UCLA

Presentations

Title

Sharing, Reusing, and Repurposing Data

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The Conundrum of Sharing Research Data

If the rewards of the data deluge are to be reaped, then researchers who produce those data must share them, and do so in such a way that the data are interpretable and reusable by others.*



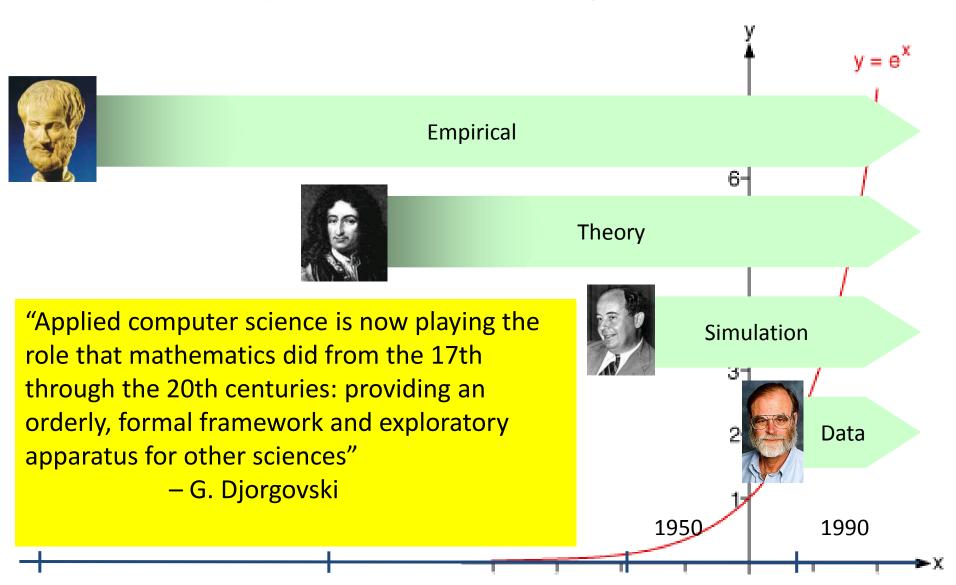
*Borgman, C.L. (2012). The Conundrum of Sharing Research Data. JASIST, 63(6):1059–1078

Overview

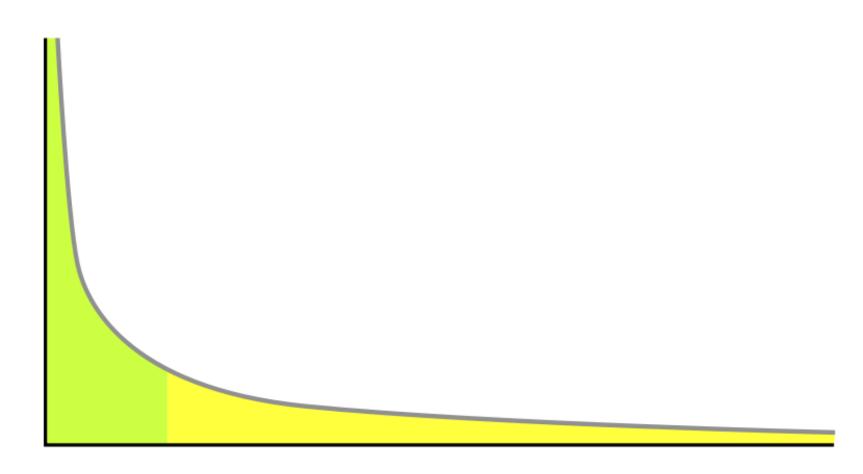


- Paradigm shift
- Arguments for sharing data
- Science friction, data friction
- Success factors for reusing and repurposing data

New problem solving methods



The long tail of data



Number of researchers

Slide: The Institute for Empowering Long Tail Research

Big Data

Little Data

No Data

No Data is the Norm

Data sharing imperatives

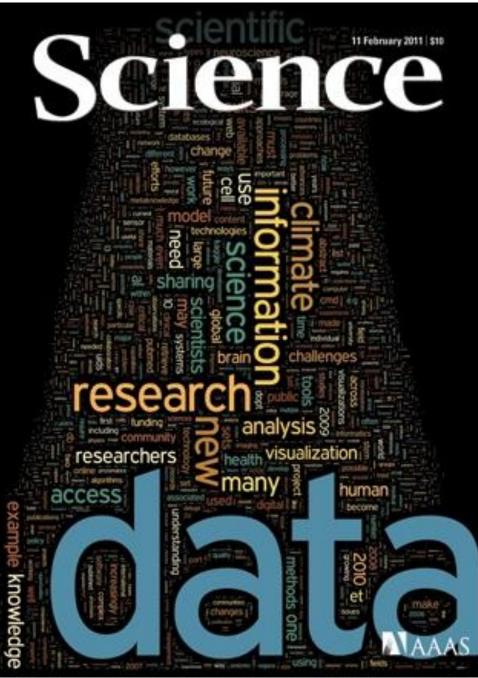
- Research Councils of the UK
 - Open access publishing requirements
 - Provisions for access to data
- Wellcome Trust
 - Open access publishing
 - Data sharing requirements
- National Science Foundation
 - Data sharing requirements
 - Data management plans
- U.S. Federal policy-2013
 - Open access to publications
 - Open access to data



