DOING REPRODUCIBLE SCIENCE AN OPINIONATED INTRODUCTION

OPEN SCIENCE STUDENT

SUPPORT GROUP

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- Brain and Cognitive Science student working in Andrea Protzner's Brain Dynamics Lab
- In the spirit of Thomas Kuhn, I am a scientific revolutionary who wants to make open science normal science
- Developing workflows, practices, and tools to do reproducible science is one way I hope to make normal science more open
- That's my dog Thor, he's passionate about open science too



REPRODUCIBLE SCIENCE

- A scientific pipeline whose steps, processes, procedures, and results can be reproduced by other scientists (or future you)
- A separate concept from replicable science
 - The robustness of a given scientific finding as determined by the degree to which it can be repeatedly obtained
- Reproducible science makes it easier for other scientists and yourself to:
 - Verify the veracity of your findings
 - Replicate your research
 - Your findings are more likely to replicate when they are informed by <u>Open</u> <u>Theory</u>



IMPORTANCE

- Scientists are untrustworthy. Some are:
 - Careerists interested in fame, money, or cultural capital over good science (e.g., Sigmund Freud)
 - Frauds running citation rings, forging data, p-hacking, or self-plagiarizing (e.g., <u>Daryl Bem</u>, <u>Hans Eysenck</u>, <u>Mark Griffiths</u>, etc.)
- Many are:
 - Humans making basic errors or using heuristics to guide their decision-making (e.g., You and Me)
 - More than 50% of papers report impossible statistics



IMPORTANCE CONT'

- Thus, scientific findings should be treated as possible but untrustworthy anecdotes unless they can be verified by other scientists
- By making science reproducible we allow our results to be verified, increasing their trustworthiness
 - Trustworthy != True
 - False positives, undetectable data forgery, etc., are still possible with open data and materials
- Reproducibility also makes science more efficient by reducing redundant labour



HOW DOES IT WORK?

- Science can be made reproducible by:
 - Sharing materials, data, etc.
 - Documenting your scientific pipeline
- Use free open-source software wherever possible in your scientific pipeline. This ensures your work is accessible to:
 - Lower-income scientists
 - Yourself once you lose access to all the licenses the university is paying for you
- Cite all software and packages you use $\ensuremath{\textcircled{}}$



HOW DOES IT WORK? CONT'

- Catalog the scientific pipeline used to obtain your results. Methods of reproducibility include:
 - Written descriptions
 - Photographic and Video guides
 - Software along with instructions on how results were obtained (using text, pictures, videos)
 - Reproducible code
 - Packaging code and data
 - Continuous Integration, Continuous Deployment, Unit Testing
 - Machine-readable Hypothesis Testing



TIPS

- Guiding principle: Keep it simple for yourself and other scientists
- Use OS independent software that works on Mac, Windows, and Linux
- Name project folders and files in their order of operation (e.g., 01_data, 02_cleaning, etc.)
 - All code/syntax should run linearly
- Make your code/syntax human readable, follow a style guide:
 - Google Style Guides (for Python, R, and many other languages)
 - Tidyverse Style Guide (for R/tidyverse)
 - Comment your code/syntax liberally, explain what it's doing and why



TIPS CONT'

- Make your data, code, and instructions machine readable (i.e., processable by computers):
 - Never take screenshots of data or code in place of sharing in a machine-readable format, seriously
 - .csv is the gold-standard for data, .json has uses too
 - .txt or .md are ideal for plain-text
 - Native file formats for any programming language are best for code or reproducible manuscripts
- Do not write scripts that install packages or change settings on someone else's computer, it's rude and disruptive



OTHER TIPS

- Do not hard code information that may change in the future or whose computation should be checked for reproducibility. This includes:
 - Citation styles
 - Statistics and other numbers
 - Tables and Figures
- Write your work as a reproducible manuscript to avoid hard coding information
- Make templates for yourself to make project management easier in the future
- Use <u>GitHub Releases</u> and/or <u>OSF Registrations</u> to take snapshots of your project at different stages



OTHER TIPS CONT'

- Set a seed before running any code/syntax that relies on a randomization function
- Use Internet Archive URLs or save webpage data if you are web mining



BENEFITS

- Your work will be more trustworthy
 - There's proof you actually did what you said you did
- The chances of errors in your work being identified will increase
 - (especially if you have a nemesis who wants to disprove your ideas)
- Other researchers (and future you) can repurpose your scientific pipeline for their own projects
- Collaboration will be easier
- You will learn and apply skills that will help land you a well-paying job



