

# “Opening Up” (Cognitive Neuro)science

Rik Henson

MRC CBU, Cambridge

*With thanks to: Rogier Kievit, Niko Kriegeskorte, Dorothy Bishop, Anthony Isles, Amy Orben, Marcus Munafo...*

# The Problem

STANFORD GRADUATE SCHOOL OF BUSINESS

Education

**Discredited "Mozart Effect" Remains Music to American Ears**



Science | Wed Mar 28, 2012 7:09pm BST

**In cancer science, many "discoveries" don't hold up**

NEW YORK | BY SHARON BEGLEY



53 landmark papers on cancer  
**47 did not replicate**

For a comedian's recent perspective:

<https://www.facebook.com/LastWeekTonight/videos/896755337120143>

## Power failure: why small sample size undermines the reliability of neuroscience

*Katherine S. Button<sup>1,2</sup>, John P. A. Ioannidis<sup>3</sup>, Claire Mokrysz<sup>1</sup>, Brian A. Nosek<sup>4</sup>, Jonathan Flint<sup>5</sup>, Emma S. J. Robinson<sup>6</sup> and Marcus R. Munafò<sup>1</sup>*

## Scanning the horizon: towards transparent and reproducible neuroimaging research

*Russell A. Poldrack<sup>1</sup>, Chris I. Baker<sup>2</sup>, Joke Durnez<sup>1,3</sup>, Krzysztof J. Gorgolewski<sup>1</sup>, Paul M. Matthews<sup>4</sup>, Marcus R. Munafò<sup>5,6</sup>, Thomas E. Nichols<sup>7</sup>, Jean-Baptiste Poline<sup>8</sup>, Edward Vul<sup>9</sup> and Tal Yarkoni<sup>10</sup>*

### Article

## Reproducible brain-wide association studies require thousands of individuals

<https://doi.org/10.1038/s41586-022-04492-9>

Received: 19 May 2021

Accepted: 31 January 2022

Published online: 16 March 2022

Open access

 Check for updates

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# In Neuroimaging...



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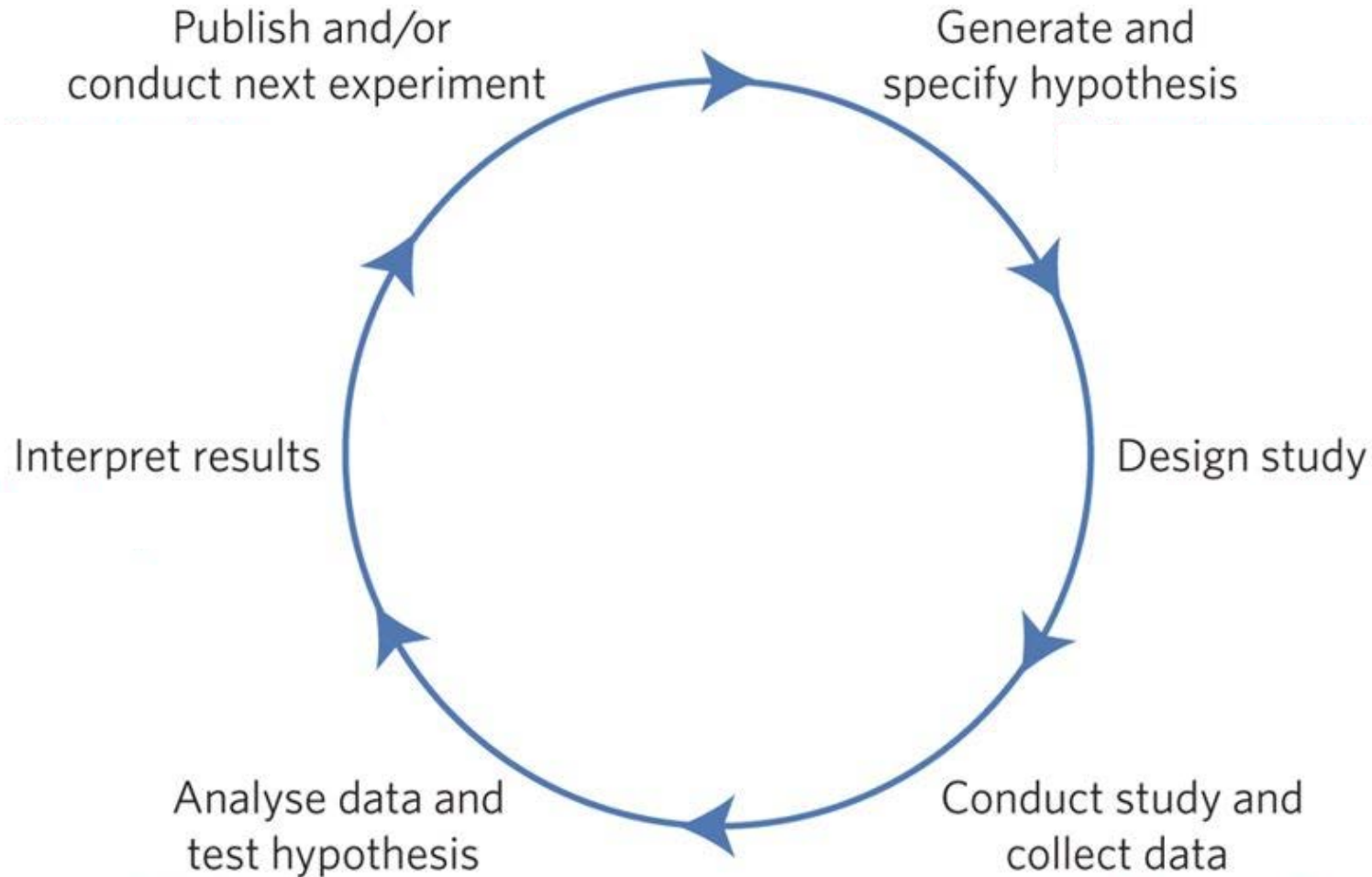
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*"...the high dimensionality of fMRI data, the relatively low power of most fMRI studies and the great amount of flexibility in data analysis contribute to a potentially high degree of false-positive findings."*

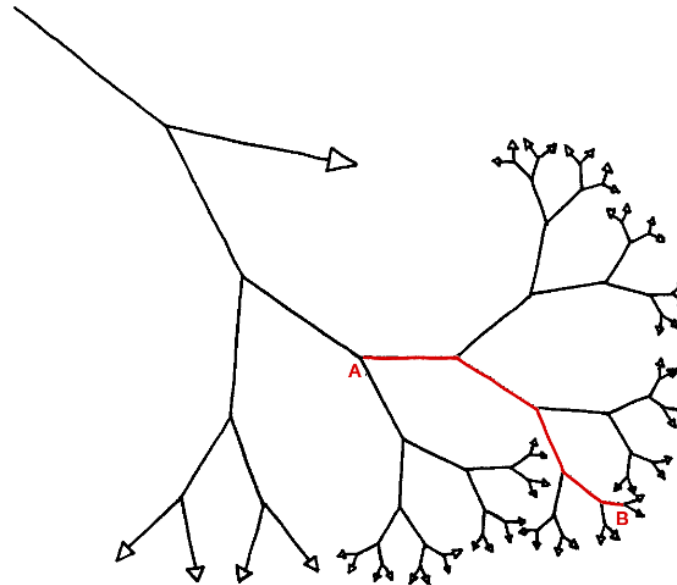
# The Problems



# Overview

- Registration
  - Study Registration (eg OSF)
  - Registered Reports
  - Pre-Registration Posters
- Statistical analysis
- Sharing Data and Code
- Publication
- Research Culture

## The Garden of Forking Paths by Jorge Luis Borges

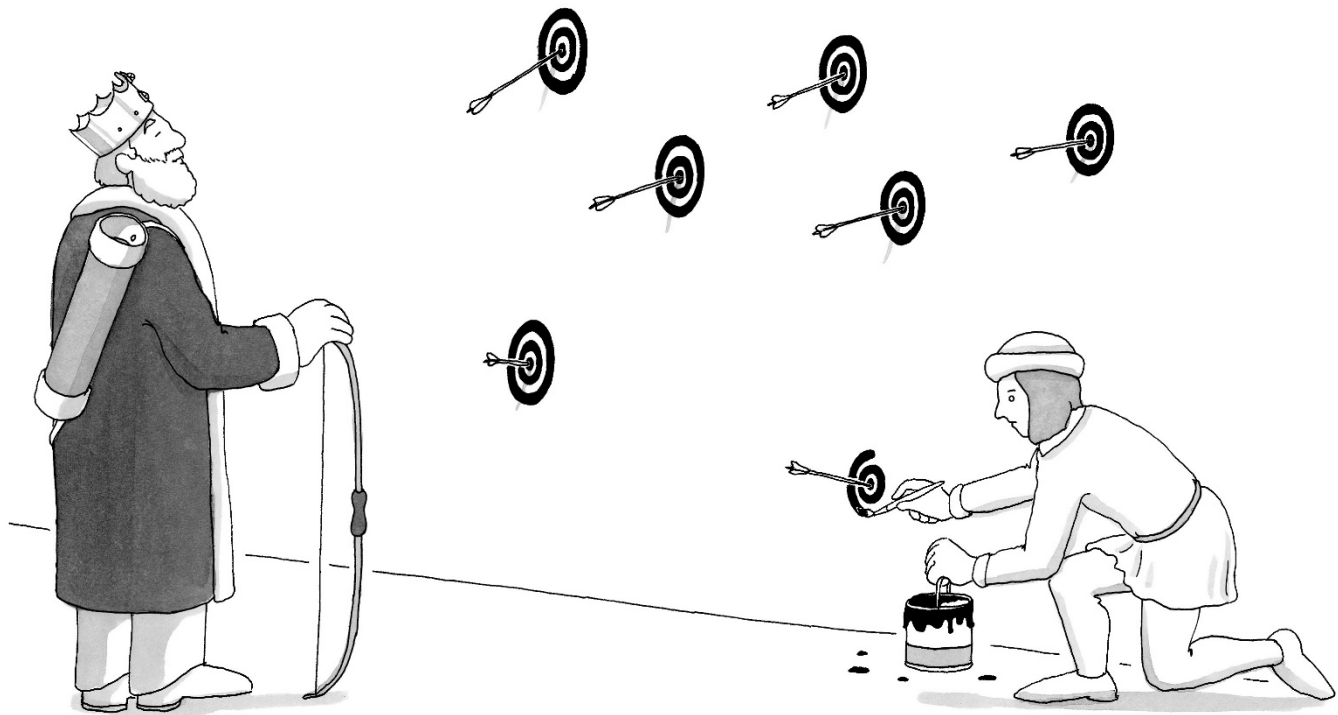


Particularly likely in neuroimaging, given so many analysis choices...?

Multiverse analyses?

# HARKing

HARK = “Hypothesising After the Results are Known”



Hankin

CartoonCollections.com



# OSF Registration

OSF Registries | New Registration x +

osf.io/registries/ost/new?view\_only=

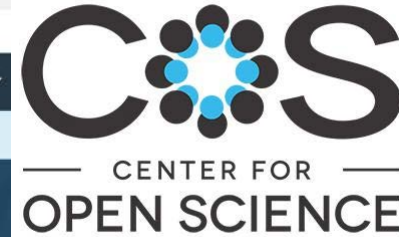
Inbox - rik.henson... CamCAN Websites myrefs Notifications Notifi... Imported

OSF REGISTRIES

Add New My Registrations Help Donate

Notice: The site will undergo maintenance between Aug 19, 2022 1:00 AM and Aug 19, 2022 2:00 AM (+0100 UTC). Thank you for your patience.

## Add New Registration



You are submitting to OSF Registries. [Click here](#) to learn more about other hosted registries.

### STEP 1

Do you have content for registration in an existing OSF project?

