FAIR Data Management
FAIR Data Sharing

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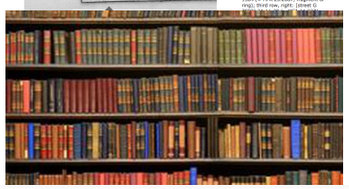
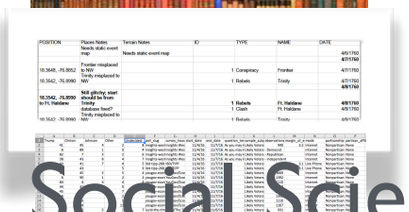
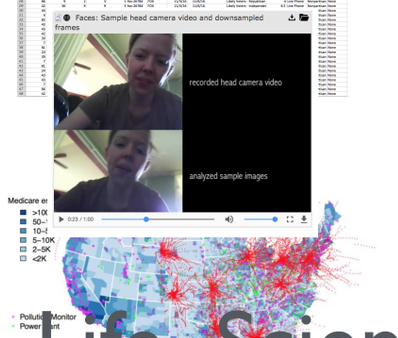
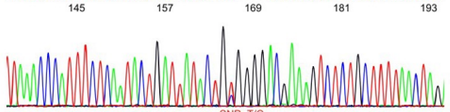
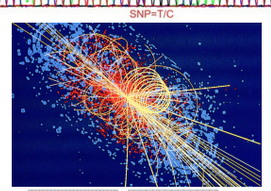
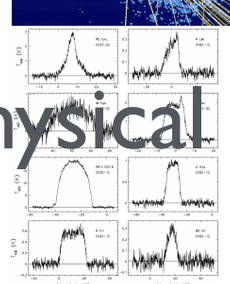
Critical Perspectives of on the Practice of Digital Archeology, Harvard University, February 3, 2017

primary data	order, transform, & analyze them	gain knowledge, make decisions
Digital Humanities	Catalog Classify Visualize Quantify Summarize Geo reference	Learn about the whole from a part.
Social Sciences	Inference Missing data Forecast Causal Inference Coding Annotations Associations Likelihoods Compare with theory	Tell a story. Make a prediction. Ultimately explain.
Life Sciences		
Physical Sciences		

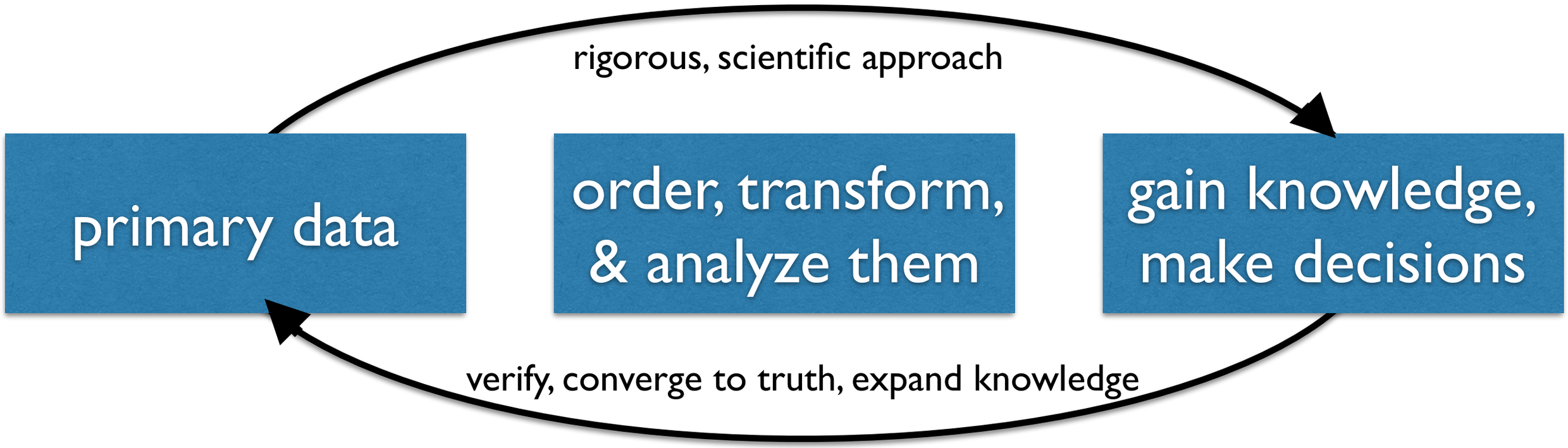







Nullius in Verba: “Take nobody’s word for it”

(Royal Society, Philosophical Transactions, 1965)



- **Replication:** Independent scientific experiments to validate findings
- **Reproducibility:** Calculation of quantitative results by others using original datasets and methods

(Definitions by Stodden, Leisch, Peng, Implementing Reproducible Research, 2014)
* Replication and reproducibility definitions vary across disciplines

“Answering even a simple scientific question requires lots of choices that can shape the results”

The Economist | World politics | Business & finance | Econ

Problems with scientific research

How science goes wrong

Scientific research has changed the world. Now it needs

Oct 19th 2013 | From the print edition

A SIMPLE idea underpins science: “trust, but verify”. Resu
challenge from experiment. That simple but powerful idea h
knowledge. Since its birth in the 17th century, modern scie
beyond recognition, and overwhelmingly for the better.

