DataTags and OpenDP
Sharing Sensitive Data

Workshop of the OECD's Committee for Scientific and Technological Policy

Revision of Recommendation concerning access to research data from public funding

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DataTags

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Dataverse Team

A **datatag** is a set of security features and access requirements for file handling.

A **datatags repository** is one that stores and shares data files in accordance with a standardized and ordered levels of security and access requirements

A DataTags repository must:

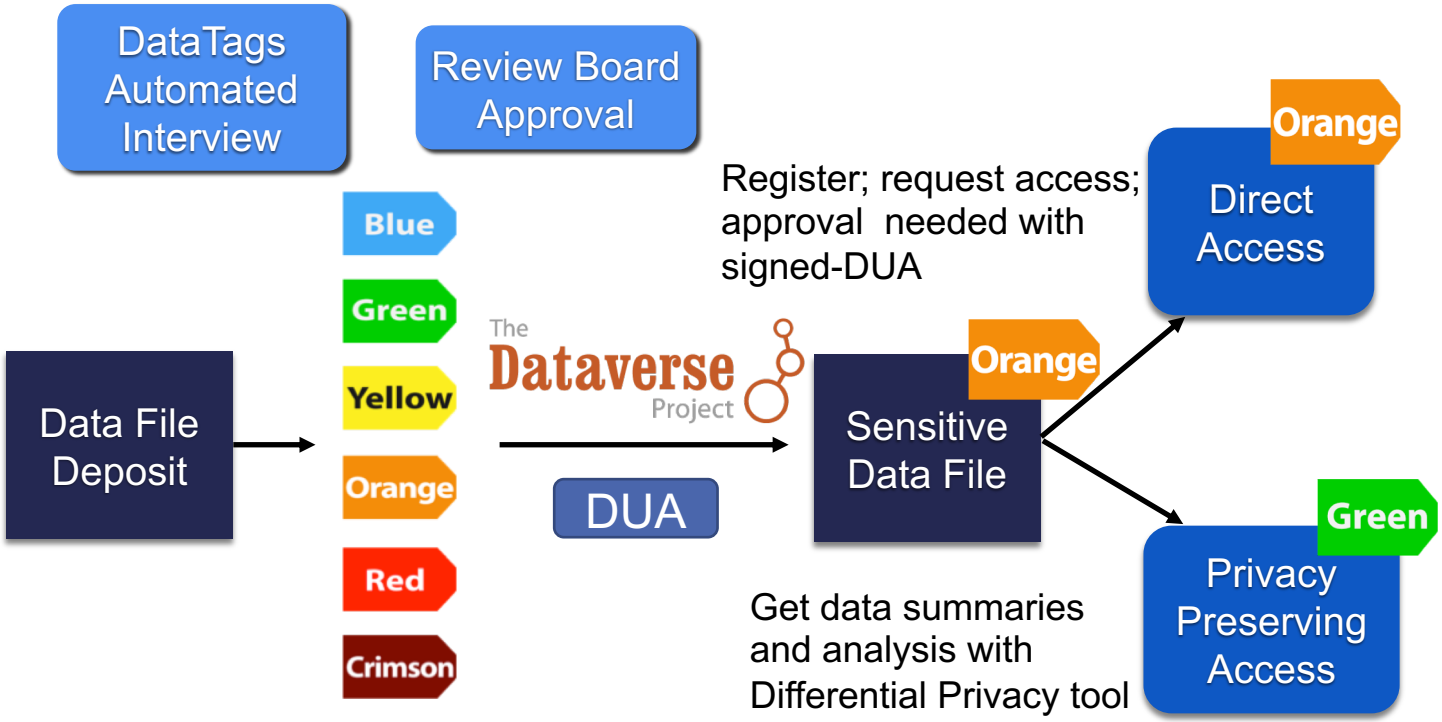
1. Support more than one **datatag**
2. Each file in the repository must have one and only one **datatag**
    a. additional requirements cannot weaken the file security
    b. and cannot required the same or more security than a more restrictive datatag
3. A recipient of a file from the **repository** must:
    a. satisfy file's access requirements,
    b. produce sufficient credentials as requested,
    c. and agree to any terms of use required to acquire the file.
4. Provide **technological guarantees** for requirements 1, 2 and 3.

