

Thesis Seminar

The publication process, problems, and possible solutions

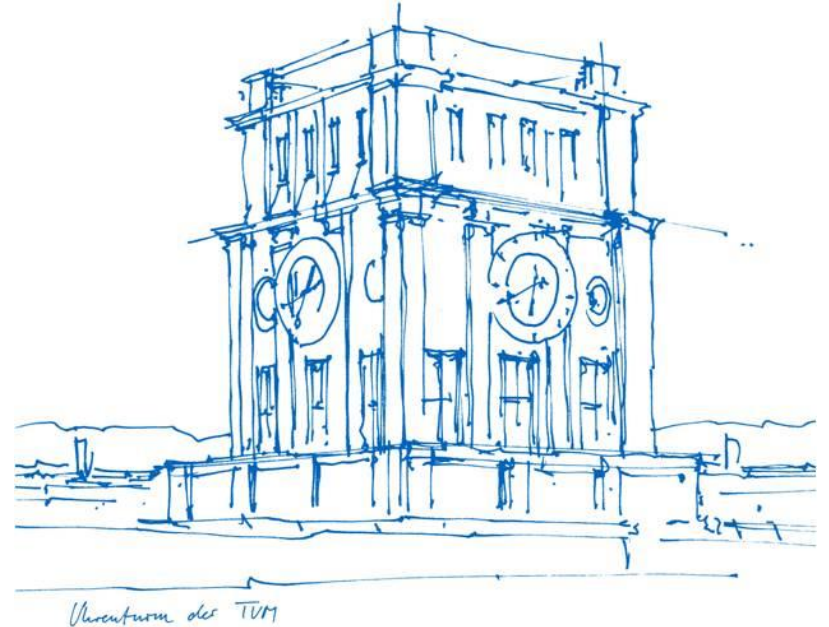
Dr. Theresa Treffers

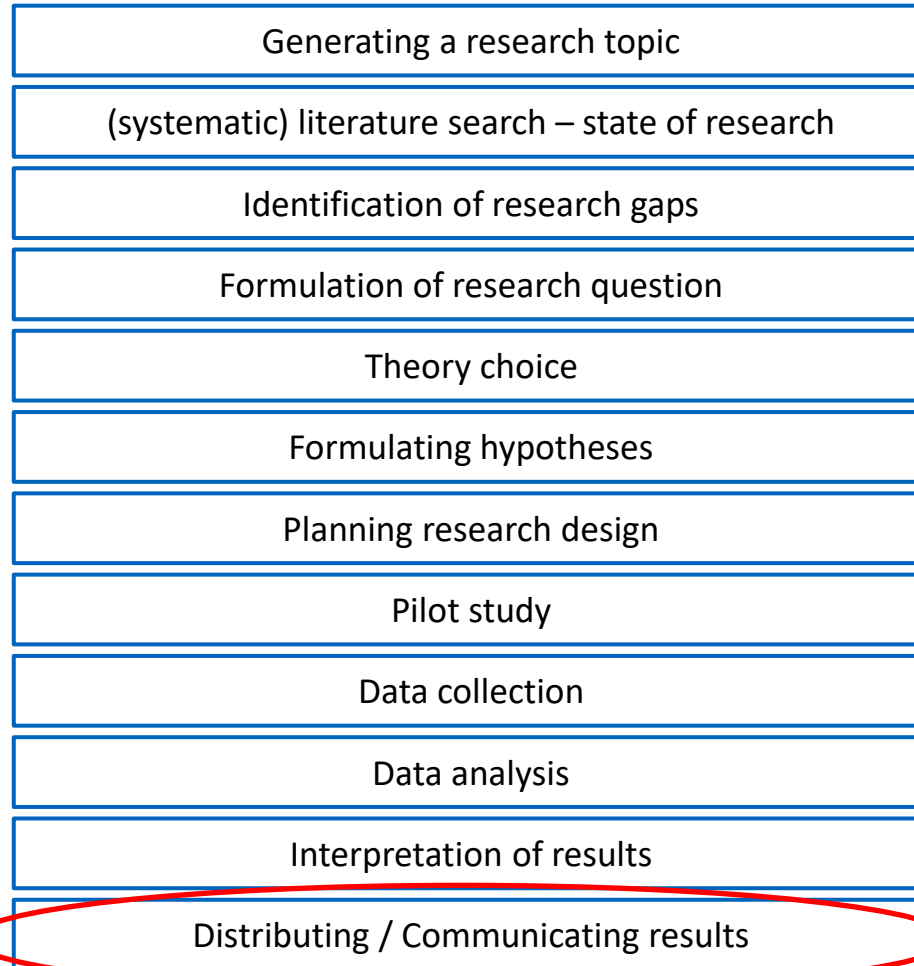
Technical University of Munich

TUM School of Management

Chair for Strategy and Organization

Prof. Dr. Isabell M. Welp



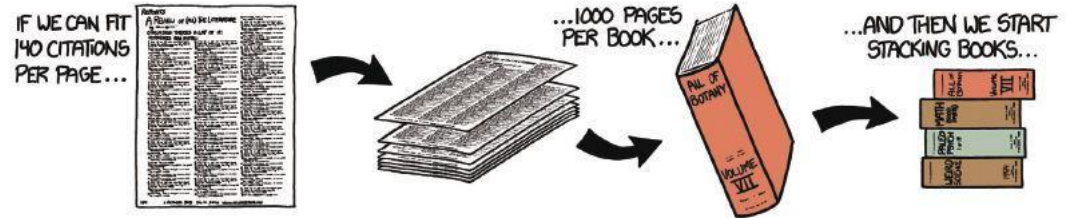


(double-blind) peer-review publication process

Writing for publications...

HOW MUCH SCIENCE IS THERE?

SCIENTIFIC PUBLISHING HAS BEEN ACCELERATING—A NEW PAPER IS NOW PUBLISHED ROUGHLY EVERY 20 SECONDS. LET'S IMAGINE A BIBLIOGRAPHY LISTING *EVERY* SCHOLARLY PAPER EVER WRITTEN. HOW LONG WOULD IT BE?



What is the (double blind) peer-review process?



The Peer-Review Process

- 1 Authors submit paper through journal system
- 2 Editor scans paper; either desk-reject or assign to action editor (AE)
- 3 Either editor or AE assign paper to 2-3 reviewers
- 4 Reviewers return reviews and recommendations (reviews usually include narrative (which authors and AE see), and private comments to AE plus numerical ratings (which AE, but not authors, see))
- 5 AE writes decision letter (reject, revise&resubmit, conditional accept, accept)
- 6 Decision letter sent to author and reviewers
- 7 In case of revision, revised paper sent through the same process above

Why do papers get rejected?



- ! **Unclear contribution**
- ! **Insufficient theory development**
- ! **Methodological issues**

Why do papers get rejected?

! Unclear contribution:

- no contribution (currently or potentially) **to the literature**
- no important **theoretical** (new ideas), **empirical** (new findings), and/or **practical** (new guidance) **contributions**
- authors don't discuss contributions **clearly and explicitly**
- authors don't provide **solid rationale** for their stated contributions
- authors can't improve their discussion of, and **justification** for, the stated contributions

Contribution:

- **State the literature you contribute to**
- **Cite the scholars on your table**
- **State exactly what is it you contribute to this table**
- **Clarify the relevance of your contributions**

Why do papers get rejected?

! Insufficient theory development:

- **research question** is **not meaningful**, important, and novel
- no meaningful, clear, and well-supported **theoretical framework or theory(ies)** guiding the study
- study doesn't **advance new knowledge** in area
- **constructs** aren't well-developed (defined, justified, explained)
- **other constructs should be included** (alternative explanations)
- **hypotheses aren't clear** and well-justified (based on theory, logic, and prior research)

Theory development:

- **State the theoretical lens that you apply**
- **Define constructs and develop hypotheses**
- **Acknowledge literature (what was done)**
- **Acknowledge limitations (what should be done)**

Why do papers get rejected?

! **Methodological issues:**

- Methods are not often the sole reason for rejection, but they certainly could be a major reason for rejection
- **Sampling and study design** (does elicitation of phenomena allow for proper test of research questions and hypotheses?)
- **Measurement issues** (reliability, validity, levels, adequacy of sources)
- **Analytical issues** (do the analyses yield valid inferences vis-à-vis hypotheses? can alternative analyses be better?)

Method development:

- **provide more evidence in support of validity**
- **test alternatives (new or additional data and/or analyses)**
- **acknowledge limitations**

Practices that harm science (1/4)

#	Management problem	Source / Paper	Especially relevant paper section
1	HARKing: Overview	Murphy, K. R., & Aguinis, H. (2019). HARKing: how badly can cherry-picking and question trolling produce bias in published results?. <i>Journal of business and psychology</i> , 34(1), 1-17.	Abstract + Introduction (not entitled, but directly following the abstract)
2	Types of HARKing: Cherry picking	Murphy, K. R., & Aguinis, H. (2019). HARKing: how badly can cherry-picking and question trolling produce bias in published results?. <i>Journal of business and psychology</i> , 34(1), 1-17.	Different forms of harking, Prevalence of Cherry-Picking and Question Trolling
3	Types of HARKing: Question trolling	Murphy, K. R., & Aguinis, H. (2019). HARKing: how badly can cherry-picking and question trolling produce bias in published results?. <i>Journal of business and psychology</i> , 34(1), 1-17.	Different forms of harking, Prevalence of Cherry-Picking and Question Trolling
4	P-Hacking	Aguinis, H., Cascio, W. F., & Ramani, R. S. (2017). Science's reproducibility and replicability crisis: International business is not immune. <i>Journal of International Business Studies</i> , 48, 653–663.	Reporting of p values

Practices that harm science (2/4)

#	Management problem	Source / Paper	Especially relevant paper section
5	The Chrysalis Effect	O'Boyle Jr, E. H., Banks, G. C., & Gonzalez-Mulé, E. (2017). The chrysalis effect: How ugly initial results metamorphosize into beautiful articles. <i>Journal of Management</i> , 43(2), 376-399.	pp. 376-378
6	Publication bias / File drawer problem	Dalton, D. R., Aguinis, H., Dalton, C. A., Bosco, F. A., & Pierce, C. A. 2012. Revisiting the file drawer problem in meta-analysis: An empirical assessment of published and non-published correlation matrices. <i>Personnel Psychology</i> , 65: 221-249.	File Drawer Problem
7	Reproducibility vs. replicability and the replication crisis	Aguinis, H., Cascio, W. F., & Ramani, R. S. (2017). Science's reproducibility and replicability crisis: International business is not immune. <i>Journal of International Business Studies</i> , 48, 653–663.	Introduction
8	Lack of transparency	Aguinis, H., & Solarino, A. M. (2019). Transparency and replicability in qualitative research: The case of interviews with elite informants. <i>Strategic Management Journal</i> , 40(8), 1291-1315.	3.1 Transparency criteria in qualitative research

Practices that harm science (3/4)

#	Management problem	Source / Paper	Especially relevant paper section
9	Quantity over quality = Publish or perish	Wright, P. M. (2016). Ensuring research integrity: An editor's perspective. <i>Journal of Management</i> , 42(5), 1037 –1043.	The Publishing Pressure Cooker (Introduction + Publish or perish)
10	Plagiarism	Clarke, R. (2006). Plagiarism by academics: More complex than it seems. <i>Journal of the Association for Information Systems</i> , 7(2), 91-121.	Definition, Plagiarism by Academics
11	‚Coercive citation‘	Wilhite, A. W., & Fong, E. A. (2012). Coercive citation in academic publishing. <i>Science</i> , 335(6068), 542-543.	complete
12	Crisis of confidence	Harley, B. (2019). Confronting the crisis of confidence in management studies: Why senior scholars need to stop setting a bad example. <i>Academy of Management Learning & Education</i> , 18(2), 286-297.	p. 286-291

Practices that harm science (4/4)

#	Management problem	Source / Paper	Especially relevant paper section
13	Arigorium	Antonakis, J. (2017). On doing better science: From thrill of discovery to policy implications. <i>The Leadership Quarterly</i> , 28(1), 5-21.	Introduction (p. 1-7) + Disease 4: Arigorium
14	WEIRD sample	Henrich, J., Heine, S. J., & Norenzayan, A. (2010). Most people are not WEIRD. <i>Nature</i> , 466(7302), 29-29.	complete
15	Endogeneity & Causality	Antonakis, J., Bendahan, S., Jacquart, P., & Lalive, R. (2010). On making causal claims: A review and recommendations. <i>The Leadership Quarterly</i> , 21(6), 1086-1120.	Introduction (p. 1-8)
16	Common-method variance	Podsakoff, N. P. (2003). Common method biases in behavioral research: a critical review of the literature and recommended remedies. <i>Journal of Applied Psychology</i> , 88(879), 10-1037.	p.879-886
17	Data fabrication or falsification	Fanelli, D. (2009). How many scientists fabricate and falsify research? A systematic review and meta-analysis of survey data. <i>PloS one</i> , 4(5), e5738.	Introduction + Discussion

Proscribed practices (not directly harming science) (1/3)



#	Management problem	Source / Paper	Especially relevant paper section
18	Theorrhea	Antonakis, J. (2017). On doing better science: From thrill of discovery to policy implications. <i>The Leadership Quarterly</i> , 28(1), 5-21.	Introduction (p- 1-7) + Disease 3: Significosis
19	Neophilia	Antonakis, J. (2017). On doing better science: From thrill of discovery to policy implications. <i>The Leadership Quarterly</i> , 28(1), 5-21.	Introduction (p- 1-7) + Disease 2: Theorrhea
20	Significosis	Antonakis, J. (2017). On doing better science: From thrill of discovery to policy implications. <i>The Leadership Quarterly</i> , 28(1), 5-21.	Introduction (p- 1-7) + Disease 1: Significosis
21	Disjunctivitis	Antonakis, J. (2017). On doing better science: From thrill of discovery to policy implications. <i>The Leadership Quarterly</i> , 28(1), 5-21.	Introduction (p- 1-7) + Disease 5: Disjunctivitis
22	Construct Proliferation	Shaffer, J. A., DeGeest, D., & Li, A. (2016). Tackling the problem of construct proliferation: A guide to assessing the discriminant validity of conceptually related constructs. <i>Organizational Research Methods</i> , 19(1), 80-110.	p. 80-83

Proscribed practices (not directly harming science) (2/3)

#	Management problem	Source / Paper	Especially relevant paper section
23	Interpretation	Bliese, P. D., & Wang, M. (2019). Results Provide Information About Cumulative Probabilities of Finding Significance: Let's Report This Information. <i>Journal of Management</i> (forthcoming).	Abstract + Introduction (not entitled, but directly following the abstract)
24	Evaluitis	Leeuw, Frans L. "Evaluation: a booming business but is it adding value?." <i>Evaluation Journal of Australasia</i> , 9(1), 3-9.	The problem
25	How the review process damages our writing	Tourish, D. (2019). The triumph of nonsense in management studies. <i>Academy of Management Learning & Education</i> .	Paragraph on "How the review process damages our writing" (p. 13-19)
26	Fisher t-tests and it's problems	Fay, M. P. (2010). Confidence intervals that match Fisher's exact or Blaker's exact tests. <i>Biostatistics</i> , 11(2), 373-374. Rice, K. (2010) A decision-theoretic formulation of Fisher's approach to testing. <i>Am. Statistn</i> , 64, 345–349.	Further readings for self-study

Proscribed practices (not directly harming science) (3/3)

#	Management problem	Source / Paper	Especially relevant paper section
27	Self-plagiarism	Clarke, R. (2009). Journal self-citation XIX: Self-plagiarism and self-citation-A practical guide based on underlying principles. <i>Communications of the Association for Information Systems</i> , 25(1), 19.	III. SELF-PLAGIARISM
28	Covert Research Practices	Roulet, T. J., Gill, M. J., Stenger, S., & Gill, D. J. (2017). Reconsidering the value of covert research: The role of ambiguous consent in participant observation. <i>Organizational Research Methods</i> , 20(3), 487-517.	Abstract + Introduction (not entitled, but directly following the abstract)
29	Data Slicing	Pfleegor, A. G., Katz, M., & Bowers, M. T. (2019). Publish, perish, or salami slice? Authorship ethics in an emerging field. <i>Journal of Business Ethics</i> , 156(1), 189-208.	Responsible and Ethical Authorship + Ethically Questionable Authorship Practices in Sport Scholarship
30	Multiple use of the same data set	Colquitt, J. A. (2013). Data overlap policies at AMJ. <i>Academy of Management Journal</i> , 56(2), 331-333.	complete

Unfortunately, the problem is much bigger...

Current challenges in management research

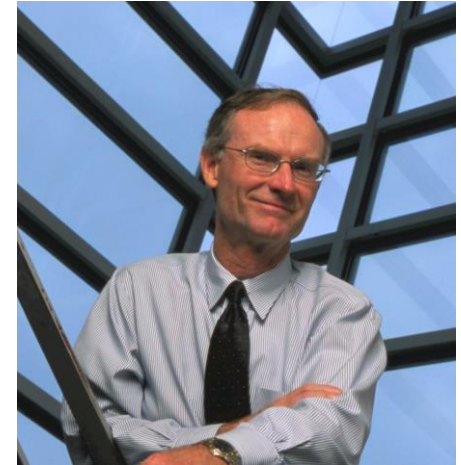
1993 Presidential Address:

WHAT IF THE ACADEMY ACTUALLY MATTERED?

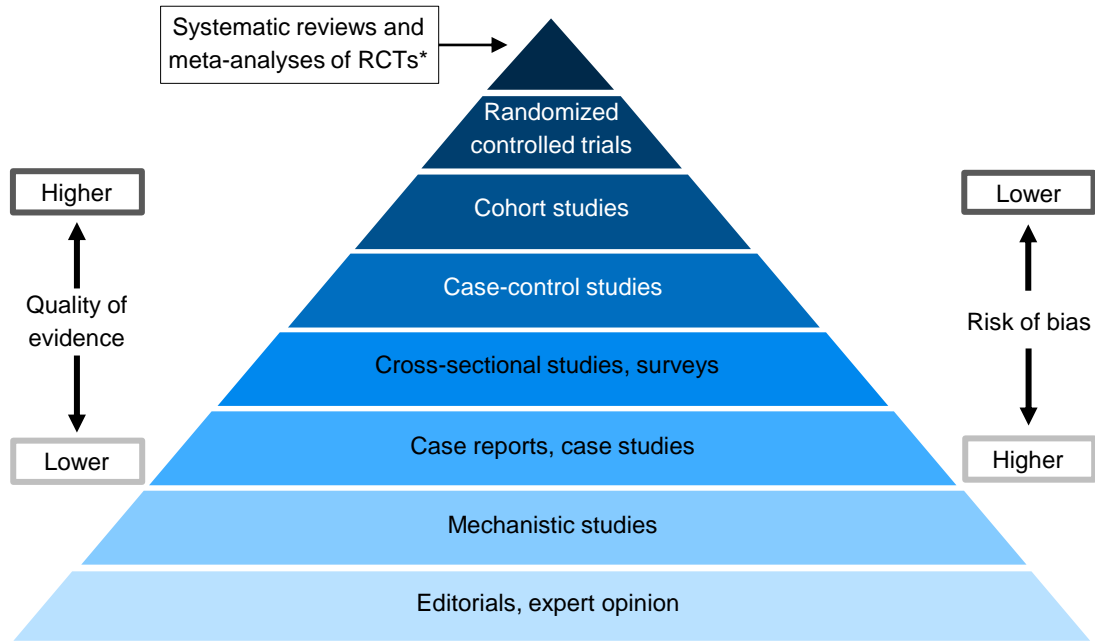
DONALD C. HAMBRICK
Columbia University

mal innovation, minimal visibility, minimal impact.¹ Each August, we come to talk with each other; during the rest of the year we read each others' papers in our journals and write our own papers so that we may, in turn, have an audience the following August: an incestuous, closed loop.