“When possible, make data, methods, and code open to verify”

INSIGHTS

Design principles for synthetic ecology p. 1425 ▸

Whacking hydrogen into metal p. 1429

SCIENTIFIC INTEGRITY

Self-correction in science at work

Improve incentives to support research integrity

By **Bruce Alberts,[1] Ralph J. Cicerone,[2] Stephen E. Fienberg,[3] Alexander Kamb,[4] Marcia McNutt,[5]* Robert M. Nerem,[6] Randy Schekman,[7] Richard Shiffrin,[8] Victoria Stodden,[9] Subra Suresh,[10] Maria T. Zuber,[11] Barbara Kline Pope,[12] Kathleen Hall Jamieson[13,14]**

Week after week, news outlets carry word of new scientific discoveries, but the media sometimes give suspect science equal play with substantive discoveries. Careful qualifications about what is known are lost in categorical headlines. Rare instances of misconduct or instances of irreproducibility are translated into concerns that science is broken. The October 2013 *Economist* headline proclaimed “Trouble at the lab: Scientists like to think of science as self-correcting. To an alarming degree, it is not” (*1*). Yet, that article is also rich with instances of science both policing itself, which is how the problems came to *The Economist*’s attention in the first place, and addressing discovered lapses and irreproducibility concerns. In light of such issues and efforts, the U.S. National Academy of Sciences (NAS) and the Annenberg Retreat at Sunnylands convened our group to examine ways to remove some of the current disincentives to high standards of integrity in science.

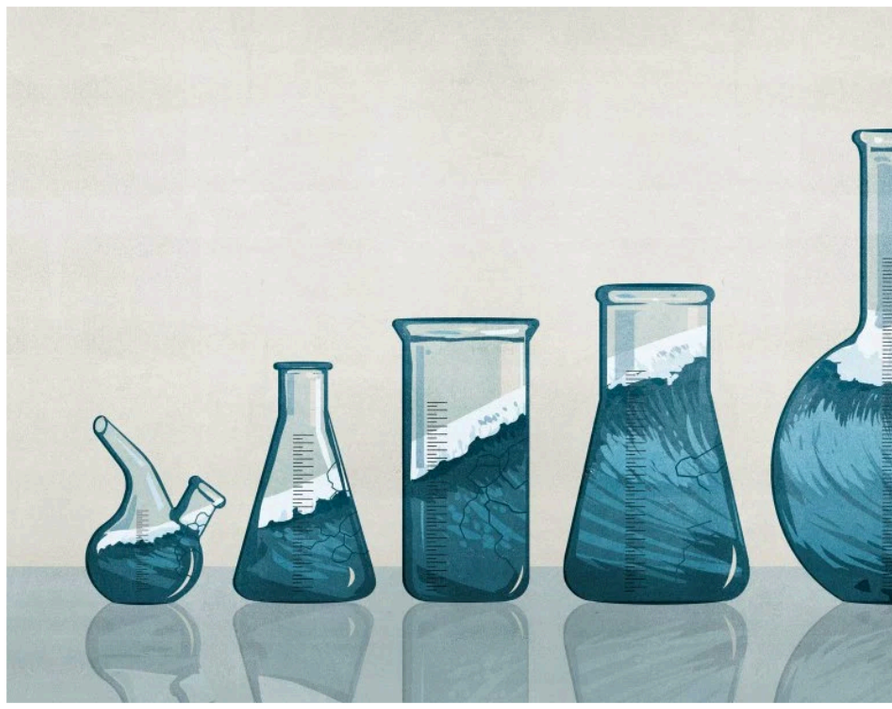
Like all human endeavors, science is imperfect. However, as Robert Merton noted more than half a century ago “the

POLICY

activities of scientis
ous policing, to a d
leled in any other
a result, as Popper
of the very few hu
the only
are syste
fairly of
(*3*). Instances in wh
address flaws in w
of success, not fail
onstrate the underl
nisms of science at
Still, as in any h
writ large does not
als. Although attem
Wakefield study a
between autism ar

“Science/research might be imperfect, but is self-correcting”

FiveThirtyEight Science



THE SCIENTIFIC METHOD | 7:00 AM | AUG 19, 2015

Science Isn’t Broken

It’s just a hell of a lot harder than we give it credit for.

