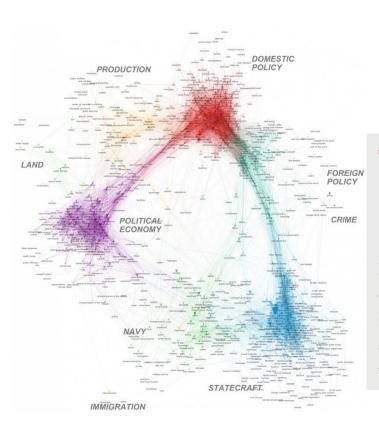
Text Mining & Reproducible Research

PEETER TINITS

#DIGMET SUMMER SCHOOL, TARTU
28.08.2019

Text mining



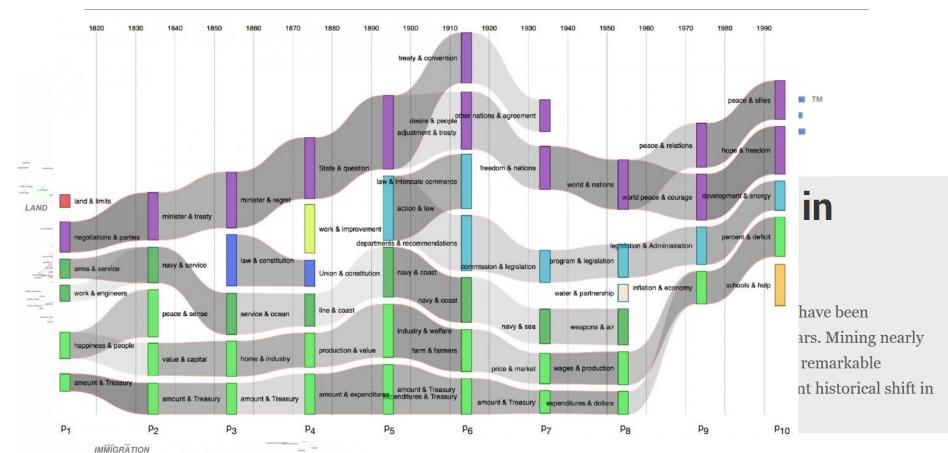
SCIENCE NO DE

Text mining strikes gold in political discourse

From George Washington to Barack Obama, US presidents have been delivering the State of the Union address for the last 225 years. Mining nearly 2 million words, researchers at Columbia University trace a remarkable stability amid the discourse streams and identify a significant historical shift in the American notion of governance.

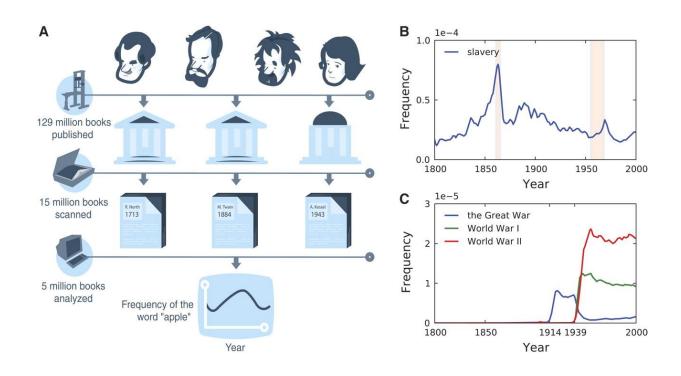
Rule et al. 2015 Lexical shifts, substantive changes, and continuity in State of the Union discourse, 1790–2014