Colleagues, if we believe highly in what we do, if we believe in the significance of advanced thinking and research on management, then it is time we showed it. We must recognize that our responsibility is not to ourselves, but rather to the institutions around the world that are in dire need of improved management, as well as to those individuals who seek to be the most effective managers they possibly can be. It is time for us to break out of our closed loop. It is time for us to matter.



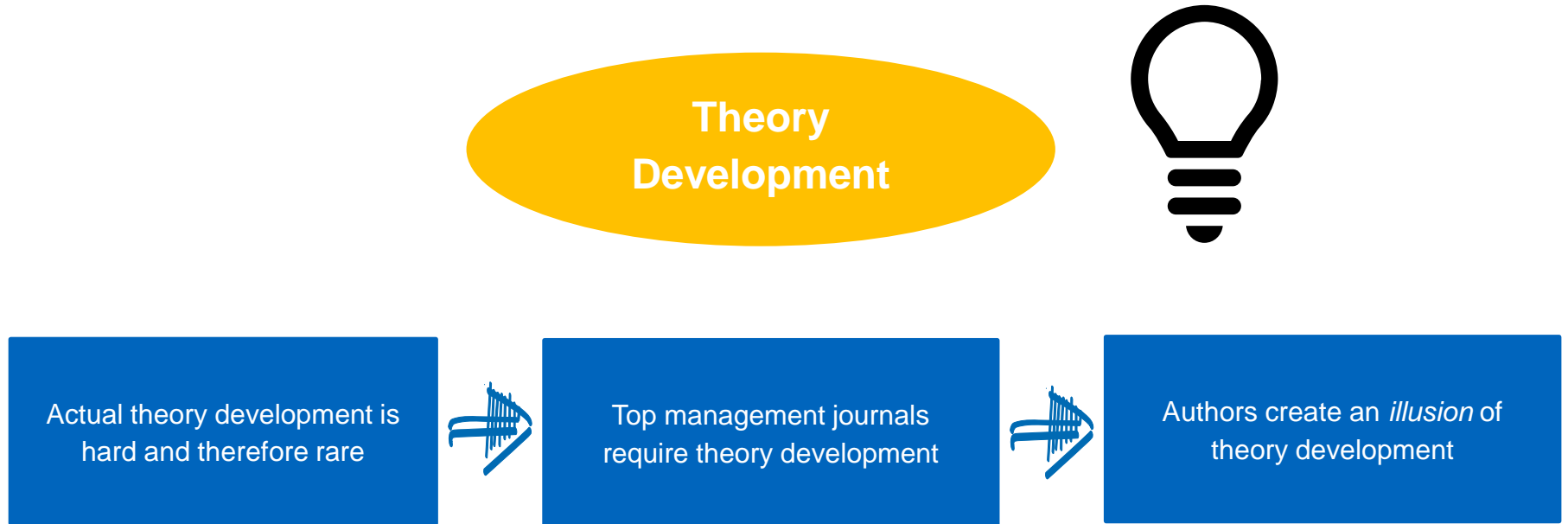
The hierarchy of evidence pyramid



RCT: randomized controlled trial

- At each **ascending** level, the **quality** of evidence is likely to **improve** (i.e., the risk of bias decreases) and the **quantity** of available studies usually **declines**
- Confidence in **causal relationships** generally **increases** toward the **top of the pyramid**
- However, within each level, the **quality varies** based on study **design** and **implementation**

How to get published in management journals



The problem of peer review

I must find something wrong in order for my review to count



Reviewer

$p < .05$

tautologous hypothesis and nonsense results



Convolved language

author may feel that work is not his own anymore



Continuous revision

Spending much time for very little added value



I do whatever I can to please the reviewer in order to get published



Author

The game of publishing

Arbitrariness

Peters & Ceci, 1982: resubmitted 12 psychology papers to journals who already published them, 16/18 reviewers recommended rejection

Big words

Lonely employees dislike workplaces where they feel lonely

„Employees who experience higher levels of workplace loneliness will be less affectively committed to their organization“ (Ozcelik & Barsade, 2018)

**Researchers become
genuine imposters**

What is to be done



Publishing should not be seen as a game



Management journals should rethink their position towards the need for theory development



Strict structural form should be abandoned – authors should be allowed to show they are human



Regain a sense of proportionate effort – Firm decision whether to publish a paper after two revision rounds

Scientific Misconduct in Empirical Research



**After 16 retractions,
management professor
Lichtenthaler resigns post**

**Cornell finds that food
marketing researcher Brian
Wansink committed
misconduct, as he announces
retirement**

Equally disturbing, some 60% of respondents reported knowledge of faculty who have “dropped observations or data points from analyses based on a gut feeling that they were inaccurate,” what many would consider an example of “data trimming.”

- **Scientific misconduct**, data falsification and honest mistakes are **quite common in empirical management research**
- **No objective and independent tests** broadly used to verify empirical findings
- **Authors identify** and demonstrate **three such objective tests** from peer-reviewed literature

Recent headlines to science, research and universities

„Scientific research has
changed the world.

Now it needs to change
itself.“



Recent headlines to science, research and universities

theguardian

Google™ Custom Search

Nobel winner declares boycott of top science journals

Randy Schekman says his lab will no longer send papers to Nature, Cell and Science as they distort scientific process

Ian Sample, science correspondent
The Guardian, Monday 9 December 2013 19.42 GMT



Randy Schekman, centre, at a Nobel prize ceremony in Stockholm. Photograph: Rob Schoenbaum/Zuma

Current worries of nobel prize winners

*„Today, he said, **he would not get an academic job.** It's as simple as that. I don't think I would be regarded as productive enough.“*



Peter Higgs
Physics nobel prize winner 2013

*Just as Wall Street needs to break the hold of bonus culture, so **science must break the tyranny of the luxury journals.**"*



Randy Schekman
Medicine nobel prize winner 2013

*"Daniel Kahneman Sees 'Train-Wreck Looming' for Social Psychology": "I **believe that you should collectively do something about this mess.**"*



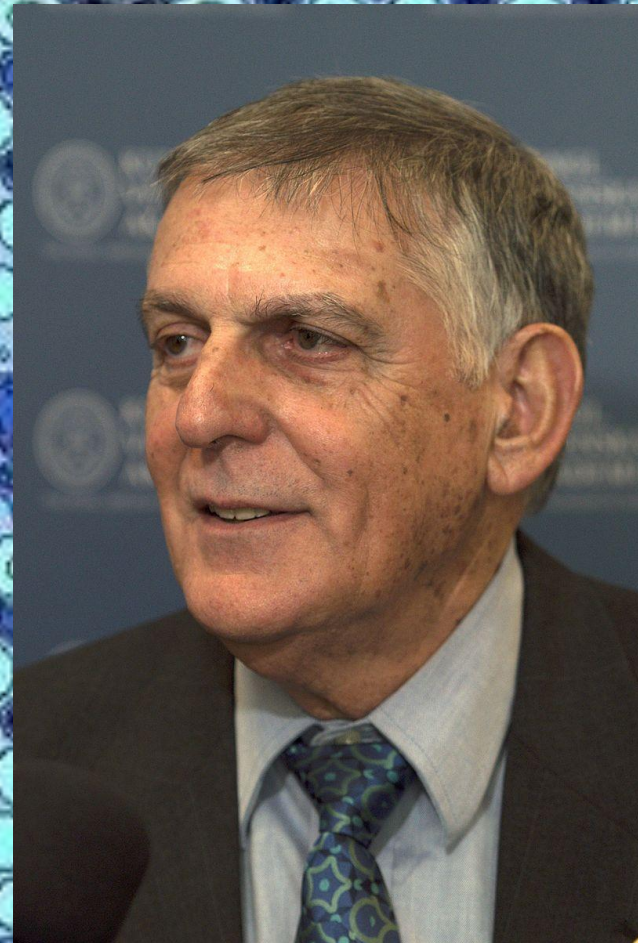
Daniel Kahneman
Economy nobel prize winner 2002

Dan Shechtman

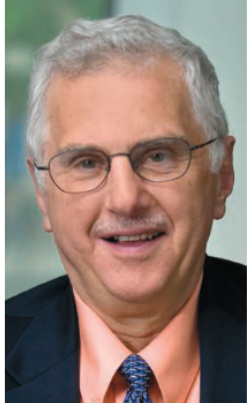
**Nobel prize winner chemistry
2011 for the finding of
quasicrystals**

„There are no quasicrystals, only
quasi-scientists“ ...

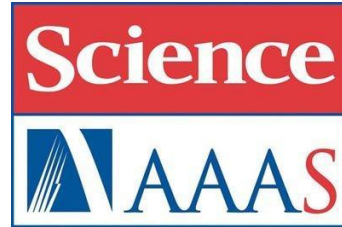
... said the chemistry Nobel prize winner
Linus Carl Pauling (passed in 1994).



Current headlines to science, research and universities



Bruce Alberts is Editor-in-Chief of *Science*.



Impact Factor Distortions

THIS EDITORIAL COINCIDES WITH THE RELEASE OF THE *SAN FRANCISCO DECLARATION ON RESEARCH Assessment (DORA)*, the outcome of a gathering of concerned scientists at the December 2012 meeting of the American Society for Cell Biology.* To correct distortions in the evaluation of scientific research, DORA aims to stop the use of the “journal impact factor” in judging an individual scientist’s work. The *Declaration* states that the impact factor must not be used as “a surrogate measure of the quality of individual research articles, to assess an individual scientist’s contributions, or in hiring, promotion, or funding decisions.” DORA also provides a list of specific actions, targeted at improving the way scientific publications are assessed, to be taken by funding agencies, institutions, publishers, researchers, and the organizations that supply

Current headlines to science, research and universities

San Francisco Declaration on Research Assessment

Putting science into the assessment of research

There is a pressing need to improve the ways in which the output of scientific research is evaluated by funding agencies, academic institutions, and other parties.

To address this issue, a group of editors and publishers of scholarly journals met during the Annual Meeting of The American Society for Cell Biology (ASCB) in San Francisco, CA, on December 16, 2012. The group developed a set of recommendations, referred to as the *San Francisco Declaration on Research Assessment*. We invite interested parties across all scientific disciplines to indicate their support by adding their names to this Declaration.

Current headlines to science, research and universities

How should medical science change?

In December, 2013, Randy Schekman received a Nobel Prize in Physiology or Medicine for his codiscovery (with James Rothman and Thomas Südhof) of the cellular machinery regulating vesicle traffic. He used the occasion to launch a ferocious attack against what he called “luxury journals”—*Nature*, *Science*, and *Cell*. Although he didn’t mention *The Lancet*, *JAMA*, or *The New England Journal of Medicine*, it probably isn’t unreasonable to think he would include us in his definition of “luxury journal”. This is what he wrote in *The Guardian*: “These luxury journals are supposed

production and reporting of research evidence” which made the extraordinary claim that as much as 85% of research investment was wasted.³ It seemed an unbelievable figure. But the useful discussion their paper triggered led to a spate of seminars and meetings to explore what could be done about what all agreed was a wholly unsatisfactory situation—wrong questions being asked by scientists, poor study designs being applied, research that was inaccessible, and findings that were distorted by selective reporting and other types of bias. The Series we now publish addresses these issues in far



Published Online
January 8, 2014
[http://dx.doi.org/10.1016/S0140-6736\(13\)62678-1](http://dx.doi.org/10.1016/S0140-6736(13)62678-1)

See Online/Series
[http://dx.doi.org/10.1016/S0140-6736\(13\)62229-1](http://dx.doi.org/10.1016/S0140-6736(13)62229-1),
[http://dx.doi.org/10.1016/S0140-6736\(13\)62227-8](http://dx.doi.org/10.1016/S0140-6736(13)62227-8),
[http://dx.doi.org/10.1016/S0140-6736\(13\)62297-7](http://dx.doi.org/10.1016/S0140-6736(13)62297-7),
[http://dx.doi.org/10.1016/S0140-6736\(13\)62296-5](http://dx.doi.org/10.1016/S0140-6736(13)62296-5), and
[http://dx.doi.org/10.1016/S0140-6736\(13\)62228-X](http://dx.doi.org/10.1016/S0140-6736(13)62228-X)

THE LANCET

„How should the entire scientific enterprise change to produce reliable and accessible evidence that addresses the challenges faced by society and the individuals who make up their societies“

(The Lancet, 2014, January 8).

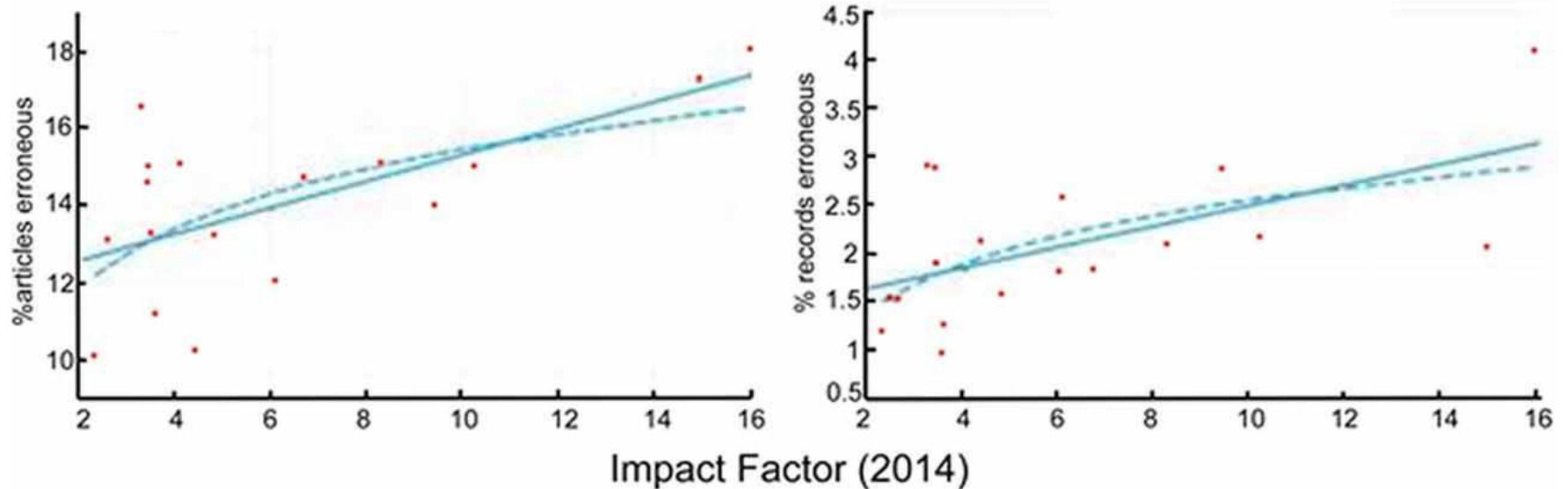


Prestigious Science Journals Struggle to Reach Even Average Reliability

In which journal a scientist publishes is considered one of the most crucial factors determining their career. The underlying common assumption is that only the best scientists manage to publish in a highly selective tier of the most prestigious journals. However, data from several lines of evidence suggest that the methodological quality of scientific experiments does not increase with increasing rank of the journal. **On the contrary, an accumulating body of evidence suggests the inverse: methodological quality and, consequently, reliability of published research works in several fields may be decreasing with increasing journal rank.** The data supporting these conclusions circumvent confounding factors such as increased readership and scrutiny for these journals, focusing instead on quantifiable indicators of methodological soundness in the published literature, relying on, in part, semi-automated data extraction from often thousands of publications at a time. With the accumulating evidence over the last decade grew the realization that the very existence of scholarly journals, due to their inherent hierarchy, constitutes one of the major threats to publicly funded science: hiring, promoting and funding scientists who publish unreliable science eventually erodes public trust in science.

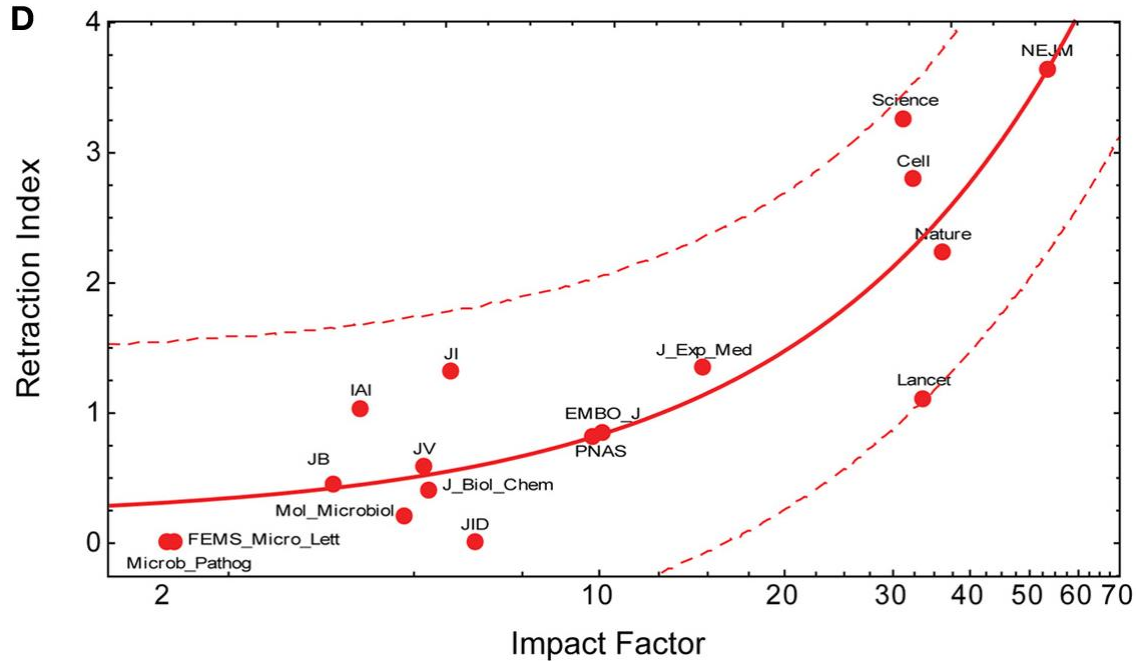
In the literature covering unretracted, peer-reviewed articles, one can identify at least eight lines of evidence suggesting that articles published in **higher ranking journals are methodologically either not stronger or, indeed, weaker than those in lower ranking journals.** In contrast, there is no evidence that articles published in higher ranking journals are methodologically stronger. Methodology here refers to several measures of experimental and statistical rigor with a potential bearing on subsequent replication or re-use. There is currently one article with evidence that higher ranking journals are better at detecting duplicated images (Bik et al., 2016).

