

A Modern Approach to Privacy Law and Policy

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Privacy Tools for Data Sharing: Lessons Learned and Directions Forward Workshop

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Motivation: Data Sharing Challenges

❖ **Variations in Legal Requirements**

Approaches may require time-consuming, case-by-case review and negotiation processes and result in markedly different treatment.

❖ **Inadequacy of Current Practice**

Approaches in common use (e.g., de-identification, notice and consent) are increasingly being shown to fail to provide reasonable protection.

❖ **Reduced Utility**

Data are heavily redacted or withheld altogether due to privacy concerns, often overly restricting access and limiting future analyses on the data.

An Interdisciplinary Collaboration

Computer Science

Kobbi Nissim, Aaron
Bembenek, Mark Bun,
Marco Gaboardi, Thomas
Steinke, Salil Vadhan



CRCS Center for Research on
Computation and Society
at Harvard John A. Paulson School of Engineering and Applied Sciences

Law & Policy

Urs Gasser,
David O'Brien,
Alexandra Wood



**BERKMAN
KLEIN CENTER**
FOR INTERNET & SOCIETY
AT HARVARD UNIVERSITY

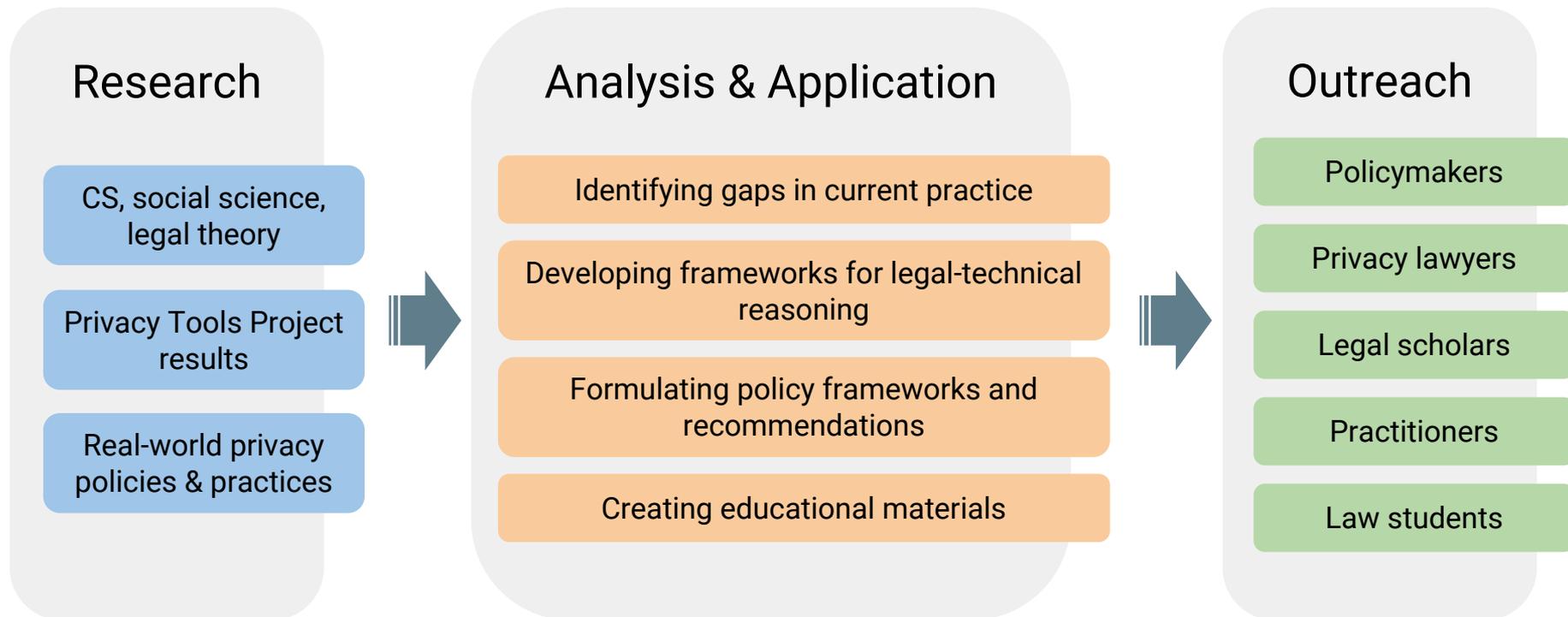
Social Science & Information Science

Micah Altman



Approach

Transferring project findings and insights from the technical literature to lawyers, policymakers, and practitioners



Framework for a Modern Privacy Analysis

Gaps in US Regulatory Framework

- ❖ Regulations are sector- and context-specific and highly variable.
- ❖ Guidance for implementing privacy safeguards is limited.
- ❖ Standards and guidance encourage use of a narrow subset of available privacy safeguards.
- ❖ Different actors treat similar privacy risks vastly differently.