Text mining



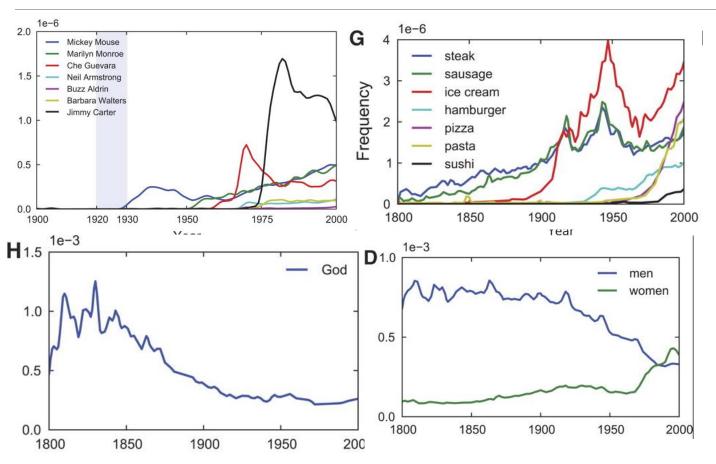
Rule et al. 2015 Lexical shifts, substantive changes, and continuity in State of the Union discourse, 1790–2014

"All digitized texts"



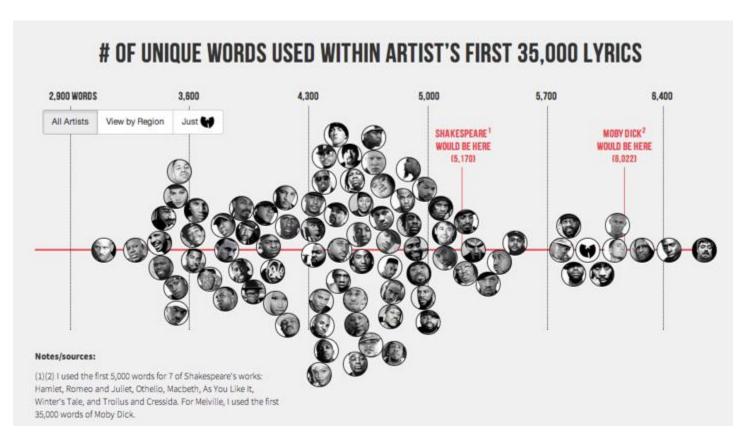
Michel et al. 2008 Quantitative Analysis of Culture Using Millions of Digitized Books

"All digitized texts"



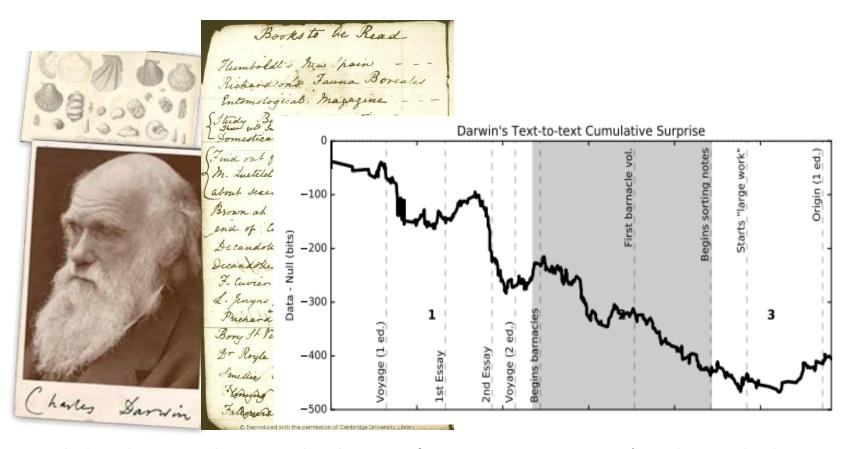
Michel et al. 2008 Quantitative Analysis of Culture Using Millions of Digitized Books

