## Data and Code Sharing for Open Science

CS seminar, EPFL, October 16, 2019

Mercè Crosas, Ph.D.

Chief Data Science and Technology Officer, IQSS

Harvard University's Research Data Officer, HUIT

@mercecrosas

#### This Talk

- 1. Open Science awareness
- 2. Open Science implementation considerations
- 3. Our Contribution to data sharing and reproducibility:

  Dataverse
- 4. Our Contribution to sensitive data sharing and analysis: DataTags and OpenDP

#### **National Academies of Sciences New Report**

The National Academies of

CONSENSUS STUDY REPORT

#### OPEN SCIENCE BY DESIGN Realizing a Vision for 21st Century Research



Researcher at the center, both contributes to open science and takes advantage of the open science practices:

- Knowledge generation: collect data, conduct research using tools compatible with open sharing; use automated workflow tools to ensure accessibility of research outputs.
- Validation: prepare data and tools for reproducibility and reuse and participate in replication studies.
- **Dissemination:** use **appropriate licenses** for sharing research outputs; report all results and supporting information.
- Preservation: deposit research outputs in FAIR archives and ensure long-term access to research results

#### **OpenAire New White Paper**



Researcher-centric, with openness as default in scientific communication:

- Publishing all kinds of scientific products
- Publishing semantic links
- Publishing research products as packages of workflows
- Innovation in publishing and dissemination practices
- Quality control for securing the quality, reproducibility,
   FAIRness of research results
- Assessment and reward by intelligently combining diverse/open/auditable metrics

## New Survey by Pew Research

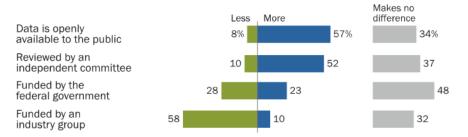
"Americans say open access to data and independent review inspire more trust in research findings"

Trust and Mistrust of American Views on Scientific Experts.

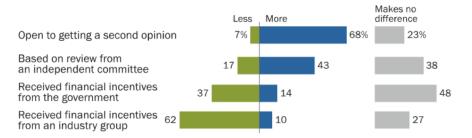
Pew Research Center, August 2, 2019

#### Majority of Americans say they are more apt to trust research when the data is openly available

% of U.S. adults who say when they hear each of the following, they trust scientific research findings ...



% of U.S. adults who say when they hear each of the following, they trust a science practitioner's recommendation ...



Note: Respondents who did not give an answer are not shown.

Source: Survey conducted Jan. 7-21, 2019.

"Trust and Mistrust in Americans' Views of Scientific Experts"

#### PEW RESEARCH CENTER

#### What Open Science Is and Isn't

- Open science can be more beneficial to humanity
  - Wider reach of data and knowledge, greater impact to human welfare
- Open Science does not guarantee good science
  - Facilitates reproducibility validation, but data/research might still be low quality
- Open Science does not always mean fully public data
  - Research data should be discoverable, but protected if sensitive

#### **Open Science Implementation**

An open science technology implementation must consider and include:

- Incentives to share data, code, and other research outputs
- Standards for metadata discovery
- Public metadata (at a minimum for citation) even when data are restricted
- Sufficient information to reuse the data, code, workflows (all research outputs)
- Support for data and code terms and use agreements
- Integration of archival repositories with computational platforms (clouds, tools)
- Solutions for collaborations that access sensitive, private data

## **Dataverse – Achievements**

Facilitating data and code sharing for Open Science

#### 15 years of Dataverse

- Launched in 2006 by Harvard's Institute for Quantitative Social Science
- Now 48 Dataverse installations sites throughout 6 continents
- Each site can support multiple Universities or research organizations
- Harvard Dataverse is the largest Dataverse repository site with:
  - 3,400 dataverses (collections of datasets)
  - 91,000 datasets
  - 560,000 files
  - 8.9 million downloads
- An active, growing open-source community with 100 contributors, 41 releases

#### Sites harvest metadata from each other



- 48 Dataverse sites
   can be federated via
   the Open Archives
   Initiative Protocol for
   Metadata Harvesting.
- Data searchable in one Dataverse site while hosted in another.

