Research Data Infrastructures
- Challenges, Desires, Incentives -

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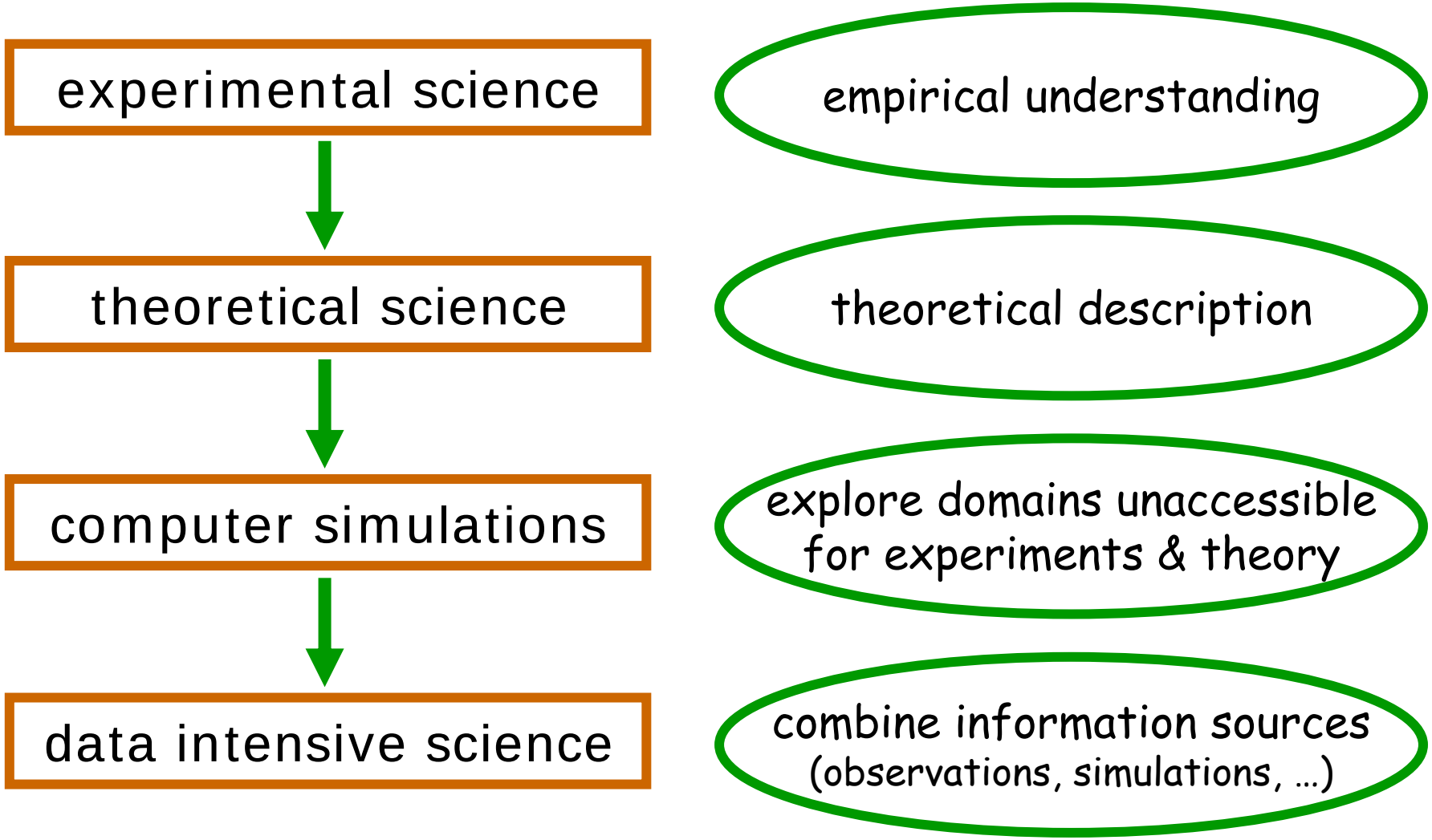
Wikipedia‘s definition

- **Data infrastructure** is a digital infrastructure promoting data sharing and consumption.