By CHRISTIE ASCHWANDEN

Graphics by RITCHIE KING

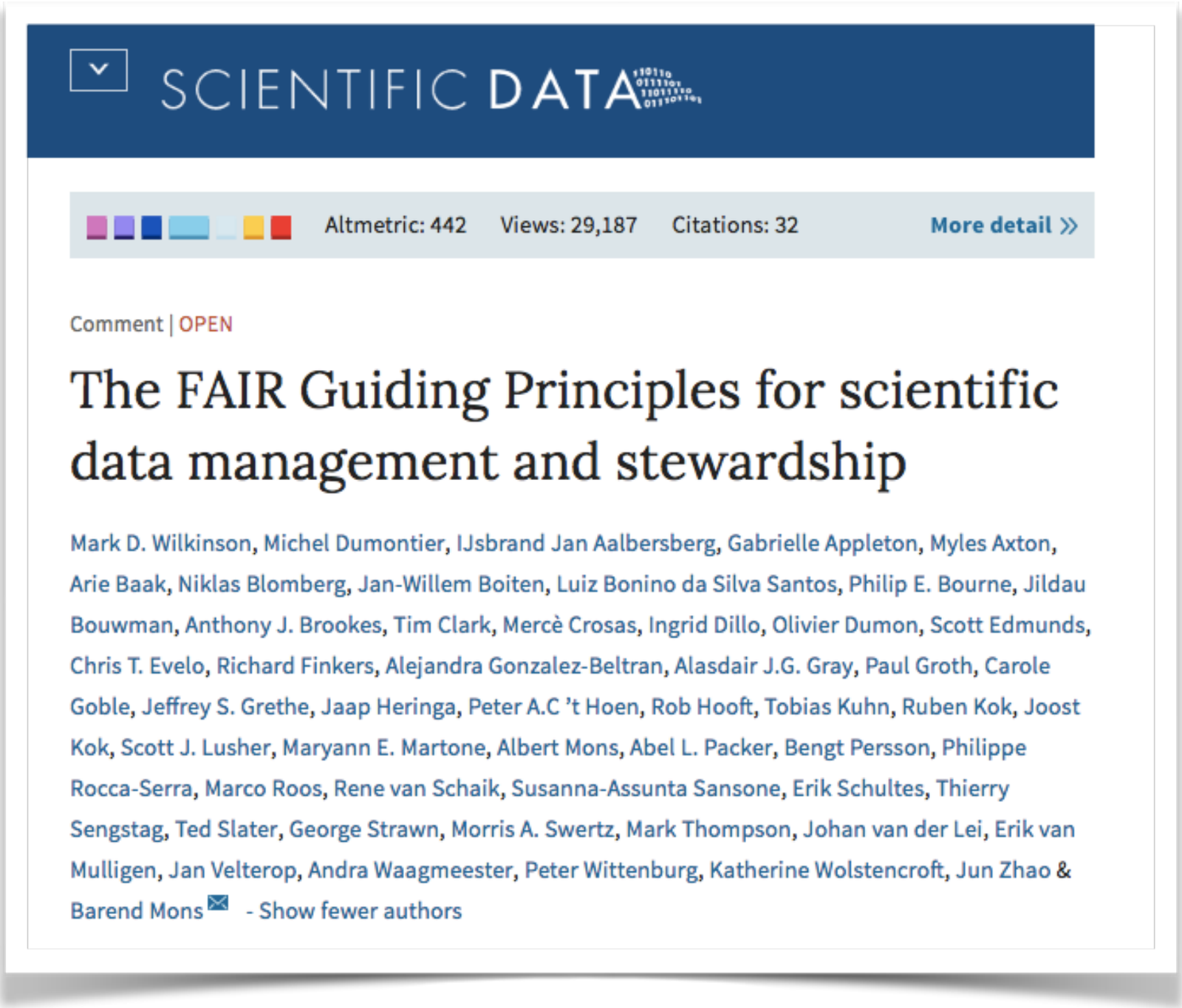
If you follow the headlines, your confidence in science may have taken a hit lately.Peer review? More like self-review. An investigation in November uncovered a scam in which researchers were rubber-stamping their own work, circumventing peer review at five high-profile

“It’s not unreliable, but more challenging that we give it credit for”

Caring for and sharing your data (and code) enable you and others to correct and reuse them

1. Love your data 2. Share your data 3. Conduct science with reuse in mind 4. Publish workflow 5. Link data to publications 6. Publish your code 7. State how you want to get credit 8. Foster and use repositories 9. Reward colleagues who share 10. Boost Data Science

Data should be Findable, Accessible, Interoperable, Reusable (FAIR) by machines



Wilkinson et al, 'The FAIR Guiding Principles scientific data management and stewardship," Nature Scientific Data, 2016; NIH Data Commons Principles; Joint Declaration of Data Citation Principles (Force11)

"**FAIR Principles** put specific emphasis on enhancing the ability of machines to automatically find and use the data, in addition to supporting its reuse by individuals."