#### **An National Example: DataverseNO**





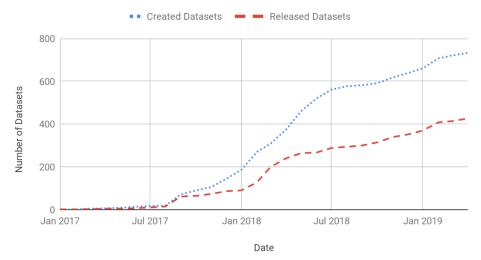
- 8 universities in Norway as members; the other 3 to join soon
- Policies and guidelines common to all DataverseNO members
- Global and local support
- Applied for Core Trust Seal certificate
- https://site.uit.no/dataverseno/

@mercecrosas

11

#### A Consortium Example: Texas Dataverse





## Texas Data Repository

- 11 universities in Texas as members
- Led by Texas Digital Library Consortium
- Texas Data Repository steering committee focuses on outreach
- Working on Core Trust Seal
- Differing policies/practices per member
- https://www.tdl.org/texas-data-repository/

#### Combines technology, standards, incentives

- A data citation with a globally unique persistent identifier and credit for data author
- Standard metadata, with variable-level metadata, plus rich custom metadata
  - Schema.org JSON-LD, DataCite, Dublin Core, DDI, OAI-ORE, OpenAire, PROV
- Tiered data access, depending on data restrictions:
  - Fully Open, CC0; Guestbook; Restricted w/ Data Use Agreement
- Data publishing workflows, with anonymous review, and reviewer/curator roles
- Multiple versions of a dataset
- Branding and customization for each dataverse (collection of datasets)
- Overall, follows FAIR guiding principles

#### SCIENTIFIC DATA

Article Open Access | Published: 10 April 2019

## A data citation roadmap for scholarly data repositories

Martin Fenner, Mercè Crosas, Jeffrey S. Grethe, David Kennedy, Henning Hermjakob, Phillippe Rocca-Serra, Gustavo Durand, Robin Berjon, Sebastian Karcher, Maryann Martone & Tim Clark

```
Scientific Data 6, Article number: 28 (2019) | Download Citation 

2512 Accesses | 3 Citations | 53 Altmetric | Metrics >>>
```

#### **Abstract**

This article presents a practical roadmap for scholarly data repositories to implement data citation in accordance with the Joint Declaration of Data Citation Principles, a synopsis and harmonization of the recommendations of major science policy bodies. The roadmap was developed by the Repositories Expert Group, as part of the Data Citation Implementation Pilot (DCIP) project, an initiative of FORCE11.org and the NIH-funded BioCADDIE (https://biocaddie.org) project. The roadmap makes 11 specific recommendations, grouped into three phases of implementation: a) required steps needed to support the Joint Declaration of Data Citation Principles, b) recommended steps that facilitate article/data publication workflows, and c) optional steps that further improve data citation support provided by data repositories. We describe the early adoption of these recommendations 18 months after they have first been published, looking specifically at implementations of machine-readable metadata on dataset landing pages.

#### **Data Citation Implementation**

- Global unique persistent identifiers (GUID)
- Allow to cite specific version or subset
- GUID resolves to dataset landing page
- Landing page w/ instructions to access data

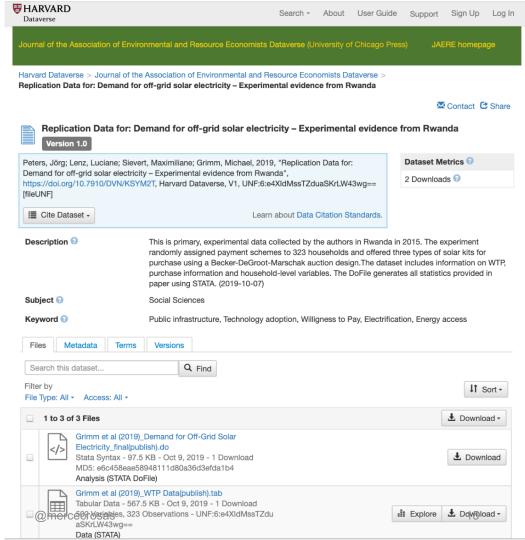
14

- Machine-readable citation metadata
- Schema.org JSON-LD in landing page
- Standard Bibliographic formats

@mercecrosas

#### **Dataset Landing Page**

- Dataset and file citation
- Data files review and downloads
  - Automatic transforms tabular files to multiple formats
  - Extracts variable metadata
- Dataset Metadata
- Metrics: Make Data Count
- Terms and data use agreements
- Versions and Provenance
- APIs to access metadata & data
- Machine-actionable



## Dataverse – Next

Improving research reproducibility and data reuse

8,000 of the 90,000 datasets in Harvard Dataverse contain the files to reproduce the published results documentation data code



