#### Practical information

- 15 minutes breaks between sessions
- Remember to mute when listening
- Use chat actively during sessions
- Use "raise hand" functionality during discussions or to indicate wish to ask questions
- No recording of sessions are planned
- Working groups are organised using the breakout session functionality in Zoom
  - Participants are allocated to sessions by host
- Supposed to be an interactive course!
- Picture of all participants wanted, will start with this



# Motivation: Why do we need data management?

Øystein Godøy

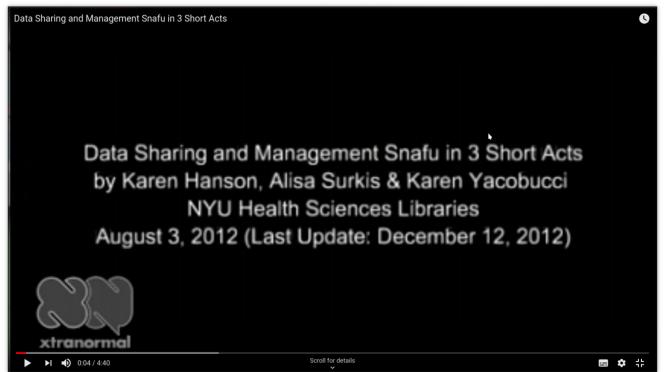


### **Outline**

- Data Sharing and Management Snafu in 3 Short Acts
  - https://www.youtube.com/watch?v=N2zK3sAtr-4
- Why do we need data management?
- Science life cycle/Data life cycle
- How to change data sharing culture.
- What are the FAIR data principles?
  - How do they help with good data management?
- External boundary conditions by funding agencies and publishers.
- Scientific data as service.
- Data management plan.



# Data Sharing and Management Snafu in 3 Short Acts



https://www.youtube.com/watch?v=N2zK3sAtr-4

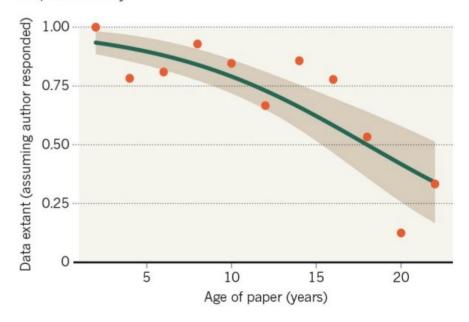
NorDataNet

### Why do we need data management?

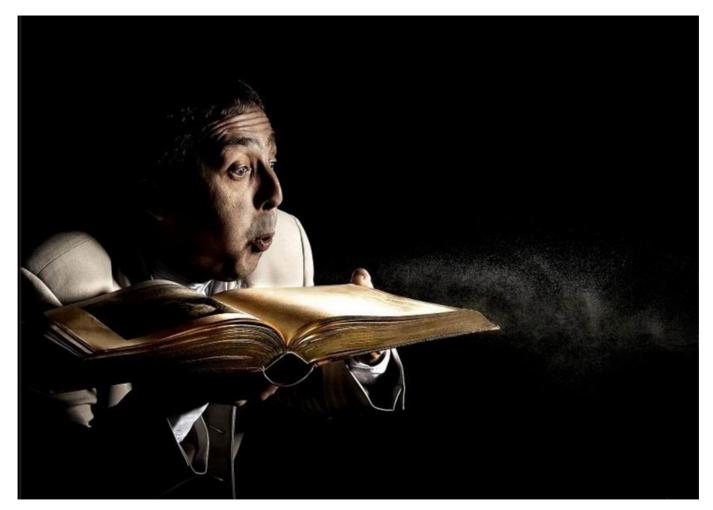
- Loosing scientific data
  - Decline can mean 80% of data are unavailable after 20 years.
    - Gibney and Van Noorden (2013), Nature

#### MISSING DATA

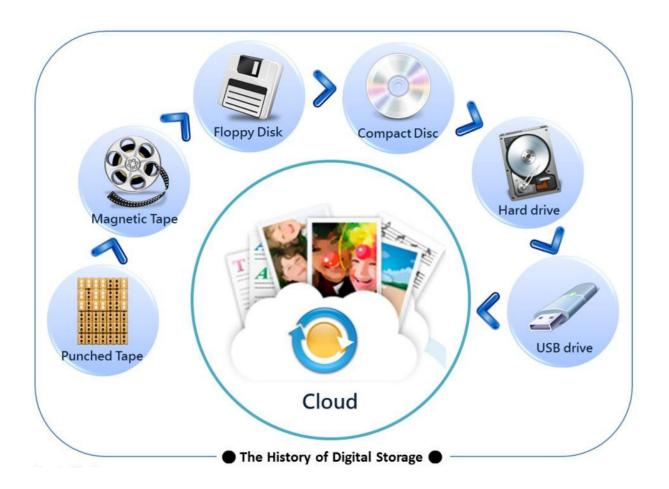
As research articles age, the odds of their raw data being extant drop dramatically.













