

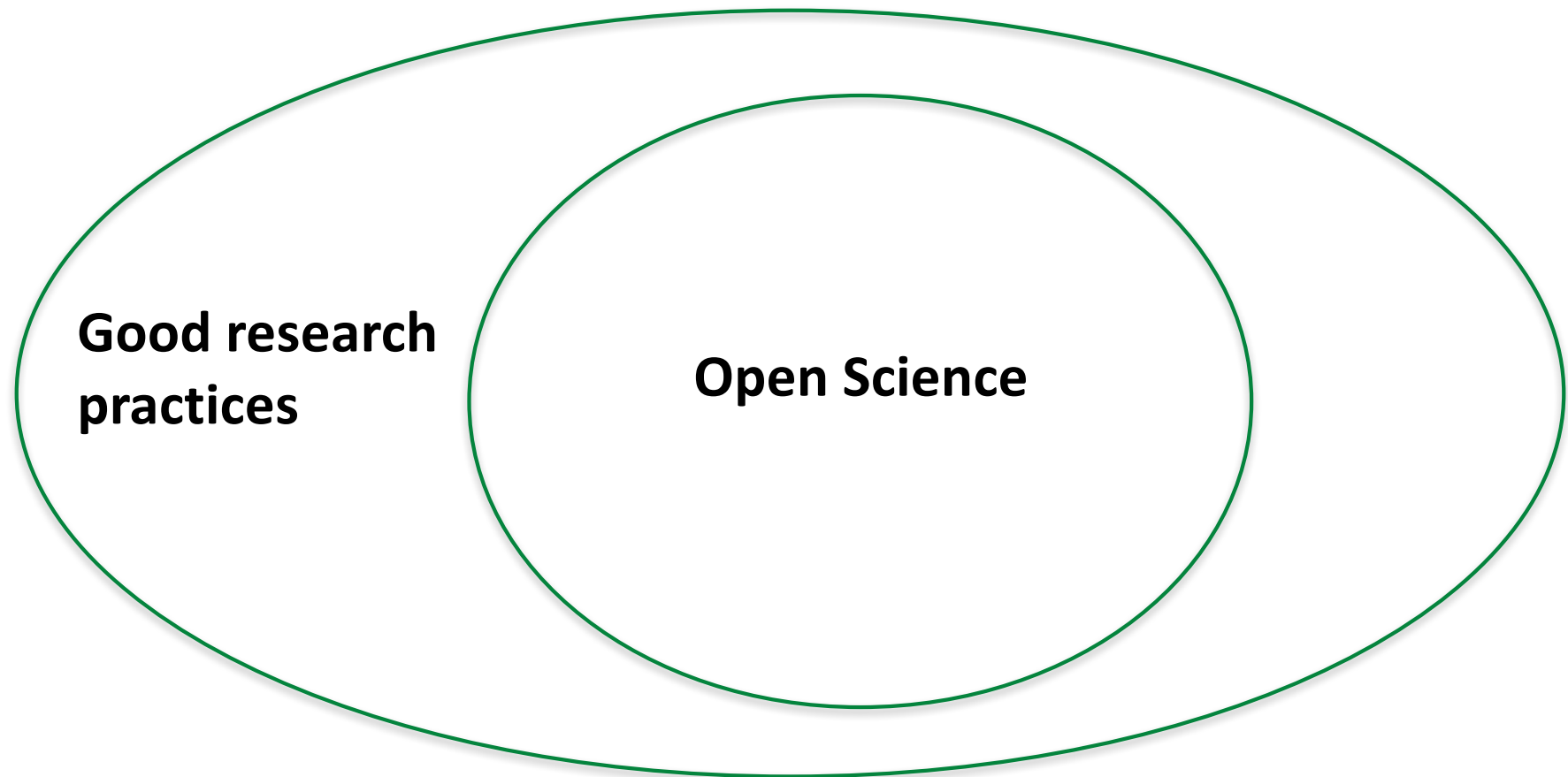
# Open Science – Introduction

Open Science Workshop

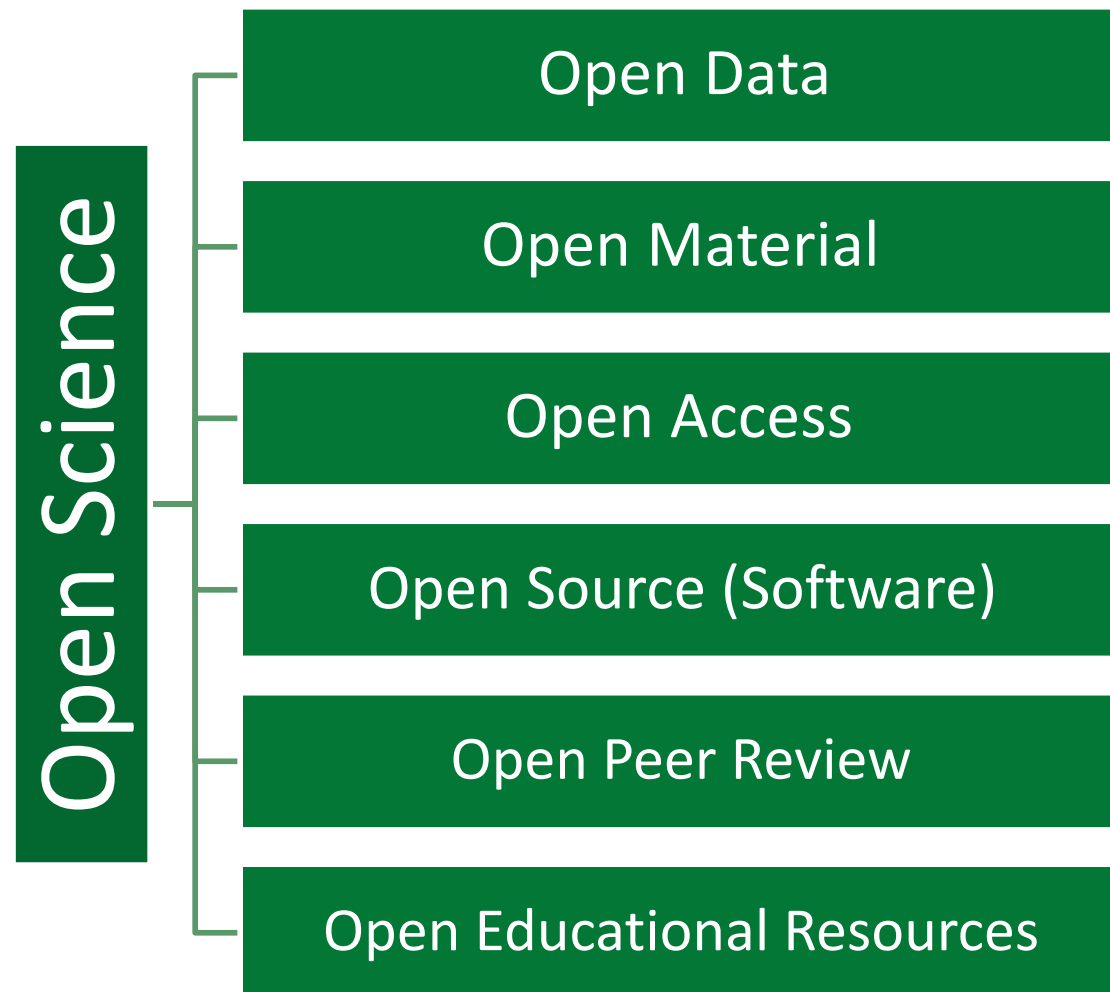


more information available on the last slide

# Open Science € Good Science

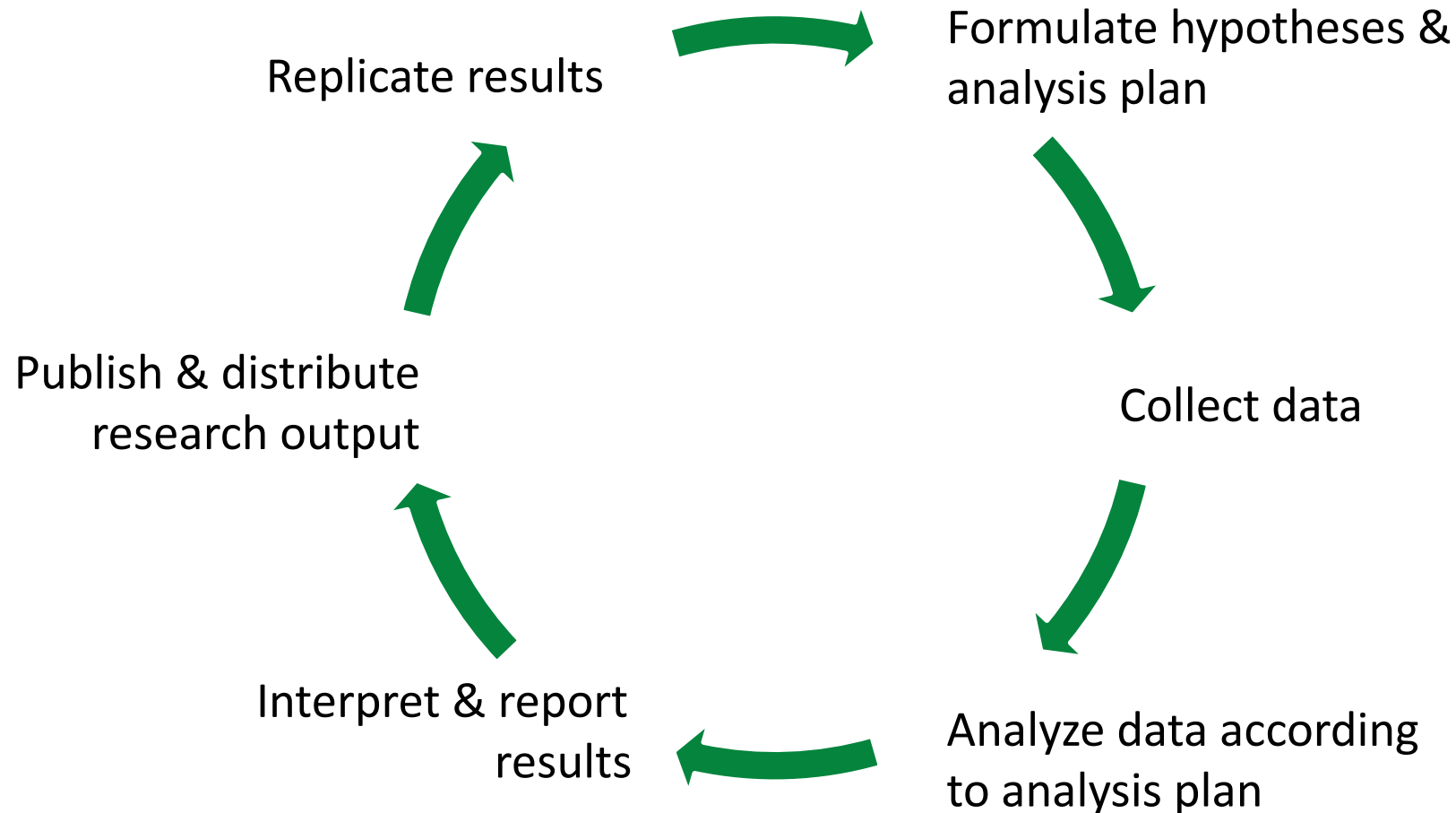


# Pillars of Open Science



# Why transparency?

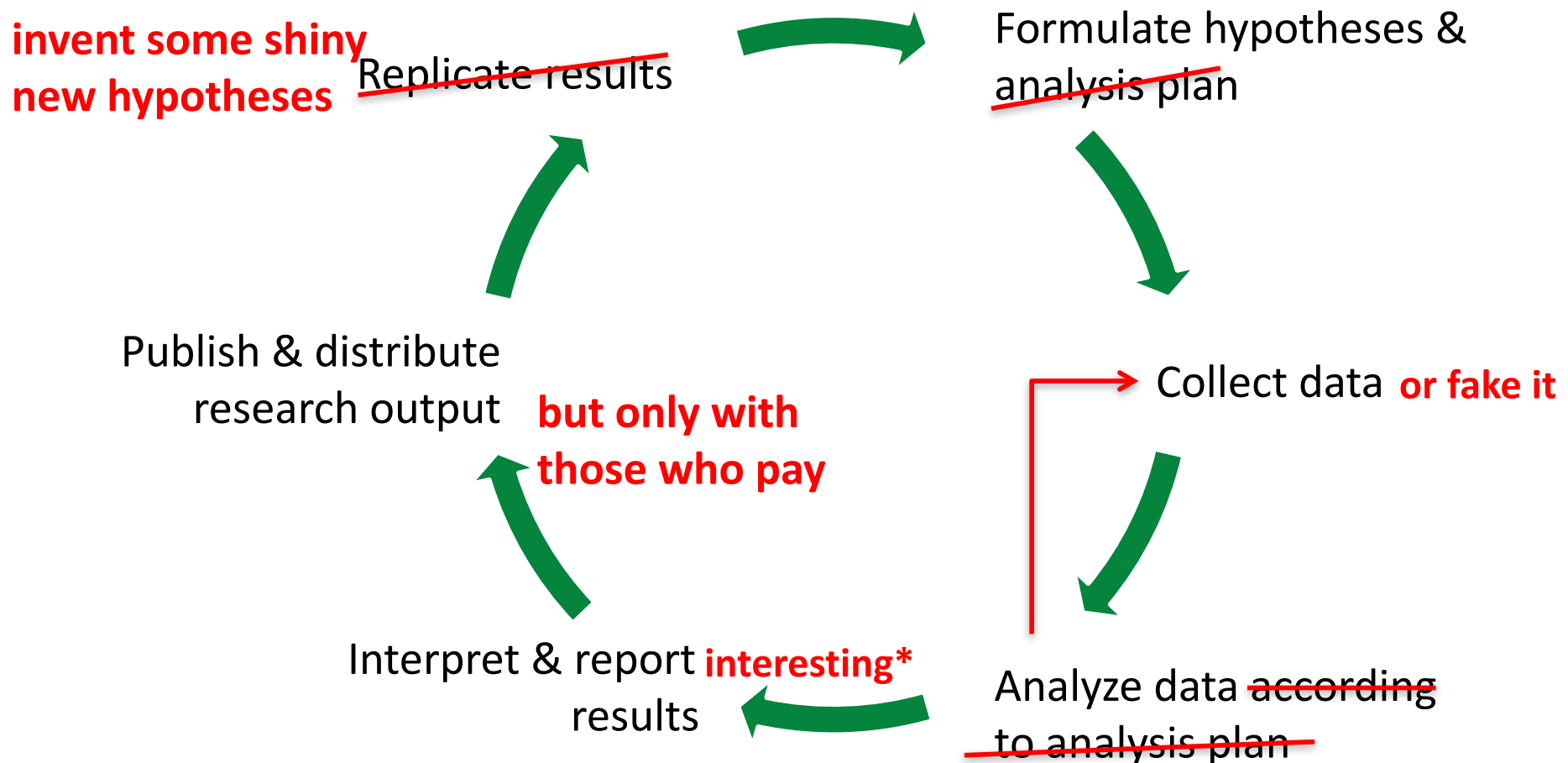
## The Confirmatory Research Process





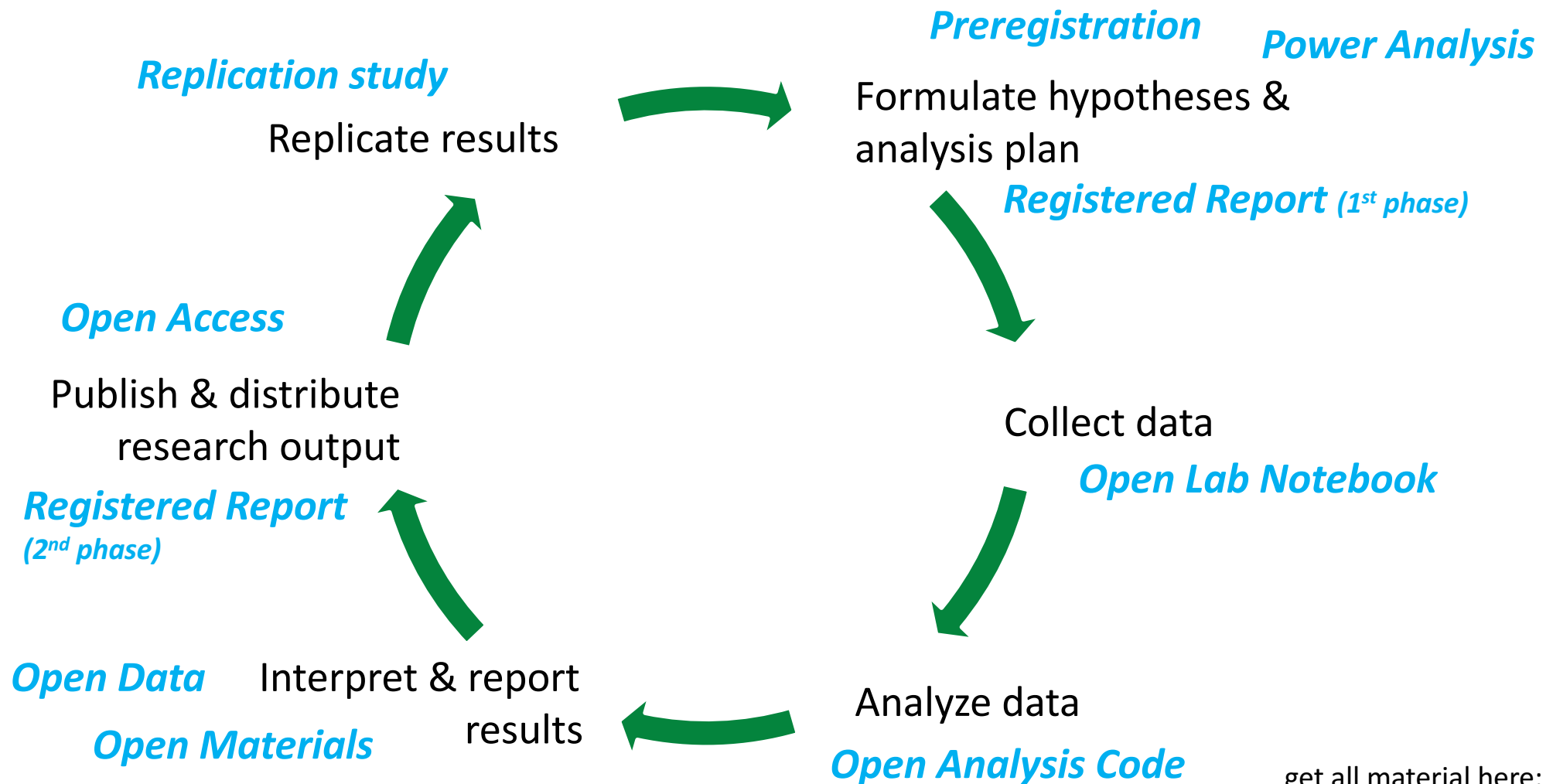
# Why transparency?

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