Framework for Privacy-Aware Data Releases

Modeled on information security and lifecycle frameworks:

1. Developing a catalog of privacy controls
2. Identifying information uses, threats, and vulnerabilities
3. Designing data releases by aligning uses and risks with controls—at each stage of the information lifecycle

Catalog of Privacy Controls

Procedural, technical, educational, economic, and legal means for enhancing privacy—at each stage of the information lifecycle

	Procedural	Economic	Educational	Legal	Technical
Access/Release	Access controls; Consent; Expert panels; Individual privacy settings; Presumption of openness vs. privacy; Purpose specification; Registration; Restrictions on use by data controller; Risk assessments	Access/Use fees (for data controller or subjects); Property rights assignment	Data asset registers; Notice; Transparency	Integrity and accuracy requirements; Data use agreements (contract with data recipient)/ Terms of service	Authentication; Computable policy; Differential privacy; Encryption (incl. Functional; Homomorphic); Interactive query systems; Secure multiparty computation

Guide to Selecting Appropriate Privacy Controls

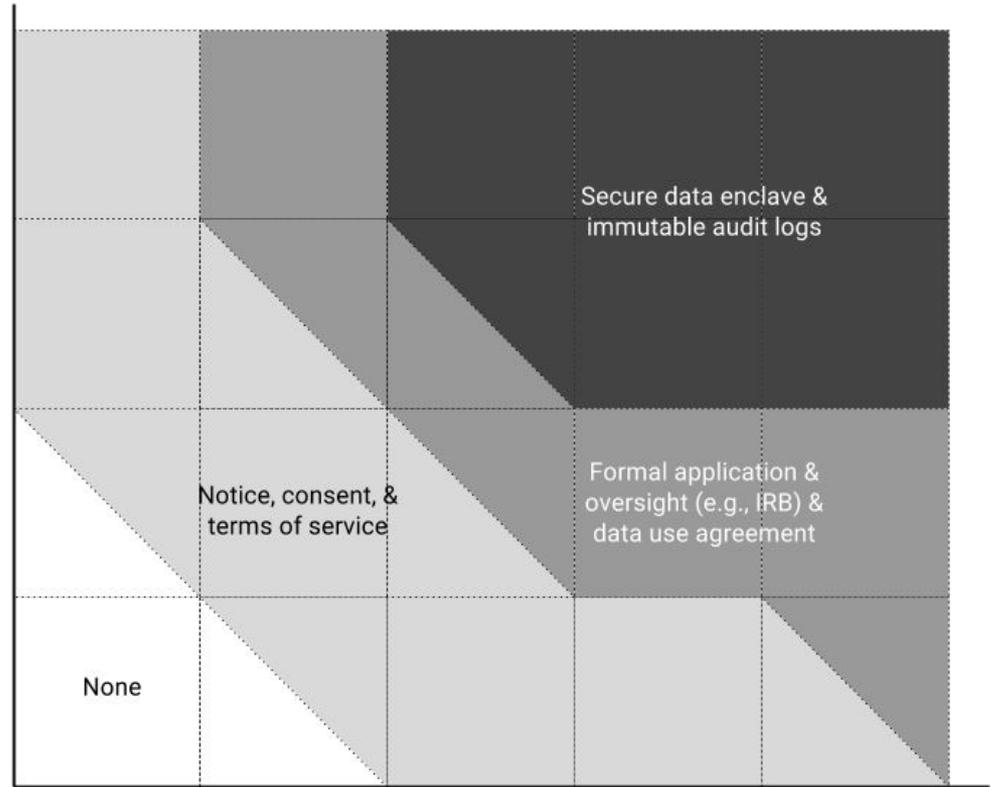
**Post-transformation
Identifiability
(Difficulty of Learning
about Individuals)**

Direct or
Indirect Identifiers
Present

Direct and
Indirect Identifiers
Removed

Heuristic (S)DL
Techniques Applied
(e.g., aggregation,
generalization, noise
addition)

Rigorous (S)DL
Techniques
Applied by Experts
(e.g., differentially
private statistics, secure
multiparty computation)



Negligible

Minor & Fleeting (e.g., temporary embarrassment)

