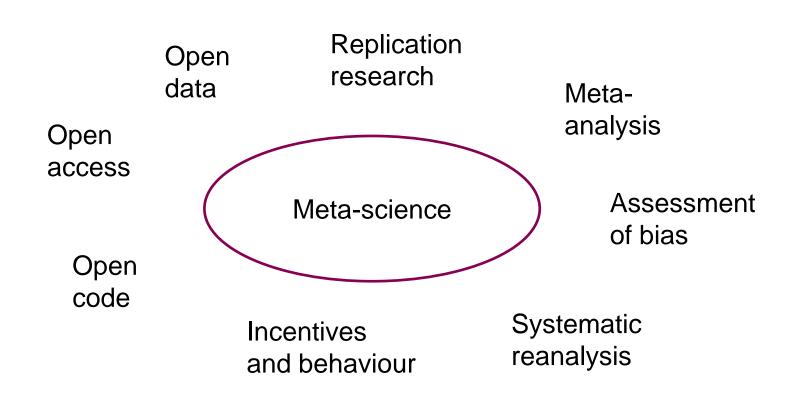


Open data in research on humans – promises and challenges

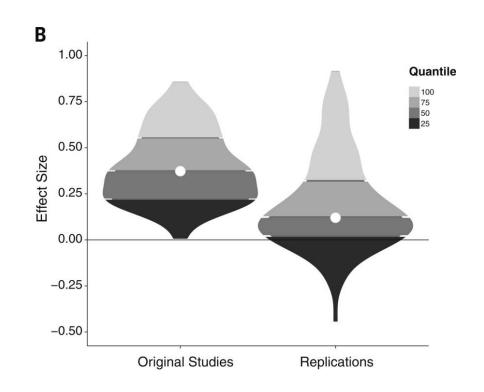
Gustav Nilsonne 2019-08-22



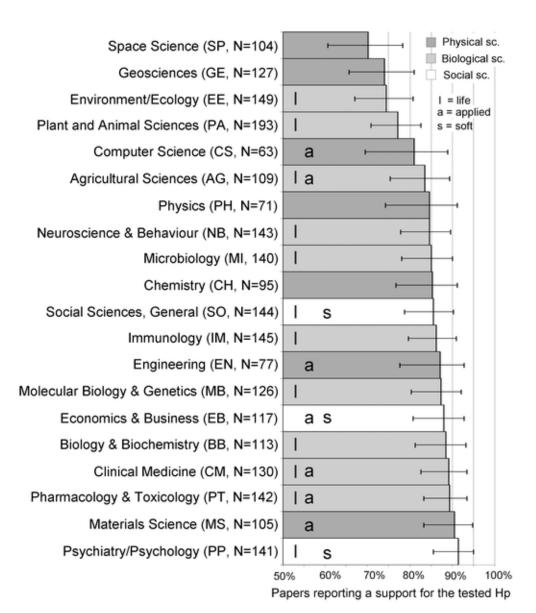


Reproducibility Project: Psychology

- Aim: to obtain an initial estimate of reproducibility in experimental psychology
- Attempted replication of 100 studies
- 36/97 studies showed a statistically significant effect in the same direction as the original study
- Effect size halved on average



Share of positive findings in different fields



ROYAL SOCIETY OPEN SCIENCE

rsos.royalsocietypublishing.org

Research





Cite this article: Ingre M, Nilsonne G. 2018 Estimating statistical power, posterior probability and publication bias of psychological research using the observed replication rate. *R. Soc. open sci.* **5**: 181190.

http://dx.doi.org/10.1098/rsos.181190

Estimating statistical power, posterior probability and publication bias of psychological research using the observed replication rate

