REPRODUCIBLE RESEARCH TOOLS BIOS 692



https://mdozmorov.github.io/BIOS692/

Image credit: Bill & Melinda Gates Foundation, http://k12education.gatesfoundation.org/learning/tools-being-talked-about/

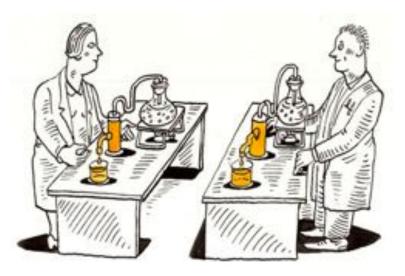
Overview of the course

- We will discuss general principles for reproducible research, but will focus primarily on the practical use of relevant tools (particularly in Linux environment, make, git, data manipulation/ graphics in R, and reports in Markdown/knitr)
- The goal is to ensure that all aspects of computational research you will do (software, data analyses, papers, presentations, posters) are integrated within reproducible framework
- Doing things properly (writing clear, documented, well-tested code) is time consuming, but it will help you many times down the road
- Ultimately, you'll be more efficient, and your work will have greater impact

Introduction

- What is reproducible research?
- Why do we care?
- The cost of reproducibility
- Reproducibility in data science

WHAT IS REPRODUCIBLE RESEARCH?



What is reproducible research?

- Reproducibility is the ability of an entire experiment or study to be duplicated, either by the same researcher or by someone else working independently
- Reproducibility is one of the main principles of the scientific method
- Reproducible research is the ultimate standard for strengthening scientific evidence by independent:
 - Investigators
 - Data
 - Methods
 - Laboratories
 - Instruments

The first reproducible research Galileo Galilei



Galileo's notes directly integrated his data (drawings of Jupiter and its moons), key metadata (timing of each observation, weather, and telescope properties), and text (descriptions of methods, analysis, and conclusions)

Basics of reproducibility Lab notebook

- Complete record of procedures (how), reagents (what), observations (data), and thoughts to pass on to other researchers
- Explanation of why experiments were initiated, how they were performed, and the results
- Legal document to prove patents and defend your data against accusations of fraud
- Scientific legacy in the lab

WHY DO WE CARE?

More data = more chance for errors

- High-throughput biology generates volumes of data
- Data-generating technologies are increasingly used to make clinical recommendations and treatment decisions
- A problem may be overlooked .. Published .. Get in clinical trials

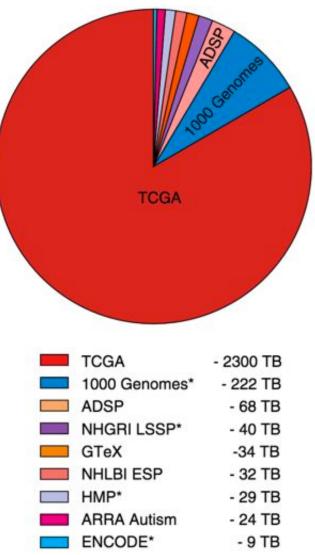


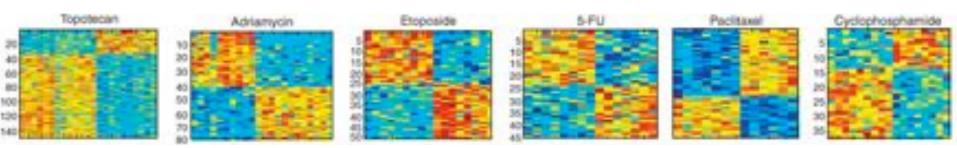
Image credit: Muir et al., "The Real Cost of Sequencing." Genome Biol. 2016 https://genomebiology.biomedcentral.com/articles/10.1186/s13059-016-0917-0

Clinical trials based on flawed and fraudulent data

Genomic signatures to guide the use of chemotherapeutics

Anil Potti^{1,2}, Holly K Dressman^{1,3}, Andrea Bild^{1,3}, Richard F Riedel^{1,2}, Gina Chan⁴, Robyn Sayer⁴, Janiel Cragun⁴, Hope Cottrill⁴, Michael J Kelley², Rebecca Petersen⁵, David Harpole⁵, Jeffrey Marks⁵, Andrew Berchuck^{1,6}, Geoffrey S Ginsburg^{1,2}, Phillip Febbo^{1–3}, Johnathan Lancaster⁴ & Joseph R Nevins^{1–3}

- Described drug response "gene signatures" in NCI60 cell lines
- Demonstrated these "signatures" correspond to patient-specific signatures and can be used to predict patient response to the drugs