# Why transparency?



Increase trust in science, don't waste public resources



Get constructive feedback



Be international and inclusive



Increase the speed of discovery

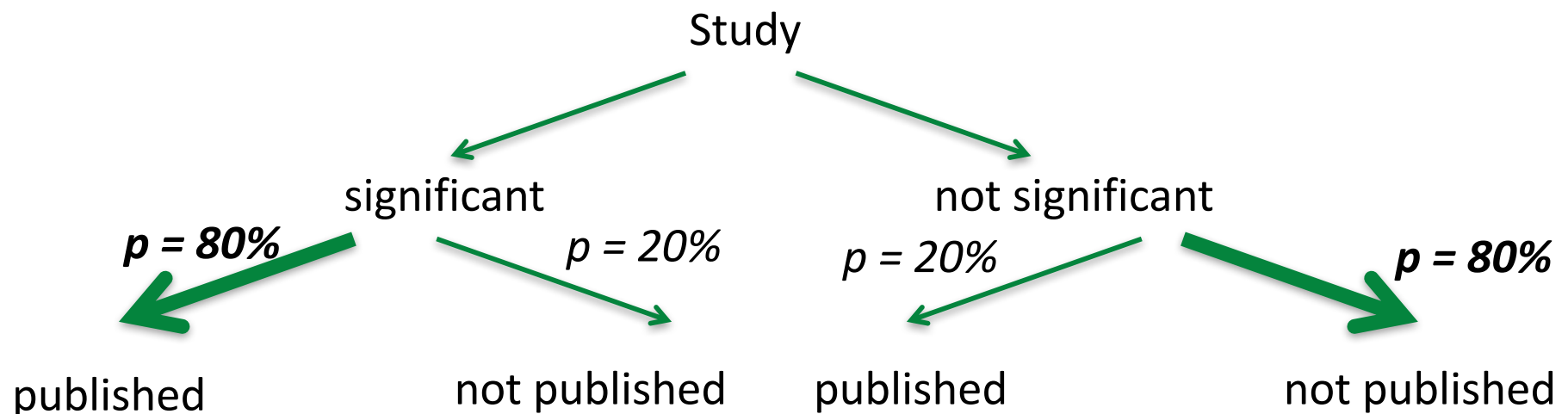
However...

The present situation is not so bright...

# Publication Bias

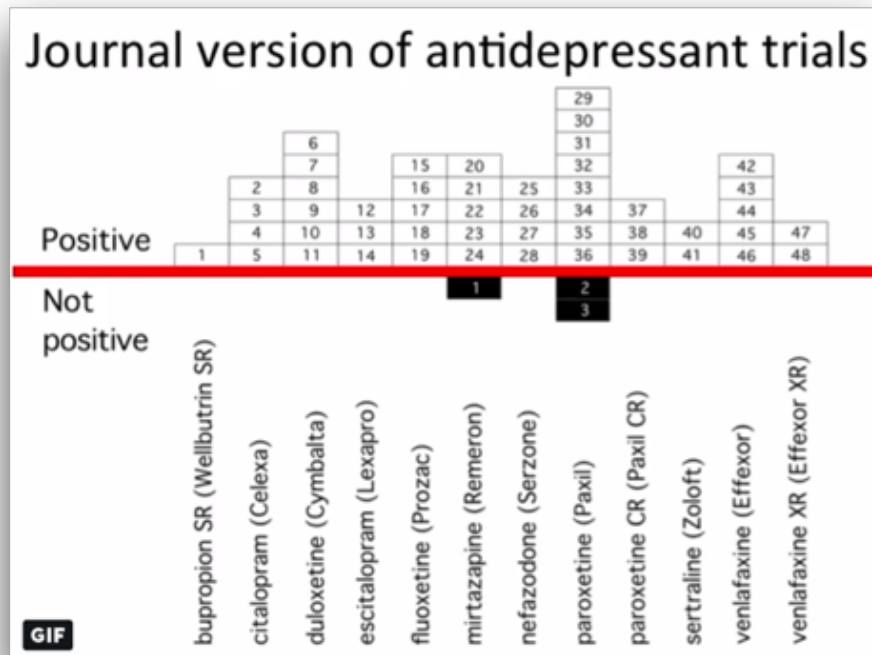
## Definition:

“The phenomenon in which studies with positive results are more likely to be published than studies with negative results.”