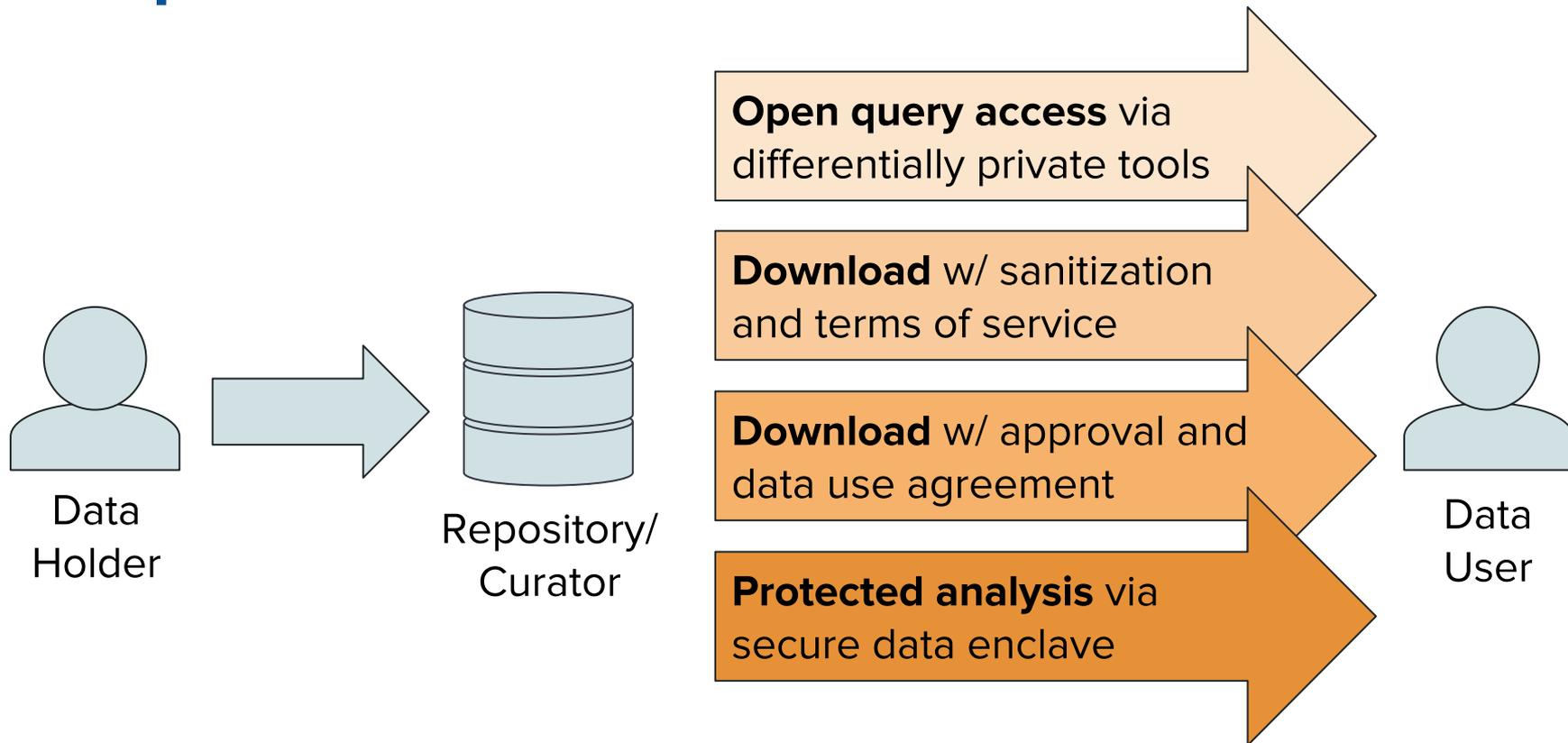
Significant & Lasting (e.g., long-term reputational harm)

Life Altering (e.g., divorce, imprisonment)

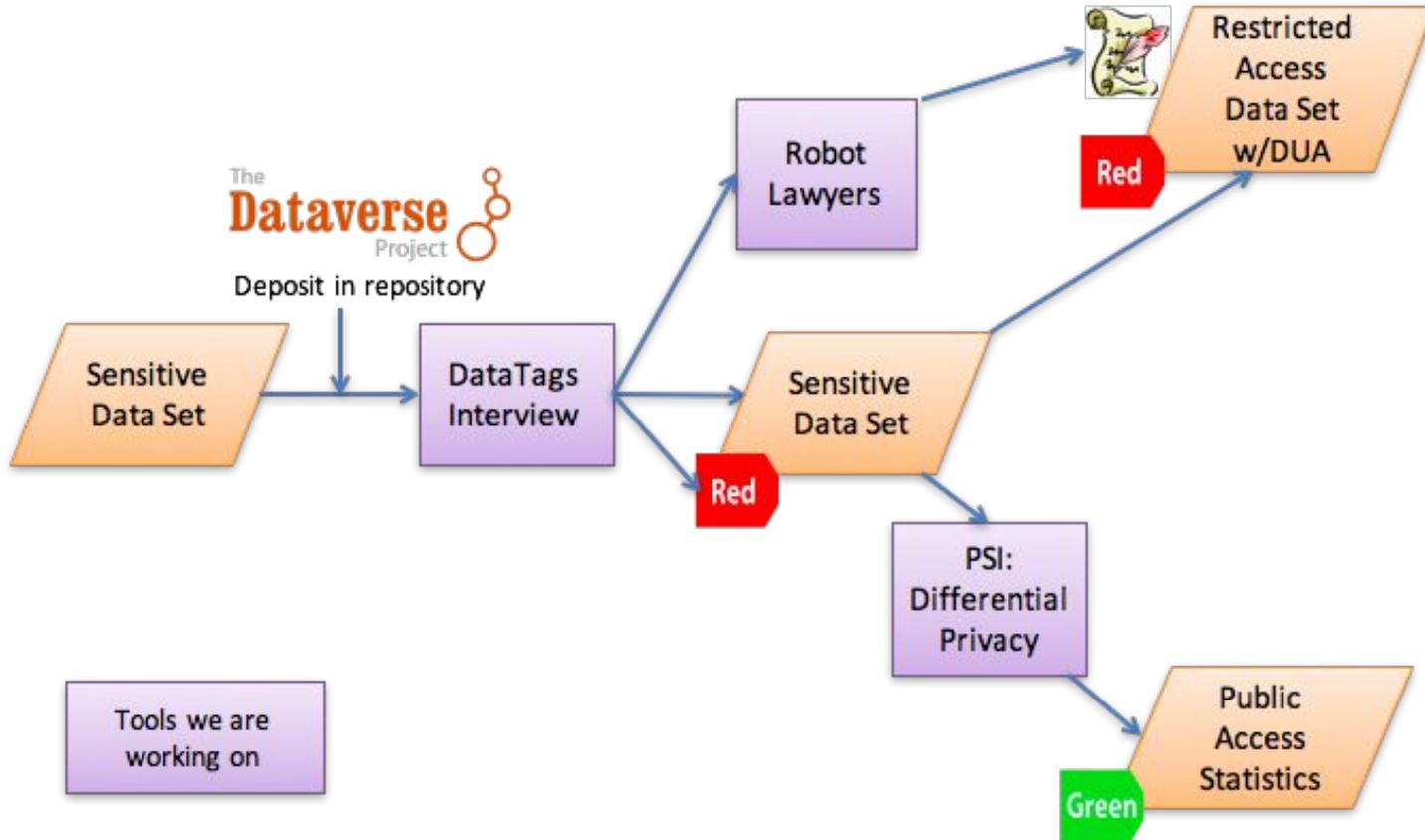
Life Threatening (e.g., domestic or gang violence)

