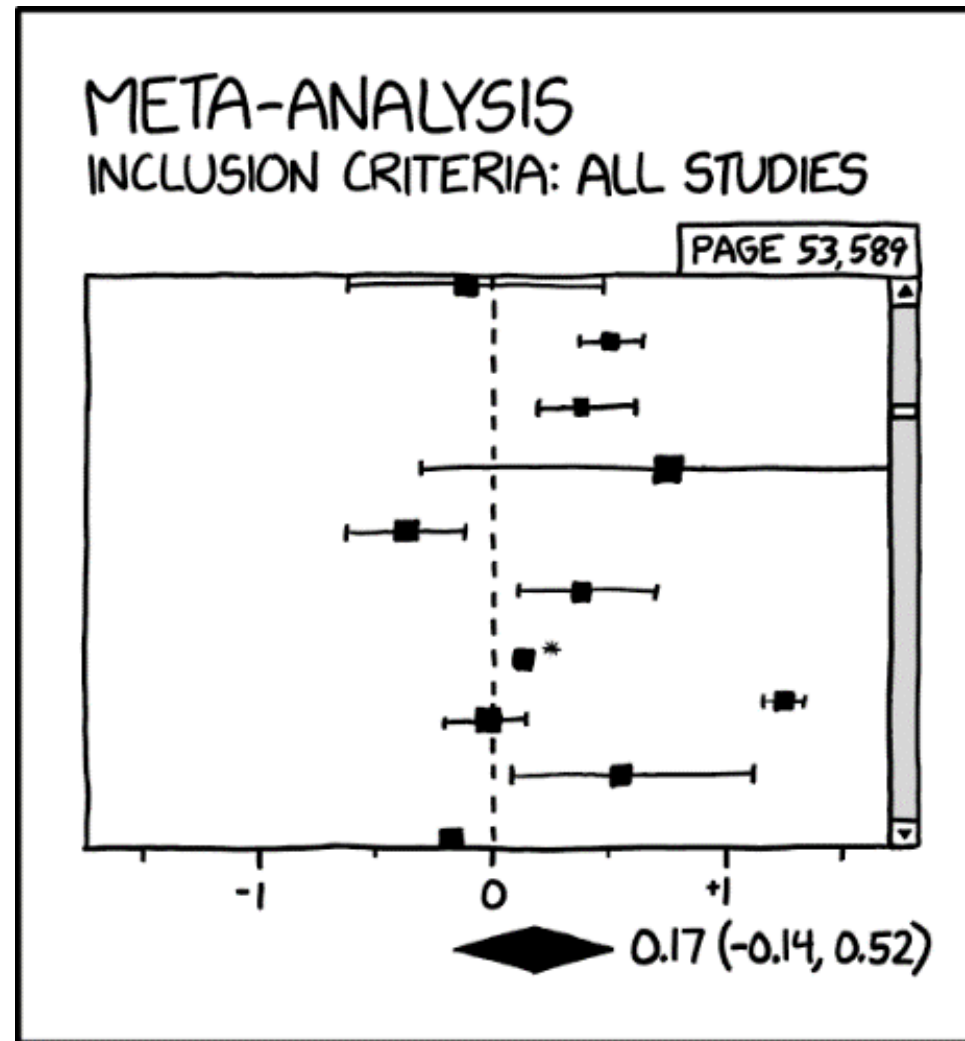




BAD NEWS: THEY FINALLY DID A META-ANALYSIS OF ALL OF SCIENCE, AND IT TURNS OUT IT'S NOT SIGNIFICANT.

<https://xkcd.com/2755/>

I have the (Developer) Power!

Supporting Power Analysis and Statistical Reporting in Usable Security and Privacy

Anna-Marie Ortloff¹, Christian Tiefenau², Matthew Smith^{1,2}

¹: University of Bonn, ²: Fraunhofer FKIE



Behavioural Security Group



A Qualitative Study on How Usable Security and HCI Researchers Judge the Size and Importance of Odds Ratio and Cohen's d Effect Sizes

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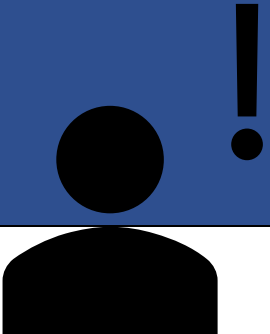
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Bonn, Germany
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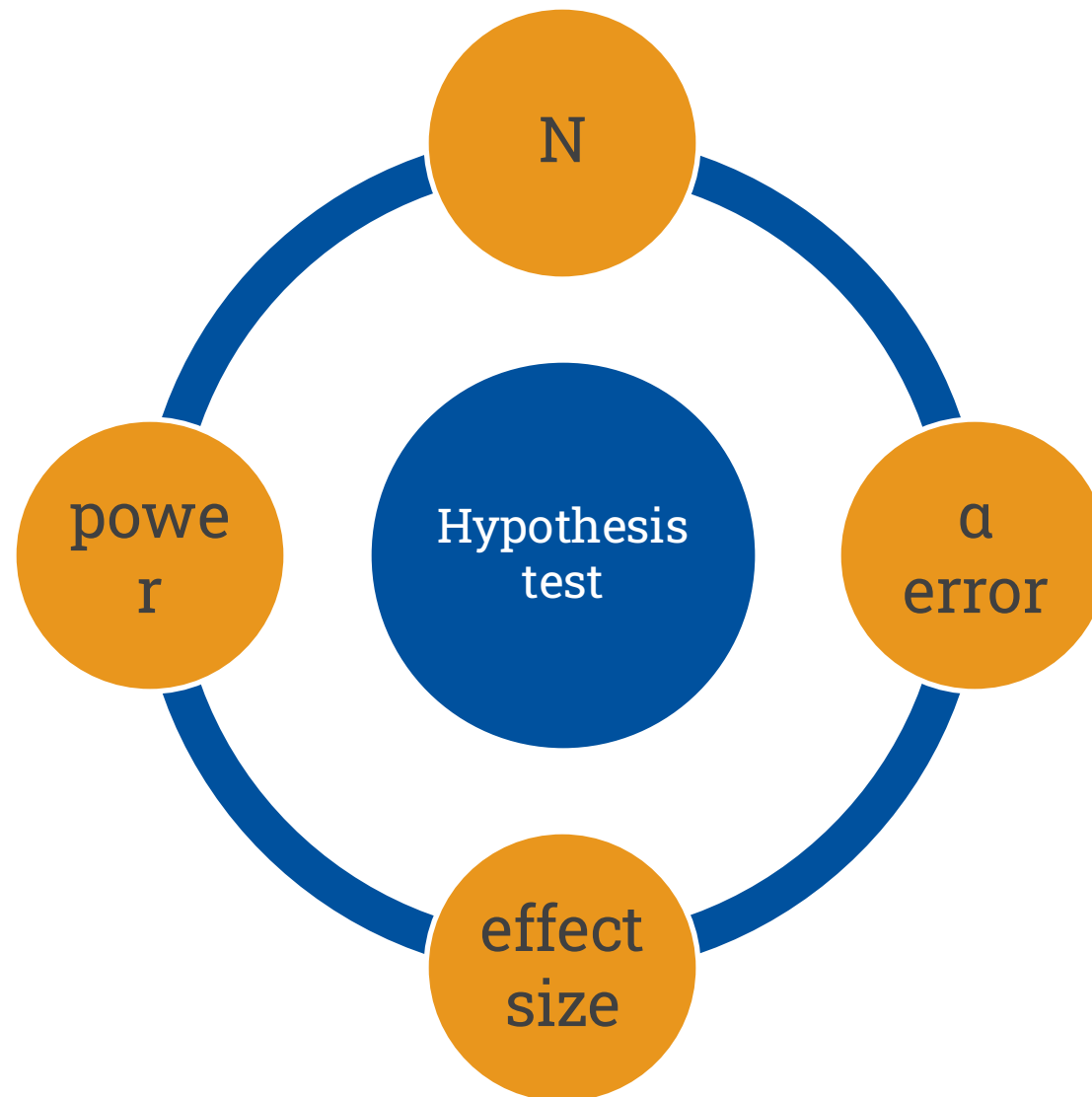
CHI'23

CHI'25

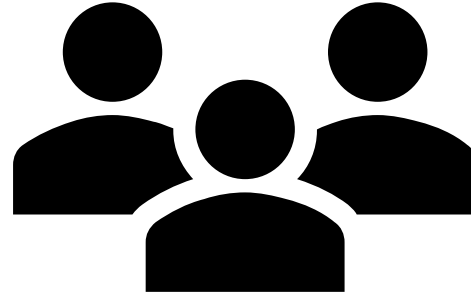
[Find recommendations here]



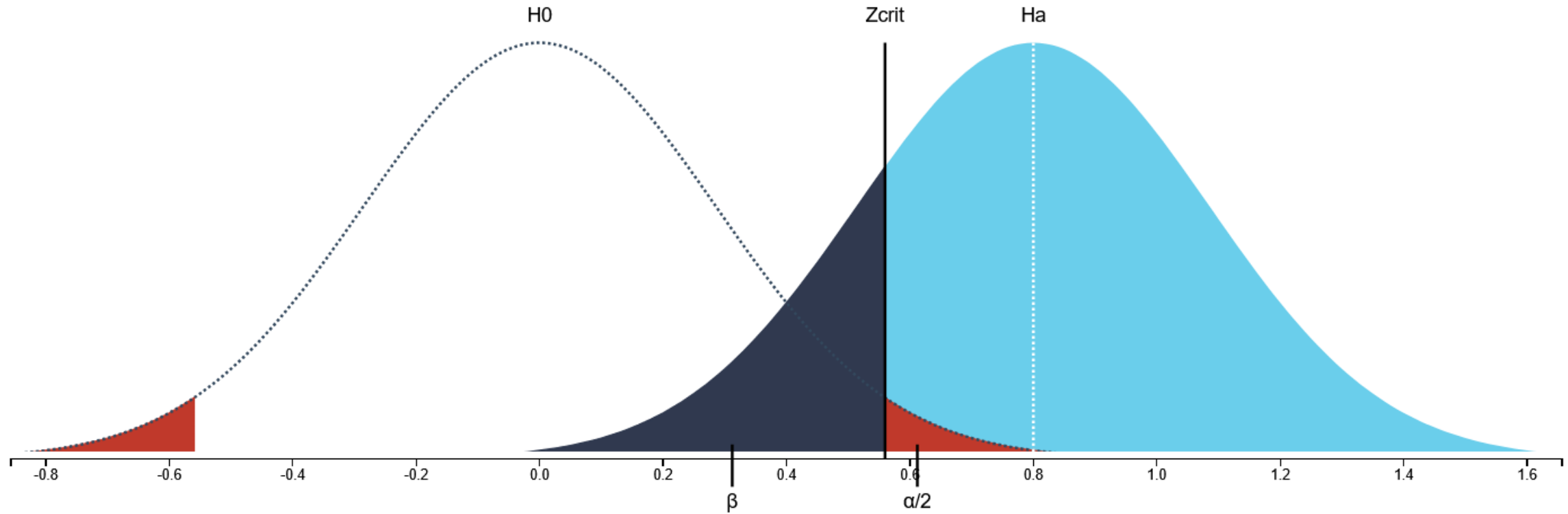
Background



Sample Size N

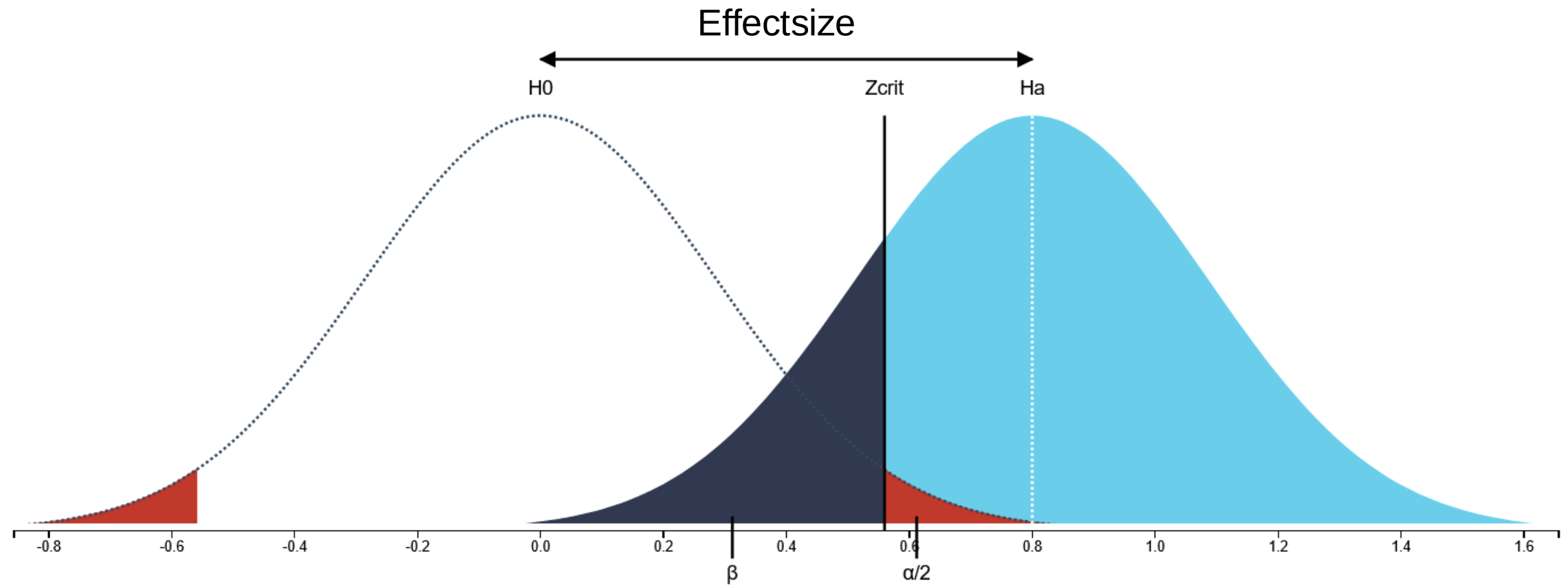


The probability of detecting an effect, if a true effect exists.



<https://rpsychologist.com/d3/nhst/>

The strength of the relationship
of predictor variables with outcome variables.



<https://rpsychologist.com/d3/nhst/>

Odds ratio

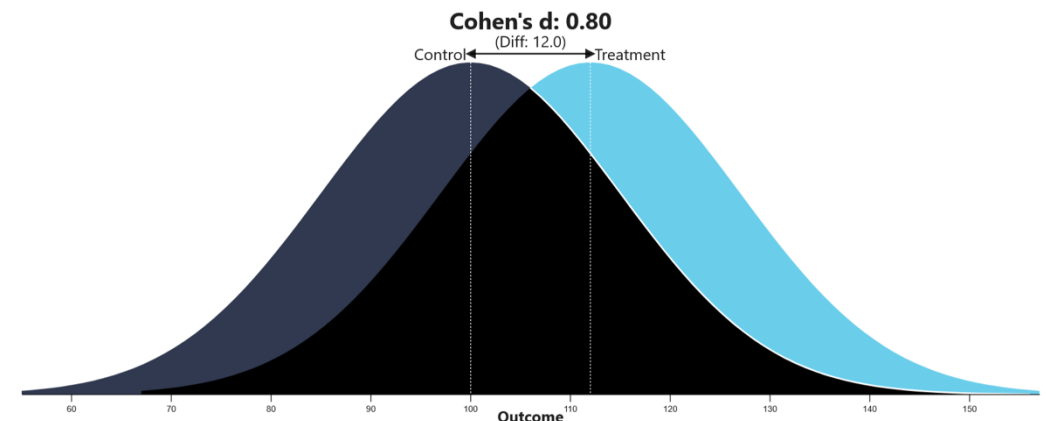
- Ratio of two odds

	Adopted PWM	Did not adopt PWM
Group 1	10	30
Group 2	20	20

- OR=1 means the odds of an outcome are the same in both groups.

