

CAWS Symposium 2017

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More than 3Rs

Improving the validity and reproducibility of animal research

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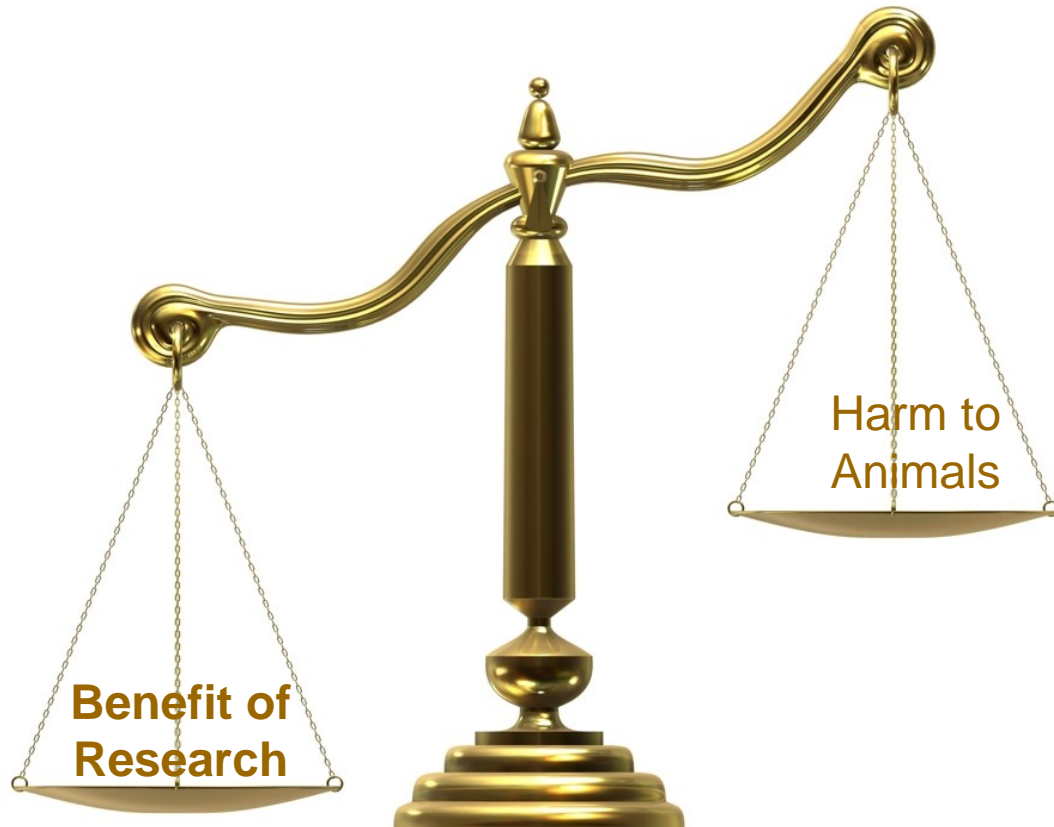
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Harm-benefit analysis

Principle of proportionality:

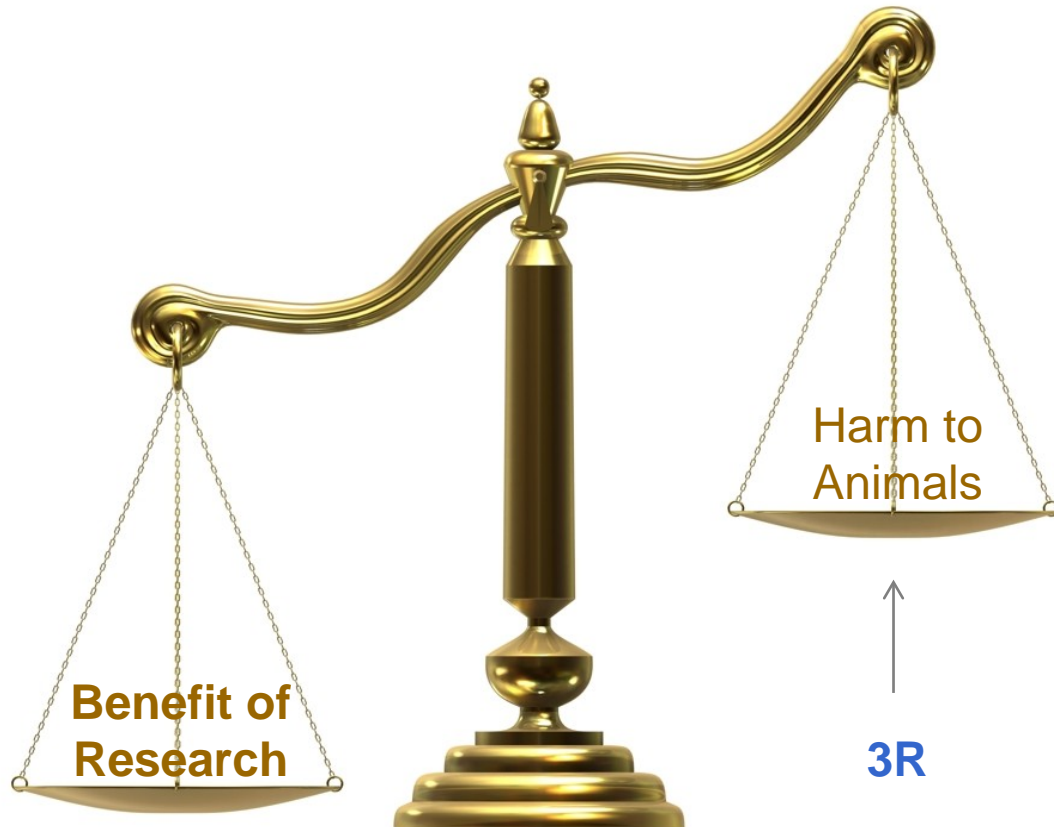
- Do the expected benefits outweigh the harms imposed on the animals?



Harm-benefit analysis

Preconditions:

- Is the study **necessary** for achieving the expected benefit?



Harm-benefit analysis

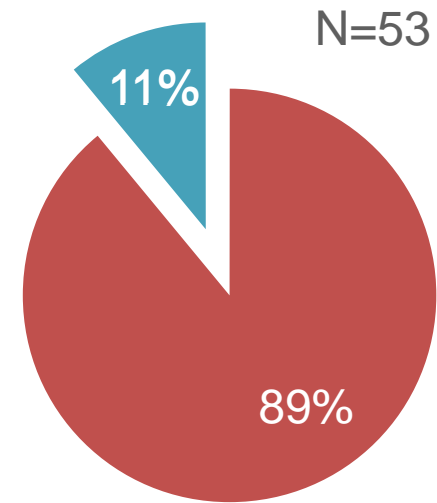
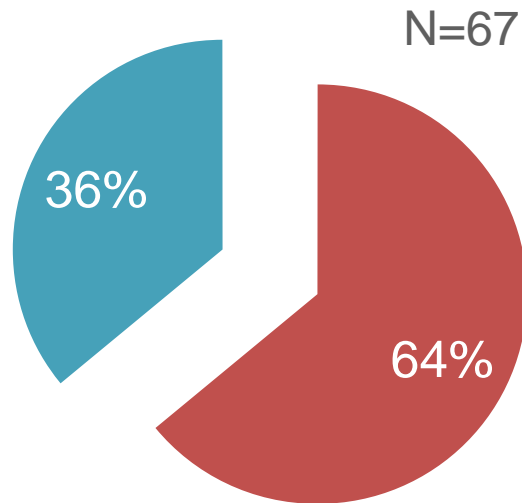
Preconditions:

- Is the study **necessary** for achieving the expected benefit?
- Is the study **suitable** for achieving the expected benefit?

scientifically valid?
reproducible?



Spectacular failures to replicate pre-clinical studies



IS THERE A REPRODUCIBILITY CRISIS?

A *Nature* survey lifts the lid on
how researchers view the 'crisis'
rocking science and what they
think will help.