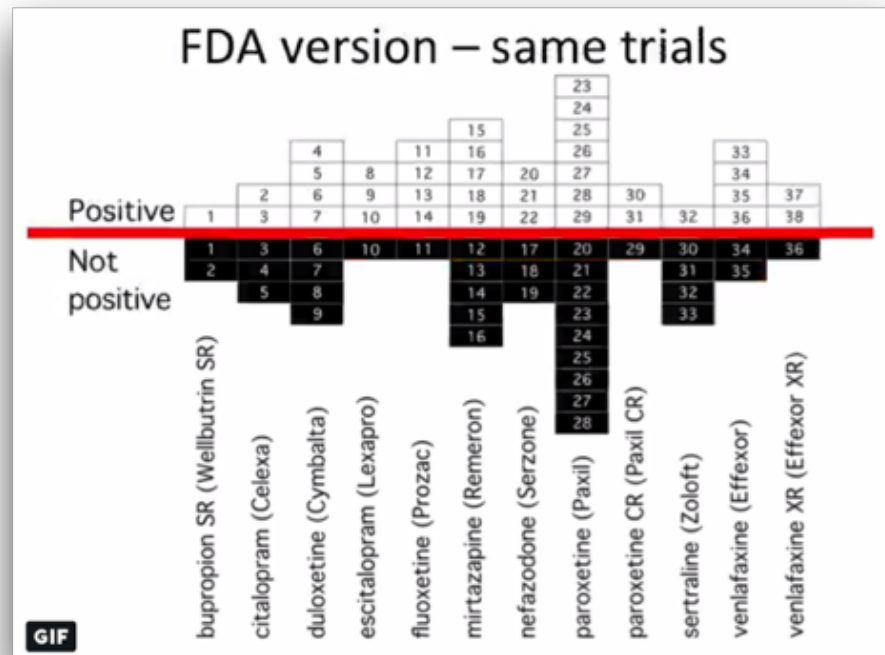


# Publication Bias: Efficacy of anti-depressants (Turner et al. 2008; Meta-Analysis with $k = 74$ )

Trials published in journals:  
48 positive, 3 negative

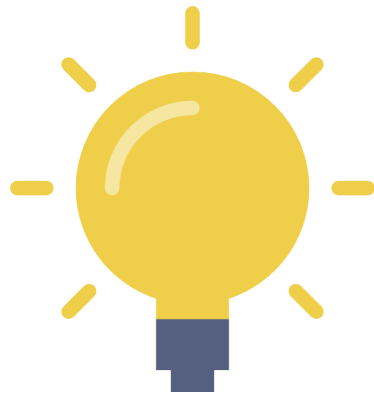


Trials registered at FDA:  
38 positive, 36 negative

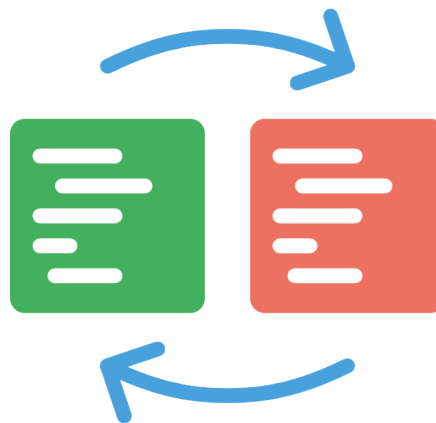


# Questionable Research Practices

## Scientific Misconduct



**Fabrication**  
(Making up data)



**Falsification**  
(Distorting data)



**Questionable  
Research Practices**  
(p-hacking, HARKing,  
selective reporting...)

# Questionable Research Practices

***p*-hacking** (*n.*). Tune your data analysis in a way that you achieve a significant *p*-value in situations where it would have been non-significant.

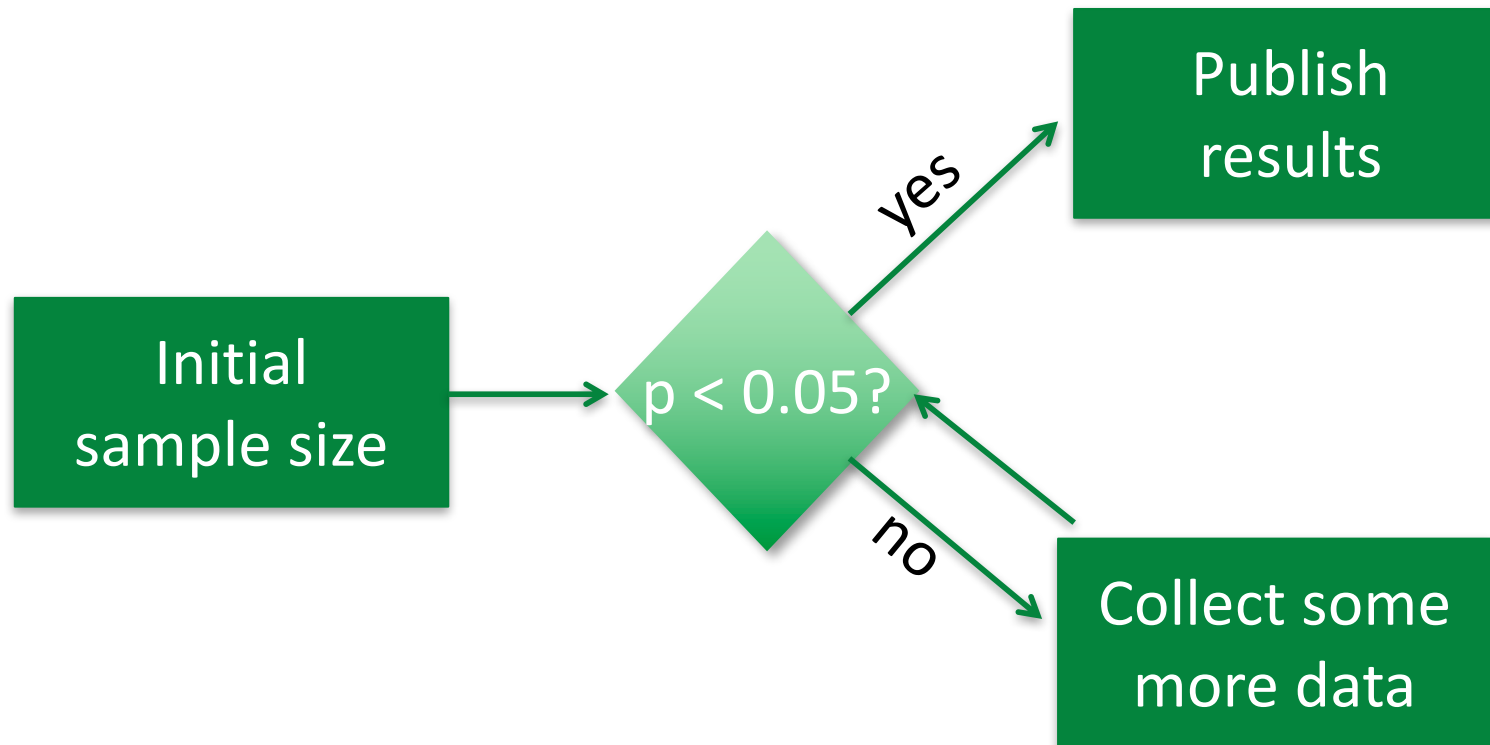
**Questionable research practices (QRPs)** (*n.*). Practices of data collection and data analysis that are not outright fraud, but also not really kosher.





# Tools for $p$ -Hacking

## Optional Stopping



# Tools for $p$ -Hacking

## Optional Stopping

### Repeated Significance Tests on Accumulating Data

By P. ARMITAGE, C. K. MCPHERSON and B. C. ROWE

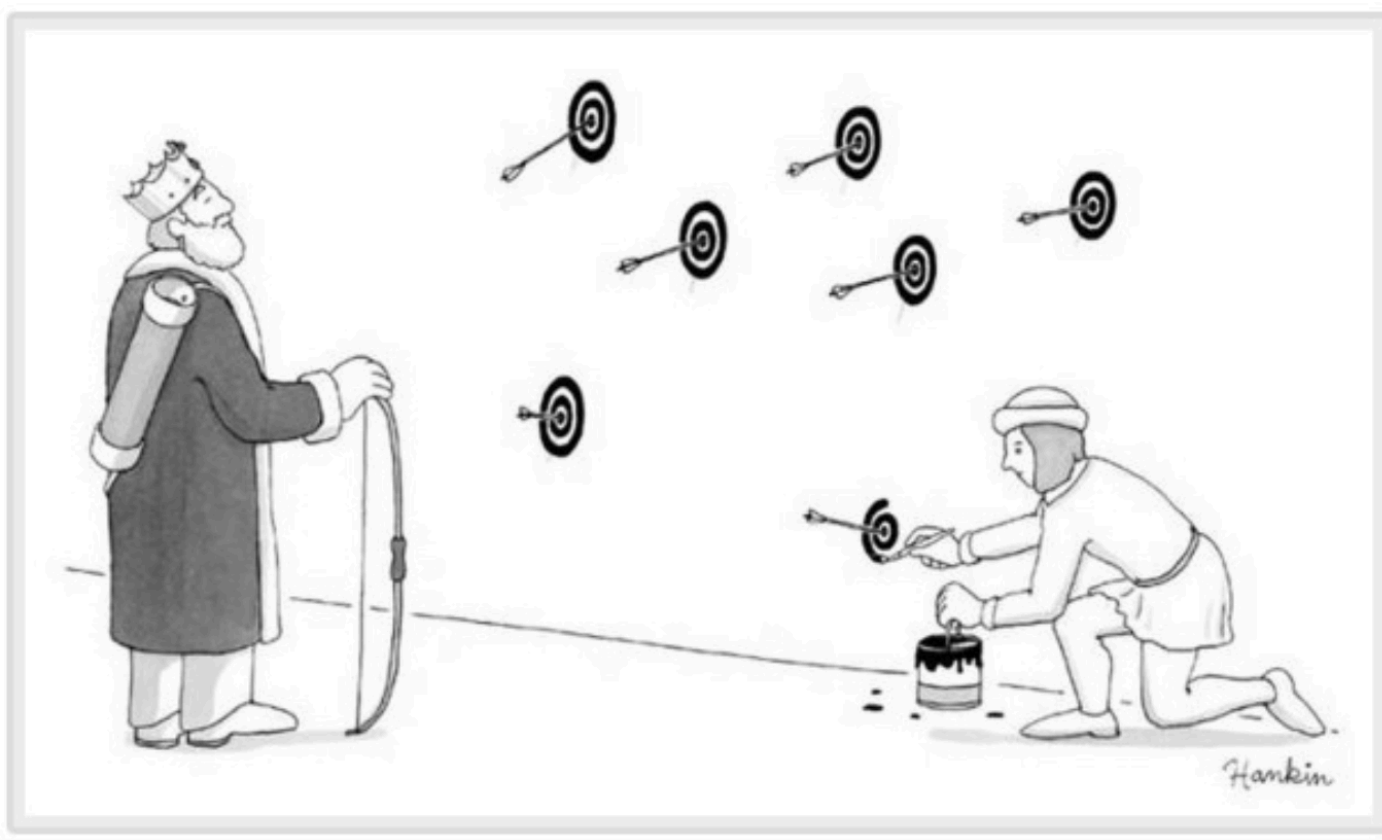
$n$	$2\alpha$ $k$		0.10 1.645		0.05 1.960		0.02 2.326		0.01 2.576	
	$Q$	$S$	$Q$	$S$	$Q$	$S$	$Q$	$S$	$Q$	$S$
1	0.10000	0.0970	0.05000	0.0545	0.02000	0.0200	0.01000	0.0100	0.00500	0.0050
2	0.16015	0.1650	0.08312	0.0885	0.03453	0.0345	0.01726	0.0173	0.00863	0.0086
3	0.20207	0.1980	0.10726	0.1115	0.04561	0.0456	0.02280	0.0228	0.01140	0.0114
4	0.23399	0.2295	0.12617	0.1260	0.05454	0.0545	0.02760	0.0276	0.01380	0.0138
5	0.25963	0.2590	0.14169	0.1420	0.06201	0.0620	0.03120	0.0312	0.01560	0.0156
160	0.63315		0.40829		0.20834		0.10417		0.05208	
180	0.64301		0.41677		0.21359		0.10729		0.05364	
200	0.65165		0.42429		0.21828		0.11014		0.05520	
250	0.670		0.440		0.228		0.11720		0.05840	
500	0.720		0.487		0.259		0.132		0.066	
750	0.746		0.513		0.276		0.144		0.072	
1,000	0.763		0.530		0.288		0.152		0.076	