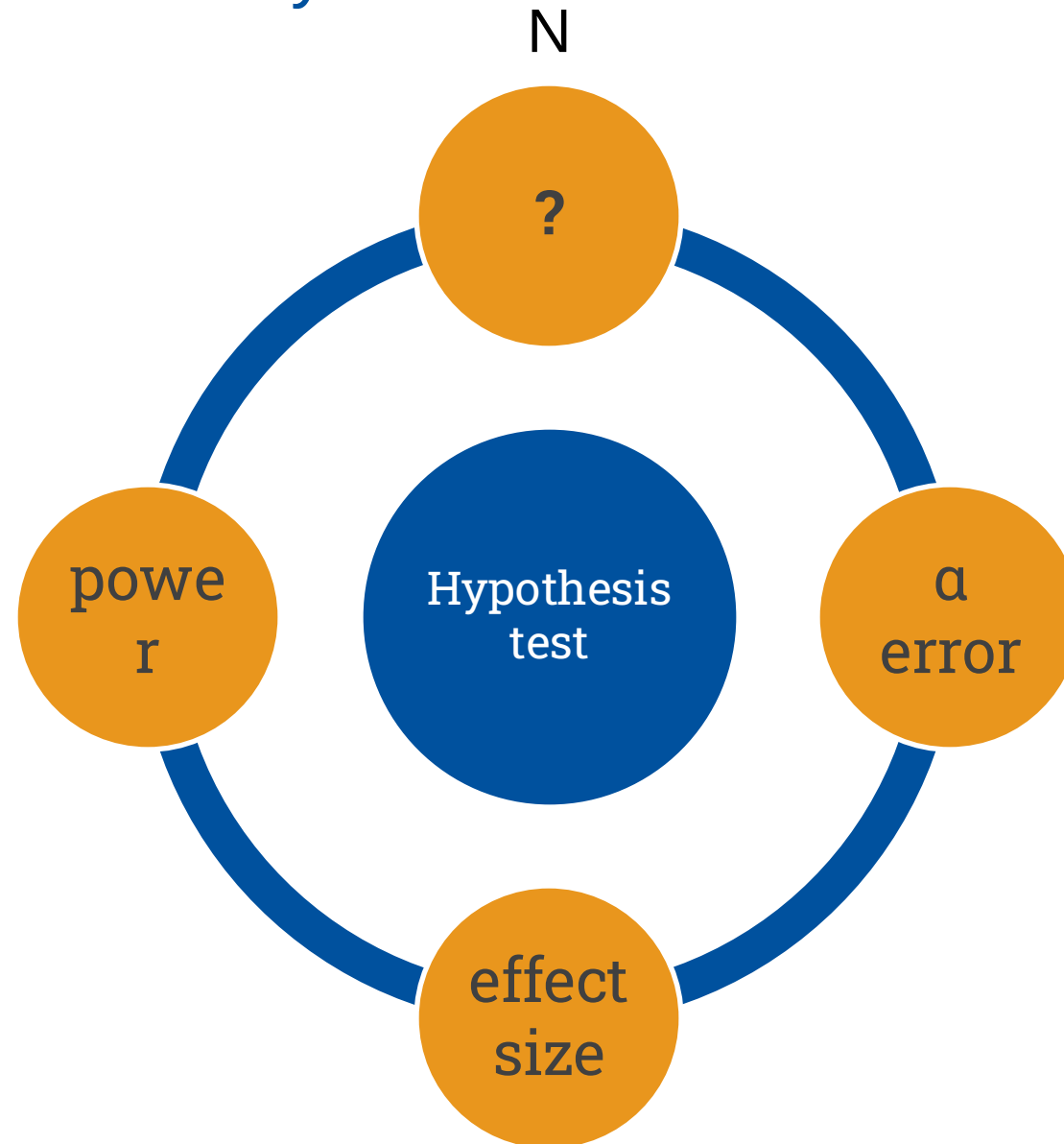
Cohen's d

- Normalized difference in means
- $d = 0.8$ is difference of 0.8 standard deviations between the means

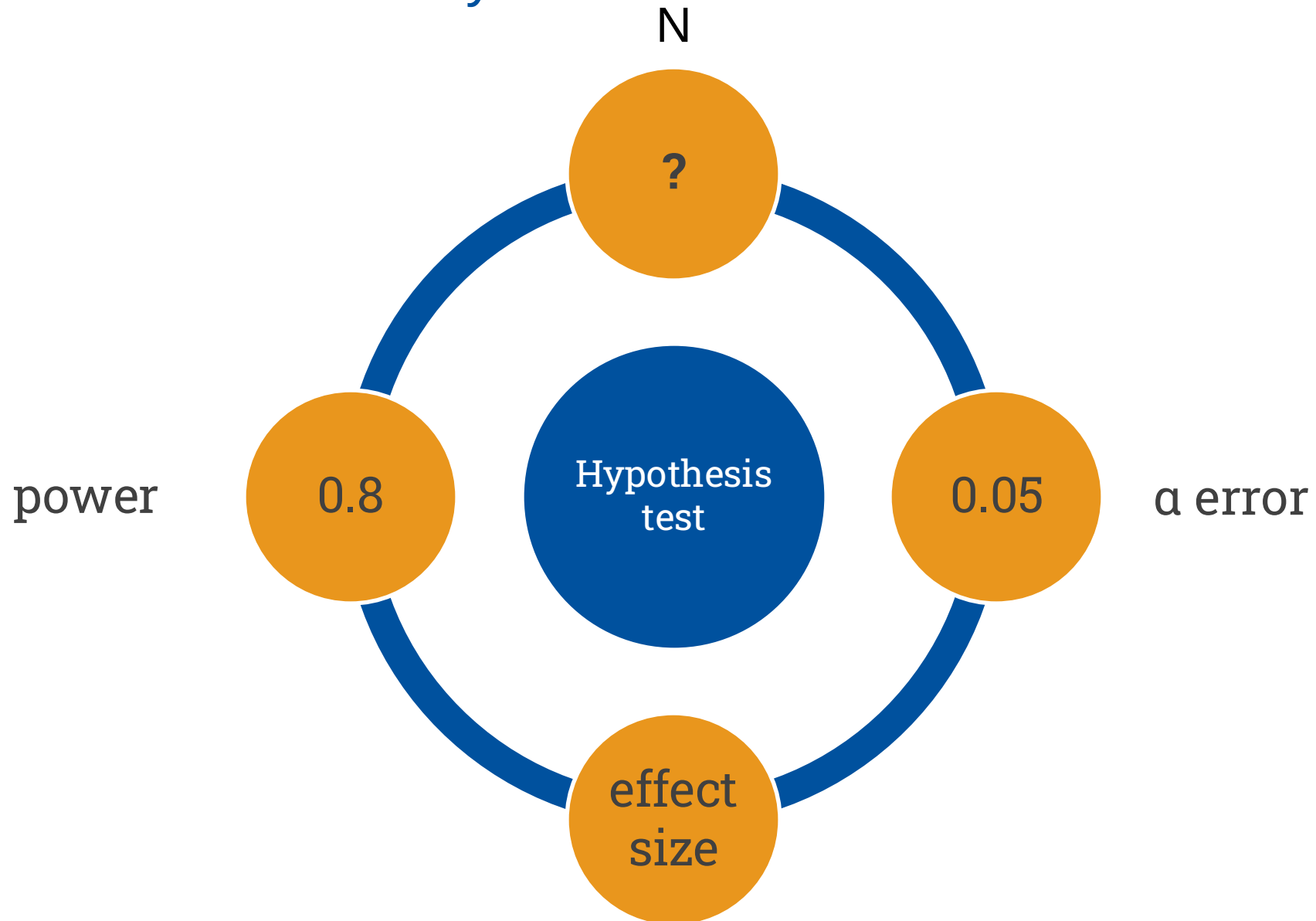


<https://rpsychologist.com/cohend/>

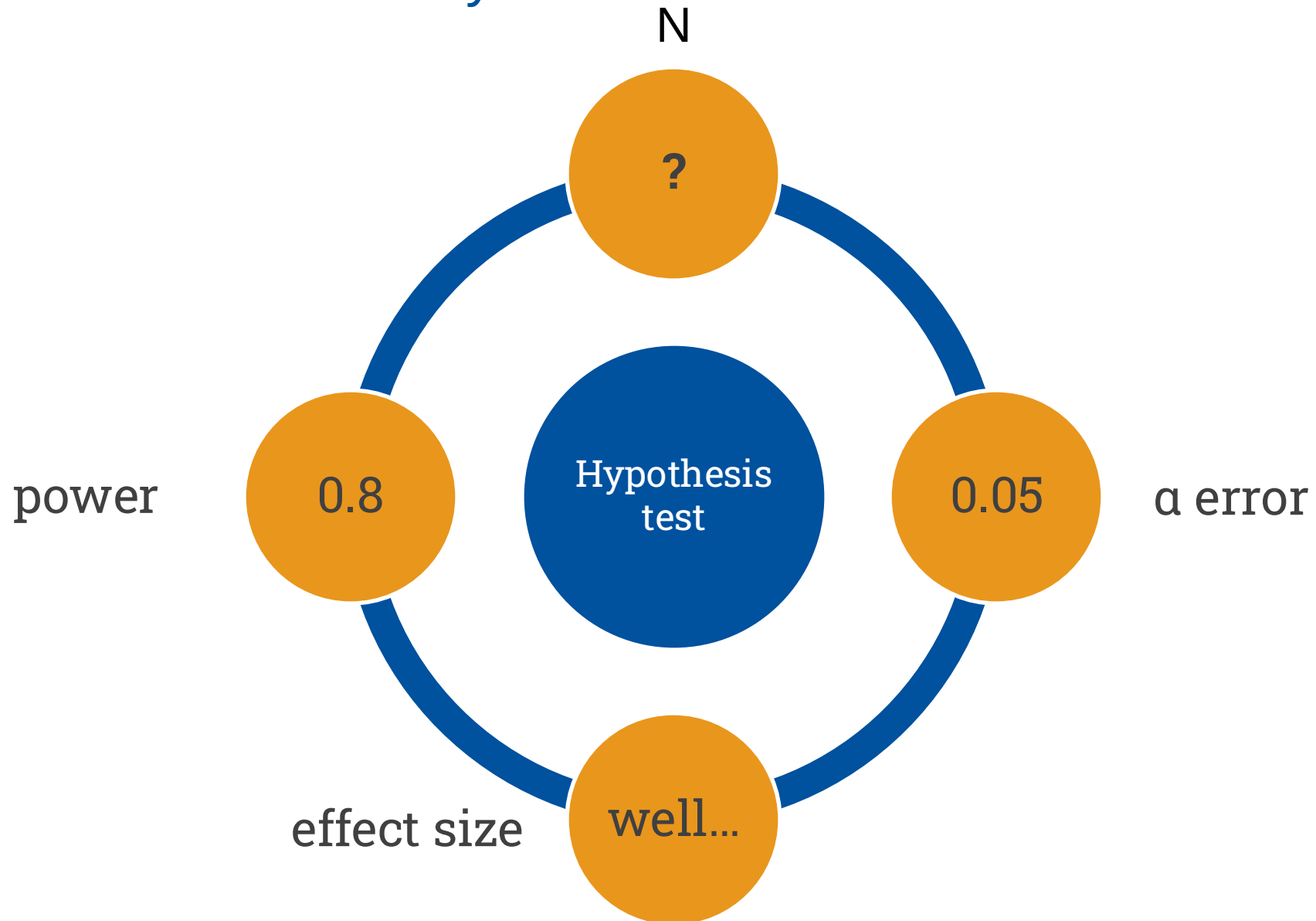
A Priori Power Analysis



A Priori Power Analysis

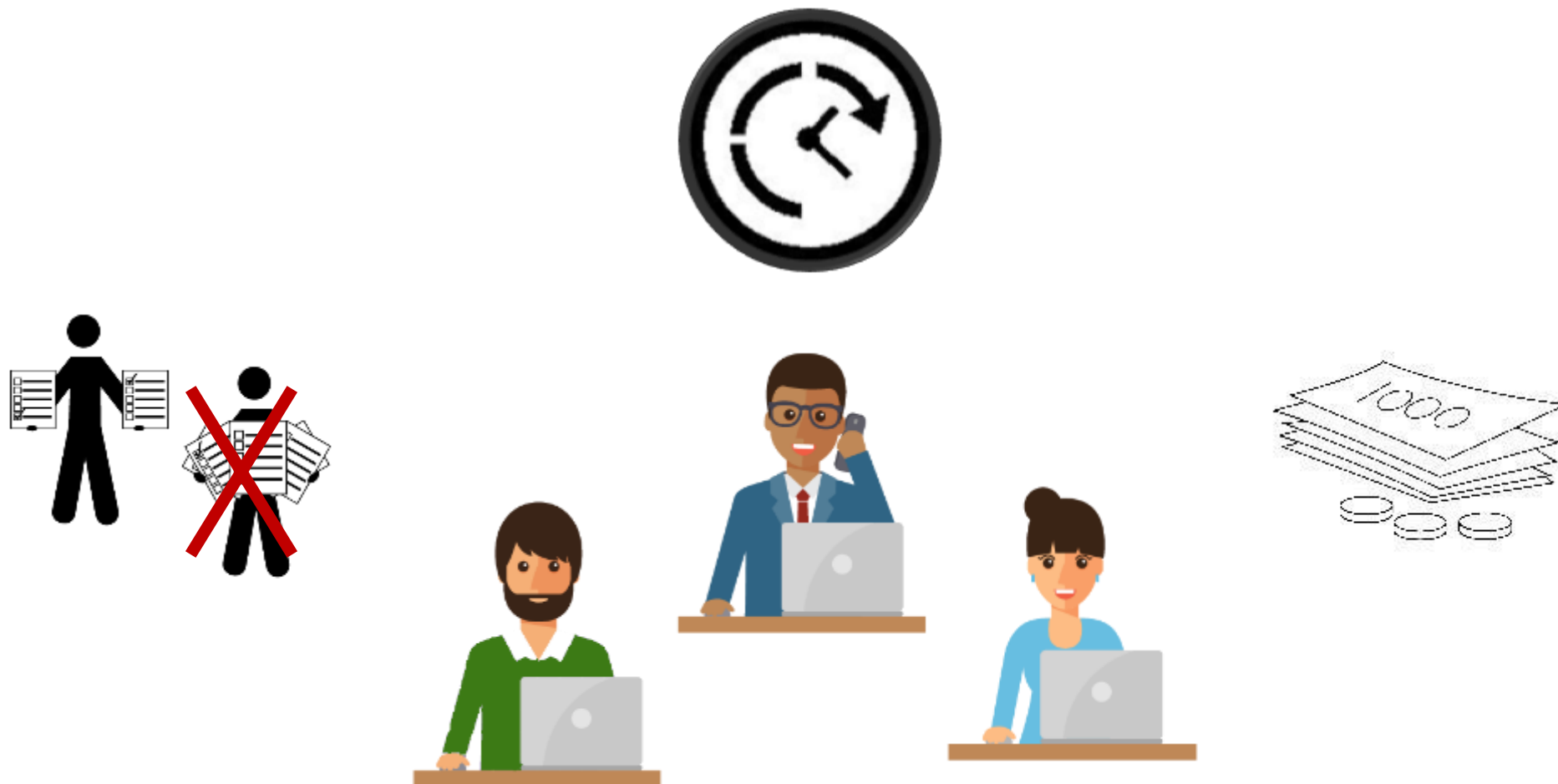


A Priori Power Analysis



Do we have the power?