"**Good data management** is not a goal in itself, but rather is the key conduit leading to knowledge discovery and innovation, and to subsequent data and knowledge integration and reuse by the community after the data publication process."

FAIR Data Principles in Brief

- **To be Findable:**
  - (meta)data are assigned a globally unique and persistent identifier
  - data are described with rich metadata
  - metadata clearly and explicitly include the identifier of the data it describes
  - (meta)data are registered or indexed in a searchable resource
- **To be Accessible:**
  - (meta)data are retrievable by their identifier using a standardized communications protocol
  - the protocol is open, free, and universally implementable
  - the protocol allows for an authentication and authorization procedure, where necessary
  - metadata are accessible, even when the data are no longer available
- **To be Interoperable:**
  - (meta)data use a formal, accessible, shared, and broadly applicable language for knowledge representation.
  - (meta)data use vocabularies that follow FAIR principles
  - (meta)data include qualified references to other (meta)data
- **To be Reusable:**
  - meta(data) are richly described with a plurality of accurate and relevant attributes
  - (meta)data are released with a clear and accessible data usage license (meta)data are associated with detailed provenance (meta)data meet domain-relevant community standards

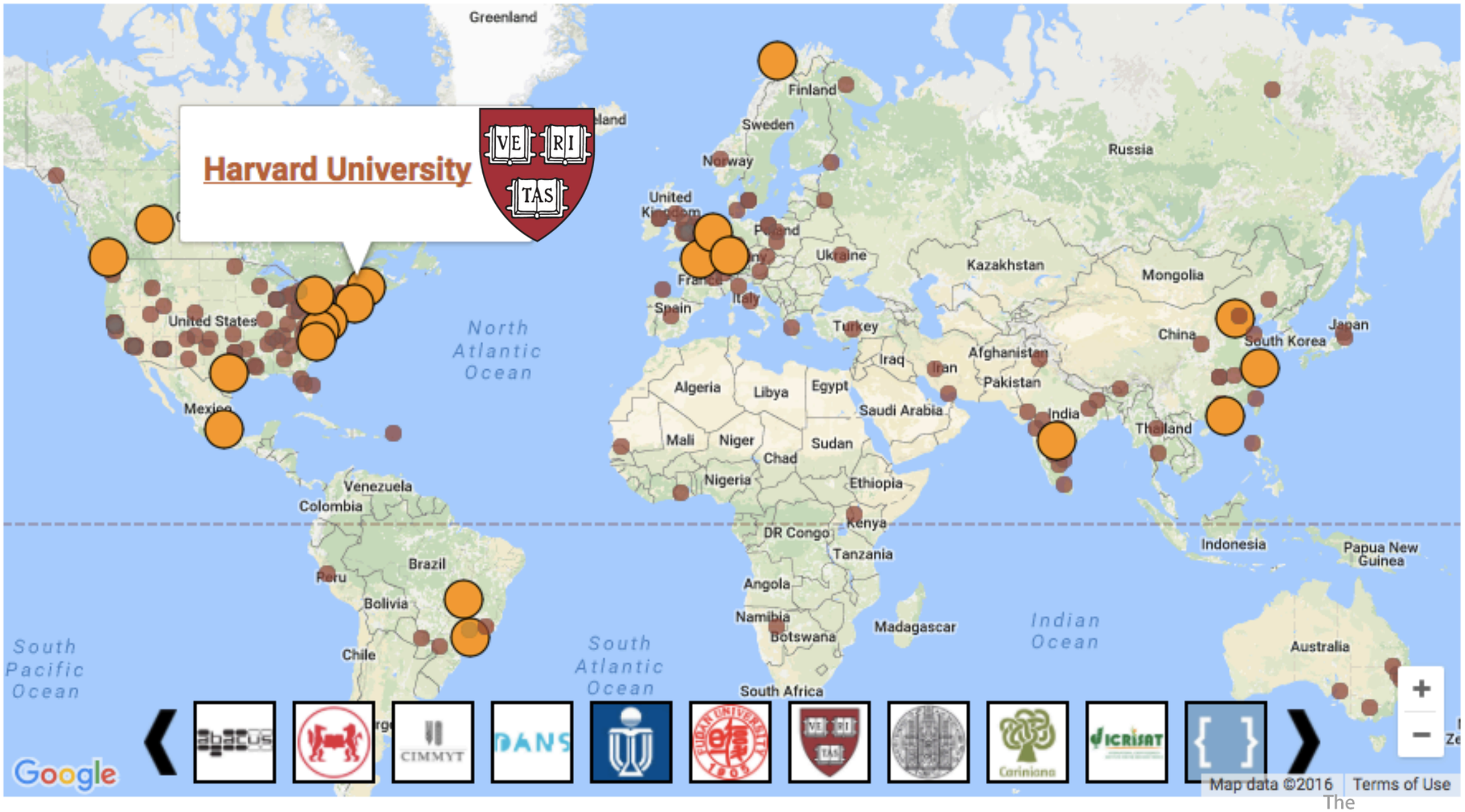
We built Dataverse to incentivize data sharing, with “good data management” in mind