BY MONYA BAKER

52%
Yes, a significant
crisis

38%
Yes, a slight
crisis

7%
Don't know

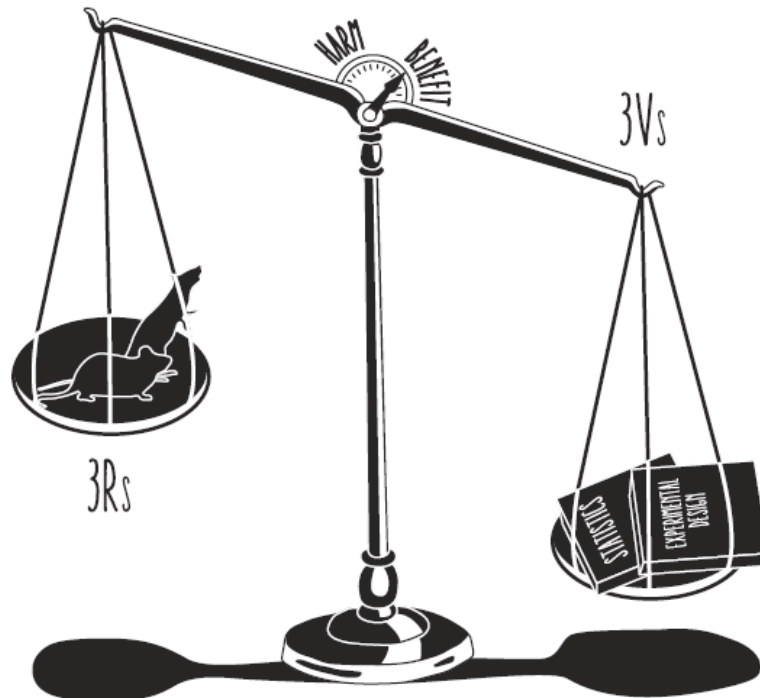
3%
No, there is no crisis

1,576
RESEARCHERS SURVEYED

Potential problems with scientific validity

The 3Vs

- Poor construct validity → poorly validated animal models and outcomes
- Poor internal validity → poor experimental design and conduct
- Poor external validity → poor representation of target population



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Focus on Reproducibility

REVIEW

Introducing Therioepistemology: the study of how knowledge is gained from animal research

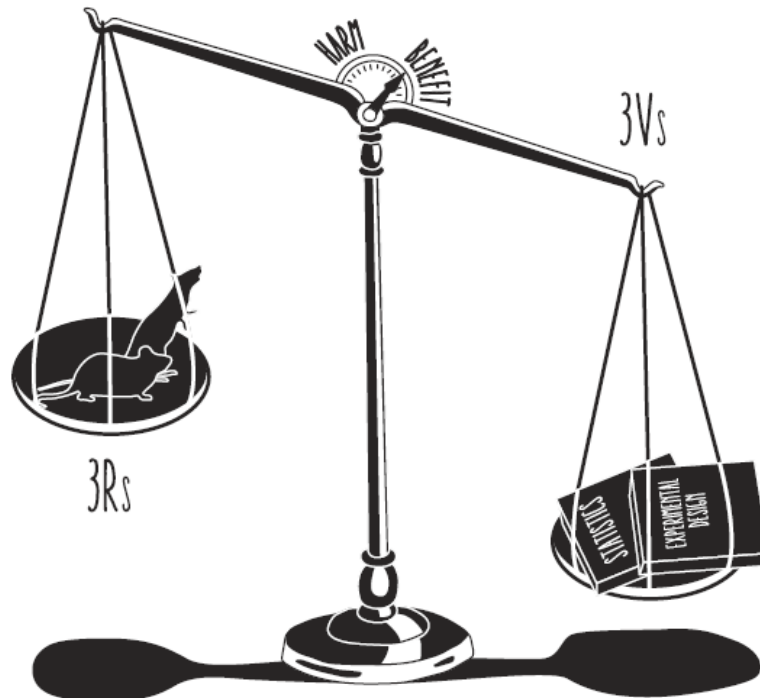
Joseph P Garner^{1,2}, Brianna N Gaskill³, Elin M Weber¹, Jamie Ahloy-Dallaire¹ & Kathleen R Pritchett-Corning^{4,5}

LabAnimal Volume 46, No. 4 | APRIL 2017 103-113

Potential problems with scientific validity

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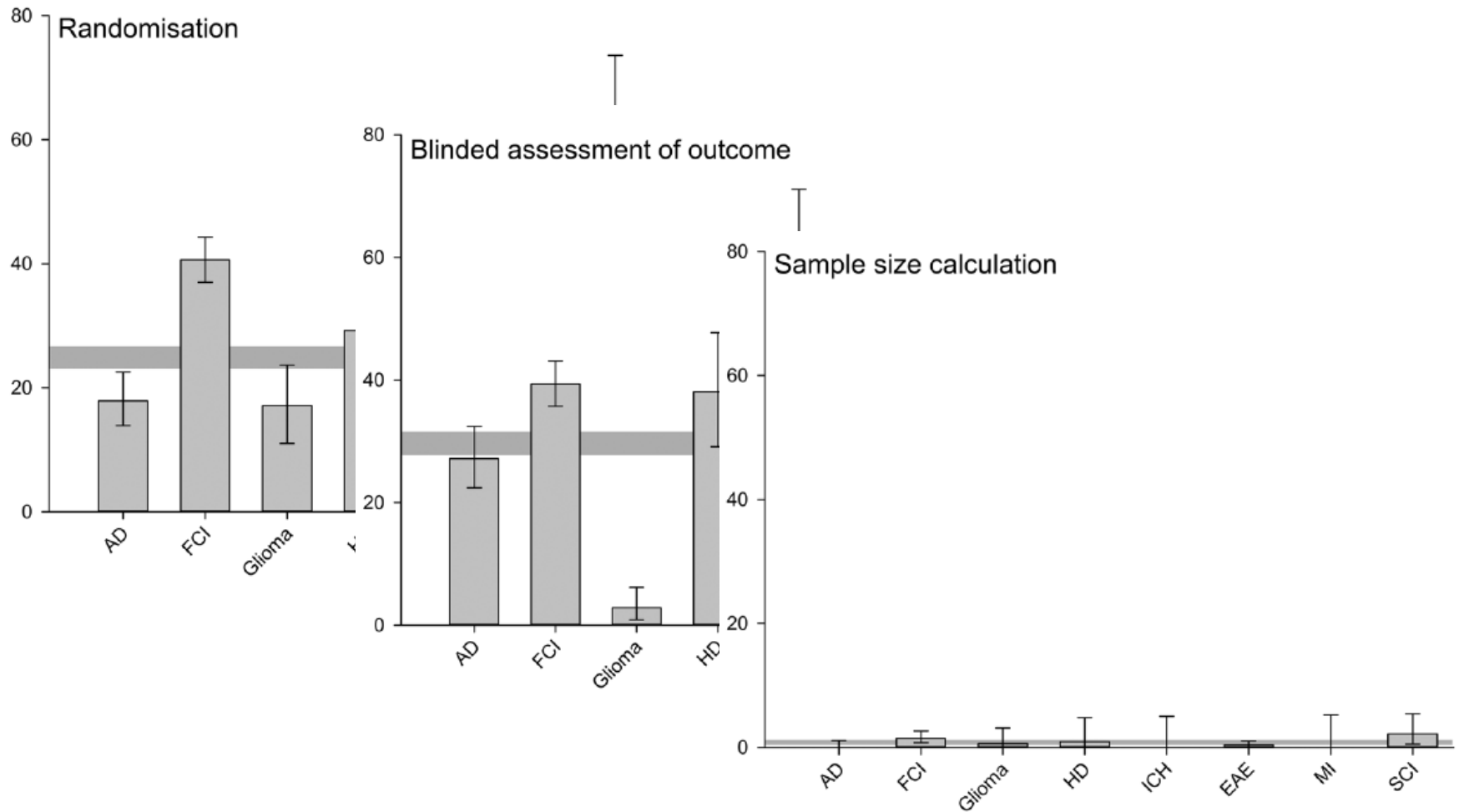
Internal validity

Measures to avoid risks of bias

- Randomisation → selection bias
- Blinding → detection bias
- Sample size calculation → statistical power
- Inclusion and exclusion criteria → attrition bias
- Definition of primary outcome variable → selective reporting
- Statistical analysis plan → analytical flexibility
- ...