User Guide

Support

Sign Up Log In

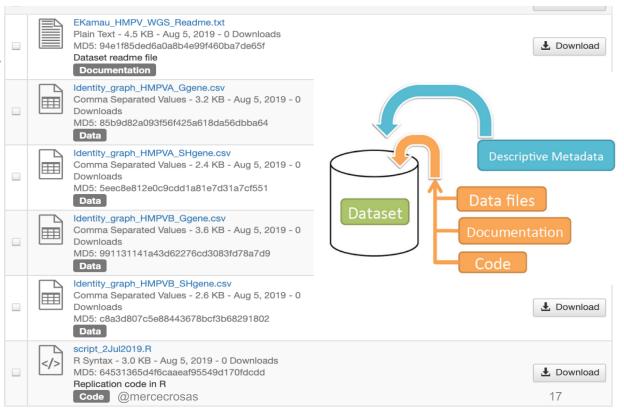


Virus Epidemiology and Control (VEC) Dataverse (Kemri Wellcome Trust Research Programme, Kilifi, Kenya)

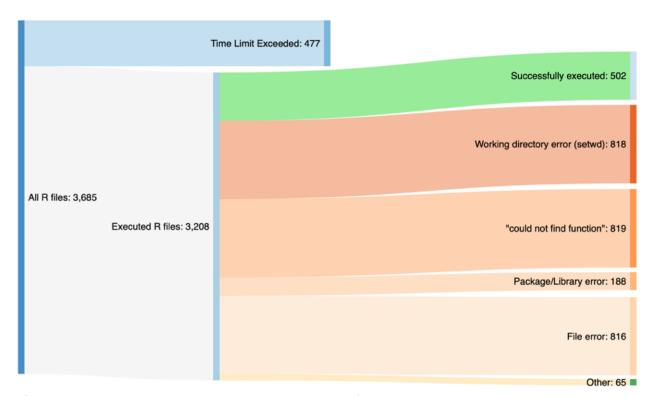
Population dynamics of viral pathogens informing intervention strategies

Harvard Dataverse > KWTRP Research Data Repository > Virus Epidemiology and Control (VEC) Dataverse >

Replication Data for: Whole genome sequencing and phylogenetic analysis of Human metapneumovirus strains from Kenya and Zambia



#### Re-execution of R Code in published datasets



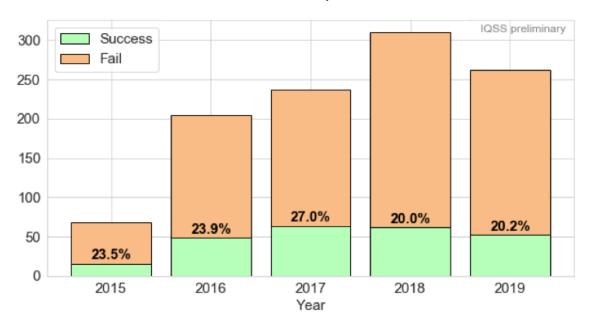
84.4% R code files fail to execute

77.5% of datasets with R files contain non-executable code

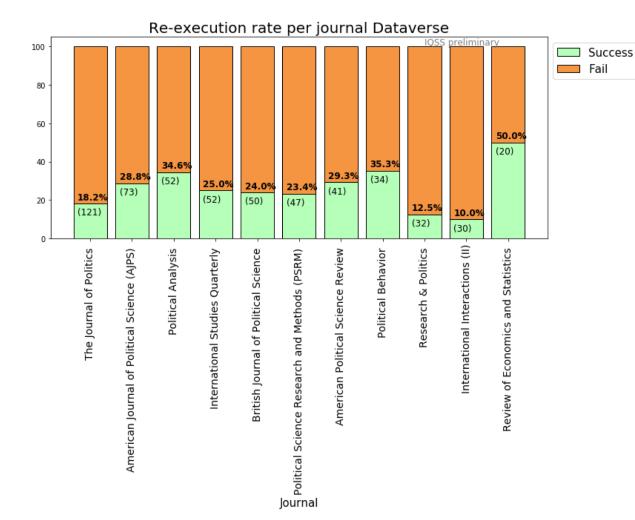
Source: Current study on code execution for datasets published in Dataverse by Ana Trisovic (Harvard's IQSS) funded by Sloan Foundation

#### Re-execution of R code per year

Re-execution of R code in datasets published in Harvard Dataverse



Re-execution of R code in Journals
Replication Data
published in
Harvard Dataverse