- An **open-source** platform to share and archive data
- Developed at Harvard’s Institute for Quantitative Social Science since 2006
- Gives **credit and control** to researchers
- Builds a **community** to:
  - define new standards and best practices
  - foster new research and collaboration in data sharing
- Has brought **data publishing** into the hands of researchers

The Dataverse Project

Dataverse is now a widely used repository platform

21 installations around the world
Used by researchers from > 500 institutions
60,000 datasets in Harvard Dataverse repository
http://dataverse.org



The Dataverse Project

Dataverse has a growing, engaged community of developers and users

38
GitHub
contributors

332
members in the
community list

23
community calls

with 239
participants from

8 countries

Annual
Community Meeting,

with 200 attendees

The Dataverse Project

Dataverse implements FAIR Data Principles

- **Data Citation with global persistent IDs:**
  - Generate DOI automatically
  - attribution to data authors and repository
  - registration to DataCite
- **Rich Metadata:**
  - citation metadata
  - domain-specific descriptive metadata
  - variable and file metadata (extracted automatically)
- **Access and usage controls:**
  - open data as default, with CC0 waiver
  - custom terms of use and licenses, when needed
  - data can be restricted, but citation & metadata always publicly accessible
- **APIs and standards:**
  - SWORD, OAI-PMH, Dataverse native open API
  - Dublin Core and DDI metadata standards
  - PROV ontology standard to capture provenance of a dataset (coming soon)

The Dataverse Project

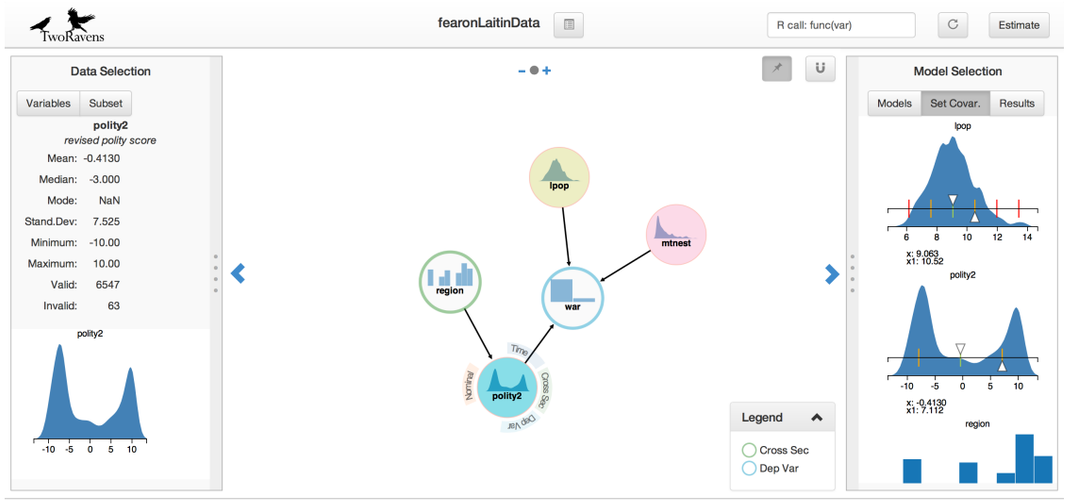
Standard file formats and automatic metadata extraction allow data exploration