→ **The commands of good research practice**

Poor reporting of risks of bias in animal research



Perspective

Improving Bioscience Research Reporting: The ARRIVE Guidelines for Reporting Animal Research

Carol Kilkenny^{1*}, William J. Browne², Innes C. Cuthill³, Michael Emerson⁴, Douglas G. Altman⁵

1 The National Centre for the Replacement, Refinement and Reduction of Animals in Research, London, United Kingdom, **2** School of Veterinary Science, University of Bristol, Bristol, United Kingdom, **3** School of Biological Sciences, University of Bristol, Bristol, United Kingdom, **4** National Heart and Lung Institute, Imperial College London, United Kingdom, **5** Centre for Statistics in Medicine, University of Oxford, Oxford, United Kingdom

In the last decade the number of bioscience journals has increased enormously, with many filling specialised niches reflecting new disciplines and technologies. The emergence of open-access journals has revolutionised the publication process, maximising the availability of research data. Nevertheless, a wealth of evidence shows that across many areas, the reporting of biomedical research is often inadequate, leading to the view that even if the science is

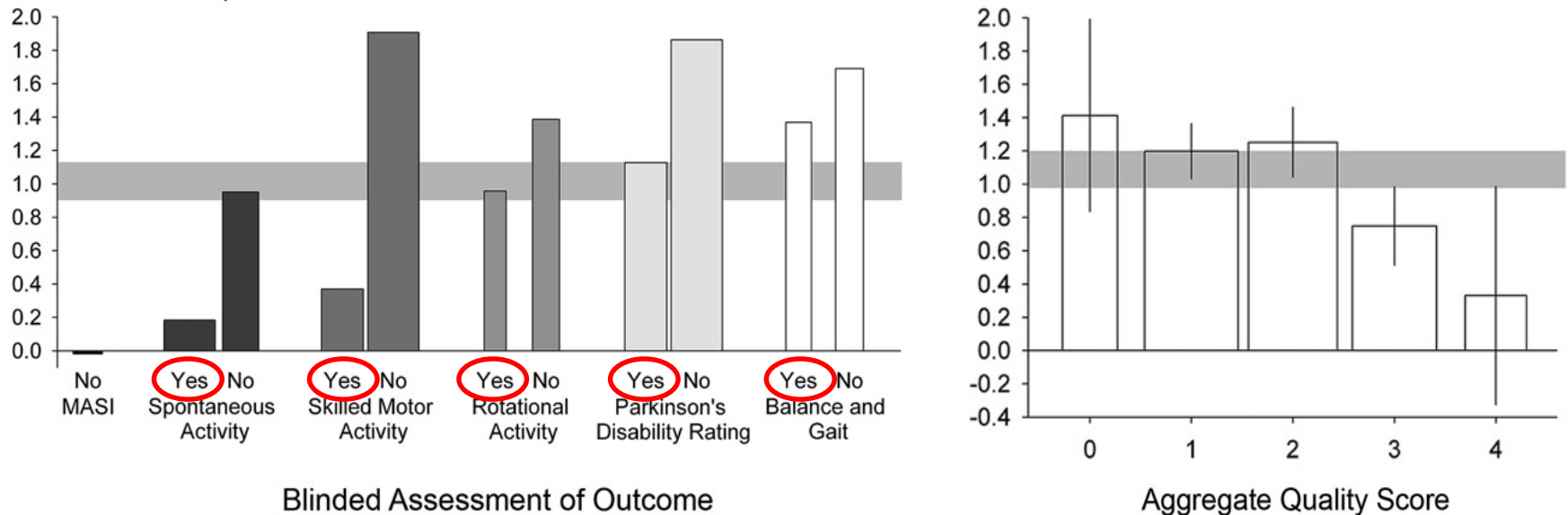
animals used (i.e., species/strain, sex, and age/weight). Most of the papers surveyed did not report using randomisation (87%) or blinding (86%) to reduce bias in animal selection and outcome assessment. Only 70% of the publications that used statistical methods fully described them and presented the results with a measure of precision or variability [5]. These findings are a cause for concern and are consistent with reviews of many research areas,

the study and the reliability and validity of the findings. There should also be enough information to allow the experiment to be repeated [23]. The problem therefore is how to ensure that all relevant information is included in research publications.

Using Reporting Guidelines Measurably Improves the Quality of Reporting

Poor reporting is associated with overstated effects

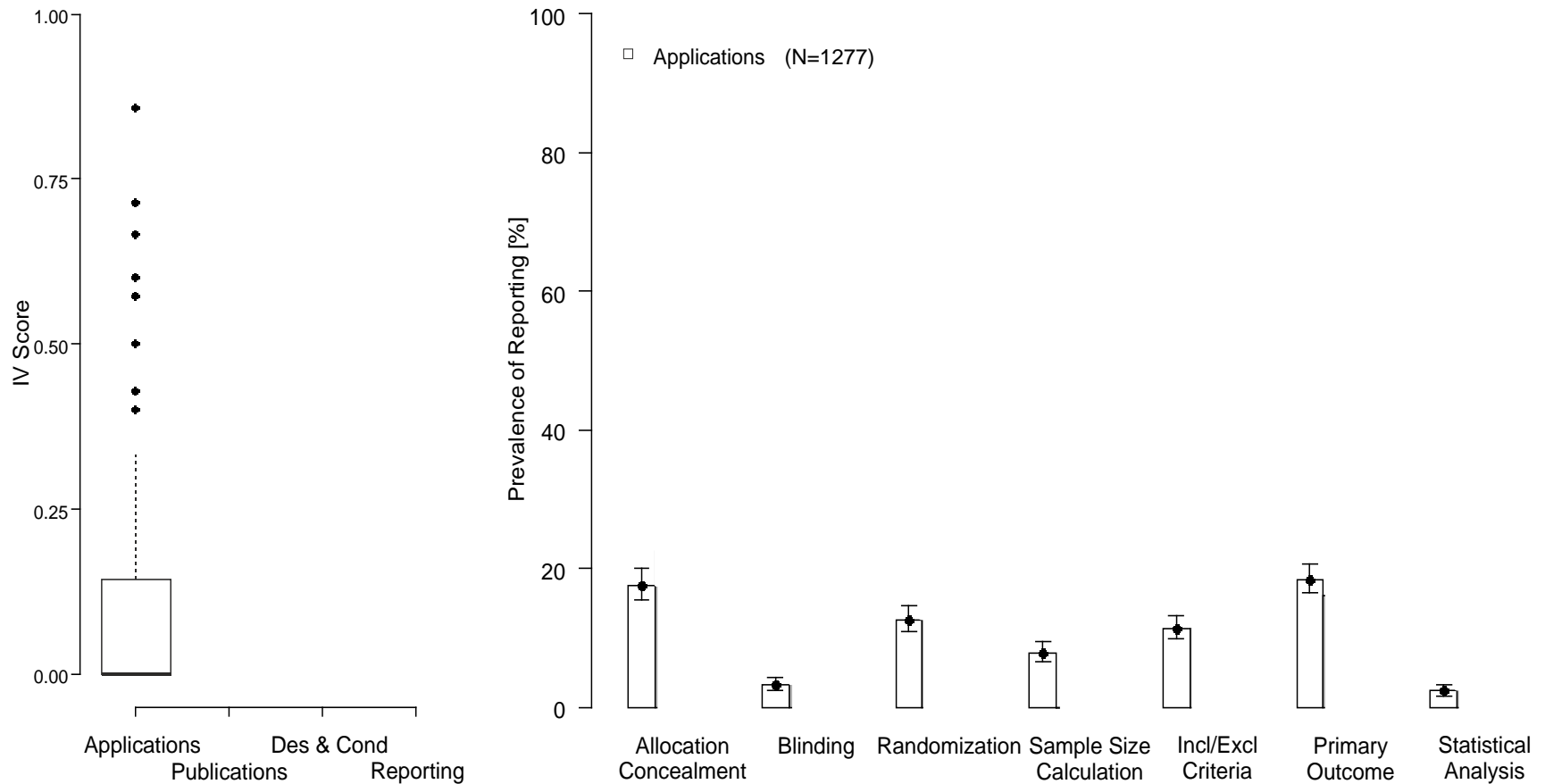
Efficacy of Dopamine agonists in animal models of Parkinson's Disease



