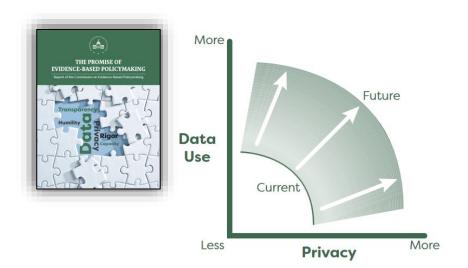
CHALLENGES TO PRIVACY AND CONFIDENTIALITY ACROSS THE BLENDED DATA LIFECYCLE

Mayank Varia **Boston University** May 22, 2023

WHY USE PRIVACY TECH?





Case study repository

Created by David Buckley, last modified on Apr 21, 2023

- 1. Boston Women's Workforce Council: Measuring salary disparity using secure multi-party computation
- 2. European Statistical System: Developing Trusted Smart Surveys
- 3. Eurostat: Processing of longitudinal mobile network operator data
- 4. Indonesia Ministry of Tourism:
 Confidentially sharing datasets between
 two mobile network operators via a

 trusted execution environment.

"Privacy, when implemented correctly, fosters more information sharing, not less."

Marc Groman

"The benefits of using data for official statistics can be realized while minimizing privacy risks to those entrusting sensitive data to National Statistics Offices."

INPUT PRIVACY

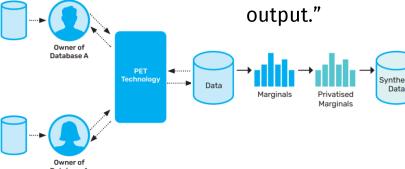
OUTPUT PRIVACY

DESIGN

Change work phases

Ongoing work phases

"Allow two or more parties to submit data into a calculation without the other respective parties seeing data in clear."



"Generally known as statistical disclosure control, [it] aims to conceal sensitive individual data from being identified or re-identified from the disseminated output."

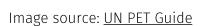
ECIFY NEEDS

Examples:

- Differential privacy
- Synthetic data generation

Selected examples:

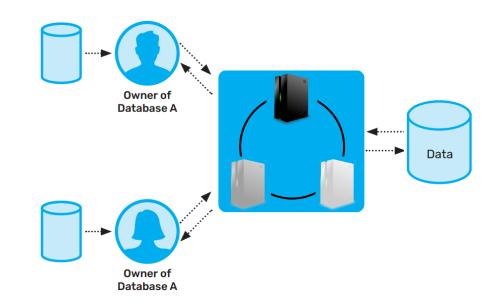
- · Secure multi-party computation
- Homomorphic encryption
- Zero knowledge proofs



SECURE MULTI-PARTY COMPUTATION (SMC)

"Enables different participating entities in possession of private sets of data to link and aggregate their data sets ... without transferring or otherwise revealing any private data to each other or anyone else."

- 2022 U.S. Senate bill S.3952



SMC DEPLOYMENTS IN THIS TALK

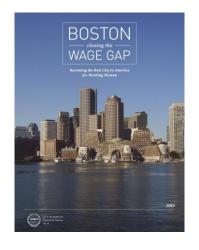
 Boston Women's Workforce Council (non-blended data)



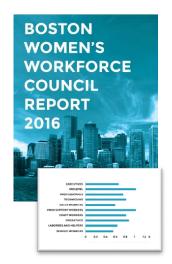
2. Massachusetts child advocacy study (blended data)



BOSTON WAGE GAP STUDY







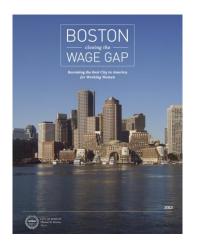






Lesson: Deploying SMC is a policy decision, promote from the top down

DATA CONTRIBUTORS



"Employers agree to anonymously contribute their wage data with the Council, which then creates a snapshot of what the wage gap looks like in our city."

100% TALENT

The Boston Women's Compact





Lesson: Easier to deploy SMC when all data contributors have similar interest in the output, and are similarly concerned about input privacy



Boston Women's Workforce Council

100% Talent Data Submission



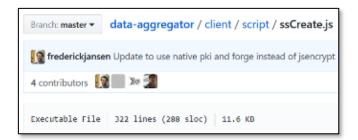


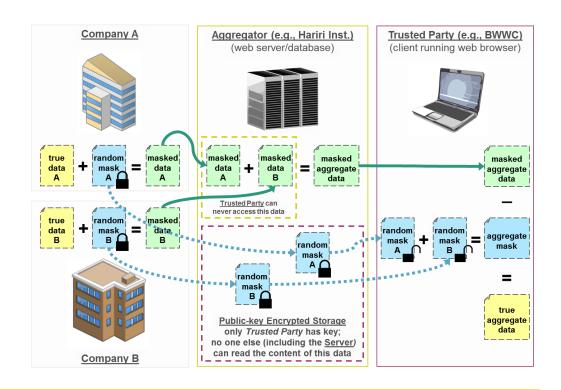
Number Of Employees

		Hispanic or Latinx		White		Black/African American		Native Hawaiian or Pacific Islander		Asian		American Indian/Alaska Native		Two or More Races (Not Hispanic or Latinx)		Unreported	
	Female	Male	Female	Male	Female	Male	Female	Male	Female	Male	Female	Male	Female	Male	Female	Male	
Executive/Senior Level																	
Officials and Managers																	
First/Mid-Level Officials																	
and Managers																	
Professionals																	
Technicians																	
Sales Workers																	
Administrative Support																	
Workers																	
Cra																	

block Lesson: Use standardized data formats & easily accessible applications

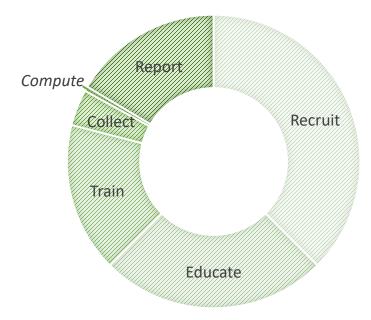
SIMPLICITY & TRANSPARENCY DRIVE ADOPTION