YES

NO

### STEP 2

Which type of registration would you like to create? \*

OSF Preregistration

Create draft

- Various levels of specificity (ideally all analysis scripts!)
- Pre-registration time-stamps a public commitment, but i) not binding for publication, and ii) not reviewed before data collection



# Registered Reports

- Peer Review before data collection/analysis
- Guaranteed Publication regardless of results

## Filing Drawer problem



# Registered Reports

- Peer Review before data collection/analysis
- Guaranteed Publication regardless of results



- Can report non-registered findings, but clear division between “confirmatory” and “exploratory” results
- Some of many Cognitive Neuroscience journals allowing RRs:  
*Cortex, Frontiers, Journal of Cognitive Neuroscience, Nature Human Behaviour, Psychological Science, Quarterly Journal of Experimental Psychology, Brain Neuroscience Advances...*
- (not currently: *Nature, Science, J. Neuroscience, Neuroimage, APA journals...* 😞)

# Pre-Reg Posters

## Trends in Cognitive Sciences

Home / News / Preregistration posters: early findings about presenting research early

## Scientific Life

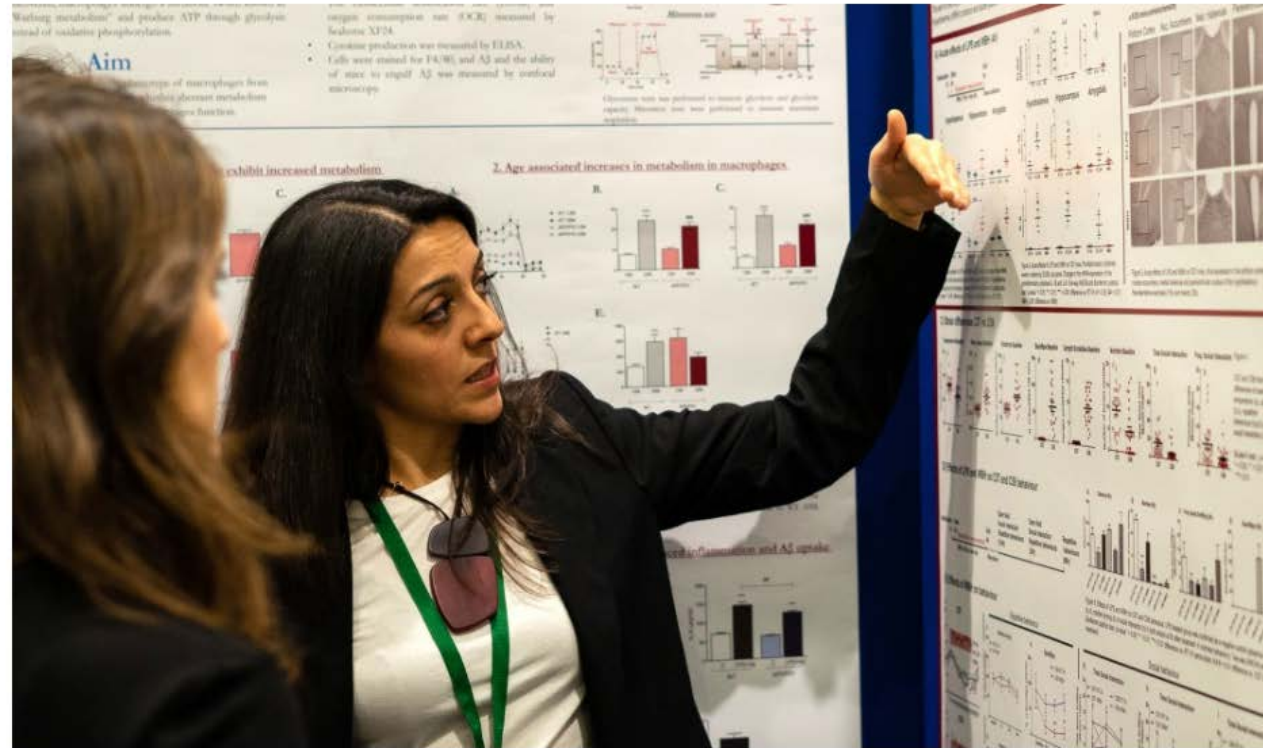
### Title TBA: Revising the Abstract Submission Process

Roni Tibon,<sup>1,\*</sup> CBU Open Science Committee,<sup>1</sup> and Richard Henson<sup>1</sup>

Academic conferences are among the most prolific scientific activities, yet the current abstract submission and review process has serious limitations. We propose a revised process that would address these limitations, achieve some of the aims of Open Science, and stimulate discussion throughout the entire lifecycle of the scientific work.

## PREREGISTRATION POSTERS: EARLY FINDINGS ABOUT PRESENTING RESEARCH EARLY

21st Aug 2019



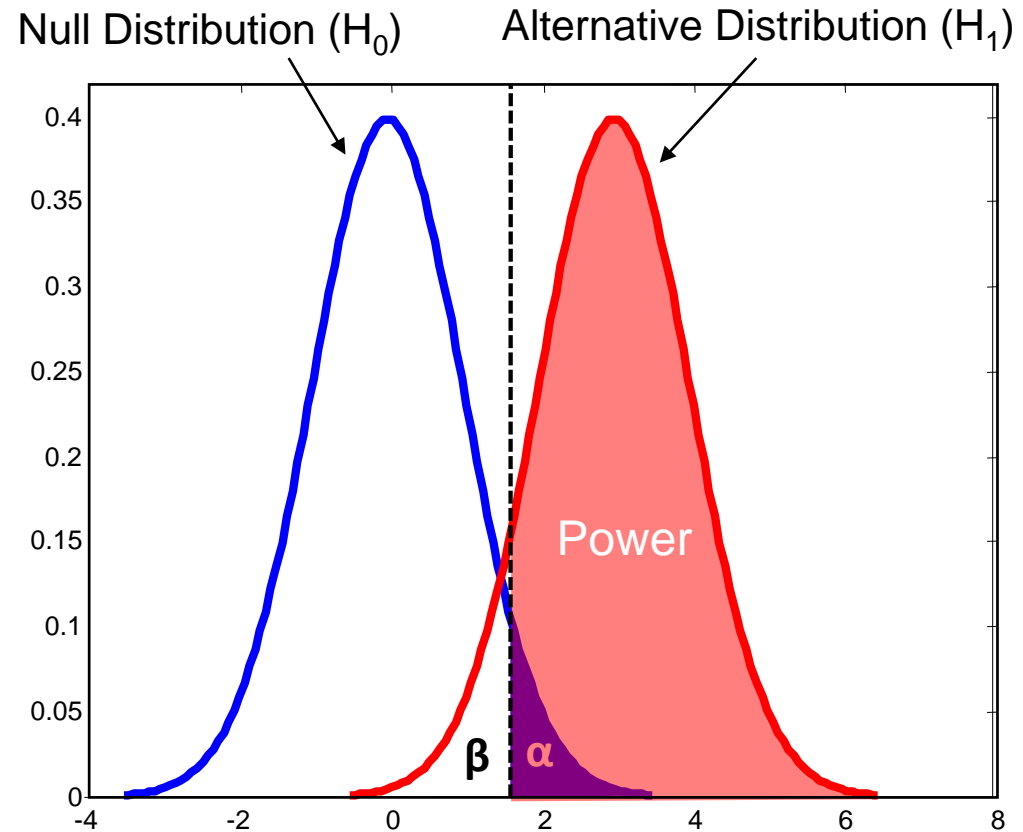
- Chance to get feedback (eg, “Is hypothesis interesting? Sufficient controls? Appropriate analysis?”) before submitting a website registration or RR...

# Overview

- Registration
  - Study Registration (eg OSF)
  - Registered Reports
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- Statistical analysis
  - Power and PPV
  - Bayesian Statistics
  - Sequential Designs
- Sharing Data and Code
- Publication
- Research Culture

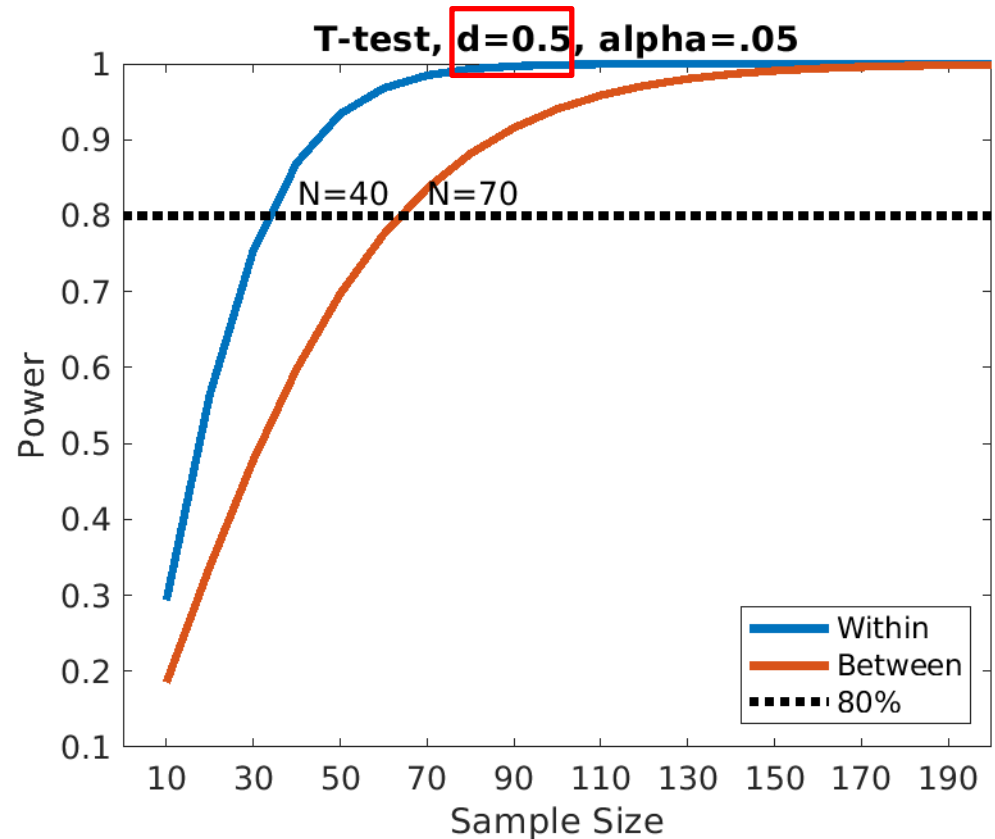
# Power

- Power = probability of rejecting  $H_0$  when  $H_1$  is true
- Must specify:
  - Sample size  $n$
  - Level  $\alpha$   
(allowed false positive rate)
  - Standard deviation  $\sigma$   
(population variability)
  - Effect magnitude  $\Delta$
- Last two can be replaced with
  - Effect size:  $\delta = \Delta/\sigma$
  - E.g, according to Cohen:
    - $\delta=0.8$  is a large effect size
    - $\delta=0.5$  is a medium effect size
    - $\delta=0.2$  is a small effect size



# Power Curves

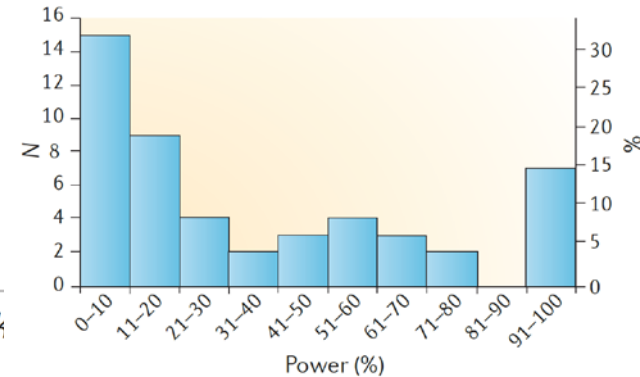
- Assuming medium effect size ( $d=0.5$ ) for a frequentist T-test:
- Within-participant (repeated measures) tests more powerful than between-participant tests (latter need  $N \sim 140$  participants total for  $>80\%$  power)
- G\*Power:  
<https://www.psychologie.hhu.de/arbeitsgruppen/allgemeine-psychologie-und-arbeitspsychologie/gpower>



# In Neuro...

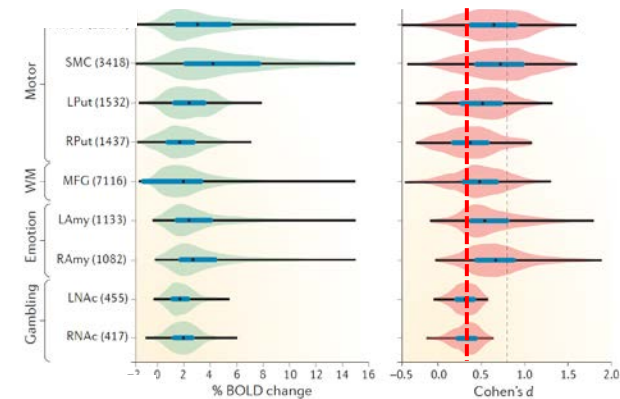
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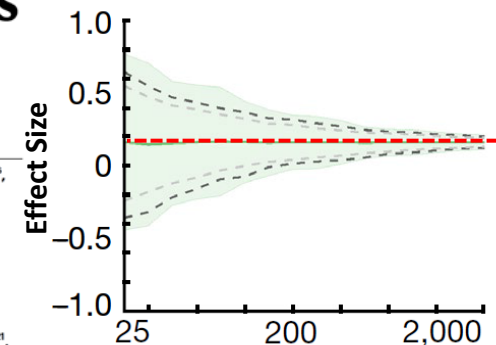
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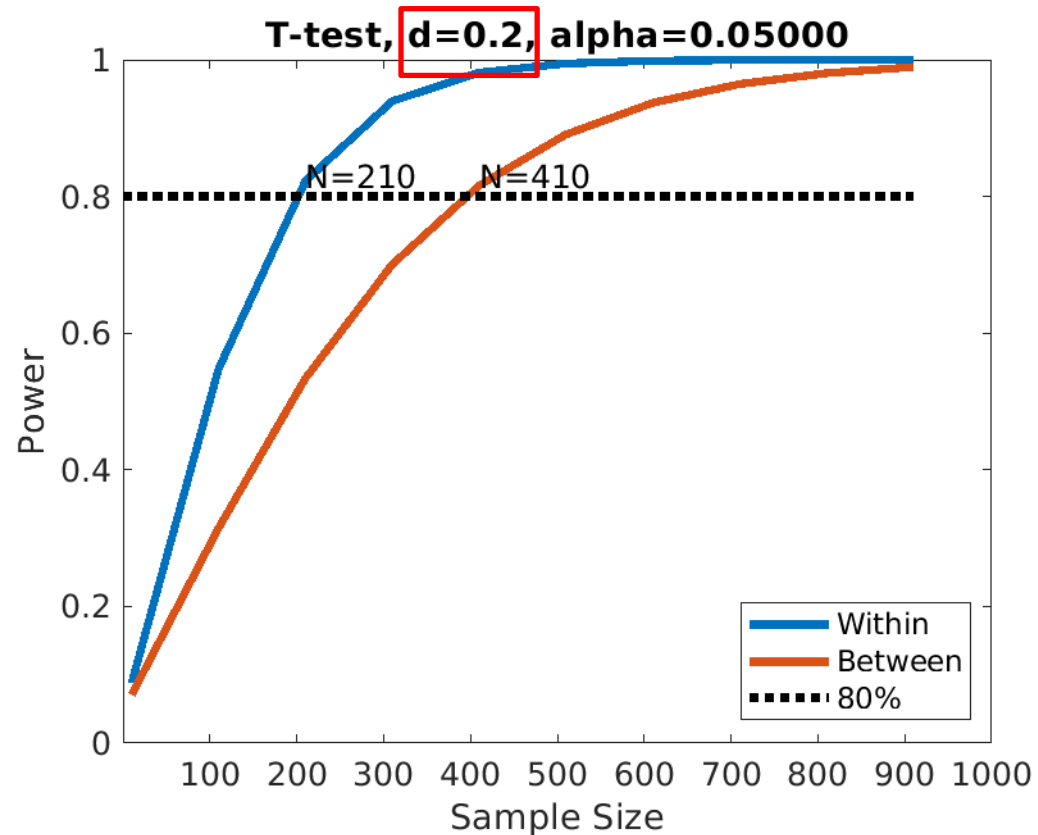
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# Power Curves