→ Studies reporting blinded outcome assessment found smaller treatment effects

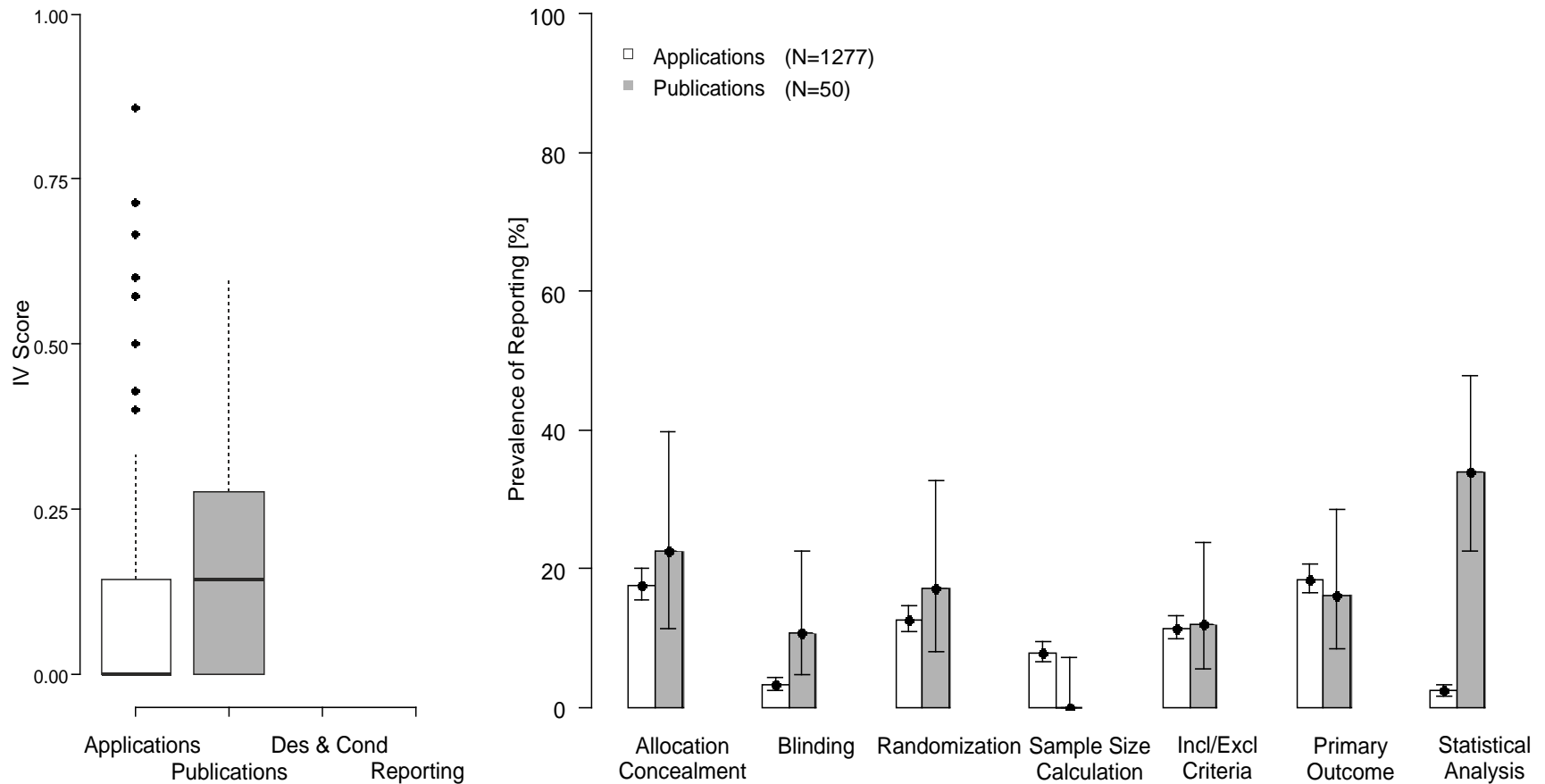
Reporting in applications for animal experiments in CH

Internal validity score and measures to avoid risks of bias



Reporting in publications of animal experiments in CH

Internal validity score and measures to avoid risks of bias



Conclusions

- Results confirm poor reporting of measures against risks of bias
- Authorities approving animal experiments and editors publishing papers lack information to assess risks of bias

Vogt et al. 2016 **PLoS Biol**

Use and reporting of measures against bias in CH

Online Survey

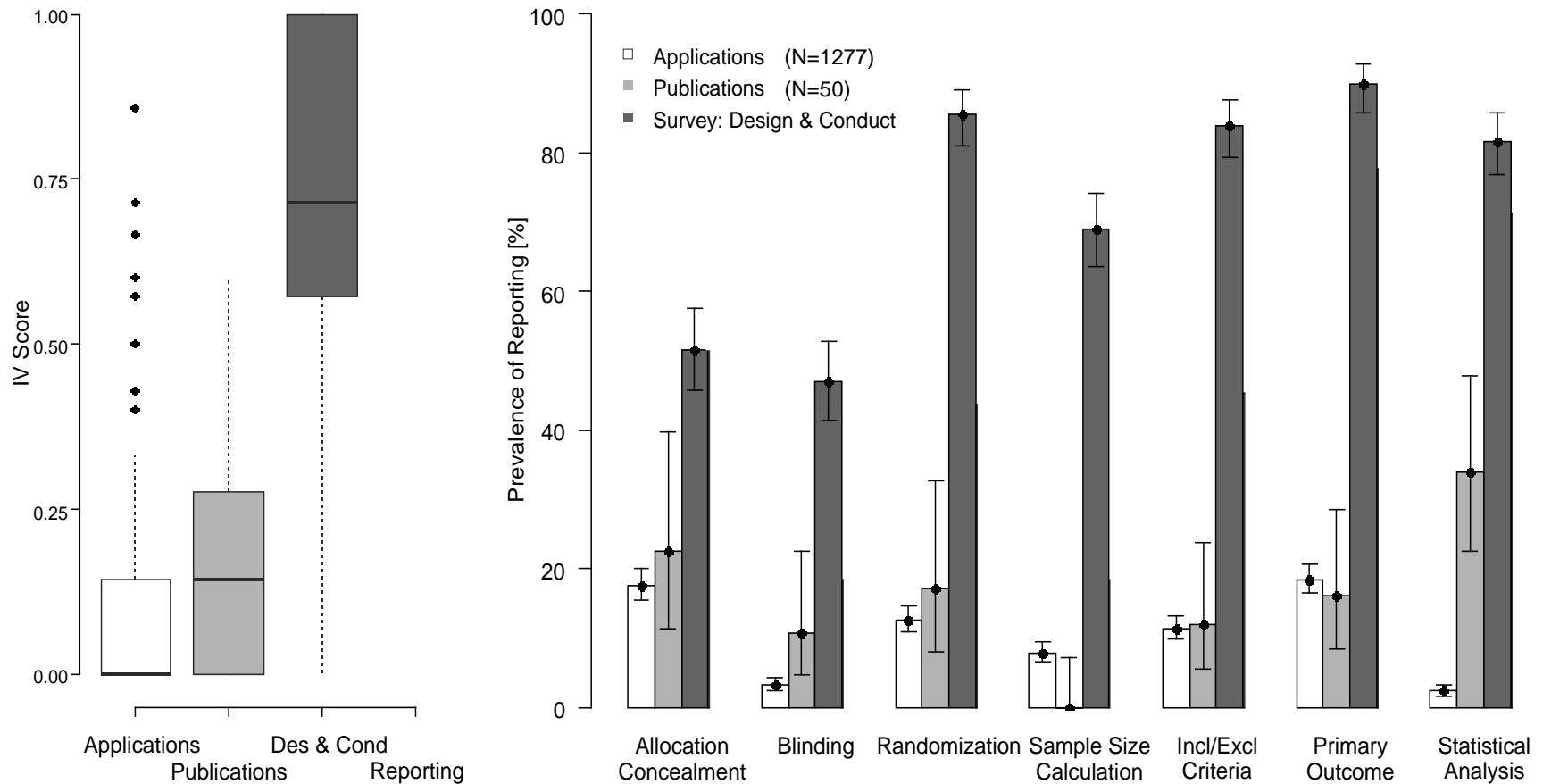
- Total sample (project leaders, experimenters): N=1891
- Return rate: 28% (N=530)
- Fully completed questionnaires: N=302 (16% of total sample)

→ “*what measures against bias do you generally use in your research?*”

→ “*which of these measures have you reported in your latest publication?*”