Level of Expected Harm from Uncontrolled Use

Example Tiered Access Model



Inspired by Privacy Tools Project Model



Applications of the Framework

Policy Commentaries

- ❖ Occupational Safety & Health Administration
- ❖ White House Office of Science & Technology Policy
- ❖ Federal Trade Commission
- ❖ Department of Health & Human Services
- ❖ National Committee on Vital and Health Statistics
- ❖ National Institute of Standards & Technology
- ❖ Commission on Evidence-based Policymaking
- ❖ Future of Privacy Forum

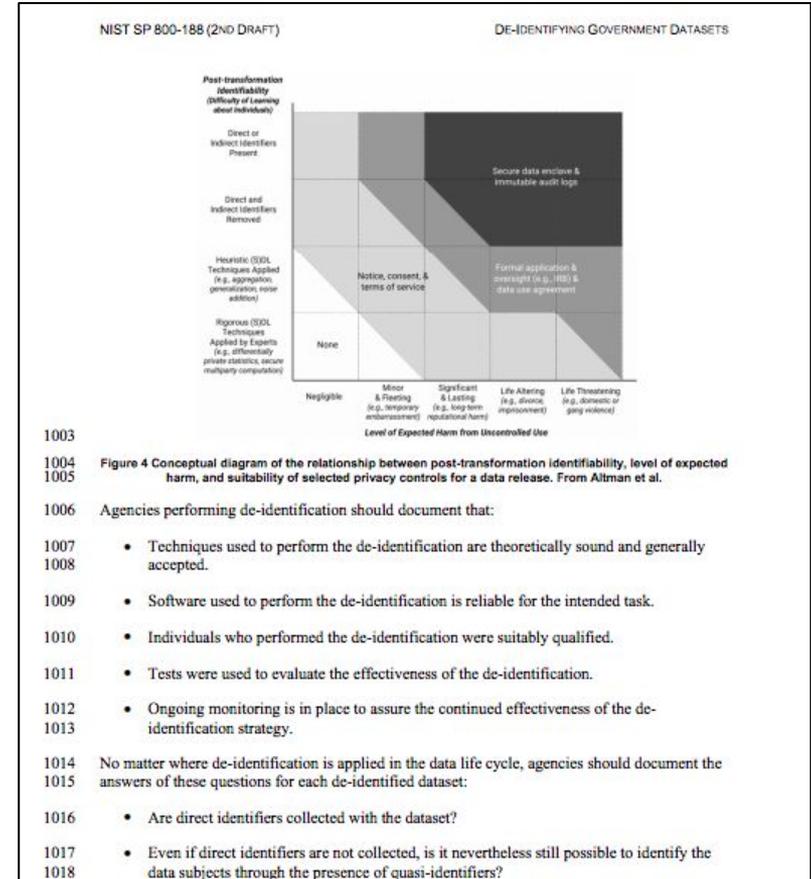
Scholarly Publications

- ❖ Open government data
- ❖ Long-term commercial and government big data activities
- ❖ Research ethics in commercial settings