BENEFITS CONT'

- You can automate the least creative tasks of the scientific process, leaving you <u>more</u> <u>time for theorizing</u>
 - Citations can be automagically generated to different formats (APA, MLA, etc.) using <u>CSL files</u>
 - Statistics, tables, and plots can be automagically generated to reflect changes in your data
 - You can create living scientific documents that are automagically published to the web



BARRIERS

- Reproducibility requires data sharing, and not all data can be shared
 - Solution: Share <u>synthetic data</u> that has similar statistical properties to your closed data
- Making your science reproducible may require learning new software or APIs
 - This can be difficult working around a busy schedule, but the payoff is worth it
 - Collaborators might not be willing to switch to or learn these either
 - Solution: Thoroughly documenting your scientific pipeline in a software agnostic way is a good practice regardless, so do this in the meantime



BARRIERS CONT'

- Less robust reproducibility methods may lead to irreproducible results in the future
 - Certain methods in software can break or disappear after updates
 - Solution: Use virtual environments, package version control, etc., in your projects
- More robust reproducibility methods may be less accessible to scientists with less technical ability than you
 - Solution: Make it so things "just work" without requiring the user to troubleshoot APIs they are unfamiliar with



DEMONSTRATION

- If you are viewing these slides after the fact, please see the recorded presentation for the demonstration
- Ephemeral demo link: https://osssg-demo.netlify.app



• Are your results based on a quantitative analysis?

- If yes, please work through this checklist
- Does your analysis use code?
 - If no, does the software you're using output code? (Most GUI statistics software does)
 - Do you provide code and other documentation sufficient to reproduce all your results?
 - Do you reference the version of all hardware, software, and code used for analysis in your manuscript?
 - Is your code and other documentation version controlled? (Git)
 - Is your code and other documentation deposited in a standard code hosting repository? (GitHub, OSF)
 - Is your code and other documentation in a human and machine-readable format? (written as plain text)
 - Do you use package version control for each of the programming languages in your project?
 - Do you provide a self-contained code execution environment? (Binder, Docker, etc.)

• Do your results rely on a series of pre-processing steps?

- Do you provide code for preprocessing?
 - If no, do you clearly describe all steps, provide a flow-chart of the steps, or refer to a pre-existing publicly available workflow?
- Do your results rely on a series of analysis steps?
 - Do you provide code for each of these steps?
 - If no, do you clearly describe all steps, provide a flow-chart of the steps, or refer to a pre-existing publicly available workflow?
- Do you provide code for automated execution of each step in your scientific pipeline?
 - If no, do you provide detailed descriptions for manual execution of your scientific pipeline?
- Has your code been peer-reviewed for reproducibility and accuracy?
 - If no, are you using Continuous Integration and/or Unit Testing to check for reproducibility?



- Do your results include analysis of data collected by yourself or your collaborators?
 - Is your data available at a digital repository?
 - Do the authors include all meta-data to reproduce the analysis?
 - Do you provide a full account of the procedures used to collect, preprocess, clean, or generate the data?
 - Do you provide research materials and description of procedures necessary to conduct an independent replication of the research?
- Are there any results based on data or materials that cannot be shared for legal or ethical reasons?
 - Do you provide synthetic data with similar statistical properties instead?
 - Do you explain the restrictions on the dataset or materials and how they preclude public access?
 - Do you provide a public description of the steps others should follow to request access to the data or materials?
- Do you provide access to all data and materials for which the constraints do not apply?



• Is your manuscript written in a reproducible format? (R Markdown, Jupyter Notebook)

- Do you report statistics and other numbers using inline code?
- Do you generate, report, and embed tables and figures using code?
- Is all your project code called within your manuscript?
 - If yes, consider splitting project code into separate script files, then calling those within the manuscript for readability
- Is your manuscript output suitable for the forum you plan to share it in? (Preprint, Journal, Talk, Website)
 - If no, do you document the additional steps needed to finish formatting your manuscript?
- Have you gotten anyone to try to reproduce your results?
 - Were they successful?



CHALLENGES

- Learn more about it!
 - Work through the <u>The Turing Way</u>, an open source community-driven guide to reproducible, ethical, inclusive and collaborative data science
 - Listen to one of the reproducible science podcasts linked to at the end of this presentation
- Talk about it!
 - Talk to your collaborators about how you can introduce reproducible workflows into your own projects
- Try it out!
 - Attend our Writing Reproducible Manuscripts workshop in two weeks
 - Try to reproduce the results of the first analysis you ever did
 - See how well your current project fairs against our Reproducibility Checklist
- Implement it!
 - Write your thesis project as a reproducible manuscript
 - Set aside time to check out the coding and reproducibility resources linked to at the end of this presentation
 - Pick one item on the reproducibility checklist and implement it in your next project



THANK YOU! COMMENTS, QUESTIONS?