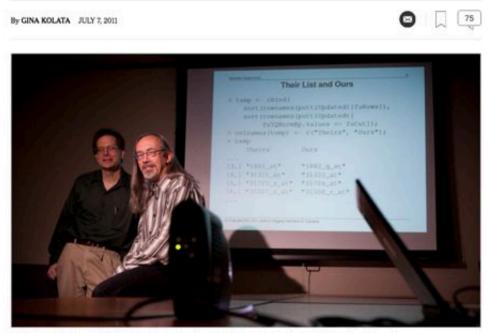


Retraction Watch "The Importance of Being Reproducible: Keith Baggerly tells the Anil Potti story" 2011, http://retractionwatch.com/2011/05/04/the-importance-of-being-reproducible-keith-baggerly-tells-the-anil-potti-story/

Bioinformatics statisticians spot

errors

How Bright Promise in Cancer Testing Fell Apart



Keith Baggerly, left, and Kevin Coombes, statisticians at M. D. Anderson Cancer Center, found flaws in research on tumors. Michael Stravato for The New York Times

When Juliet Jacobs found out she had lung <u>cancer</u>, she was terrified, but realized that her hope lay in getting the best treatment medicine could offer. So she got a second opinion, then a third. In February of 2010, she ended up at <u>Duke University</u>, where she entered a research study whose promise seemed stunning.

"Off-by-one" error

Published

Replicated

• • •	
[3,]	1881_at
[4,]	31321_at
[5,]	31725_s_at
[6,]	32307_r_at

1882_g_at 31322_at 31726_at 32308_r_at

More data added

- Sample ID Response
 - 1 GSM44303 RES
 - 2 GSM44304 RES
 - 3 GSM9653 RES
 - 4 GSM9653 RES
 - 5 GSM9654 RES
 - 6 GSM9655 RES
 - 7 GSM9656 RES
 - 8 GSM9657 RES
 - 9 GSM9658 SEN
- 10 GSM9658 SEN

RES/SEN – resistant/sensitive

- 11 GSM9694 RES
- 12 GSM9695 RES
- 13 GSM9696 RES
- 14 GSM9698 RES
- 15 GSM9699 SEN
- 16 GSM9701 RES
- 17 GSM9708 RES
- 18 GSM9708 SEN
- **19 GSM9709 RES**
- 20 GSM9711 RES

Summary of the Duke case

- A total of 162 co-authors
- 40 papers
- Two-thirds are partially or completely retracted



Inside information on cancer research and drug development

publication date: Nov 19, 2010

Duke University issued the following press release Nov. 19:

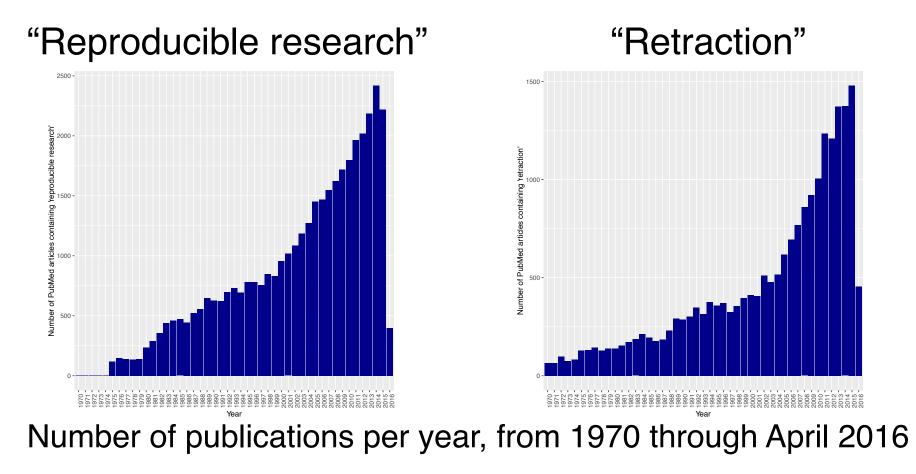
Duke Accepts Potti Resignation; Retraction Process Initiated with Nature Medicine

DURHAM, NC -- Anil Potti, MD, has voluntarily resigned from his positions as associate professor of medicine at Duke University School of Medicine and at the university's Institute for Genome Science & Policy. Dr. Potti's resignation is effective immediately.

In addition, Dr. Potti's collaborator, Joseph Nevins, Ph.D., has initiated a process intended to lead to a retraction request regarding a paper previously published in Nature Medicine. This process has been initiated due to concerns about the reproducibility of reported predictors, and their possible effect on the overall conclusions in this paper. Other papers published based on this science are currently being reviewed for any concerns.

The three clinical trials based on this science for which new enrollment was suspended in mid-July, have been closed.

