If Data Sharing is the Answer, What is the Question?

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Data sharing policies

- European Union
- U.S. Federal research policy
- Research Councils of the UK
- Australian Research Council
- Individual countries, funding agencies, journals, universities
Why Share Research Data?

• To reproduce research
• To make public assets available to the public
• To leverage investments in research
• To advance research and innovation
Lack of incentives to share data

• Rewards for publication
• Effort to document data
• Competition, priority
• Control, ownership

http://www.buildingsrus.co.uk/.../target1.htm
Data
Center for Embedded Networked Sensing

- NSF Science & Tech Ctr, 2002-2012
- 5 universities, plus partners
- 300 members
- Computer science and engineering
- Science application areas

Slide by Jason Fisher, UC-Merced, Center for Embedded Networked Sensing (CENS)
Engineering researcher:  
“Temperature is temperature.”

Biologist: “There are hundreds of ways to measure temperature. ‘The temperature is 98’ is low-value compared to, ‘the temperature of the surface, measured by the infrared thermopile, model number XYZ, is 98.’ That means it is measuring a proxy for a temperature, rather than being in contact with a probe, and it is measuring from a distance. The accuracy is plus or minus .05 of a degree. I [also] want to know that it was taken outside versus inside a controlled environment, how long it had been in place, and the last time it was calibrated, which might tell me whether it has drifted.."
Data are representations of observations, objects, or other entities used as evidence of phenomena for the purposes of research or scholarship.


http://www.genome.gov/dmd/nci晃node=Photos/Graphic晃id=85327
If Data Sharing is the Answer, What is the Question?

- Research Design, 2015-2018
- Methods
- Questions
- Findings
- Conclusions
- DANS study
- Recommendations
Research Design

• Goals
  – Explicate data, sharing, reuse, openness, infrastructure across scientific domains
  – Identify new models of scientific practice

• Dimensions
  – Mixtures of domain expertise
  – Factors of scale
  – Centralization of data collection and analysis
Qualitative Methods

• Document analysis
  – Public and private documents and artifacts
  – Official and unofficial versions of scientific practice

• Ethnography
  – Observing activities on site and online
  – Embedded for days or months at a time

• Interviews
  – Questions based on our research themes
  – Compare multiple sites over time
# Current Research Sites

<table>
<thead>
<tr>
<th>Domain</th>
<th>Focus</th>
<th>Topic</th>
</tr>
</thead>
<tbody>
<tr>
<td>Astronomy sky surveys</td>
<td>Place: sky and universe</td>
<td>Survey of night sky</td>
</tr>
<tr>
<td>Deep subseafloor biosphere</td>
<td>Place: under ocean floor</td>
<td>Microbial life and environment</td>
</tr>
<tr>
<td>Craniofacial research</td>
<td>Problem: Craniofacial birth defects in humans</td>
<td>Genomics of four model organisms</td>
</tr>
<tr>
<td>Computational science</td>
<td>Problem: Data analysis at scale</td>
<td>Computing platform for astronomy, physics, turbulence, soil science, genomics...</td>
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Research Question 1

How do the *mixtures of domain expertise* influence the collection, use, and reuse of data – and vice versa?

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Sloan Digital Sky Survey (SDSS-I/II)

- Survey from 2000-2008
- 160+ TB data total
- Tens of millions of dollars
- Open data
- Proprietary software

Telescope for the Sloan Digital Sky Survey, Apache Point, New Mexico
Large Synoptic Survey Telescope (LSST)

- Survey from 2022-2032
- 15 TB data per night
- 1+ Billion dollars
- Data open to partners
- Open source software

https://news.slac.stanford.edu/sites/default/files/images/image/lsst_h_0.jpg
Mixtures: Astronomy sky surveys

- Domains
  - Astronomy
  - Computer science

- Project characteristics
  - Mature discipline
  - Abundant data
  - Trusted archives
  - Shared tools, methods
  - Established infrastructure for data access and use
Center for Dark Energy Biosphere Investigations