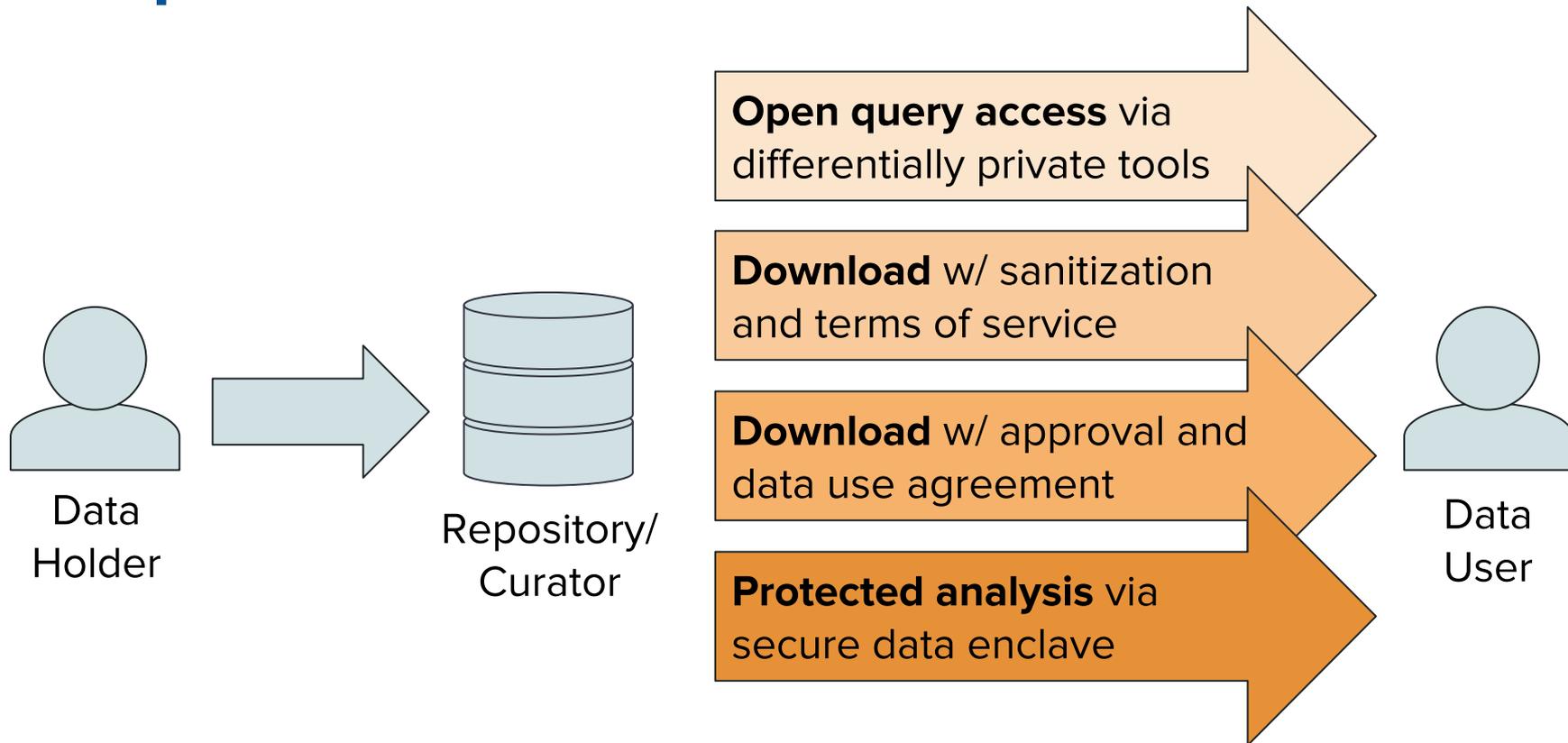
Applications of the Framework

Policy Commentaries

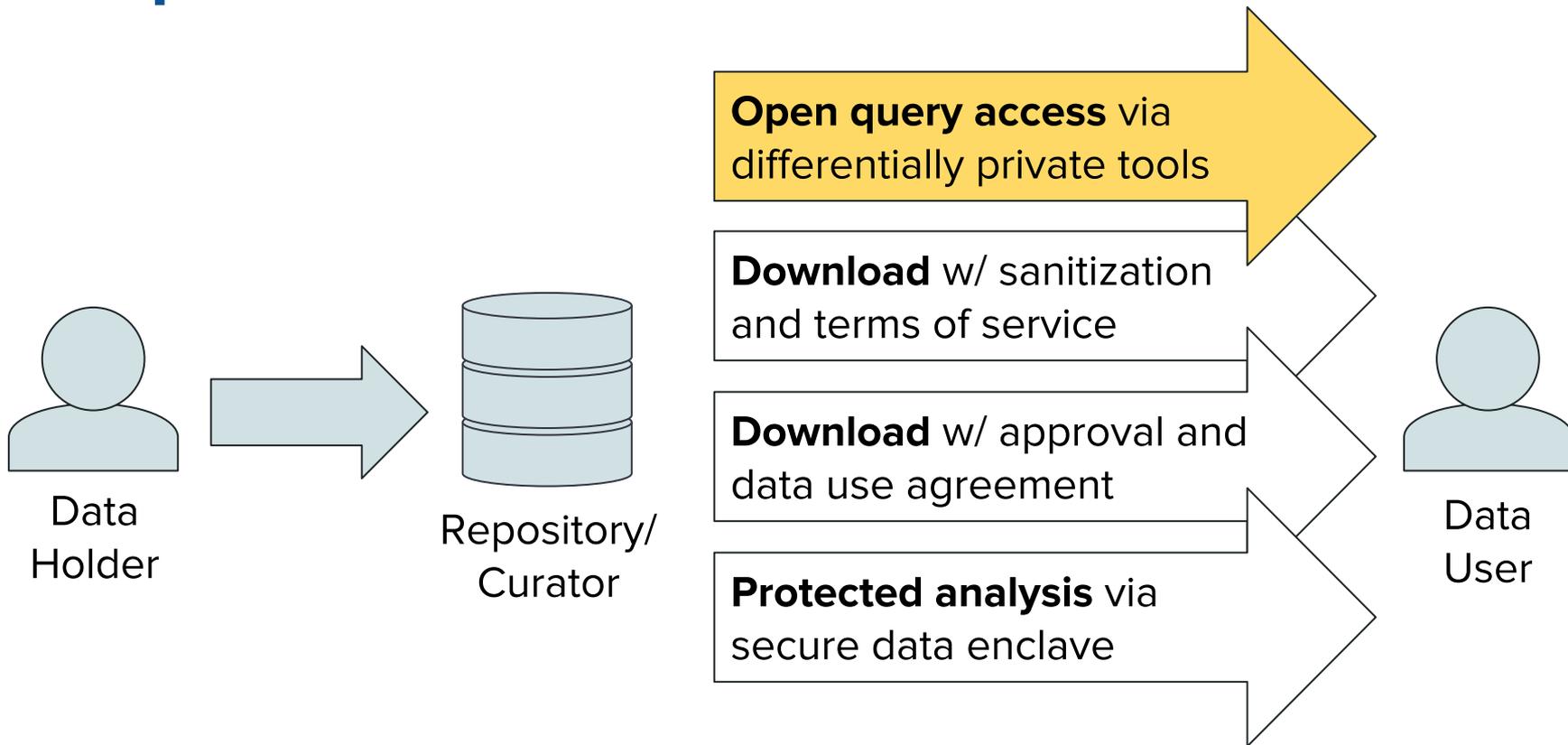
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Example Tiered Access Model