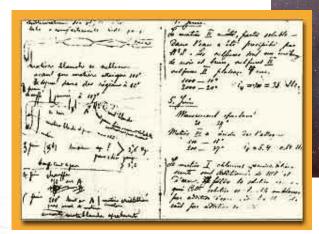






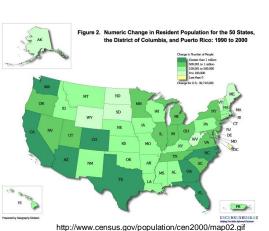


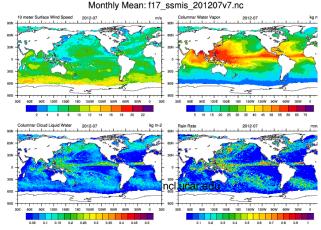
What are data?



Marie Curie's notebook aip.org

hudsonalpha.org





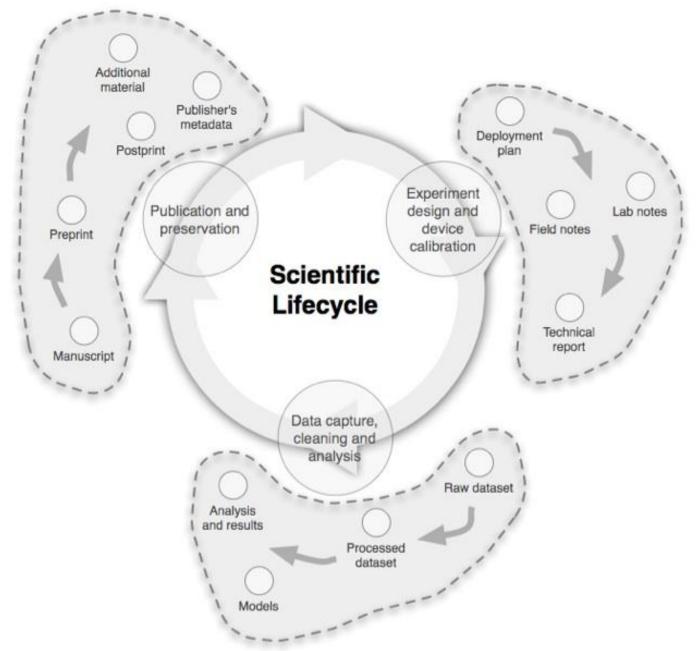
Date:1/2.07.75 Place:Sakaltutan Zafor

He will grow old in his present house; new house is for sons - 5 sons. Not sure they want to live in village. He will only build another if they want him to. eS came from Germany and did the plastering. He arranged the carpentry in Kayseri. Çok para gitti. (much money went) Has a tractor.

Date: July 1980 Place: Sakaltutan Zafor:

Household now Zafor and wife; Nazif Unal and wife and youngest son, still a boy. They run two dolmuß; one with a driver from Süleymanli. Goes in and out once a day. He gets 8,000 a month. Zafor then said, keskin deoil. { not sharp - i.e.? not profitable} I said he did very well on 8,000 TL with only two journeys a day. Nazif Unal has "bought" a Durak (dolmuß stop) from Belediye and works all day in Kayseri.

http://onlineqda.hud.ac.uk/Intro_QDA/Examples_of_Qualitative_Data.php



Pepe, A., Mayernik, M. S., Borgman, C. L. & Van de Sompel, H. (2010). From Artifacts to Aggregations: Modeling Scientific Life Cycles on the Semantic Web. Journal of the American Society for Information Science and Technology, 61(3): 567–582.

Overview

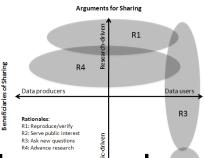


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- Success factors for reusing and repurposing data

Why share research data?