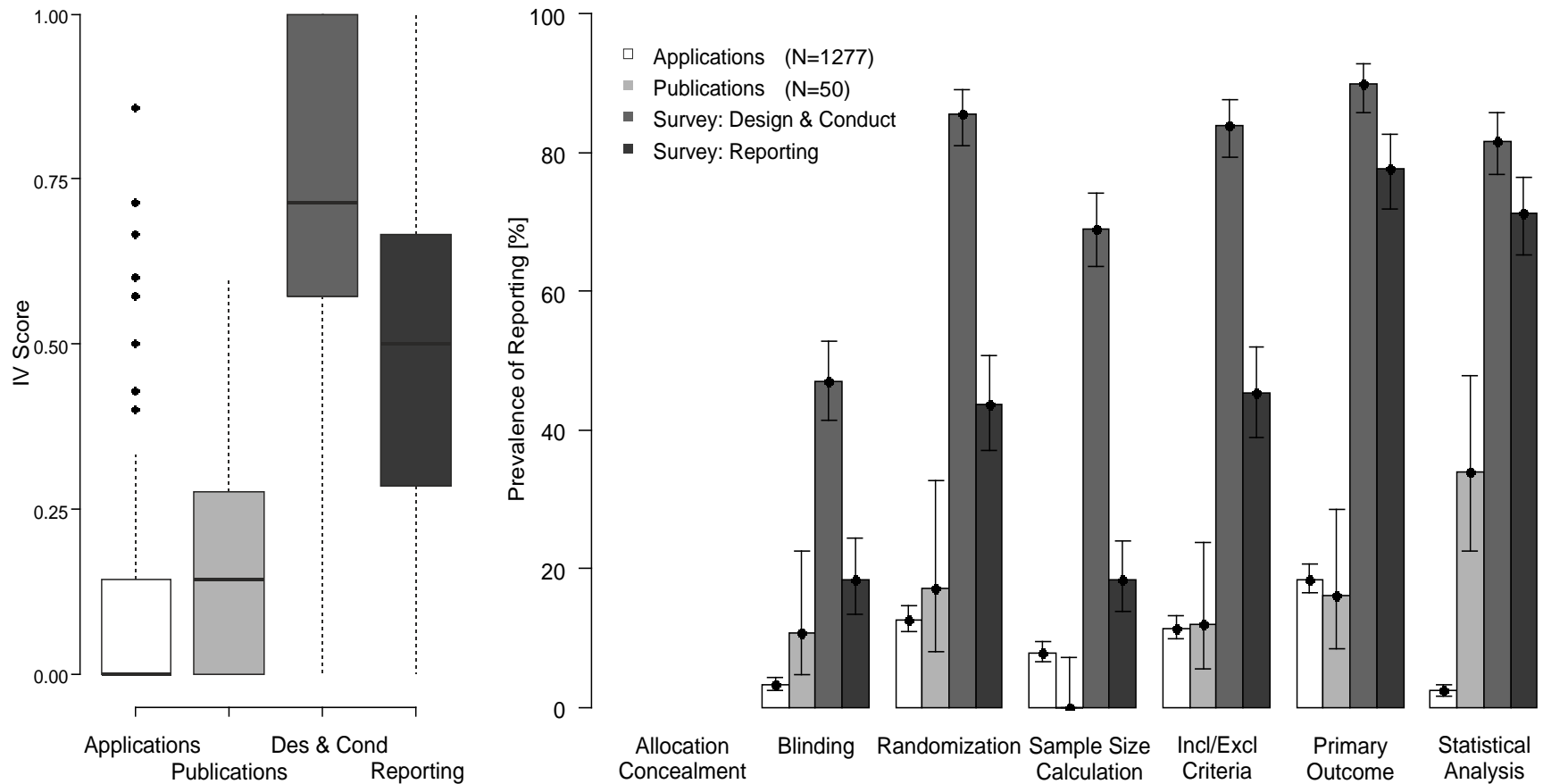
Use of measures against risks of bias in CH

Internal validity score and measures to avoid risks of bias



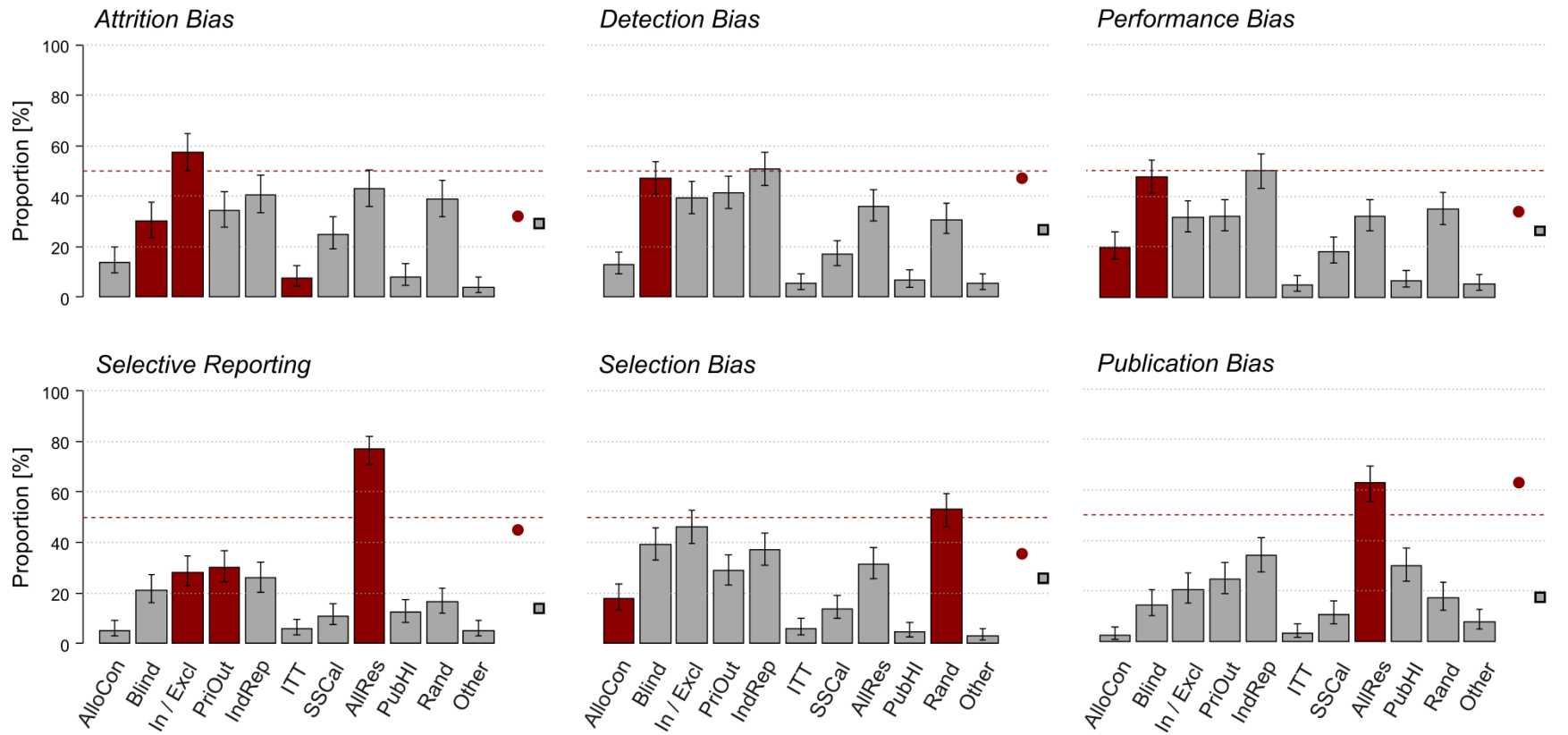
Reporting of measures against risks of bias in CH

Internal validity score and measures to avoid risks of bias



Use of measures against risks of bias in CH

Participants' knowledge of measures to avoid risks of bias



Conclusions

B

- Results confirm poor reporting of measures against risks of bias
- Authorities approving animal experiments and editors publishing papers lack information to assess risks of bias

Vogt et al. 2016 **PLoS Biol**

- Researchers overestimate scientific rigor of their own research
- Researchers lack problem awareness and knowledge to avoid risks of bias

Reichlin et al. 2016 **PLoS ONE**

Problems associated with lack of scientific rigor

- Inferential statistics applied to hypothesis-free fishing expeditions
- Searching for p-values < 0.05 (P-hacking)
- Hypothesizing After the Results are Known (HARKing)
- Selective reporting of outcomes
- Publishing “positive” results only (Publication bias)

Problems associated with lack of scientific rigor

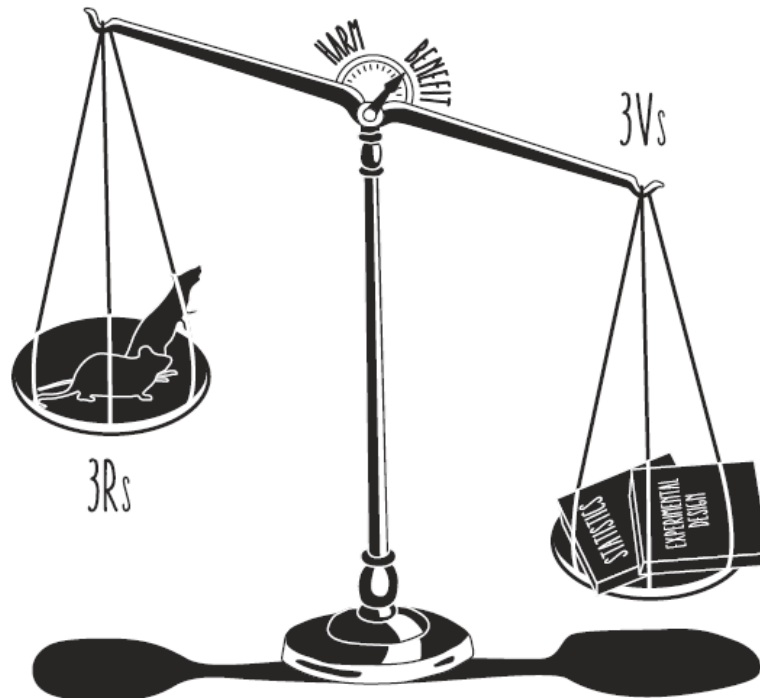
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→ **Pre-registration of experimental design and statistical analysis plan**

Potential problems

The 3Vs

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- **Poor external validity → poor representation of target population**



External validity and reproducibility

The birth of “reproducibility” as principle to distinguish facts from fiction



Robert Hooke setting up an air pump to replicate observations made by a Dutch scientist, witnessed by Royal Society members.

