

Python Data Collection and Management for Public Policy Research

Day 7: Intro to Data, Data Ethics

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Agenda for Today

- What is Data?
 - Data Formats
 - Data Types
 - Data Storage
- Ethics of Computational Social Science

What is Data?

What is data?

- A representation of the world
- Data necessarily involve subjective decisions on how to:
 - How/what we decide to measure
 - How/what to sample.
 - How/what to share.
 - How to store.
- Storing data will necessarily involve some information loss.

- **Tabular data:** data in rectangular form, with rows and columns.
- **Time series data:** observations indexed in time order, often used in event analysis, and trend analysis.
- **Graph data:** Data representing relationships between entities (social networks, and citation networks)
- **Hierarchical data:** Data organized in a tree-like structure.



Common file formats for storing tabular data:

- Comma- or tab-separated values (.csv, .tsv)
 - Each line is an observation
 - Variables are separated by a comma or tab
 - Free, wide support
- Proprietary data formats (.dta, .xlsx, etc.)
 - Difficult to read without closed-source software (Stata, Microsoft Excel).
 - Want to avoid to facilitate replication of our research!

Relational databases (e.g., SQL)

- Data organized into tables (e.g., author, article, newspaper, etc. for a database of newspaper articles)
- Tables are related through "keys" (e.g., articles written by the same author will have a numeric key indicating the author record in the "author" table)
- Allows for fast retrieval of data from large datasets

Building on Data Storage Types

- Learn SQL ([watch here](#) )
- Use SQLite in Python ([watch here](#) )

Ethics of Computational Social Science (CSS)

Why are we learning about ethics in a Python course?

- Computational methods are powerful tools, misusing these tools can harm people.
- Many harms from misuse of data are unintentional, awareness is key!
- Goals and agendas of funding agencies, corporations do not align with **research stakeholders**.

Ethical Challenges Faced by CSS

Challenges faced by CSS [Leslie, 2023]:

- Treatment of research subjects
- Impacts of CSS research on affected individuals and communities
- Quality of CSS research and to its epistemological status
- Research integrity
- Research equity

Treatment of Research Subjects

We should aim to treat research subjects with:

1. Respect

- **Expectation of Privacy:** Subjects may not expect their data will be used for research.
- **Personal Autonomy:** Subjects may not want to be a part of our research.

2. Justice

- Risks and benefits of research participation should not be **unevenly distributed** across groups (age, race, disability, gender, sexual orientation, etc.).

3. Care

- **Risks** to subjects should be minimized.
- **Benefits** to subjects should be maximized.

Treatment of Research Subjects: Anonymization

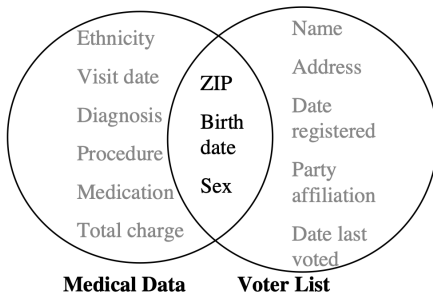


Figure 1 Linking to re-identify data

Source: [Sweeney, 2002]

- To protect subjects, it is good practice to remove **personally identifiable information (PII)**
- But is this enough? There is still a risk of **re-identification** through **data linkage**.

Treatment of Research Subjects: Harm Minimization Strategies

- Obscure/remove personally identifiable information (PII).
- Aggregate data to less specific units.
- Apply differential privacy: introduce noise to protect data while retaining usefulness.
- Obtain informed consent:
 1. Explain study's purpose and procedures.
 2. Encourage questions; ensure clear responses.
 3. Secure explicit consent, freely given.
 4. Confirm participants' right to withdraw anytime.

Impacts on Individuals and Communities

Consider individuals and communities as **stakeholders** in research:

- Think of who is likely to benefit from research (stakeholders). What do they care about?
- **Interests and values** of subjects/stakeholders are often not considered by researchers.
- Do funding agencies/corporations and subjects/stakeholders have **mismatched agendas**?
- Just allocation of **risks and benefits** of research.

- Challenges with **algorithmic influences** on data collection.
 - Companies like Meta, Google and ByteDance use algorithms to target delivery of content for engagement.
 - How does this affect the authenticity of social phenomena captured?
- The **illusion of data veracity** due to large volume of data.
 - Misconception that large data sets are inherently representative or accurate.
 - Overreliance on big data can obscure the need for robust methodological rigor and validation.

- Asymmetrical resources and influence:
 - Disparities in access to data and computational resources can skew research outcomes.
 - Potential for dominant stakeholders to dictate research agendas and priorities.
- Dependence on corporations for resources and data.
 - Conflicts of interest may arise when corporate interests drive research directions.
 - Ethical dilemmas in maintaining research independence and integrity.

Research Equity

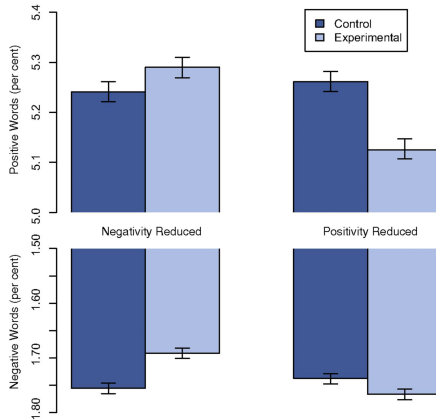
- Reinforcement of **digital divides** and **data inequities**.
 - Research driven by data predominantly collected from more privileged or accessible groups.
 - Potential to overlook marginalized populations, thus perpetuating inequality.
- **Aggregation biases** mask subgroup differences.
 - Generalized findings can obscure significant variations and perpetuate stereotypes.
 - Risk of policy and interventions failing to address or even exacerbating subgroup vulnerabilities.
- **Global inequalities** affect data sharing and collaboration.
 - Power imbalances between high- and low-resource settings can lead to exploitative data practices.
 - Inequitable distribution of research benefits and burdens across global divides.