## Why bother with structured data management?



#### Benefits

- Maximise public investment in data collection and production
- Promote scientific collaboration
- Promote interdisciplinary science
- Promote scientific transparency
- Leave a legacy
- Science paradigms
  - according to Jim Gray
  - empirical science
  - theoretical science
  - computational science
  - data exploration science



### Why share data?

- Research sponsor require it
  - recognition as an authoritative source and wise investment
- Quality control
  - improved data quality due to expanded use, field checks, and feedback
- Improved visibility
  - improved connections to scientific network, peers, and potential collaborators
- Journals require it
  - Reproducible research
- Far upstream sponsors require it



CC image by SLU Madrid Campu on Flickr



## Making Your Research Easier and Cheaper

The 5 P's matter!

Prior

Planning

Prevents

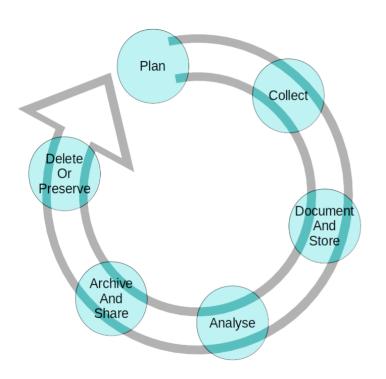
Poor

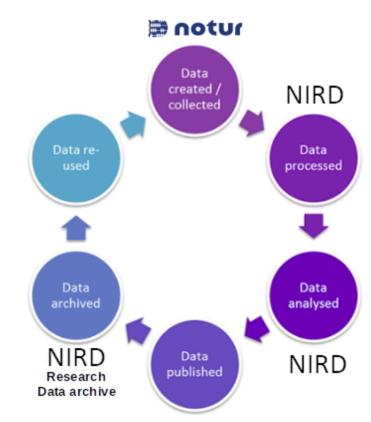
Performance!





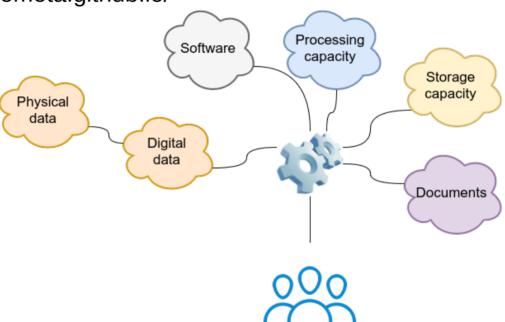
## Science life cycle/Data life cycle







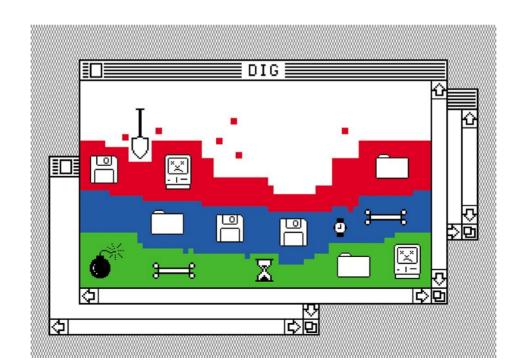
https://codemeta.github.io/





# Challenge to scientists: does your ten-year-old code still run?

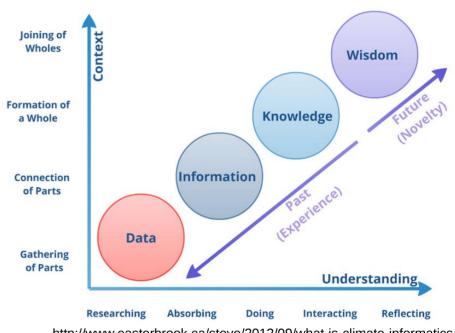
https://www.nature.com/articles/d41586-020-02462-7





