Checking Robustness in 4 Steps

Dr. Michèle B. Nuijten



Sounds like Newton/Nowton











My background.



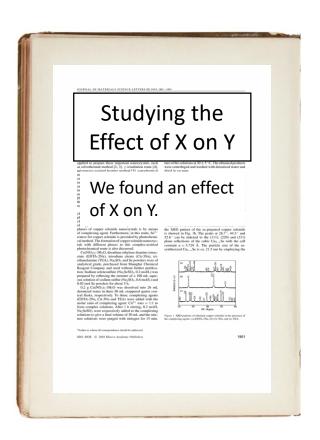




Today.

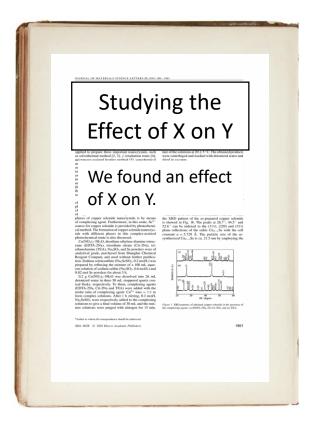
Assessing and improving robustness of psychological science in 4 steps (while using minimal resources).

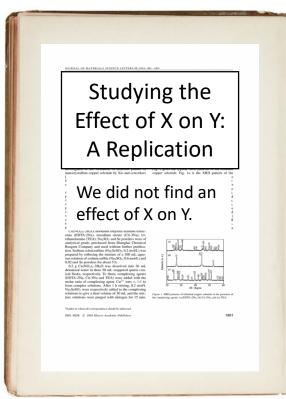
Robustness.



Robustness ≈ "Can I trust this result?"

Assessing robustness through replication?





Cons:





Focus on reproducibility first.

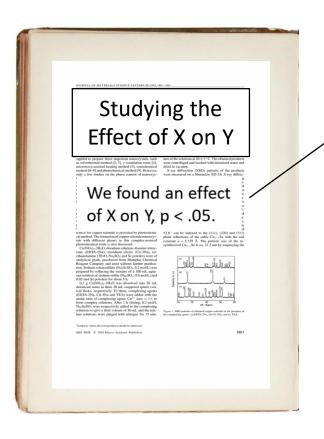
Replicability

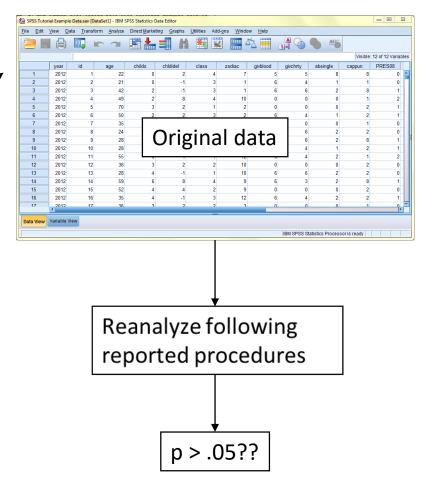
A study is successfully **replicated** if the same/a similar result is found in a new sample.

Reproducibility

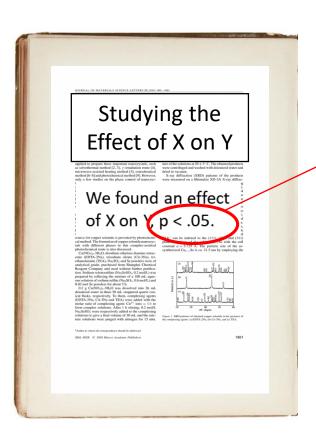
A study is successfully **reproduced** if independent reanalysis of the original data, using the same analytic approach, leads to the same results.

Reproducibility is a prerequisite for replicability.





Reproducibility is a prerequisite for replication.



- If a result is not reproducible, it has no clear bearing on theory or practice
- An irreproducible number is effectively meaningless

You don't need replication to find out whether this finding is robust. It's not.

Today.

