# Bocconi

# A (BRIEF) OVERVIEW ON PRIVACY ENHANCING TECHNOLOGIES (PETs)

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## The current paradox of data economy

- The exponential increase in the processing of personal data has created a wide array of unprecedented possibilities to gain useful insights via artificial intelligence and machine learning; at the same time, these developments expose individuals to new privacy threats.
- Privacy **vs** usability

## Privacy enhancing technologies

PETs want to bridge the gap of data protection on one side and the value of data on the other, also called the **privacy-utility tradeoff**.

## Legislator and public authorities on PETs

The US Senate in "<u>Promoting Digital Privacy Technologies Act</u>"

- Proposes the definition of PETs as "any software solution, technical processes, or other technological means of enhancing the privacy and confidentiality of an individual's personal data in data or sets of data".
- The Act is planning to provide support for research, broader deployment and standardization



## PETs categories

| Types of PETs                       | Key technologies   | Current and potential<br>applications*  | Challenges and limitations   |  |
|-------------------------------------|--|---|--|--|
| Data obfuscation tools              | Anonymisation /<br>Pseudonymisation                                  | Secure storage  | - Ensuring that information does not leak (risk of re-identification)  |  |
|                                     | Synthetic data   | Privacy-preserving machine<br>learning  | - Amplified bias in particular for<br>synthetic data   |  |
|                                     | Differential privacy   | Expanding research<br>opportunities   | <ul> <li>Insufficient skills and<br/>competences</li> </ul>  |  |
|                                     | Zero-knowledge proofs  | Verifying information without<br>requiring disclosure<br>(e.g. age verification)  | - Applications are still in their<br>early stages  |  |
| Encrypted data                      | Homomorphic encryption   | Computing on encrypted data   | - Data cleaning challenges   |  |
| processing tools                    | Multi-party computation  | within the same organisation  | - Ensuring that information does   |  |
|                                     | (including orivate set intersection)                                 | Computing on private data that<br>is too sensitive to disclose<br>Contact tracing / discovery                                     | not leak - Higher computation costs  |  |
|                                     |  |   |  |  |
|                                     | Trusted execution environments                                       | Computing using models that need to remain private  | <ul> <li>Higher computation costs</li> <li>Digital security challenges</li> </ul>  |  |
| Federated and distributed analytics | Federated learning   | Privacy-preserving machine  | - Reliable connectivity needed   |  |
|                                     | Distributed analytics  | learning  | <ul> <li>Information on data models<br/>need to be made available to<br/>data processor</li> </ul>   |  |
| Data<br>accountability<br>tools     | Accountable systems  | Setting and enforcing rules<br>regarding when data can be<br>accessed<br>Immutable tracking of data<br>access by data controllers | <ul> <li>Narrow use cases and lack<br/>stand-alone applications</li> <li>Configuration complexity</li> <li>Privacy and data protection<br/>compliance risks where</li> </ul> |  |
|                                     | Threshold secret sharing   |   | distributed ledger technologies  |  |
|                                     | Personal data stores / Personal<br>Information Management<br>Systems | Providing data subjects control over their own data   | <ul> <li>are used</li> <li>Digital security challenges</li> <li>Not considered as PETs in the strict sense</li> </ul>  |  |

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## (Fully) homomorphic encryption (FHE)



## Secure multiparty computation (MPC)



## Secure multiparty computation



## Secure multiparty computation vs FHE





MPC

FHE

## Is this even possible?

## FHE



"I don't think we'll see anyone using Gentry's solution in our lifetimes."

### \* HE is getting faster 8 times every year

e.g. Bootstrapping time: the most time-consuming operation in HE



- Still slow in computation
- Relatively cheap in communication
- Only possible (currently) for simple functions

### MPC



- MPC is a very mature technology
- Several general purpose protocols
- It allows collaborative computing, hence it requires communication between parties
- Generally cheap in computation

## Zero-knowledge proofs



- ZKPs enable a party to demonstrate a statement to other parties without revealing any information beyond the proof.
- ZK Proofs facilitate the development of an entirely new range of applications, e. g. blockchain, "proof of humanity", national security purpose

## Zero-knowledge proofs – The SIEVE project

(Securing information for encrypted verification and evaluation)



Defense Advanced Research Projects Agency: ``Recent research has substantially increased the efficiency of ZK proofs, enabling real-world use, primarily by cryptocurrencies. While useful for cryptocurrencies, the ZK proofs created are specialized for **this task and do not necessarily scale for transactions that are more complex.** For highly complex proof statements like those that the Department of Defense (DoD) may wish to employ, novel and more efficient approaches are needed.''

Prove whether or not vulnerabilities exist in software without revealing how they are caused;
 Assess properties of cyberspace operations capabilities without accessing source code;
 Assure anti-tamper cyber-security techniques without revealing code details.

## ZK proofs and post-quantum signature schemes

FAEST: new post-quantum signature scheme

- Only based on symmetric primitives (AES-SHAKE)
- Small keys
- Faster in signing and shorter than SPHINCS+
- Based on MPC tools

It will be submitted to the next NIST PQC standardization call on the 1<sup>st</sup> June.