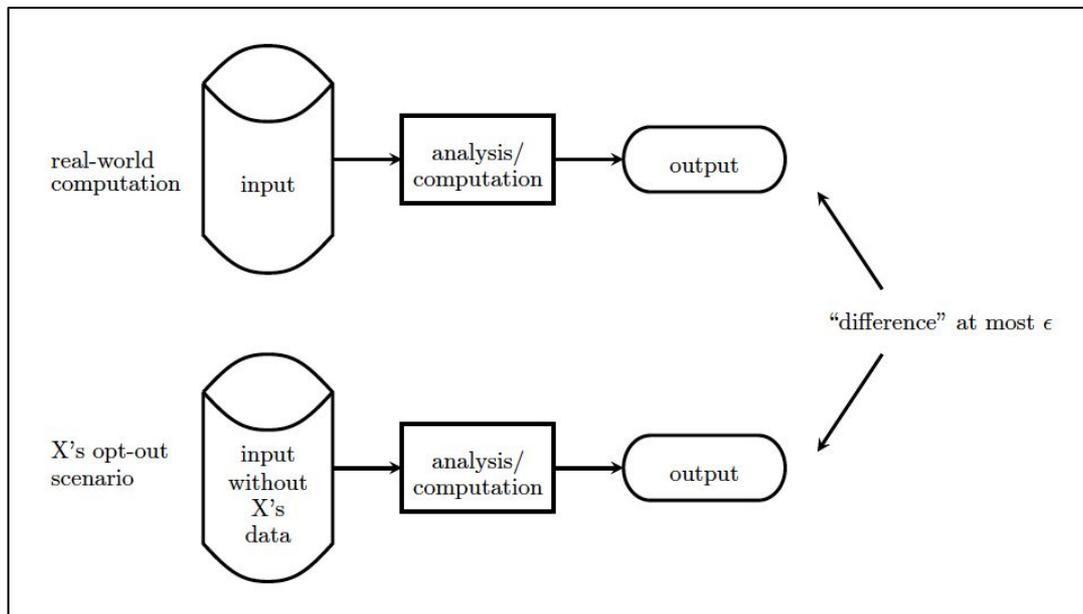
Example Tiered Access Model



Explaining Differential Privacy to Non-technical Audiences

Primer for a Non-technical Audience

Introduces differential privacy using *simplified and intuitive, but mathematically accurate, illustrations.*