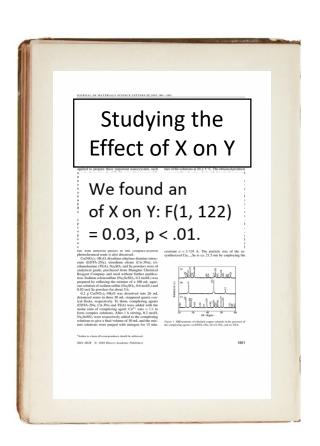
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The 4-Step Robustness Check

- Check the internal consistency of the statistical results
- Reanalyze the data using the original analytical strategy
- 3. Check if the result is robust to **alternative analytical choices**
- 4. Perform a **replication** study in a new sample

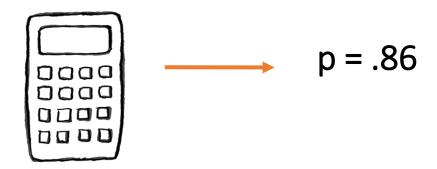
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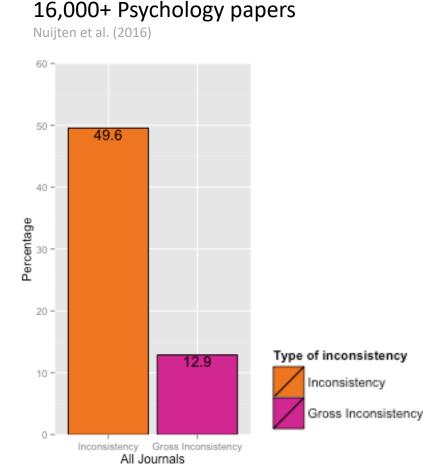


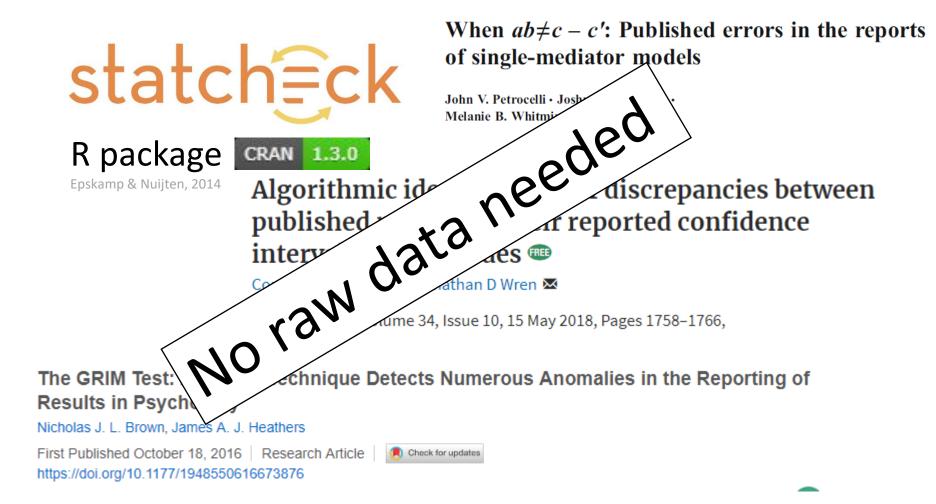
= Statistical sanity check

Also as expected, when priming condition was crossed with age group and time of memory prediction, interaction effects emerged for both the photo recall predictions, F(1, 122) = 0.03, p < .01 and the learned recall predictions, F(1, 135) = 3.75, p < .06.