With long enough sampling and optional stopping, it is **guaranteed** to get a significant result!

100%

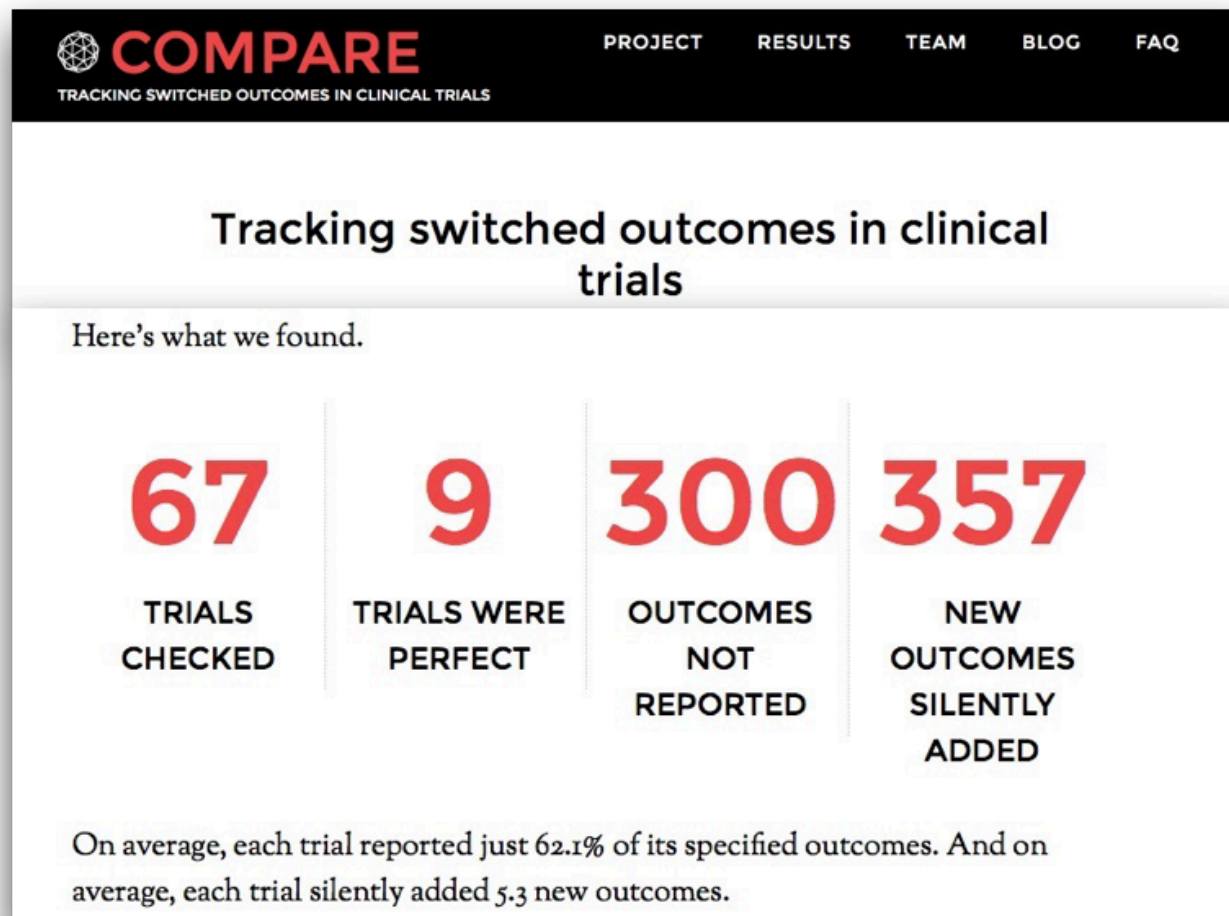
# Tools for $p$ -Hacking

**HARKing: Hypothesizing after the results are known**



# Tools for $p$ -Hacking

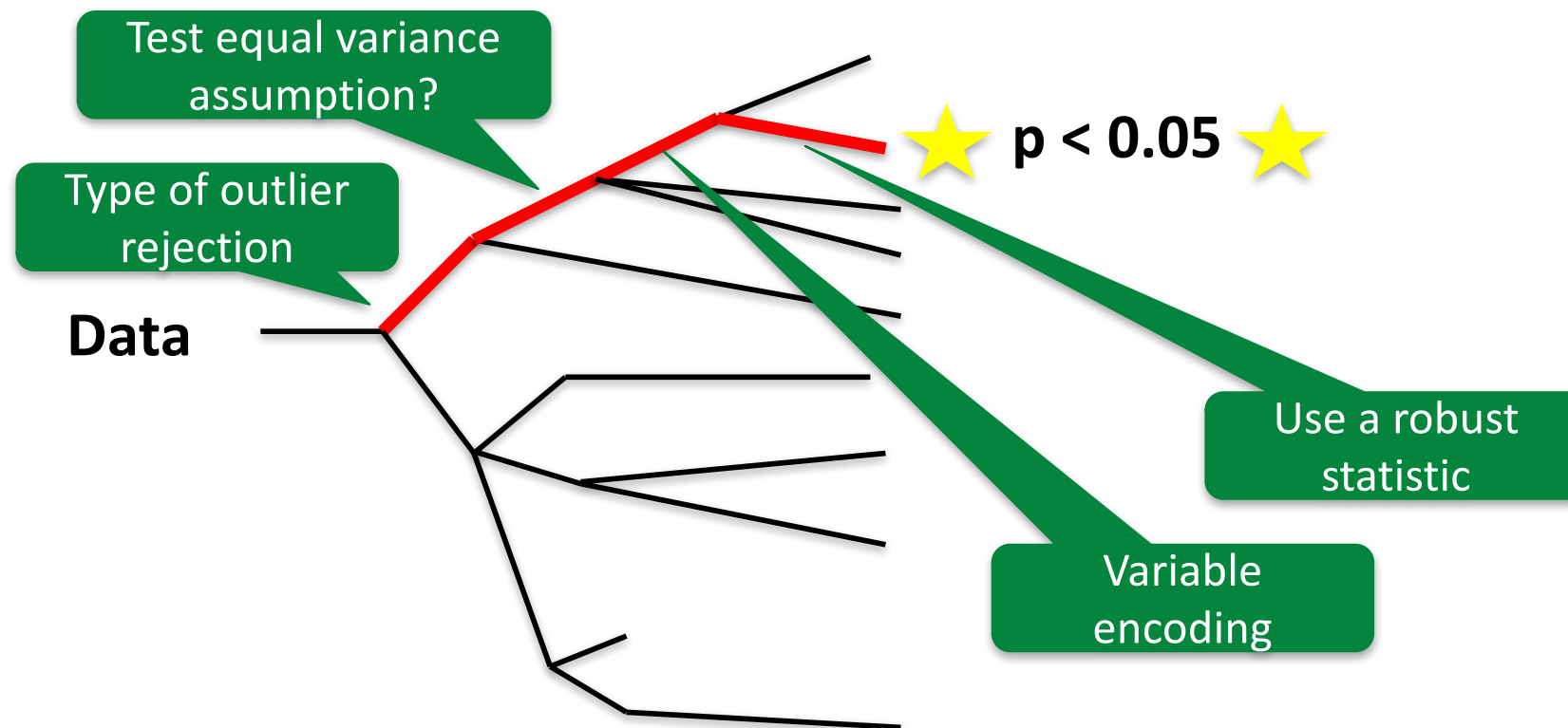
## Outcome switching



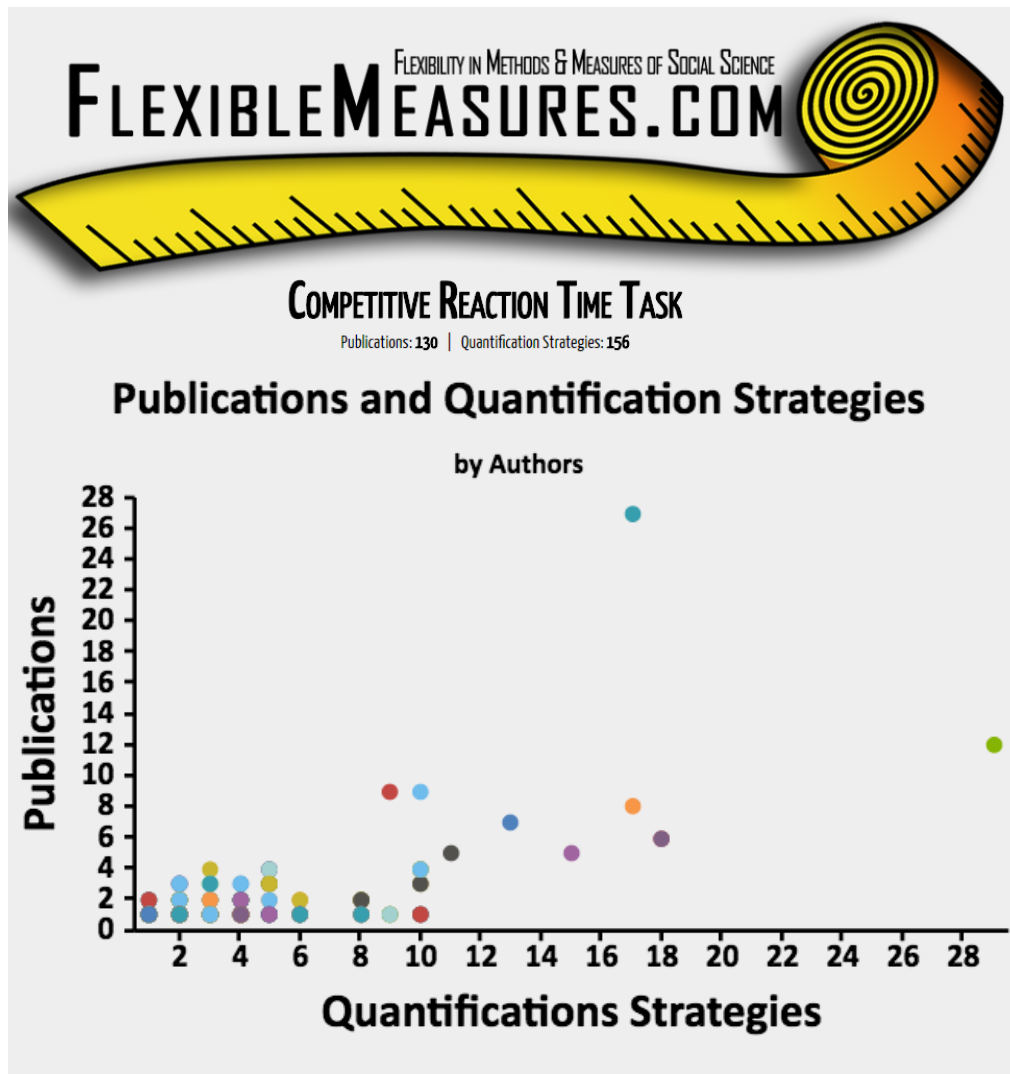
- 2 outcome variables:  
false positive rate  
5% → 9.5%
- 5 outcome variables  
with one-sided  
testing:  
false positive rate  
5% → 41%

# Tools for *p*-Hacking

## The garden of forking paths / Analytical degrees of freedom



# Tools for $p$ -Hacking



There are currently 130 publications in which results are based on the Competitive Reaction Time Task, and they reported 156 different quantification strategies in total!