Vocabulary of Rap



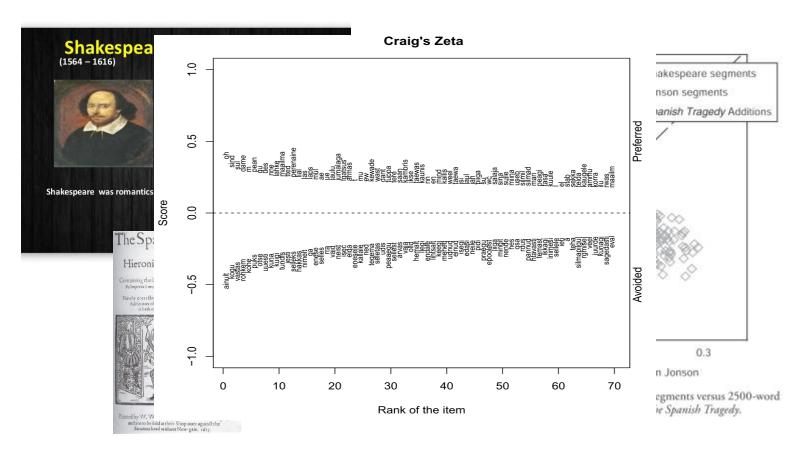
https://pudding.cool/projects/vocabulary/

Darwin's reading habits



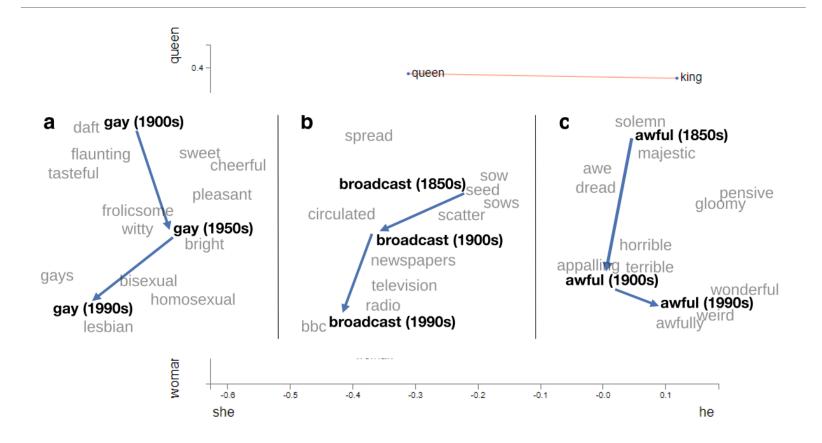
Murdock et al. 2017. Exploration and exploitation of Victorian science in Darwin's reading notebooks.

Finding the author

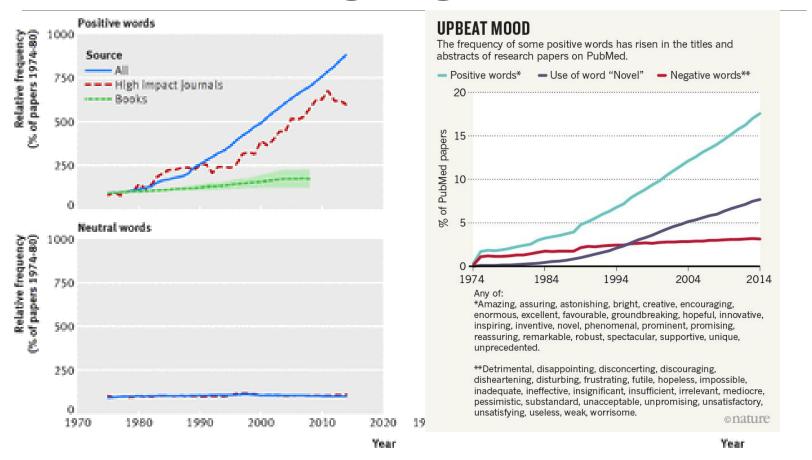


Craig & Kinney 2009. Shakespeare, Computers, and the Mystery of Authorship

Word semantics



Scientific language



Text mining

textual data <-> questions

Questions matter.

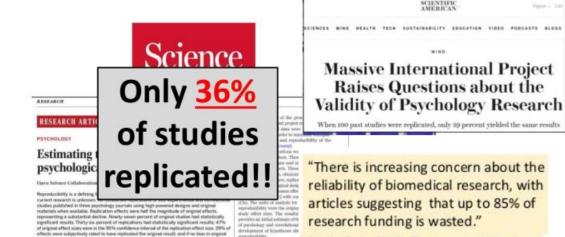
Technology can be easy (if data is available).