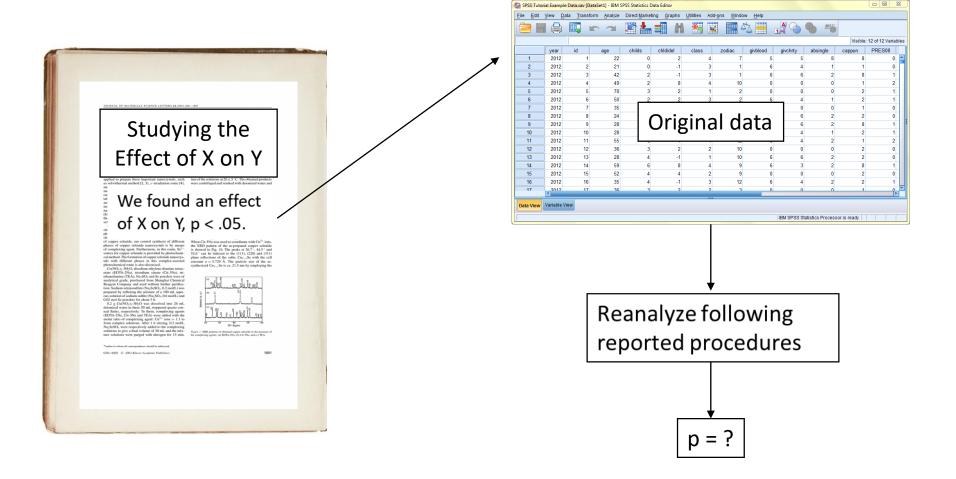




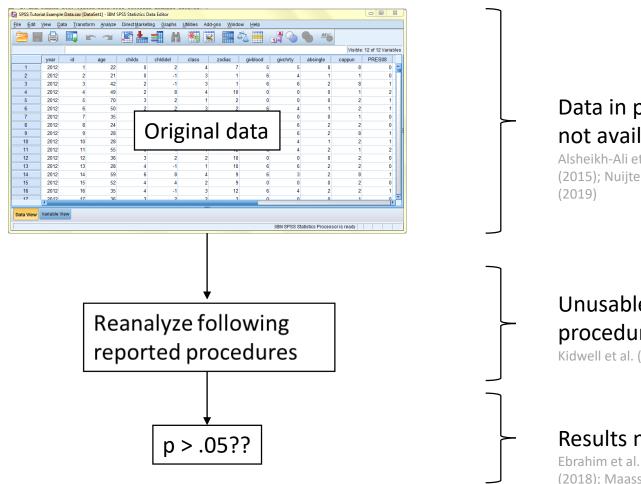




2. Reanalyze the data using the original analytical strategy.



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Data in psychology often not available

Alsheikh-Ali et al. (2011); VanPaemel et al. (2015); Nuijten et al. (2017); Hardwicke et al. (2019)

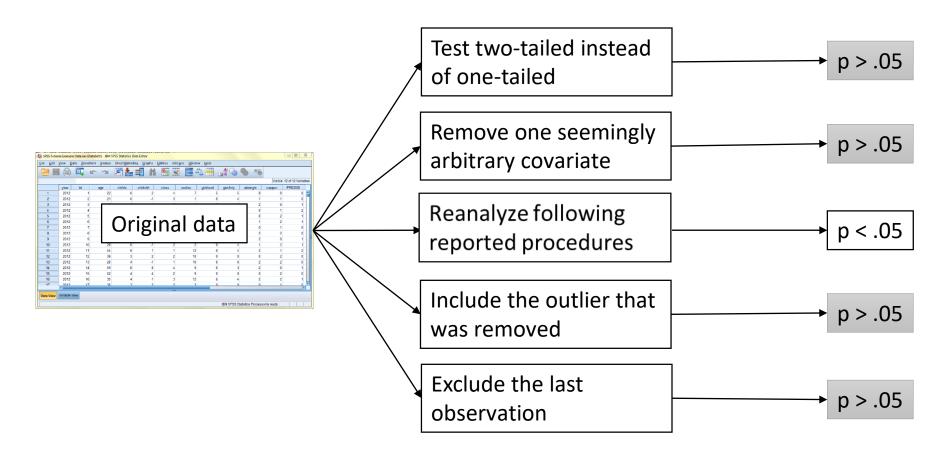
Unusable data or analytical procedure unclear

Kidwell et al. (2016); Hardwicke et al. (2019)

Results not reproducible

Ebrahim et al. (2014); Hardwicke et al. (2018); Maassen et al. (forthcoming)

3. Check if the result is robust to alternative analytical choices.



3. Check if the result is robust to alternative analytical choices.

False-Positive Psychology: Undisclosed Flexibility in Data Collection and Joseph P. Simmons, Leif D. Nelson, Un Simons, Leif D. Nelson, Leif D. Nelson, Un Simons, Leif D. Nelson, Un Simons, Leif D. Nelson, Un Simons, Leif D. Nelson, Leif D. Nelson, Un Simons, Leif D. Nelson, Leif D Analysis Allows Presenting Anything as Significant Measuring the Prevalence of Questionable © The Author(s) 2012 Reprints and permission: The garden of forking paths: Why multiple comparisons can be a problem, or "number of the research when there is no "fishing expedition" or "number of the research of the res Research Practices With Incentives for The garden of forking paths: Why multiple comparisons can be a problem, and the research of "p-hacking" and the research of time when there is no hypothesis was posited ahead of time. **Truth Telling** Leslie K. John , George Loewenstein, and Drazer Marketing Unit, Harvard Business School; Department of Social & D and ³Sloan School of Management and Departments of Ec Institute of Technology

19

4. Perform a replication study in a new sample.

- 1. Check the **internal consistency** of the statistical results
- **/**
- 2. Reanalyze the data using the original analytical strategy



- 3. Check if the result is robust to alternative analytical choices
- 4. Perform a replication study in a new sample

Failed replication more likely to have bearing on the effect

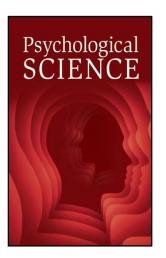
Today.