The higher the impact factor of the journal the less reliable the data, the less methodologically good the experiments, the more fraud



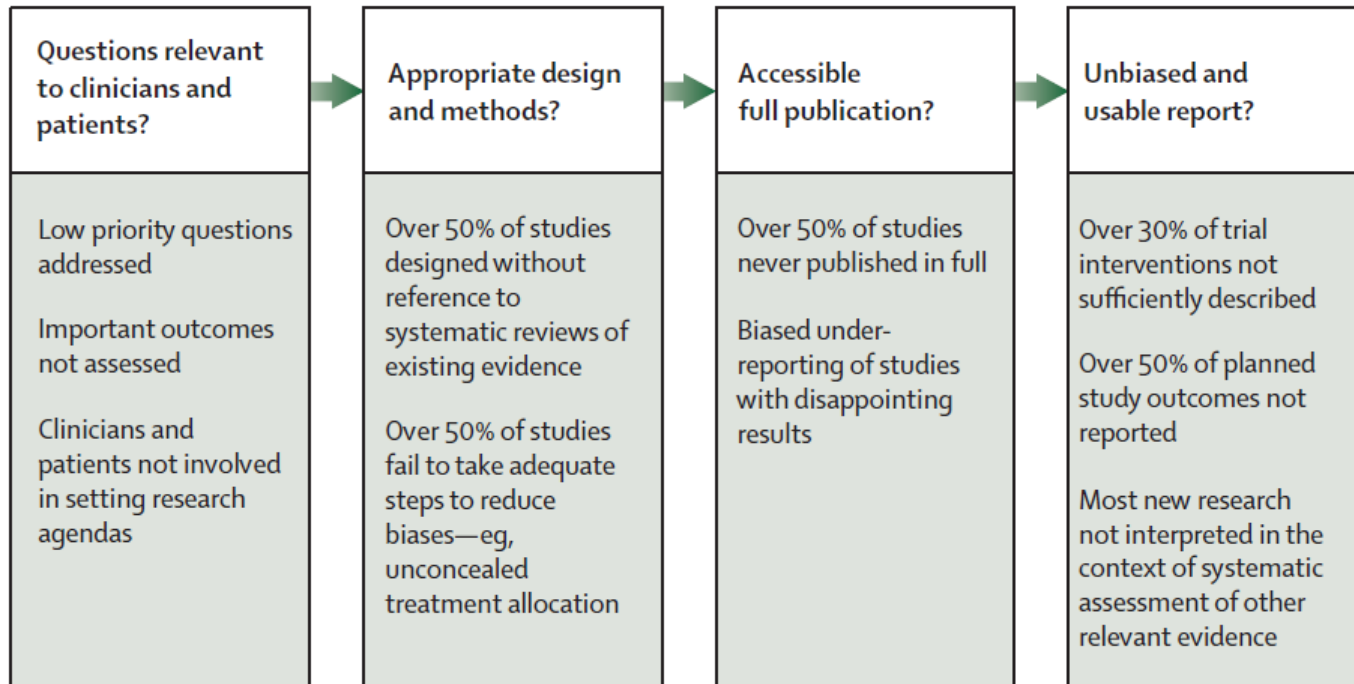
Impact factors and rankings: the problem of their objectivity and resulting consequences

Is the impact factor reliable enough as a quality indicator for research?



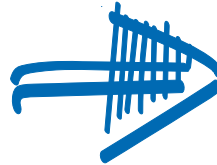
Positive correlation between retraction rate and impact factor (Brembs et al., 2013).

Criticism on all stages of the research value chain



Data basis: Pros and cons of obligatory specifications of objective measurement results and quality criteria of measurement instruments

„Publish or perish“



„Significance chasing“

„HARKing“

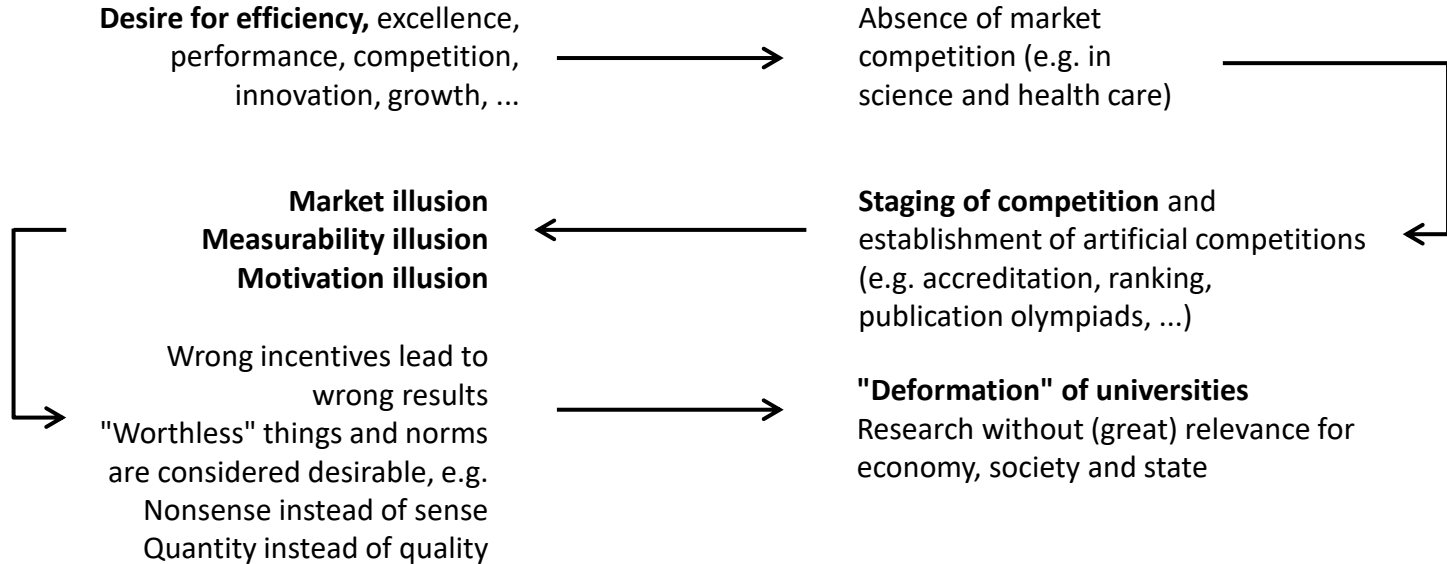
Improper methodology
and evaluation

“publish or perish”-principle results in the publication of more and more nonsense.” (Binswanger, 2015, p.19)

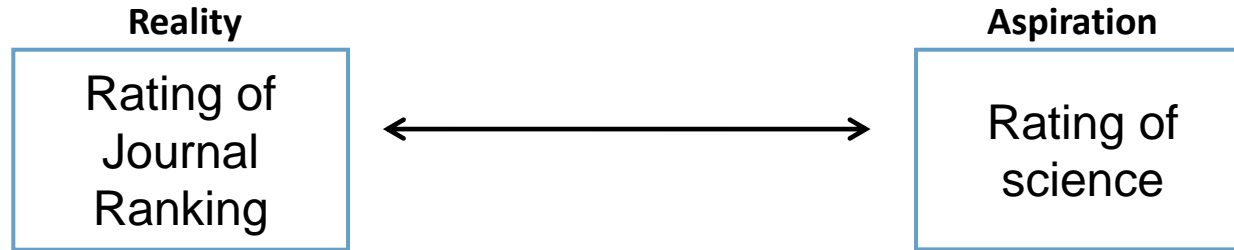
“It can be proven that most claimed research findings are false.”

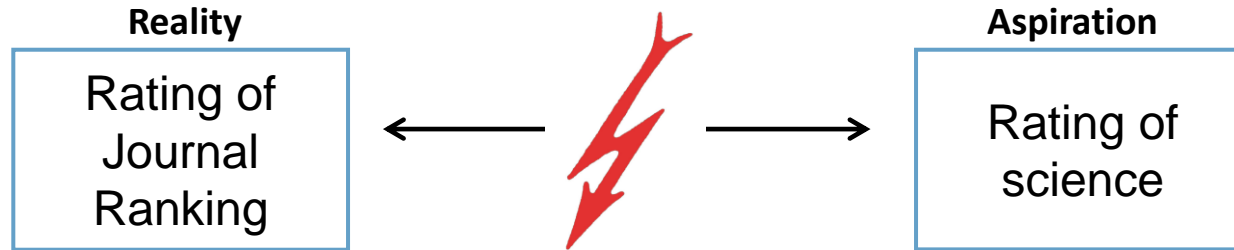
(Ioannidis, 2005, S. 0696)

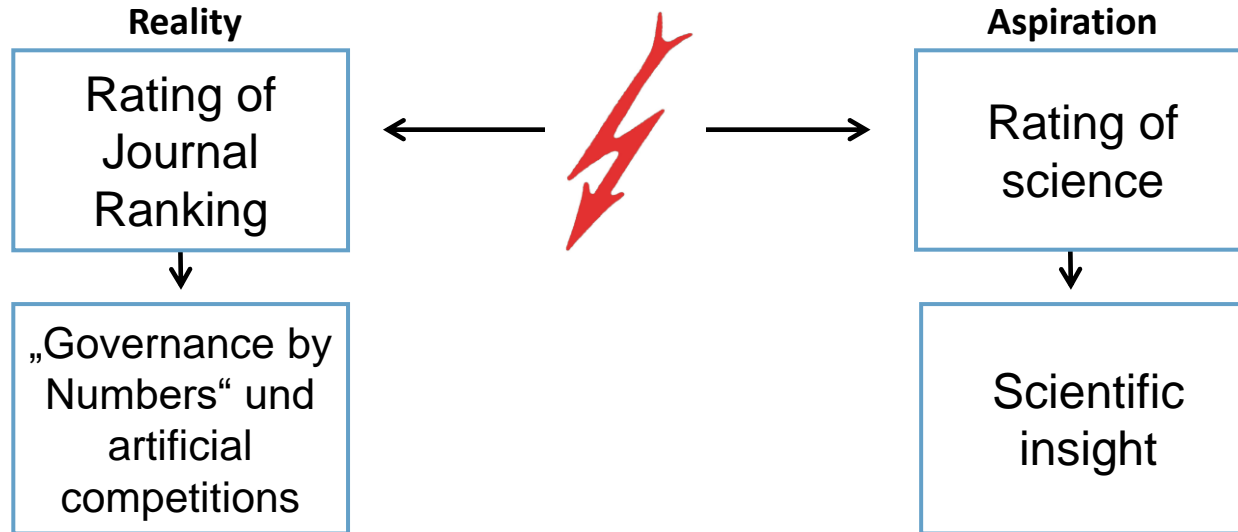
How artificial competitions harm science

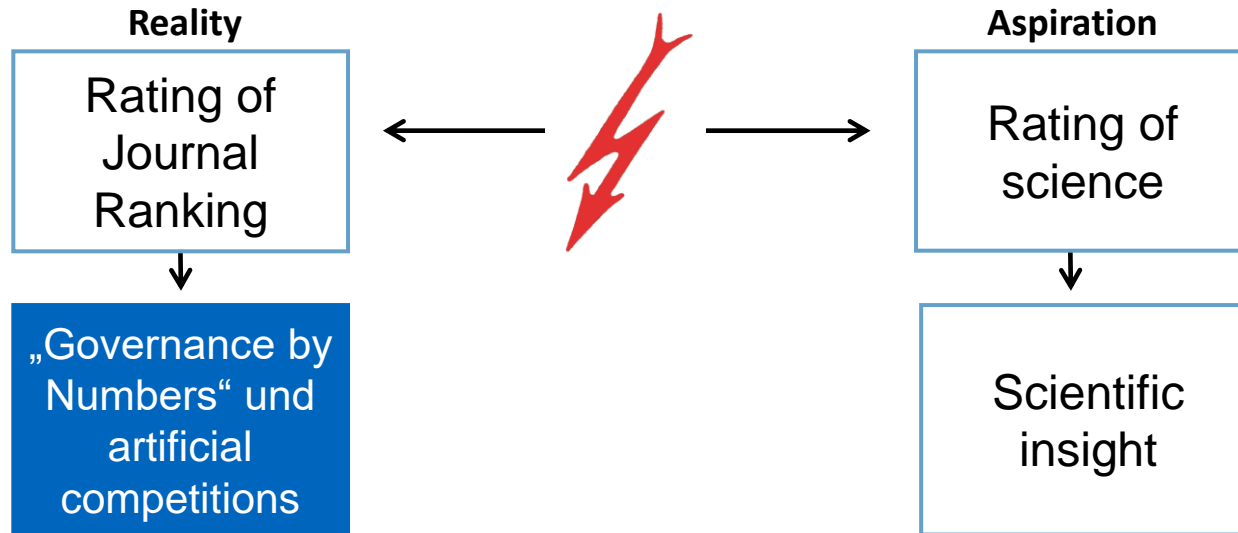


“publish or perish”-principle results in the publication of more and more nonsense.” (Binswanger, 2015, p.19)





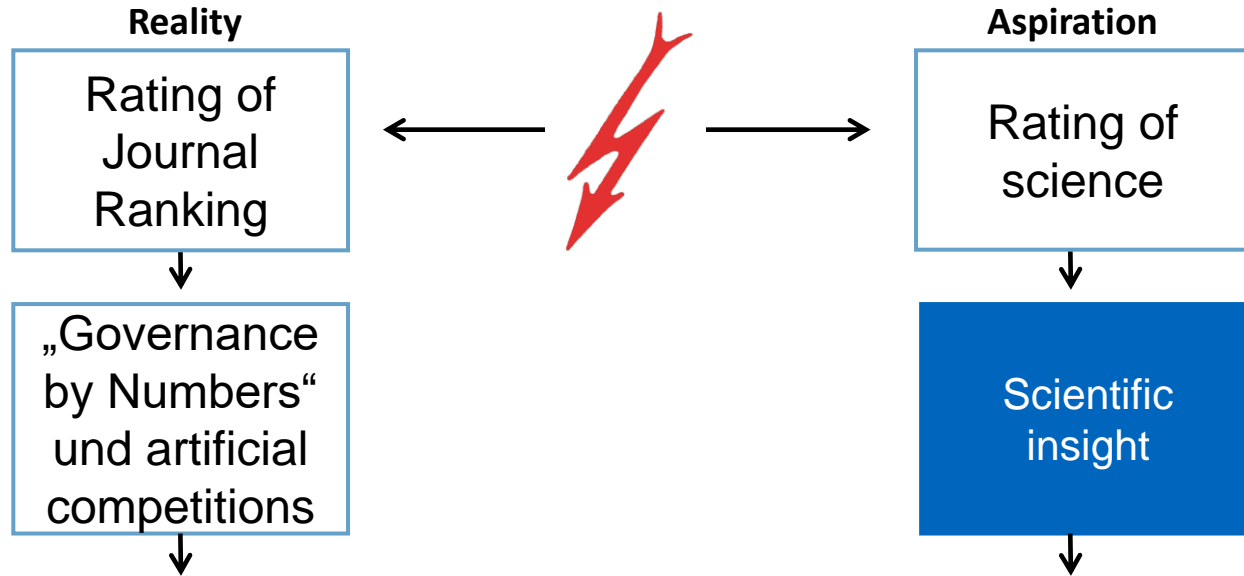




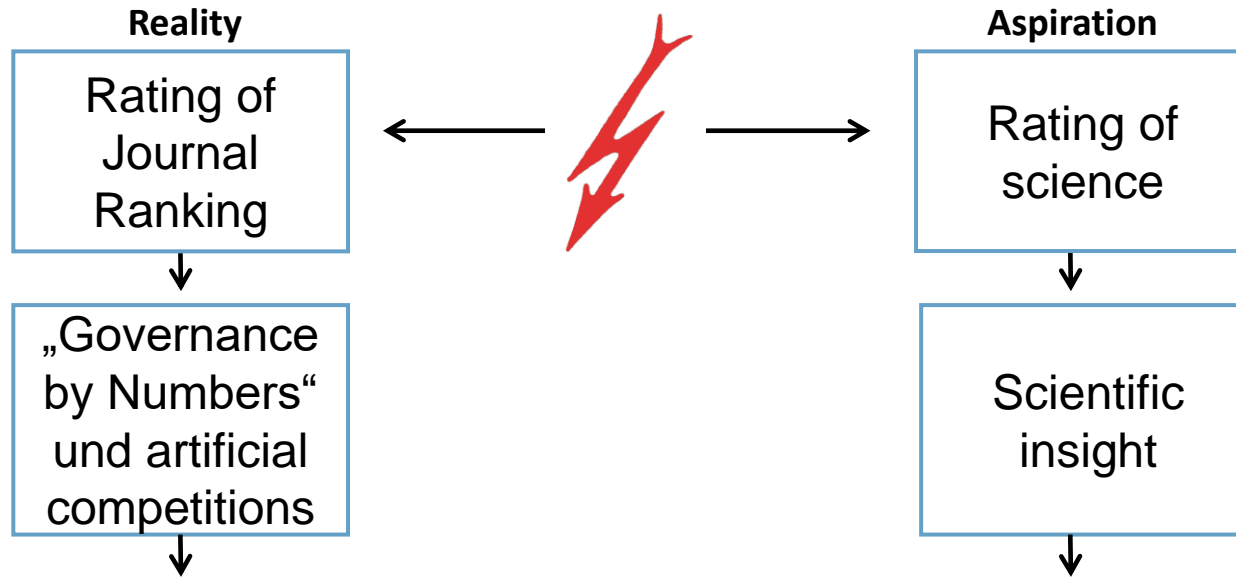
$$\textit{Performance} = \frac{\textit{Third party funds}}{\textit{Researchers involved}}$$

$$\textit{Performance} = \frac{\textit{Number of publications}}{\textit{Resources deployed}}$$

$$\textit{Performance} = \frac{\textit{Number of citations}}{\textit{Resources deployed}}$$



$$\textit{Performance} = \frac{\textit{Scientific Insight}}{\textit{Resources deployed}}$$



Problems of science

- Research without relevance and risk
 - Errors and lack of replicability
 - Slowness
 - Lack of transparency and access to data
 - Black box of peer review
 - Non-intended effects/manipulability
 - One-dimensionality of measurement
-



"What you are suggesting would mean that we would have to start (sic!) reading the applicants' publications again!"

