



About me – Jiří Novák

- —From Prague, Czech Republic,
- —PhD in Statistics
- —Research topics
 - —Data privacy and data anonymization
 - —Statistical disclosure control



- —Responsible for data confidentiality of Czech Census 2021
- —Currently working at FHNW on a SNF project: *Harnessing event and longitudinal data* in industry and health sector through privacy preserving technologies
- —PhD student of Carolin Strobl with topic: *Data anonymization of longitudinal psychological data*

Data anonymization

My work is Data Anonymization in the context of the field of Statistical Disclosure Control (SDC)

Statistical Disclosure Control seeks to protect statistical data in such a way that they can be released without giving away confidential information that could be linked to specific individuals or entities.

Importance of Data Anonymization

-1. Principle

Statistical records of individual persons, businesses, or events used to produce Statistics are strictly confidential.

-2. Legal

Laws mandate protection of business and personal data, regulating publication of private information.

-3. Quality

Trust in confidentiality ensures respondents provide accurate information.

-4. Ethical

Disclosing information that can be linked to specific individuals or entities is unethical.

Relevance to Open Science

Open Science, Open Access, Open Data are important trends in the scientific community.

Research data that results from publicly funded research should be FAIR:

Findable, Accessible, Interoperable, Reusable

- therefore replicable, transparent, trustworthy, verifiable and accountable
- Principle: As open as possible, as closed as necessary

Commission Recommendation (EU) 2018/790 on access to and preservation of scientific information

University of Zurich

Key Concepts

Disclosure

- Disclosure occurs when new information is revealed via released data.
- (1) **Identity disclosure**: Revealing the identity of an individual.
- (2) **Attribute disclosure**: Revealing sensitive attributes of an individual.
- (3) Inferential disclosure: Making inferences about an individual based on the released data.
- (4) **Membership disclosure**: Determining whether a specific individual is included in a dataset

Re-identification risk

—Risk that an intruder can link a record in the released data to a specific individual in the population.

Data utility

— Usefulness of the data for the intended purpose.

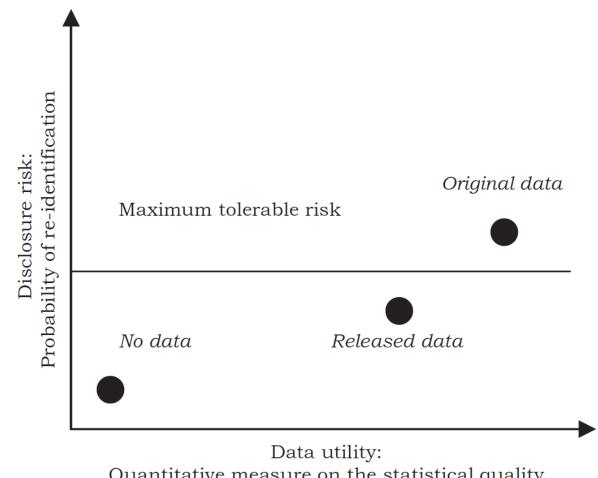
Risk vs. Utility

The goal is to find a balance between risk and utility, so there is a **risk-utility trade-off**.

— **Risk**: the probability of a disclosure event occurring.



— **Utility**: the usefulness of the data for the intended purpose.



Quantitative measure on the statistical quality

R-U confidentiality map (Duncan et al.,2001)

Disclosure control methods

1. Masking original data

a) Non-perturbative masking

Generalize data to hide identities without changing actual values.

b) Perturbative masking

Add noise or alter data values.

2. Generating synthetic data

a) Parametric methods

Statistical models based on the data's assumed distribution

b) Non-parametric methods

Techniques that do not assume an underlying distribution

c) Generative Adversarial Networks (GANs)

Machine learning models that generate synthetic data by training two neural networks in tandem.

Mostly AI

MOSTLY•AI

https://mostly.ai/

MOSTLY AI's synthesizer uses a proprietary AI-driven approach that draws on deep learning principles similar to those found in generative adversarial networks (GANs).

You can use point-and-click environment, or python code.

Libraries used

a) os and sys

Set the working directory to the folder

a) Matplotlib

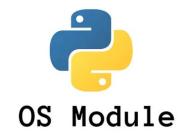
For visualization

b) Pandas

For data manipulation

c) MostlyAI

Synthetize data with Mostly AI



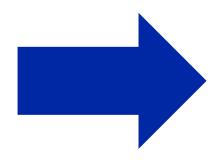






MOSTLY AI

Challenges and error



Open word document Problems.docx

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Thank you for the attention



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