Assessing and **improving** robustness of psychological science in 4 steps (while using minimal resources).

1. Check the **internal consistency** of your own statistical results

 Use statch=ck and related tools for self-checks / in the peer review process







http://statcheck.io

2. Facilitate **reanalyis** of the data

EFFORT

- Share data
- Share well-documented data
- Share analysis scripts
- "In-house" code review (co-authors = co-pilots)
- Code review during peer review
- Fully reproducible dynamic manuscripts (R Markdown, Code Ocean, Docker, etc.)

3. Report whether your result is robust to **alternative analytical choices**

21-word solution

Simmons et al. (2011)

These 21 words in a Methods section can say it succinctly:

"We report how we determined our sample size, all data exclusions (if any), all manipulations, and all measures in the study."

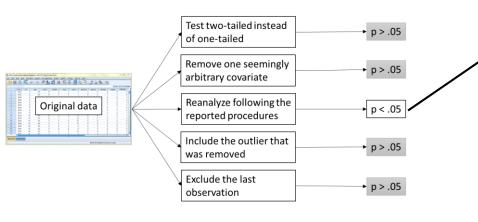
3. Check and report whether your result is robust to

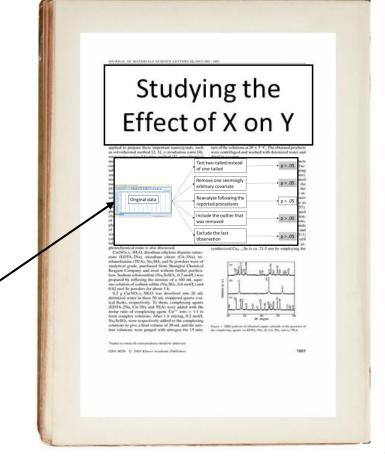
alternative analytical choices

Journals could require sensitivity analyses

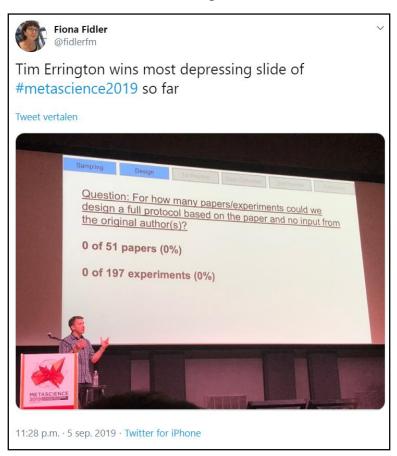
Multiverse analysis

Steegen et al. (2016)





4. Facilitate **replication** in a new sample



Write detailed methods sections/appendices and share materials & protocols!

Discussion.

Assessing and improving robustness of psychological science in 4 steps (while using minimal resources).

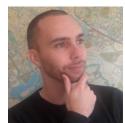
- If you're interested in the robustness of a specific study
- Context matters: an inconsistency in the 3rd decimal doesn't automatically mean you shouldn't replicate
- Regardless of the logic of the 4-step robustness check:

All published research should always be reproducible!

Meta-Science Symposium

November 22 2019, Tilburg University, The Netherlands

- Keynotes of
 - John Ioannidis
 - Ana Marusic
 - Sarah de Rijcke
- Parallel sessions focused on meta-scientific questions
- Questions? <u>metaresearch@tilburguniversity.edu</u>
- Or go to Olmo van den Akker or Jelte Wicherts





• Sneak preview: July 2020, Tilburg University, 2-3 day metascience conference. Details will follow!

Thank you!

A 4-step robustness check to **assess** and **improve** psychological science.

- Check the internal consistency of the statistical results
- Reanalyze the data using the original analytical strategy
- 3. Check if the result is robust to alternative analytical choices
- Perform a replication study in a new sample



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