Though not a magic bullet.

Reproducible research

Replication crisis





results is assumed, combining original and replication results left 68% with statistically significant effects. Correlational tests suggest that replication success was better predicted by the strength of regional evidence than its characteristics of the original and replication beams.





Katsete korratavus kütab psühholoogias kirgi



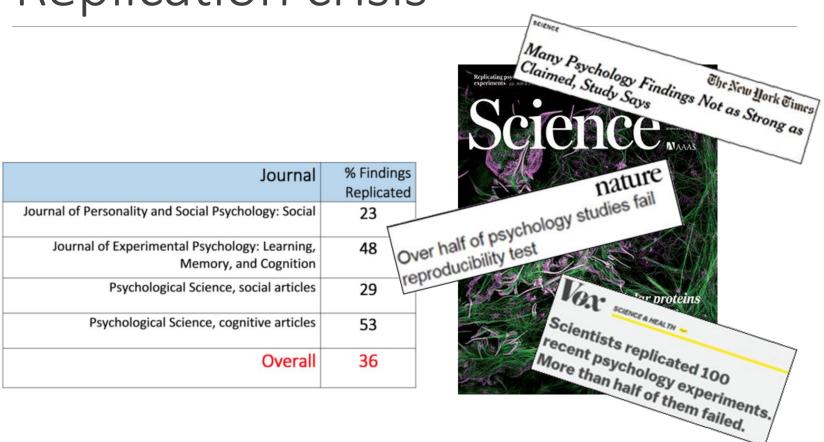
Analüüs: teaduskirjandus on kiivas, kriisist pole mõtet rääkida

Bustin, S. A. (2015). The reproducibility of biomedical research: Sleepers awake!

Biomolecular Detection and

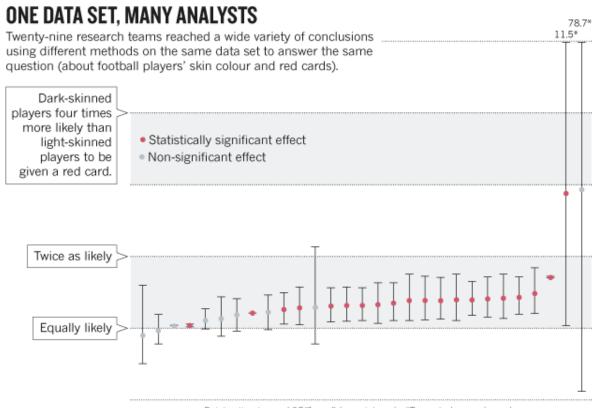
Quantification

Replication crisis



Open Science Collaboration 2015. Estimating the reproducibility of psychological science.

Same data, different conclusions



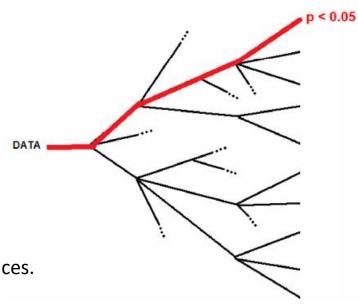
Point estimates and 95% confidence intervals. *Truncated upper bounds.

Silberzahn et al. 2018. Many Analysts, One Data Set: Making Transparent How Variations in Analytic Choices Affect Results

Garden of forking paths

"Researcher degrees of freedom"



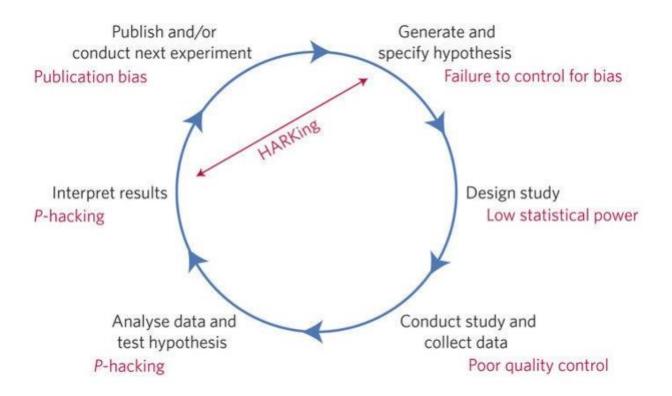


Problem is, if you pretend that there were no choices.