Rationales

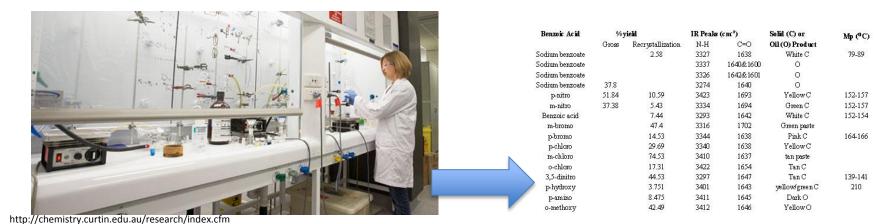


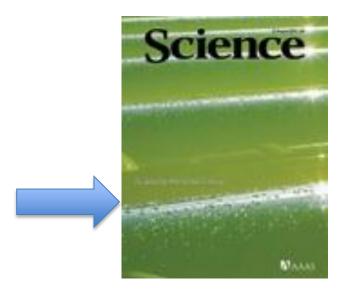


- 2. To make results of publicly funded research available to the public
- 3. To enable others to ask new questions of extant data
- 4. To advance the state of research and innovation

Borgman, C.L. (2012). The Conundrum of Sharing Research Data. JASIST, 63(6):1059–1078

1. Reproduce or verify research







Scientific Gold Standard



REPLICATION—THE CONFIRMATION OF RESULTS AND CONCLUSIONS FROM ONE STUDY obtained independently in another—is considered the scientific gold standard.

Jasny, B. R., Chin, G., Chong, L. & Vignieri, S. (2011). Again, and again, and again. Science, 334(6060): 1225.







Victoria Stodden, Columbia

- Deductive sciences
 - Check the proof
- Experimental sciences
 - Redo the field work
- Computational sciences
 - Start with the dataset
 - Reconstruct workflow

Reproducibility?

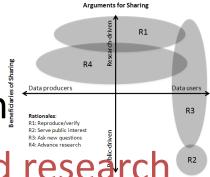
Analytic validity	Do different labs, techniques, and platforms measure the same thing?
Repeatability	Can other scientists access the data and protocols, repeat the analyses, and get the same results?
Replication	Do many different data sets and their combination (meta-analysis) get consistent results?
External validation	Do different data sets by different teams, preferably prospectively and with large-scale evidence, get consistent results?
Clinical validity	Does the discovered information predict clinical outcomes?
Clinical utility	Does the use of the discovered information improve clinical outcomes?



Why share research data?

Rationales





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2. Public monies serve the public good



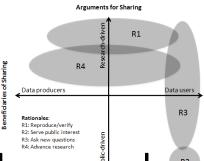




Why share research data?

Rationales





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- 4. To advance the state of research and innovation

3. Others can ask new questions



data





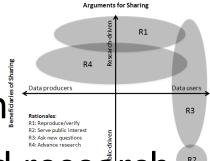
discovery

http://annualreport.ucdavis.edu/2008/images/photos/discovery.jpg

Why share research data?

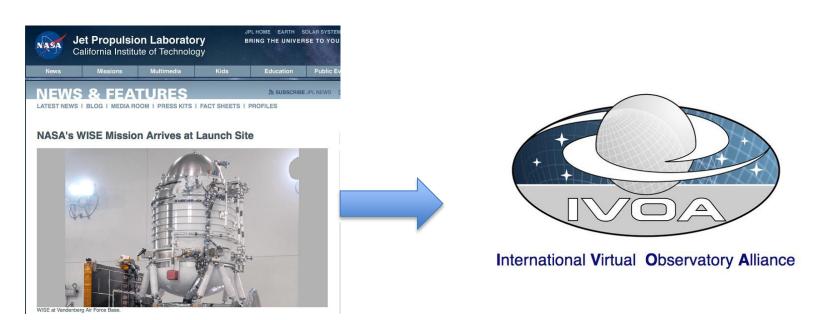
Rationales





- 2. To make results of publicly funded research available to the public
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4. Data curation advances research





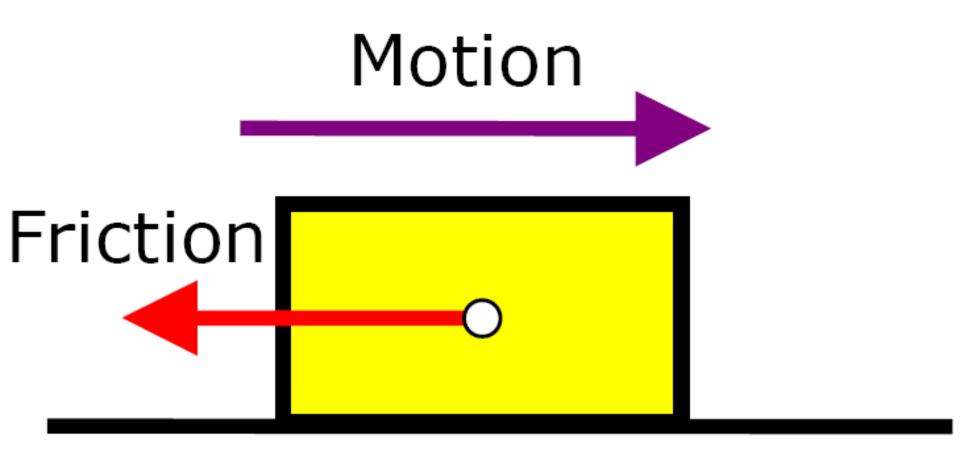




Overview



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http://www.stmary.ws/highschool/physics/home/notes/dynamics/friction/imgE2.gif

