# Understanding De-identification Guidance and Practices for Research Data

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## Motivating examples



**Academic** researchers are studying restrictions on reproductive care.

They survey women in areas where abortion is criminalized about barriers to access.

Public data has benefits

- Replication, meta-analysis
- Transparency for public funds
- Required by journals/funders



**Evaluators** are contracted to assess the impact of foreign aid programs in conflict zones.

They survey residents about perceptions of organized crime and terrorism.

But de-id is challenging

- Traditional methods flawed
- Diff. privacy has accessibility & acceptability barriers

## Our research



#### Analyzing guidance

Thematic analysis of 38 de-id guides (pub. post-2018)

- What techniques?
- Framing of outcomes
- Usability



## **Conducting interviews**

Interviews of 26 experienced researchers and reviewers • How do they de-id data? • Perceptions of threats

- Challenges





Guides still skew towards traditional methods

- 28: pseudonymization •
- 17: *k*-anonymity •
- 11: differential privacy

## Gaps in threat coverage

- Listing salary and medical diagnosis as non-identifying info
- Examples where deleted data can be deduced from context



Perceiving unlikely threats, practitioners use heuristic methods, fail to prevent singling out

"You could crosstab all variables in theory, but that would be like millions of crosstabs. It's not necessarily a scientific process. It's **more knowing** what to look for."

## Funders & repositories sometimes push for weaker de-id

They felt if you've **removed all** the really obvious things like name, state, town of residence, and date of birth then that's probably enough.





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• 36 out of 38 guides: generalization (coarsening)