# P-Hacking

## **Intentional?**

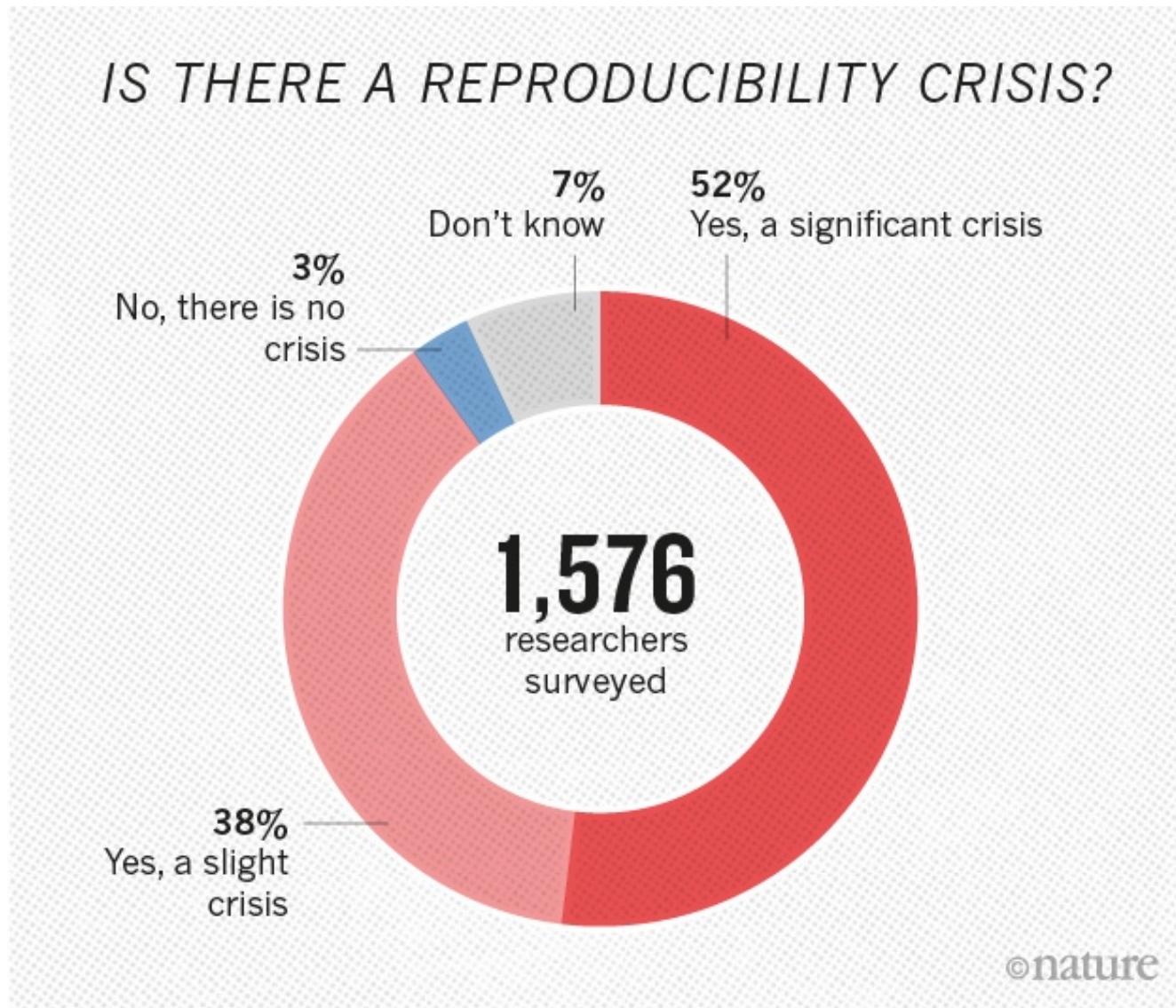
- Evil researcher who only cares about his/her career and not at all about truth-seeking?

## **Unintentional?**

- Wrong education?
- Wrong/uncritical standards of the field?
- Pushed by supervisors, reviewers, or editors?
- → Distorting effects on the published record are probably comparable, but the ethical evaluations differs strongly.



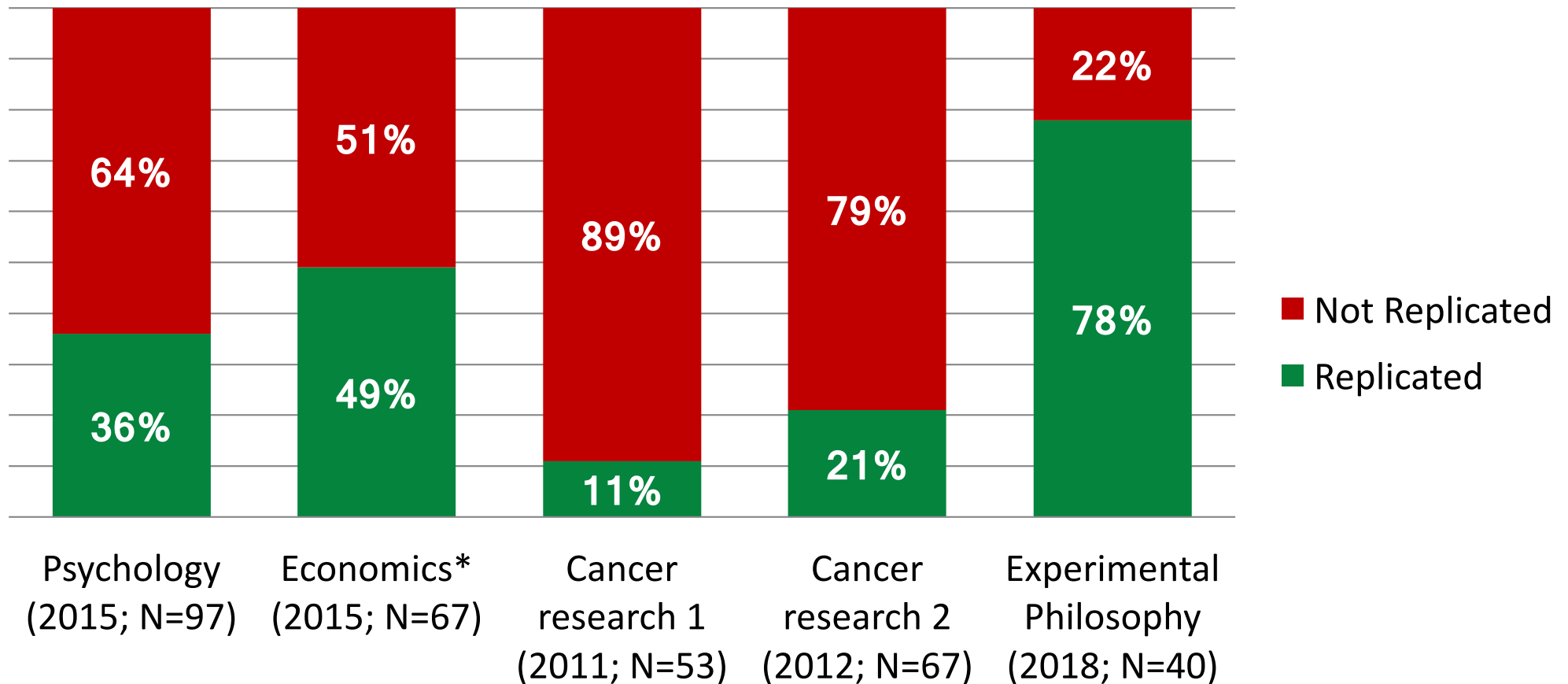
# We might have a reproducibility crisis



**90% YES**  
(there is a crisis)



# We might have 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.

# 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**

# The Open Research Process

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

# The Open Research Process

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.”

COS(2018): [www.cos.io/rr/](http://www.cos.io/rr/)

### Why?

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

# The Open Research Process

Study Design

Data Collection

Publication &  
Distribution

Replication

## Open Lab Notebook

“Researchers use a lab notebook to document their hypotheses, experiments, and initial analysis or interpretation of these experiments. The lab notebook serves as an organizational tool, a memory aid, and can have a role in protecting intellectual property that comes from the research.”

Goyal, Malviya, & Kapoor (2012)

### Why?

- Gain & share procedural knowledge
- Increase authenticity
- Protect your intellectual property

# 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.”

GO FAIR (2018)



### 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

# The Open Research Process

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.”



OSF (2016)

### Why?

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

# The Open Research Process

Study Design

Data Collection

Publication &  
Distribution

Replication

## Open Access

Literature which is “digital, online, free of charge, and free of most copyright and licensing restrictions”.

Suber (2015)



### Why?

- Enable faster progress in research by opening the access to knowledge
- Give back value to the community that funded you and not only to publishers



# The Open Research Process

Study Design

Data Collection

Publication &  
Distribution

Replication

## Open Analysis Code

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

British Ecological Society (2017)

### Why?

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

# The Open Research Process

Study Design

Data Collection

Publication &  
Distribution

Replication

## 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.”

Schmidt (2009)

### Why?

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

# Let's not rest on our laurels: Current challenges.

1. Blind spots
2. High openness, low quality
3. Empirical evidence for effectiveness of reforms
4. Incentive structures

# Let's not rest on our laurels: Current challenges.

## **1. Blind spots**

2. High openness, low quality
3. Empirical evidence for effectiveness of reforms
4. Incentive structures

# Leave your bubble!

- Blind spots within psychology?
- Project: Analyse conference programs of the subsections of the DGPs („Fachgruppen“) for keywords such as *replication*, *reproducibility*, *open science*.
- Leave your bubble and spread the word to your community.