Member of an appointment committee

"As a scientist, I agree with all of your criticisms of counting, measuring, and weighing science - but: **politically, success in rankings, artificial competitions, and publication rankings that we win are of use to us, even if they measure the wrong thing and lead us to do irrelevant things.**"



President of a university of excellence

TOP 100

Handelsblatt VWL-Ranking 2015 - Beste Forschungsleistung

Die Handelsblatt-Studie orientiert sich an international gängigen Standards zur Evaluierung wirtschaftswissenschaftlicher Forschung. Betrachtet werden Publikationen in Fachzeitschrift Konjunkturforschungsstelle der ETH Zürich erstellt die Rangliste im Auftrag des Handelsblatts

Rang 2015	Rang 2013	Vorname	Nachname	Universität	Alter	Fach
1	3	Peter	Egger	Zürich ETH	45	Internationale Ökonomie
2	12	Holger	Struik	Göttingen Uni	50	Theoretische Makroökonomie
3	5	Hans	Gersbach	Zürich ETH	55	theoretische Makroökonomie; Politische Ökonomie
4	8	Patrick W.	Schmätz	Köln Uni	46	theoretische Makroökonomie; angewandte Mikroökonomie
5	1	Roman	Inderst	Frankfurt / Main Uni	45	Industrieökonomie; Bankbetriebslehre & Finanzierung
6	2	Matthias	Suter	Köln Uni	46	Experimentelle Wirtschaftsforschung
7	10	Ernst	Fehr	Zürich Uni	58	Experimentelle Wirtschaftsforschung
8	11	Ulrich	Müller	Princeton University	49	Ökonometrie
9	4	Lutz	Kilian	University of Michigan	51	Ökonometrie; angewandte Makroökonomie
10	36	Joachim	Wagner	Lüneburg Leuphana Uni	60	angewandte Mikroökonomie
11	49	Ansgar	Belke	Duisburg-Essen Uni	50	Internationale Ökonomie; Geldtheorie und -politik
12	32	Jan-Christoph	Rülke	ESS Uni	37	Internationale Ökonomie; Geldtheorie und -politik; Ausenwirtschaftstheorie und -politik
13		Martin	Huber	Fribourg Uni	35	Ökonometrie; angewandte Mikroökonomie
14	13	Armin	Falk	Bonn Uni	47	Experimentelle Wirtschaftsforschung
15	6	Marcel	Fratzscher	Berlin HU	44	Internationale Ökonomie; angewandte Makroökonomie
16	7	Axel	Dreher	Heidelberg Uni	42	Internationale Ökonomie
17	17	Niklas	Potrafke	München LMU	35	Politische Ökonomie
18	83	Yves	Bretmoser	Berlin HU	38	Verhaltensökonomie; theoretische Mikroökonomie
19	14	Johannes	Hörner	Yale University	42	theoretische Mikroökonomie; Spieltheorie
20	21	Thomas	Eichner	Hagen FernUni	45	Finanzwissenschaft
21	65	Mario	Larch	Bayreuth Uni	39	Ökonometrie; Internationale Ökonomie
22	18	Kai A.	Konrad	München LMU	54	Finanzwissenschaft
22	74	Björn	Bartling	Zürich Uni	41	Verhaltensökonomie; Experimentelle Wirtschaftsforschung; angewandte Mikroökonomie
24		Peter	Nunnenkamp	itw Institut für Weltwirtschaft Kiel	-	-
25	33	Tim	Friehe	Marburg Uni	37	Rechtsökonomie; angewandte Mikroökonomie

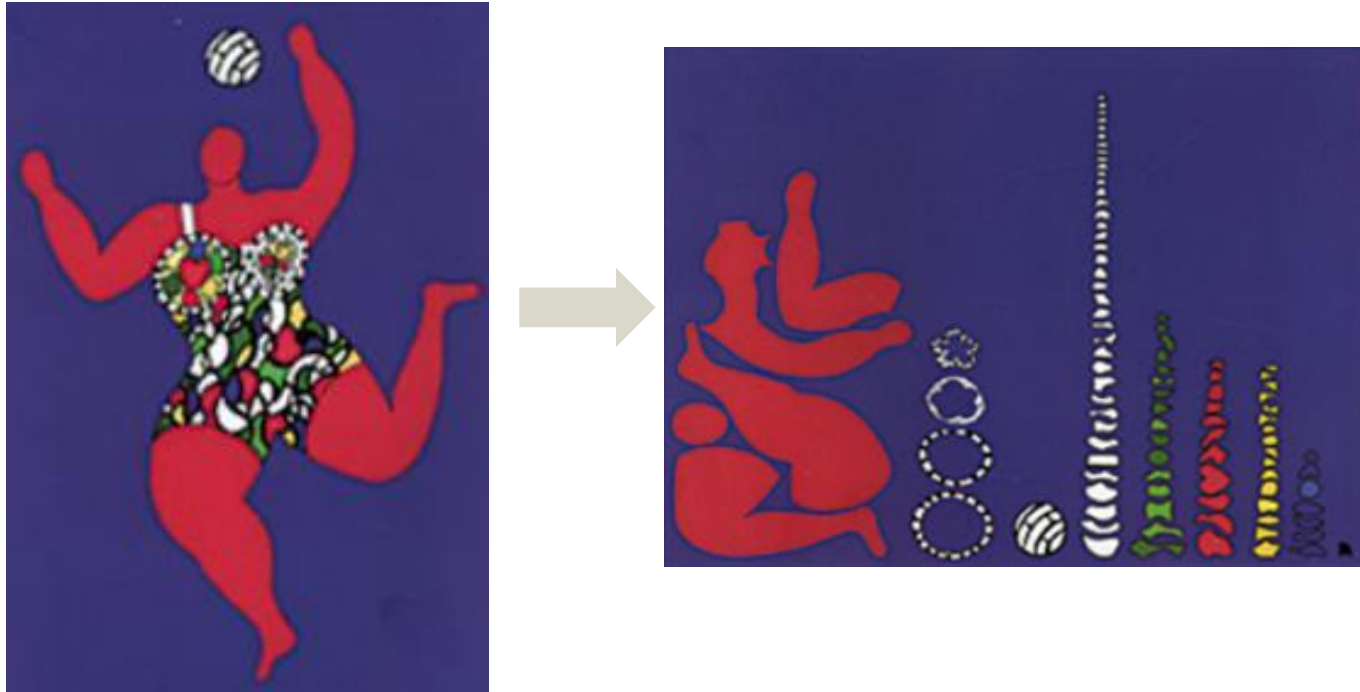
Who has benefited from the Excellence Initiative?

Growth of positions in administration vs. in science (2005-2012)



“These numbers show the real winners and losers of the increased cash flow in this past decade: the students and scientists lose out, while university administrations benefit the most from the billions.”

New KPIs? How to evaluate originality?



1

Input control

“The primary means for controlling the quality of scholarly activities is through selecting its members“

2

Content discussion

Who has recognized something that is true that others have not yet recognized?
 (Hint: Not those who celebrate "success" in the mainstream).

3

Digitization and social media

Openness
 Access
 Transparency
 Multidimensionality
 Democratization
 Complementarity ("Alt-, & Compmetrics")

San Francisco Declaration on Research Assessment



The San Francisco Declaration on Research Assessment, which has been endorsed by many leading institutions, clearly states: “***Do not use journal-based metrics, such as Journal Impact Factors, as a surrogate measure of the quality of individual research articles, to assess an individual scientist’s contributions, or in hiring, promotion, or funding decisions.***”

The recently released “Statement by three national academies (Académie des Sciences, Leopoldina and Royal Society) on good practice in the evaluation of researchers and research programmes” also asserts that “***[i]mpact factors of journals should not be considered in evaluating research outputs***”

Summary of the challenges of management science

- 1) Irrelevance of research
- 2) Focus on mainstream research and thus neglect of riskier projects
- 3) Low reliability and reproducibility of research results
- 4) Lack of transparency and access to data
- 5) Black box of the peer review system
- 6) Non-intended effects and manipulability
- 7) ...



„Scientific research has changed the world.
Now it needs to change itself.“

How to treat these diseases?

What journals can do

Publish:

- Registered reports
- Results-masked articles
- Replication studies
- Null results studies
- Exploratory studies
- Other types of articles (e.g. commentaries, critiques, adversarial collaborations)

What authors can do

- Theorize clearly
- Design realistic experiments
- Do high-quality qualitative research
- Do not ignore endogeneity
- Be transparent with data, methods, and reporting
- Declare conflicts of interest

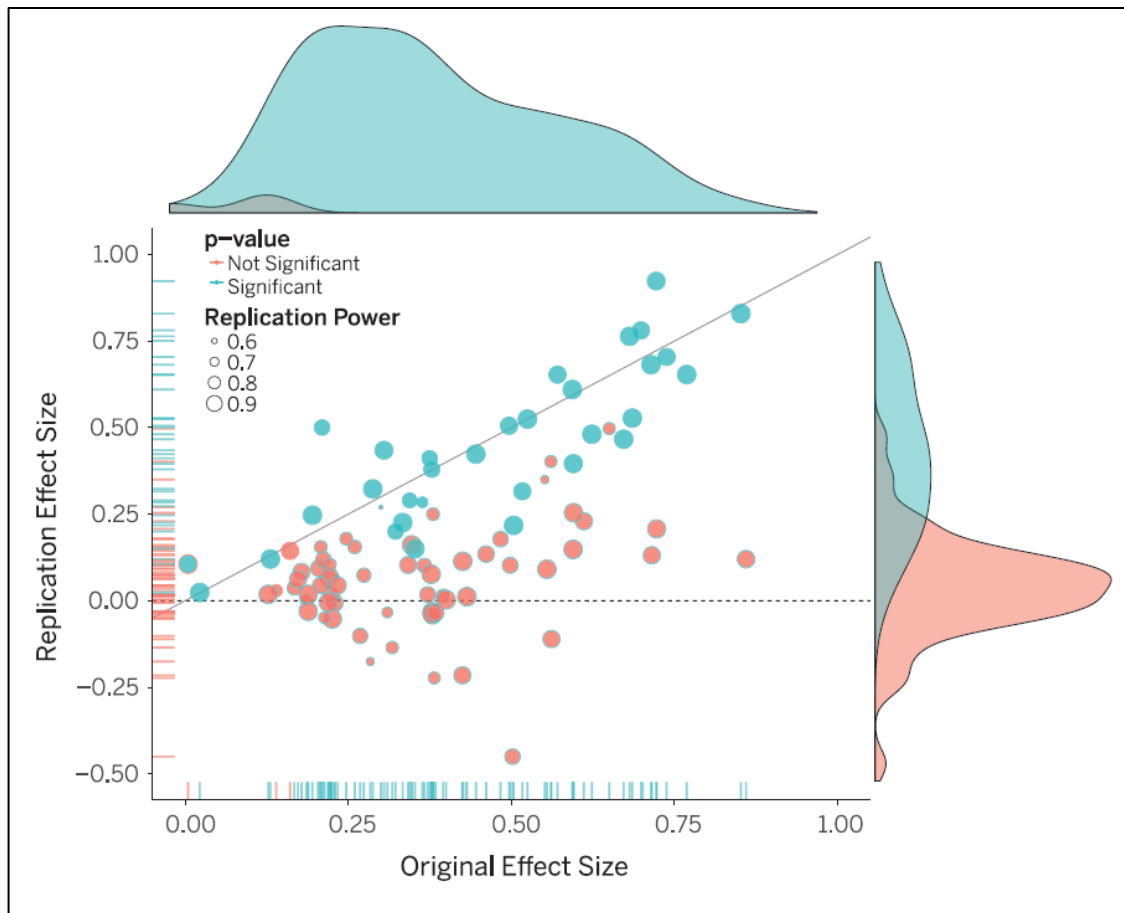
Solutions for better science in general

- Ten Principles of the Leiden Manifesto or DORA
- Coincidence
- Article Based Metrics
- Post Publication Reviews (open)
- Independent Thinking
- Look for Impact
- Replication studies
- Open Science
- Pre-Registration

Solutions for better science in general

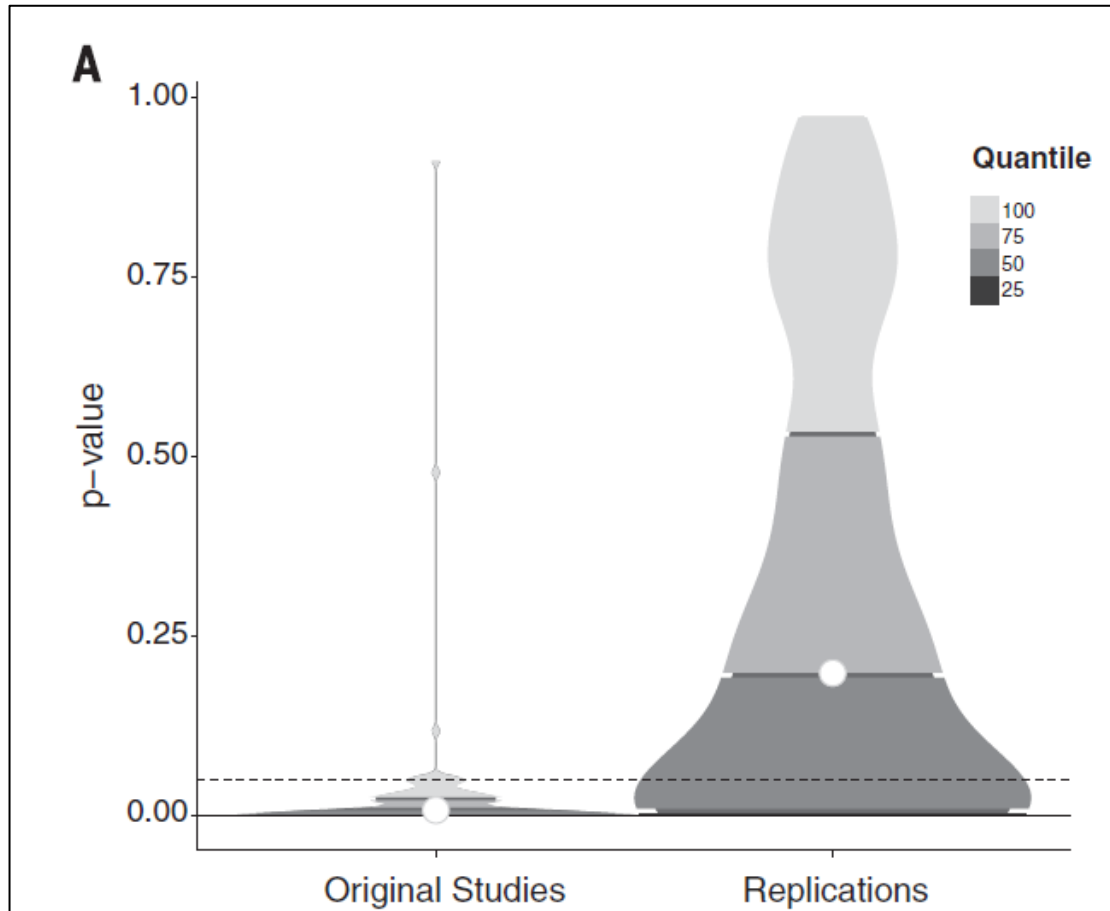
- Ten Principles of the Leiden Manifesto or DORA
- Coincidence
- Article Based Metrics
- Post Publication Reviews (open)
- Independent Thinking
- Look for Impact
- **Replication studies**
- Open Science
- Pre-Registration

Reproducibility project in psychology (1/3)



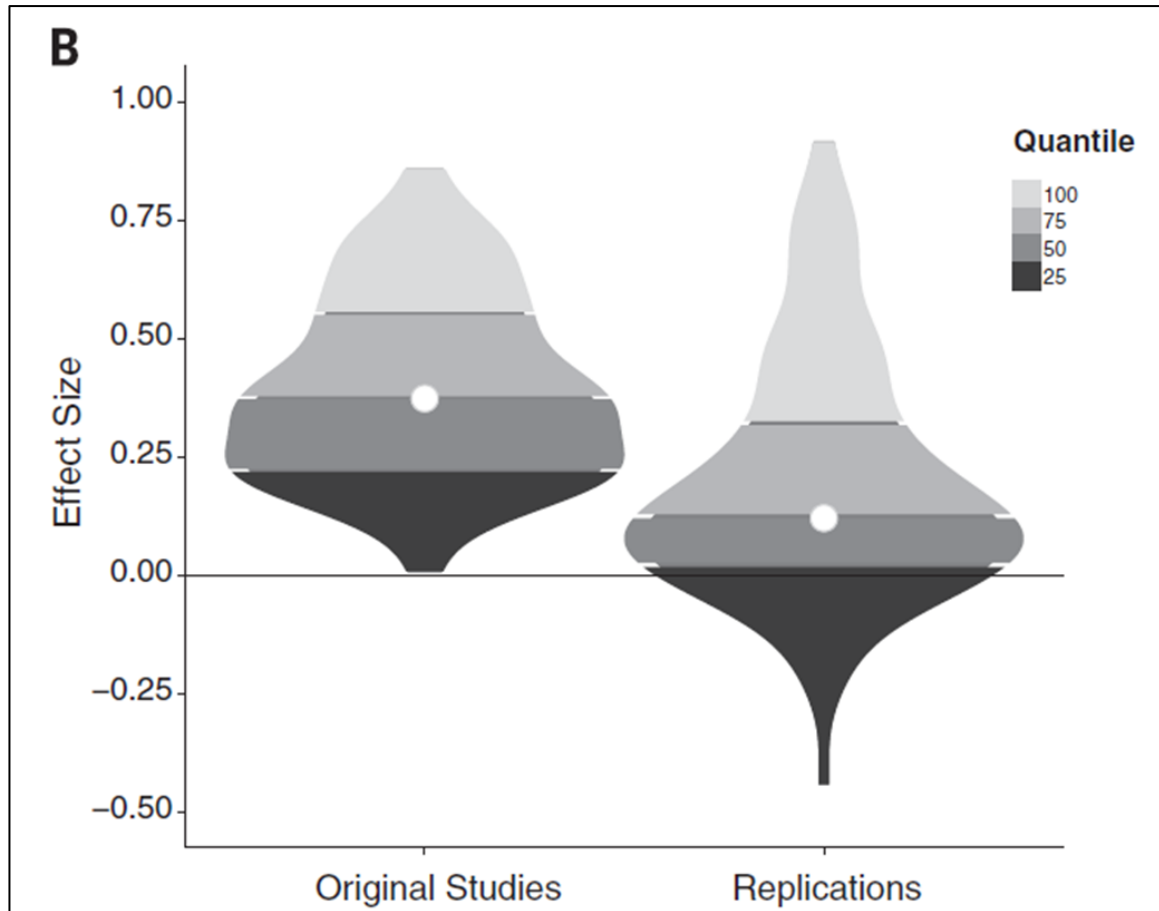
Source: Open Science Collaboration (2015).

Reproducibility project in psychology (2/3)



Evidence for
p-hacking?

Reproducibility project in psychology (3/3)



Source: Open Science Collaboration (2015).

Classification of replications

Types of Replication		
	Same Measurement and Analysis	Different Measurement and/or Analysis
Same Data Set	Checking of analysis	Reanalysis of data
Same Population	Exact replication	Conceptual extension
Different Population	Empirical generalization	Generalization and extension

Source: Tsang and Kwan (1999).

	Same Research Design	Different Research Design
Same Data and Sample	Checking for errors and/or falsification of results	
Same Population (Same Context) with Different Sample	Reliability and representativeness of data	
Different Population (Different Context)		

Figure 2. Narrow replication

	Same Research Design	Different Research Design
Same Data and Sample		Robustness to different measures, methods, and models
Same Population (Same Context) with Different Sample		Robustness to different measures, methods, and models
Different Population (Different Context)	Generalize to new population (subjects, industry, time period, etc.)	Generalize to new population and assess robustness

Figure 3. Quasi-replication

Source: Bettis et al. (2016).

Replication versus robustness

Table 1. A Proposed Standard for Classifying Any Study as a Replication.

	Sampling distribution for parameter estimates	Sufficient conditions for discrepancy	Types	Methods in follow-up study versus methods <i>reported</i> in original			Examples
				Same specification	Same population	Same sample	
Replication	<i>Same</i>	<i>Random chance, error, or fraud</i>	Verification	Yes	Yes	Yes	<i>Fix faulty measurement, code, data set</i>
			Reproduction	Yes	Yes	No	<i>Remedy sampling error, low power</i>
Robustness	<i>Different</i>	<i>Sampling distribution has changed</i>	Reanalysis	No	Yes	Yes/No	<i>Alter specification, recode variables</i>
			Extension	Yes	No	No	<i>Alter place or time; drop outliers</i>

Notes: The “same” specification, population, or sample means the same as *reported* in the original paper, not necessarily what was contained in the code and data used by the original paper. Thus for example if code used in the original paper contains an error such that it does not run exactly the regressions that the original paper said it does, new code that fixes the error is nevertheless using the “same” specifications (as described in the paper).

Source: Clemens (2017).

