

# Reproducible Research

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“Research is reproducible if it can be reproduced by others”

One of the main principles of the scientific method

# Definition of Reproducible Research

A **complete description** of the data and the analysis of that data — including computer programs — so the results can **be exactly reproduced by others**.

Amstat News, 1 January 2011

Someone unfamiliar with your project should be able to look at your computer files and understand in detail **what you did** and **why**.

William S. Noble



# Forensic Bioinformatics

*The Annals of Applied Statistics*  
2009, Vol. 3, No. 4, 1309–1334  
DOI: 10.1214/09-AOS1291  
© Institute of Mathematical Statistics, 2009

## DERIVING CHEMOSENSITIVITY FROM CELL LINES: FORENSIC BIOINFORMATICS AND REPRODUCIBLE RESEARCH IN HIGH-THROUGHPUT BIOLOGY

BY KEITH A. BAGGERLY<sup>1</sup> AND KEVIN R. COOMBES<sup>2</sup>

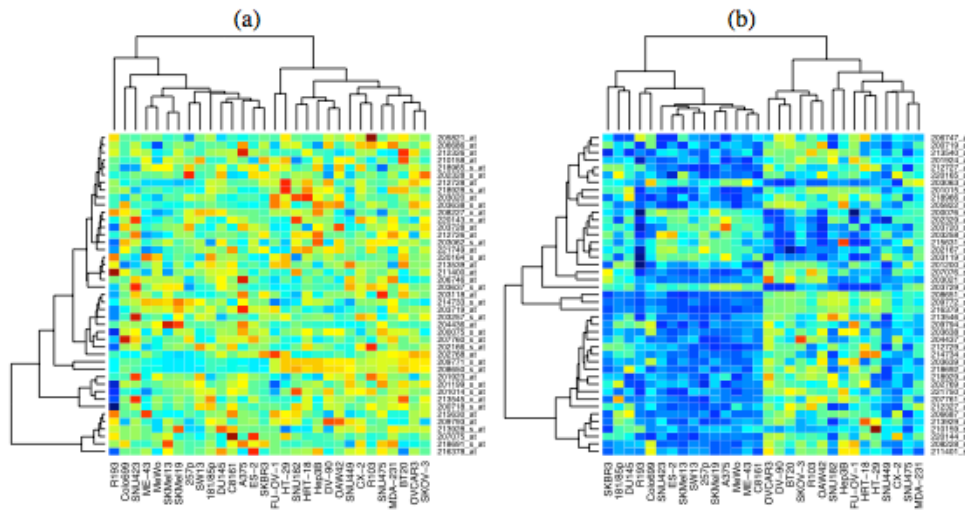
*University of Texas*

High-throughput biological assays such as microarrays let us ask very detailed questions about how diseases operate, and promise to let us personalize therapy. Data processing, however, is often not described well enough to allow for exact reproduction of the results, leading to exercises in “forensic bioinformatics” where aspects of raw data and reported results are used to infer what methods must have been employed. Unfortunately, poor documentation can shift from an inconvenience to an active danger when it obscures not just methods but errors. In this report we examine several related papers purporting to use microarray-based signatures of drug sensitivity derived from cell lines to predict patient response. Patients in clinical trials are currently being allocated to treatment arms on the basis of these results. However, we show in five case studies that the results incorporate several simple errors that may be putting patients at risk. One theme that emerges is that the most common errors are simple (e.g., row or column offsets); conversely, it is our experience that the most simple errors are common. We then discuss steps we are taking to avoid such errors in our own investigations.



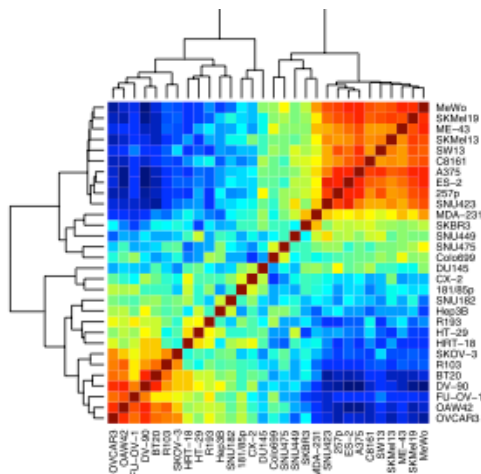
Keith Baggerly, Ph.D.