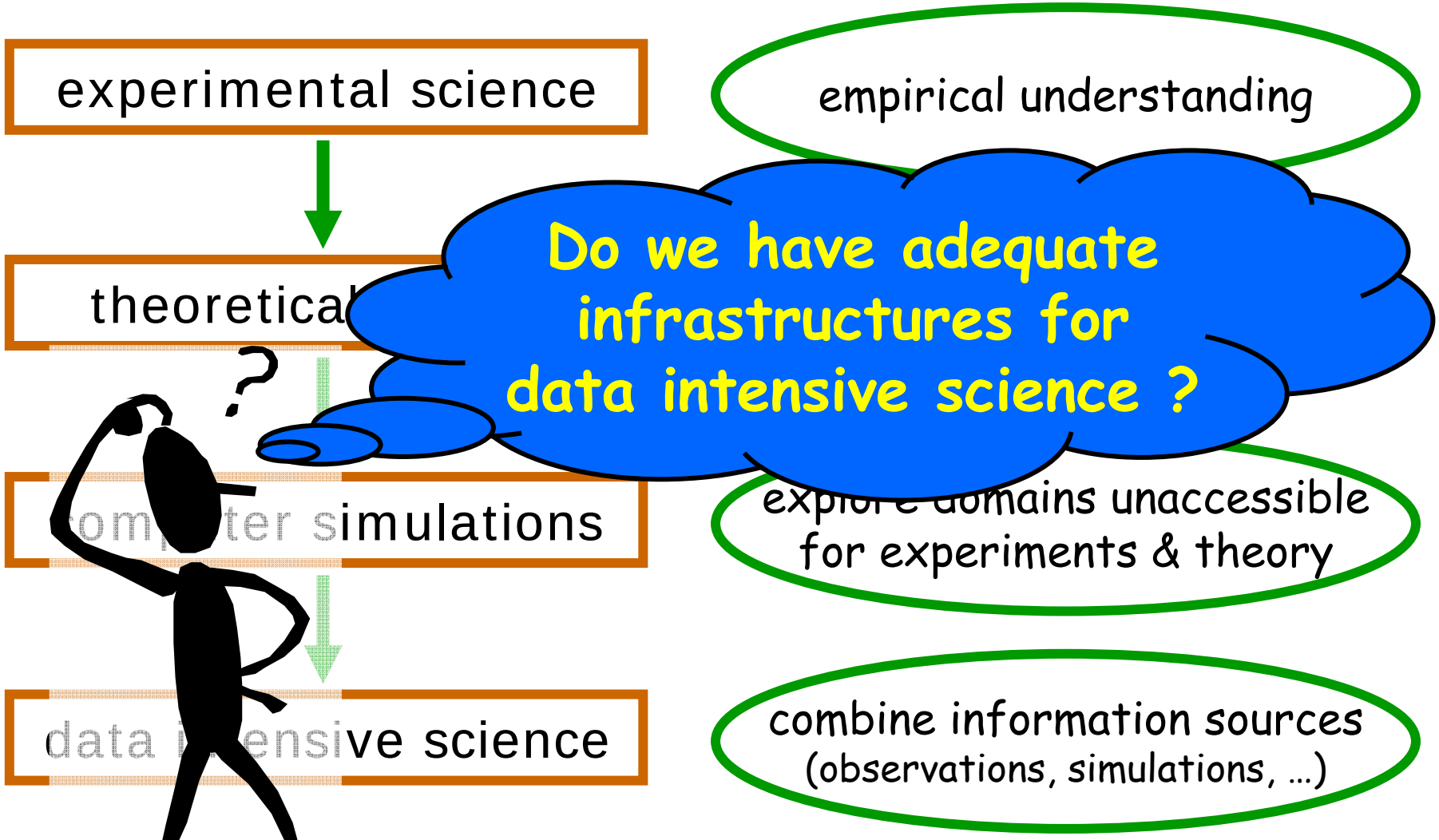
  Similarly to other infrastructures, it is a structure needed for the operation of a society as well as the services and facilities necessary for an economy to function, the data economy in this case.