Rita Greer 2007 The Scientists
Source: **Wikimedia Commons**



Standardization and poor reproducibility

The standardization fallacy

«Standardization is the attempt to increase reproducibility at the expense of external validity.»

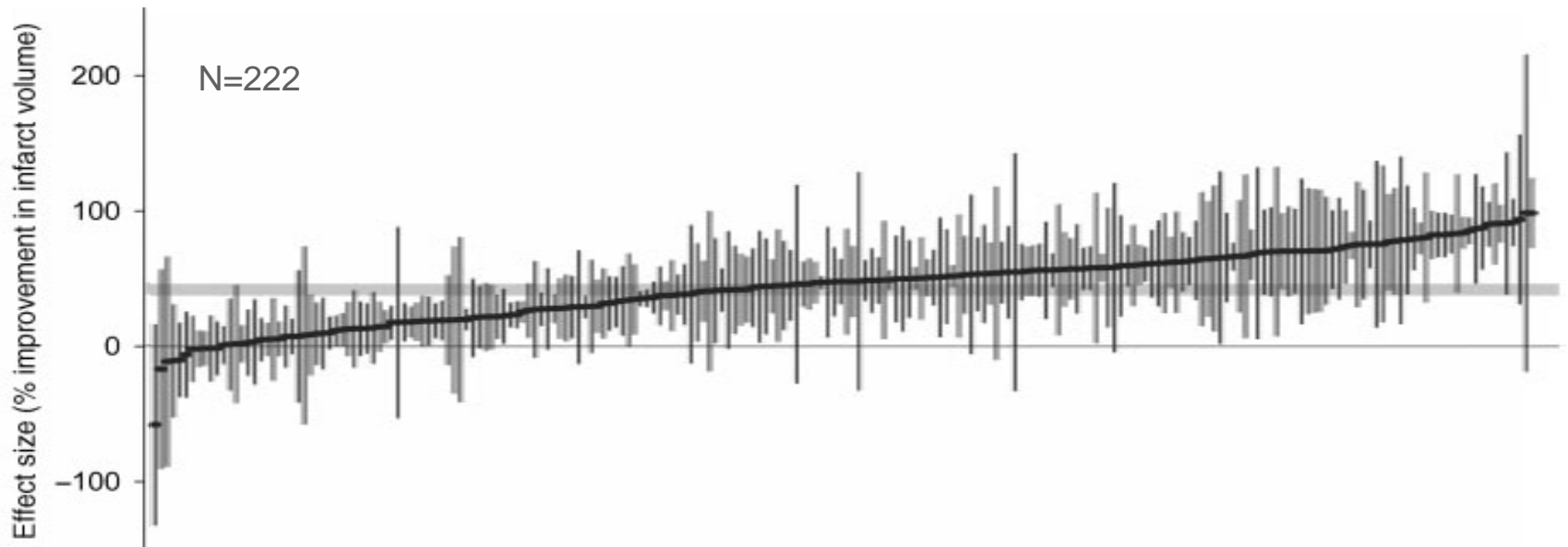
Würbel 2000 **Nat Genet**

*«A highly standardized experiment supplies direct information only in respect of the narrow range of conditions achieved by standardization. **Standardization, therefore, weakens rather than strengthens our ground for inferring a result, when, as is the case in practice, these conditions are somewhat varied.**»*

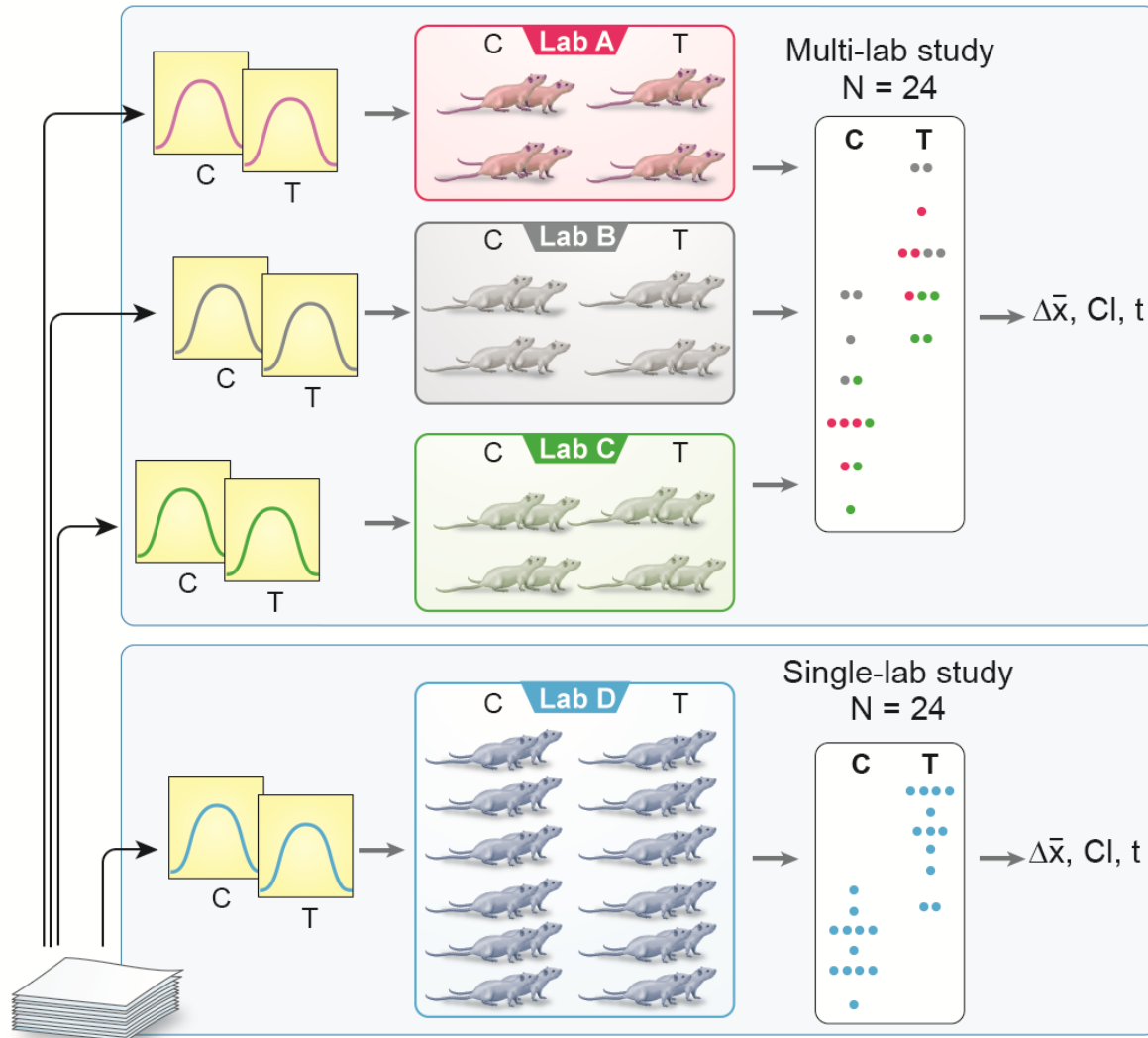
Fisher 1935 **The Design of Experiments**

