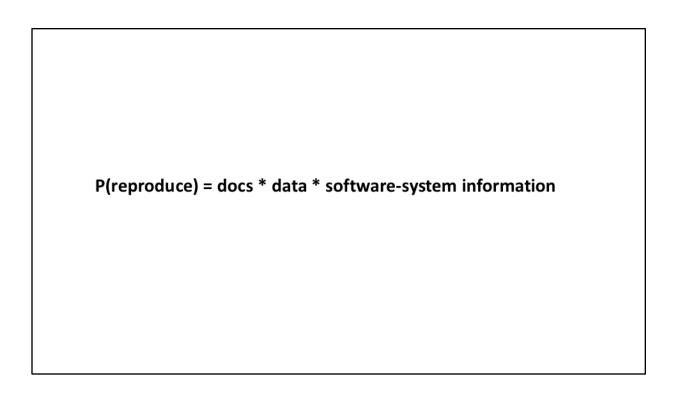


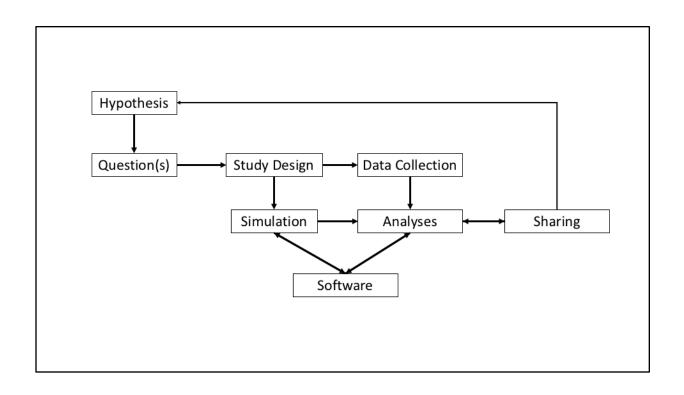
- Introduce youreself = Matt, external fellow, former REU, Originally from LA, PhD NAU, main research is community genetics of ecological networks
- Close computers, no cell phones
- Wordcloud = we're all here to learn something (BASICS, PLOTTING, MORE COMFORTABLE)
- STATISTICAL ANALYSIS = Not teaching stats, but R facilitates learning both through the language itself and through the open-source community
- Focus = R is for Reproducibility



Your moment of zen. Reproducibility is a goal not a state.

- Science is driven by the exchange of information and knowledge.
- Currently, there is a lot that we can do
- A recent study (Stodden *et al.* 2018) demonstrated that only 26% of studies published in *Science* could be reproduced. This was even more striking given that the study was conducted after the *Science* had instituted it's open data policy.
- Luckily, advances in open-source computer languages, such as **R** and **python**, provide a way to produce computations that can more easily document scientific research in a transparent, easily shared way.
- In this course, we will cover how to conduct **reproducible** scien-tific research using the **R** programming language and supporting software that will enable researchers to more clearly and easily docu-ment projects.
- Goal = coding in **R** using the *RStudio* IDE and the *git* version control system.







A Tool for Working Smarter (Transparency): code is a part of science and should be shared

- 1. Setup your project so that there is a clear architecture (**RStudio**)
- 2. Work so that your computation from initial data to finished results will be coded (wherever possible), including data cleaning and processing steps (R)
- 3. Keep track of versions of your code (git)
- 4. Make initial data available (whenever possible, **Github**)
- 5. Keep track of software dependencies (packrat)
- 6. Be organized, succinct in style, coding and documentation (*formatR*, *Rmarkdown*)

- Philosophy: Reproducibility & Benefaction
 - Science 44% w/ artefacts, 24% reproducible
 - FAIR
- Experiment Design and Observation
- Data Digitizing and Storage
- Data Wrangling (IO, assign, object, index, type, paths)
- Check & Test (conditionals)
- Computation Analysis, Modeling, etc.
- Results
 - Plotting
- Best Practices
 - Notebooks (data + code + annotation)
 - Style (formatR)
 - Data and code = Open Access
 - Github, Dataverse, Figshare, Dryad, EDI, NCBI
- Preservation
 - Capsules (ReproZip)
 - VMs (system independence)
 - Provenance (provR)

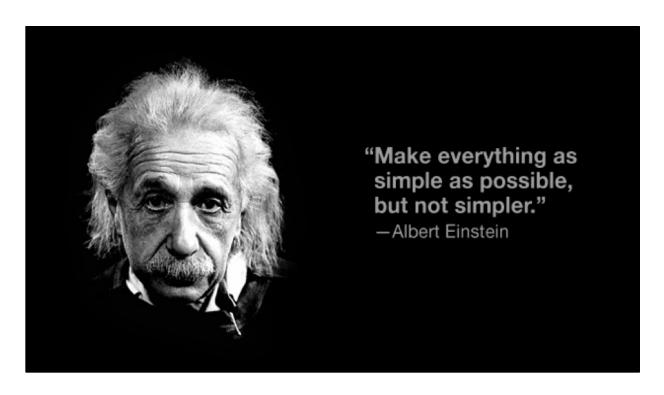
- Code cleaning (cleanR)
- Continuous integration (Travis)
- Engagement (Websites, Blogs, Twitter, etc.)
- Life-long learning ROpenSci, stackoverflow, Tidyverse



- Spreadsheets and databases for data entry, not analysis. If "Data wrangling" and analyses are coded, they are more transparent and easier to redo.



- Scientific "Scripts" are goal oriented, usually "one-offs", leading to complicated "brittle machines"
- Aiming for shareable code is not the main goal, but it is important
- We want to turn on the light now, but making it easier for others to turn on the light is more beneficial



Style



R is like an iceburg. This class will go as deep as we can.