#### DIKW chain

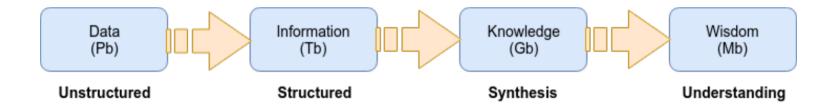
- How to transition from data to knowledge and understanding...
  - The illustration is a common redrawing of Russ Ackoff "From Data to Wisdom"
    - Journal of Applied Systems Analysis, Volume 16, 1989 p 3-9
- DIKW is necessary to
  - Take care of data for the future
  - Ensure data is the basis for knowledge
    - Now and in the future
  - Knowledge based management depends on national, regional and global interaction



http://www.easterbrook.ca/steve/2012/09/what-is-climate-informatics/

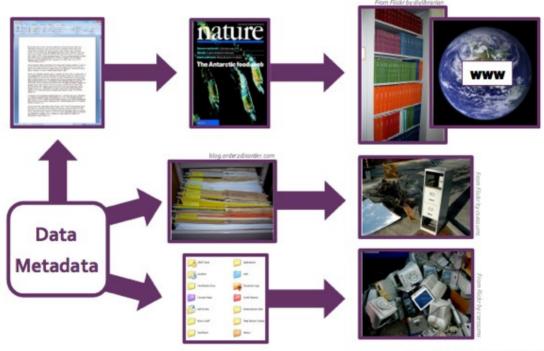


### **DIKW** chain





## The reality today





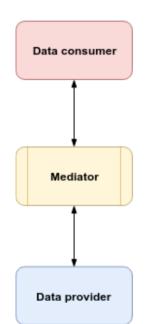


## How to change data sharing culture?





### Bridging between data management actors



#### Unwilling

- Do not want to change behaviour, existing tools have worked well.
- · Want to continue as before.
- Does not see the benefit of standardisation, until explicitly explained/demonstrated or through new

#### Willing

- · Wants to translate between provider and consumer.
- Still relies on some sort of standardisation in order to be cost effective.
- Must know dimensions, structures, content, missing values, units, aggregation levels, ...

#### Unwilling and skeptical to potential users

- Do not want to change behaviour, legacy system(s).
- Want to continue as before.
- Understands own requirements (knows the data well).

- A prerequisite for efficient data sharing across communities is application of proper metadata and standards
- Although standards exist, they are often not used by data providers who doesn't see the benefit
- Lacking understanding for the importance of use metadata
  - Enabling reuse across communities and generations
  - Lacking understanding for the importance of semantic standardisation
- Need a business model crediting all involved parties
  - Scientists, institutions, data centres, ....
- It is about leaving a legacy



# The FAIR Guiding Principles for scientific data management and stewardship

#### To be Findable:

- F1. (meta)data are assigned a globally unique and persistent identifier
- F2. data are described with rich metadata (defined by R1 below)
- F3. metadata clearly and explicitly include the identifier of the data it describes
- F4. (meta)data are registered or indexed in a searchable resource

#### To be Accessible:

- A1. (meta)data are retrievable by their identifier using a standardized communications protocol
- A1.1 the protocol is open, free, and universally implementable
- A1.2 the protocol allows for an authentication and authorization procedure, where necessary
- A2. metadata are accessible, even when the data are no longer available

#### • To be **Interoperable**:

- I1. (meta)data use a formal, accessible, shared, and broadly applicable language for knowledge representation
- I2. (meta)data use vocabularies that follow FAIR principles
- 13. (meta)data include qualified references to other (meta)data

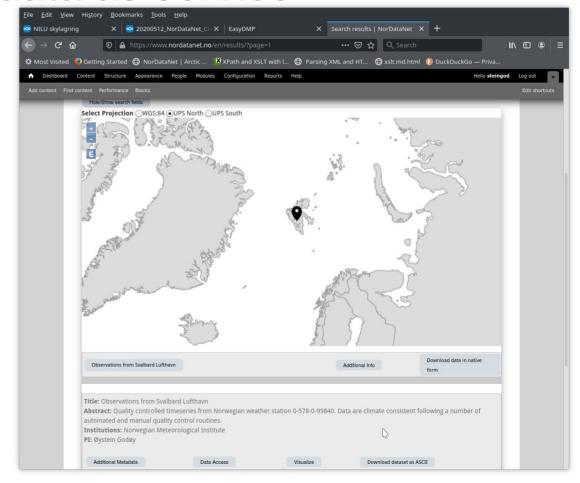
#### To be Reusable:

- R1. meta(data) are richly described with a plurality of accurate and relevant attributes
- R1.1. (meta)data are released with a clear and accessible data usage license
- R1.2. (meta)data are associated with detailed provenance
- R1.3. (meta)data meet domain-relevant community standards



### Scientific data as service

- From science to service
  - Goddard, Science 23
    Sep 2016: Vol. 353,
    Issue 6306, pp. 13661367 DOI:
    10.1126/science.aag308





### Data management plans

A data management plan (DMP) is a written document that describes the data you expect to acquire or generate during the course of a research project, how you will manage, describe, analyze, and store those data, and what mechanisms you will use at the end of your project to share and preserve your data.



### Data management plans

- Are increasingly being required by funding agencies
  - e.g. RCN, EU
- And by e-infrastructure providers
  - e.g. Sigma2