# Let's not rest on our laurels: Current challenges.

1. Blind spots
- 2. High openness, low quality**
3. Empirical evidence for effectiveness of reforms
4. Incentive structures

# Paper does not match preregistration



**Sanjay Srivastava**

@hardsci

Folge ich

This week's homework in my grad open science seminar was to find a preregistered article, compare it to the preregistration, and write a reaction paper. Here are representative quotes from students' reax papers

 Tweet übersetzen

"The article that was published as being "preregistered" was actually completed before the preregistration was submitted. They reported the methods in terms of the experimental procedures, numbers of subjects etc. However, no analyses were preregistered... I don't really understand why this can be considered a preregistration."

"I realized that they had only preregistered one of several main questions that they were reporting on. I was appalled to see that the paper had no mention of preregistering the other components of the study, including their major finding..."

"The pre-registration left it unclear when and how the decision to conduct study X was made in relation to the other studies. Further, the power analysis and sample size justification presented in the pre-registration differed from [the sample size] presented in the paper... Although I anticipated a pre-registration would increase my faith in the credibility of the resulting paper, I believe this example had the opposite effect."

# Open-washing

„However, the analysis plan was posted to OSF but unfortunately not actually registered“

„Also, the Study 3 design was part of the registration, but it did not include an analysis plan.“

→ half-way preregistration?

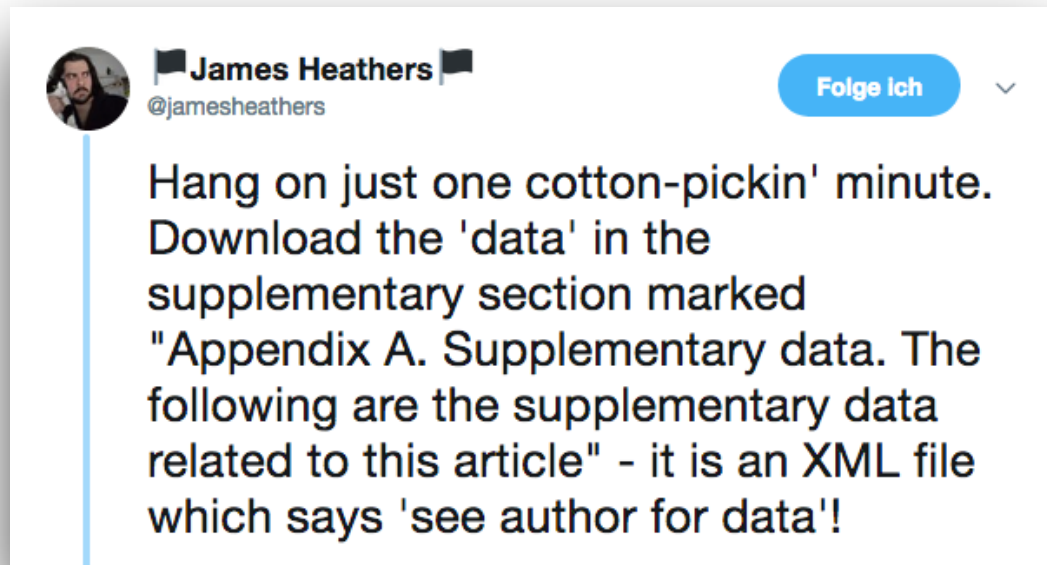
„One of the five studies (Study 3) was preregistered“

→ Preregister one trivial research question, get the badge for the whole paper?





# Open-washing



```
1 <?xml version="1.0" encoding="UTF-8" standalone="no"?>
2   <data:data>
3     <datasets>
4       <author>moldham@liverpool.ac.uk</author>
5       <type>0</type>
6       <dataset>
7         <value>1</value>
8         <reason>Data will be made available on request
9         </reason>
9       <comments/>
10    </dataset>
```

# Who is responsible for checking/enforcing the badges?

- **„Self-disclosure model“:**

Authors sign the statement „I have a preregistration and my paper matches the prereg“, but verification is left to community (in post-publication peer review)

→ badge means: „This is verifiable *in principle*“  
(but somebody still has to do it)

- **„verification model“:**

Reviewers and or editors do the verification

→ badge means: „This has been verified and can be trusted“  
(but extra burden for reviewers and editors)

- Registered Reports as a much better model? Preregistration is the paper, no mismatch possible. Reviewers check it during stage 1 review.

# FAIR data

- **F**indable: Metadata and data should be easy to find for both humans and computers.
- **A**ccessible: Once the user finds the required data, she/he needs to know how can they be accessed, possibly including authentication and authorisation.
- **I**nteroperable: The data usually need to be integrated with other data. In addition, the data need to interoperate with applications or workflows for analysis, storage, and processing.
- **R**eusable: Metadata and data should be well-described so that they can be replicated and/or combined in different settings.

# Open Data vs. FAIR data

- FAIR data can be *not* open
  - e.g., if a data set is findable, reuseable, etc., but only accessible within a closed research group
- Open Data can be *not* FAIR
  - e.g., an undocumented data dump in an uncurated repository, such as OSF, which is neither findable, nor reuseable, nor interoperable
- FAIR dimensions are quality criteria that can be applied to data sets. Ideally, a data set is open *and* FAIR.

Open-Washing =  
Hey, let's game the new system!

Endorse open science on Twitter and your CV,  
try to get badges with minimum effort,  
pretend openness but do not deliver.

# Let's not rest on our laurels: Current challenges.

1. Blind spots
2. High openness, low quality
- 3. Empirical evidence for effectiveness of reforms**
4. Incentive structures

# Meta-Science: Gather empirical evidence

- Hypothesis: Open science practices increase the credibility, the veridicality\*, and the replicability of research.
- A critic could say:  
*Where is the empirical evidence? You rush implementing all these interventions and reforms without having any evidence that they actually have the desired effect.*

\* the degree to which a theory or interpretation accurately represents reality

# Meta-Science: Gather empirical evidence

Rowhani-Farid *et al.* *Research Integrity and Peer Review* (2017) 2:4  
DOI 10.1186/s41073-017-0028-9

Research Integrity and  
Peer Review

REVIEW

Open Access

## What incentives increase data sharing in health and medical research? A systematic review



Anisa Rowhani-Farid\* , Michelle Allen and Adrian G. Barnett

**Results:** Only one incentive (using open data badges) has been tested in health and medical research that examined data sharing rates. The number of opinion pieces ( $n = 85$ ) out-weighed the number of article-testing strategies ( $n = 76$ ), and the number of observational studies exceeded them both ( $n = 106$ ).

**Conclusions:** Given that data is the foundation of evidence-based health and medical research, it is paradoxical that there is only one evidence-based incentive to promote data sharing. More well-designed studies are needed in order to increase the currently low rates of data sharing.



# Let's not rest on our laurels: Current challenges.

1. Blind spots
2. High openness, low quality
3. Empirical evidence for effectiveness of reforms
4. **Incentive structures**



Richard Horton,  
Editor von *The Lancet*

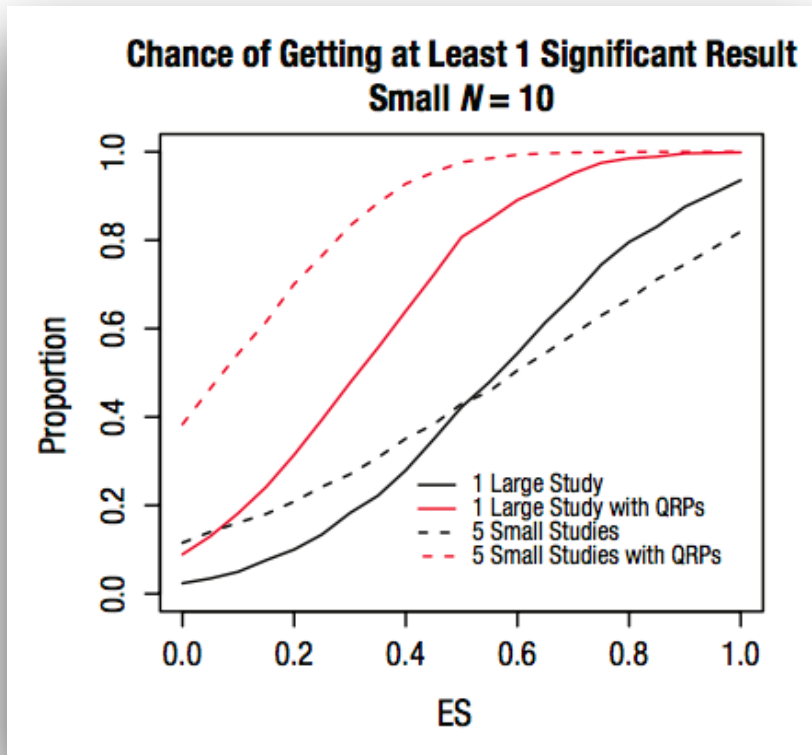
Much of the scientific literature,  
perhaps half, may simply be untrue.

Part of the problem is that no one is  
incentivised to be **right**.

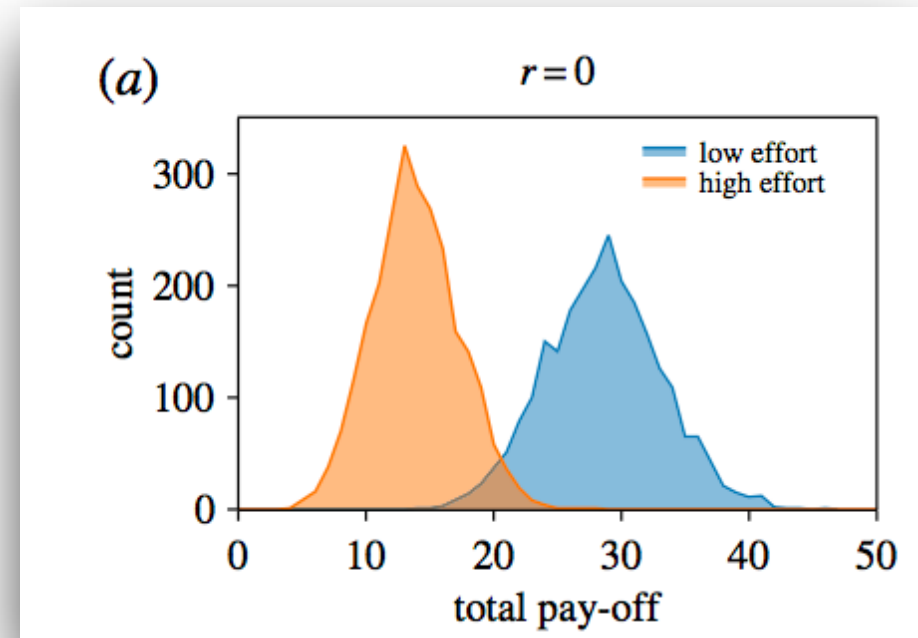
# Quantity, not quality

Actual (not desired) relevance at professorship hiring committees:	Rank
<b>Number</b> of peer-reviewed publications	1
Fit of research profile to the advertising institution	2
Quality of research talk	3
<b>Number</b> of publications	4
<b>Volume</b> of acquired third-party funding	5
<b>Number</b> of first authorships	6
...	...