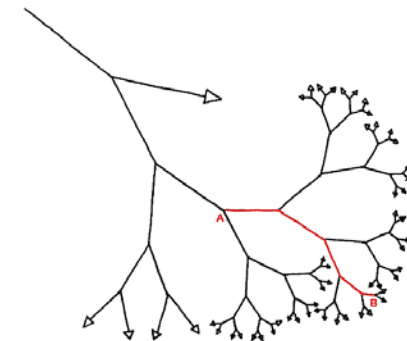
- Assuming medium effect size ( $d=0.5$ ) for a frequentist T-test:
- Within-participant (repeated measures) tests more powerful than between-participant tests (latter need  $N \sim 140$  participants total for  $>80\%$  power)
- With small effect size  $d=0.2$ , approaching total of  $N \sim 1000$  for between-participant test
- G\*Power:  
<https://www.psychologie.hhu.de/arbeitsgruppen/allgemeine-psychologie-und-arbeitspsychologie/gpower>



# Multiple Comparisons

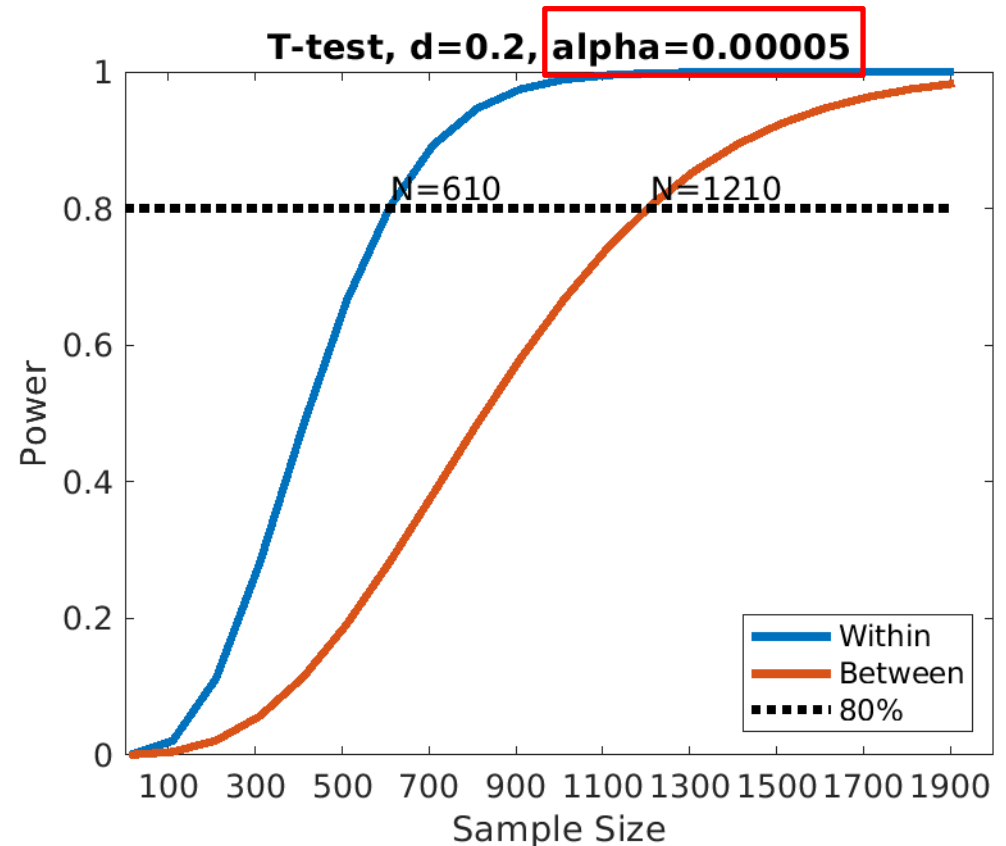


**The Garden of Forking Paths**  
by Jorge Luis Borges



# Power Curves

- Assuming medium effect size ( $d=0.5$ ) for a frequentist T-test:
- Within-participant (repeated measures) tests more powerful than between-participant tests (latter need  $\sim 140$  participants total for  $>80\%$  power)
- With small effect size  $d=0.2$ , correction for 1000 tests (“resels”) approaches total of  $N\sim 2500$  for between-participant test



More sophisticated treatment of multiple comparisons, within- and between-participant variance (e.g, #trials and #participants):

fMRIPower: <http://fmripower.org>

PowerMap: <http://sourceforge.net/projects/powermap>

NeuroPower: <http://neuropower.shinyapps.io/neuropower>

# Positive Predictive Value (PPV)



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Essay

## Why Most Published Research Findings Are False

John P. A. Ioannidis

### Summary

There is increasing concern that most current published research findings are false. The probability that a research claim is true may depend on study power and bias, the number of other studies on the same question, and, importantly, the ratio of true to no relationships among the relationships probed in each scientific field. In this framework, a research finding is less likely to be true when the studies conducted in a field are smaller; when effect sizes are smaller; when there is a greater number and lesser preselection of tested relationships; where there is greater flexibility in designs, definitions, outcomes, and analytical modes; when there is greater financial and other interest and prejudice; and when more teams are involved in a scientific field in chase of statistical significance. Simulations show that for most study designs and settings, it is more likely for a research claim to be false than true. Moreover, for many current scientific fields, claimed research findings may often be simply accurate measures of the

factors that influence this problem and some corollaries thereof.

### Modeling the Framework for False Positive Findings

Several methodologists have pointed out [9–11] that the high rate of nonreplication (lack of confirmation) of research discoveries is a consequence of the convenient, yet ill-founded strategy of claiming conclusive research findings solely on the basis of a single study assessed by formal statistical significance, typically for a  $p$ -value less than 0.05. Research is not most appropriately represented and summarized by  $p$ -values, but, unfortunately, there is a widespread notion that medical research articles

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

should be interpreted based only on  $p$ -values. Research findings are defined here as any relationship reaching

is characteristic of the field and can vary a lot depending on whether the field targets highly likely relationships or searches for only one or a few true relationships among thousands and millions of hypotheses that may be postulated. Let us also consider, for computational simplicity, circumscribed fields where either there is only one true relationship (among many that can be hypothesized) or the power is similar to find any of the several existing true relationships. The pre-study probability of a relationship being true is  $R/(R + 1)$ . The probability of a study finding a true relationship reflects the power  $1 - \beta$  (one minus the Type II error rate). The probability of claiming a relationship when none truly exists reflects the Type I error rate,  $\alpha$ . Assuming that  $c$  relationships are being probed in the field, the expected values of the  $2 \times 2$  table are given in Table 1. After a research finding has been claimed based on achieving formal statistical significance, the post-study probability that it is true is the positive predictive value, PPV.

# PPV Arithmetic



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|                     | Hypothesis True (H+)   | Hypothesis False (H-)  |
|---------------------|--|--|
| Positive Finding D+ | $P(D+   H+)$<br>("hit", "sensitivity")<br><i>Power</i> $1-\beta$ | $P(D+   H-)$<br>("false alarm", "Type I error")<br><i>FPR</i> $\alpha$ |
| Negative Finding D- | $P(D-   H-)$<br>("miss", "Type II error")                        | $P(D-   H+)$<br>("correct rejection", "specificity")                   |

# PPV Arithmetic



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|                     | Hypothesis True (H+)   | Hypothesis False (H-)  |
|---------------------|--|--|
| Positive Finding D+ | $P(D+   H+)$<br>(“hit”, “sensitivity”)<br><i>Power</i> $1-\beta$ | $P(D+   H-)$<br>(“false alarm”, “Type I error”)<br><i>FPR</i> $\alpha$ |
| Negative Finding D- | $P(D-   H-)$<br>(“miss”, “Type II error”)                        | $P(D-   H-)$<br>(“correct rejection”, “specificity”)                   |
| (Prior)             | $P(H+)$  | $P(H-)$  |

$$PPV = P(H+ | D+)$$

$$= p(D+ | H+) \times p(H+) / p(D+)$$

*Bayes Rule*

$$p(D+) = p(D+/H+) \times p(H+) + p(D+/H-) \times p(H-)$$

*Summation Rule*

$$PPV = p(D+ | H+) \times p(H+) / (p(D+/H+) \times p(H+) + p(D+/H-) \times p(H-))$$

$$R = p(H+)/p(H-)$$

*(a priori) Odds Ratio of Hypothesis being true*

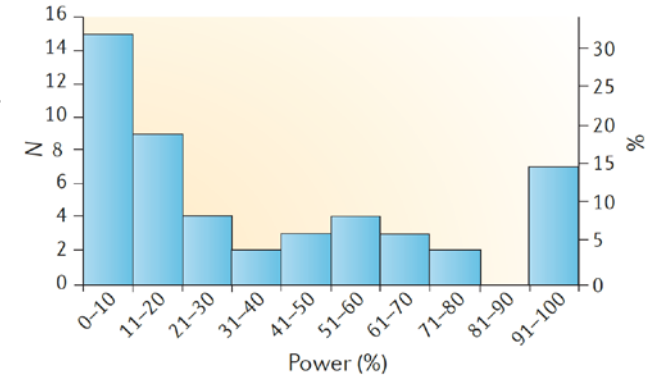
$$PPV = (1-\beta) \times R / ((1-\beta) \times R) + \alpha$$

# PPV Arithmetic

- Why most studies false, ie  $PPV < \frac{1}{2}$ ?

$$0.5 > PPV = \frac{(1-\beta) R}{(1-\beta) R + \alpha} \quad \Rightarrow \quad \alpha > (1-\beta) R$$

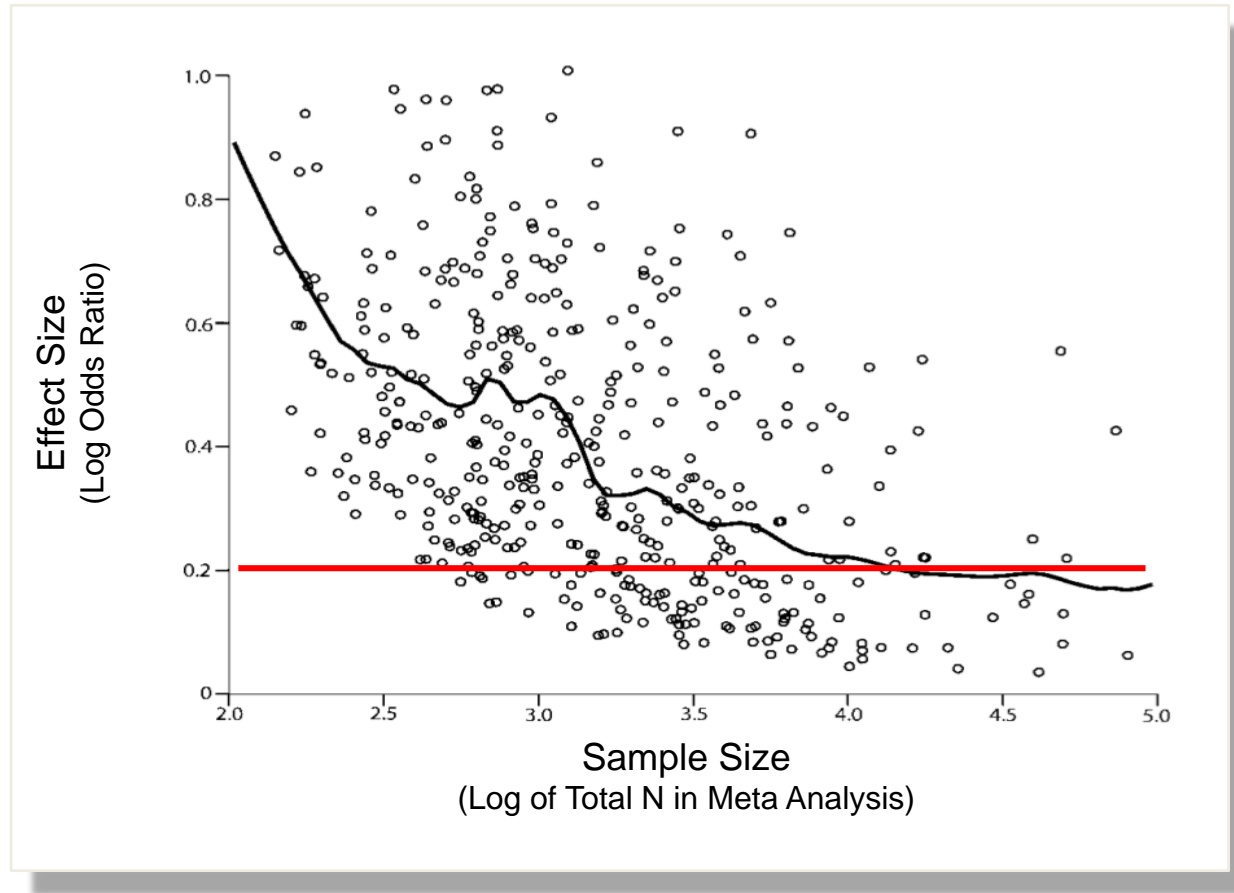
- Assume power 20%, ie  $(1-\beta) = 0.2$  (and  $\alpha=0.05$ )...
- $0.05 > 0.20 R \rightarrow R < 0.05/0.20$
- So  $PPV < 0.5$  if  $H1:H0 < 1:4$ ; discovery science ?
- Worse once consider bias...