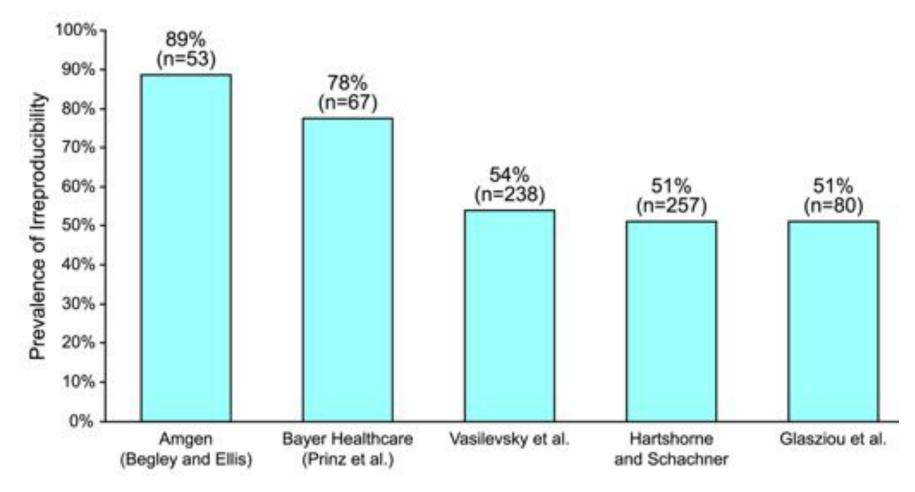
PubMed stats on "Reproducible research" vs. "Retraction"



THE COST OF REPRODUCIBILITY

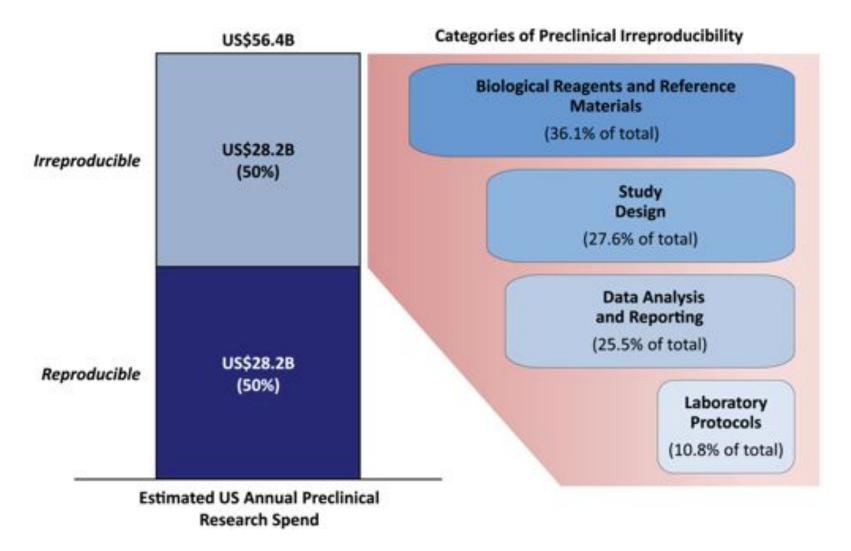


Irreproducibility ranges from 51% to 89%



Leonard Freedman, Iain Cockburn, and Timothy Simcoe, "The Economics of Reproducibility in Preclinical Research." PLOS Biol 2015 http://journals.plos.org/plosbiology/article?id=10.1371/journal.pbio.1002165

Cost of irreproducibility



Leonard Freedman, Iain Cockburn, and Timothy Simcoe, "The Economics of Reproducibility in Preclinical Research." PLOS Biol 2015 http://journals.plos.org/plosbiology/article?id=10.1371/journal.pbio.1002165

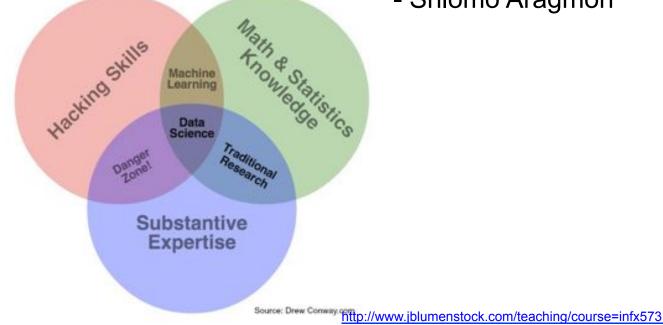
REPRODUCIBILITY IN DATA SCIENCE

Reproducibility in data science

A data scientist is often referred to as someone who knows more statistics than a computer scientist and more computer science than a statistician