Some evidence from the field of economics

Table 2—: Coding Rates

	Replications	Extension	Robustness	Any	<i>N</i>
<i>Volume Sample</i>	28.6% (20)	48.6% (34)	40.0% (28)	60.0% (42)	70
<i>Citing Sample</i>	3.4% (52)	7.8% (121)	4.7% (73)	11.0% (170)	1,546

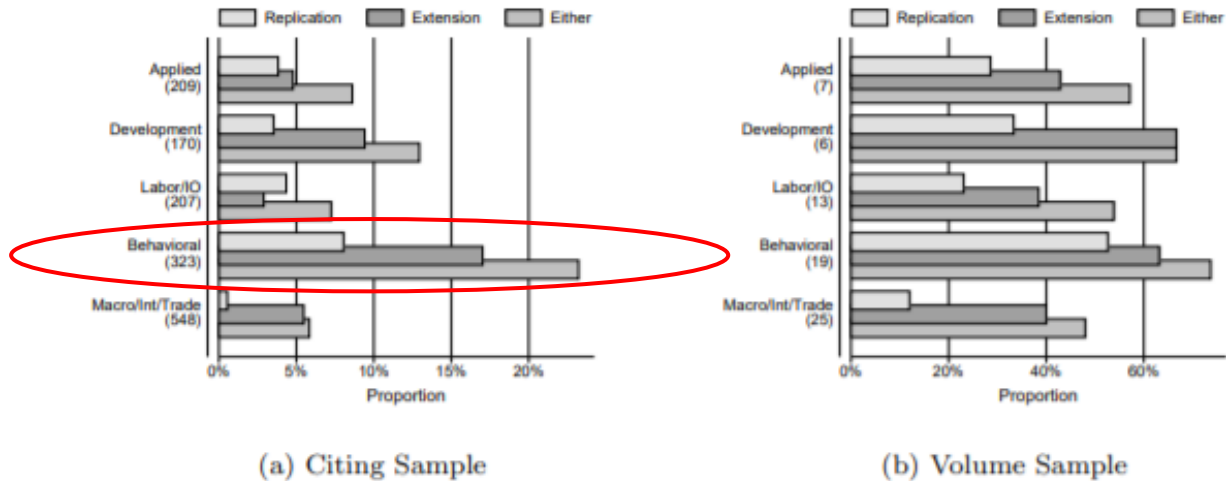


Figure 1. : Replications/Extensions by Field

Source: Berry et al. (2017).



Contents lists available at [ScienceDirect](https://www.sciencedirect.com)

Research Policy

journal homepage: www.elsevier.com/locate/respol



Replication studies in economics—How many and which papers are chosen for replication, and why? ☆



Frank Mueller-Langer^{a,b,*}, Benedikt Fecher^{c,d}, Dietmar Harhoff^b, Gert G. Wagner^{c,d,e}

A B S T R A C T

We investigate how often replication studies are published in empirical economics and what types of journal articles are replicated. We find that between 1974 and 2014 0.1% of publications in the top 50 economics journals were replication studies. We consider the results of published formal replication studies (whether they are negating or reinforcing) and their extent: Narrow replication studies are typically devoted to mere replication of prior work, while scientific replication studies provide a broader analysis. We find evidence that higher-impact articles and articles by authors from leading institutions are more likely to be replicated, whereas the replication probability is lower for articles that appeared in top 5 economics journals. Our analysis also suggests that mandatory data disclosure policies may have a positive effect on the incidence of replication.

The study sample

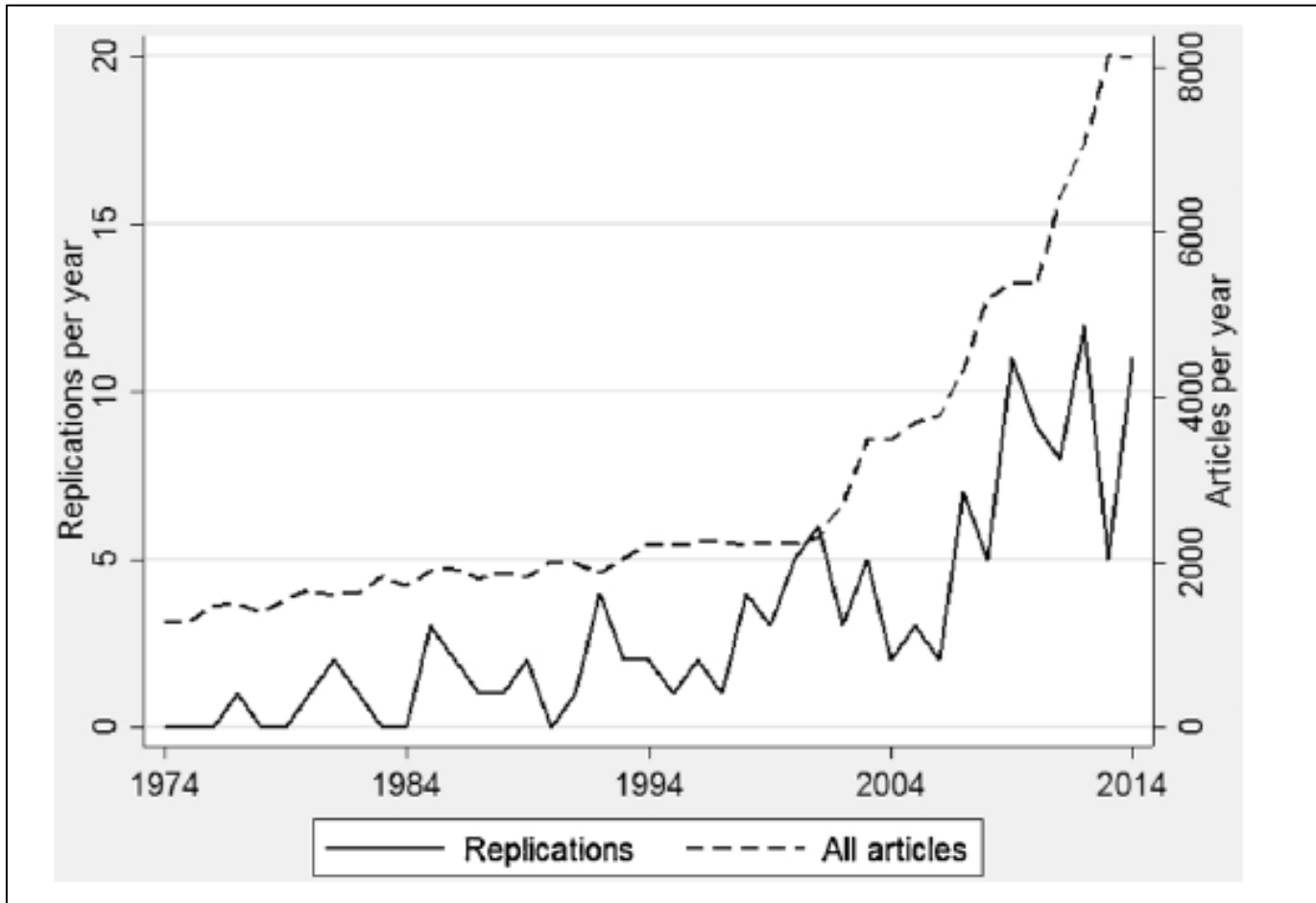
Table 1
Descriptive statistics.

	mean	sd	min	max	N
Dependent variable					
Replicated article	0.105		0	1	1243
Negated replicated article	0.049		0	1	1243
(Partially) reinforced replicated article	0.056		0	1	1243
Main variables of interest					
Total citations before publication of replication	20.89	64.18	0	1508	1243
Lag between publication of replicated article and replication	4.851	3.601	0	23	1243
Journal impact factor	3.516	1.153	2.137	6.033	1243
Top five economics journal	0.512		0	1	1243
Top 50 university	0.606		0	1	1243
Mandatory data disclosure policy	0.230		0	1	1243
Data or program code available	0.169		0	1	1243
Control variables					
Self-created data	0.124		0	1	1243
Confidential or proprietary data	0.012		0	1	1243
Article published in conference proceedings	0.118		0	1	1243
Number of references	29.28	17.83	0	130	1243
Number of pages	19.51	10.94	1	65	1243
Number of authors	2.057	1.097	1	16	1243
h-index of best author	17.42	12.90	0	106	1225
Third-party funding	0.185		0	1	1243
Funder's support for data availability	0.598	1.366	0	5	1243

10.5 % of the study sample is a replication.

About half of the replications negated the result of the replicated article.

Amount and share of replications over time



Source: Mueller-Langer et al. (2021).

Replicated versus nonreplicated articles

Table 2
Extended descriptive statistics.

	Replicated articles		Nonreplicated articles	
	<i>CitesPreReplication</i> mean	<i>LagReplication</i> mean	<i>CitesPreReplication</i> Mean	<i>LagReplication</i> mean
Main variables of interest				
<i>Top5Journal</i> = 1	100.57	6.71	24.51	6.10
<i>Top5Journal</i> = 0	25.23	3.64	7.40	3.45
<i>Top50University</i> = 1	73.07	5.04	19.67	5.07
<i>Top50University</i> = 0	26.69	5.17	11.32	4.47
<i>MandatoryDisclosure</i> = 1	35.22	3.89	12.06	4.04
<i>MandatoryDisclosure</i> = 0	64.66	5.26	17.58	5.07
<i>DataOrCode</i> = 1	29.00	4.26	20.15	3.87
<i>DataOrCode</i> = 0	65.99	5.22	15.45	5.02

The replicated articles have far more prereplication citations than similar articles in the same issue of the journal!

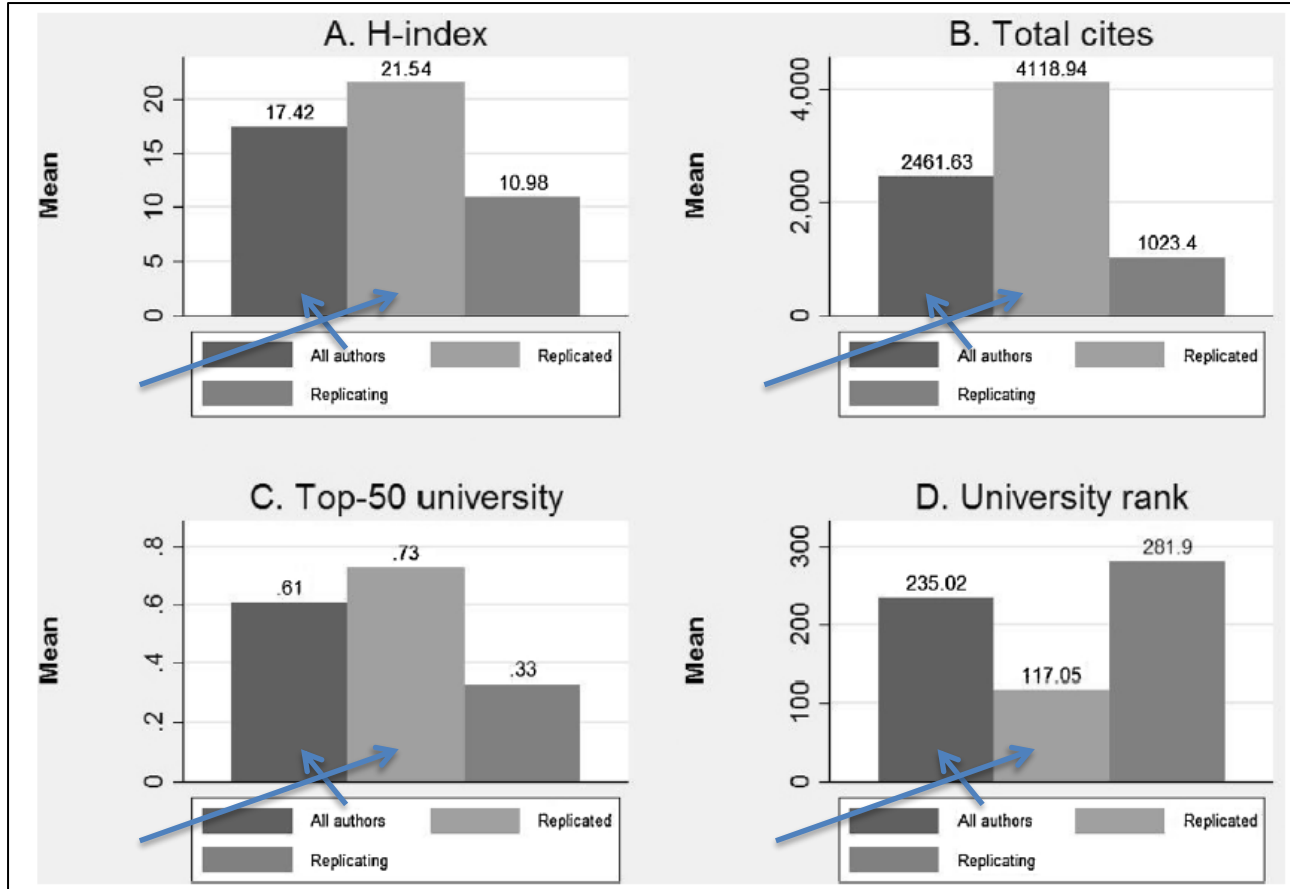
Source: Mueller-Langer et al. (2021).

Determinants of being replicated

Issues/Sample:	[1] All	[2] All	[3] All	[4] All	[5] All	[6] All	[7] With sci. repl. art.	[8] With neg. repl. art.	[9] With reinf. repl. art.
Dependent variable:	Repl. article	Repl. article	Repl. article	Repl. article	Repl. article	Repl. article	Scient. repl. art.	Negated repl. art.	Reinf. repl art.
Log total citations before publication of replication		0.055*** (0.006)	0.056*** (0.006)	0.053*** (0.006)	0.053*** (0.006)	0.053*** (0.006)	0.053*** (0.007)	0.067*** (0.010)	0.040*** (0.007)
Log lag between publication of replicated article and replication		-0.106*** (0.014)	-0.106*** (0.014)	-0.100*** (0.014)	-0.101*** (0.014)	-0.101*** (0.015)	-0.105*** (0.013)	-0.142*** (0.022)	-0.073*** (0.016)
Top 5 economics journal = 1			-0.096*** (0.024)	-0.092*** (0.025)	-0.081** (0.030)	-0.090*** (0.025)	-0.072*** (0.018)	-0.037 (0.043)	-0.062** (0.021)
Log impact factor			-0.040 (0.022)	-0.036 (0.023)	-0.042 (0.027)	-0.038 (0.026)	-0.072** (0.025)	-0.008 (0.040)	-0.071 (0.037)
Top 50 university = 1				0.038 (0.020)	0.038 (0.020)	0.038 (0.020)	0.044* (0.019)	0.019 (0.027)	0.045* (0.021)
Mandatory data-disclosure policy = 1					-0.011 (0.016)				
Data or program code available = 1						-0.002 (0.012)	-0.016 (0.017)	-0.014 (0.025)	0.003 (0.008)
Observations	1225	1225	1225	1225	1225	1225	973	563	714
Pseudo R-squared	0.0723	0.111	0.113	0.120	0.120	0.120	0.139	0.129	0.145
Log Pseudo Likelihood	-384.5	-368.4	-367.4	-364.9	-364.8	-364.9	-290.1	-168.2	-193.8
Journal Dummies	YES	YES	YES	YES	YES	YES	YES	YES	YES
Year Dummies	YES	YES	YES	YES	YES	YES	YES	YES	YES

Source: Mueller-Langer et al. (2021).

Who replicates vs. who gets replicated



Source: Mueller-Langer et al. (2021).

Creating Repeatable Cumulative Knowledge in Strategic Management: A Call For a Broad and Deep Conversation among Authors, Referees, and Editors

Richard A. Bettis, Sendil
Ethiraj, Alfonso Gambardella,
Constance Helfat, and Will Mitchell

2016

Strategic Management Journal



Repeatability problems of statistical studies



- “A third of the most-cited clinical research seems to have replication problems, and this seems to be as large, if not larger than the vast majority of less-cited clinical research” (Ioannidis 2005: 224).
- “Much of the scientific literature, perhaps half may simply be untrue” (Horton, 2015) Editor-in-Chief of the top medical journal *The Lancet*.
- Most relevant study reported on 100 replications from three important psychology journals in 2008: 97 percent of the original studies had statistically significant results, while only 36 percent of the replications did (Open Science Collaboration, 2015).
- Problem: only evidence from a study with a significant coefficient is publishable in top journals, also if other studies using a different sample of similar data find that the appropriate model coefficient is not statistically significant.

Horton R. 2015. Offline: what is medicine's 5 sigma. **385**. Available at: <http://www.thelancet.com>.

Ioannidis PAJ. 2005. Contradicted and initially stronger effects in highly cited clinical research. *Journal of the American Medical Association* **294**(4): 218–228.

Open Science Collaboration. 2015. Estimating the reproducibility of psychological science. *Science* **349**: 943.

Does a significant p -value have value?



- P-values arising from Null Hypothesis Significance Tests provide no information “regarding the reliability of the research” (Branch, 2014: 257).
- It is incorrect to interpret p as the probability that the null hypothesis is false. Instead, p is the probability that the sample value would be at least as large as the value actually observed if the null hypothesis is true (Wonnacott and Wonnacott, 1990: 294).
- The true definition of p -value permits a conclusion only about the probability of finding a result in a particular sample.
- “Searching for asterisks” (Bettis, 2012) is inconsistent with Popper’s falsifiability criterion.

Bettis RA. 2012. The search for asterisks: compromised statistical tests and flawed theory. *Strategic Management Journal* **33**(1): 108–113.

Branch M. 2014. Malignant side-effects of null-hypothesis testing. *Theory and Psychology* **24**(2): 256–277.

Wonnacott TH, Wonnacott RJ. 1990. *Introductory Statistics for Business and Economics*. Wiley & Sons: New York, NY.

Does p-size matter?



- The size of p-values (0.05, 0.01, 0.001) is often taken as a measure of the “strength” of the result, where smaller p-values are considered stronger evidence.
- A rigid p-value breakpoint between truth and irrelevancy is inconsistent with good science.
- The strength of a result in terms of economic, behavioral, or practical importance, as indicated by the size of the estimated coefficient, is sometimes ignored if the coefficient is significant.

Replication is the measure of repeatability



- Publishing only statistically significant results, while not publishing replications or non-results is inconsistent with the establishment of repeatable cumulative knowledge
- One significant coefficient in one study proves little or nothing.
- A single replication without statistical significance on the coefficient of interest does not disprove anything. Instead, it adds disconfirming evidence.
- Because the nature of statistical testing is probabilistic, we can only make statements about the balance of evidence.

What to do:

New policies at *strategic management journal*

- *SMJ* will publish and welcomes submissions of replication studies
- *SMJ* will publish and welcomes submissions of studies with non-results. These types of studies demonstrate a lack of statistical support in a particular sample for specific hypotheses or research propositions
- *SMJ* will no longer accept papers for publication that report or refer to cut-off levels of statistical significance (p -values). In statistical studies, authors should report either standard errors or exact p -values (without asterisks) or both and should interpret these values appropriately in the text.
- *SMJ* will require in papers accepted for publication that authors explicitly discuss and interpret effect sizes of relevant estimated coefficients.

The Replication Recipe: What makes for a convincing replication?