# Reconstructing heatmap for cisplatin signature

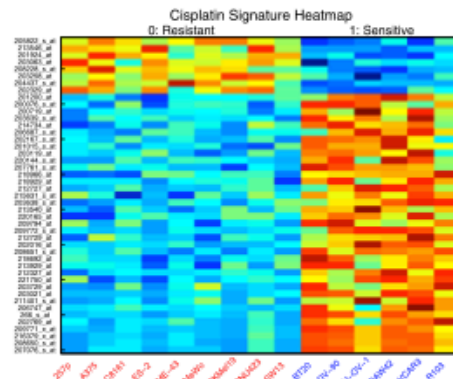


orig data: no structure

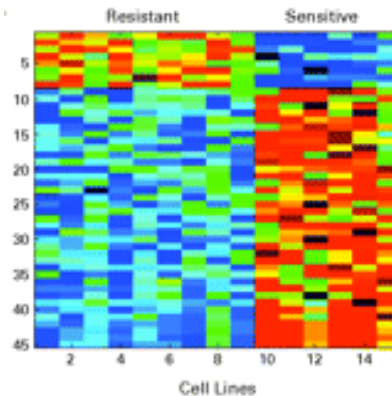
offsetting by one  
(indexing error)



pairwise sample correlations  
to detect label switches



reconstructed heatmap



original heatmap  
Hu et al. (2007)

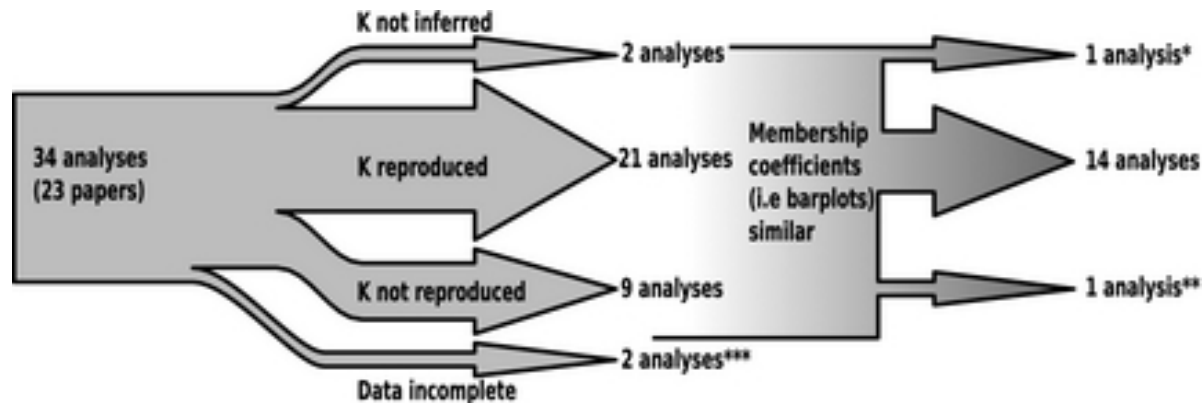
# Conclusions of Baggerly and Coombes

- The most common problems are simple:
  - confounding in the experimental design
  - mixing up sample labels
  - mixing up the gene labels (off-by-one errors)
  - mixing up the group labels (sensitive/resistant)
- Most of these mix-ups involve simple switches or offsets.
- These mistakes are easy to make, particularly if working with Excel/oocalc
- Or if working with 0/1 labels instead of names

These easy-to-fix errors are often hidden (incomplete documentation and code not shared)

# Population Genomics

UBC Reproducibility Group could not reproduce the results in 30% of published analyses using the population genetic package STRUCTURE, using the same data as provided by the authors



<http://onlinelibrary.wiley.com/doi/10.1111/j.1365-294X.2012.05754.x/abstract>



# Gene name errors are widespread

- Automatic conversion of gene symbols to dates and floating-point numbers in Excel

Gene symbol	converted to
SEPT2 (Septin 2)	2-Sep
SEPT2	2006/09/02
MARCH1 [Membrane-Associated Ring Finger (C3HC4) 1, E3 Ubiquitin Protein Ligase]	1-Mar
2310009E13	2.31E+13

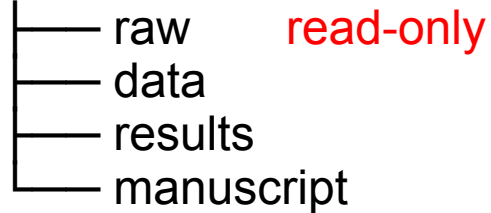
- 20% of papers in leading genomics journals with supplementary Excel gene lists have erroneous gene name conversions
- 39.7% of Excel files with gene lists deposited to NCBI GEO (4321 screened) contain gene name errors

# Practical aspects

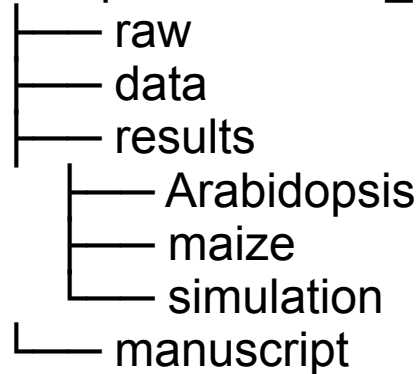
- How do organize your folders?
- How you name your files?