- Joshua Blumenstock

Data Scientist = statistician + programmer + coach + storyteller + artist - Shlomo Aragmon



http://bigdata-madesimple.com/what-everybody-ought-to-know-about-data-scientist/

DATA SCIENTIST



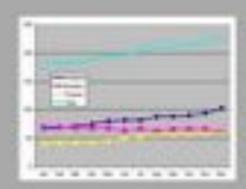
What my friends think I do



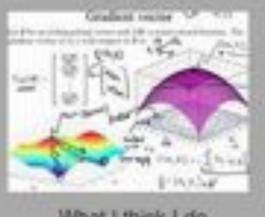
What my mom thinks I do



What society thinks I do



What my boss thinks I do



What I think I do



Steps in reproducible research

The most important is the mindset, when starting, that the end product will be reproducible.

– Keith Baggerly

- Experimental design
- Data generation
- Data analysis
- Results interpretation
- Dissemination of results

Common approach: write report around results

Point and click approach

Use MS Excel for data entry/cleaning/preparation, and possibly statistical analysis

Problems

- With point-and-click, there's no way to record/save the steps that generated the (copy/pasted) results
- Data files are kept separately from the analysis code, and from reports
- After modifications of one of the files, it becomes unclear which version corresponds exactly to the reported results
- Every time something changes, you have to regenerate the figures/results/reports by hand – very time consuming

Better approach: write report that generates results

- The report is automated via code
- Data is attached to the welldocumented code
- History of any changes should be preserved

The final report should be self-sufficient and reproducible with a single command

The importance of software

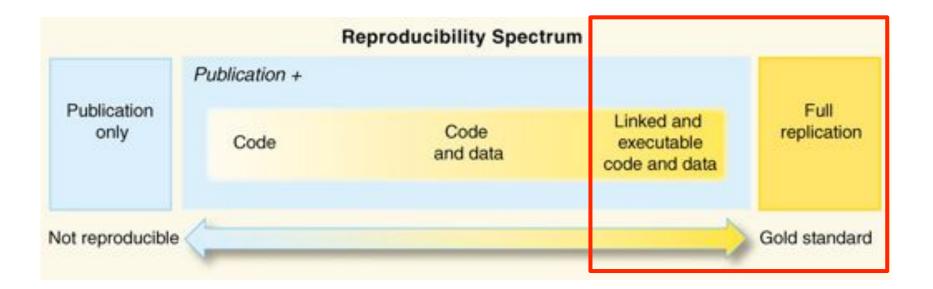
An article about computational science in a scientific publication is **not** the scholarship itself, it is merely **advertising** of the scholarship. The actual scholarship is the complete software development environment and the complete set of instructions which generated the figures.

J. B. Buckheit and D. L. Donoho. (1995)

http://www-stat.stanford.edu/~donoho/Reports/1995/wavelab.pdf

Software + data ⇔ reproducible report

Distribute fully automated report with data and code



Quality of reproducible report

- Are the tables and figures reproducible from the code and data?
- Does the code actually do what you think it does?
- In addition to what was done, is it clear why it was done? (e.g., how parameter settings were chosen?)
- Is your code scalable to accommodate more data/ methods?

Six degrees of reproducibility

- 5: The results can be easily reproduced by an independent researcher with at most 15 min of user effort, requiring only standard, freely available tools (C compiler, R, Python, etc.)
- **4:** Easy reproducibility, but require some proprietary source packages (MATLAB, SAS, etc.)
- **3**: Reproducibility requires considerable effort
- **2:** Reproducibility requires extreme effort
- **1**: The results cannot seem to be reproduced
- **0**: The results cannot be reproduced

Scientific computing

...SCIENTISTS AND THEIR SOFTWARE

A survey of nearly 2,000 researchers showed how coding has become an important part of the research toolkit, but it also revealed some potential problems.

> **45%** said scientists spend more time today developing software than five years ago."

> **SUZ** of scientists spend at least one fifth of their time developing software.

> Only **Mark** of scientists have a good understanding of software testing.

> Only **34%** of scientists think that formal training in developing software is important. ... PRACTICING SAFE SOFTWARE > Five tips to make scientific code more robust.

Use a version-control system:
Put source code, raw data files, parameters and other primary
material into it to record what you did, and when.

Track your materials:

Know the source of your software. Keep a record of what raw data were processed to produce a particular result, what tools were used to do the processing, and how the tools were set up.

🖶 Write testable software:

Build large codes from smaller, easily testable chunks.

Test the software: And get somebody else to read it and look for bugs.

Encourage sharing of software:

Make the code that you use in research freely available, when possible.