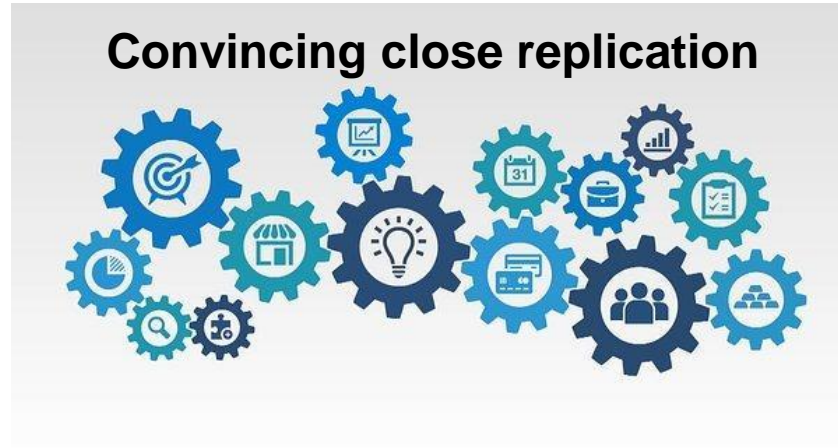
Brandt, M. J., et al.

2014

Journal of Experimental
Social Psychology



Replicate to build a cumulative knowledge base



Test assumed underlying theoretical process



Test the robustness of an effect



Assess average effect size of an effect



Refinement of old, psychological theories

The Replication Recipe

-

Standard criteria for a convincing close replication

Ingredient #1: Carefully defining the effects and methods that the researcher intends to replicate

How does the precise effect you intend to replicate look like?



- Verbal description of the effect
- Effect size, size's confidence intervals

Is it important and necessary to replicate this particular effect?



- Theoretical importance to a particular field or direct or indirect value to society
- Existing confidence in the reliability of the effect

What methods were used to detect the effect in the original study?

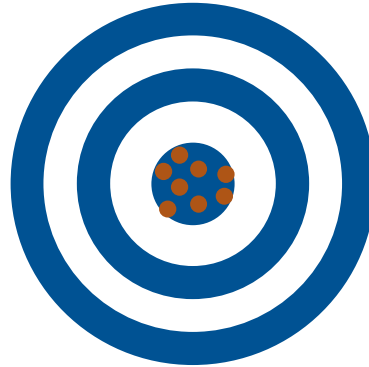


- Sample size & sampling procedure
- Demographics of the participants
- Study design

Ingredient #2: Following exactly the methods of the original study

Designing the replication study

- Use of original study materials (contact the authors)
- OR
- Recreate the necessary methods (ask experts to provide feedback & conduct pilot studies)



Documentation

- Identify & justify differences
- Categorize which parts of the study are
 - ... exactly the same as,
 - ... close to
 - ... conceptually different tothe original study

Ingredient #3: Having high statistical power



Why is a high statistical power important?

- Allowing a strong chance to confirm the effect size from the original study as significant
- Avoiding incorrect conclusions that original effects are false positives

How to derive a sufficient sample size?

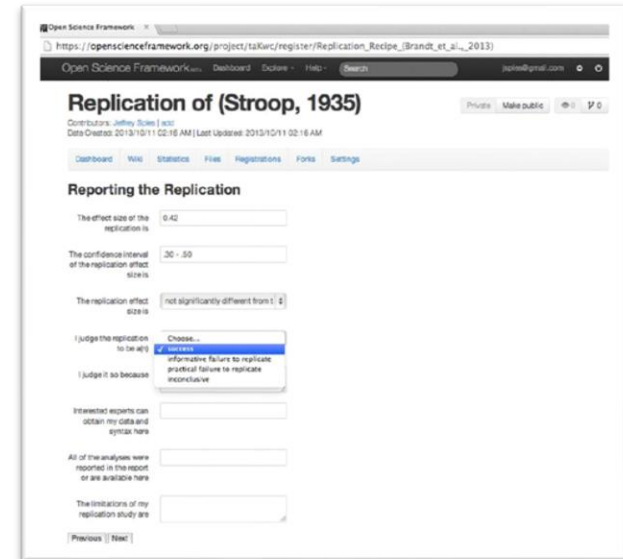
- Power calculations
(Aberson, 2010; Cohen, 1992; Faul, Erdfelder, Lang, & Buchner, 2007; Maxwell, Kelley, & Rausch, 2008; Scherbaum & Ferreter, 2009; Shieh, 2009; Zhang & Wang, 2009)
- Take 2.5 times the original sample size
(Simonsohn, 2013)

Ingredient #4: Making complete details about the replication available

As much openness as ethically possible

- Pre-registration of replication attempts
- Openness in methods, sample, and procedure
- Make data, syntax, analyses and all results available
- Accessible to readers, editors and reviewers

Reporting a replication on the Open Science Framework



The screenshot shows a web browser window displaying the Open Science Framework interface. The page title is "Replication of (Stroop, 1935)". The URL is [https://osf.io/project/taKwz/register/Replication_Recipe_\(Brandt_et_al_2013\)](https://osf.io/project/taKwz/register/Replication_Recipe_(Brandt_et_al_2013)). The page includes a navigation menu with "Dashboard", "Wiki", "Statistics", "Files", "Registrations", "Fora", and "Settings". The main content area is titled "Reporting the Replication" and contains several input fields and a dropdown menu. The fields are: "The effect size of the replication is" (0.42), "The confidence interval of the replication effect sizes" (20 - .00), "The replication effect size is" (not significantly different from 1: 0), "I judge the replication to be a(n)" (Choose...), "I judge it to be because" (Choose...), "Interested experts can obtain my data and syntax here" (text input), "All of the analyses were reported in the report or are available here" (text input), and "The limitations of my replication study are" (text input). A dropdown menu is open under "I judge the replication to be a(n)", showing options: "Choose...", "affirmative failure to replicate", "practical failure to replicate", and "inconclusive".

Ingredient #5: Evaluating replication results and comparing them critically to the results of the original study



Evaluation & Comparison

- 1) Test the size, direction and confidence interval of the effect (replication effect is significantly different from the null).
- 2) Test whether it is significantly different from the original effect.
- 3) Meta-analytic aggregation of the replication study's effect with the original and other close replications.



Solutions for better science in general

- Ten Principles of the Leiden Manifesto or DORA
- Coincidence
- Article Based Metrics
- Post Publication Reviews (open)
- Independent Thinking
- Look for Impact
- Replication studies
- Open Science**
- Pre-Registration

What is Open Science?



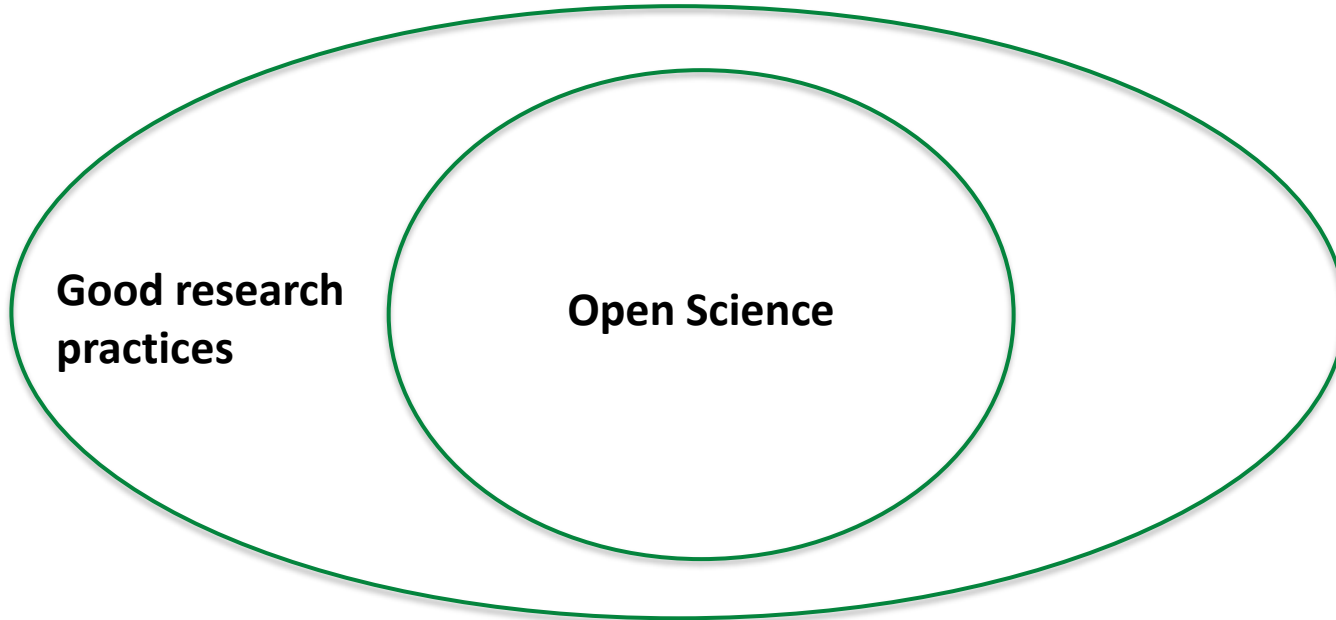
Good Research Practice

Good researchers

- Strive for excellence and take responsibility
- Respect the law, research ethics, and professional standards
- **Support a culture of transparency, openness, and honesty towards other researchers and the public**
- **Maximize public benefit and avoid resource waste**
- Continue learning and mentor others



Open Science



Basic principles of Open Science

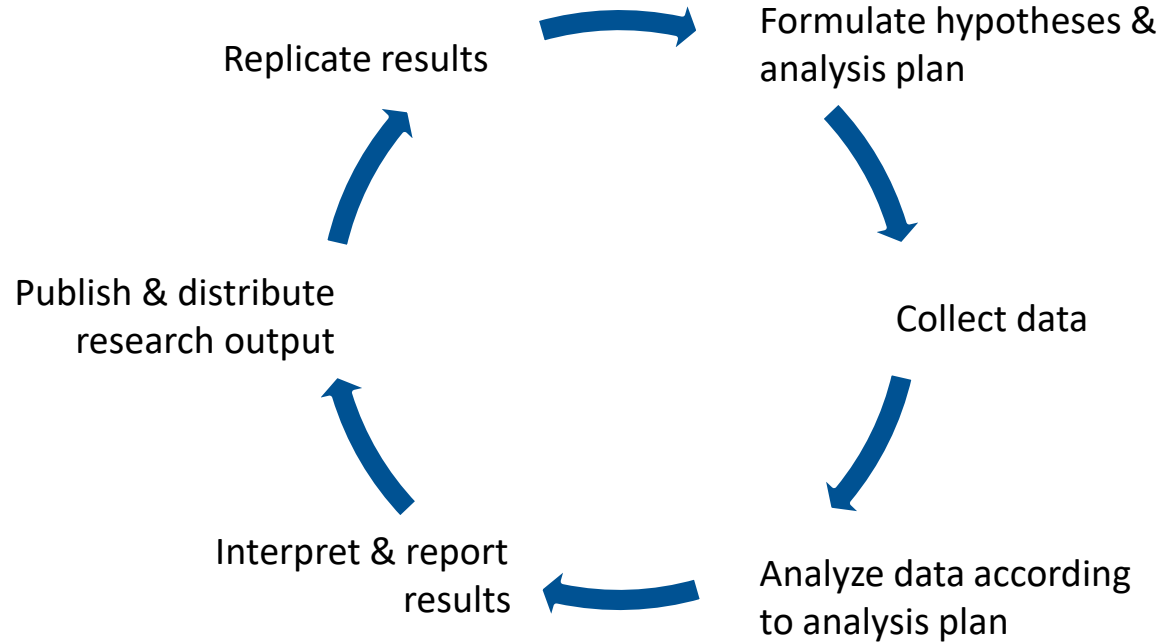
“free to use, re-use and re-distribute for all”

- Transparency
- Reproducibility
- Reusability
- Open Communication

“Open science is the idea that scientific knowledge of all kinds should be openly shared as early as is practical in the discovery process.” Michael Nielsen

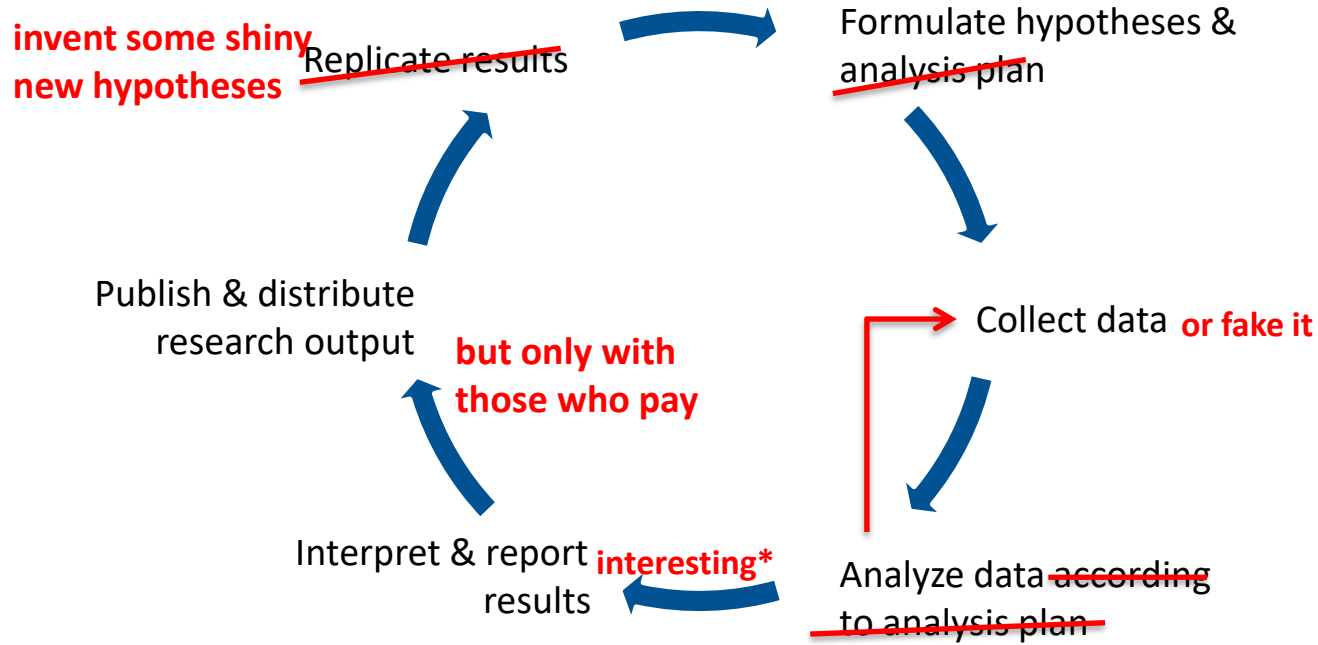


The confirmatory research cycle



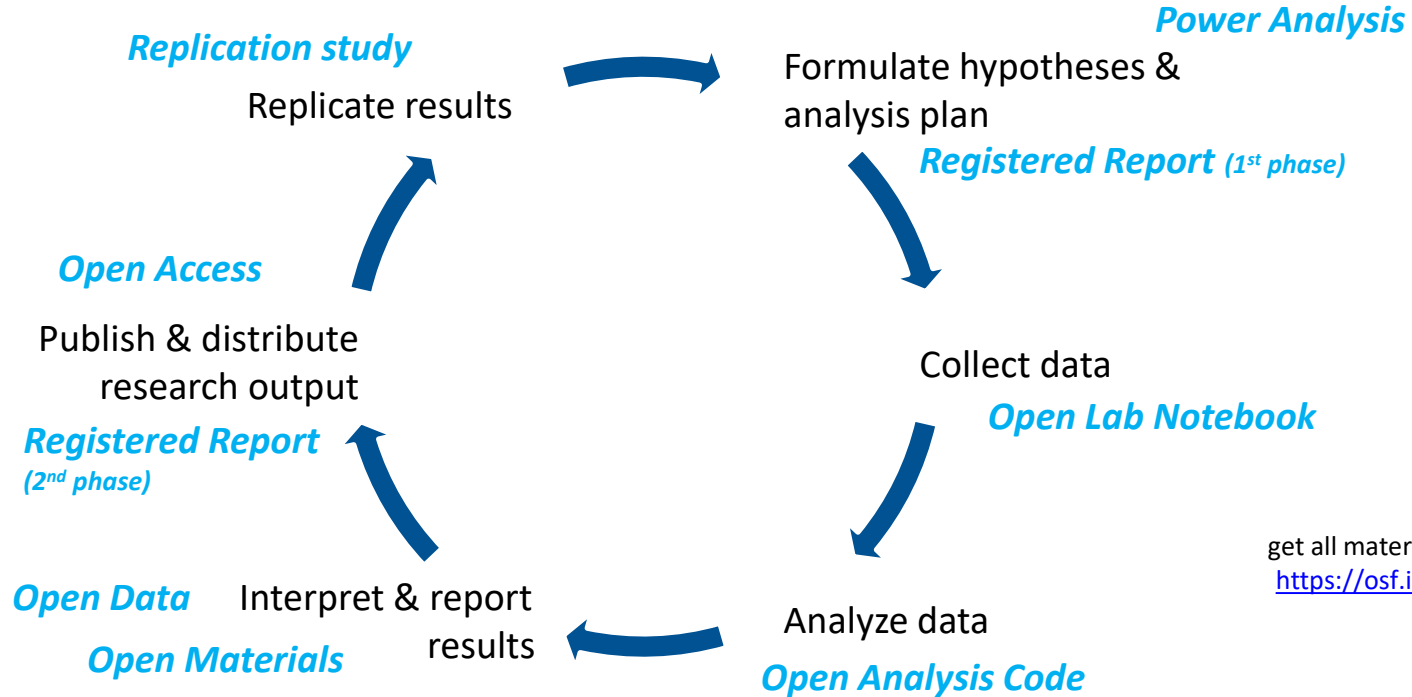
The confirmatory research cycle

How can you know that it does not look like this?



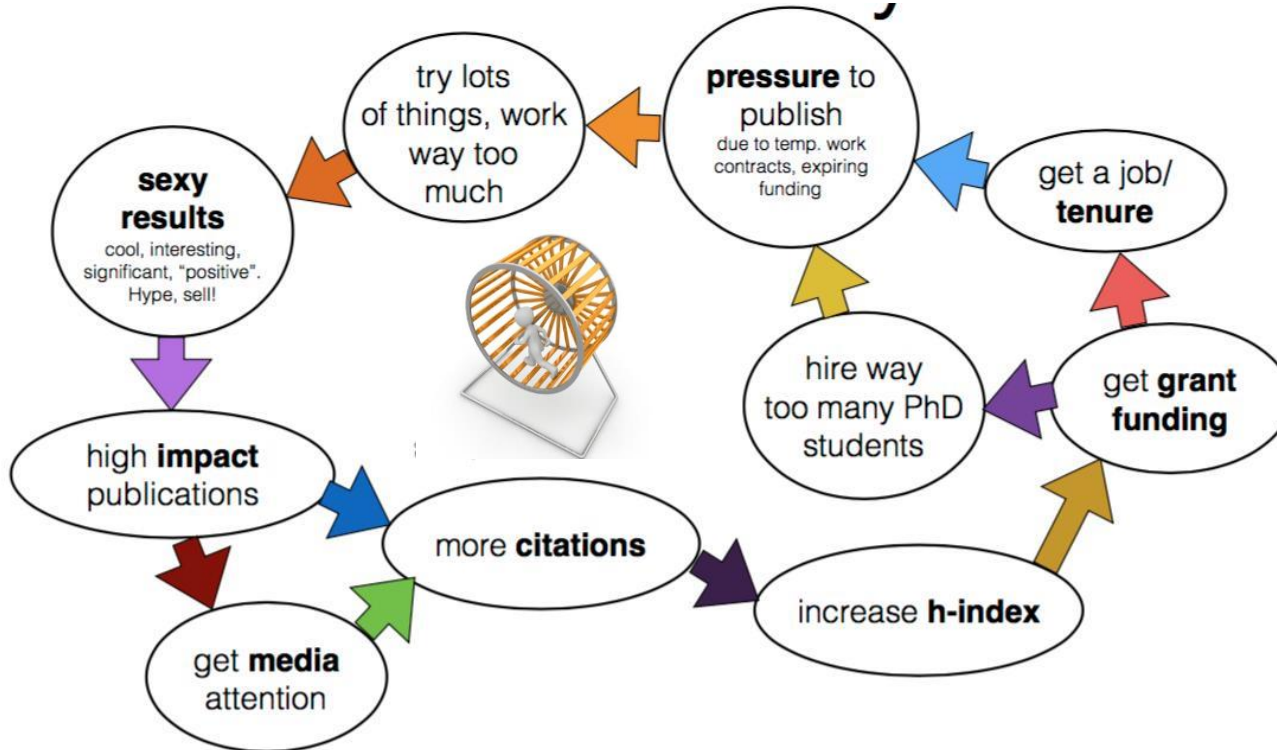
* $p < .05$; that fit a theory; that are surprising / publishable...