Michael Ingre^{1,2} and Gustav Nilsonne^{1,3,4}

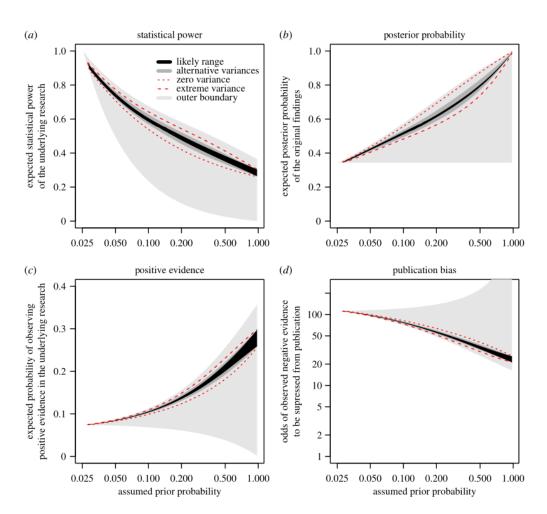
¹Department of Clinical Neuroscience, Karolinska Institutet, Solna, Sweden

²Institute for Globally Distributed Open Research and Education (IGDORE), Sweden

³Stress Research Institute, Stockholm University, Stockholm, Sweden

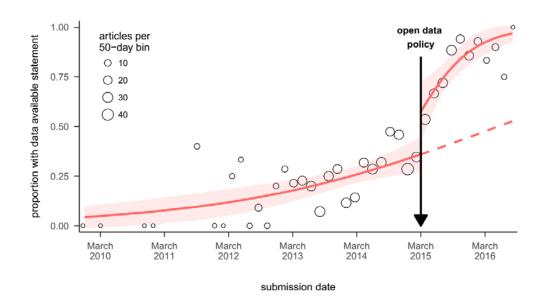
⁴Department of Psychology, Stanford University, Stanford, CA 94305, USA

Properties of a literature with 90% positive findings and 36% reproducibility rate



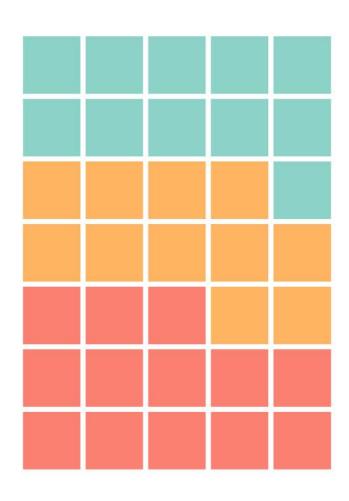
Computational reproducibility: a case study

- The journal Cognition introduced an open data policy in 2015
- We retrieved data from 35 papers and attempted to reproduce the main reported results



Hardwicke et al, Royal Society Open Science 2018

One third fully reproducible from data and published methods description

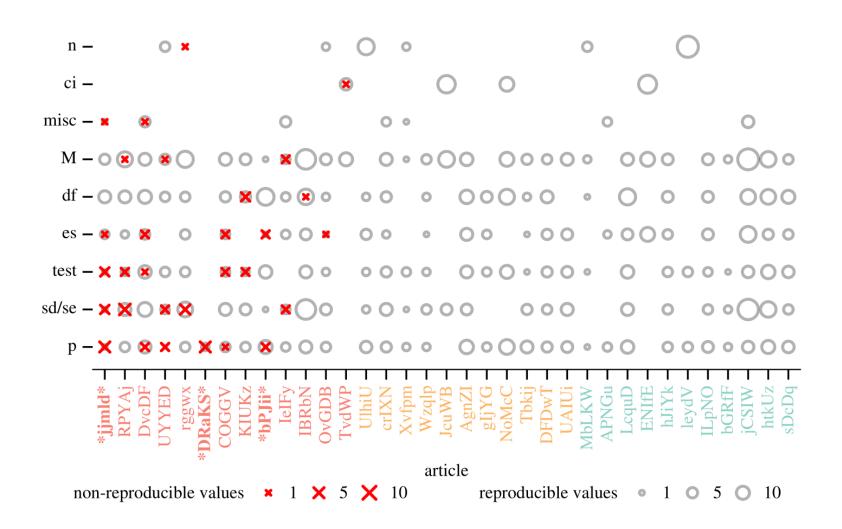


11 (31%) Reproducible

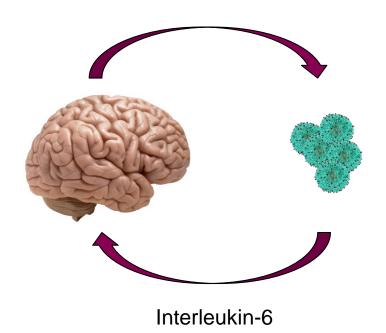
11 (31%)
Reproducible
with author assistance

