitr")
library("papaja")

Silly Heading

data("sleep")
result <- t.test(extra ~ group, data = sleep)</pre>

This is an example of students' sleep data taken from help(t.test).

 $\Delta M = -1.58, 95\%$ CI [-3.37, 0.21], t(17.78) = -1.86, p = .079

I can now assert that what I *believe* to be true — that there is a difference in means between the groups — is **not** supported by the data.

Git/GitHub—Version Control

Version control is a system that records changes to a set of files over time so that you can recall specific versions later.

It guarantees that code and data are exactly the same version as used for publication.

Make—Dependency Tracking

Make is a "recipe" language that describes how files depend on each other and how to resolve these dependencies.

```
spaghetti_arrabiata.pdf: spagetti_arrabiata.Rmd arrabiata_sauce.csv pa
Rscript -e 'rmarkdown::render("spaghetti_arrabiata.Rmd")'
pasta.csv: cook_pasta.R
Rscript -e 'source("cook_pasta.R")'
arrabiata_sauce.csv: cook_sauce.R canned_tomatoes.csv
Rscript -e 'source("cook_sauce.R")'
```

Docker—Containerization

Docker is a lightweight virtual computer. Dockerfiles are "recipes" that describe what to install on that virtual computer:

FROM rocker/verse:3.6.1
ARG BUILD_DATE=2019-11-11
RUN install2.r --error --skipinstalled\
here lavaan
WORKDIR /home/rstudio



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Unambiguous



Unambiguous Standardized



Unambiguous Standardized Portable



Unambiguous Standardized Portable Automated

Simplifying the tools

These tools require extensive training and need much time to configure correctly.

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These tools require extensive training and need much time to configure correctly. The R package 'repro' abstracts away the concrete technical implementation:

```
repro:
    packages:
        - ggplot2
        - aaronpeikert/repro@adb5fa569
scripts:
        - R/clean.R
data:
        mycars: data/mtcars.csvrepro:
```

The function **repro::automate()** automatically infers Dockerand Makefile.

Disadvantages

- requires complex software infrastructure
- depends on for-profit services
- diverges from the standard manuscript workflow

Your melange may vary

Different requirements regarding: archivation + number of machines

Your melange may vary

Different programming languages:

- Python
- ► R

Julia

Matlab



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Dependency tracking + software management = distributed computation

distributed computation Dependency tracking enables intelligent task scheduling Software management guarantees compatible software environment

distributed computation on Cloud Computing infrastructure

distributed computation on High Performance Computing cluster (HPC)

HPC—Container

- repro supports Singularity as a Docker alternative
- developing environment matches HPC environment exactly
- full freedome to use any software, even when not supported by HPC admin

HPC—Dependency tracking

making dependencies between tasks explicit enables:

- intelligent caching
- automatic parralelization
- dynamic job scheduling

Make is well supported by several job schedulers.

Pure R solutions like the packages targets + futureverse offer even more convinience and are compatible with repro

Focus: Modularity



repro is a modular system

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- repro is a modular system
- potential integration of other workflows

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- repro is a modular system
- potential integration of other workflows
- "Lego system of reproducibility tools"

Focus: Longterm Archive

All software is bundled into the container, therefore all we need is:

- ► container software
- storage infrastructure

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What happens when Docker and co. are not supported anymore?

Containers can be converted into a full system image ensuring support for decades.

References

Slides: https://github.com/aaronpeikert/repro-talk Package: https://github.com/aaronpeikert/repro-thesis Workflow: https://doi.org/10.31234/osf.io/8xzqy

Thank you

Questions?