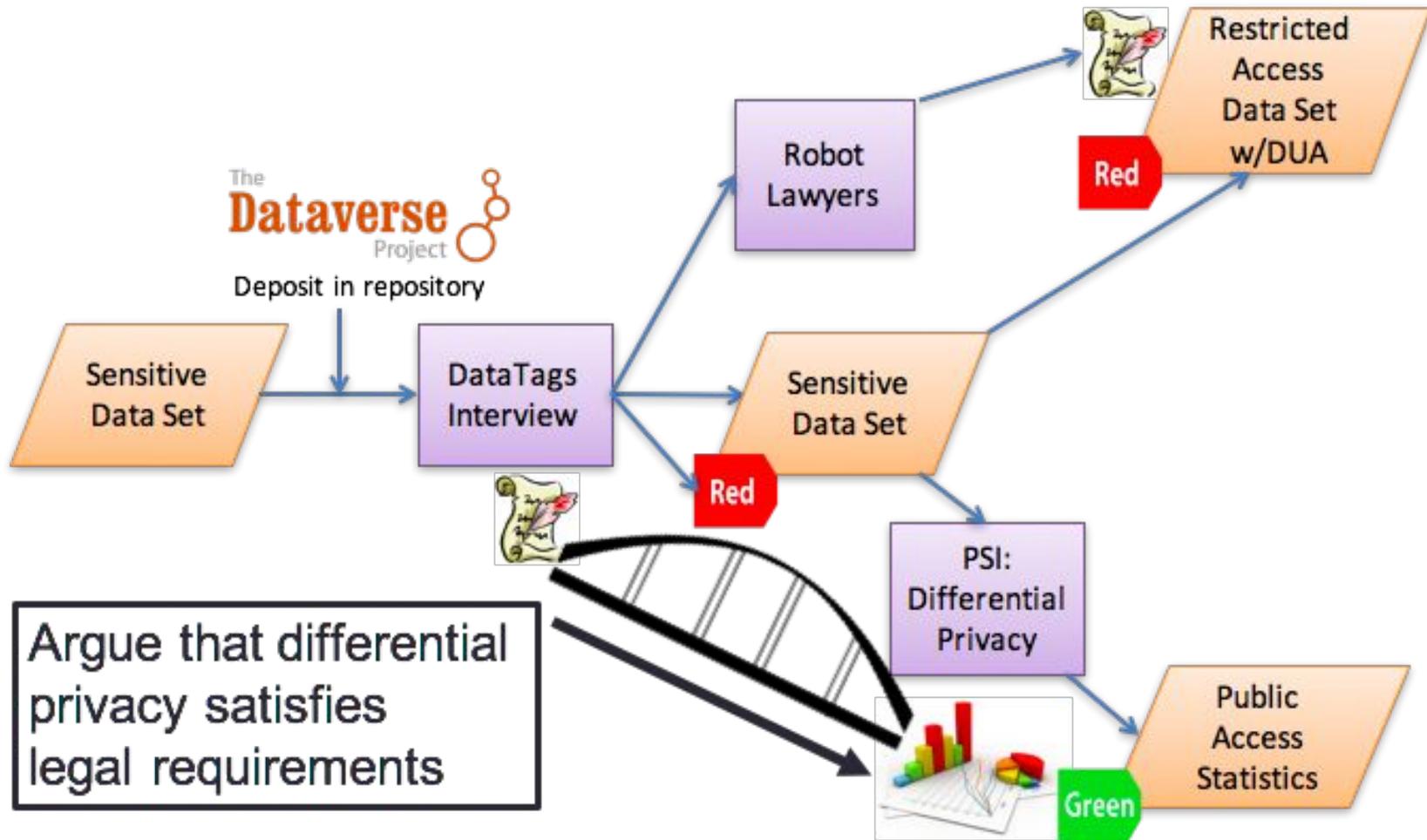
Primer for a Non-technical Audience

- ❖ Provides a foundation to guide future decisions when:
 - Analyzing and sharing statistical data about individuals,
 - Informing individuals about the privacy protection they will be afforded, and
 - Designing policies and regulations for robust privacy protection.

Bridging Gaps between Differential Privacy and Legal Standards for Privacy



Deposit in repository



Argue that differential privacy satisfies legal requirements

Challenges for Formal Privacy Models

With the emergence of new technologies based on formal privacy models, can we claim they satisfy existing regulatory requirements?

Information privacy laws create uncertainty because they are generally:

- ❖ **context-specific,**
- ❖ **subject to interpretation,**
- ❖ allow for some **degree of flexibility,** and
- ❖ rely on **traditional, often heuristic, conceptions of privacy.**

Is it possible to bridge these very different languages?



$M: X^n \rightarrow T$ satisfies ϵ -differential privacy if

$\forall x, x' \in X^n$ s.t. $dist_H(x, x') = 1 \forall S \subseteq T,$

$$\Pr_M[M(x) \in S] \leq e^\epsilon \Pr_M[M(x') \in S].$$

Approach #1: Formal Modeling

Goal: Rigorously arguing that a technological privacy solution satisfies the requirements of a particular law.

Proposed approach:

1. Extracting a formal mathematical requirement from the law.
2. Proving mathematically that a technological privacy solution satisfies the requirement derived from the law.

Illustration: Formally Modeling FERPA

We extracted a formal model of the privacy desiderata for the [Family Educational Rights & Privacy Act \(FERPA\)](#).

We used a **game-based privacy definition**:

- ❖ This provides a concise and fairly intuitive abstraction of FERPA's requirements.
- ❖ If a formal model, such as differential privacy, satisfies the definition, then we have a strong argument that it satisfies the requirements of FERPA.