13 (37%)
Not fully reproducible despite author assistance

Irreproducible numbers found in all categories; *p*-values most common



Does the immune system make us sleepy at night? Lessons from a meta-analysis



IL-6 in Units/ml

Ill-6 in Units/ml

Ilights off EEG

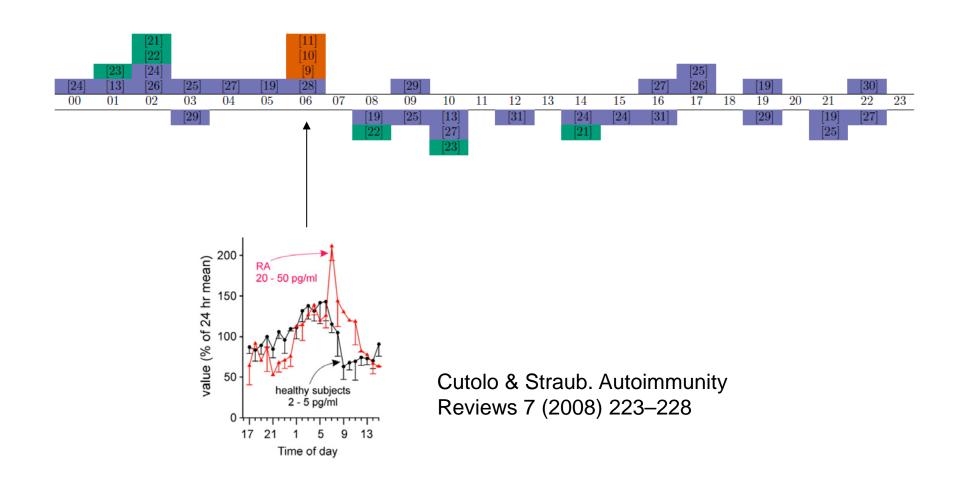
Ilights off EEG

Time

Test subject 1 ** test subject 2 ** test subject 3

Bauer et al. Clin Investig. 1994 72(4):315

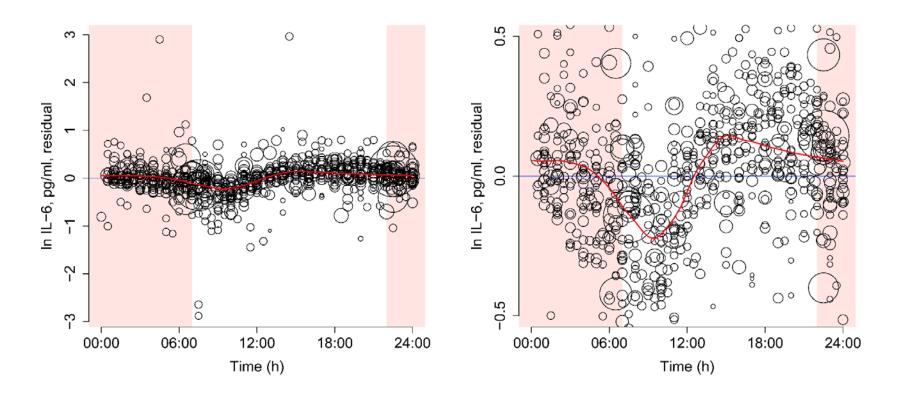
Interleukin-6 over the day: reported peaks and troughs



Data for meta-analysis

- 43 studies with 56 datasets from a total of 1100 persons could be excluded in a meta-analysis
- Individual participant data were available from 3 datasets
- Another 36 studies fulfilled inclusion criteria, but could not be included because data were unavailable (k = 25), because data had been previously reported (k = 7) or because they reported data far from a physiological range (k = 4)