Developer-Centered Usable Security (DCUS)



Literature Collection

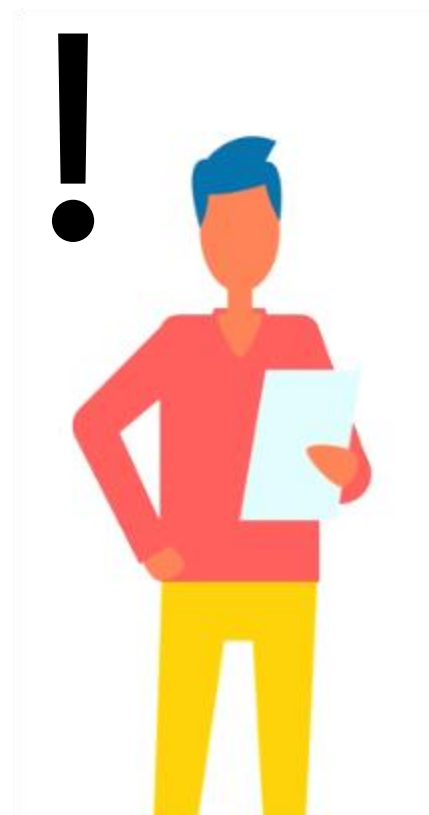
- SOUPS, USENIX Security, S&P, CCS, ICSE, USP Tracks of CHI
- 2010 - 2021
- Include user study
- Participants: software developers, similar expert users, or proxies
- Domain of usable security and privacy



- 54 papers
- including 64 studies, 467 hypothesis tests, 413 variables

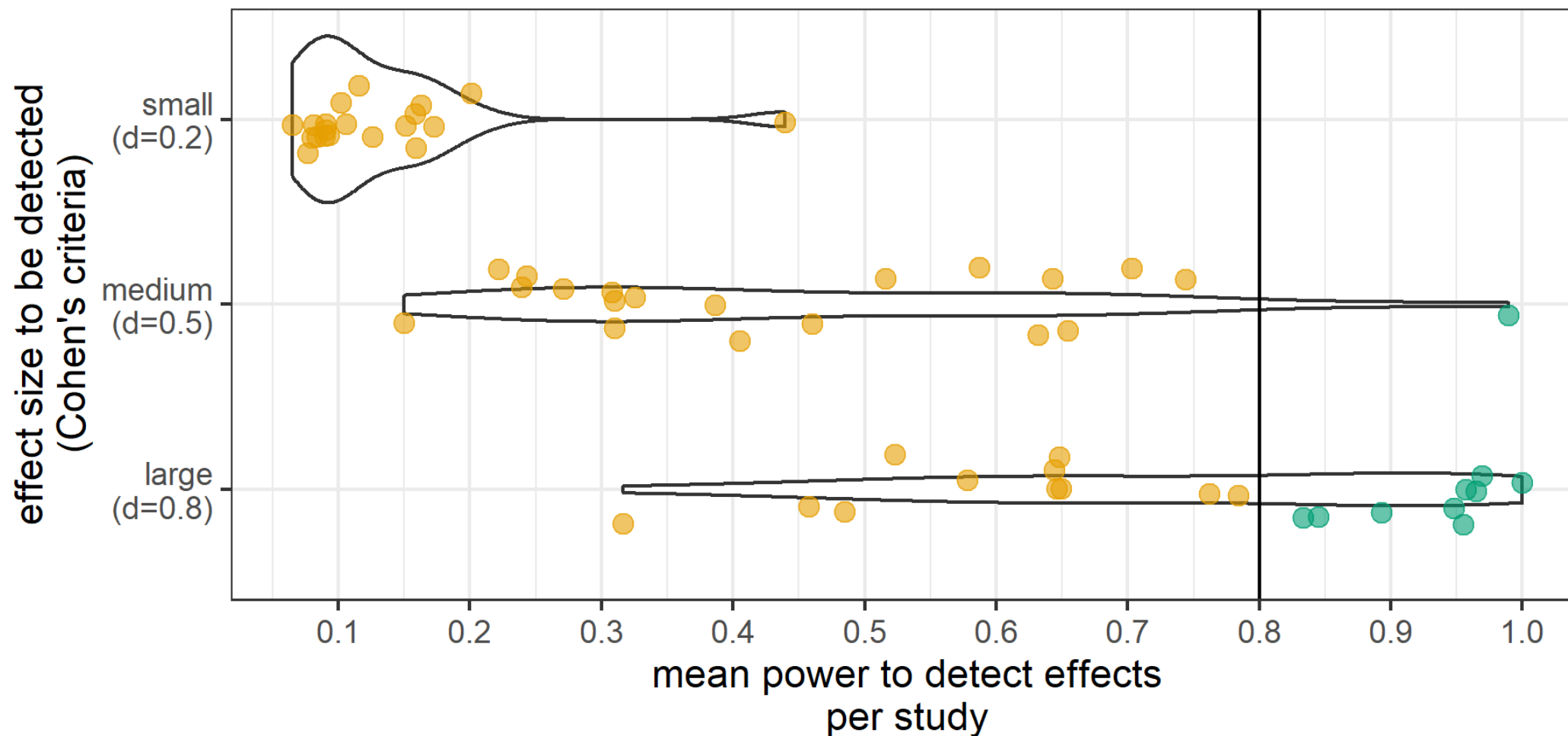
Data Structure

- Relevant information on power and effect sizes in these studies



Power Meta-Analysis (simulated a-priori power analysis)

Is power sufficient? ● no ● yes

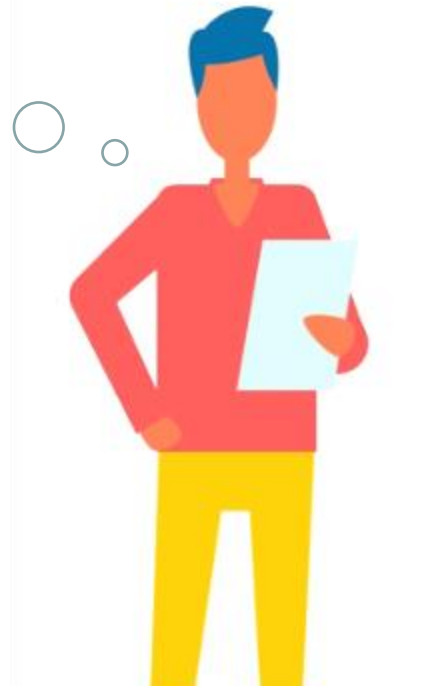


In SOUPS and CHI USP publications from 2020/2021
only 8 of 74 (10.8%) quantitative papers used a priori
power analysis

How can we get the power?

Which effect size
should I use?

?!



- Use general guidelines for large, medium, small effects
- Use context specific guidelines for large, medium, small effects
- Literature research
- Do a pilot study
- Decide on the smallest effect size of interest

I have the power!

This is the companion website for the paper *SoK: I Have the (Developer) Power! Sample Size Estimation for Fisher's Exact, Chi-Square, Wilcoxon Rank-Sum, Wilcoxon Signed-Rank and t-tests in Developer-Centered Usable Security*.

For more information, see [About](#).



<https://powerdb.info/>

[Click here to download the database as a SQLite file](#)

Operator

☒ All selected (AND) ☐ Any selected (OR)

Variable

Variable category

security

Participants type

Test

fishers

Data collection method

SEARCH

Website developed by Ahmad M. Assaf • University of Bonn
Copyright © University of Bonn 2023



Results:

1. Test: Fisher's Exact Test

Paper: *On Conducting Security Developer Studies with CS Students: Examining a Password-Storage Study with CS Students, Freelancers, and Company Developers*
(2020) Naiakshina et al.

Participants: professional software developer (N=36)

Dependent Variable

DV Categories

Independent Variable

IV Categories

security

"a binary variable secure indicating whether participants used any kind of security in their code"

1. Level: yes

Any kind of security was used in the code

2. Level: no

No kind of security was used in the code

Categories for security

1. security

prompting

condition set by researchers 'whether the participant is asked to store the password securely'

1. Level: true

participant was prompted for security

2. Level: false

participant wasn't prompted for security

Categories for prompting

1. study related variable

Effect sizes: odds ratio=46.33 | Cohen's d=2.11

Introduction

Introduction

Fisher's Exact Test ▾

Chi-Squared Test ▾

McNemar's Test ▾

Independent t-Test ▾

Paired t-Test ▾

Wilcoxon Rank-Sum Test ▾

Wilcoxon Signed-Rank Test ▾

A Guide to Power Analysis for Hypothesis Tests with One Categorical Independent Variable with Two Groups

Anna-Marie Ortloff Christian Tiefenau Matthew Smith
University of Bonn University of Bonn University of Bonn, Fraunhofer FKIE

What can I expect? ▾

What is Power Analysis? ^

Four parameters are relevant to power analysis: [power](#), the [significance criterion](#) (i.e. the α error level), the [reliability](#) of the sample results or sensitivity of the test, and the [effect size](#) [2]. These four parameters are interdependent, such that when three of them are available, it is possible to calculate the fourth. Such calculations are referred to as power analysis. In general, there are four different kinds of power analysis, each used to determine one of the parameters from the other three, although it is also possible to determine both α and power if a ratio for α and β is given together with the other two parameters - this is termed compromise power analysis [5]. The other four flavors are summarized, e.g. by Cohen [2] in Chapter 1.5.

The Four Parameters Explained ^

All of the mentioned parameters are explained here. Click the cards below to see the definition.

☰ Power ▾

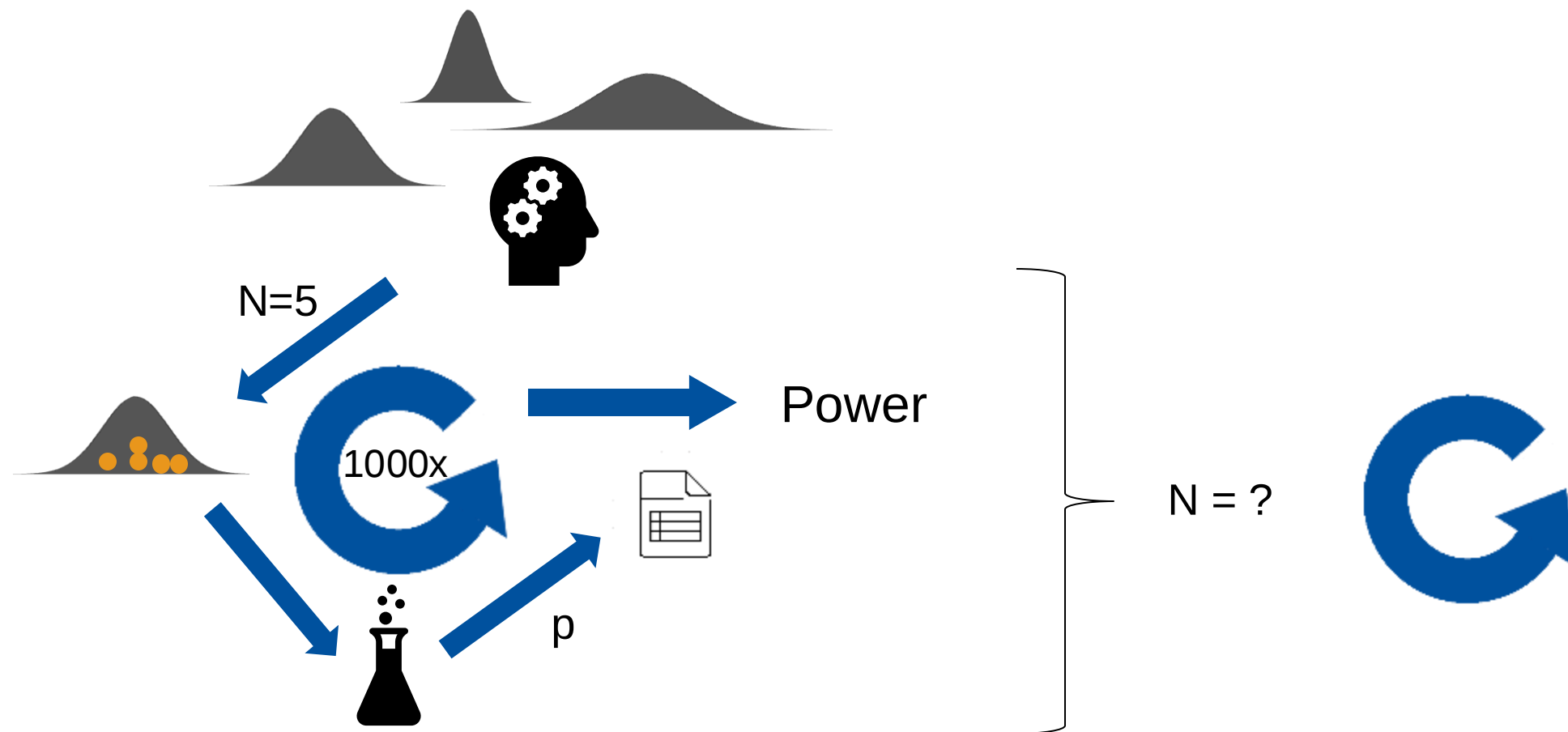
☰ Significance Criterion ▾

☰ Reliability (Sample Size) ▾

☰ Effect Size ▾

Example Power Analysis

Simulation for Power Analysis



“Not the Right Question?” A Study on Attitudes Toward Client-Side Scanning with Security and Privacy Researchers and a U.S. Population Sample

Lisa Geierhaas*, Florin Martius*, Arthi Arumugam[†], Matthew Smith[‡]