„The rules of the game“



„Evolution of bad science“



Ideal strategy for a high quantity of publications:

small  $n$  + many studies + questionable research practices (QRPs), such as  $p$ -hacking

Bakker, M., van Dijk, A., & Wicherts, J. M. (2012). The Rules of the Game Called Psychological Science. *Perspectives on Psychological Science*, 7(6), 543–554. <http://doi.org/10.1177/1745691612459060>

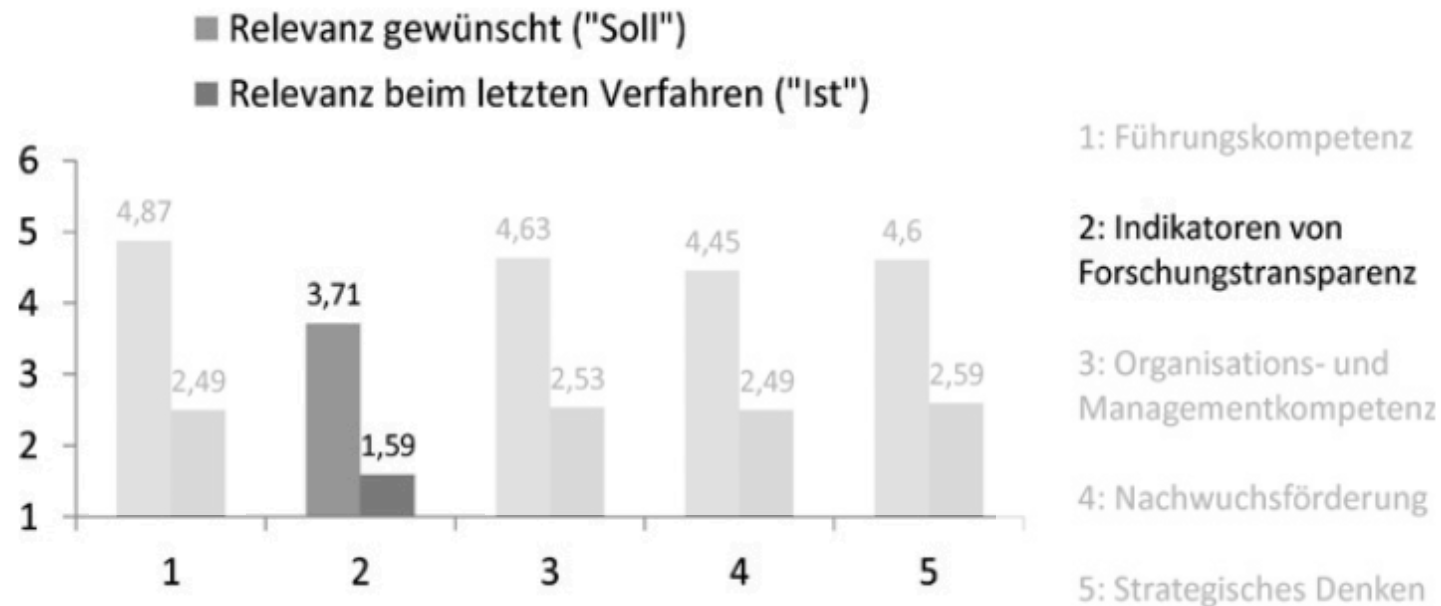
Smaldino, P. E., & McElreath, R. (2016). The natural selection of bad science. *Royal Society Open Science*, 3(9), 160384–17. <http://doi.org/10.1098/rsos.160384>

# Quantity, not quality

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<b>Number</b> of publications	4
<b>Volume</b> of acquired third-party funding	5
<b>Number</b> of first authorships	6
...	...
<b>Quality assessment</b> of the best three publications	17
...	...
<b>Indicators of research transparency</b>	<b>41 (of 41)</b>

# Quality, not quantity

Kriterien mit der größten Diskrepanz zwischen  
„Soll“ und „Ist“



# Roadmap

# Fast adoption vs. High (FAIR) quality?


- Low hurdles, one small step at a time
- Reward small steps  
*Sharing something - even badly documented data - is better than sharing nothing.*
- Learning by doing  
*With increasing practice, hopefully the quality gets better, too.*
- But: (Initially) Low quality  
*Barely reusable data sets; trying to reproduce a result is a pain in the ass or impossible; data reuse very limited.*
- Risk of „open-washing“  
*Pretending openness without actual value.*



- High hurdles  
*Mainly enthusiasts/computer scientists will be able and motivated to use it*
- Reward big steps  
*Curated repositories with input quality control.*
- Instant high quality  
*The data sets which are open are instantly FAIR.*



# Hiring committees: Make „open science“ a desirable or essential job characteristic



An der Fakultät für Psychologie und Pädagogik der Ludwig-Maximilians-Universität München ist zum Wintersemester 2016/2017 eine

**Professur (W3) für Sozialpsychologie (Lehrstuhl)**

...

Das Department Psychologie legt Wert auf transparente und replizierbare Forschung und unterstützt diese Ziele durch Open Data, Open Material und Präregistrierungen. Bewerber/innen werden daher gebeten, in ihrem Anschreiben darzulegen, auf welche Art und Weise sie diese Ziele bereits verfolgt haben und in Zukunft verfolgen möchten.


Since 2015: All professorship job descriptions use this requirement

**Full Professor (W3)  
of Social Psychology**

to be filled as soon as possible.

...

The Department of Psychology aims for transparent and reproducible research (including Open Data, Open Materials, and Preregistrations). Applicants are asked to illustrate how they have pursued these goals in the past and/or how they plan to do so in the future.

 **Ulrich Dirnagl**  
@dirnagl Folge ich

If you are applying for a professorship at the Charite you now need to tell us about your contributions to your scientific field, open science, team science, interactions with stakeholders. Past and future plans. As a structured narrative.

Original (Englisch) übersetzen

Topics

Main Focus: Science \*  
[g.] Apoptosis

Main Focus: Clinic  
[g.] Clinical Psychotherapy

Please describe in short what you believe is your scientific contribution in your scientific field.  
[scientific contribution]

Remaining characters: 1000

What do you consider to be the 3 most important papers you have published? Please briefly justify this selection and mention your respective contribution. How have the work accepted in the scientific field, what impact did they have on the advancement of knowledge or the clinical practice (therapies, guidelines)? \*

[PubMed-ID] OR [DOI]

[Description of first publication] [Own share of the first publication]

The Charite attaches great importance to transparent, replicable research and supports the objectives of Open Science (Open Access, Open Data). This includes the registration of studies in registries (clinicaltrials.gov, DRKS, etc.), the preregistration of studies, and the publication of negative and zero results. How have you been pursuing these goals so far and what are your plans for the future?

Remaining characters: 1000

Charite is interested in team science and collaborations. Please describe in short most important collaboration projects within recent five years.  
[g.] Kiehl Institute

[Description]

Please describe in short your interactions with relevant actors in biomedicine, e.g. industry, patient care, policy panel, etc.

Remaining characters: 1000

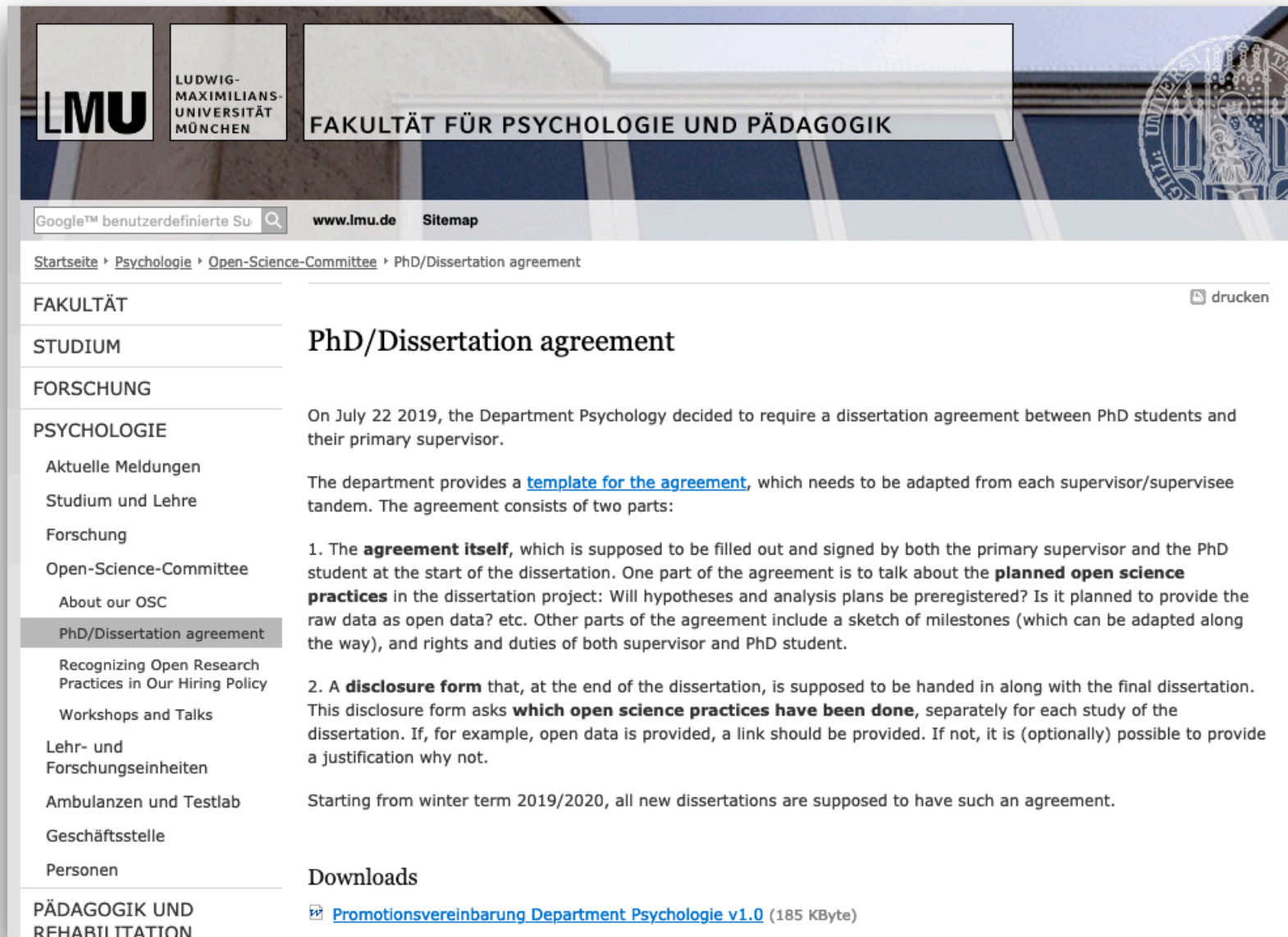
relevant patents  
[patent number]

[Description]

01:21 - 4. März 2018

See more such prof job ads at: <https://osf.io/7jbnt/>


# Dissertation agreement




The screenshot shows the website of the Faculty of Psychology and Pedagogy at Ludwig-Maximilians-Universität München (LMU). The page is titled "PhD/Dissertation agreement". It features a left sidebar with a navigation menu and a main content area with text and a download link.

**LMU** LUDWIG-MAXIMILIANS-UNIVERSITÄT MÜNCHEN

FAKULTÄT FÜR PSYCHOLOGIE UND PÄDAGOGIK

Google™ benutzerdefinierte Su  [www.lmu.de](http://www.lmu.de) [Sitemap](#)

[Startseite](#) > [Psychologie](#) > [Open-Science-Committee](#) > [PhD/Dissertation agreement](#)  drucken

**FAKULTÄT**

**STUDIUM**

**FORSCHUNG**

**PSYCHOLOGIE**

Aktuelle Meldungen

Studium und Lehre

Forschung

Open-Science-Committee

About our OSC

**PhD/Dissertation agreement**

Recognizing Open Research Practices in Our Hiring Policy

Workshops and Talks

Lehr- und Forschungseinheiten

Ambulanzen und Testlab

Geschäftsstelle

Personen

**PÄDAGOGIK UND REHABILITATION**

## PhD/Dissertation agreement


On July 22 2019, the Department Psychology decided to require a dissertation agreement between PhD students and their primary supervisor.

