

# Academic integrity

## Where it may inadvertently be lacking

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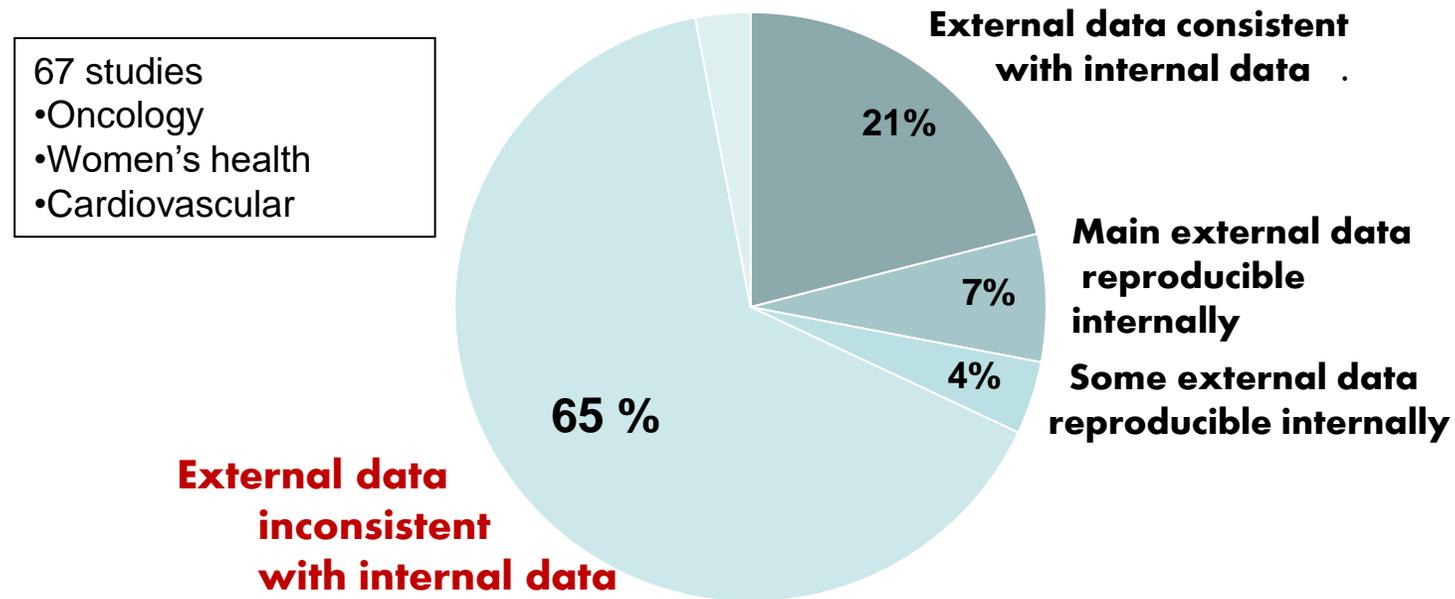
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Partnership for the Assessment and Accreditation of Scientific Practice

Work in the author lab funded by IMI as part of EQIPD consortium and by the US NIH Innovative Medicines Initiative 2 Joint Undertaking (grant agreement no. 777364); this Joint Undertaking receives support from the European Union's Horizon 2020 research and innovation program and EFPIA

# Can you trust published data?



Poor replicability meanwhile found in several other studies in experimental medicine and many other disciplines including social sciences

# What are the root causes?

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- Outright fraud minor component (intentionally falsified data)
- Lack of detail in Methods section makes replication difficult

But the big-3 apparently are:

- Biases in design, conduct, analysis and reporting of studies
- Low statistical power (sample sizes too small)
- Poor understanding of statistical concepts

Let's focus on 2 examples where "good intentions" can get uncomfortably close to fraud

# Randomness principle

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- A p-value reports the probability of seeing an effect as large as observed, or larger, if the two samples had been selected randomly from populations with the same mean/median
  - Does not tell whether an observed effect is true or of relevant magnitude
- Only meaningful if all factors other than primary variable are randomly distributed among groups
  - P-values cannot be interpreted at face value if major bias exists (violation of randomness principle)

# Violation of randomness

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- Violations (biases) can occur unconsciously or be investigator-induced
- Unconscious biases
  - Sampling error
  - Selection bias
  - Other biases
- Investigator-induced violations are also referred to as p-hacking
  - Includes HARKing