This was always the path we were going to take.

Andrew Gelman & Eric Loken 2013. The garden of forking paths: Why multiple comparisons can be a problem, even when there is no "fishing expedition" or "p-hacking" and the research hypothesis was posited ahead of time. (Unpublished.)

Problems all over the research cycle



Munafo et al. 2017. A manifesto for reproducible science

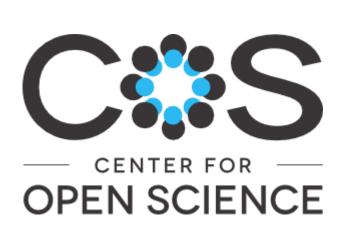
Open science solutions

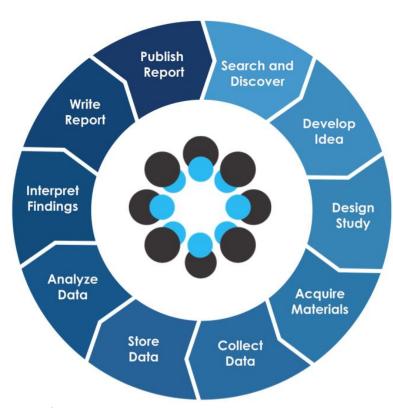
Openness

Fairness

Transparency

Open science solutions





https://osf.io/

Standards and practices

Community Page

Best Practices for Scientific Computing

Greg Wilson¹*, D. A. Aruliah², C. Titus Brown³, Neil P. Chue Hong⁴, Matt Davis⁵, Richard T. Guy⁶*, Steven H. D. Haddock⁷, Kathryn D. Huff⁸, Ian M. Mitchell⁹, Mark D. Plumbley¹⁰, Ben Waugh¹¹, Ethan P. White¹², Paul Wilson¹³

1 Mozilla Foundation, Toronto, Ontario, Canada, 2 University of Ontario Institute of Technology, Oshawa, Ontario, Canada, 3 Michigan State University, East Lansing, Michigan, United States of America, 4 Software Sustainability Institute, Edinburgh, United Kingdom, 5 Space Telescope Science Institute, Baltimore, Maryland, United States of America, 6 University of Toronto, Toronto, Ontario, Canada, 7 Monterey Bay Aquarium Research Institute, Moss Landing, California, United States of America, 8 University of California Berkeley, Berkeley, California, United States of America, 9 University of British Columbia, Vancouver, British Columbia, Canada, 10 Queen Mary University of London, London, United Kingdom, 11 University College London, London, United Kingdom, 12 Utah State University, Logan, Utah, United States of America, 13 University of Wisconsin, Madison, Wisconsin, United States of America

https://doi.org/10.1371/journal.pbio.1001745

Standards and practices

Commu

Best

PERSPECTIVE

Greg Wi

Steven

Ethan P.

1 Mozilla Fou Michigan, Ur States of Am 8 University University of 13 University

Good enough practices in scientific computing

Greg Wilson¹⁶*, Jennifer Bryan²⁶, Karen Cranston³⁶, Justin Kitzes⁴⁶, Lex Nederbragt⁵⁶, Tracy K. Teal^{6©}

- 1 Software Carpentry Foundation, Austin, Texas, United States of America, 2 RStudio and Department of Statistics, University of British Columbia, Vancouver, British Columbia, Canada, 3 Department of Biology, Duke University, Durham, North Carolina, United States of America, 4 Energy and Resources Group, University of California, Berkeley, Berkeley, California, United States of America, 5 Centre for Ecological and Evolutionary Synthesis, University of Oslo, Oslo, Norway, 6 Data Carpentry, Davis, California, United States of America
- These authors contributed equally to this work.
- * gvwilson@software-carpentry.org

https://doi.org/10.1371/journal.pbio.1001745

FAIR data

What is FAIR DATA?