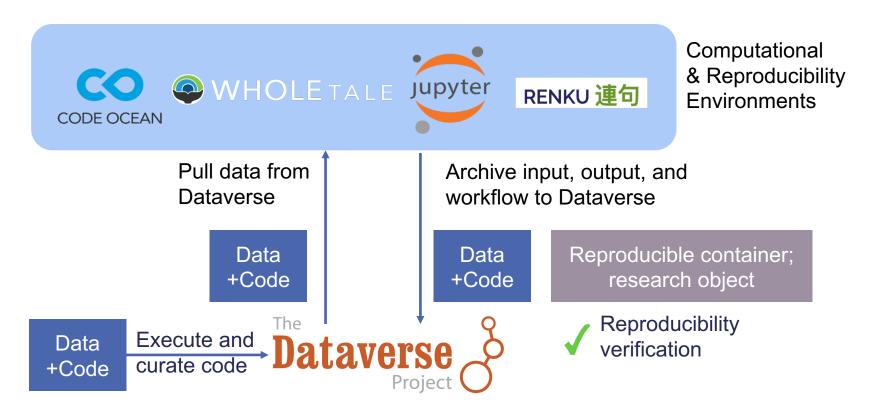
#### New Features: computational reproducibility

- Include reproducibility as part of review workflow
- Integrate Dataverse with computational tools to facilitate code execution
- Deposit a capsule (container with data and code) once verified for reproducibility
- When possible, automate code execution upon publishing the data and code

#### Also, in the process of evaluating integration with:

- Data Curation Tools
- Research Objects
- Citation and metadata for software

#### Integration with computational tools



## **DataTags**

Mercè Crosas, Latanya Sweeney, Michael Bar-Sinai 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

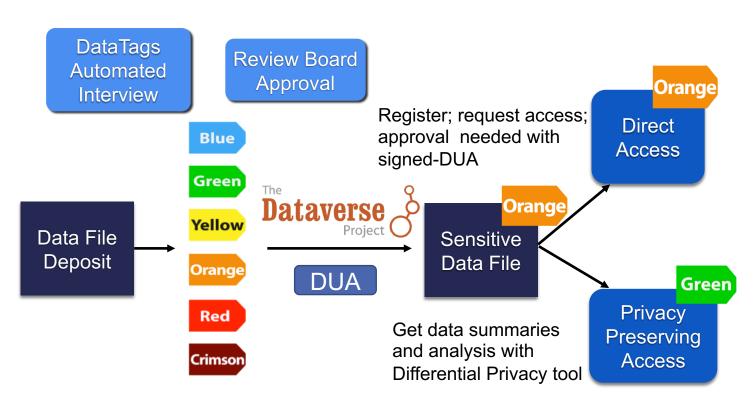
DataTag	Access	Authorization	Data Use Agreement	Encryption
Blue	Public			
Green	Public	+ Register		
Yellow	Restricted	+ Approval Needed	+ Click-thru DUA	+ Encrypted transit
Orange	Restricted	+ Approval Needed	+ Signed DUA	+ Encrypted transit + Encrypted storage
Red	Restricted	+ Approval Needed	+ Signed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

mercecrosas

26

#### 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

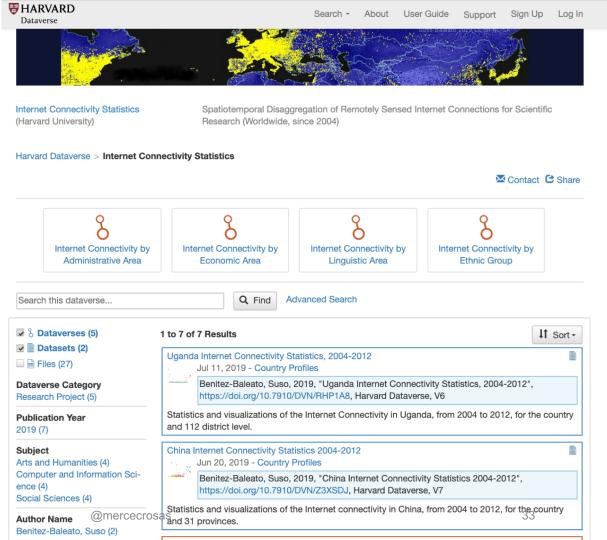
can grow from the continuing research developments in the field

# **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
- Map with DP epsilon (noise) = 0.01, error added only 10<sup>-4</sup>
- Error = 10<sup>-5</sup> with Epsilon = 0.2, smaller epsilon, more privacy protection



#### **Conclusions**

- Dataverse integrated with computational environments and tools is a solution for implementing Open Science, for a university, for a consortium, or nationally
- Dataverse with DataTags and OpenDP is a solution for sharing sensitive data
- A technology solution should include standards and incentives to share openly
- A technology solution should automatically assist in improving the quality of research outputs and tracking provenance
- But data and code authors must do their part to make their research outputs highquality and reusable

### **Thanks**









The Global Dataverse Community Consortium Supporting Dataverse repositories around the world.



Harvard University Privacy Tools Project

<u>dataverse.org</u> | <u>dataversecommunity.global/</u> | <u>scholar.harvard.edu/mercecrosas</u>