Open Science in the research process

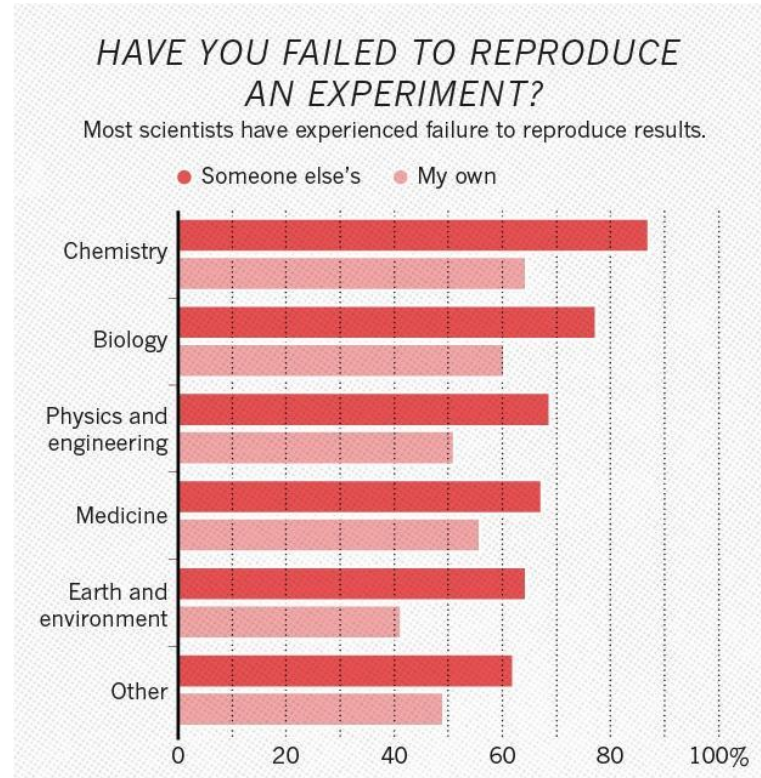


get all material here:
<https://osf.io/zirhu/>

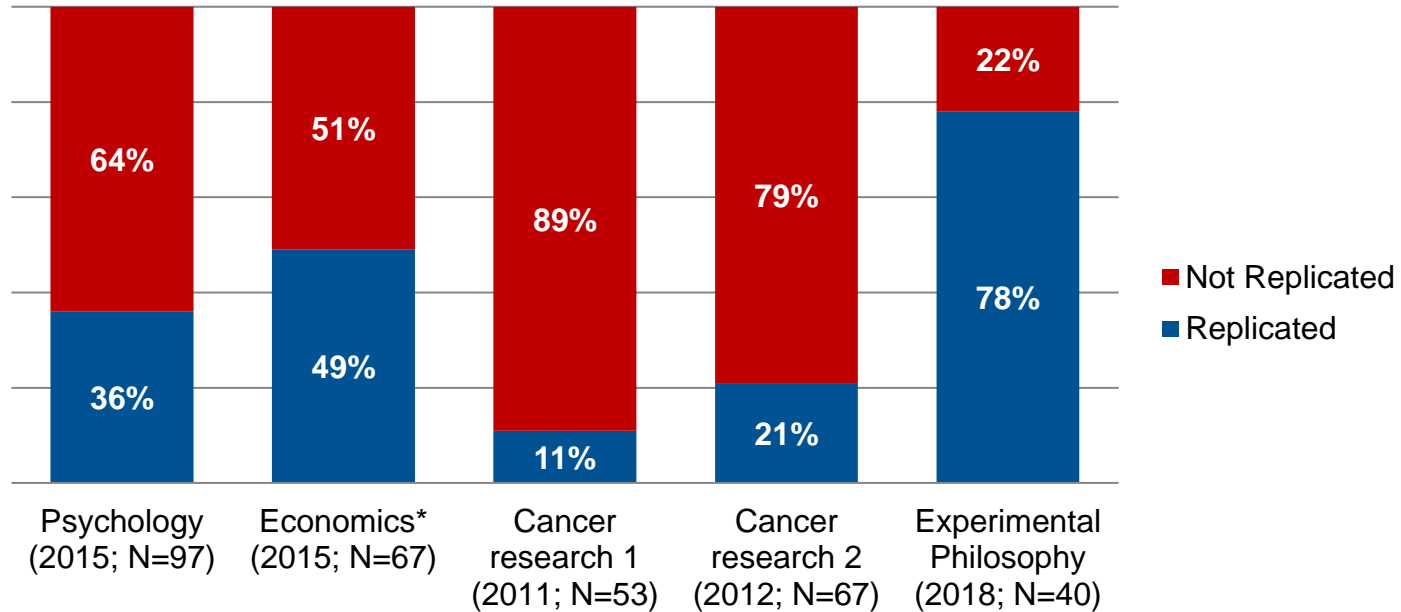
The science hamster wheel



Is there a Replication Crisis?

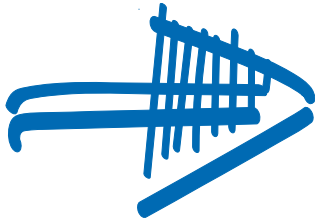


Is there a Replication Crisis?



* The data on economics is about *reproducibility*; i.e. the attempt to get the same results if you apply the original data analysis on the original data set.

Current problems in science overall



Scientific progress is slowed down



Published results cannot be trusted






Resources are wasted

What can you do?

Publication bias
QRPs / p-hacking

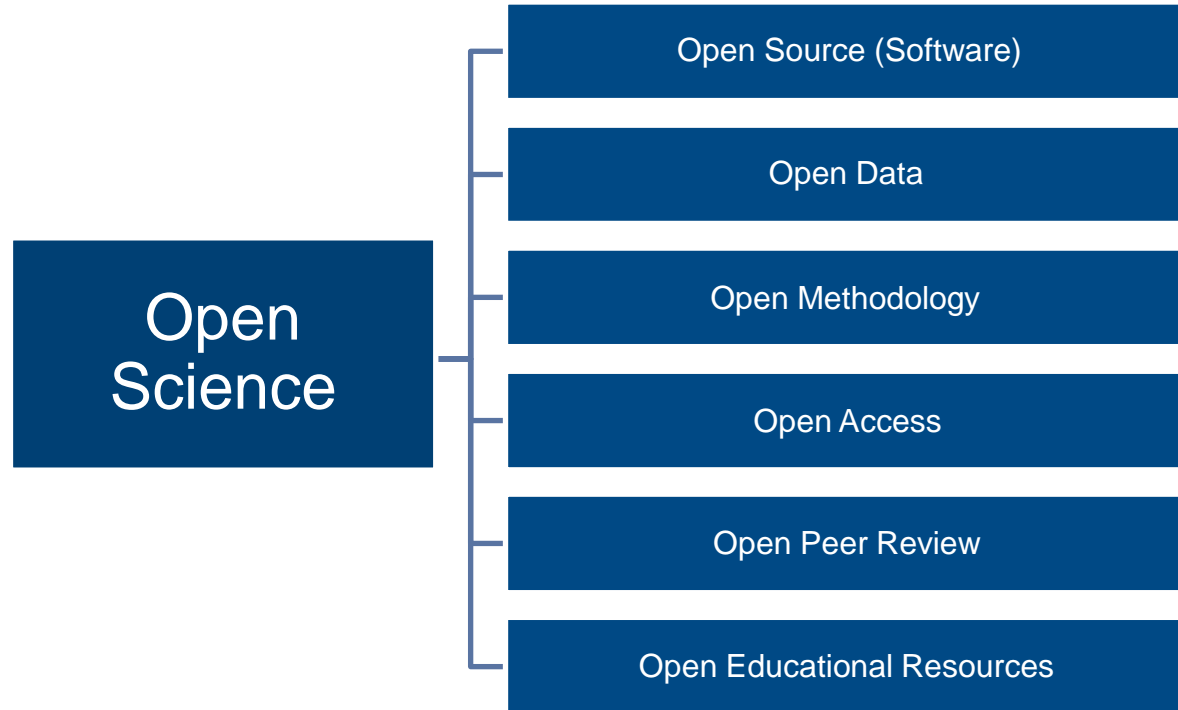


-  Scientific progress is slowed down
-  Published results cannot be trusted
-  Resources are wasted



- (1) Identify Questionable Research Practices**
- (2) Practice Open Science: Make your own research trustworthy**
- (3) Help to change incentive structures**

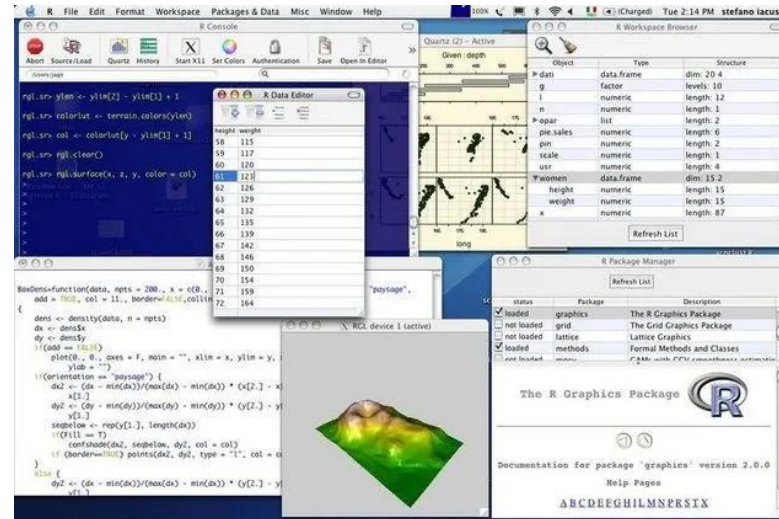
Overview: Open Science



Open Source

= decentralized software development model that encourages open collaboration

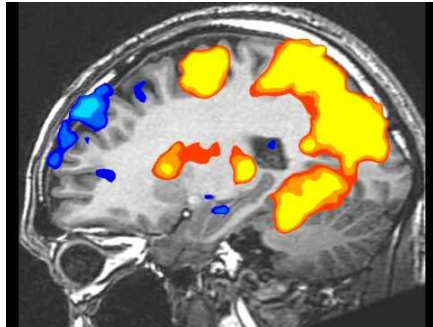
- Example: The R Project for Statistical Computing



Open Data

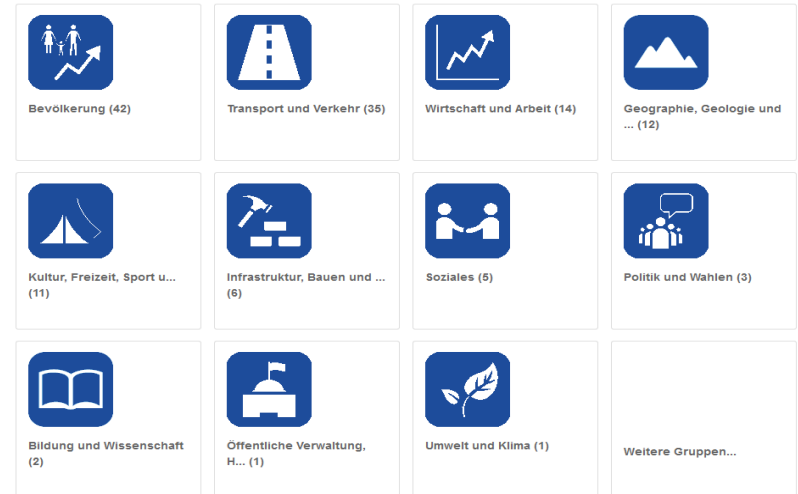
= data is freely available to everyone to use and republish as they wish, without restrictions from copyright, patents or other mechanisms of control

- **Examples** from other fields:
 - fMRI Data Center
 - Open-Data-Portal München



Open-Data-Portal München - Statistik

151	8	14	8
Datensätze	Organisationen	Gruppen	Zugehörige Elemente



Open Methodology

= using methods and documenting the entire process in a practicable and relevant manner

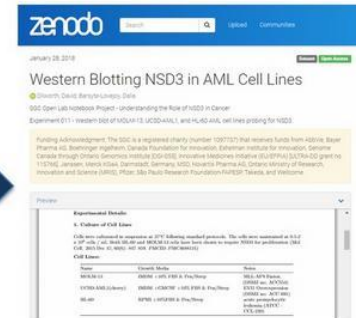
- Example:

<https://openlabnotebooks.org/>

We introduce our experiments at openlabnotebooks.org with links to experimental details



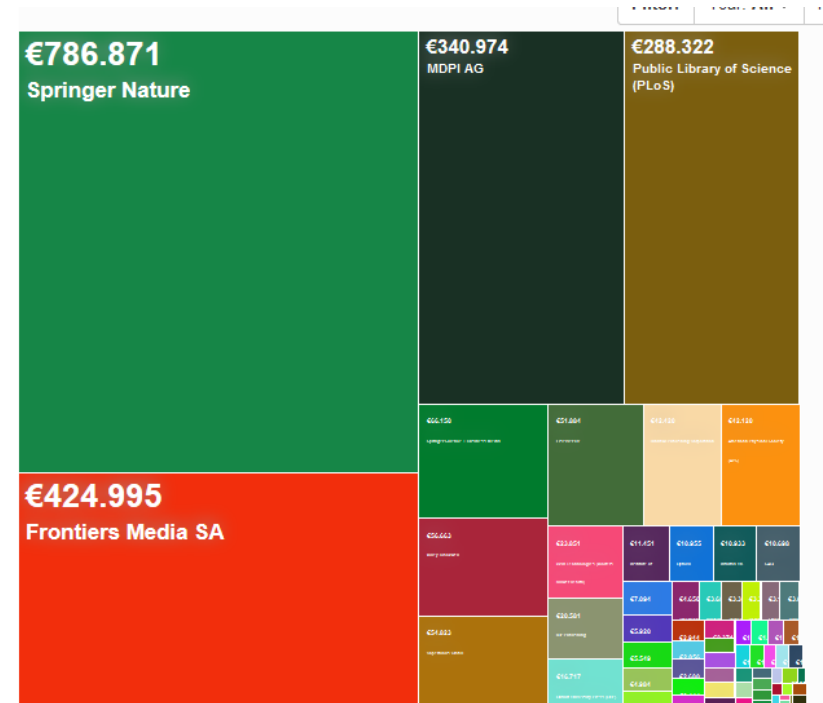
We provide all experimental details at zenodo.org



Open Access

= research outputs are distributed online, free of cost or other access barriers

- **Example:**
 - The **TUM Open Access Publishing Fund** covers article processing charges (APC) for papers published by TUM members in open access journals when certain eligibility conditions are fulfilled.
 - Have a look at the **TUM Open Access Policy:**
<https://www.ub.tum.de/en/open-access-policy>



Open Access

= research outputs are distributed online, free of cost or other access barriers



= research outputs are distributed online, free of cost or other access barriers

- Example: <https://doaj.org/>

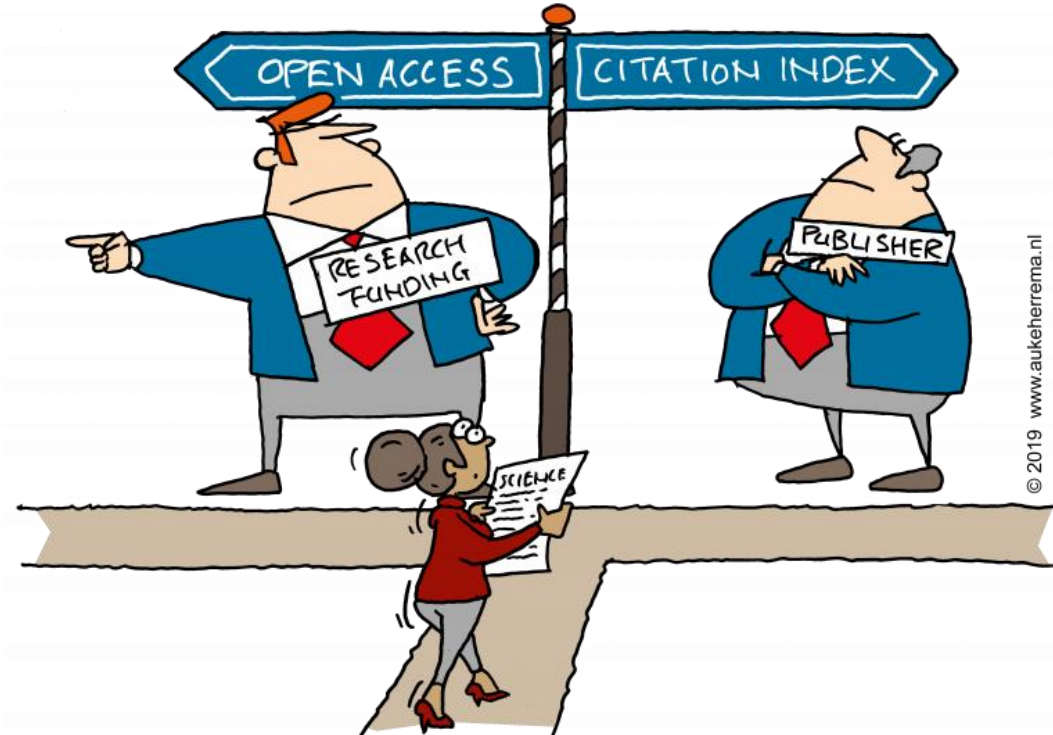
DOAJ (Directory of Open Access Journals)

DOAJ is a community-curated online directory that indexes and provides access to high quality, open access, peer-reviewed journals. DOAJ is independent. All funding is via donations, 22% of which comes from [sponsors](#) and 78% from [members and publisher members](#). All DOAJ services are free of charge including being indexed in DOAJ. All data is freely available.

DOAJ operates an education and outreach program across the globe, focussing on improving the quality of applications submitted.

[Why index your journal in DOAJ?](#)

A conflict of interests...