Science friction, data friction*

Data are unruly objects

 Data do not stand alone

 Data reuse is a function of distance from origin

Intractable problems



Data are unruly objects*

- Poorly bounded
- Malleable, mutable, mobile (Latour)
- Dynamic, evolving
- Signal to noise varies by use



*Wynholds, L. A. (2010). Linking to Scientific Data: Identity Problems of Unruly and Poorly Bounded Digital Objects. Presented at the Digital Curation Conference, 15 June 2011. http://www.ijdc.net/index.php/ijdc/article/view/174

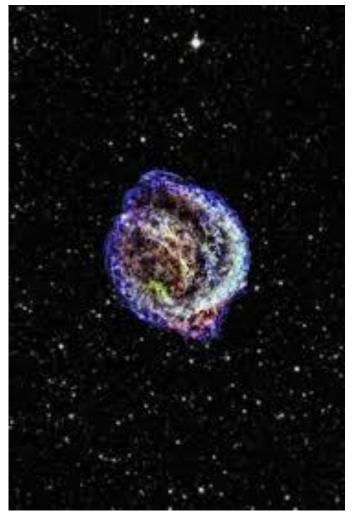
Data do not stand alone

- Data are inseparable
 - Code
 - Technical standards
 - Documentation
 - Instrumentation
 - Calibration
 - Provenance
 - Workflows
 - Local practices
 - Physical samples



Data reuse is a function of distance from origin

- Reuse by investigator
- Reuse by collaborators
- Reuse by colleagues
- Reuse by unaffiliated others
- Reuse at later times
 - Months
 - Years
 - Decades
 - Centuries



Intractable problems

- Confidentiality
- Anonymization
- Reidentification
- Intellectual property
- Economics



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How to share data

Curated data archive: NASA, UKDA, ICPSR...

Author curated data archive

University data archive: ORA

- Personal website
- ftp site
- Email on request



http://www.zippykidstore.com/

Simple Rules for the Care and Feeding of Scientific Data*

- 1. Good science requires good data
- 2. Make your science inspectable by others
- 3. Conduct your science with provenance in mind
- 4. Do not reduce your data more than necessary
- 5. Make your data available
- 6. Make your workflows available
- 7. Publish all software, even small scripts
- 8. Foster a "data community" for your community
- 9. Describe how you want to be acknowledged
- 10. Attribute the sources of data that you use

^{*}DRAFT: Radcliffe Seminar on Data Provenance, 9-10 May 2013, A. Goodman & X-L Meng

Conclusions

- Data reuse is part of open science / open scholarship
- Data sharing is a paradigm shift
- Data are not journal articles (yet)
- Data are messy
- Data sharing is a necessary but not sufficient condition for reuse
- Data reuse depends on
 - Conditions of sharing
 - Conditions of reuse
- Data friction is part of scholarship
- Better practices in managing data will increase





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 - Towards a Virtual Organization for Data Cyberinfrastructure, #OCI-0750529, C.L. Borgman, UCLA, PI; G. Bowker, Santa Clara University, Co-PI; T. Finholt, University of Michigan, Co-PI.
 - Monitoring, Modeling & Memory: Dynamics of Data and Knowledge in Scientific Cyberinfrastructures:
 #0827322, P.N. Edwards, UM, PI; Co-PIs C.L. Borgman, UCLA; G. Bowker, SCU; T. Finholt, UM; S. Jackson, UM;
 D. Ribes, Georgetown; S.L. Star, SCU)
 - Data Conservancy: OCI0830976, Sayeed Choudhury, PI, Johns Hopkins University.
 - Knowledge and Data Transfer: the Formation of a New Workforce. # 1145888. C.L. Borgman, PI; S. Traweek,
 Co-PI.
- Microsoft External Research: Tony Hey, Lee Dirks, Catherine van Ingen, Catherine Marshall
- Sloan Foundation: The Transformation of Knowledge, Culture, and Practice in Data-Driven Science: A Knowledge Infrastructures Perspective. # 20113194. C.L. Borgman, PI; S. Traweek, Co-PI. Joshua Greenberg, program director
- Project website: http://knowledgeinfrastructures.gseis.ucla.edu/index.html