Simulation of multi-lab studies using real data

Effect of hypothermia on infarct volume in animal models of stroke



Simulation of multi-lab studies using real data

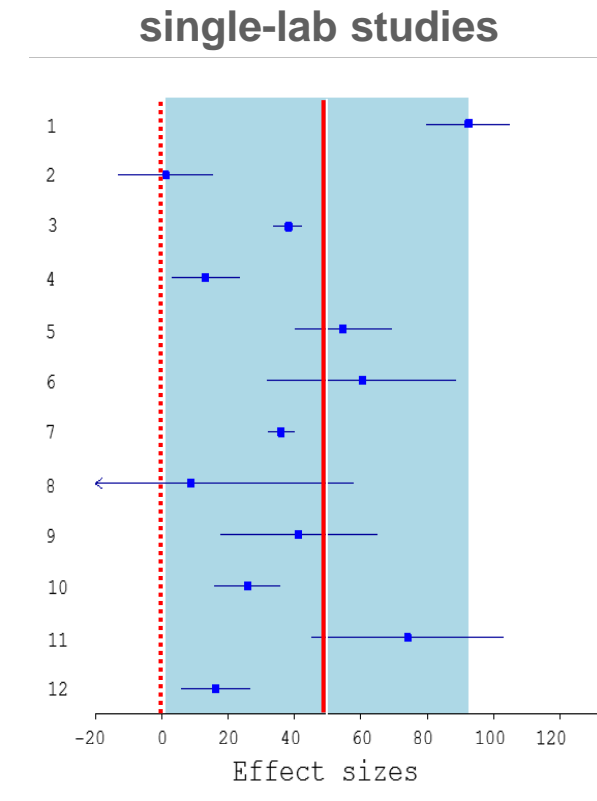


50 Studies

Voelkl et al. 2017 unpublished results

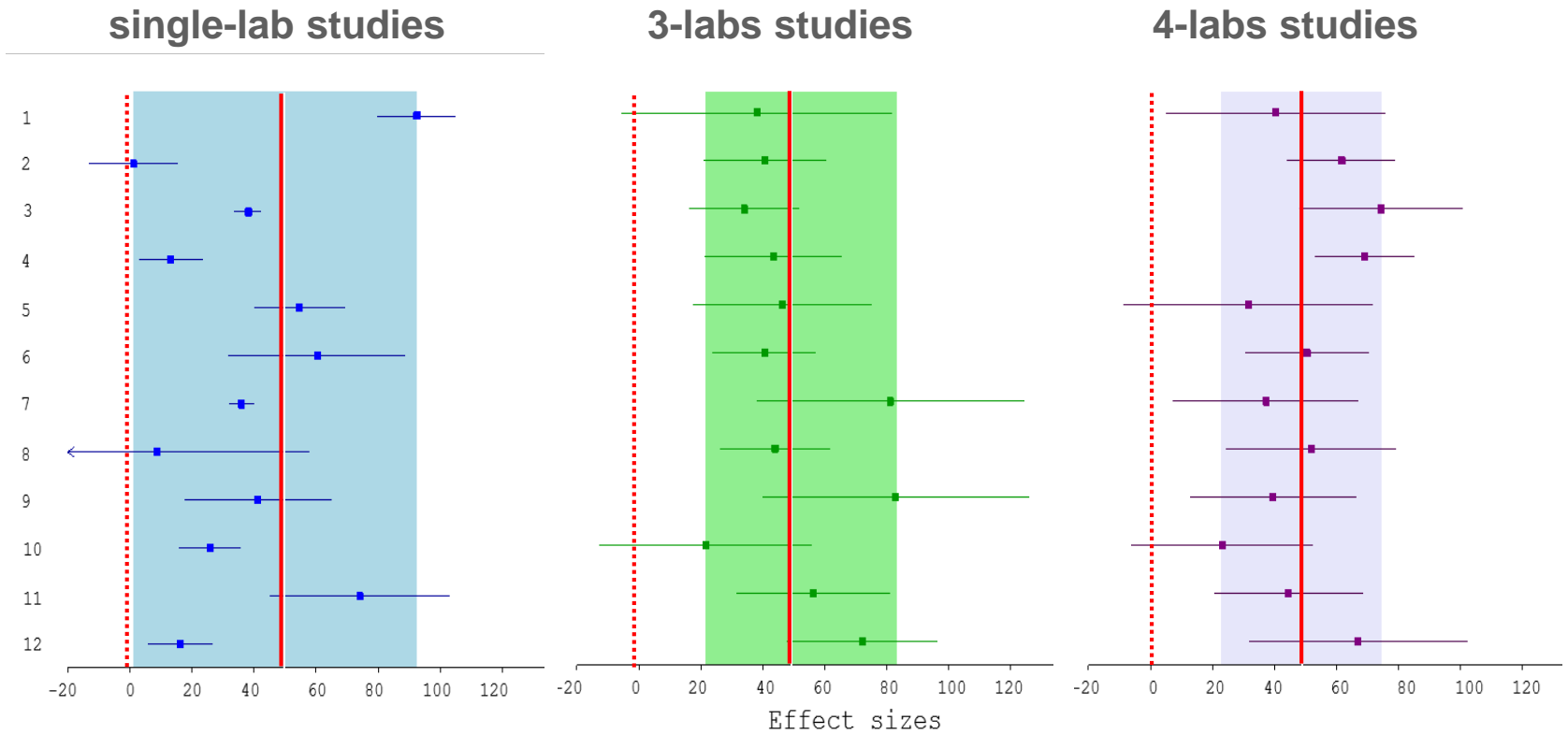
Simulation of single-lab versus multi-lab studies

Effect of hypothermia on infarct volume in rodent models of stroke



Simulation of single-lab versus multi-lab studies

Effect of hypothermia on infarct volume in rodent models of stroke



Conclusions

- Results confirm poor reporting of measures against risks of bias
- Authorities approving animal experiments and editors publishing papers lack information to assess risks of bias
- Researchers overestimate scientific rigor of their own research
- Researchers lack problem awareness and knowledge to avoid risks of bias
- Reproducibility is dependent on external validity
- Standardized single-lab studies produce results of poor external validity and poor reproducibility
- Multi-lab studies (or other forms of heterogenization of study samples) will increase external validity and reproducibility

Acknowledgments

My lab

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- Emily Sena (Edinburgh)
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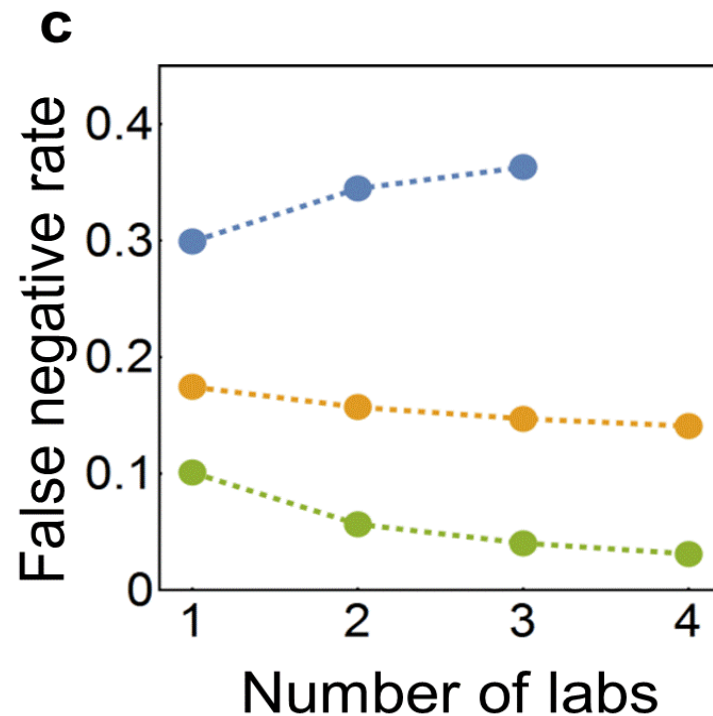
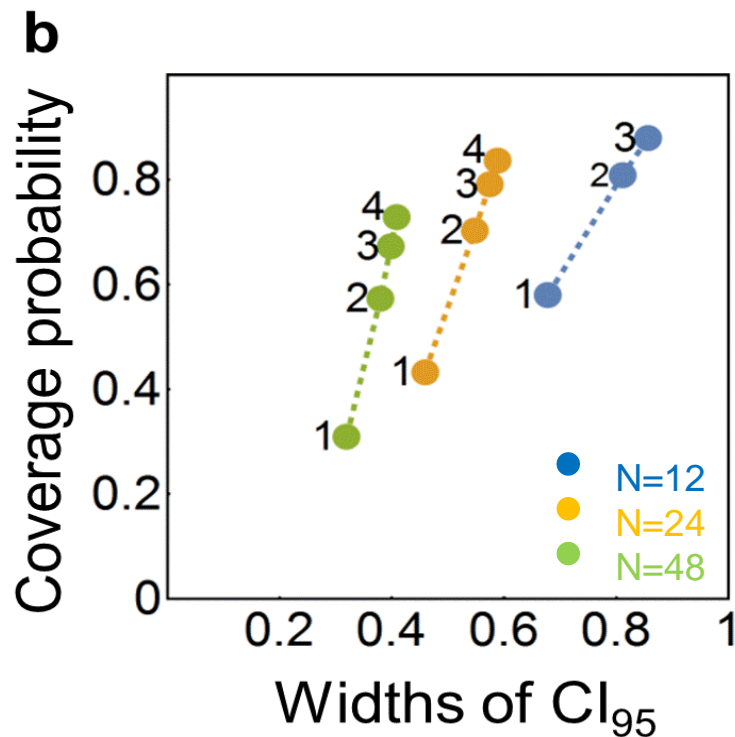
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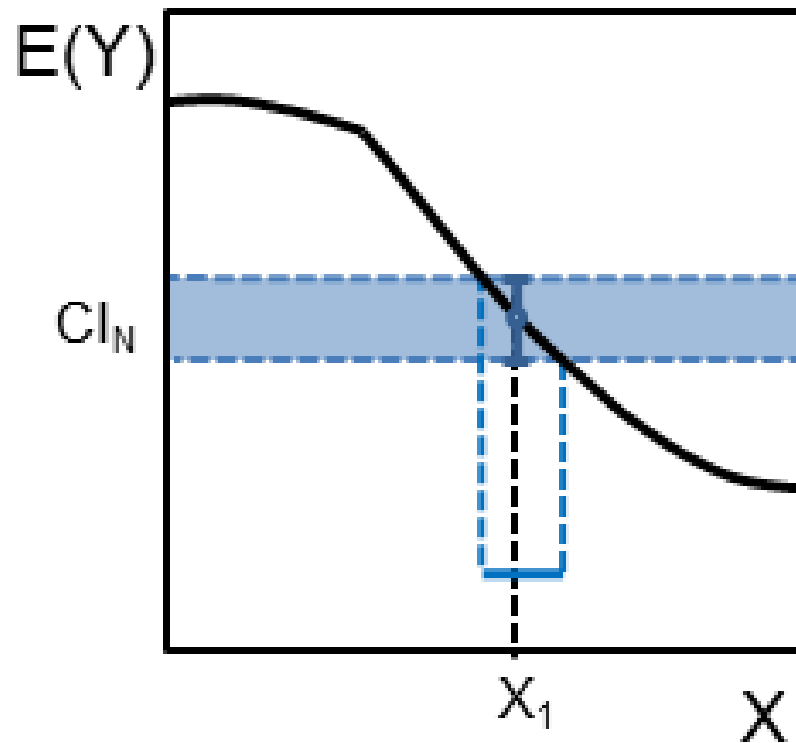
Simulation of multi-lab studies using real data