Findable C
Accessible
Interoperable
Reusable



Data and supplementary materials have sufficiently rich metadata and a unique and persistent identifier.

FINDABLE



Metadata and data are understandable to humans and machines. Data is deposited in a trusted repository.

ACCESSIBLE



Metadata use a formal, accessible, shared, and broadly applicable language for knowledge representation.

INTEROPERABLE



Data and collections have a clear usage licenses and provide accurate information on provenance.

REUSABLE

https://www.go-fair.org/fair-principles/

Open data



https://scienceport.tut.fi/openaccess/whyoa

Practically, what it means

Open data

Easy to use and access

Open scripts

Published, verifiable, reusable

Multiple independent analysis

Multilab studies: e.g. https://osf.io/j9ady/

Building on reusable elements

Learning their limits

Primary motivations are selfish

Confidence

Comfort

Easy to share

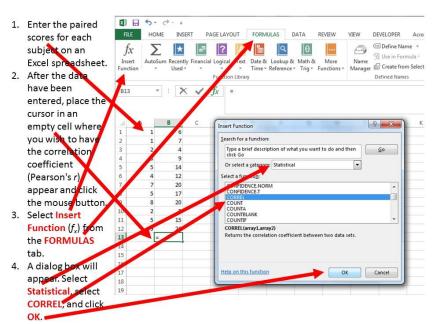
Easy to keep

Easy to reuse

Comfort

How to conduct linear regression

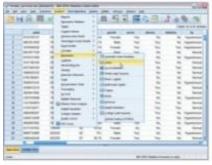
EXCEL



SPSS

versus

₹



- 1. Pay \$\$\$
- 2. Click Analysis
- 3. Click Regression
- 4. Click Linear
- 5. Check box Y...

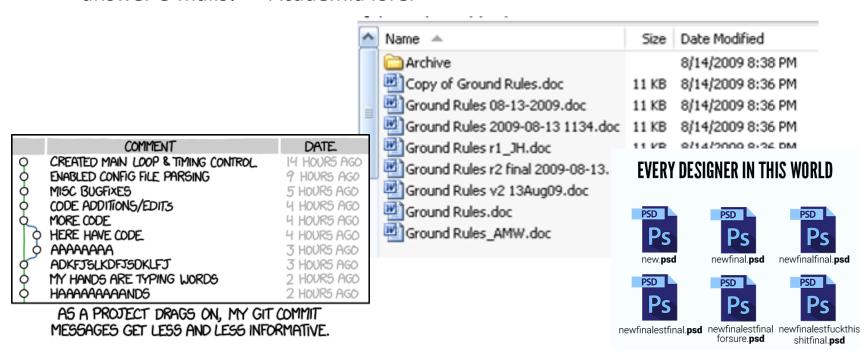


- Type "lm(y~x)"
- 2. Facebook...
- 3. YouTube...
- 4. Facebook...

Keeping track

Open scripts is also documentation

 "Your best/worst collaborator is YOU from 6 months ago, and they do not answer e-mails!" – Academia lore.



Sharing is caring

Share your work. Be successful.

Open scholarship is good for the public and for you.







Increase your visibility

Reduce publishing costs

Take back control

http://whyopenresearch.org/

Other opinions

Five selfish reasons

- Reason 1: reproducibility helps to avoid disaster
- Reason 2: reproducibility makes it easier to write papers
- Reason 3: reproducibility helps reviewers see it your way
- Reason 4: reproducibility enables continuity of your work
- Reason 5: reproducibility helps to build your reputation

Florian M. 2015. <u>Five selfish reasons to work reproducibly</u>
Open Science Training Handbook. 2018. <u>Reproducible Research and Data Analysis</u>
Gatto, L. 2019. <u>Becoming a better scientist with open and reproducible research</u>
Reproducible Research MOOC at coursera.org

Terminology

Technically:

- Replicability a new experiment
- Reproducibility can I rerun the code?

Today: Reproducibility!