Repository for seafloor cores. Photo: Peter Darch

International Ocean Discovery Program
lodb.tamu.org

- NSF Science & Tech Ctr, 2010-2020
- 35 institutions
- 90 scientists
- Biological sciences
- Physical sciences
Mixtures: Deep subseafloor biosphere

• Domains
  – Biological sciences
  – Physical sciences
  – 50+ self-identified specialties

• Project characteristics
  – Emergent scientific problem area
  – Scarce data
  – Disparate, exploratory methods
  – Building capacity for data collection
  – Sharing established infrastructures
FaceBase Consortium

- National Institute for Dental and Craniofacial Research
- Genetics, imaging data: craniofacial development
- 11 projects: clinical, biology, bioinformatics
- 4 model organisms: human, primates, mice, zebrafish
- Make data available on hub www.facebase.org
FaceBase Spokes and Hub

1 coordinating center
10 spokes

Photos: Irene Pasquetto
Mixtures: Craniofacial deformities

• Domains
  – Genomics, bioinformatics
  – Molecular, developmental biology
  – Dentistry, plastic surgery

• Project characteristics
  – Urgent medical problem
  – Species-specific data
    • Humans
    • Primates
    • Mice
    • Zebrafish
  – Competing tools, methods
  – Multiple established infrastructures
Research Question 2

What *factors of scale* influence research practices, and how?

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Scale factors

- Temporal
- Spatial
- Personnel
## Scale factors

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<th>Scale factors</th>
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<tr>
<td>Astronomy sky surveys</td>
<td>Uncertainty due to long temporal frame; paradigm shifts</td>
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<tr>
<td>Deep subseafloor biosphere</td>
<td>Scarce data are sparse data; high variety; difficult to standardize</td>
</tr>
<tr>
<td>Craniofacial research</td>
<td>High variety in genomes studied, models, methods, duration of analysis; difficult to standardize</td>
</tr>
<tr>
<td>Computational sciences</td>
<td>High variety in data, methods, tool expertise; difficult to standardize</td>
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Research Question 3

How does the degree of centralization of data collection and analysis influence use, reuse, curation, and project strategy?

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## Centralization factors

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<tr>
<td>Astronomy sky surveys</td>
<td>Centralized data collection and initial processing; decentralized use and analysis</td>
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<tr>
<td>Deep subseafloor biosphere</td>
<td>Common data source, shared repositories of cores; decentralized analysis</td>
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<tr>
<td>Craniofacial research</td>
<td>Decentralized data collection; efforts to integrate data for centralized analysis reveal lack of commonalities</td>
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Conclusions so far

• General
  – Data sharing is not one problem, but many
  – Factors interact: domain mixtures, scale, centrality

• Research themes
  – Domains consist of subdomains with fluid boundaries
  – Volume might be least important scale factor
  – Centrality contradictions
    • Centralized data collections become decentralized in analysis
    • Decentralized data collections are hardest to integrate for analysis
DANS promotes **sustained access** to digital research data files and encourages researchers to **archive** and **reuse** data.

**ARCHIVING**
Deposit your datasets in EASY or send research data and publications to NARCIS.

**REUSE**
Find datasets, publications, researchers, projects and institutions via NARCIS and EASY.

**TRAINING & CONSULTANCY**
Let DANS advise you on data management and certification of digital archives.

**SPOTLIGHT**
DANS and Inria sign the MoU for Software Heritage

DANS will collaborate with Inria (France) on the development of the Software Heritage Initiative. On June 30 Inria officially announced the collaboration with DANS with the launch of the Software Heritage website.

**NEWS**
Your data paper in Research Data Journal for the Humanities and Social Sciences

RDJ, published by Brill publishers and DANS, is a peer reviewed e-only open access journal. Authors can submit their data papers online.
Why do users put data in DANS?

• Meet legal requirements
• Preserve data for long term
• Get credit for data
• Control access to data
• Use as background service
• Motivate citizen science participants

Recommendations so far

- Identify practices of subdomains and interactions
- Seek right level of abstraction for data sharing, integration, curation, reuse
- Invest in data curation early in project design
- Promote infrastructure solutions
  - Shared tools and services
  - Data discovery mechanisms
  - Iterative stewardship
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