# P-hacking

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- Various design choices may be fine if pre-specified
- Post-hoc changes in design, conduct, analysis and reporting introduce bias and violate randomness principle
- This makes resulting p-values difficult to interpret, irrelevant or even misleading
  - Bias for finding an effect even if it is not there
  - Even if effect true, trend for exaggerated effect size estimate

# P-hacking examples

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- After  $n = 6$  yielded  $p = 0.055$ , add 2 additional experiments
  - The new  $n = 8$  is biased by the trend in  $n = 6$  and no longer a random sample
  - Variation: Stop adding experiments as soon as  $p < 0.05$
- Post-hoc decision to log-normalize data
  - Log normalization can be justified or even required when raw data are skewed and only get closer to a normal distribution on a log scale
- Post-hoc change of denominator
  - From fmol/mg protein to fmol/g wet weight
- Switch to a different statistical test
  - paired vs. unpaired test
- “Outlier” removal (attrition bias)

# HARKing

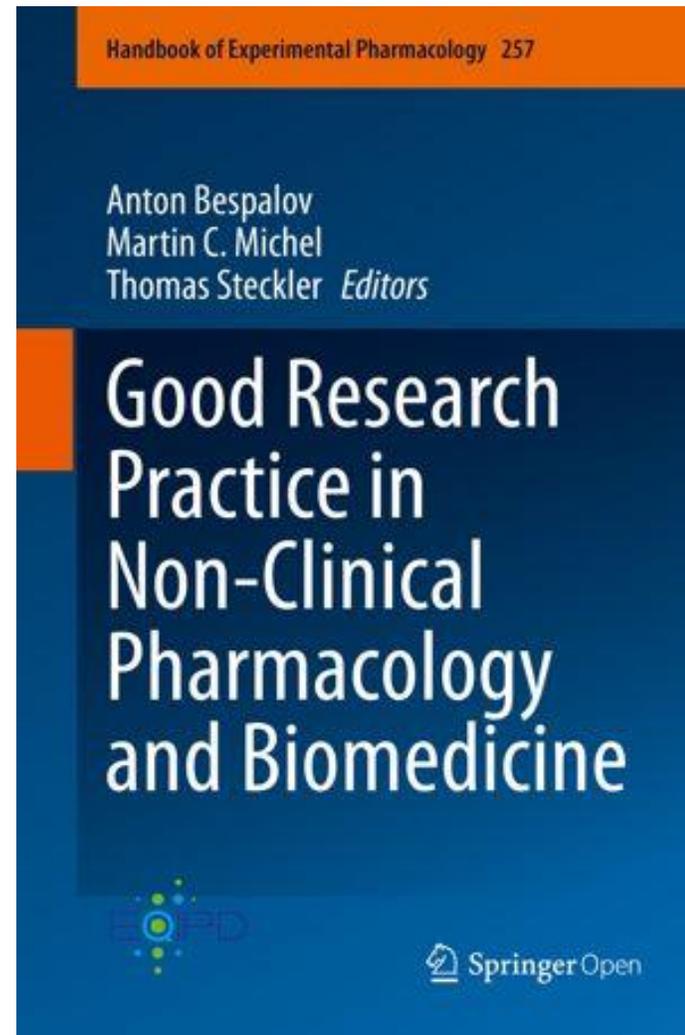
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## Hypothesizing **A**fter **R**esults are **K**nown

- Redefining study aim after results have been seen
- Introduces bias into reporting
  - Focus on “positive” results in reporting
  - Making possible chance finding appear as study aim
- Ignores impact of low prior probability on False Discovery Rate
  - Increases probability for false positives
- HARKing becomes fraud if paper claims that a HARKed hypothesis had been pre-existing

# Open access book on reproducibility

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# Conclusions

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- Biases at any level of study design, conduct, analysis and reporting violate the randomness principle
  - This makes p-values difficult to interpret, possibly misleading
- Pre-specification of all critical elements of study design, conduct and analysis, and randomization and blinding are key defenses against unintentional bias
- P-Hacking and HARKing are intentionally introduced biases
  - Often lead to misleading finding/conclusions or at least exaggerated effect sizes
  - Get uncomfortably close to fraud

**DON'T**



**JUST DO IT.**