= modifications of the traditional scholarly peer review process, including

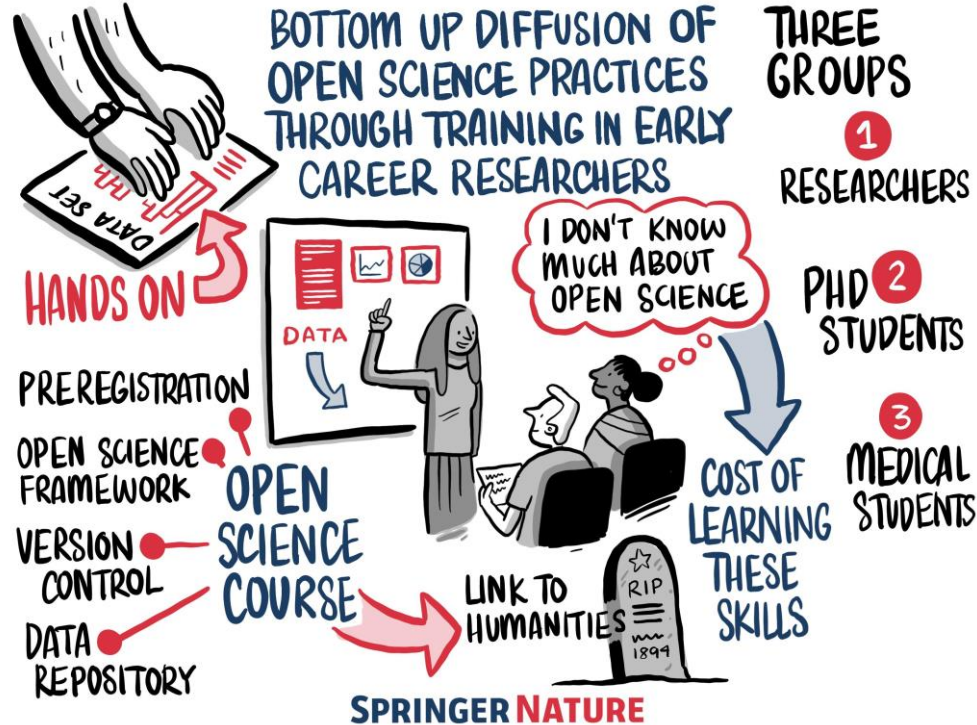
- Open identities
- Open reports
- Open participation



→ **Example:**
Public post-publication
peer review



How can we make Open Science work?



Summary - The Open Research Process

Study Design

Data Collection

Publication &
Distribution

Replication

Study Design

Data Collection

Publication &
Distribution

Replication

Preregistration

“The specification of a research design, hypotheses, and analysis plan prior to observing the outcomes of a study”



Nosek & Lindsay (2018)

Why?

- Prevent HARKing
- Reduce analytical flexibility
- Make selective reporting visible
- Get early feedback
- Take credit for your ideas
- Regulatory agencies require it

Study Design

Data Collection

Publication &
Distribution

Replication

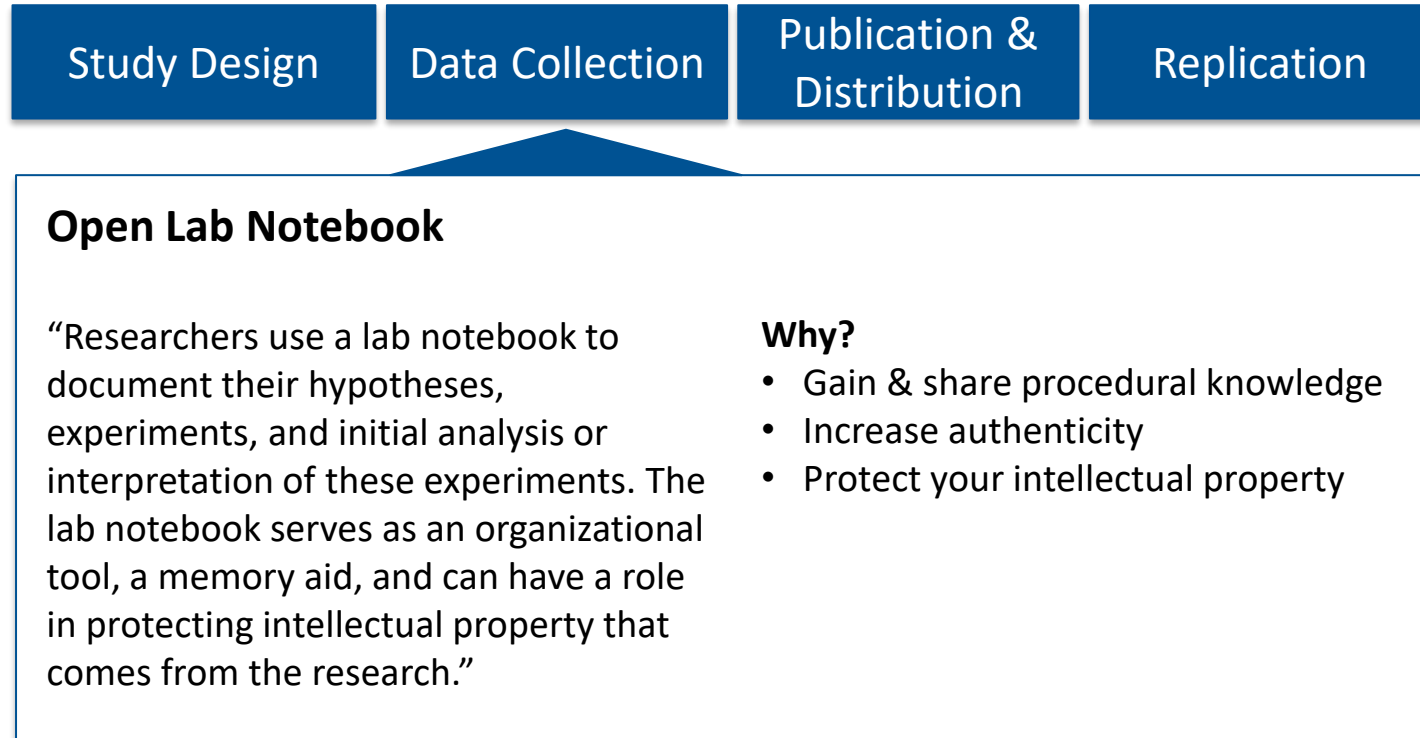
Apply for Registered Report

“Registered Reports are a form of empirical journal article in which methods and proposed analyses are pre-registered and peer-reviewed prior to research being conducted. High-quality protocols are then provisionally accepted for publication before data collection commences.”

Why?

- Advantages of preregistration
- Guaranteed publication independent of results
- Peer review for your design

The Open Research Process



The Open Research Process

Study Design

Data Collection

Publication &
Distribution

Replication

Open Data

“Open data should be available to everyone to access, use, and share.”



Why?

- Make your analyses reproducible
- Enable re-use of data for answering other research questions
- Never lose valuable data in a file drawer
- Funding agencies require it

Study Design

Data Collection

Publication &
Distribution

Replication

Open Materials

“Making components of the research methodology needed to reproduce the reported procedure and analysis publicly available.”



Why?

- Make your study reproducible
- Enable re-use of materials for other experiments

The Open Research Process



The Open Research Process

Study Design

Data Collection

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Replication

Open Analysis Code

“Clean, repeatable, script-based workflow [...] that links raw data through to clean data and to final analysis outputs.”

Why?

- Enable others to reproduce your analyses
- Understand your own code (after some time)
- Recreate your results with one click

The Open Research Process



Replication

“replication is a scientific method to verify research findings and [...] refers to a repetition of a research procedure to check the accuracy or truth of the findings reported.”

Why?

- Enhance credibility of your research
- Gain confidence in your findings & solidify the basis of your research

Food for thoughts: Open Science can save the planet (Kamila Markram)



Open science: Michael Nielsen at TEDxWaterloo



Solutions for better science in general

- Ten Principles of the Leiden Manifesto or DORA
- Coincidence
- Article Based Metrics
- Post Publication Reviews (open)
- Independent Thinking
- Look for Impact
- Replication studies
- Open Science

→ Pre-Registration

What is Preregistration?



Preregistration

Preregistration is the process of **specifying key study** and **analysis details and decisions** before conducting the experiment

The main goal of preregistering one's research is to make it easier to distinguish

- a) between what you set out to do (confirmation) and
- b) what was discovered along the way (exploration).



Confirmatory Research

Hypothesis testing

- Results are held to the highest standards
- Data-independent
- Minimizes false positives
- P-values retain diagnostic value
- Inferences may be drawn to wider population

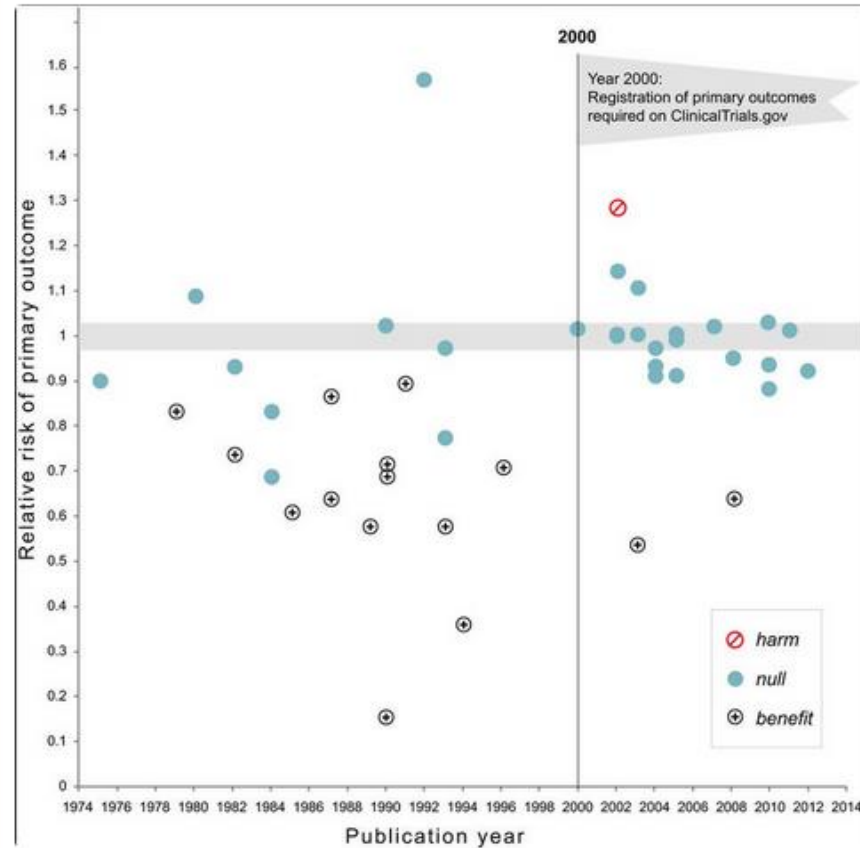
Exploratory Research

Hypothesis generating

- Results deserve to be replicated and confirmed
- Data-dependent
- Minimizes false negatives in order to find unexpected discoveries
- P-values lose diagnostic value
- Not useful for making inferences to any wider population

Preregistration allows the researcher to make a clear distinction between both modes of research.

Pre-registration causes drugs to stop working



Arguments against preregistration

1. Preregistration limits experimentation

However:

Preregistration does not forbid exploration!
It only makes transparent, what is planned.

2. Preregistration is very time-consuming

However:

It is.
**But you gain a peace of mind through it and at some point
you would need to do the work anyways.**

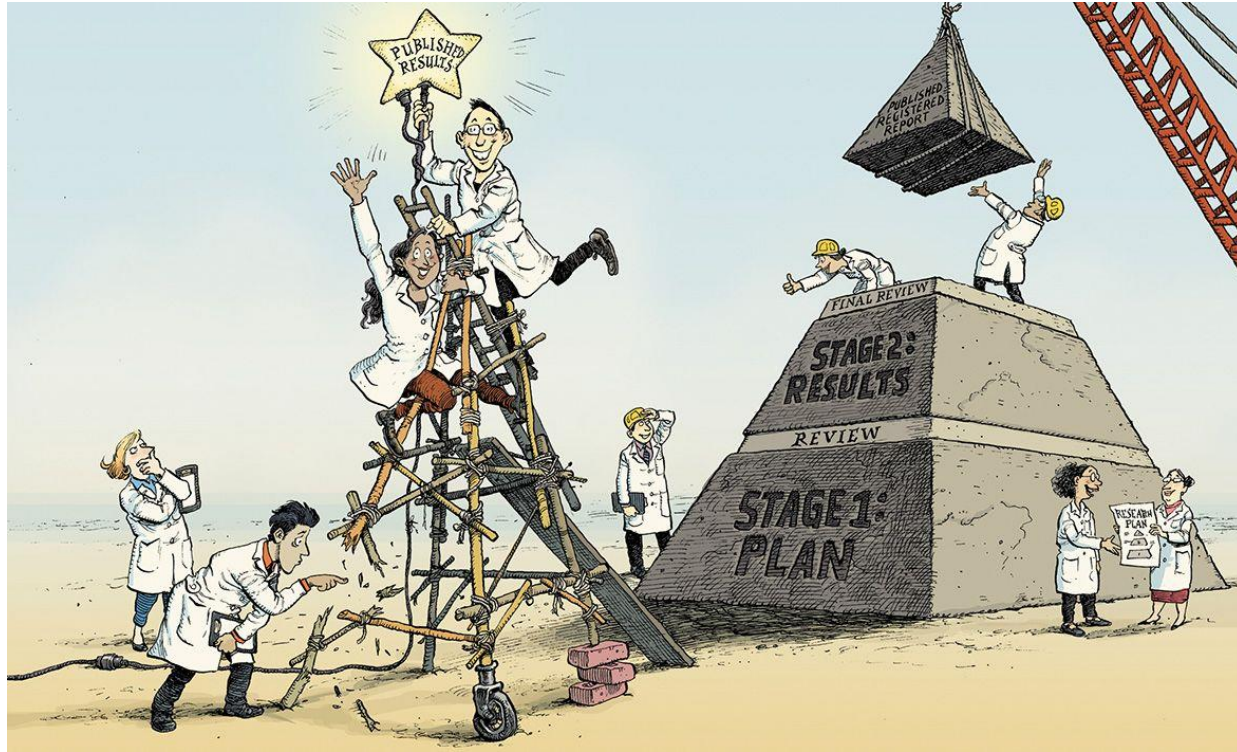
Registered Report

= a form of empirical article in which the introduction, methods, and proposed analyses are pre-registered and reviewed prior to research being conducted.

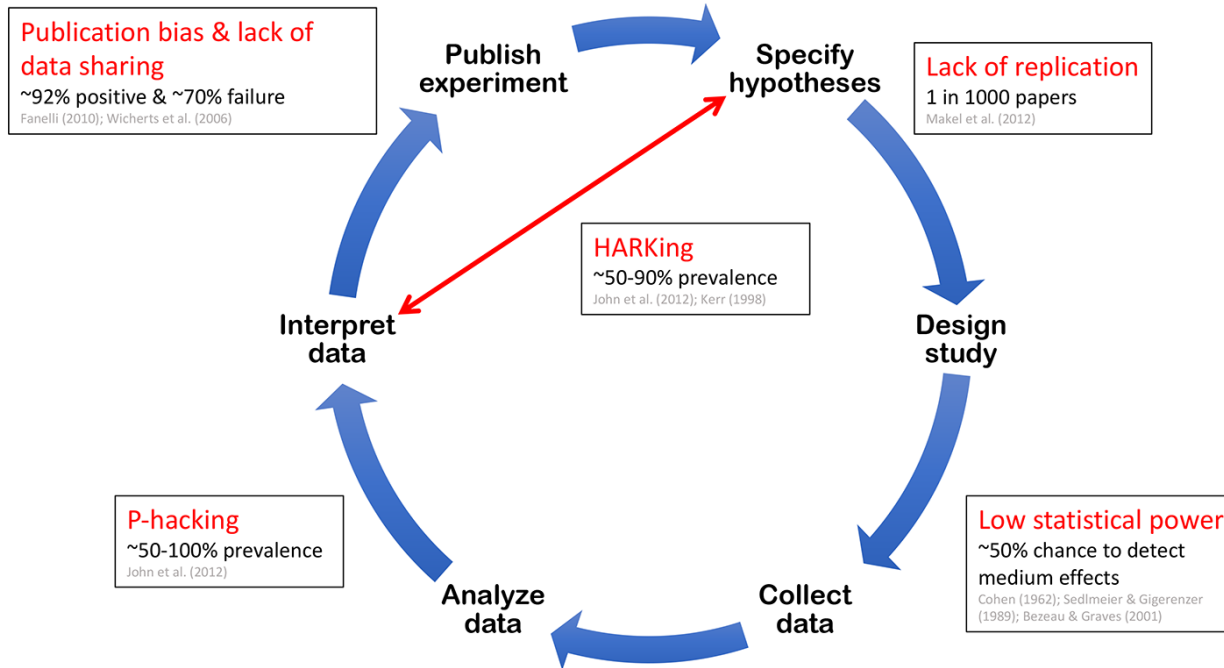
- High quality protocols are then provisionally accepted for publication, if the authors follow through with the registered methodology.



Registered Report – what it can prevent



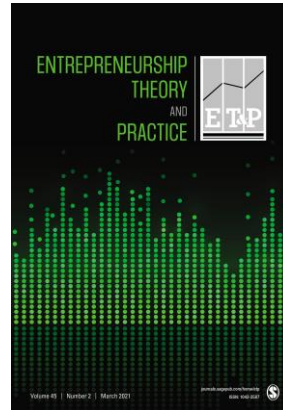
Registered Report – what it can prevent



Some leading journal examples



Replications



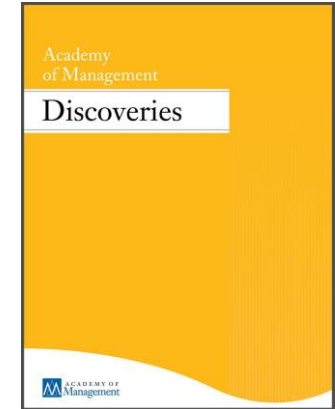
Replications

Preregistered reports



Replications

Abductive research



Replications

Preregistered reports

Abductive research

TABLE 1
Differences between Deduction, Induction, and Abduction

	Deductive Reasoning	Inductive Reasoning	Abductive Reasoning
Objective	<ul style="list-style-type: none"> - To demonstrate that if premises are true, it is impossible for the conclusion to be false - To demonstrate the situational validity of a generalizable rule or claim 	<ul style="list-style-type: none"> - To generate a knowledge claim where “it is improbable that the conclusion is false if the premises are true” (Hurley, 2000) - To demonstrate the probable generalizability of a situational reality 	<ul style="list-style-type: none"> - To generate plausible, conjecturable explanations - Discovery
Strength of knowledge claim	Strongest (certain)	Strong (probable)	Weak (plausible)
Role of theory	Provides <i>a priori</i> explanations (hypotheses) to be challenged empirically	Provides a guiding framework and systematic approach to generate a generalizable explanation from the data	Provides assumptions to be challenged and frames anomalies to be explored and suggests the variables on which to sample
How data are used	<ul style="list-style-type: none"> - To disconfirm the null - To disconfirm alternatives 	To confirm a generalizable outcome when premises are met	<ul style="list-style-type: none"> - To describe phenomena - To elicit tentative claims - To narrow range of possible explanations
Type of reasoning and how used	Necessary reasoning Used to test falsifiability of presumed means-ends linkages	Probabilistic reasoning Used to demonstrate generalizable means-ends linkages or processes	Contrastive reasoning Used to identify patterns indicative of alternative dynamics, processes, mechanisms, or means-ends linkages

Primary sources: Campos (2011), Folger and Stein (2017), Okhuysen and Behfar (2017).