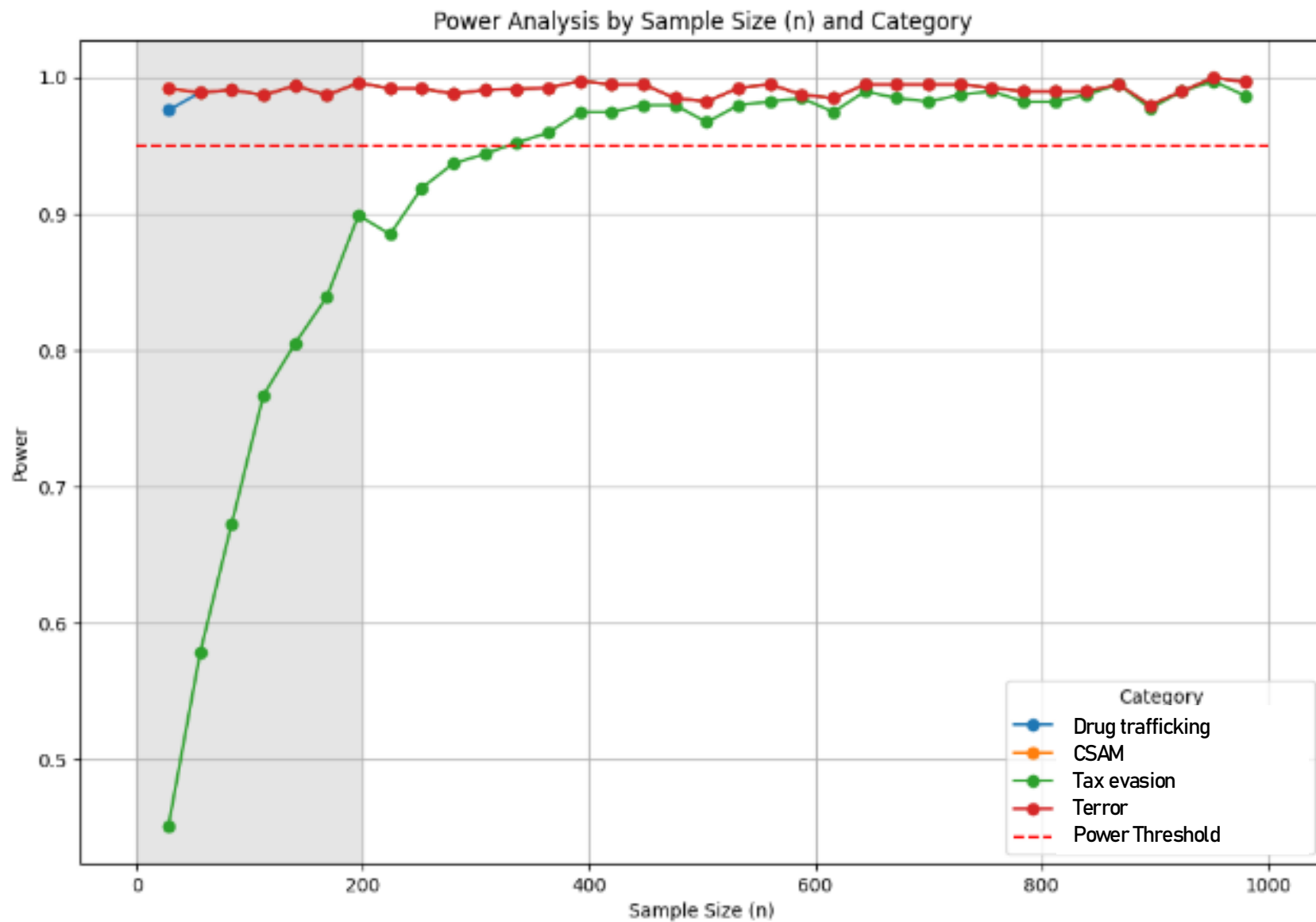
**University of Bonn, {geierhaa,martius}@cs.uni-bonn.de*

[†]University of Bonn, arumugam@uni-bonn.de

[‡]University of Bonn, Fraunhofer FKIE, smith@cs.uni-bonn.de

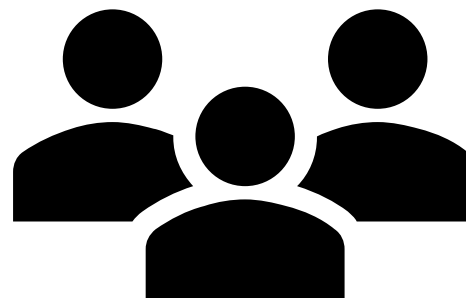
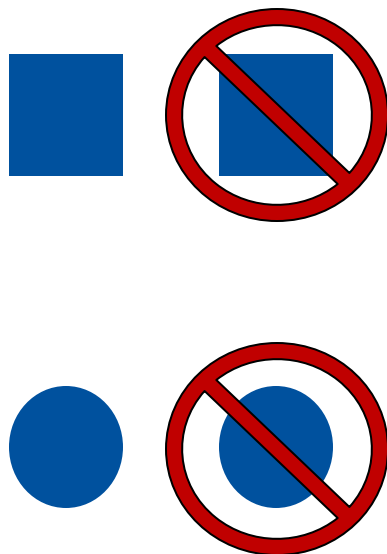
S&P'25

Results from Simulation

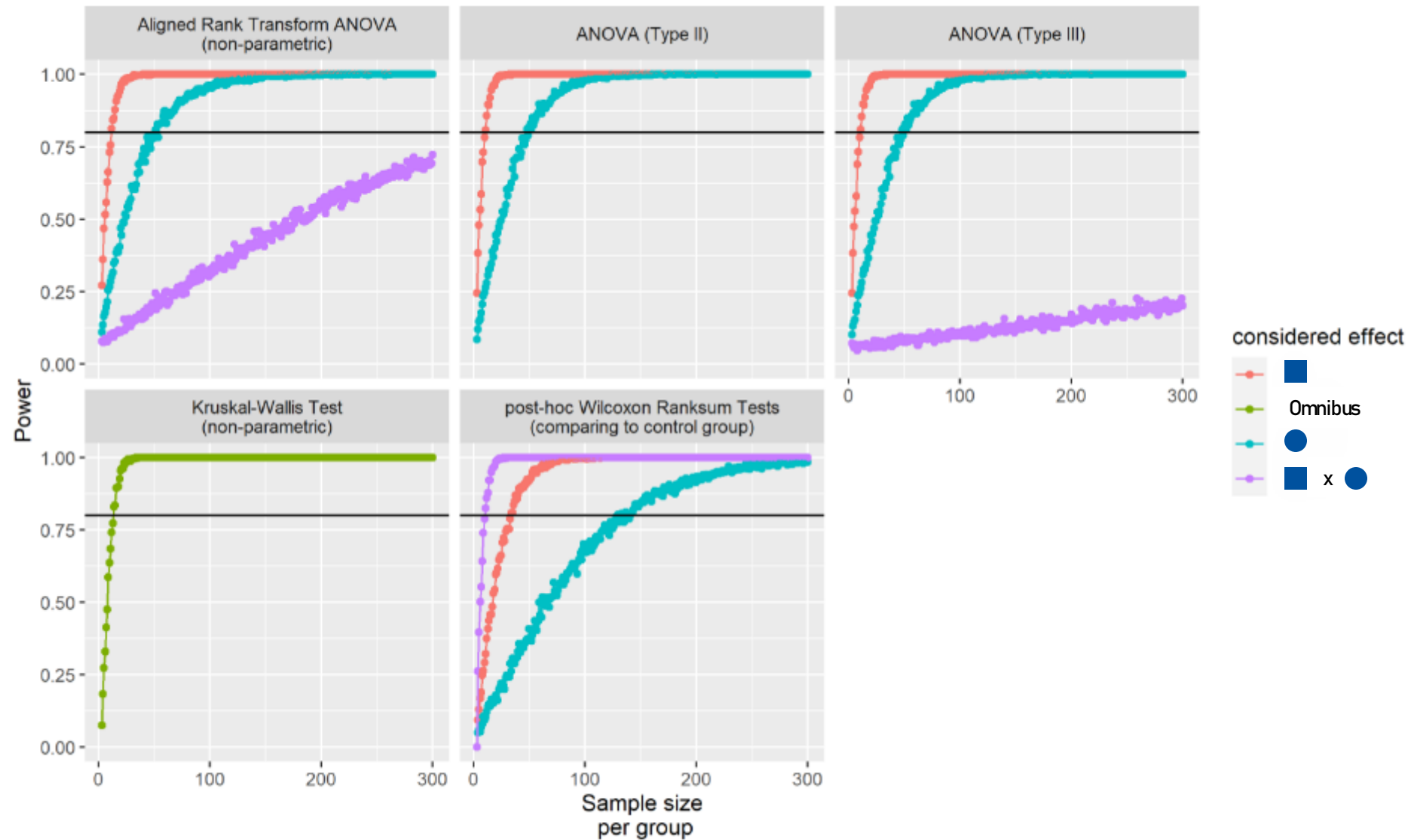


Example 2

- Currently unpublished work
 - Comparing two types of intervention in a fully crossed design
- Effect size estimate based on
 - small-N pilot
 - prior work with raw data/frequencies published in the paper



Results from Simulation



Simulation

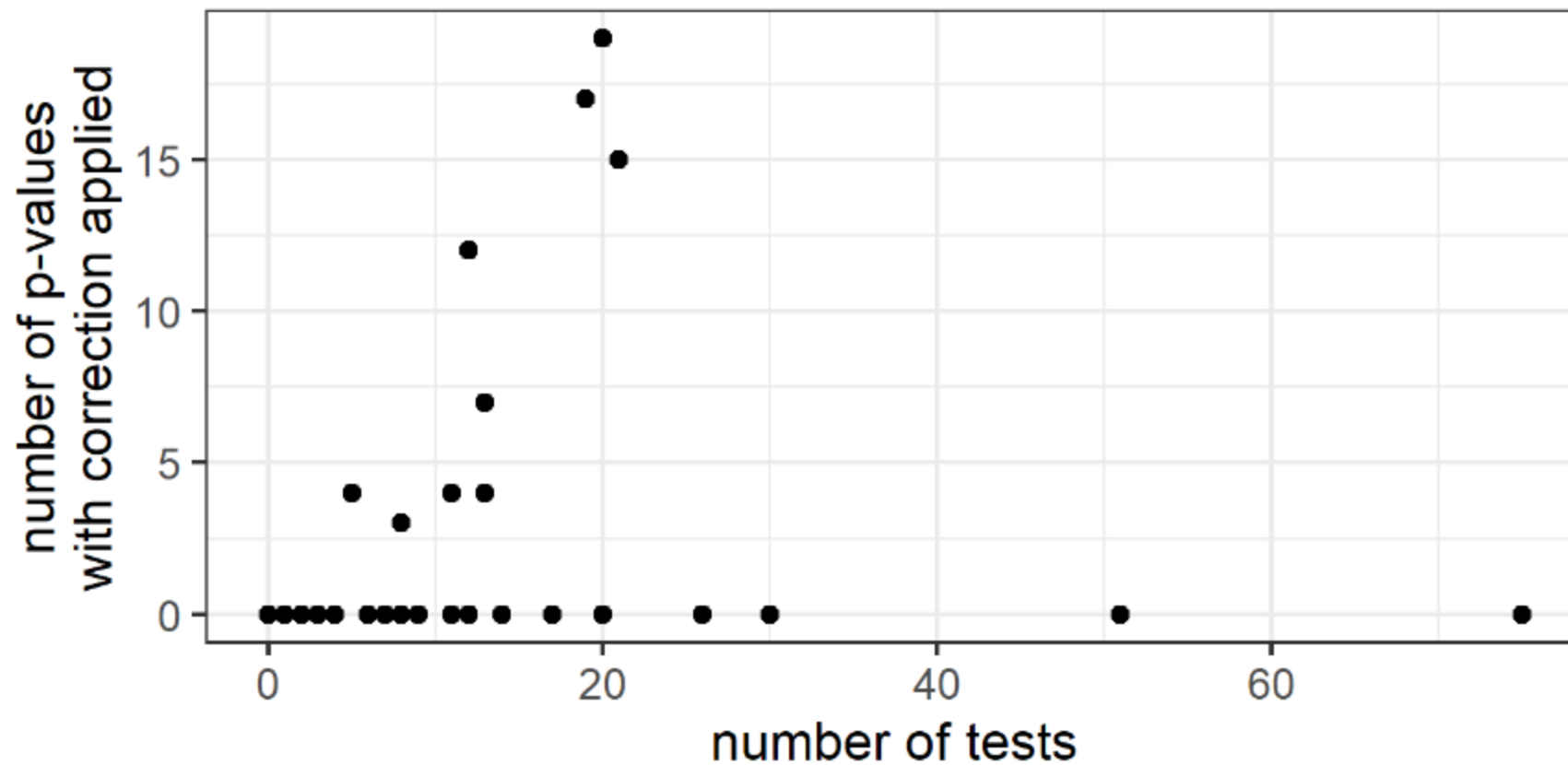
- One way: 14 / group
- Multi-way:
 - ● 48 / group
 - ■ 11 / group
- Post-hoc comparisons
 - ● 130 / group
 - ■ 34 / group
 - ●+■ 11 / group

Direct

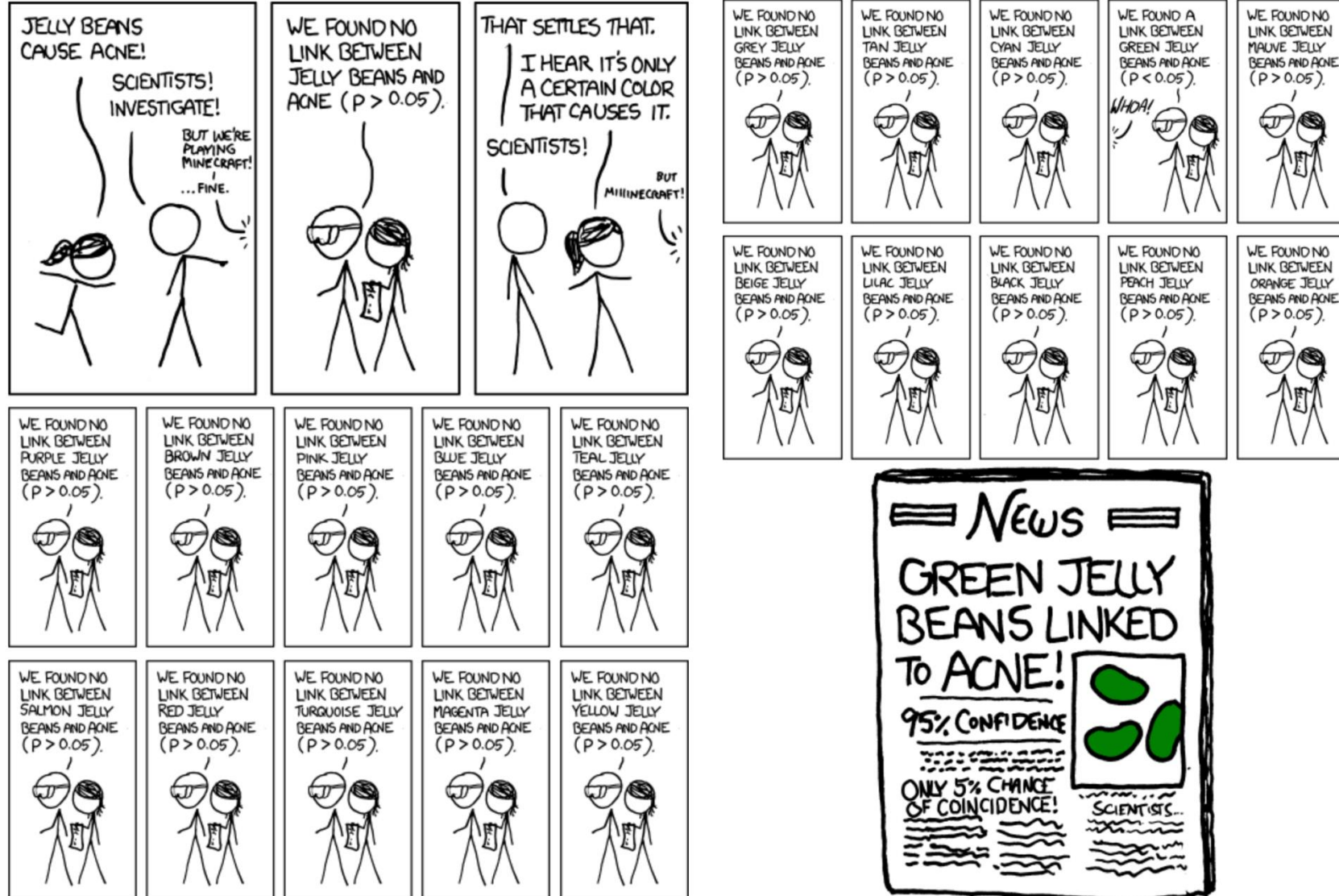
- One-way: 15 / group
- Multi-way
 - ● 55 / group
 - ■ 14 / group
- Post-hoc comparisons:
 - ● 159 / group
 - ■ 29 / group
 - ●+■ 9 / group

