

Analyzing Cyber Security Research Practices through a Meta-Research Framework

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Analyzing cyber security research practices

- › Cyber security research should be **valid** and **sound**
 - ›› Use appropriate methods, evaluate and communicate well, ...
 - ›› Ensure reliable results, correct findings and well-justified claims
 - ›› Accurately reflect real-world security & propose effective solutions
- › Understanding *how* we do research can help us *improve* it

Analyzing Cyber Security Research Practices through a **Meta-Research** Framework

Meta-research

“Do research on research”

- › Critically **evaluate** research practices
- › Understand if research is **sound** and **reliable**
- › Assess presence/mitigation of research **biases**

Analyzing Cyber Security Research Practices through a Meta-Research Framework

Ioannidis et al. [Ioannidis 2015] introduced a framework for categorizing **meta-research** work

- › *Methods* performing research as best as possible
- › *Reporting* communicating research well
- › *Reproducibility* verifying research by reproducing it
- › *Evaluation* fairly evaluating research by peer review
- › *Incentives* rewarding research correctly and fairly

Analyzing Cyber Security Research Practices through a Meta-Research Framework

We **categorize** cyber security meta-research work

“Do meta-research on meta-research”

- › **Goal:** gain a better understanding of our community's efforts to examine its own research practices
- › **Process:** apply the framework by Ioannidis et al.
- › **Result:** characterize main areas of meta-research work
- › Encourage our community to continue self-reflection

Classification

Methods
Reporting
Reproducibility
Evaluation
Incentives

Methods

Reporting

Reproducibility

Evaluation

Incentives

Methods: *performing*

Goal: Conducting research using
the best scientific methods & practices (available)

Risk: Experiments and results
are not truly representative or accurate

Interest: high

Methods: *performing*

- › Best practices and pitfalls
 - › Common in many domains of cyber security research
Malware, machine learning, hardware, systems, social networks, ...
 - › Correct and open data sets, proper metrics and benchmarks, ...
 - › Possibly “flawed” prior work as examples
 - › Serve as a reference for future studies

Methods: *performing*

- › Qualitative methods
 - ›› Usually for usable security and privacy
 - ›› Special care/scrutiny to show validity of research methods
 - ››› Rest of security community: unfamiliar
 - ›› Best practices and guidelines, specific to qualitative methods
 - ››› But not always followed [Gro20,Kau21]

Methods: *performing*

- › Ethical considerations
 - ›› Existing frameworks for ethical review (e.g., IRB) may be unadapted
 - ›› Community has to set own ethical standards (+ provide guidelines)
 - ››› Security: Menlo Report
 - ››› Ethics increasingly enforced at top-tier conferences
 - ›› Controversial studies serve as use cases for lessons learned

Methods

Reporting

Reproducibility

Evaluation

Incentives

Reporting: *communicating*

Goal: Reaching the intended audience(s)
with research results relevant to them

Risk: Results are misinterpreted or misrepresented

Interest: medium

Reporting: *communicating*

› Publication bias

›› Under or overrepresented research

››› e.g., omission of negative results (shown in security user studies [Gro20])

››› e.g., more attack than defense papers?

› Preregistration

›› Stabilize research questions, hypotheses, methods, analyses, ...
before actual experiments take place

›› Very uncommon in security and privacy research

››› Due to exploratory or vulnerability-driven nature of studies?

Methods

Reporting

Reproducibility

Evaluation

Incentives

Reproducibility: *verifying*

Goal: Repeating a study to confirm its results and increase the likelihood that its hypothesis is correct

Risk: Failing to repeat a study puts validity of its results into question → “replication crisis” (?)

Interest: high

Reproducibility: *verifying*

› Artifacts

- ›› Sharing data sets and tools

- ››› Artifact evaluation (badges)

- ››› Still often fail to meet replicability criteria [Dem22]

Methods
Reporting
Reproducibility

Evaluation

Incentives

Evaluation: *evaluating*

Goal: Judging the quality of a research paper to maintain the integrity of science

Risk: Subjectivity could lead to published subpar papers and unpublished state-of-the-art-advancing papers

Interest: medium

Evaluation: *evaluating*

› Peer review

- ›› Top-tier security conferences [Son22]
 - ››› *Novelty* as only shared evaluation metric
 - ››› Various reasons to reject (“toxic culture of rejection”? [Lee22])
 - ››› Sense of ‘randomness’

Evaluation: *evaluating*

› Peer review

- › Trend towards journal-style model (i.e., revisions)
- › Good reviewing practices encouraged (meta-reviews, awards, ...)

- › A lot of trials, but also a lot of reversals?

Methods
Reporting
Reproducibility
Evaluation
Incentives

Incentives: *rewarding*

Goal: Evaluating the quality, value, and impact of research and providing the right incentives and support

Risk: Incentivizing “wrong” research (practices), improperly supporting “good” research

Interest: *medium-low*

Incentives: *rewarding*

› Rankings

›› Conference (tiers)

››› More restrictive = more prestigious

››› “Underappreciated” research? (e.g., replication studies)

›› Researchers, institutions

››› Criticism: biased or non-representative of quality

Discussion and conclusion

More meta-research work is being published

- › Strong focus on improving **methods**
 - › Best practices, analyzing data collection tools, data sets, ...
 - › Lack of central repository may make awareness & adoption difficult
 - › Enforcement: left as a task for peer review?
- › **Other categories**: less work, but more clarity
 - › Enforced or encouraged explicitly, with noticeable evolutions
e.g., ethical considerations, artifact badges, stricter peer review
 - › Less (academic) reflection?

Meta-research is a collective community effort

- › Venues like CSET support discussion of research practices
- › Research communities can **learn from each other**
 - ›› Meta-reviews: gone in Internet measurement, back in security?
 - ›› *Introspectively: framework from biology can be reused in security*
 - ›› Some concerns are common to all fields (e.g., incentivization)
- › But all communities have their **own accents**
 - ›› Badging as artifact encouragement; lack of preregistration

Cyber security meta-research

contributes to more reliable and
trustworthy **cyber security research**

and therefore helps to improve
cyber security itself

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