# Folder organization

ComparGenomics\_mea



ComparGenomics\_mea

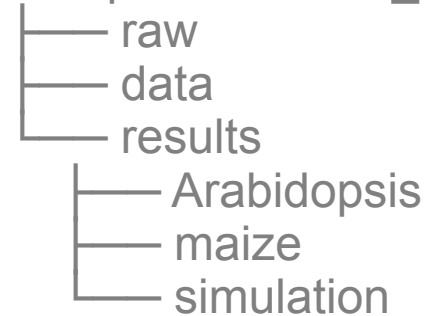


- Result folders contain scripts
- Each file exists only once - use aliases (shortcuts / symbolic links)

# ...different tastes...

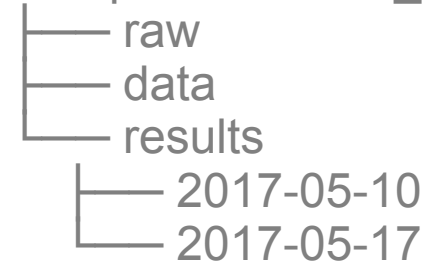
## Descriptions

ComparGenomics\_mea



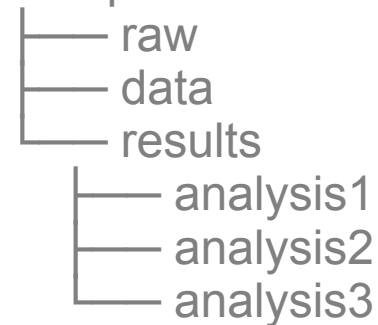
## Dates

ComparGenomics\_mea



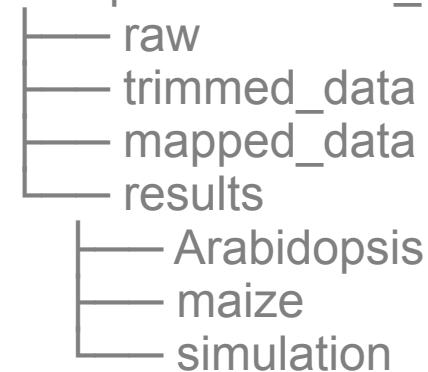
## Numbered

ComparGenomics\_mea



## Descriptions 2

ComparGenomics\_mea



# Description files

- We can never fit in a single filename all the metadata to describe a file
- Each folder contains a Readme.txt/Notebook/Description file that describes the folder
- Format: plain text or Markdown

ComparGenomics\_mea

Readme.txt

High-level description: Introduction, why, thoughts, conclusions

|— raw

SampleDescription.txt

|— data

|— results

Readme.txt

low-level description, describes all analyses, links to plots

|— Arabidopsis

|— maize

|— simulation

|— manuscript

# 3 Principles for file naming

1. machine readable
2. human readable
3. plays well with default ordering
  - put something numeric first

# File naming example

01_marshall-data.r	01.r
02_pre-dea-filtering.r	02.r
03_dea-with-limma-voom.r	03.r
04_explore-dea-results.r	04.r
90_limma-model-term-name-fiasco.r	90.r
helper01_load-counts.r	helper01.r
helper02_load-exp-des.r	helper02.r
helper03_load-focus-stainfo.r	helper03.r
helper04_extract-and-tidy.r	helper04.r

# File naming example 2

2017-06-26\_BRAFWTNEGASSAY\_Plasmid-Cellline-100-1MutantFraction\_H01.csv  
2017-06-26\_BRAFWTNEGASSAY\_Plasmid-Cellline-100-1MutantFraction\_H02.csv  
2017-06-26\_BRAFWTNEGASSAY\_Plasmid-Cellline-100-1MutantFraction\_H03.csv  
2017-06-26\_BRAFWTNEGASSAY\_FFPE-CRC-1-41-A01.csv  
2017-06-26\_BRAFWTNEGASSAY\_FFPE-CRC-1-41-A02.csv  
2017-06-26\_BRAFWTNEGASSAY\_FFPE-CRC-1-41-A03.csv