Standardized Access and Security Levels

DataTag	Access	Authorization	Data Use Agreement	Encryption
Blue	Public			
Green	Public	+ Register		
Yellow	Restricted	+ Approval Needed	+ Click-thru DUA	+ Encrypted transit
Orange	Restricted	+ Approval Needed	+ Signed DUA	+ Encrypted transit + Encrypted storage
Red	Restricted	+ Approval Needed	+ Signed DUA + Two-factor Auth	+ Encrypted transit + Encrypted storage
Crimson	Restricted	+ Approval Needed	+ Signed DUA + Two-factor Auth	+ Encrypted transit + Multi-encrypted storage

Sweeney, Crosas, Bar-Sinai, 2015. Sharing Sensitive Data with Confidence: The DataTags System, Technology Science

DataTags, Dataverse, Privacy-Preserving Tools



OpenDP

Salil Vadhan, James Honaker, Gary King, Mercè Crosas

Harvard Privacy Tools team

OpenDP: A New Project for Sensitive Data

A **community effort** to build a trustworthy and **open-source** suite of **differential privacy** tools that can be **easily adopted** by custodians of sensitive data to make it available for statistical research.

- To be launched in **2020** by Sloan Foundation funding
- Initially led by **Harvard Privacy Tools** project

*A tool (algorithm) is **differentially private** if its output cannot reveal whether any individual's data was included in the original dataset or not*.

OpenDP: Use Cases Motivation

- **Archival data repositories** (e.g., Dataverse) enabling secondary reuse and reproducibility
- **Government agencies** making data available to the public, both for official statistics and open data mandates
- **Companies** sharing data for academic research or internal research.
- **Focus on "centralized model" for Differential Privacy:**
  - A central aggregator has accessed to the raw data. The aggregator transforms the data with a differentially private mechanism.

OpenDP: Principles

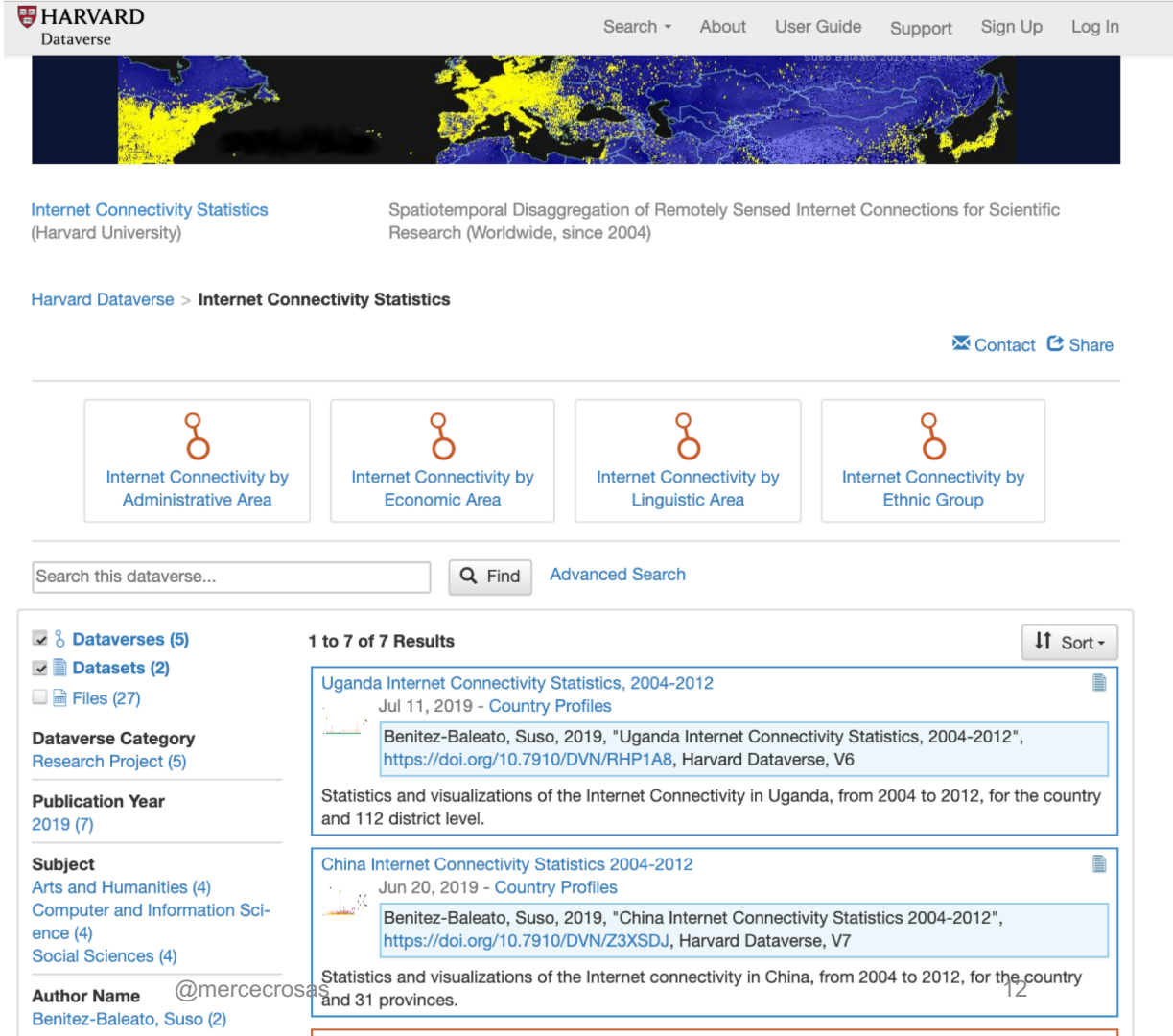
- **Open Source**
  - worldwide open-source community
  - processes and recognition for contribution
- **Security & Privacy**
  - careful vetting of any security-critical or privacy-critical code
  - can ship code to the sensitive data
- **Scalability**
  - handle petabyte-scale data
- **Extensibility**
  - can grow from the continuing research developments in the field