Var1	Var2	Var3	Var4

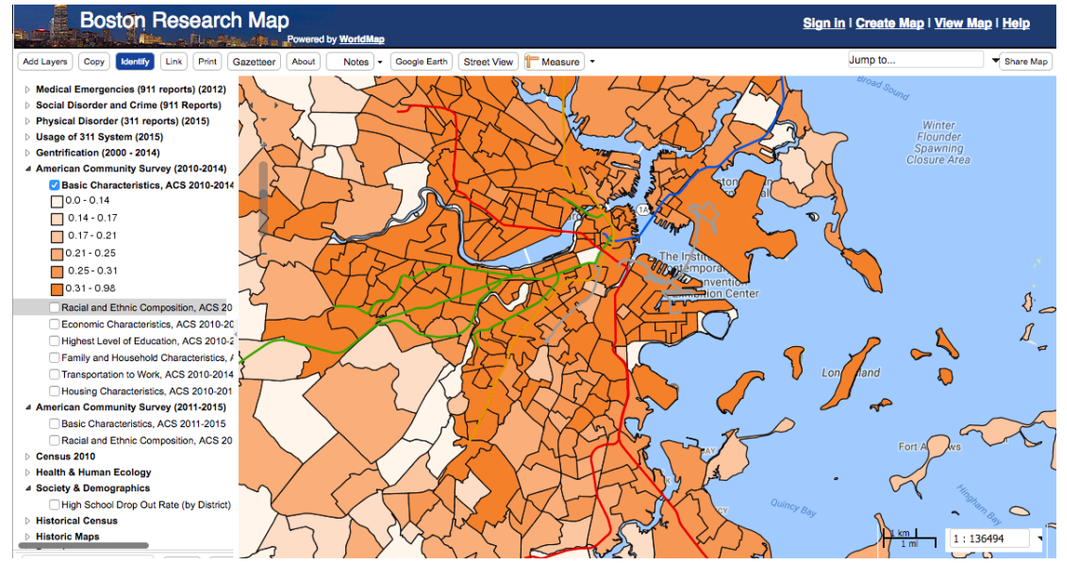
geospatial variable

Var1	Var2	Var3	Var4

TwoRavens: summary stats & analysis

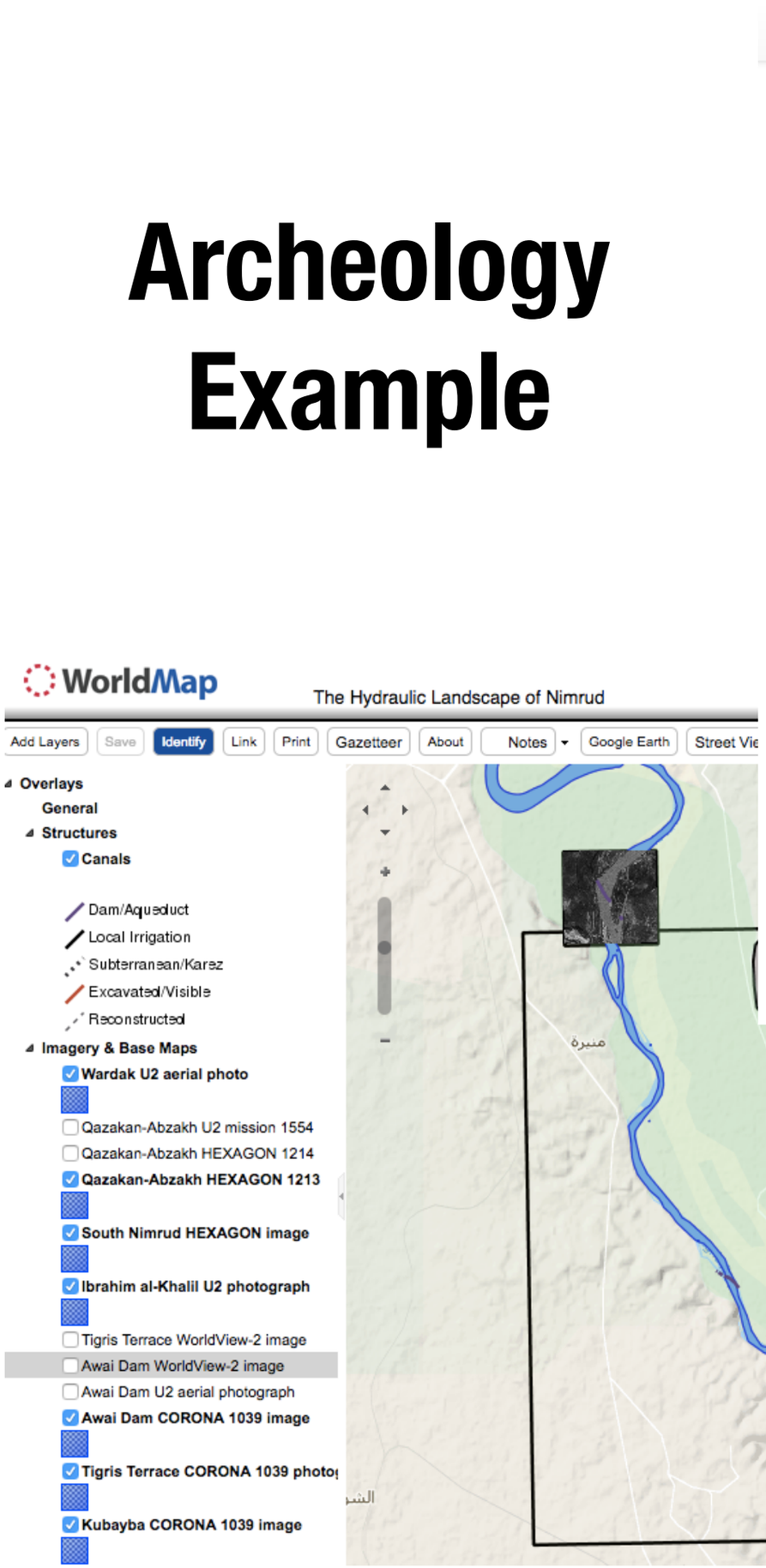


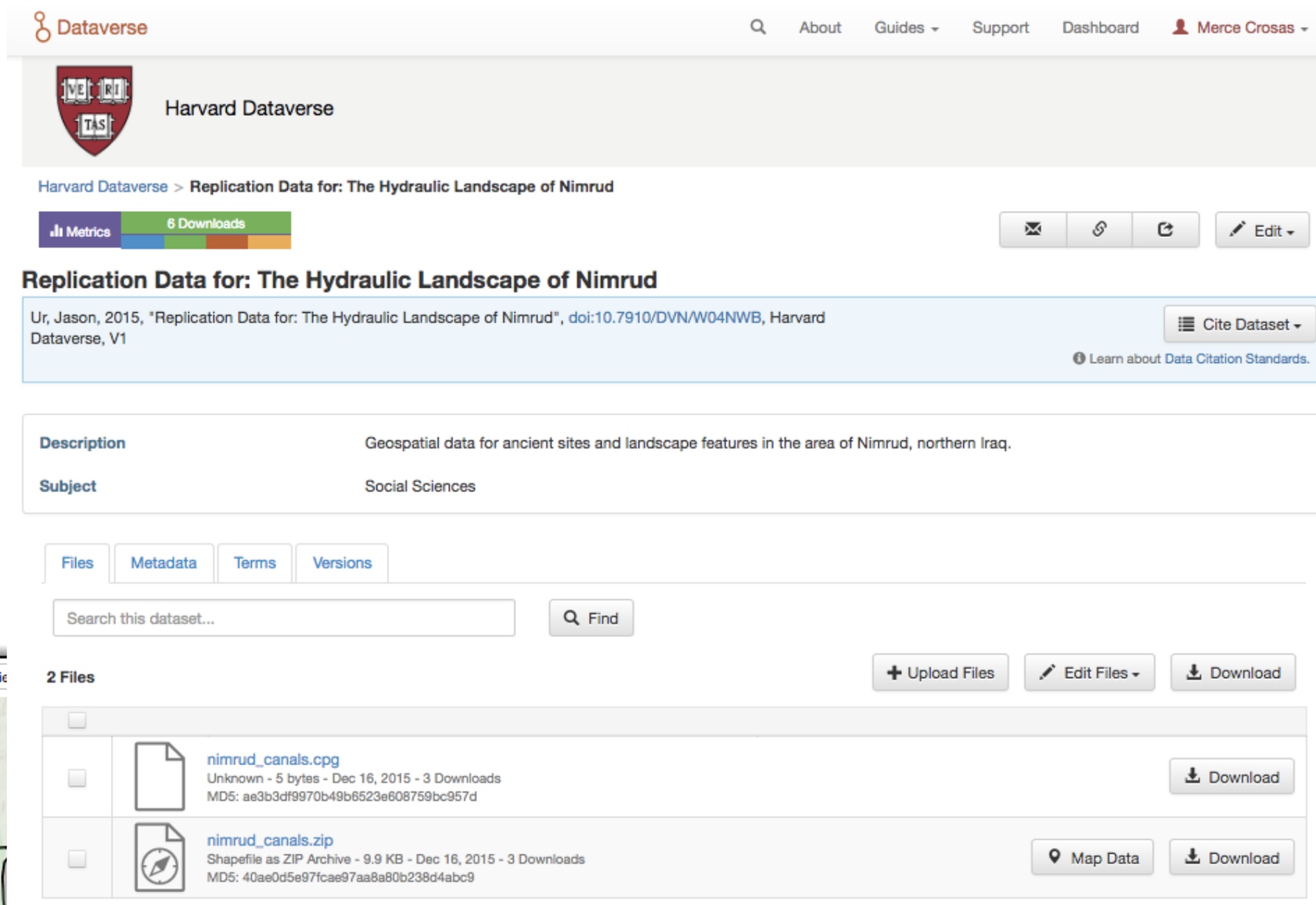
WorldMap: geospatial exploration



The Dataverse Project

Archeology Example





In the works: data citation roadmap to improve data discoverability

- Force11 Data Citation Implementation Pilot
- Landing page for dataset with machine-actionable standard citation metadata
- Working with Google to include dataset metadata in schema.org

Cold Spring Harbor Laboratory

bioRxiv beta

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A Data Citation Roadmap for Scholarly Data Repositories