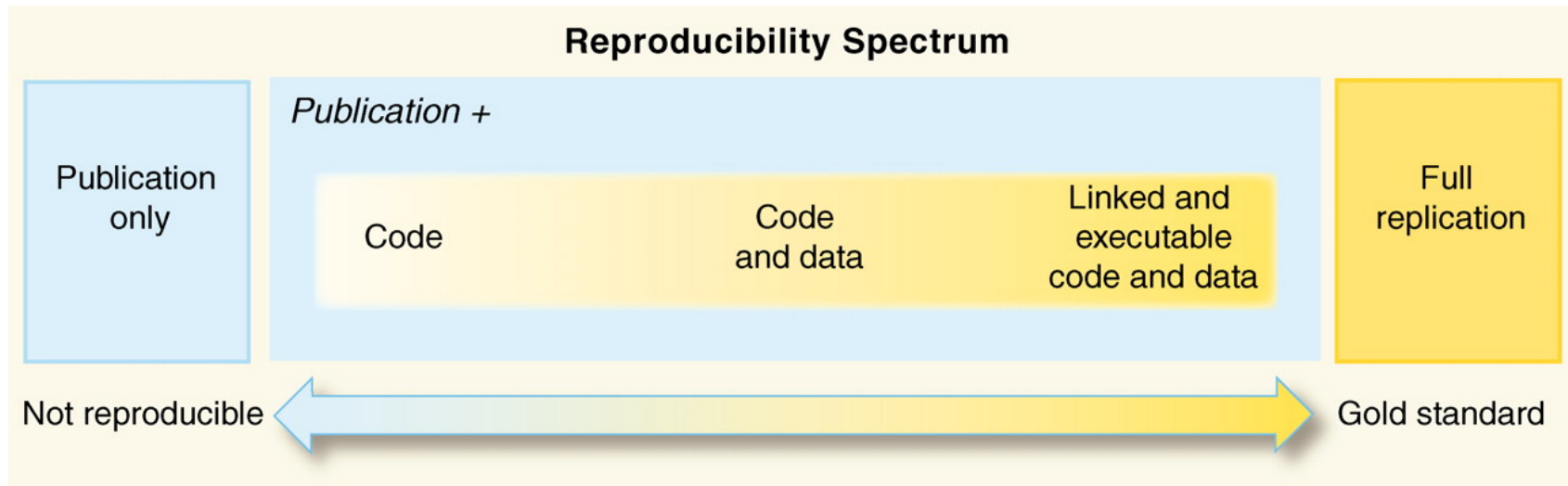
## Machine-readable

- easy to search for later
- easy to narrow file lists based on names
- easy to extract info from file names (e.g. by splitting)
- no spaces, punctuation, accents

Jenny Brian



# The spectrum of reproducibility



<http://science.sciencemag.org/content/334/6060/1226.full>

- A minority of the papers available today provide code and data
- Making articles reproducible takes time and effort

Partial reproducibility is better than nothing!

# Research as an iterative process

Everything you do, you will probably have to do over again

modifications in preprocessing / analysis  
more / new data  
new group members

Reproducible research -> less friction, time-saver in the longer run

# Handy tools

- Electronic notebooks  
allow to mix text, code and plots in the same document

R: rmarkdown, knitr, Jupyter Notebook

Python: Jupyter Notebook

- "Workflow managers"  
run analyses using GUI  
take care of many aspects (history, ..)

Galaxy, sushi (FGCZ)

- Virtualization  
Virtual machines, Docker

# Sources & Links

Presentation by Frédéric Schütz (SIB Lausanne)

Good Enough Practices in Scientific Computing  
<http://arxiv.org/abs/1609.00037>

Ten Simple Rules for Reproducible Computational Research  
<http://journals.plos.org/ploscompbiol/article?id=10.1371/journal.pcbi.1003285>

## **More things about reproducibility**

Article collection from Nature  
<http://www.nature.com/news/reproducibility-1.17552>

Nature poll about reproducibility crisis  
<http://www.nature.com/news/1-500-scientists-lift-the-lid-on-reproducibility-1.19970>

Practical Data Science for Stats - a PeerJ Collection  
<https://peerj.com/collections/50-practicaldatascistats/>

# Dynamic reports

- Combining thoughts and code
- Dynamic creation of figures  
One can delete all figures and recreate them at will
- R: rmarkdown, knitr, Jupyter Notebook  
python: Jupyter/IPython Notebook
- Demo  
<http://rmarkdown.rstudio.com/lesson-1.html>  
<http://jupyter.org/>
- Problems:
  - reduced development environment (<-> Rstudio)
  - long-running calculations
  - big files
  - code to wrangle/polish data might be long