# **Case study report -** UN Committee on Big Data and Data Science for Official Statistics

Boston Women's Workforce Council: Measuring salary disparity using secure multi-party computation



| Purpose                        | To measure the gender and racial wage gaps throughout the greater Boston area every 1-2 years  |
|--------------------------------|--|
| Datasets                       | Real demographic and payroll data from companies<br>and non-profit organisations, large and small,<br>throughout the greater Boston area   |
| PETs used                      | Secure Multi Party Computation   |
| Application                    | Secure vector addition   |
| Details of computation         | Organisations contribute a spreadsheet containing<br>more than 600 cells of data. The Boston Women's<br>Workforce Council receives the summation of each<br>cell across all participating organisations.   |
| Parties and trust relationship | More than 100 participating organisations act as input<br>parties; the Boston Women's Workforce Council<br>serves in a compute and output party role; Boston<br>University serves as a compute party. Participants<br>trust BU and the BWWC to behave semi-honestly, with<br>the ability to audit and verify code. |
| Implementation status          | Production   |
| Resources                      | Boston Women's Workforce Council reports<br>Data submission website<br>Open-source code repository on GitHub<br>Publications about the PET used in this project appear<br>at SOUPS 2019, COMPASS<br>2018, SecDev 2016, and the Communications of the<br>ACM.   |

# **Case study report -** UN Committee on Big Data and Data Science for Official Statistics

Italian National Institute of Statistics and Bank of Italy: Enriching data analysis using privacy-preserving record linkage

| Purpose                        | To enable enriched socio-economic analysis by<br>augmenting data held by Bank of Italy with data<br>held by ISTAT (and vice versa).  |
|--------------------------------|--|
| Datasets                       | Socio-demographic and financial datasets<br>(linkable via tax code common key)   |
| PETs used                      | Secure Multi Party Computation   |
| Application                    | Private Set Intersection with Analytics  |
| Details of computation         | ISTAT and Bank of Italy perform an Exact PSI<br>using the shared tax code key. The intersection<br>is encrypted and transferred to the third<br>"linker" party. ISTAT and Bank of Italy submit<br>queries to the linker, which can perform<br>aggregation and counts against the data on-<br>demand, with outputs transmitted to ISTAT and<br>Bank of Italy. |
| Parties and trust relationship | Bank of Italy and ISTAT act as input, compute,<br>and output parties; third "linker" party serves in<br>a compute role. Organisations trust each other<br>("honest but curious" threat model).   |
| Implementation status          | Pilot  |
| Resources                      |  |

#### ISTAT

| HOUSEHOLD        |                       |                 |   |
|------------------|-----------------------|-----------------|---|
| A <sub>iD</sub>  | Number of<br>children | Age<br>class    |   |
| a <sub>id1</sub> | a <sub>11</sub>       | a <sub>12</sub> | - |
| a <sub>id2</sub> | a <sub>21</sub>       | a <sub>22</sub> |   |

### BANK OF ITALY

| B <sub>ID</sub>             | Income<br>Class        | Type of<br>Mortgage<br>Payment | Payer<br>type          |
|-----------------------------|------------------------|--------------------------------|------------------------|
| b <sub>id1</sub>            | b <sub>11</sub>        | <b>b</b> <sub>12</sub>         | <b>b</b> <sub>13</sub> |
| <br><b>b</b> <sub>id2</sub> | b <sub>21</sub>        | b <sub>22</sub>                | b <sub>23</sub>        |
| <br>b <sub>id3</sub>        | <b>b</b> <sub>31</sub> | b <sub>32</sub>                | b <sub>33</sub>        |

MORTGAGE



| $\mathbf{C}_{\mathrm{ID}} = \mathbf{A}_{\mathrm{ID}} = \mathbf{B}_{\mathrm{ID}}$ | Number of children | Age class       | Income class           | Payer Type             |
|--|--------------------|-----------------|------------------------|------------------------|
| C <sub>id1</sub>   | a <sub>11</sub>    | a <sub>12</sub> | <b>b</b> <sub>21</sub> | <b>b</b> <sub>23</sub> |
| C <sub>id2</sub>   | a <sub>21</sub>    | a <sub>22</sub> | <b>b</b> <sub>31</sub> | <b>b</b> <sub>33</sub> |

## **Future directions**



All the stages in this line are still very active!!!

## PETs maturity, opportunities and challenges

- PETs are at different stages of development and will likely need to be part of data governance frameworks to ensure they are used properly. Many of these tools are still in their infancy and limited to specific data processing use cases.
- Given their innovative nature and high potential, PETs warrant a comprehensive re-evaluation of the application of regulations on data collection and processing
- Policy makers will increasingly need to consider how the use of PETs may impact regulatory assessments under national privacy and data protection frameworks, taking into account the contribution of PETs to privacy protective outcomes.
- As PETs mature, there will be an increasing need for awareness raising and training to better design, build, implement, use and audit these new technologies.
- Stronger cross-border and cross-sectoral regulatory co-operation will be needed to better consider technological developments on PETs for privacy and data protection.