Lesson: SMC is simple to explain, and doing so improves trust + adoption

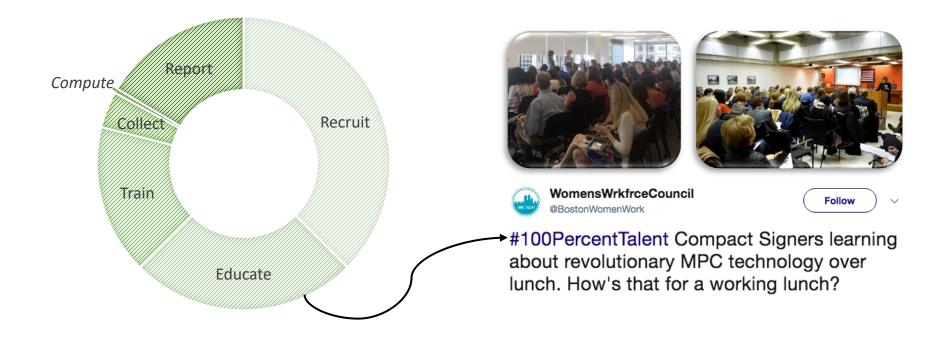
TIMELINE





Lesson: Time is spent on finding and helping people to work with data, not computer science metrics like computational efficiency

EDUCATION & TRAINING



Lesson: Identify key people to convince, together with domain experts

NDAS & LEGAL CONSIDERATIONS





Definitional question

Do encodings count as personal data?



Process question

Does computing constitute disclosure?



Liability question

Who should be blamed for an error?



Lesson: Consider the legal ramifications of data disclosure & processing

VALUE PROPOSITION

BWWCData quality

HR Personnel
Accessibility

Lawyers Liability

IT Personnel Comprehension









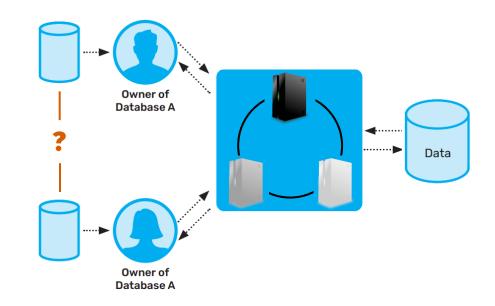


Lesson: Provide value to all stakeholders involved in the data analysis

Goal: make data-driven recommendations for reforming the juvenile justice system

Blend datasets containing data on:

- Vulnerable groups
- Educational success
- Judicial outcomes
- ...



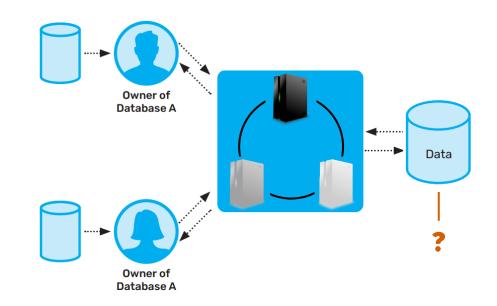


Lesson: Challenging to conduct any data analysis (with or without input privacy) in the absence of standardized data formats and common fields

Goal: make data-driven recommendations for reforming the juvenile justice system

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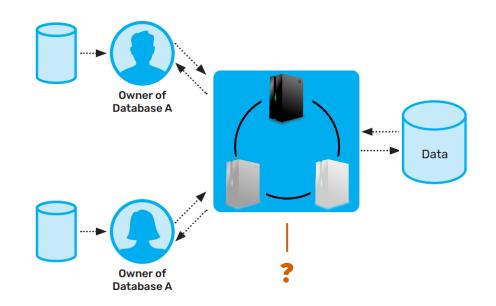


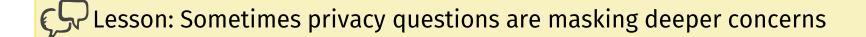
Lesson: Challenging to build consensus for data analyses when each potential contributor has its own policy goals and interest in the output

Goal: make data-driven recommendations for reforming the juvenile justice system

Blend datasets containing data on:

- Vulnerable groups
- Educational success
- Judicial outcomes
- ...

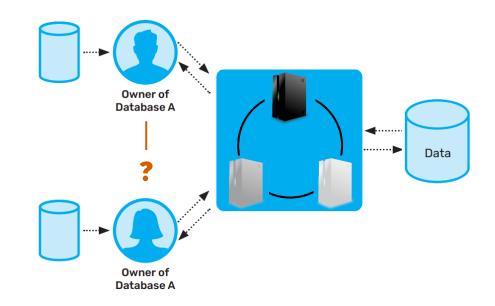




Goal: make data-driven recommendations for reforming the juvenile justice system

Blend datasets containing data on:

- Vulnerable groups
- **Educational success**
- Judicial outcomes





Lesson: Ultimately, can only move forward with a data analysis if the result provides value to all stakeholders involved

SUMMARY: LESSONS LEARNED

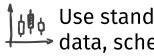
WHO

WHAT

HOW



Promote from the top down



Use standardized Provide value to all data, schemas, apps



Find orgs with interest in the result



Identify key people to convince



Simple, transparent tech drives adoption



People, not CPU, dominate runtime



Consider legal ramifications



Improve both data privacy and utility