Example: Enabling Reproducibility with Privacy Preserving Data Sharing

Suso Baleato, James Honaker, Mercè Crosas

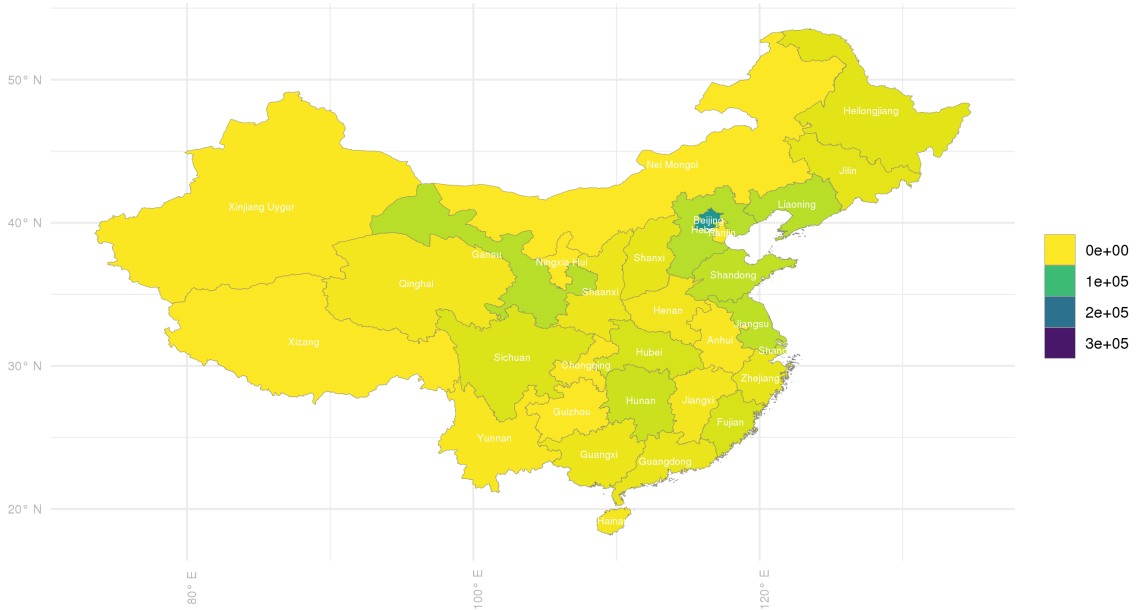
Internet Connectivity Statistics

- Data from measurements of the used **IPv4** address space across countries
- Close **correlation with internet penetration statistics** from International Telecommunications Union (ITU) and **OECD**



Differentially private data released in Dataverse

- Original data might contain **sensitive** information:
  - **DataTag = Orange**
- Statistics published in Dataverse are **differentially private**:
  - **DataTag = Green**
- With DP epsilon (noise) = 0.01, error added only 10^{-4}



Take Away

- A solution for reproducibility and publishing of **sensitive data**:
  - **Dataverse + DataTags + Differential Privacy**
  - Acceptable precision loss; e.g. Error = 10^{-5} with Epsilon = 0.2 or Error = 10^{-4} with Epsilon = 0.1 (smaller epsilon, more privacy protection)
- Internet Connectivity Statistics as a use case:
  - https://dataverse.harvard.edu/dataverse/ics
- OpenDP will be launched by Harvard Privacy Tools project:
  - https://privacytools.seas.harvard.edu

Thanks

scholar.harvard.edu/mercecrosas