Modeling FERPA: The Adversary

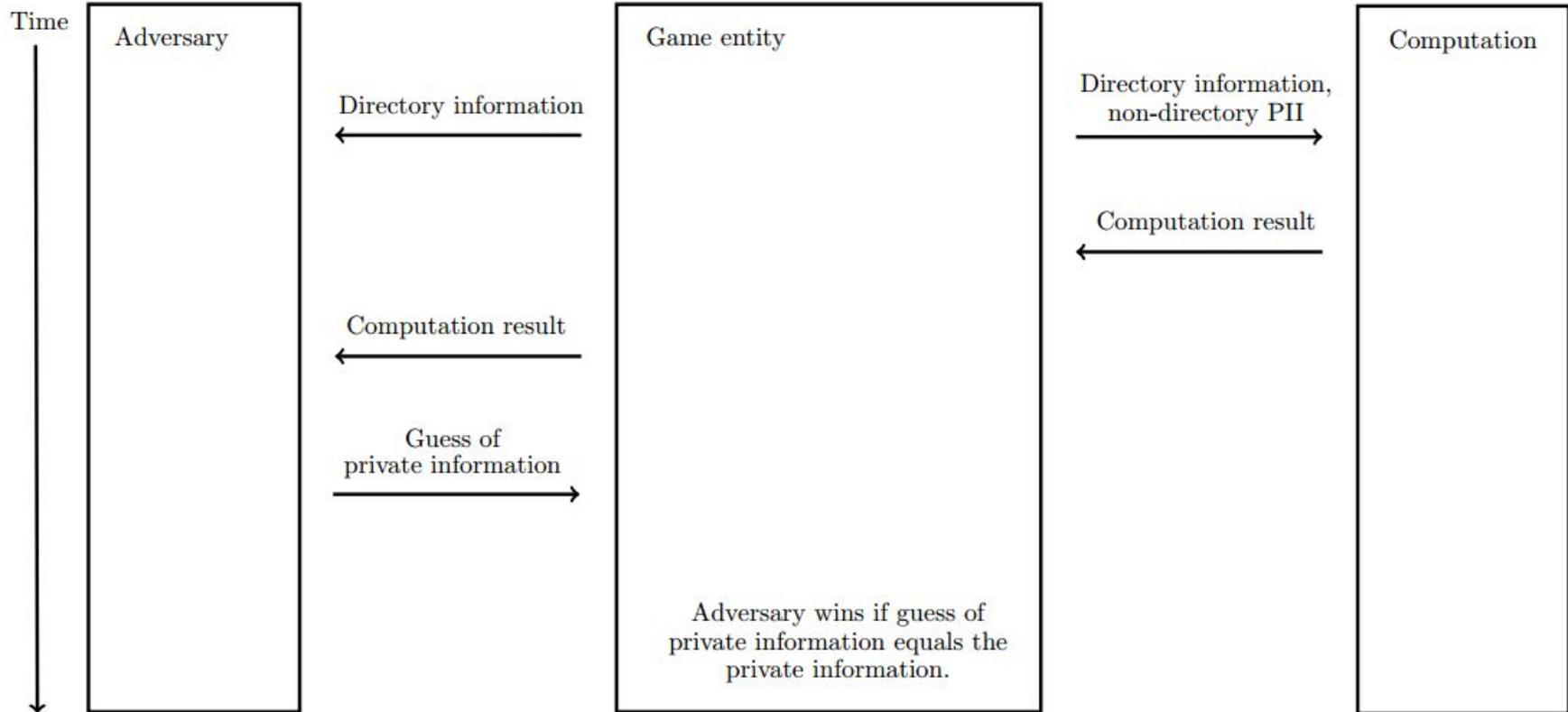
Personally identifiable information: “information that, alone or in combination, is linked or linkable to a specific student that would allow a reasonable person in the school community, who does not have personal knowledge of the relevant circumstances, to identify the student with reasonable certainty.”

This is FERPA’s **implicit adversary**.

Modeling FERPA: Directory Information

- ❖ Regulatory language interpreted conservatively, i.e., erring towards what is most beneficial for the adversary.
- ❖ E.g., the definition of **directory information** is ambiguous.
 - If we made assumptions about the definition, new interpretations could call these assumptions into question.
 - Instead, **we let the attacker to choose** what constitutes directory information.

Components of a FERPA Privacy Game



Approach #2: Interpreting the Differential Privacy Guarantee

- ❖ Legal requirements relevant to issues of privacy in computation rely on an understanding of a range of different privacy concepts.
- ❖ None of the privacy concepts that appear in the law refer directly to differential privacy.
 - However, the differential privacy guarantee can be interpreted in reference to these concepts—while accommodating differences in how these concepts are defined across contexts.

Illustration: Personally Identifiable Information

- ❖ Legal protections typically extend only to PII.
 - e.g., FERPA, HIPAA Privacy Rule, Massachusetts data security regulation, OMB memorandum
- ❖ Definitions vary significantly, but generally understood to refer to the presence of pieces of information that are linkable to the identity of an individual or to an individual's personal attributes.

PII: Interpreting Differential Privacy Guarantee

- ❖ PII does not have a precise technical meaning.
- ❖ It can be difficult to determine whether information is PII.
- ❖ Regardless of the definition of PII that is used, differential privacy can be interpreted as (essentially) ensuring that using an individual's data will not reveal any PII that is specific to her.
 - Here, **specific** is used to refer to information that cannot be inferred unless the individual's information is used in the analysis.

Lessons Learned

- ❖ There are **significant, seemingly irreconcilable, gaps** between computer science and legal approaches to privacy.
- ❖ **Progress through long-term, deep interdisciplinary collaboration:**
 - Framework for a modern privacy analysis
 - Educational materials explaining recent computer science advances
 - Novel approaches to legal-technical privacy reasoning
- ❖ **Rich area full of potential research directions:**
 - New research through cooperative agreement with the US Census Bureau
 - Opportunities to explore differential privacy and contextual integrity, approaches to setting formal privacy parameters

Selected References

Kobbi Nissim, Aaron Bembenek, Alexandra Wood, Mark Bun, Marco Gaboardi, Urs Gasser, David R. O'Brien, & Salil Vadhan, **Bridging the Gap between Computer Science and Legal Approaches to Privacy**, 31 Harvard Journal of Law & Technology — (forthcoming 2018).

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Available from <http://privacytools.seas.harvard.edu>