OPEN-SOURCE ALTERNATIVES

- Mendeley/Endnote alternative:
 - Zotero plus Zotero Connector
 - Import from <u>Mendeley</u> or <u>Endnote</u>
- Useful Zotero plugins:
 - <u>scite</u>
 - <u>pubpeer</u>
 - Better BibTeX
 - <u>zotfile</u>
 - <u>Sci-hub Downloader</u>

- SPSS alternatives with GUI interface:
 - Jamovi
 - JASP
- Code-based SPSS alternatives:
 - <u>R</u> and <u>RStudio</u>
 - <u>Python</u> and <u>RStudio v1.4+</u>
 - Julia
- E-Prime/Presentation/Qualtrics/etc. alternatives:
 - PsychoPy
 - jsPsych
 - Formr



REPRODUCIBILITY IN GENERAL

- Version control:
 - <u>Git</u>
- Data and code distribution, collaboration:
 - GitHub and GitHub Desktop
 - OSF and osfr
- Data repositories:
 - UCalgary Library Guide
 - Nature Recommended Data

Repositories

- Virtual environments:
 - Docker
 - Code Ocean
- Continuous Integration:
 - GitHub Actions
- Web hosting:
 - Netlify



REPRODUCIBILITY IN R

- Use <u>RStudio Projects</u>
- Use <u>inline R code</u> to report statistics
- Package version control:
 - <u>{renv}</u> (works at project level; use from the start)
 - {groundhog} (works at script level; use to recover a reproducible environment in scripts without one)

- <u>{holepunch}</u> (easy Docker sessions)
- Codebooks:
 - {codebook}
- Reproducible workflows:
 - <u>{worcs}</u>
 - {targets}
 - {breakerofchains}
 - <u>Rocker</u>
- Reproducibility w/ interactive data:
 - {shinymeta}

- Learn R Markdown:R Markdown
 - Cookbook
 - <u>R Markdown: The</u> <u>Definitive Guide</u>
- Manuscript writing:
 - {rmarkdown}
 - {bookdown}
 - <u>{distill}</u>
 - {rticles}
 - <u>{papaja}</u>
 - {officeverse}

- Machine-readable hypothesis testing:
 - {scienceverse}
- Unit testing:
 - {testthat}
- GitHub Actions templates:
 - <u>r-lib/actions</u>
- Zotero connector:
 - <u>rbbt</u>
 - <u>citr</u>



REPRODUCIBILITY IN PYTHON

- Use Python Projects
- Use inline python code to report statistics
- <u>The Turing Way</u> has more python reproducibility information
- Package version control:
 - {virtualenv}, {venv} (python virtual environments)
 - {recipy}
 - {sumatra}
- Manuscript writing:
 - Jupyter Notebooks
 - Alternatively, you can use any of the R packages for manuscript writing from the previous slide and run Python code within them
 using the <u>{reticulate}</u> R package
- GitHub Actions guide:
 - Documentation



RESOURCES

- Learn Git and GitHub:
 - Happy Git and GitHub for the useR
 - GitHub Learning Lab
 - <u>Resources</u>
 - GitHub Desktop Documentation
- Learn Jamovi:
 - Documentation
 - <u>Textbook</u>
- Learn JASP:
 - <u>Textbook</u>

- Learn R:
 - Online Books
 - Learn R, in R (Swirl package)
- Learn Python:
 - Python for Data Analysis
 - Automate the Boring Stuff With Python
- Learn Docker:
 - Documentation
 - Hands-on Tutorials



OPEN SCIENCE PODCASTS

- RIOT Science Club:
 - Is there a reproducibility crisis in science?
 - Improve your workflow for reproducible science
 - Five selfish reasons for working reproducibly
 - Synthetic data: A primer
 - <u>Reproducibility in neuroimaging:</u> <u>Problems and solutions</u>
 - JASP and Jamovi
 - <u>Reproducibility in psychiatric genetics</u>

- Reproducibilitea:
 - <u>Reproducibility now</u>
- Everything Hertz:
 - Predicting the replicability of research
 - Large-scale collaborative science
 - Academic hipsters
 - <u>A manifesto for reproducible science</u>
 - Data Sharing
 - Software and coding