- PPV highly dependent on Power (since  $\alpha$  small)

$$PPV = (1-\beta) \frac{R}{R + \alpha/(1-\beta)} \approx (1-\beta)$$

# Additional Bias ( $u$ )



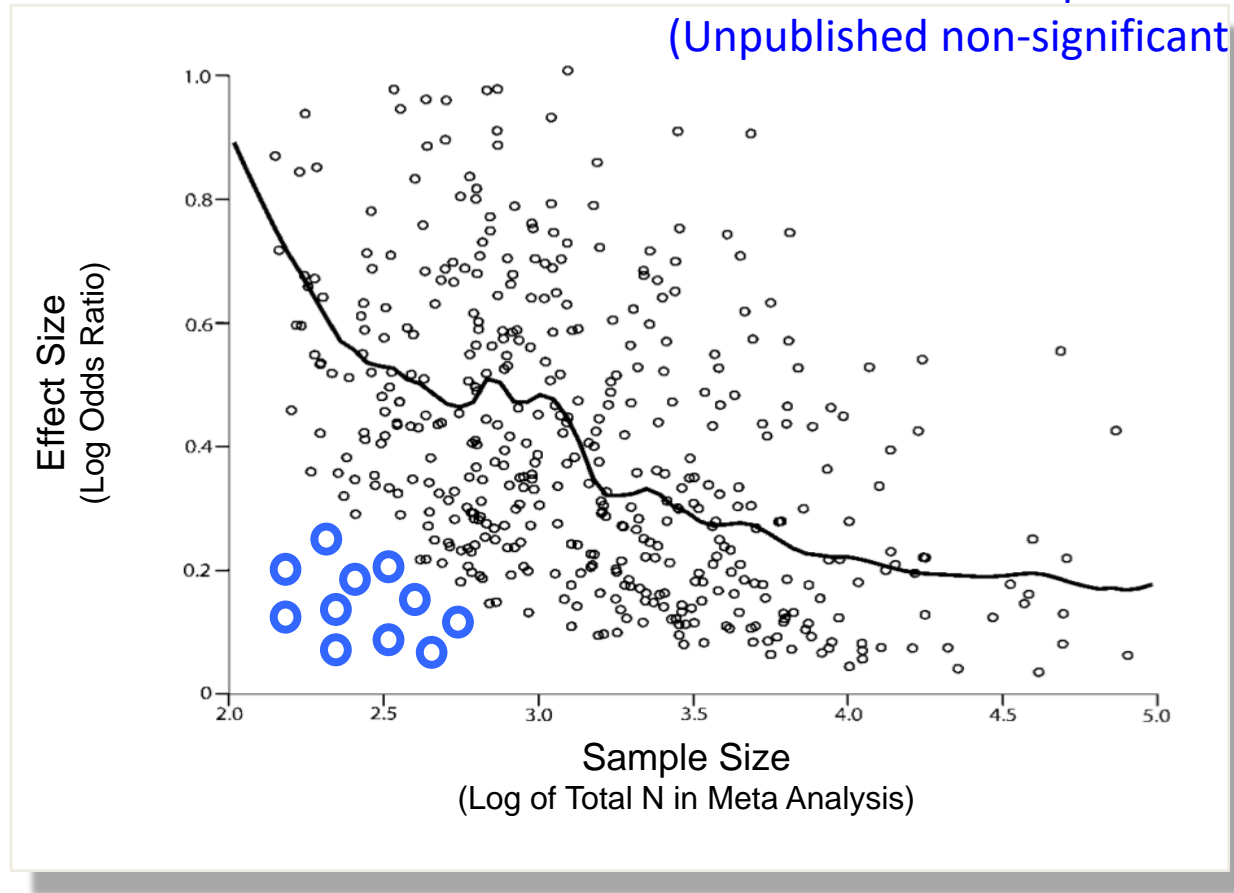


# Filing Drawer problem

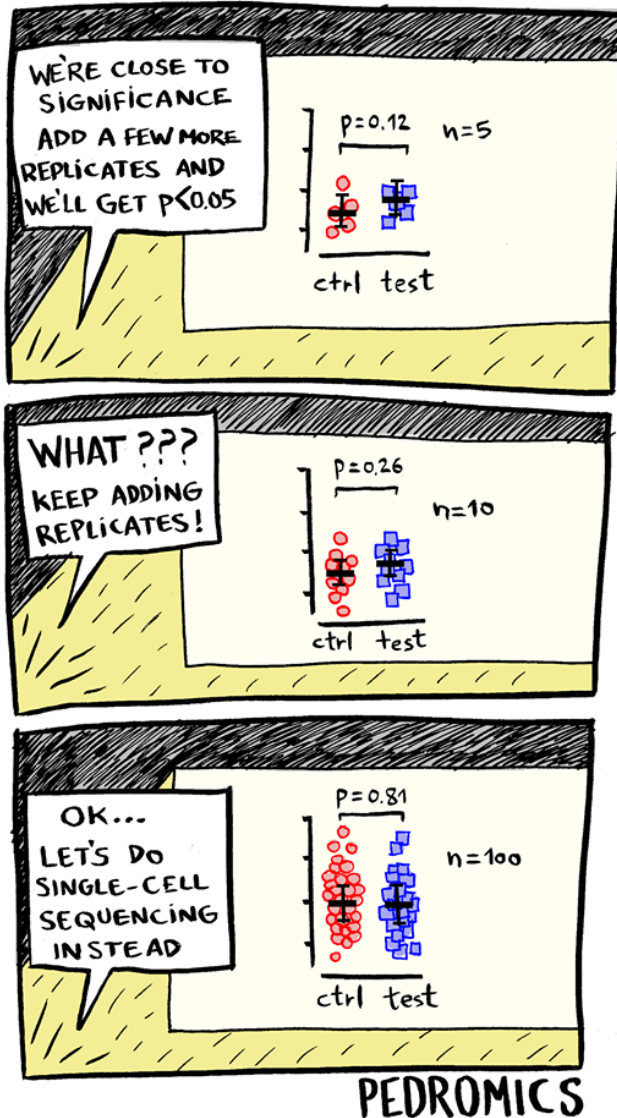


# Additional Bias ( $u$ )

File drawer problem  
(Unpublished non-significant studies)

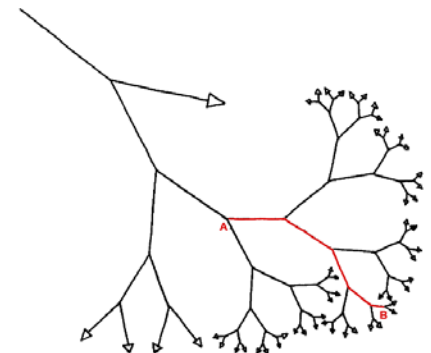


# P-Hacking

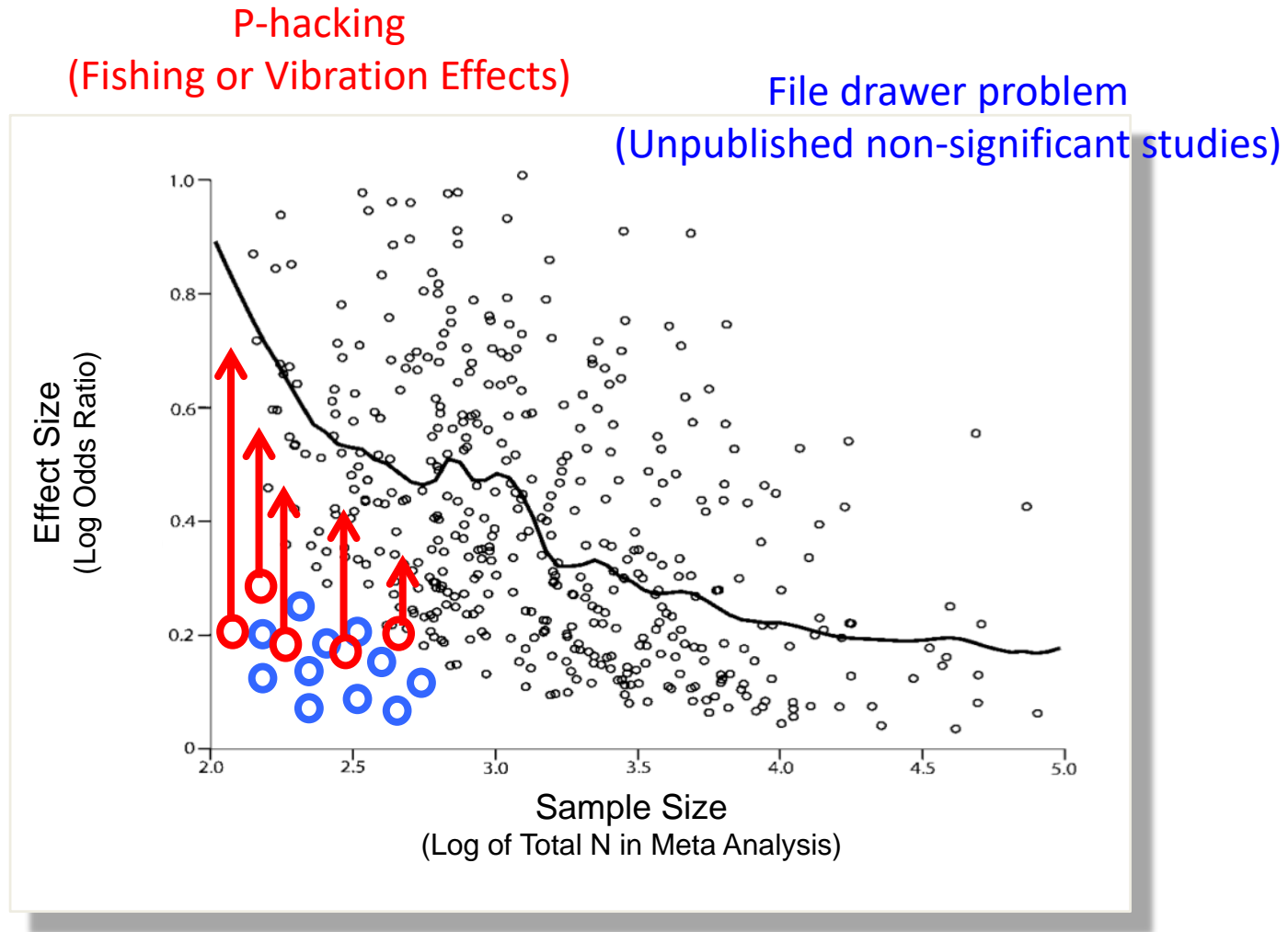


- Different statistical tests
- Different variables
- Removal of outliers
- Peeking & +/- n = numbers

The Garden of Forking Paths  
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# Additional Bias ( $u$ )



# PPV Arithmetic

- Adding bias of  $u$ :

$$PPV = \frac{(1-\beta)R + u\beta R}{(1-\beta)R + u\beta R + \alpha + u(1-\alpha)}$$

**Table 4.** PPV of Research Findings for Various Combinations of Power ( $1 - \beta$ ), Ratio of True to Not-True Relationships ( $R$ ), and Bias ( $u$ )

| $1 - \beta$ | $R$     | $u$  | Practical Example  | PPV    |
|-------------|---------|------|--|--------|
| 0.80        | 1:1     | 0.10 | Adequately powered RCT with little bias and 1:1 pre-study odds | 0.85   |
| 0.95        | 2:1     | 0.30 | Confirmatory meta-analysis of good-quality RCTs                | 0.85   |
| 0.80        | 1:3     | 0.40 | Meta-analysis of small inconclusive studies                    | 0.41   |
| 0.20        | 1:5     | 0.20 | Underpowered, but well-performed phase I/II RCT                | 0.23   |
| 0.20        | 1:5     | 0.80 | Underpowered, poorly performed phase I/II RCT                  | 0.17   |
| 0.80        | 1:10    | 0.30 | Adequately powered exploratory epidemiological study           | 0.20   |
| 0.20        | 1:10    | 0.30 | Underpowered exploratory epidemiological study                 | 0.12   |
| 0.20        | 1:1,000 | 0.80 | Discovery-oriented exploratory research with massive testing   | 0.0010 |

# Piloting

- Where obtain effect size for new study?
  - From literature? But publication bias (over-estimated)...
  - From pilot experiment? But then need large sample...
  - A priori (e.g, medium effect)? But will reviewers agree... (register!)