Coverage probability vs. false negative rate



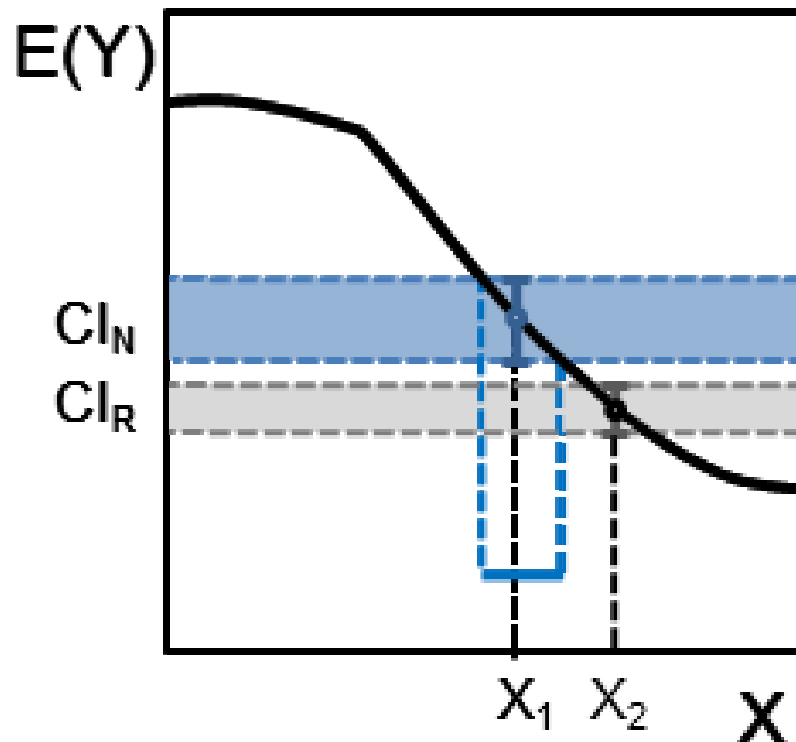
Standardization and lack of replication

The reproducibility paradox



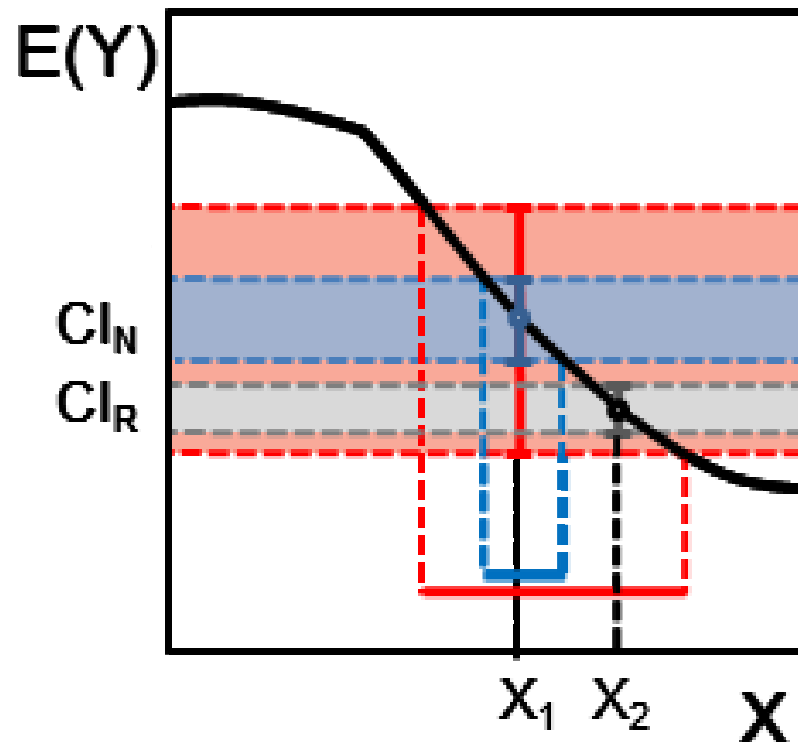
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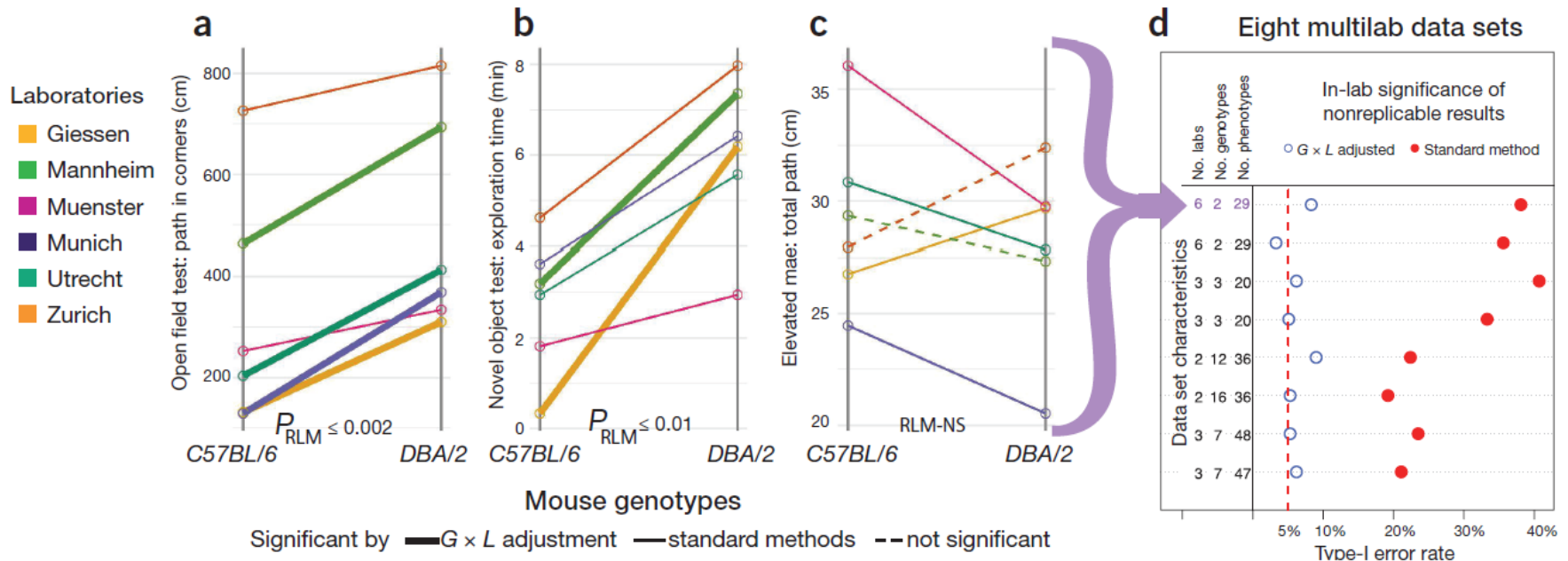
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Statistical correction for between-lab variation

Adjusting p-values based on data on genotype x laboratory interaction



Prototype web server for performing $G \times L$ adjustment:

https://stat.cs.tau.ac.il/gxl_app/