More Findings from DCUS

Many Statistical Tests

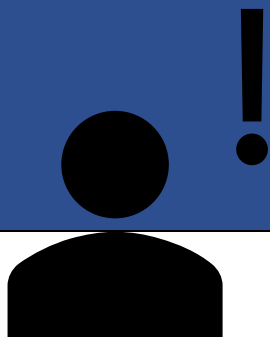


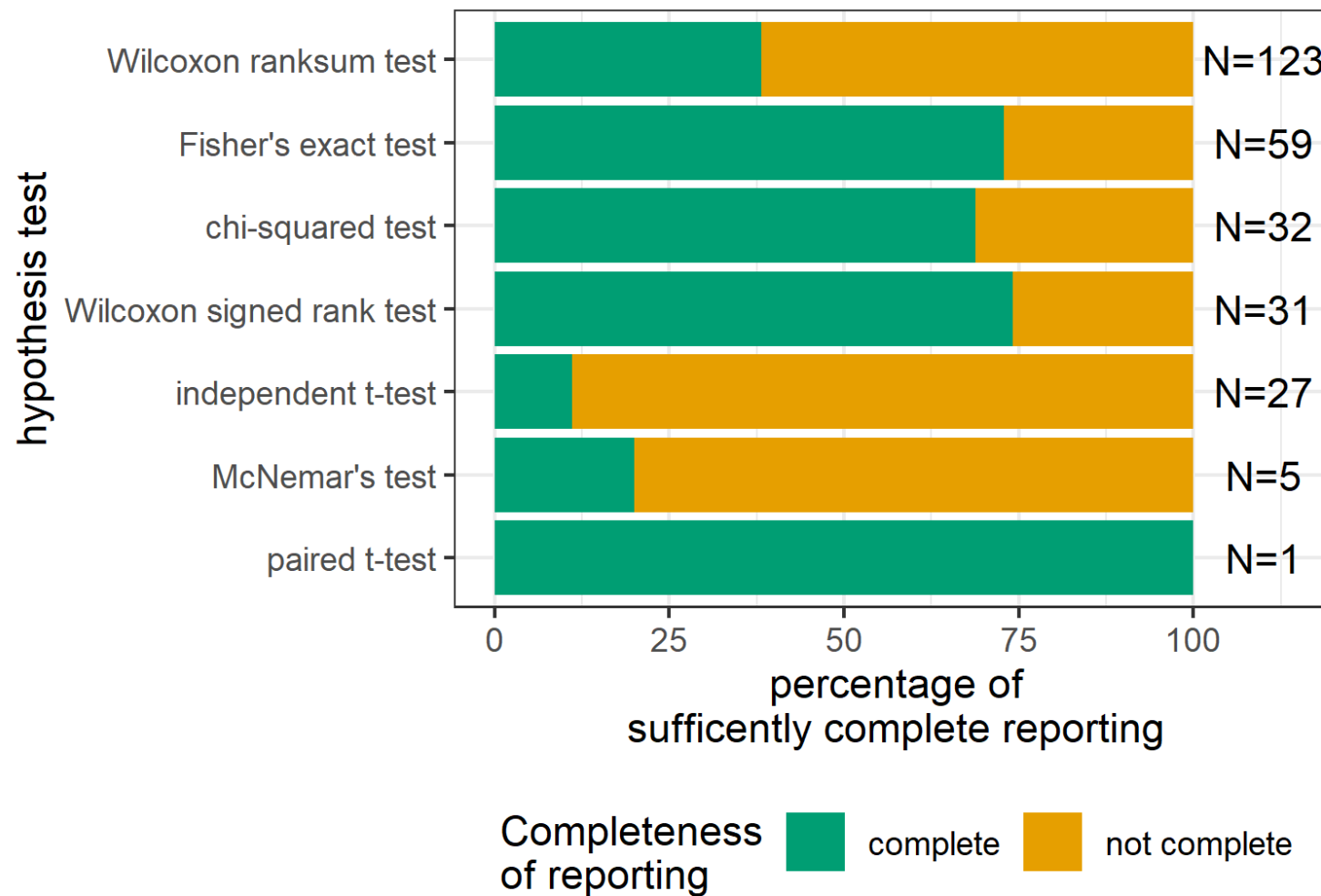
Too? Many Statistical Tests



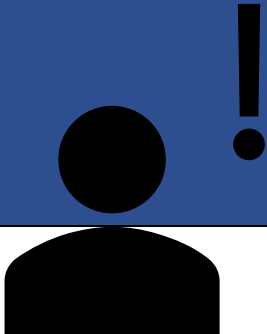
<https://xkcd.com/882/>

Specify if you are using
confirmatory or exploratory analysis methods.

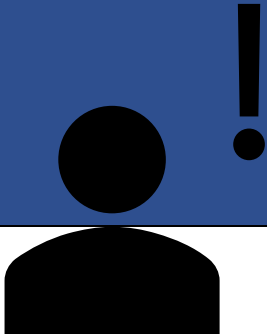


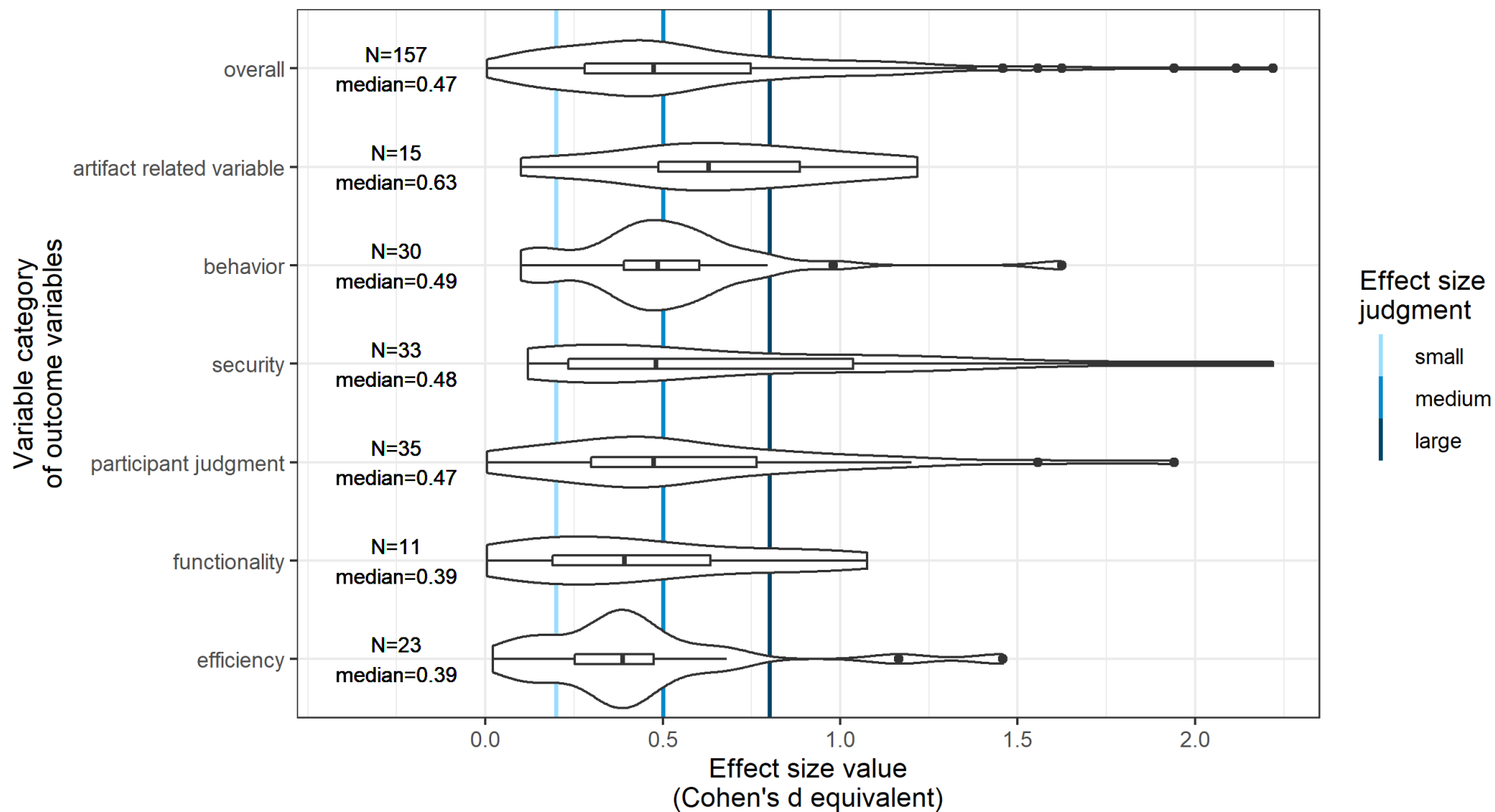


Report complete and appropriate descriptive statistics.



Make fully anonymized data sets available
when needed.





Extension: Effect sizes at CHI

Small, Medium, Large? A Meta-Study of Effect Sizes at CHI to Aid Interpretation of Effect Sizes and Power Calculation

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Theo Raimbault
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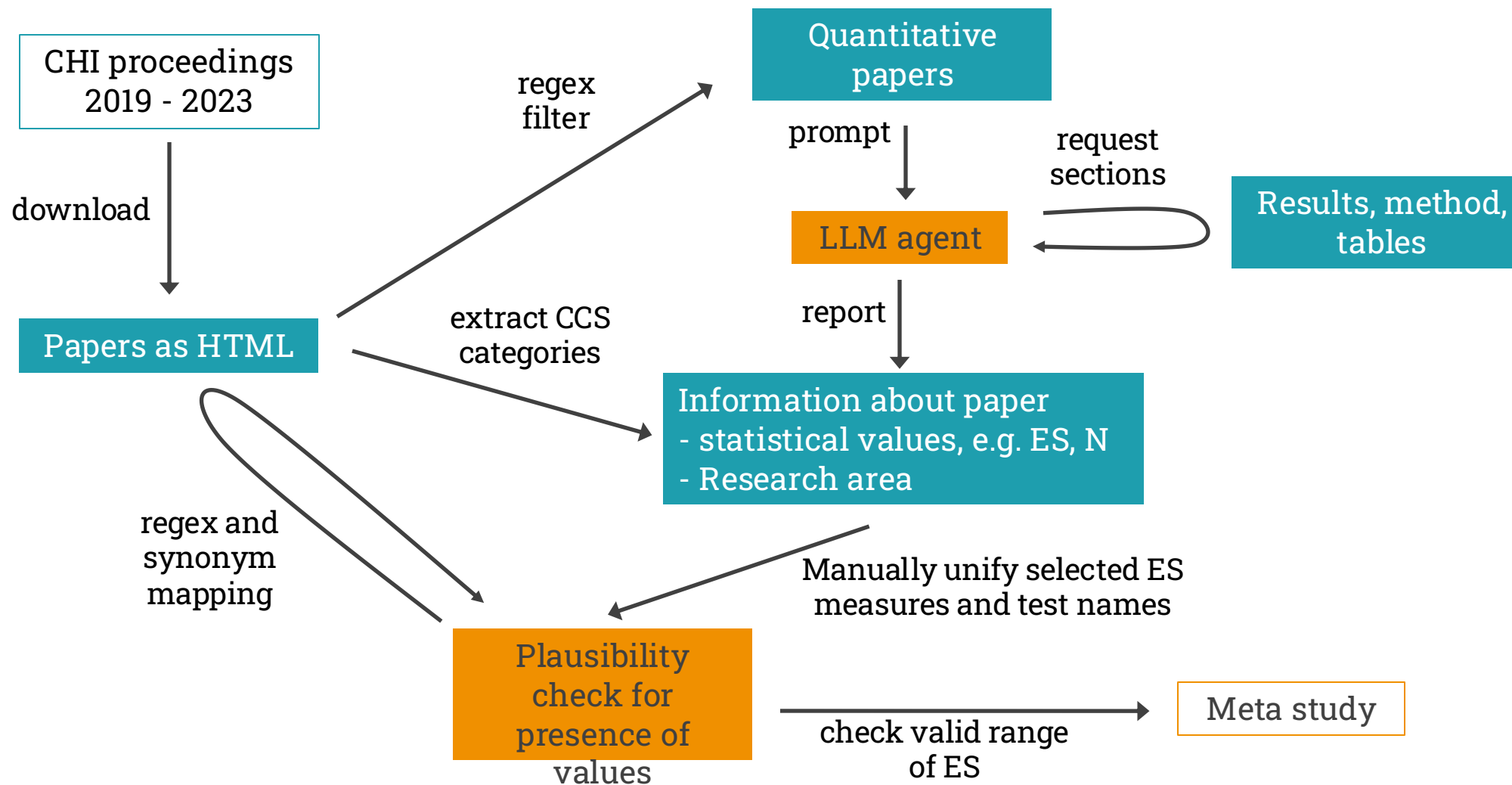
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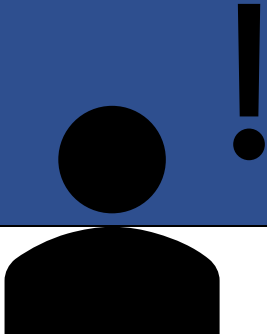
CHI'25

Find a way to standardize and make machine-readable
the research output / statistical results.