## Article

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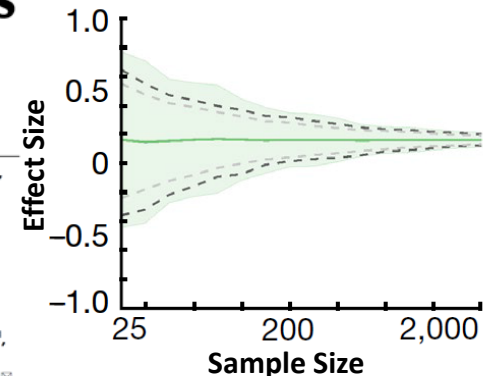
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Scott Marek<sup>1,30,31</sup>, Brenden Tervo-Clemmens<sup>2,3,30,31</sup>, Finnegan J. Calabro<sup>4,5</sup>, David F. Montez<sup>6</sup>, Benjamin P. Kay<sup>6</sup>, Alexander S. Hatoum<sup>1</sup>, Meghan Rose Donohue<sup>1</sup>, William Foran<sup>4</sup>, Ryland L. Miller<sup>1,6</sup>, Timothy J. Hendrickson<sup>7</sup>, Stephen M. Malone<sup>8</sup>, Sridhar Kandala<sup>1</sup>, Eric Feczko<sup>9,10</sup>, Oscar Miranda-Dominguez<sup>9,10</sup>, Alice M. Graham<sup>11</sup>, Eric A. Ear<sup>9,11</sup>, Anders J. Perrone<sup>9,11</sup>, Michaela Cordova<sup>11</sup>, Olivia Doyle<sup>11</sup>, Lucille A. Moore<sup>11</sup>, Gregory M. Conan<sup>9,11</sup>, Johnny Uriarte<sup>11</sup>, Kathy Snider<sup>11</sup>, Benjamin J. Lynch<sup>9,12</sup>, James C. Wilgenbusch<sup>9,12</sup>, Thomas Pengo<sup>7</sup>, Angela Tam<sup>13,14,15,16</sup>, Jianzhong Chen<sup>13,14,15,16</sup>, Dillan J. Newbold<sup>6</sup>, Annie Zheng<sup>6</sup>, Nicole A. Seider<sup>6</sup>, Andrew N. Van<sup>9,17</sup>, Athanasia Metokji<sup>6</sup>, Roselyne J. Chauvin<sup>6</sup>, Timothy O. Laumann<sup>1</sup>, Deanna J. Greene<sup>18</sup>, Steven E. Petersen<sup>6,17,19,20,21</sup>, Hugh Garavan<sup>22</sup>, Wesley K. Thompson<sup>23</sup>, Thomas E. Nichols<sup>24</sup>, B. T. Thomas Yeo<sup>13,14,15,16,25,26</sup>, Deanna M. Barch<sup>1,21</sup>, Beatriz Luna<sup>3,4</sup>, Damien A. Fair<sup>9,10,27,31,32</sup> & Nico U. F. Dosenbach<sup>6,17,19,28,29,31,32</sup>



# Classical (Frequentist) vs Bayesian Inference

- Classical “p-value” is *likelihood* of getting a statistic (derived from the data, D), given Null Hypothesis (H<sub>0</sub>) is true, i.e, that effect size is exactly zero:

$$p(D|H_0)$$

- Bayes Factor (BF) is the *relative evidence* for H<sub>1</sub> vs H<sub>0</sub> (or vice versa):

$$BF_{10} = \frac{p(D|H_1)}{p(D|H_0)}$$

- ...though requires you to specify some **priors** on H<sub>1</sub>, H<sub>0</sub> parameters...
  - “Subjective Bayesians” specify priors based on theory/data (register!)
  - “Objective Bayesians” specify priors as minimal (default) assumptions...

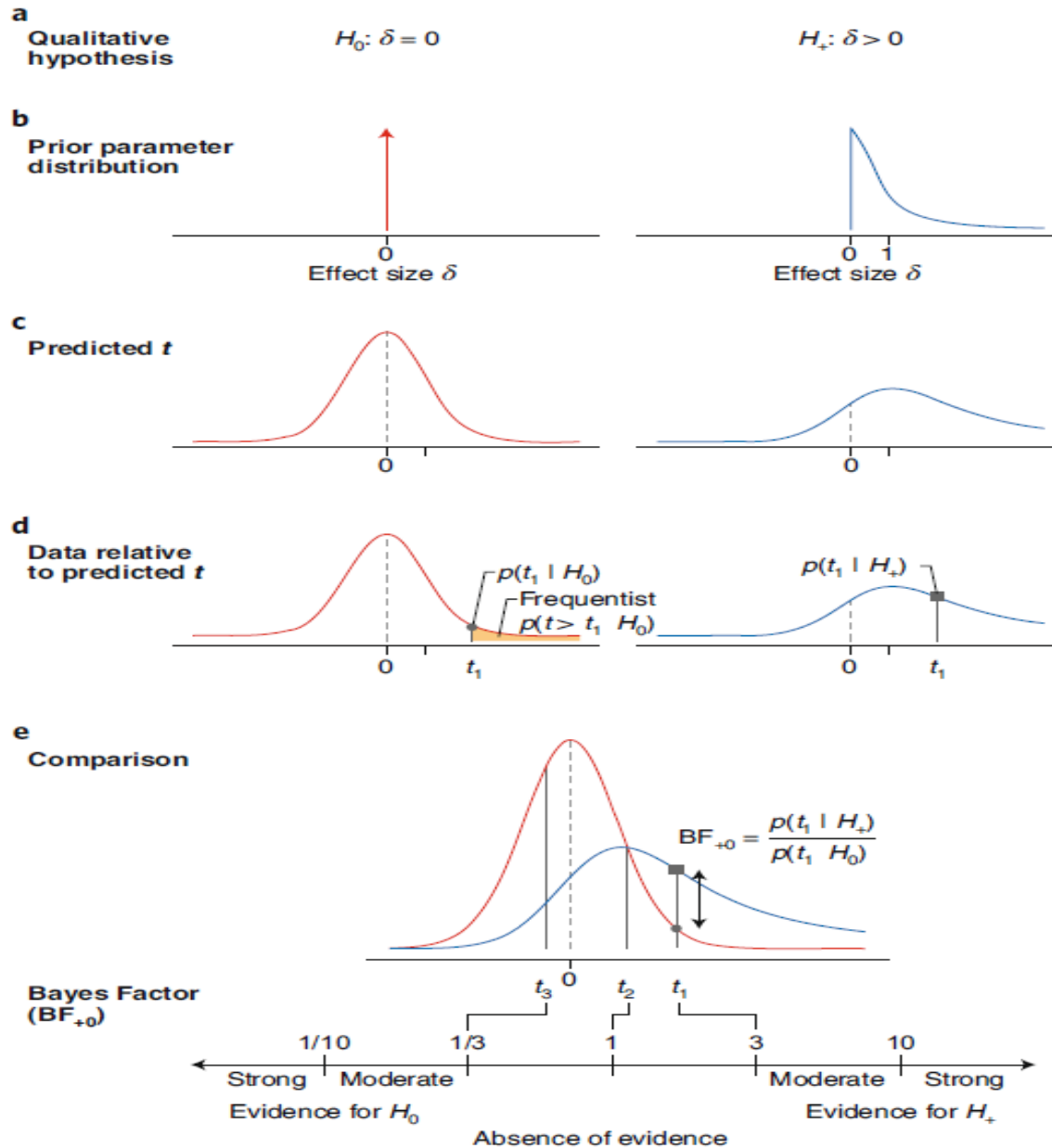
# Bayes Factors

| $BF_{10}$    | Evidence                       |
|--------------|--------------------------------|
| $> 100$      | Extreme evidence for $H_1$     |
| 30 – 100     | Very strong evidence for $H_1$ |
| 10 – 30      | Strong evidence for $H_1$      |
| 3 – 10       | Moderate evidence for $H_1$    |
| 1 – 3        | Anecdotal evidence for $H_1$   |
| 1            | No evidence                    |
| 1 – 1/3      | Anecdotal evidence for $H_0$   |
| 1/10 – 1/30  | Strong evidence for $H_0$      |
| 1/30 – 1/100 | Very strong evidence for $H_0$ |
| $< 1/100$    | Extreme evidence for $H_0$     |

- **Most journals** either require **BF of 6** or **10** for registered reports
- We often take  **$BF_{10} > 10$**  and  **$BF_{10} < 1/6$**  as sufficient



# Bayes Factors



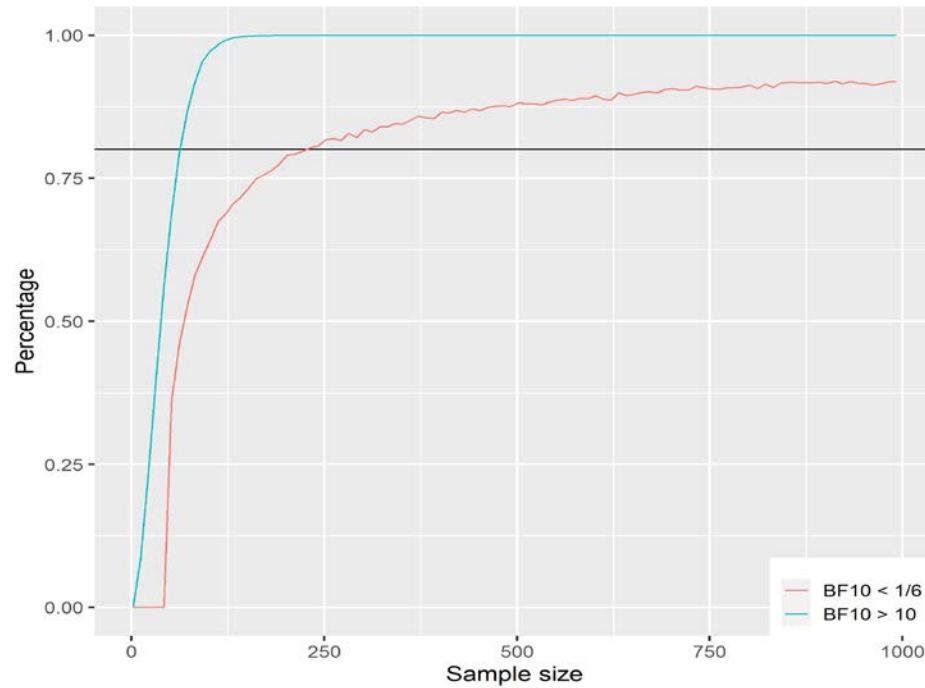
Keyesers et al. (2020). Nat. Neurosci.

# Classical (Frequentist) vs Bayesian Inference

Problems of Classical Inference (or advantages of “going Bayesian”):

- A “non-significant” p-value (e.g,  $p > .05$ ) does not mean there is no effect (“*absence of evidence is not evidence of absence*”)...
  - ...BFs can quantify evidence for Null ( $BF_{01} = 1/BF_{10}$ )
- A “significant” p-value can be found for unrealistic/trivial effect sizes...
  - ...BFs make reference to likely effect sizes...
- The more tests performed, the more likely a “Type I” error (when  $p < .05$  but  $H_0$  is true)...
  - ...BFs can be combined across data (or prior adjusted)
- You should specify sample size (stopping rule) in advance (you cannot “top-up” observations just to try to get  $p < .05$ )...
  - ...BFs reflect belief-updating, and allow Sequential Designs

# “Fixed N” design

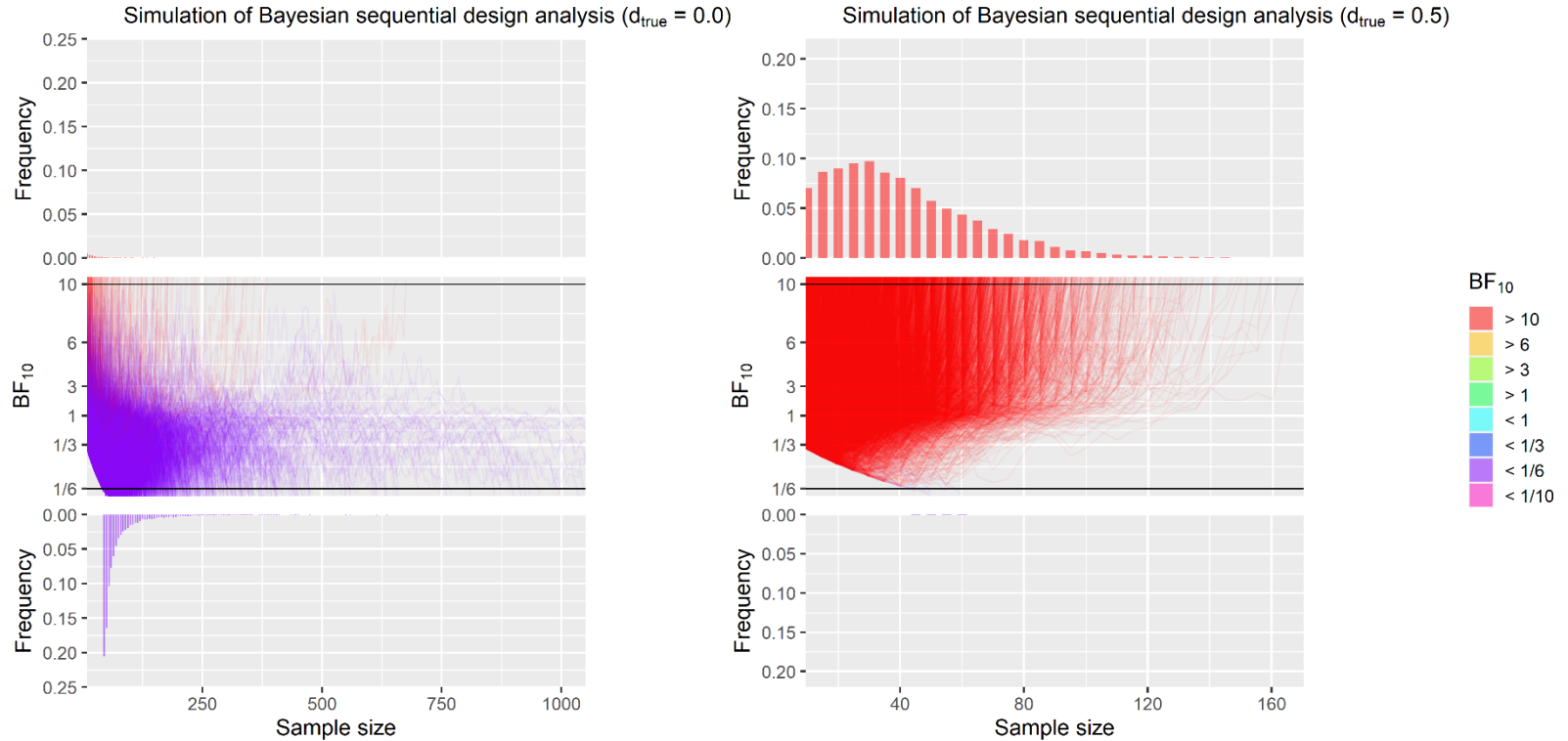


| Effect size | Max N | Misleading evidence | Strong evidence | Costs fMRI experiment |
|-------------|-------|---------------------|-----------------|-----------------------|
| 0.5         | 72    | 0.0003 %            | 80 %            | £ 39,600              |
| 0.0         | 232   | 0.0011 %            | 80 %            | £ 127,600             |