Example 1: Emotional Contagion

Experimental evidence of massive-scale emotional contagion through social networks

Adam D. I. Kramer^{a,1}, Jamie E. Guillory^{b,2}, and Jeffrey T. Hancock^{b,c}

^aCore Data Science Team, Facebook, Inc., Menlo Park, CA 94025; and Departments of ^bCommunication and ^cInformation Science, Cornell University, Ithaca, NY 14853



Source: [Kramer et al., 2014]

Summary and Ethical Concerns: Emotional Contagion

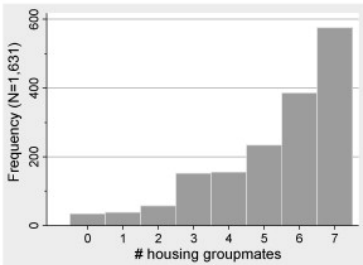
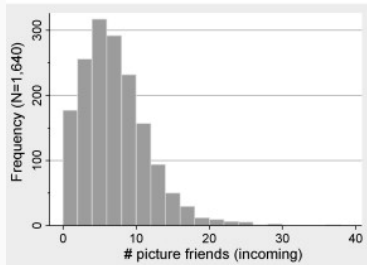
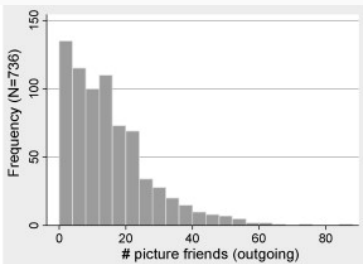
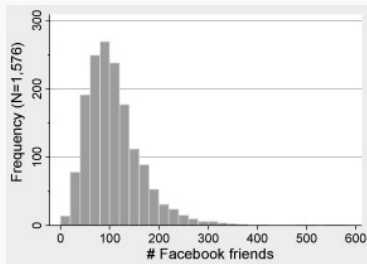
- **Study Overview:**

- Researchers manipulated the news feeds of Facebook users to test if emotional states can be transferred to others via emotional contagion.
- Participants were not informed they were part of an experiment.

- **Ethical Issues:**

- Lack of informed consent.
- Psychological manipulation without users' knowledge.
- Potential emotional harm to participants.

Example 2: Tastes, Ties, and Time



Source: [Lewis et al., 2008]

Summary and Ethical Concerns: Tastes, Ties, and Time

- **Study Overview:**

- Used Facebook data to analyze the relationship between online behavior and offline social networks.

- **Ethical Issues:**

- Privacy and data security.
- Consent process and the extent to which participants were aware of the data usage.
- Potential risks to participants' privacy given the sensitive nature of social network data.

Example 3: Montana Mail Study

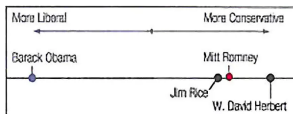


2014 Montana General Election Voter Information Guide

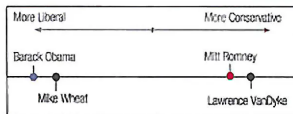
Election Date: November 4, 2014



Nonpartisan Supreme Court Justice #1 Race



Nonpartisan Supreme Court Justice #2 Race



For more information on how these figures were created, please see <http://data.stanford.edu/time>. Please note that this guide is non-partisan and does not endorse any candidate or party. This guide was created as part of a joint research project at Stanford and Dartmouth.

Paid for by researchers at Stanford University and Dartmouth College, 616 Serra Street, Stanford, CA 94305

Take this to the polls!

Source: [New York Times](#)

Summary and Ethical Concerns: Montana Mail Study

- **Study Overview:**

- Researchers sent political mailers to Montana voters, resembling official state election guides, to study political behavior.

- **Ethical Issues:**

- Deception and misrepresentation.
- Interference in a real election process without proper oversight.
- Potential to influence voter behavior and outcomes.

Example 4: Tweetment Effects on the Tweeted

Tweetment Effects on the Tweeted: Experimentally Reducing Racist Harassment

Kevin Munger¹

(a)



Rasheed

@Rasheed

@ [redacted] Hey man, just remember that there are real people who are hurt when you harass them with that kind of language

Source: [Munger, 2017]

Summary and Ethical Concerns: Tweetment Effects on the Tweeted

- **Study Overview:**

- Examined the impact of automated counter-speech on racist Twitter users by sending messages from bot accounts.

- **Ethical Issues (?):**

- Deception and manipulation.
- Psychological impacts on participants.
- Consent of targeted users.
- But does the positive normative effect matter?



Kramer, A. D., Guillory, J. E., and Hancock, J. T. (2014).

Experimental evidence of massive-scale emotional contagion through social networks.

Proceedings of the National academy of Sciences of the United States of America, 111(24):8788.



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Social networks, 30(4):330–342.



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Sweeney, L. (2002).

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International journal of uncertainty, fuzziness and knowledge-based systems, 10(05):557–570.