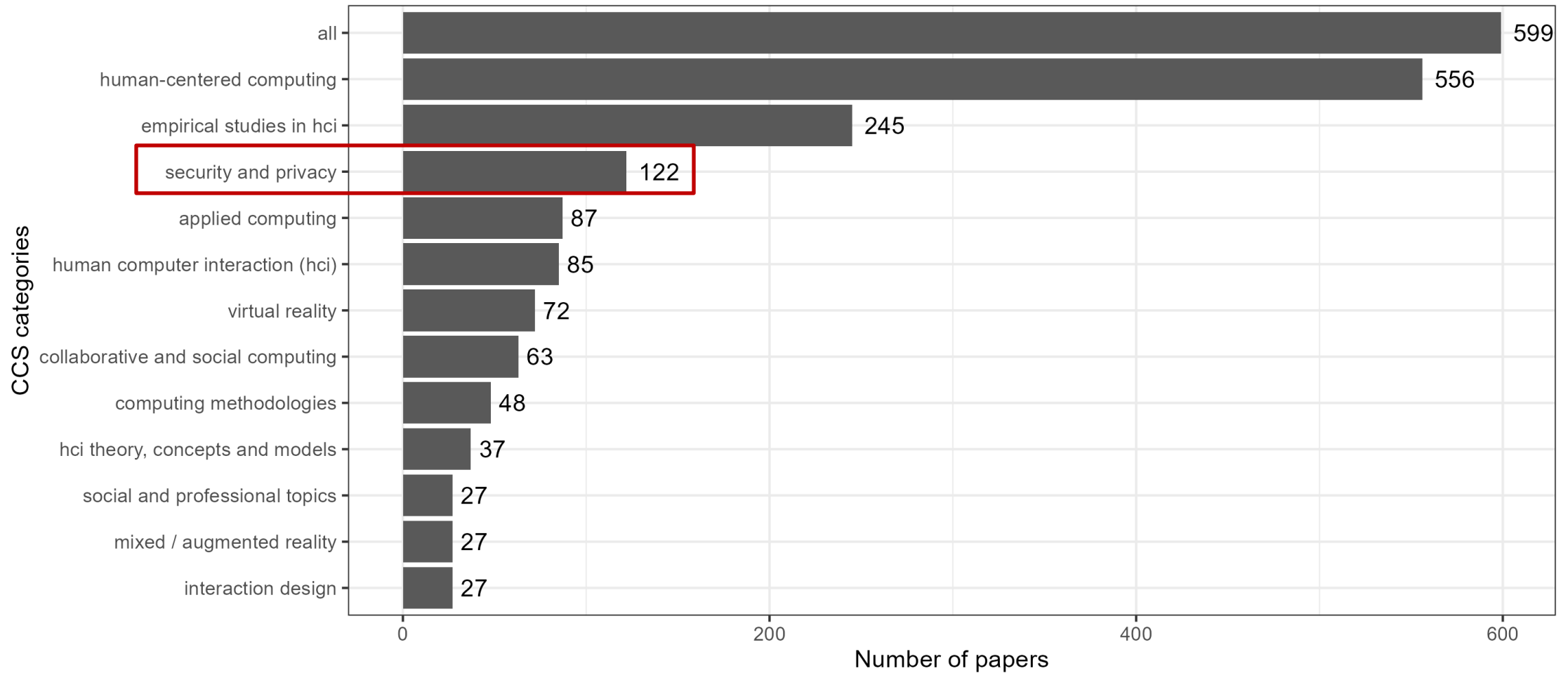




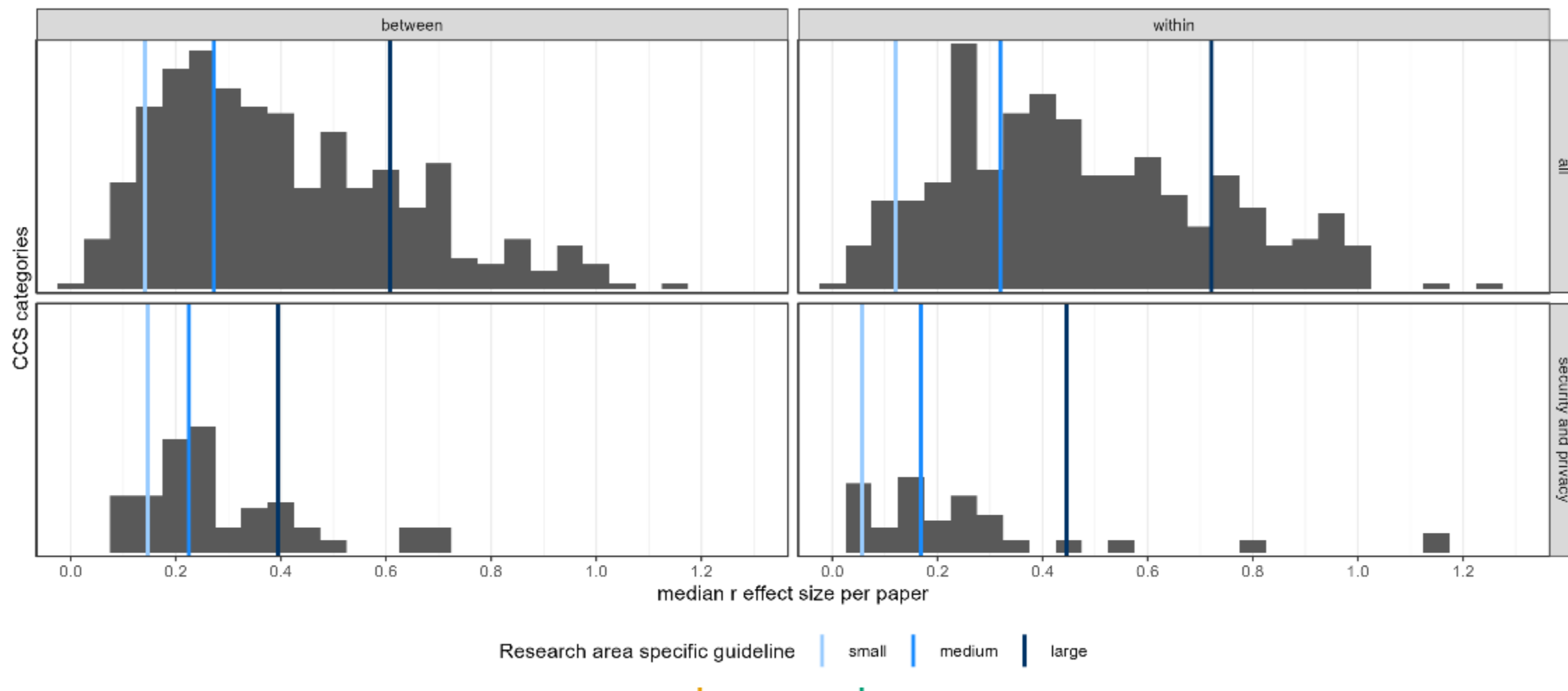
Report all effect sizes, even non-significant ones.



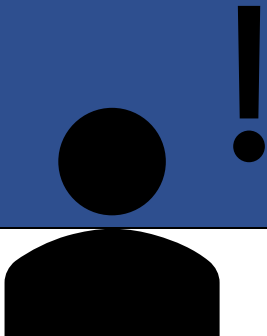
Overview of Quantitative CHI papers 2019 - 2023



Context-specific effect size guidelines



Take into account context
when interpreting effect sizes.



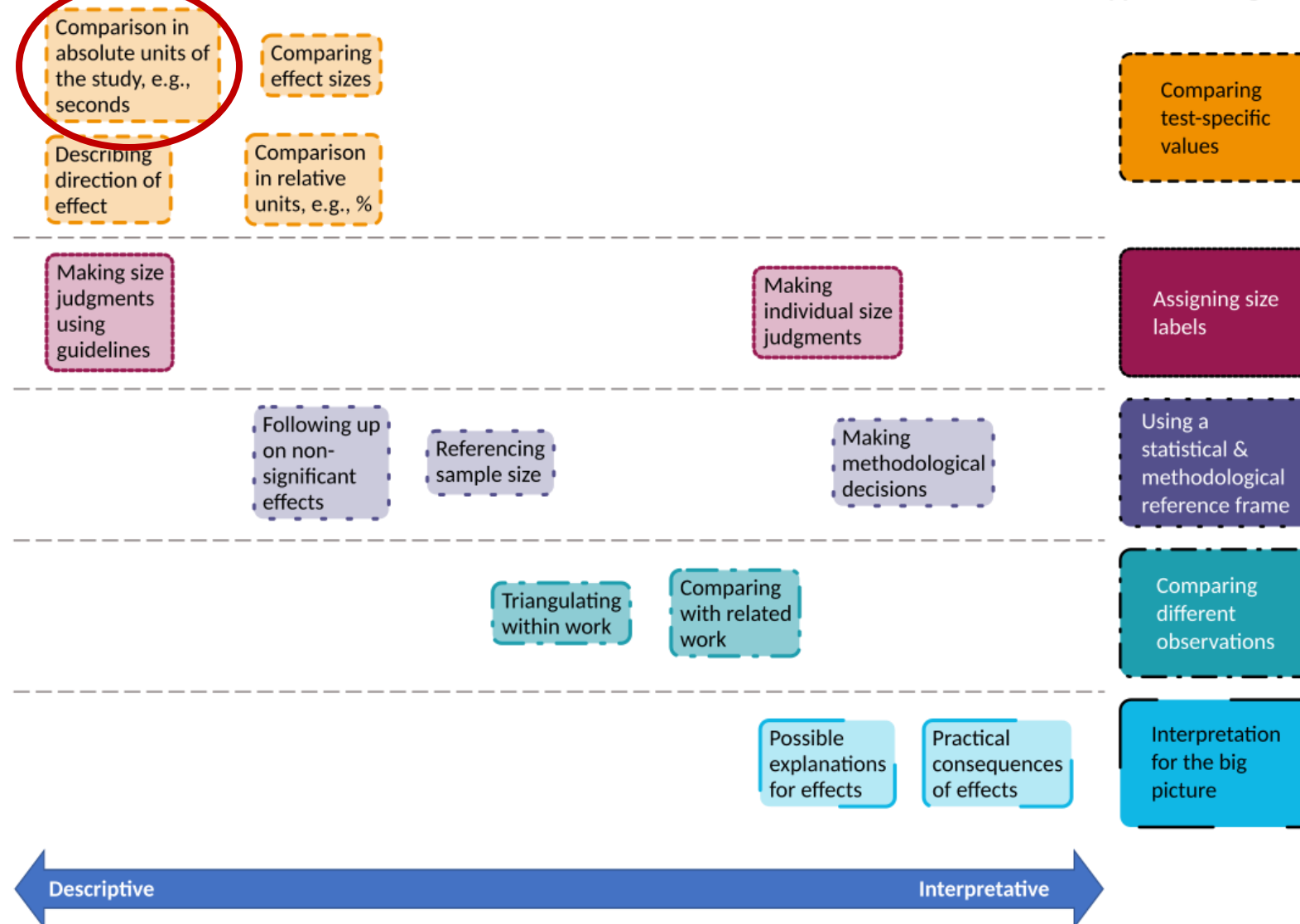
What about interpretation of effects?

<u>P-VALUE</u>	<u>INTERPRETATION</u>
0.001	]— HIGHLY SIGNIFICANT
0.01	
0.02	
0.03	
0.04	]— SIGNIFICANT
0.049	
0.050	]— OH CRAP. REDO CALCULATIONS.
0.051	]— ON THE EDGE OF SIGNIFICANCE
0.06	
0.07	]— HIGHLY SUGGESTIVE, SIGNIFICANT AT THE $P < 0.10$ LEVEL
0.08	
0.09	
0.099	]— HEY, LOOK AT THIS INTERESTING SUBGROUP ANALYSIS
≥ 0.1	

<https://www.xkcd.com/1478/>

Interpretation approaches

Approach categories

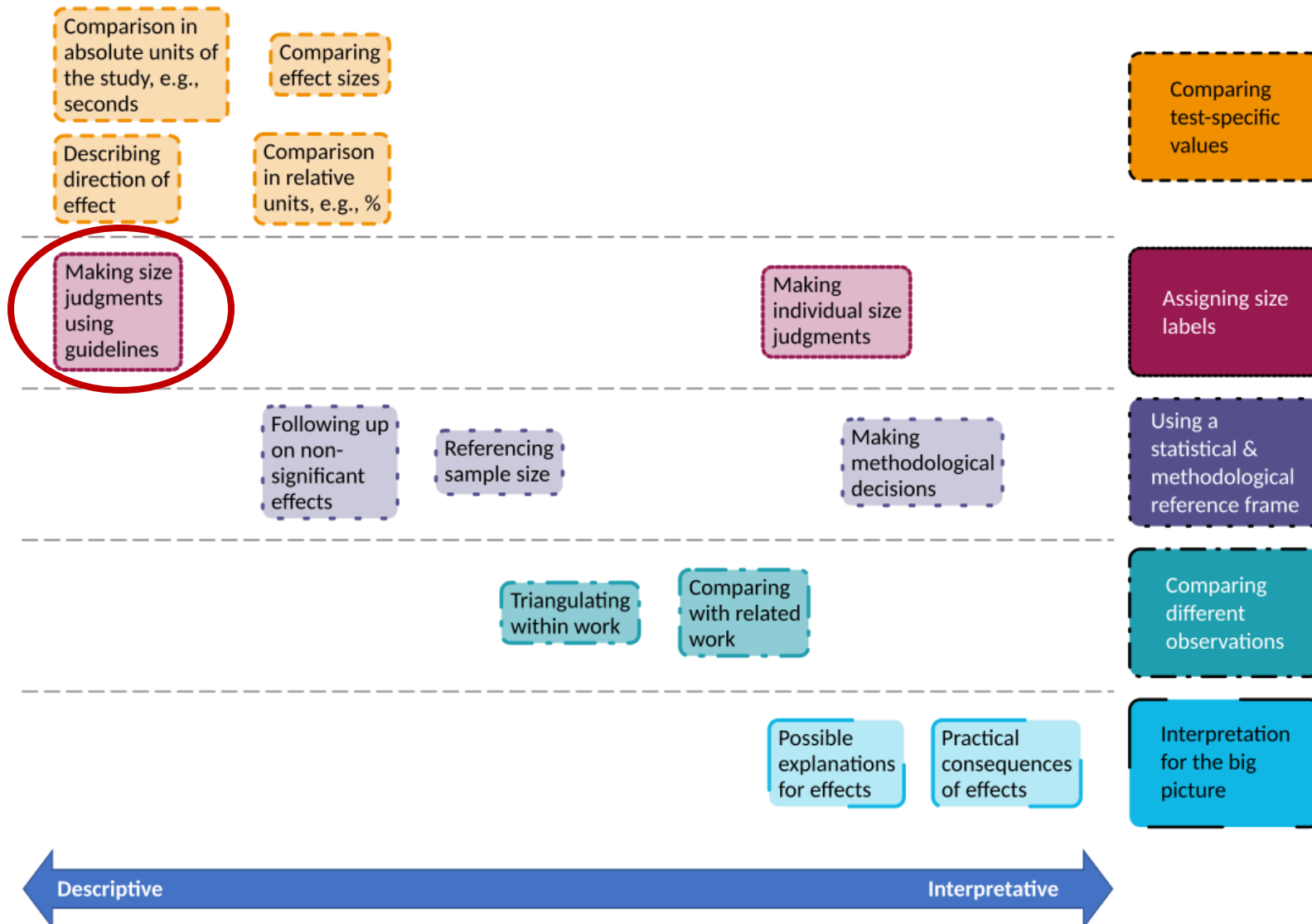


“[...] significantly increased the number of words in a sentence after which suggestions were requested – **by about 1.5 words**”