Slides thanks to Alex Quent

<https://jaquent.github.io/post/bayesian-sequential-designs-are-superior/>

# Sequential design

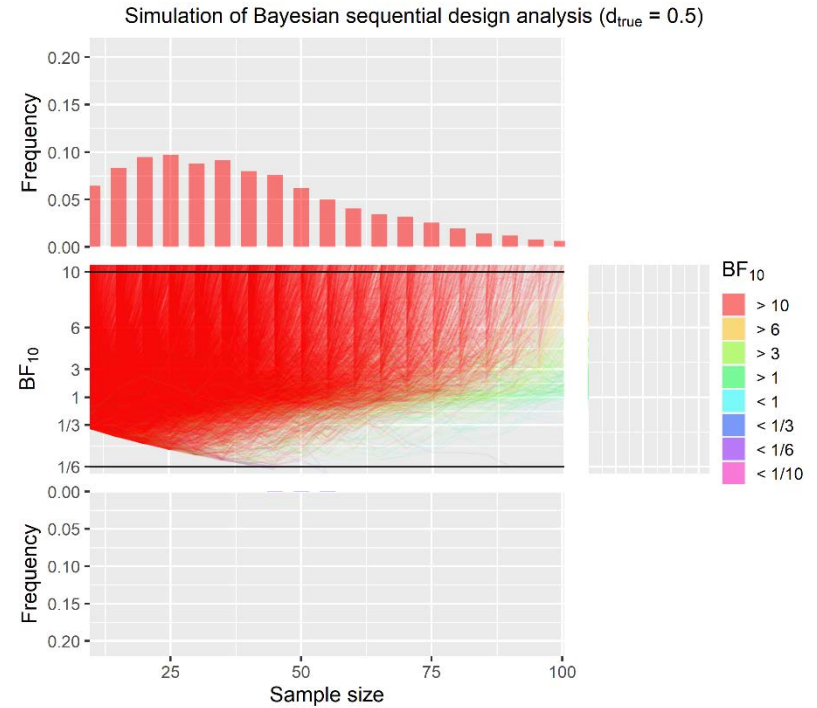
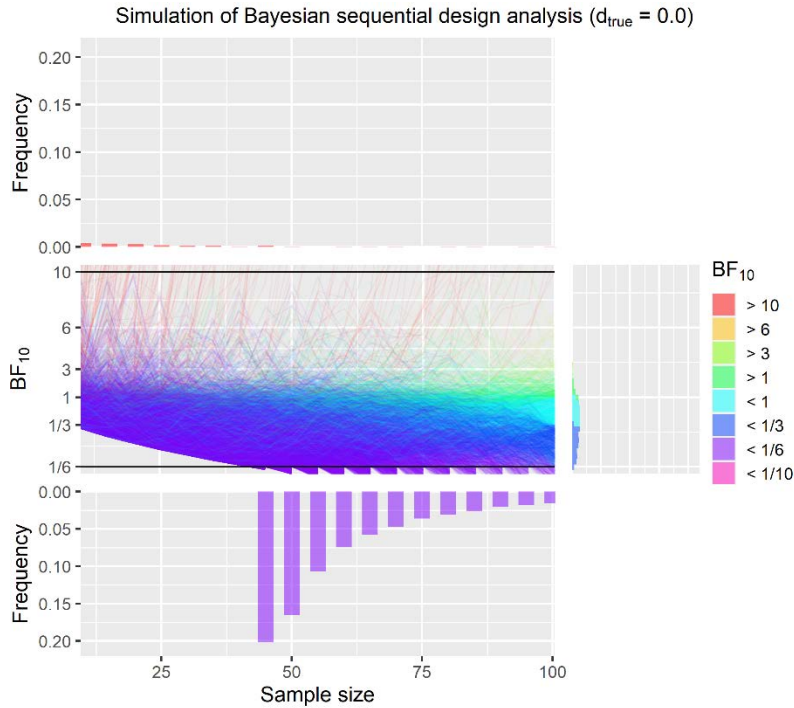


| Effect size | Max N | Mean N | Misleading evidence | Strong evidence | Costs fMRI experiment |
|-------------|-------|--------|---------------------|-----------------|-----------------------|
| 0.5         | 170   | 41     | 0.13 %              | 100 %           | £ 22,550              |
| 0.0         | 2765  | 83     | 2.95 %              | 100 %           | £ 45,650              |

Slides thanks to Alex Quant

<https://jaquent.github.io/post/bayesian-sequential-designs-are-superior/>

# “Max N” Sequential design



| Effect size | Max N | Mean N | Misleading evidence | Strong evidence | Costs fMRI experiment |
|-------------|-------|--------|---------------------|-----------------|-----------------------|
| 0.5         | 100   | 39     | 0.12 %              | 98 %            | £ 21,450              |
| 0.0         | 100   | 58     | 2.31 %              | 80 %            | £ 44,138              |

Slides thanks to Alex Quent

<https://jaquent.github.io/post/bayesian-sequential-designs-are-superior/>

# Comparison

## Fixed-N Design

| Effect size | Max N | Misleading evidence | Strong evidence | Costs fMRI experiment |
|-------------|-------|---------------------|-----------------|-----------------------|
| 0.5         | 72    | 0.0003 %            | 80 %            | £ 39,600              |
| 0.0         | 232   | 0.0011 %            | 80 %            | £ 127,600             |

## Sequential design

| Effect size | Max N | Mean N | Misleading evidence | Strong evidence | Costs fMRI experiment |
|-------------|-------|--------|---------------------|-----------------|-----------------------|
| 0.5         | 170   | 41     | 0.13 %              | 100 %           | £ 22,550              |
| 0.0         | 2765  | 83     | 2.95 %              | 100 %           | £ 45,650              |

## Sequential, Max-N design

| Effect size | Max N | Mean N | Misleading evidence | Strong evidence | Costs fMRI experiment |
|-------------|-------|--------|---------------------|-----------------|-----------------------|
| 0.5         | 100   | 39     | 0.12 %              | 98 %            | £ 21,450              |
| 0.0         | 100   | 58     | 2.31 %              | 80 %            | £ 44,138              |

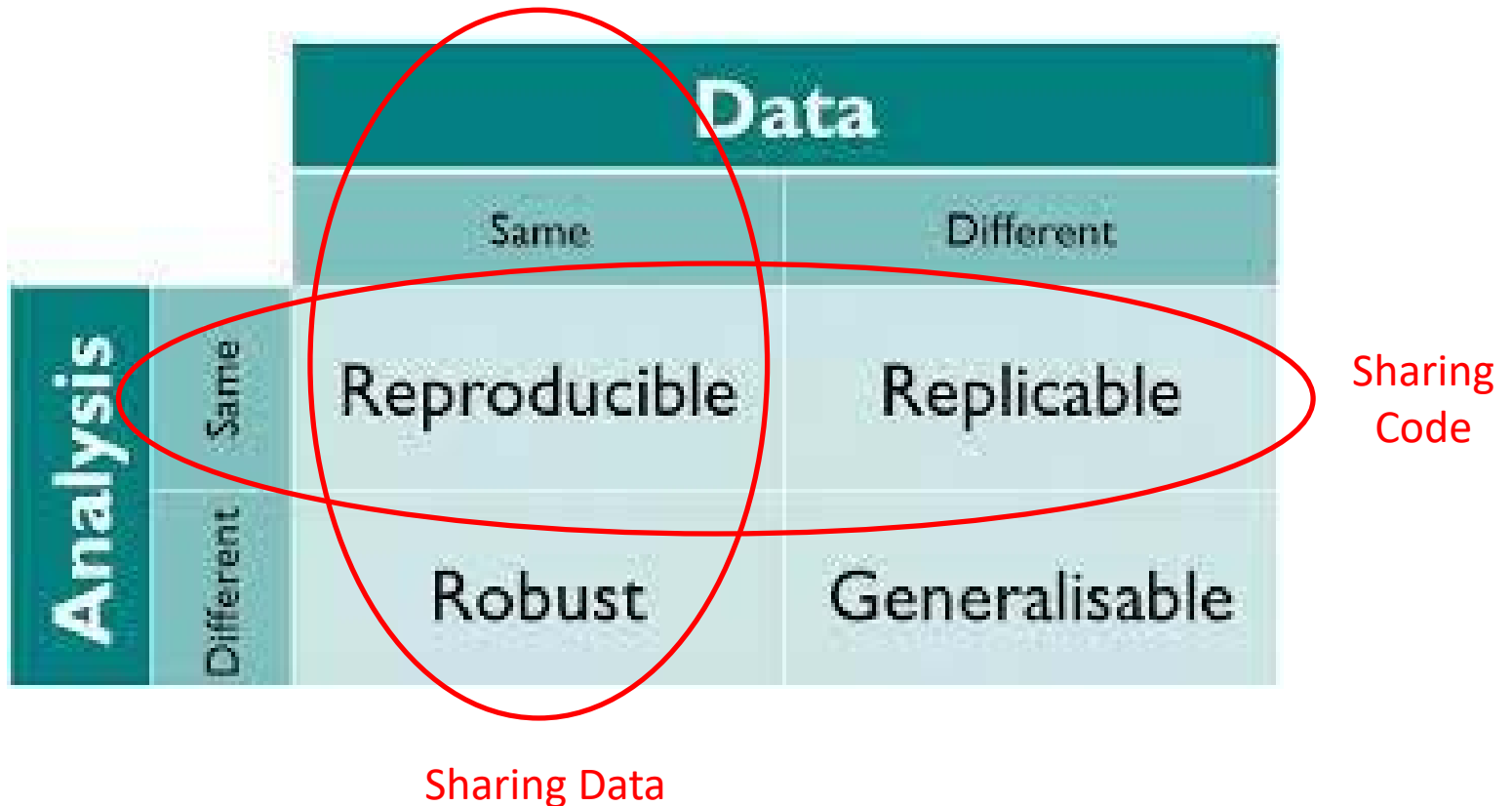
*Slides thanks to Alex Quent*

<https://jaquent.github.io/post/bayesian-sequential-designs-are-superior/>

# Overview

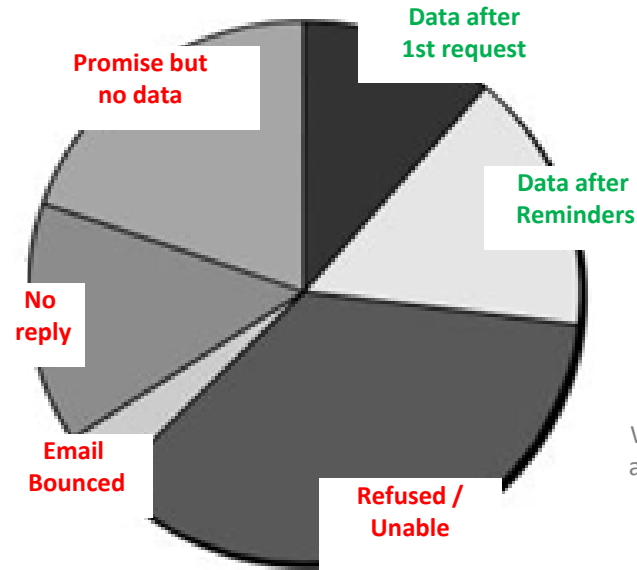
- Registration
- Statistical analysis
  - Power and PPV
  - Bayesian Statistics
  - Sequential Designs
- Sharing Data and Code
  - FAIR principles
  - Incentivising
  - GDPR
- Publication
- Research Culture

# Definitions





# Sharing Data



Wicherts, Borsboom, Kats & Molenaar (2006). The poor availability of psychological research data for reanalysis. *American Psychologist*, 61(7), 726.

- What to share?
- Where to share?
- How to share?
- When to share?

# What to Share?

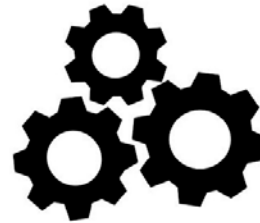
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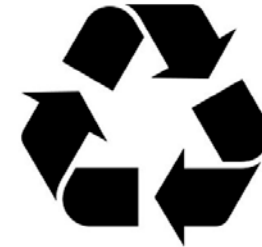
A  
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I  
nteroperable



R  
eusable



By SangyaPundir - Treball propi, CC BY-SA 4.0,  
<https://commons.wikimedia.org/w/index.php?curid=53414062>



Organization for  
Human Brain Mapping

Advancing Understanding of the Human Brain

Committee on Best Practice in Data Analysis and Sharing (COBIDAS)

<https://www.humanbrainmapping.org/i4a/pages/index.cfm?pageID=3728>

# What to Share?

- Sufficient for someone to reproduce your results
- Minimal: raw data and analysis scripts to results in paper
- Non-proprietary formats
- Conventional data formats, eg BIDS for neuroimaging
- Sufficient documentation (data paper?)