Meta-analysis

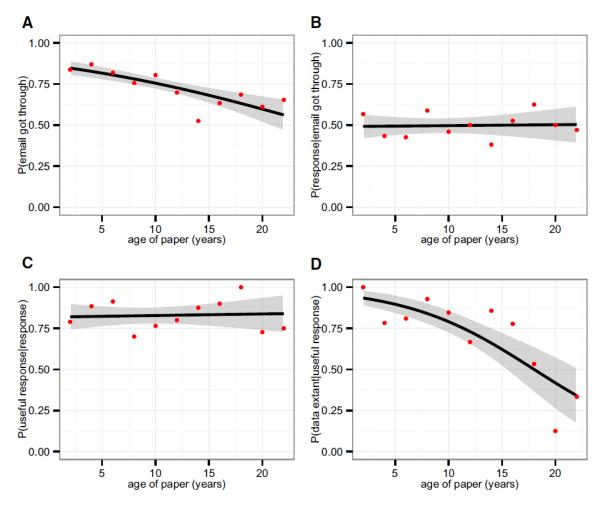


Nilsonne et al. Plos One 2016. https://doi.org/10.1371/journal.pone.0165799

Open, closed, and hidden data

- The meta-analysis showed that interlukin-6 does not increase at night
- Data could often be estimated from published plots
- Data that could not be included increase the risk of bias and reduce precision of estimates of the diurnal curve
- Lack of individual participant data limits opportunities to model residual heterogeneity
- Closed and hidden data are a waste of resources and may be questioned on ethical grounds

Survival of research data in ecology

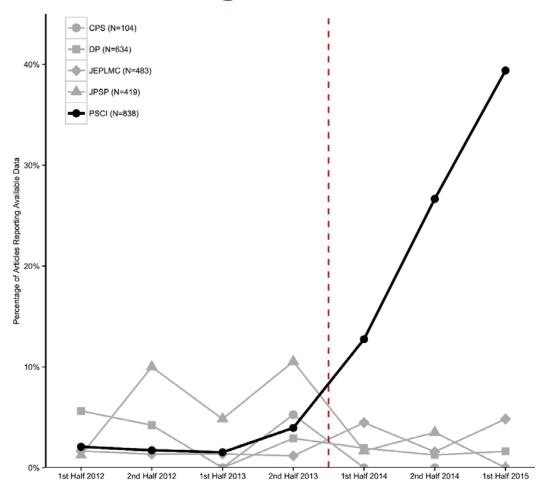


Vines et al. Current Biology 2014

Badges for open practices: an effective incentive?



Publication of open data in one journal after the introduction of badges



Kidwell et al. PLoS Biol 14(5): e1002456

Data life cycle: new concepts

- Data erosion: Functional loss of data reusability. May be due to e.g:
 - → Metadata not documented
 - → Loss of support for old file formats
 - → Loss of records
 - → Intentional destruction of records
- Loss of rawness is erosion where incomplete aggregate results are preserved but raw data are lost
- Salvage window: The time during which data can be saved with an additional effort

Scientific inference: new concepts

- A scientific field that generates data will generate increasing knowledge if the growth in available data outmatches the growth in bias risk due to unavailable data
- A field that generates increasing knowledge can be called a cumulating field
- A field that generates new data but not increasing knowledge can be called a dissipating field
- Research organizations and funders should make efforts to counteract dissipation by supporting:
 - → Publication of new data in appropriate repositories with structured formats and metadata
 - → Salvage of old data by curation and publication

Some quick thoughts on incentives

- The scientific paper is canonized as bearer of merit
 - → Prestige matters
 - → Contributions hard to disentangle
 - → Quality control operates mostly in a closed system and cannot be assessed
- Can we imagine a different system?

Conclusions

- Meta-science studies science itself
- Limited reproducibility suggests opportunities for improvement
 - → Open data
 - → Open code
 - → More rigorous scientific practices, e.g. preregistration, larger samples
- Research data need to be salvaged and preserved in a systematic manner
- Changes in the ecosystem of research give cause for optimism