– Dang et al.: Choice over Control
<https://dl.acm.org/doi/full/10.1145/3544548.3580969>

Interpretation approaches

Approach categories

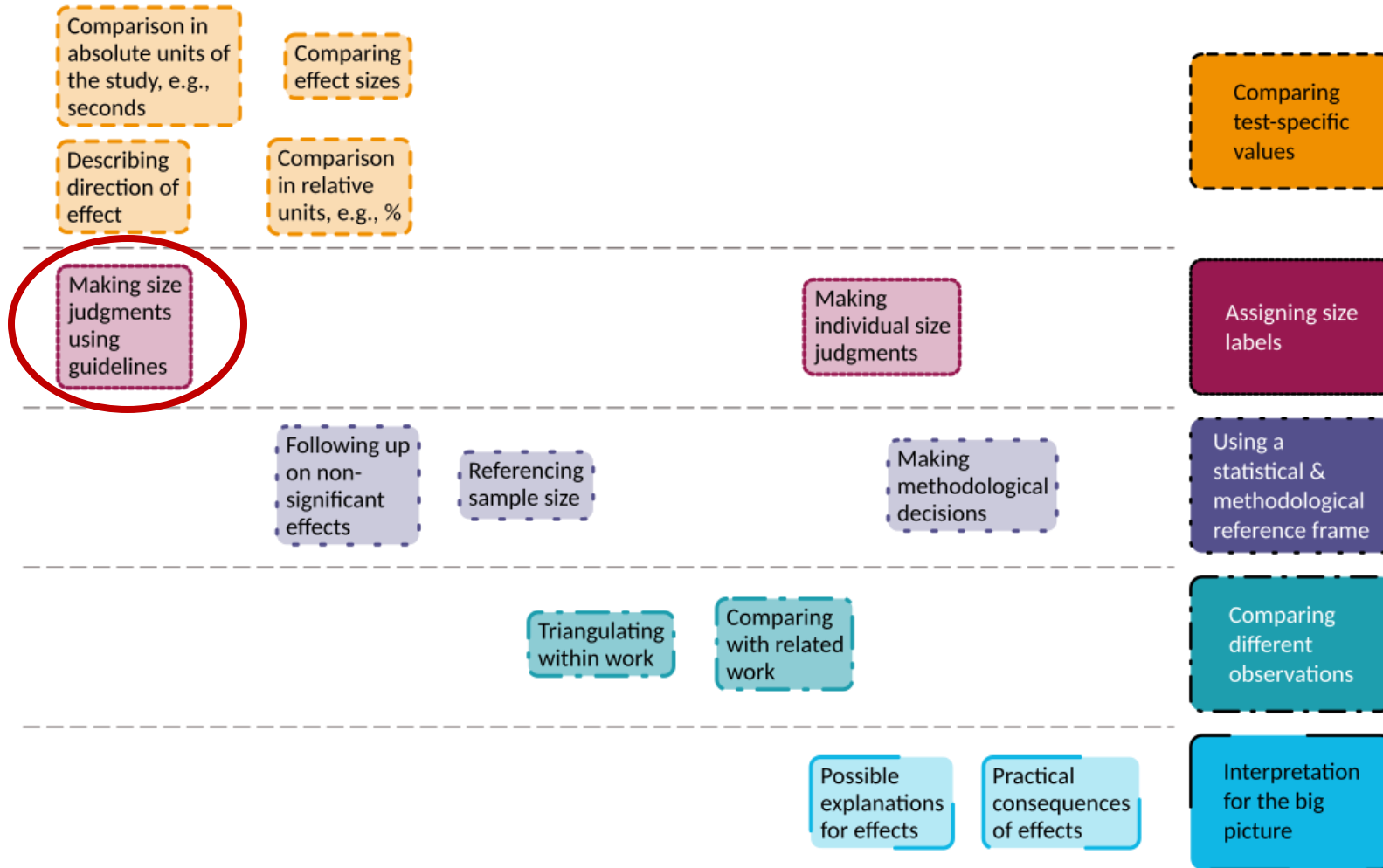


„Both plank (static core) and press-ups (dynamic upper) showed only non-significant, small effect sizes (Cohen's $d = 0.239$ and 0.298).”

- Clarke et al.: FakeForward (<https://dl.acm.org/doi/abs/10.1145/3544548.3581100>)

Interpretation approaches

Approach categories



P-VALUE	INTERPRETATION
0.001	HIGHLY SIGNIFICANT
0.01	
0.02	
0.03	
0.04	SIGNIFICANT
0.049	
0.050	OH CRAP. REDO CALCULATIONS.
0.051	ON THE EDGE OF SIGNIFICANCE
0.06	
0.07	HIGHLY SUGGESTIVE, SIGNIFICANT AT THE P<0.10 LEVEL
0.08	
0.09	
0.099	HEY, LOOK AT THIS INTERESTING SUBGROUP ANALYSIS
≥0.1	

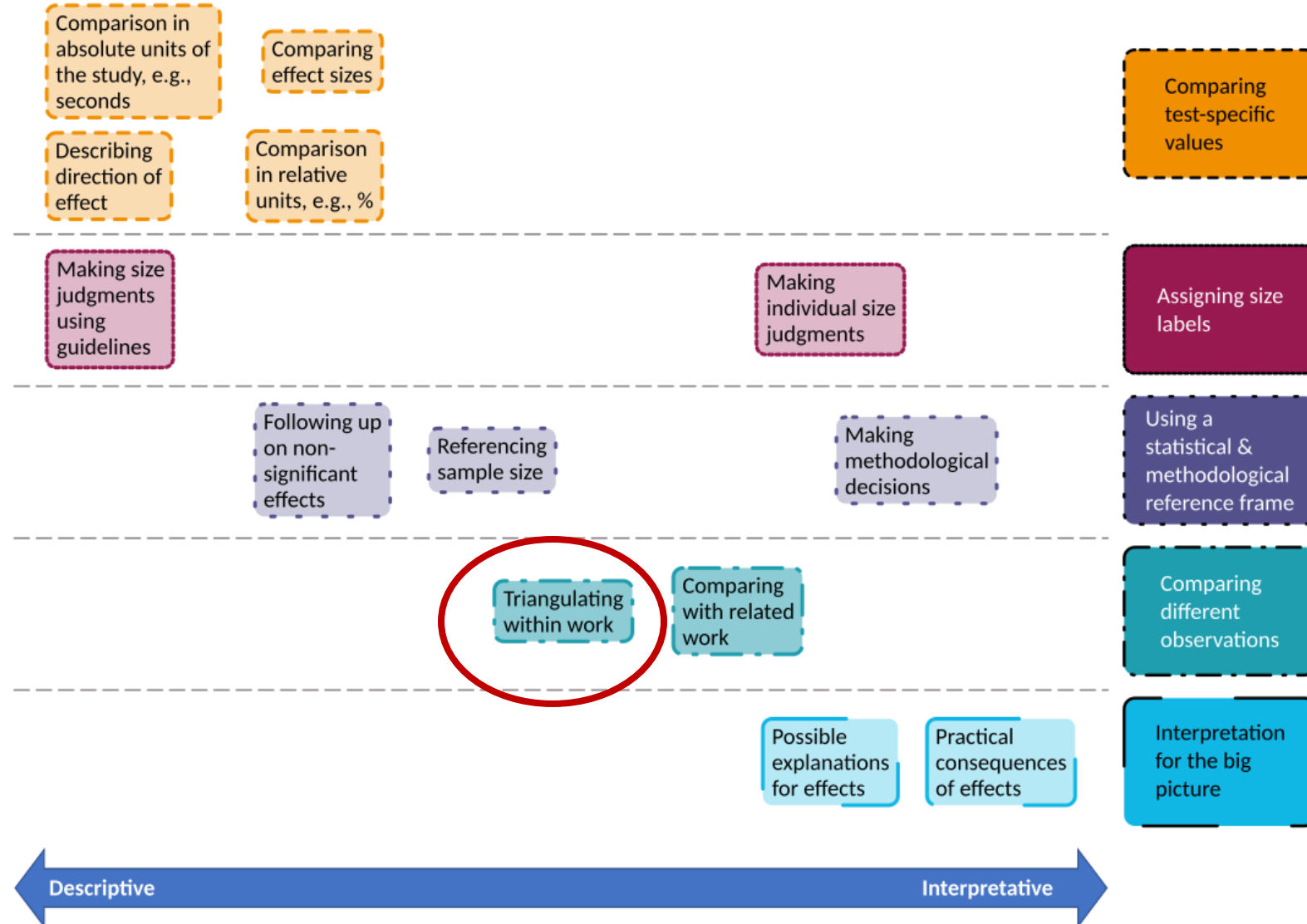
<https://www.xkcd.com/1478/>

Descriptive

Interpretative

Interpretation approaches

Approach categories

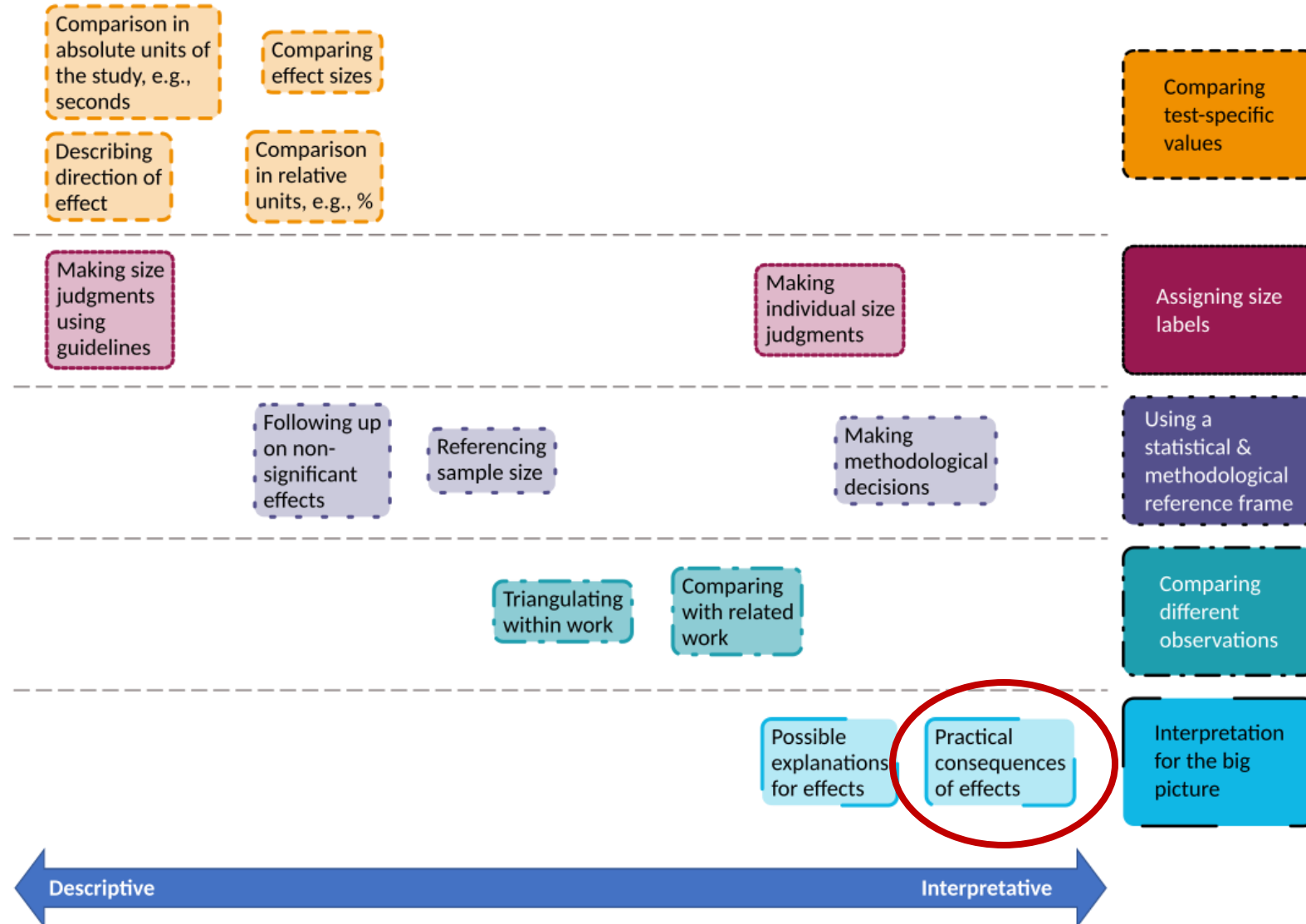


„We could not draw a definite conclusion about which algorithm was more accurate because GlanceWriter had a lower error rate but it also had a higher number of error correction actions.”

- Cui et al.: GlanceWriter (<https://dl.acm.org/doi/abs/10.1145/3544548.3581100>)

Interpretation approaches

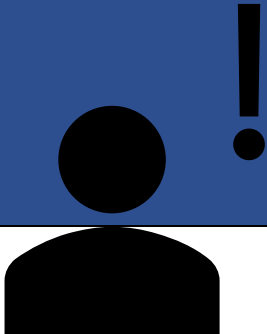
Approach categories



“While eye-tracking seems to be the favourite visual attention cue, not all VR headsets have eye-tracking capabilities, and perhaps future cheap VR headsets will never incorporate such capabilities. Our results show that [...], **cheap VR headsets using bi-directional CoV can still lead to the same amount of joint attention, therefore, being effective.**”

- Bovo et al.: Speech-Augmented Cone-of Vision
<https://dl.acm.org/doi/full/10.1145/3544548.3581283>

Interpret and Discuss Effect Sizes.