<https://bids.neuroimaging.io/>

# Where to Share?

- Small, non-personal or consented (GDPR) data:
  - open on personal website (but DOI?), university repository, OSF...
  - <http://neurovault.org> for imaging effect size maps
- Large, non-personal or consented data:
  - Public websites like <https://figshare.com/>, or <https://openneuro.org/> for neuroimaging
- Personal data with limited consent
  - managed access, electronic Data Usage Agreements (DUAs)
- Personal, unconsented data
  - only by ethical approval / collaboration agreement / DTA / re-consent
- Synthetic data with same statistical properties – anywhere!

# When to Share?

- As soon as possible, even pre-publication!
  - Unwarranted fear of scooping?

## **The Open Scoop Challenge**

*Posted 2014-02-25 by Greg Wilson in Community, Open Science.*

- During review (but reviewer anonymity?)
- Mandatory on publication!?
  - (In principle) reproducibility is a cornerstone of Science...

# Incentivising Sharing?

- Data Papers

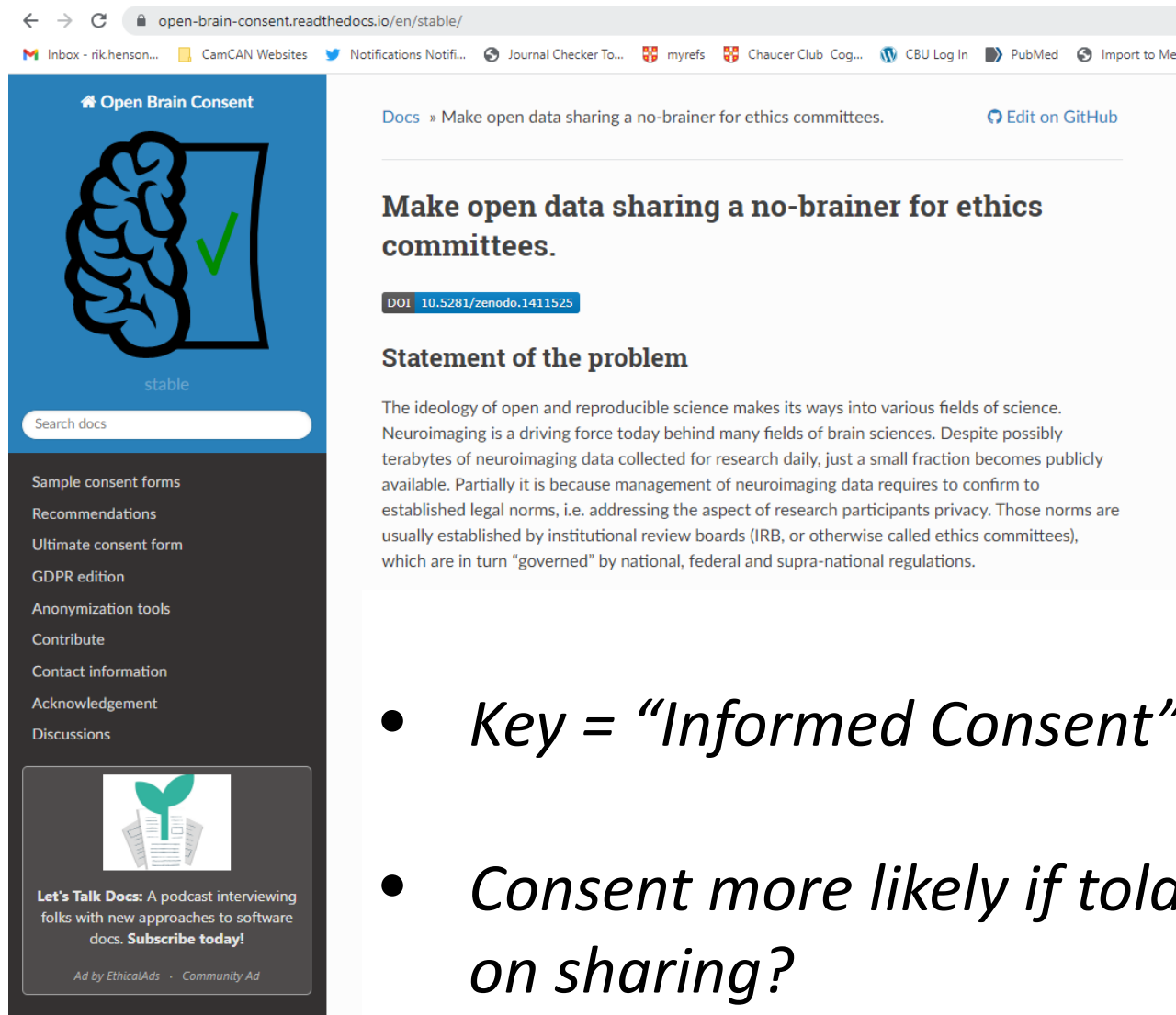


- Kite Marking

- Reproduction Papers (citation inheritance)?

The image displays three Open Badges, each with a distinct icon and a list of requirements:

- Open Data Badge** (blue icon with a bar chart):
  - A URL, doi, or other permanent path for accessing the data in a public, open-access repository
  - Sufficient information for an independent researcher to reproduce the reported results
- Open Materials Badge** (orange icon with a box):
  - A URL, doi, or other permanent path for accessing the materials in a public, open-access repository
  - Sufficient information for an independent researcher to reproduce the reported methodology
- Preregistration Badge\*** (red icon with a checkmark):



The screenshot shows a web browser displaying the Open Brain Consent website. The page title is "Make open data sharing a no-brainer for ethics committees." and it includes a DOI of 10.5281/zenodo.1411525. The page content discusses the challenges of open data sharing in neuroscience, particularly regarding privacy and ethics committees. A sidebar on the left contains navigation links for consent forms, recommendations, and tools. A footer advertisement for "Let's Talk Docs" is also visible.

Open Brain Consent

stable

Search docs

Sample consent forms

Recommendations

Ultimate consent form

GDPR edition

Anonymization tools

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Discussions

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Docs » Make open data sharing a no-brainer for ethics committees. [Edit on GitHub](#)

## Make open data sharing a no-brainer for ethics committees.

DOI 10.5281/zenodo.1411525

### Statement of the problem

The ideology of open and reproducible science makes its ways into various fields of science. Neuroimaging is a driving force today behind many fields of brain sciences. Despite possibly terabytes of neuroimaging data collected for research daily, just a small fraction becomes publicly available. Partially it is because management of neuroimaging data requires to conform to established legal norms, i.e. addressing the aspect of research participants privacy. Those norms are usually established by institutional review boards (IRB, or otherwise called ethics committees), which are in turn "governed" by national, federal and supra-national regulations.

- *Key = "Informed Consent"*
- *Consent more likely if told restrictions on sharing?*

# Managed Access



CBSU Publications x +

Not secure | www.mrc-cbu.cam.ac.uk/publications/opendata/

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#### Open Data Repository

This page shows all 88 data sets currently available in our Open Data repository

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Showing page < 3 > of 9



**The missing link? Testing a schema account of unitization**

Authors: TIBON, R., GREVE, A., HENSON, R.  
Reference: *Memory & Cognition*  
Year of publication: In Press  
CBSU number: 8267  
Open Data available, [click](#) to request

**Impact of Culture on Autobiographical Life Structure in Depression**

Authors: Jobson, L., Miskon, N., DALGLEISH, T., Hill, E., Golden-A-M., Zulkefly, S., & Mukhtar, F.  
Reference: *British Journal of Clinical Psychology*





# Managed Access

- I will receive access to de-identified data and will not attempt to establish the identity of, or attempt to contact any of the participants.
- I will not publish or disclose any information in a way that would allow the identity of any individual participants to become known.
- I will only use the data for the purposes of non-commercial, ethically approved research or teaching as specified above. I will seek approval from the MRC CBU if I wish to use the data for any other purpose.
- I agree to store the data securely.
- I will not disclose the data to any third parties beyond my immediate research team
- I will require any members of my team with whom I do share the data to comply with these terms and conditions
- I will comply with any rules and regulations imposed by my institution and its institutional review board when requesting and using the data.
- I understand that determining whether ethical approval is needed for the use of the data and gaining that approval is my responsibility.
- I understand that the CBU cannot guarantee exclusive use of these data or police potential overlaps of interest between researchers who request the data.
- I understand that it is my responsibility to check the data for errors, and that the MRC CBU is not responsible for any consequences of unreported errors in the data.
- I agree to make any errors that I discover in this data known to MRC CBU as soon as possible.
- I agree to acknowledge the MRC CBU in any output arising from the use of the data.
- I agree to make any publications that arise from use of the data open-access.
- I agree that should any data I derive from this data set appear in a publication, I will make that derived data, as well as any processing scripts used to produce that derived data, available on a suitable open-access data repository. I will also notify the MRC CBU where the data has been made available.

# (Dangers of Open Data?)



JOIN/RENEW

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SEARCH...

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Observer > 2013 > November > The Subterranean War on Science

## The Subterranean War on Science

STEPHAN LEWANDOWSKY, MICHAEL E. MANN, LINDA BAULD, GERARD HASTINGS, AND ELIZABETH F. LOFTUS

TAGS: COGNITIVE PSYCHOLOGY | EXPERIMENTAL PSYCHOLOGY | FALSE MEMORY | PREJUDICE

Science denial kills. More than 300,000 South Africans died needlessly in the early 2000s because the government of President Mbeki preferred to treat AIDS with garlic and beetroot rather than antiretroviral drugs (Chigwedere, Seage, Gruskin, Lee, & Essex, 2008). The premature death toll from tobacco is staggering and historians have shown how it was needlessly inflated by industry-sponsored denial of robust medical evidence (Proctor, 2011). The US now faces the largest outbreak of whooping cough in decades, in part because of widespread denial of the benefits of vaccinations (Rosenau, 2012). According to the World Health Organization, climate change is already claiming more than 150,000 lives annually (Patz, Campbell-Lendrum, Holloway, & Foley, 2005), and estimates of future migrations triggered by unmitigated global warming run as high as 187 million refugees (Nicholls et al., 2011). A common current attribute of denial is that it side-steps the peer-reviewed literature and relies on platforms such as internet blogs or tabloid newspapers to disseminate its dissent from the scientific mainstream. In contrast, the publication of dissenting views in the peer-reviewed literature does not constitute denial.

The tragic track record of denial has stimulated research into its political, sociological, and psychological underpinnings (Dunlap, 2013; Jacobson, Targonski, & Poland, 2007;



## About the Authors

**Stephan Lewandowsky** is with the Department of Psychology at the University of Bristol, UK, and University of Western Australia; **Michael E. Mann** is with the Departments of Meteorology & Geosciences at Penn State University; **Linda Bauld** and **Gerard Hastings** are with the Centre for Tobacco Control Research at the University of Stirling, UK; and **Elizabeth F. Loftus** is with the Department of Psychology and Social Behavior at the University of California, Irvine.

## Related



### MYTH: EYEWITNESS TESTIMONY IS THE BEST KIND OF EVIDENCE

Activities in this unit reveal how eyewitness testimony is subject to unconscious memory distortions and biases even

among the most confident of witnesses. ... More



### MYTH: TRAUMATIC MEMORIES ARE OFTEN REPRESSED AND LATER RECOVERED

This provides students with an opportunity to see that, often, analyses may lead to conclusions

# Data Usage Agreement (DUA)



MRC Cognition  
and Brain  
Sciences Unit



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CBSU Publications

## Data Request

Title

Data Set ID

Please complete the details below to request a copy of this data set. All fields are required, and requests cannot be considered if any of the fields are blank. Once we receive your request it will be evaluated by a member of staff based on criteria such as the scientific merit of the proposed data use. We will contact you with the result of this evaluation, and if your request is approved we will provide you with a link you can use to download a copy of the data.

Name

Institution

E-mail

Please outline the project for which the data are requested. Please include details of the scientific questions addressed, methods used, publication strategy, the organisation funding the research, and how data sources, funders, etc will be acknowledged

[By requesting this data set, you agree to the following terms and conditions:](#)

- I will receive access to de-identified data and will not attempt to establish the identity of, or attempt to contact any of the participants.
- I will not publish or disclose any information in a way that would allow the identity of any individual participants to become known.
- I will only use the data for the purposes of non-commercial, ethically approved research or teaching as specified above. I will seek approval from the MRC CBSU if I wish to use the data for any other purpose.
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# Open Code



<https://github.com/>

Free  
Version Control  
Multiple Users



<https://software-carpentry.org/>

R (for statistics)



Python (for anything!)