The department provides a [template for the agreement](#), which needs to be adapted from each supervisor/supervisee tandem. The agreement consists of two parts:

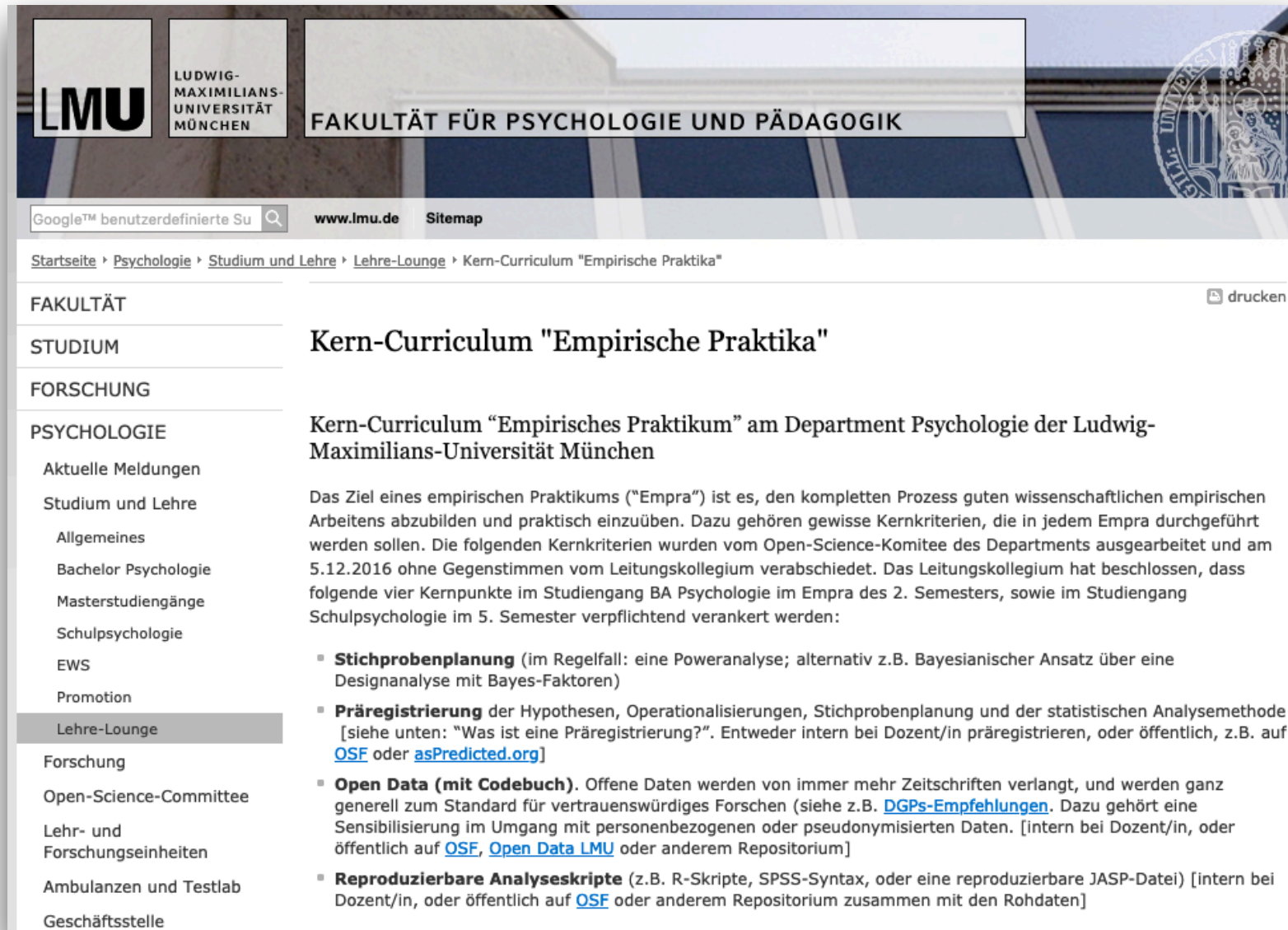
1. The **agreement itself**, which is supposed to be filled out and signed by both the primary supervisor and the PhD student at the start of the dissertation. One part of the agreement is to talk about the **planned open science practices** in the dissertation project: Will hypotheses and analysis plans be preregistered? Is it planned to provide the raw data as open data? etc. Other parts of the agreement include a sketch of milestones (which can be adapted along the way), and rights and duties of both supervisor and PhD student.
2. A **disclosure form** that, at the end of the dissertation, is supposed to be handed in along with the final dissertation. This disclosure form asks **which open science practices have been done**, separately for each study of the dissertation. If, for example, open data is provided, a link should be provided. If not, it is (optionally) possible to provide a justification why not.

Starting from winter term 2019/2020, all new dissertations are supposed to have such an agreement.

### Downloads

 [Promotionsvereinbarung Department Psychologie v1.0](#) (185 KByte)

# Open science in curricula



The screenshot shows the website of the Faculty of Psychology and Pedagogy at Ludwig-Maximilians-Universität München (LMU). The header includes the LMU logo, the faculty name, and a search bar. The main navigation menu on the left lists various sections, with 'Lehre-Lounge' currently selected. The main content area displays the 'Kern-Curriculum "Empirische Praktika"' page. This page includes a brief description of the empirical practicum's goals and a list of four key points: Stichprobenplanung, Präregistrierung, Open Data (mit Codebuch), and Reproduzierbare Analyseskripte. Each point provides details on its implementation and where to find related resources.

**LMU** LUDWIG-MAXIMILIANS-UNIVERSITÄT MÜNCHEN

FAKULTÄT FÜR PSYCHOLOGIE UND PÄDAGOGIK

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drucken

## FAKULTÄT

## STUDIUM

## FORSCHUNG

### PSYCHOLOGIE

- Aktuelle Meldungen
- Studium und Lehre
  - Allgemeines
  - Bachelor Psychologie
  - Masterstudiengänge
  - Schulpsychologie
  - EWS
  - Promotion
  - Lehre-Lounge**
- Forschung
- Open-Science-Committee
- Lehr- und Forschungseinheiten
- Ambulanzen und Testlab
- Geschäftsstelle

## Kern-Curriculum "Empirische Praktika"

### Kern-Curriculum "Empirisches Praktikum" am Department Psychologie der Ludwig-Maximilians-Universität München

Das Ziel eines empirischen Praktikums ("Empira") ist es, den kompletten Prozess guten wissenschaftlichen empirischen Arbeitens abzubilden und praktisch einzuüben. Dazu gehören gewisse Kernkriterien, die in jedem Empira durchgeführt werden sollen. Die folgenden Kernkriterien wurden vom Open-Science-Komitee des Departments ausgearbeitet und am 5.12.2016 ohne Gegenstimmen vom Leitungskollegium verabschiedet. Das Leitungskollegium hat beschlossen, dass folgende vier Kernpunkte im Studiengang BA Psychologie im Empira des 2. Semesters, sowie im Studiengang Schulpsychologie im 5. Semester verpflichtend verankert werden:

- **Stichprobenplanung** (im Regelfall: eine Poweranalyse; alternativ z.B. Bayesianischer Ansatz über eine Designanalyse mit Bayes-Faktoren)
- **Präregistrierung** der Hypothesen, Operationalisierungen, Stichprobenplanung und der statistischen Analysemethoden [siehe unten: "Was ist eine Präregistrierung?". Entweder intern bei Dozent/in präregistrieren, oder öffentlich, z.B. auf [OSF](#) oder [asPredicted.org](#)]
- **Open Data (mit Codebuch)**. Offene Daten werden von immer mehr Zeitschriften verlangt, und werden ganz generell zum Standard für vertrauenswürdigen Forschung (siehe z.B. [DGPs-Empfehlungen](#)). Dazu gehört eine Sensibilisierung im Umgang mit personenbezogenen oder pseudonymisierten Daten. [intern bei Dozent/in, oder öffentlich auf [OSF](#), [Open Data LMU](#) oder anderem Repositorium]
- **Reproduzierbare Analyseskripte** (z.B. R-Skripte, SPSS-Syntax, oder eine reproduzierbare JASP-Datei) [intern bei Dozent/in, oder öffentlich auf [OSF](#) oder anderem Repositorium zusammen mit den Rohdaten]

# Hiring committees: Require an annotated CV with limited items (e.g., $\leq 10$ )

No journal; JIF is irrelevant or misleading

Paper-level citation metrics

Basic information for judging evidential value

Open science indicators: Judging reproducibility

Data: own collection or reuse?

Authors & title	Year	Citations	Sample size per study	p-value per study	Open Science indicators	Data set	Applicants contribution
Doe, John & Smith, Peter	2001	47	n <sub>1</sub> = 21 n <sub>2</sub> = 30 n <sub>3</sub> = 19	p <sub>1</sub> = .048 p <sub>2</sub> = .050 p <sub>3</sub> = .023	<input type="checkbox"/> Open Data <input type="checkbox"/> Open Material <input type="checkbox"/> Preregistered	<input checked="" type="checkbox"/> Own data collection → URL NA <input type="checkbox"/> Archival data	<ul style="list-style-type: none"> <li>Analyzed data</li> <li>Wrote manuscript</li> </ul>
Doe, John	2016	26	n <sub>1</sub> = 180 n <sub>2</sub> = 158	p <sub>1</sub> = .012 p <sub>2</sub> = .001	<input checked="" type="checkbox"/> Open Data <input checked="" type="checkbox"/> Open Material <input checked="" type="checkbox"/> Preregistered	<input checked="" type="checkbox"/> Own data collection → URL <a href="https://osf.io/as1cd">osf.io/as1cd</a> <input type="checkbox"/> Archival data	<ul style="list-style-type: none"> <li>Designed study</li> <li>Wrote manuscript</li> </ul>

# Open Science: Who to ask?

- Ask Open Science Initiative (University of Bielefeld)  
<https://ask-open-science.org/>
- Your local Open Science Initiative  
<https://osf.io/tbkzh/wiki/home>
- Reddit Open Science  
[https://www.reddit.com/r/Open\\_Science/](https://www.reddit.com/r/Open_Science/)



# Open Science: 3 Easy Steps

## **How you can improve your OS record (almost) without effort**

1. When reviewing a paper: Keep an eye on QRPs and ask for open data and open material (<https://opennessinitiative.org/>)
2. Which parts of your research process can you make open?  
Start out with the least work-intensive part and give it a try!
3. Get a Twitter account and join the discussion

# Open Science: What you learned

- Open Science as part of good research practice
- The science hamster wheel: Incentive structures in research
- The replication crisis: Non-replicability in research and its problems
- Identify scientific misconduct: Fabrication, Falsification, Questionable Research Practices
- Methods of p-hacking: Optional stopping, HARKing, selective reporting, analytical flexibility
- The open research process: Preregistration, Registered Reports, Open Lab Notebooks, Open Data, Open Materials, Open Access Publishing, Replication Studies
- How to make a change: Open Science networks and initiatives

# Further Resources

- Arslan, R. (2018). *Open science vs. bad science: The replication crisis and possible reforms*. Presentation slides available on [osf.io/65mqz/](https://osf.io/65mqz/)
- Gelman, A., & Loken, E. (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. *Department of Statistics, Columbia University*.  
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\* Unfortunately, there is no Open Access version of this article available

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