A Qualitative Study on How Usable Security and HCI Researchers Judge the Size and Importance of Odds Ratio and Cohen's d Effect Sizes

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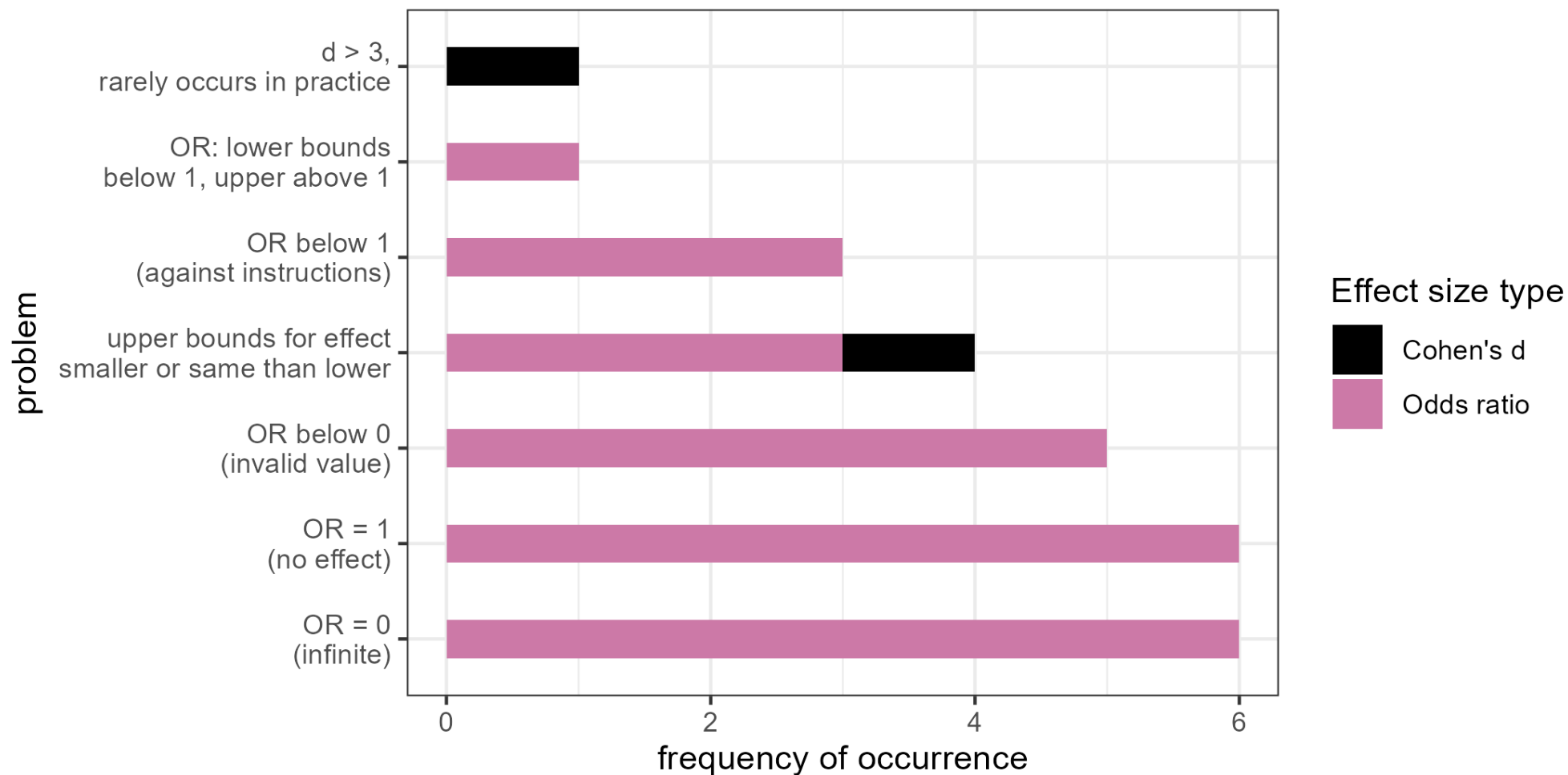
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CHI'25

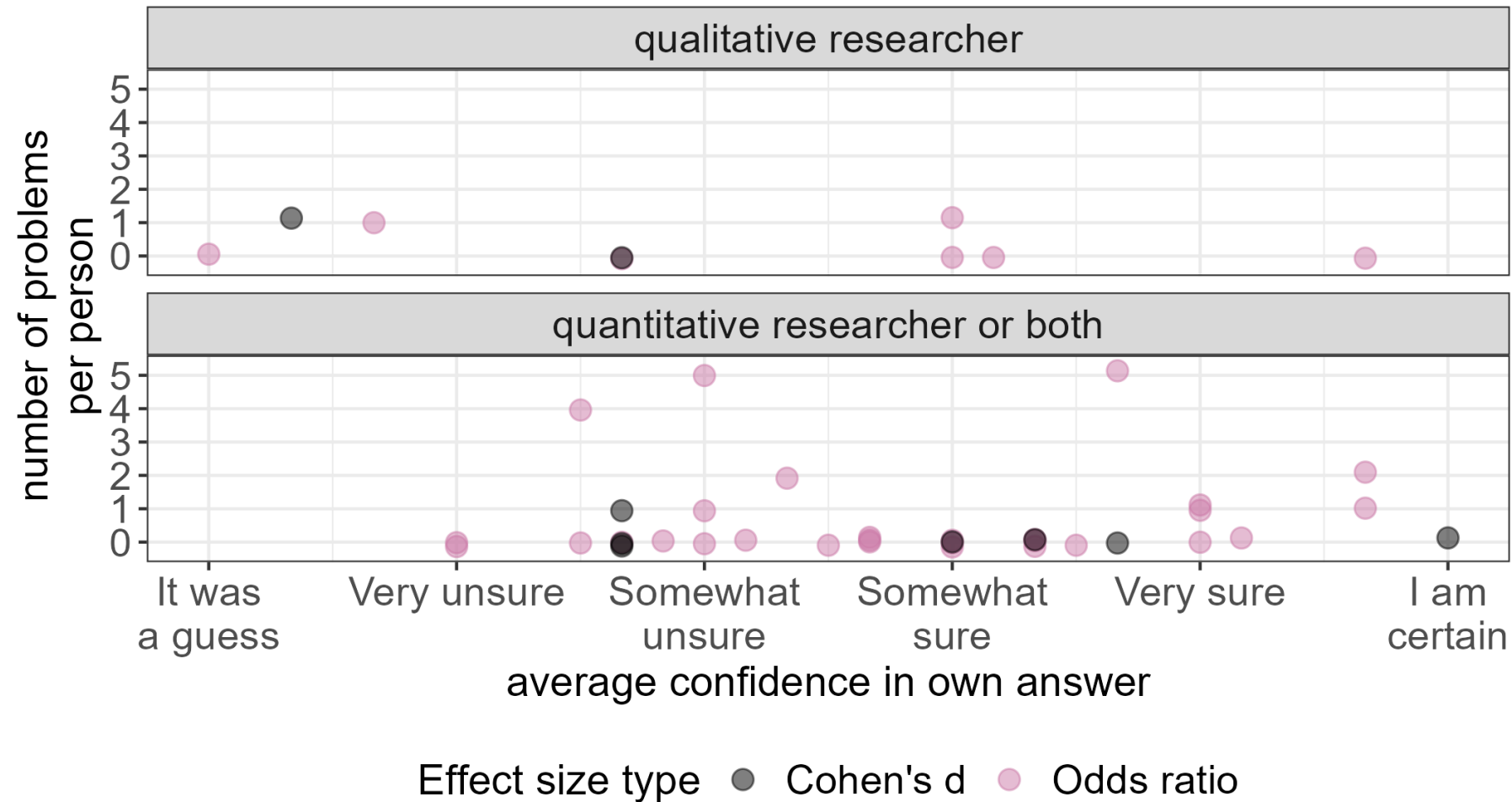
Vignette – Password Manager Example

A study investigated the difference between the adoption of a password manager in a baseline group that received a general introduction to the password manager and an intervention group that additionally was informed that using a password manager was the top recommendation made by security experts. 2100 Participants used the password manager during the study. Two weeks after the end of the study, they were asked whether they were still using the password manager or not.	RQ and study design
511 out of 1050 (48.7%) of those who received the intervention were still using it. 224 out of 1050 (21.3%) of those who did not see the intervention were still using it	Descriptive Stats
Fisher's exact test showed that this difference was statistically significant ($p < 0.001$, odds ratio = 3.50, 95% CI=[2.88, 4.25]). The effect size (Odds ratio) is 3.50.	Inferential Stats
This means that the odds for the participants to continue to use the formatting tool in the intervention group were 3.50 times higher than for participants to continue to use the password manager in the baseline group.	Effect size explanation

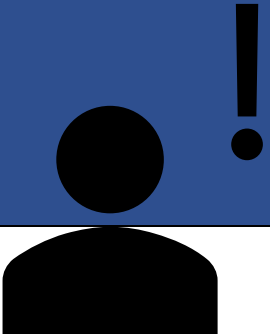
Misconceptions about OR and Cohen's d



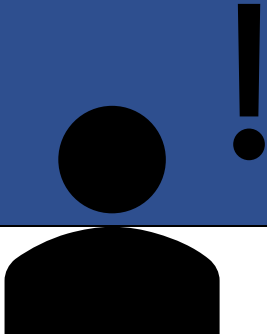
Understanding of Effect Size Measures



Explain effect size measures.



Report standardized **and** non-standardized
effect sizes.



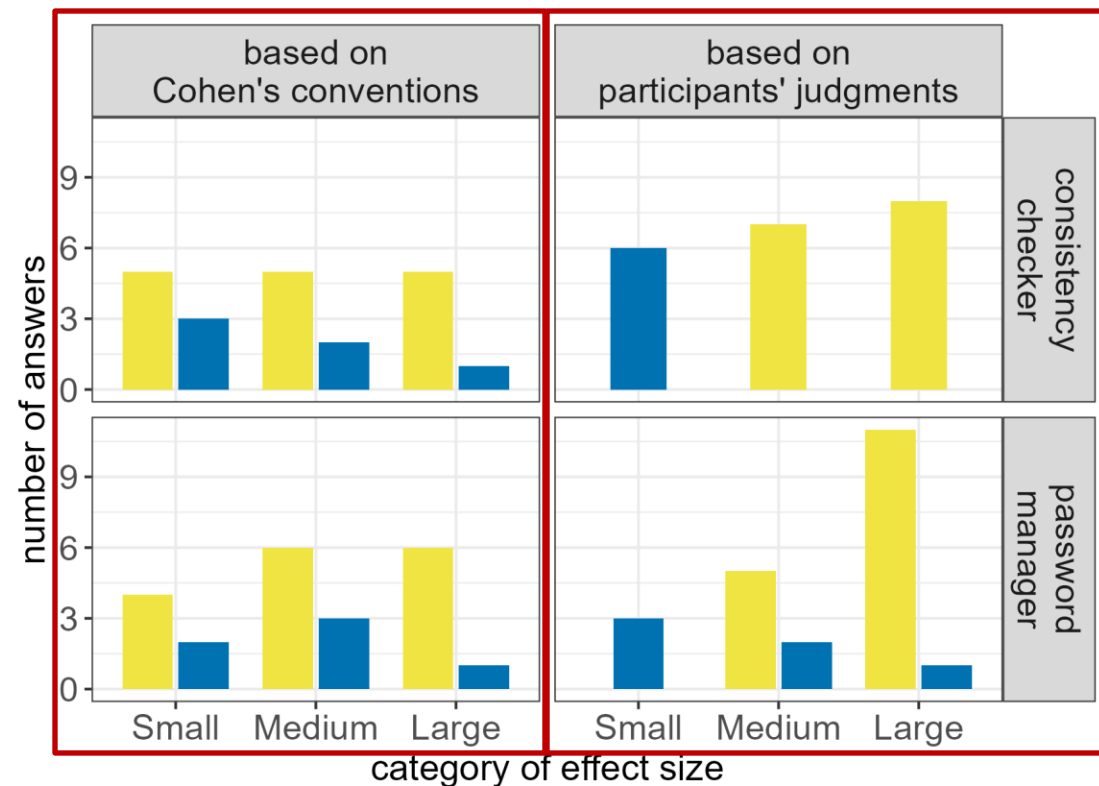
Influencing Factors on Interpretation of Effect Size

- Size
- Context
- Point of view
- Other numerical values in the vignette

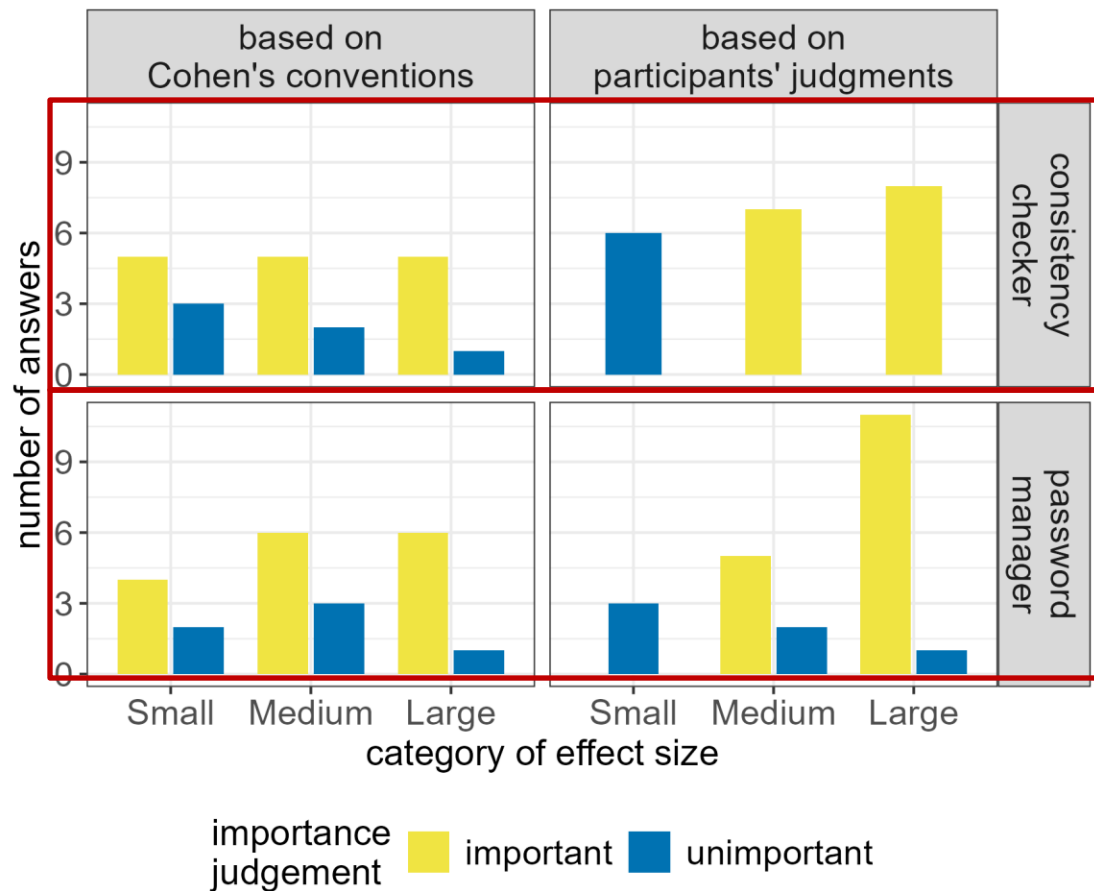
Influencing Factors on Interpretation of Effect Size

- **Size**
- **Context**
- Point of view
- Other numerical values in the vignette

Judgment Based on Size



importance judgement
 yellow important blue unimportant

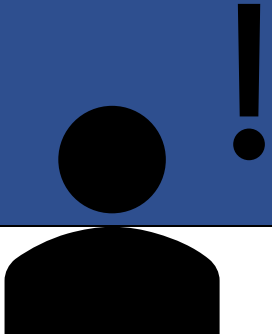


“**[Importance]** depends on the topic and the population. Using a password manager is not the same as curing cancer.” (from survey)

On effect **size**:

“Comparing an elephant to the earth, it's small, but compared to a mouse it is big” (from interview)

Interpret and Discuss Effect Sizes.



Recommendations for Reporting

- Specify if you are using confirmatory or exploratory analysis methods
- Report all effect sizes - even non-significant ones
- Report standardized and non-standardized ES
- Interpret and discuss ES
- Explain ES measures

Make Reporting Usable for Future Work

- Report complete and appropriate descriptive statistics
- Make fully anonymized data sets available when needed
- Find a way to standardize statistical research output and make it machine-readable

Recommendations up for Discussion

- Move test statistics to supplementary material
- Consider not reporting p-values and focus on confidence intervals instead

Develop / Use Reporting Guidelines.