Martin Fenner, Mercè Crosas, Jeffrey Grethe, David Kennedy, Henning Hermjakob, Philippe Rocca-Serra, Robin Berjon, Sebastian Karcher, Maryann Martone, Timothy Clark

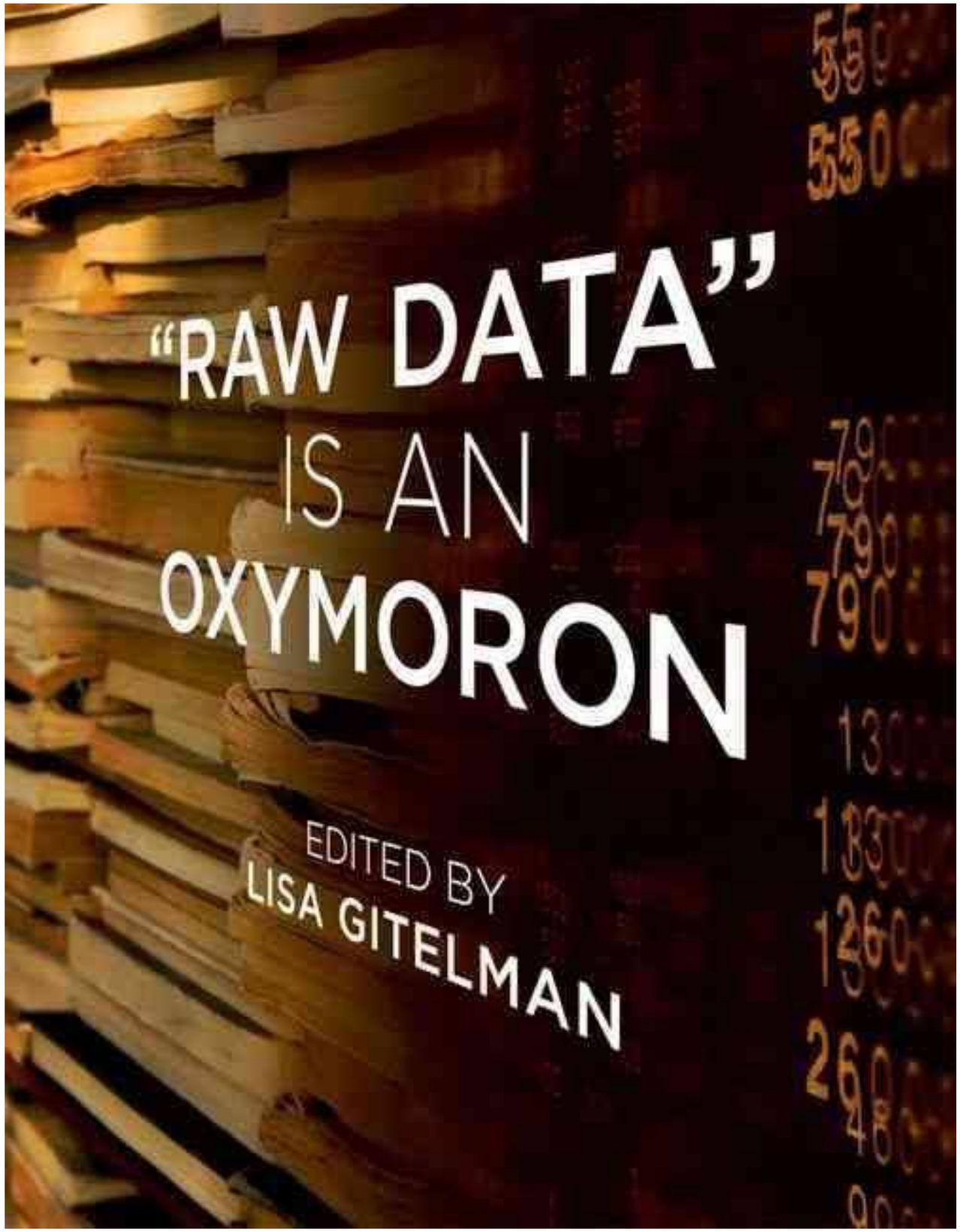
doi: https://doi.org/10.1101/097196

This article is a preprint and has not been peer-reviewed [what does this mean?].

Abstract | Info/History | Metrics | Preview PDF

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 (Data Citation Synthesis Group, 2014), a synopsis and harmonization of the recommendations of major science policy bodies. The roadmap was developed by the Repositories Early Adopters Expert Group, part of the Data Citation Implementation Pilot (DCIP) project (FORCE11, 2015), an initiative of FORCE11.org and the NIH BioCADDIE (2016) program. 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 shouldn't think of data as a natural resource but as a cultural one that needs to be generated, protected, and interpreted."

Thank you!

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