# Overview

- Registration
- Statistical analysis
- Sharing Data and Code
  - FAIR principles
  - Incentivising
  - GDPR
- Publication
  - Open Access
  - Preprints
  - Open Review
- Research Culture

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▶ 1:04:48    vimeo



Commission and its priorities

Policies, information and services



English (en) v

Search

Home > ... > Carlos Moedas > Announcements > 'Plan S' and 'cOAlition S' – Accelerating the transition to full and immediate Open Access to scientific publications

STATEMENT | 4 September 2018

# 'Plan S' and 'cOAlition S' – Accelerating the transition to full and immediate Open Access to scientific publications

With the increasing pace of scientific discovery and growing public demand for reliable information, there has never been a greater need for immediate, universal, access to the latest research findings. But with many scientific journals behind paywalls not everyone can get hold of this knowledge. 'Knowledge is power' and I firmly believe that free access to all scientific publications from publicly funded research is a moral right of citizens. Two years ago, on 27 May 2016, all Member States of the European Union committed to achieve this goal by 2020. It is one of the most important political commitments on science of recent times and puts Europe at the forefront of the global transition to open

- UKRI adopted April 2022
- Similar initiative in US





# Open Publication

- Open Access (OA): Public-funded (tax payer) money?
- Gold OA but minimal Author Processing Charge (APC)?
- Free journals (funded by government eg UKRI)?
- More radical solutions, eg Octopus, <https://www.octopus.ac/>



## An emerging consensus for open evaluation: 18 visions for the future of scientific publishing

**Nikolaus Kriegeskorte<sup>1\*</sup>, Alexander Walther<sup>1</sup> and Diana Deca<sup>2</sup>**

<sup>1</sup> Medical Research Council Cognition and Brain Sciences Unit, Cambridge, UK

<sup>2</sup> Institute of Neuroscience, Technische Universität München, Munich, Germany

\*Correspondence: [nikokriegeskorte@gmail.com](mailto:nikokriegeskorte@gmail.com)

**Edited by:**

Misha Tsodyks, Weizmann Institute of Science, Israel

**Reviewed by:**

Misha Tsodyks, Weizmann Institute of Science, Israel

bioRxiv

THE PREPRINT SERVER FOR BIOLOGY

<https://www.biorxiv.org/>

- Pros and Cons of peer-review as gateway to publication

Pro: Publication Bias (Pfizer review AZ's paper submitted to FDA?)

Con? COVID examples

# Open Review

- Post-publication of Reviews (eg PubMed Commons, F1000)...
- ...continuing dialogue linked to original paper (“conversation”)



PubMed Commons is a system that enables researchers to share their opinions about scientific publications. Researchers can comment on any publication indexed by PubMed, and read the comments of others. PubMed Commons is a forum for open and constructive criticism and discussion of scientific issues. It will thrive with high quality interchange from the scientific community. PubMed Commons is currently in a closed pilot testing phase, which means that only invited participants can add and view comments in PubMed.

[Adding comments](#) [Usage guidelines](#) [Invite an author](#) [How do I join?](#) [FAQ](#)

- Double-blind Reviews...
- ...or even identified Reviewers (or unique ID within system?)
- Publish reviews



<https://asapbio.org/publishyourreviews>

- Quality of Reviews – overworked, incentivize (£, or CVs, eg Publons)

## Transparency and Openness Promotion (TOP) guidelines

<https://www.cos.io/initiatives/top-guidelines>



↓ HARKing      ↓ P-hacking  
+ Feedback

Null results



Materials; Code

↑ Reliability  
↓ Variability



Re-analysis + pooling data

Quality control

# Overview

- Registration
- Statistical analysis
- Sharing Data and Code
- Publication
  - Open Access
  - Preprints
  - Open Review
- **Research Culture**
  - DORA
  - CRediT
  - Narrative CVs

# Publish or Perish

## 19th century scientist

I must find the  
explanation for this  
phenomenon in order  
to truly understand  
Nature...



## 21st century scientist

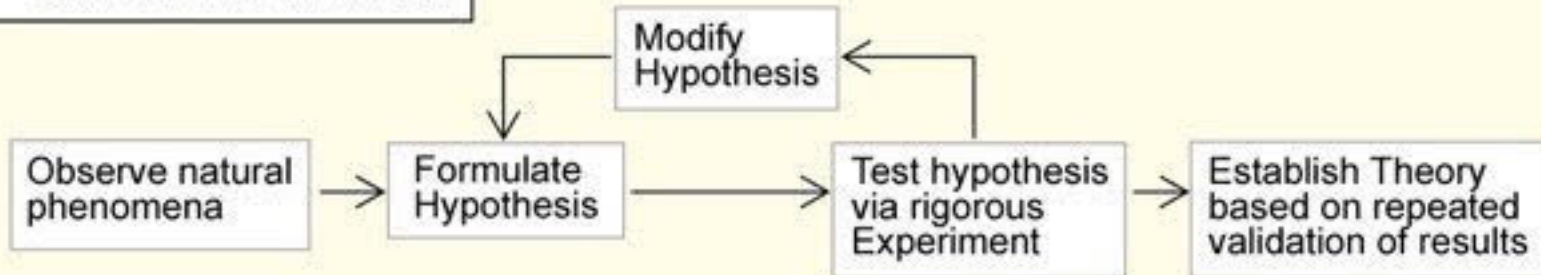
I must get the  
result that fits my  
narrative so I can  
get my paper into  
Nature..



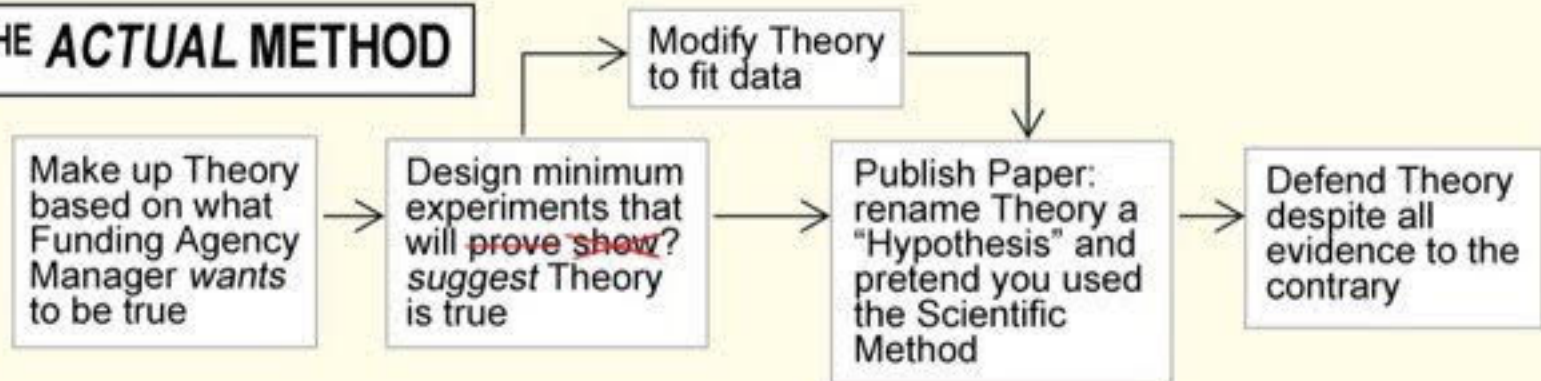
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## Real Scientific Method

### THE SCIENTIFIC METHOD



### THE ACTUAL METHOD

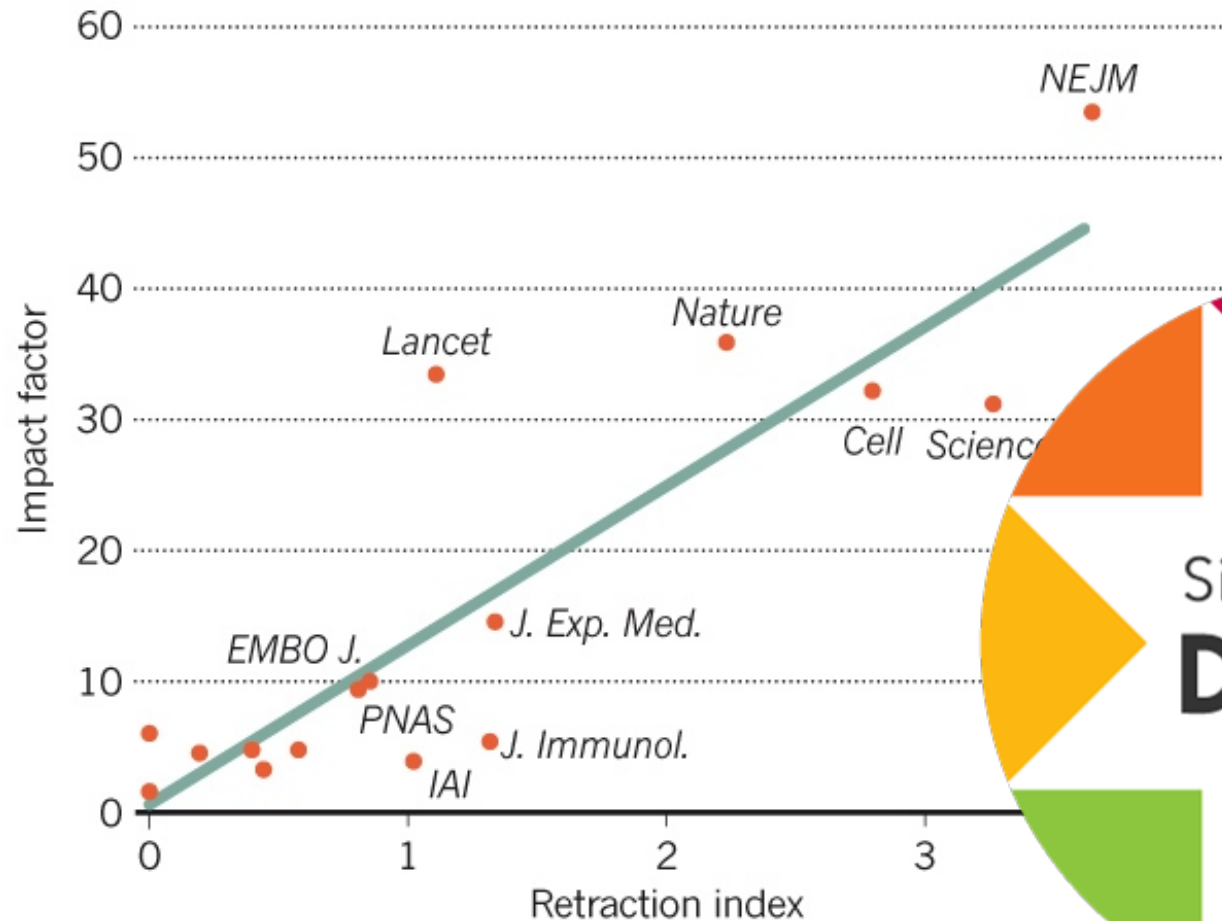


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# Impact Factor

## RETRACTION RELATION

Journals with higher impact factors also have a higher rate of retractions.





# Other Issues

- CRediT (Contributor Roles Taxonomy):

*Eg: “Zhang San: Conceptualization, Methodology, Software Priya Singh.: Data curation, Writing- Original draft preparation. Wang Wu: Visualization, Investigation. Jan Jansen: Supervision.: Ajay Kumar: Software, Validation.: Sun Qi: Writing-Reviewing and Editing”*

- Reward «team science», eg corporate authorship



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## Cam-CAN Corporate Authorship Membership

14. Project principal personnel: Lorraine K Tyler, Carol Brayne, Edward T Bullmore, Andrew C Calder, Rhodri Cusack, Tim Dalgleish, John Duncan, Richard N Henson, Fiona E Matthews, William D Marslen-Wilson, James B Rowe, Meredith A Shafto; Research Associates: Karen Campbell, Teresa Cheung, Simon Davis, Linda Geerligs, Rogier Kievit, Anna McCarrey, Abdur Mustafa, Darren Price, David Samu, Jason R Taylor, Matthias Treder, Kamen A Tsvetanov, Janna van Belle, Nitin Williams, Daniel Mitchell, Simon Fisher, Else Eising, Ethan Knights; Research Assistants: Lauren Bates, Tina Emery, Sharon Erzinçlioglu, Andrew Gadie, Sofia Gerbase, Stanimira Georgieva, Claire Hanley, Beth Parkin, David Troy; Affiliated Personnel: Tibor Auer, Marta Correia, Lu Gao, Emma Green, Rafael Henriques; Research Interviewers: Jodie Allen, Gillian Amery, Liana Amunts, Anne Barcroft, Amanda Castle, Cheryl Dias, Jonathan Dorrack, Melissa Fair, Hayley Fisher, Anna Goulding, Adarsh Grewal, Geoff Hale, Andrew Hilton, Frances Johnson, Patricia Johnston, Thea Kavanagh-Williamson, Magdalena Kwasniewska, Alison McMinn, Kim Norman, Jessica Penrose, Fiona Roby, Diane Rowland, John Sargeant, Maggie Squire, Beth Stevens, Aldabra Stoddart, Cheryl Stone, Tracy Thompson, Ozlem Yazlik; and administrative staff: Dan Barnes, Marie Dixon, Jaya Hillman, Joanne Mitchell, Laura Willis.

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# Alternative CVs

- Narrative CVs
  - Royal Society’s “Resume for Researchers” (R4R)
    - How have you contributed to: 1) knowledge, 2) develop individuals, 3) research community, 4) society?
  - Description of best work; no Impact Factors!
- Employers:
  - Read papers rather than note journal
  - Recruitment & Promotion: seek evidence of commitment to Open Practices
  - Reward team/community/support work – “scientific citizenship”

# Guidance/Hope



<https://osf.io/>



<https://www.ukrn.org/>



<https://reproducibilitatea.org/>



<https://www.bnacredibility.org.uk/>

# Overview

- Registration
  - Study Registration (eg OSF)
  - Registered Reports
  - Pre-Registration Posters
- Statistical analysis
  - Power and PPV
  - Bayesian Statistics
  - Sequential Designs
- Sharing Data and Code
  - FAIR principles
  - Incentivising
  - GDPR
- Publication
  - Open Access
  - Preprints
  - Open Review
- Research Culture
  - DORA
  - CRediT
  - Narrative CVs