Paradigms of science



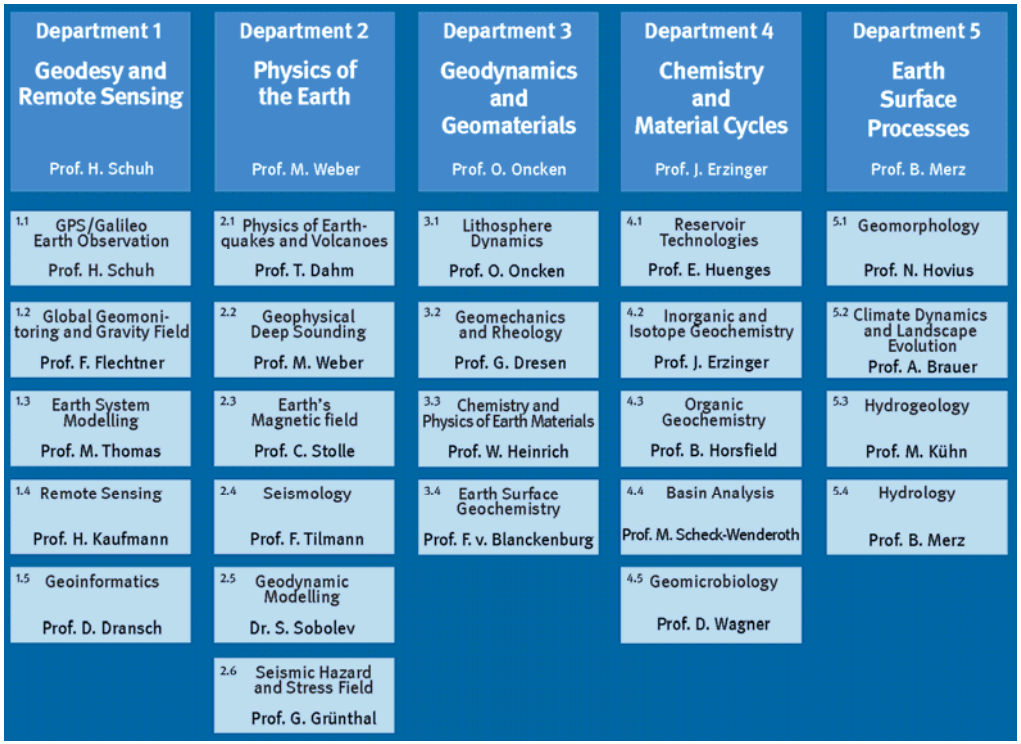
after Bell et al. (Science, 2009)

Paradigms of science



after Bell et al. (Science, 2009)

Example:
German Research Centre for Geosciences (GFZ)

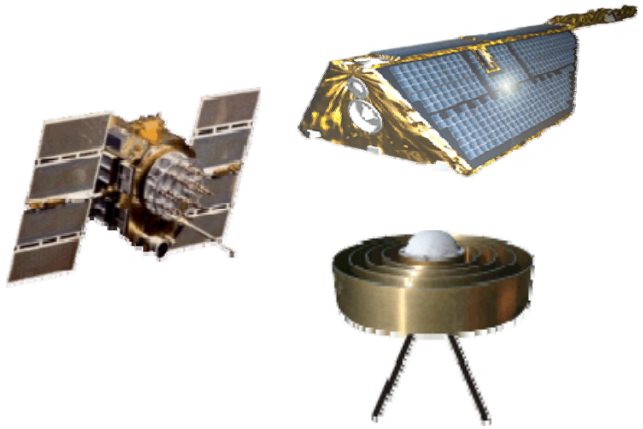


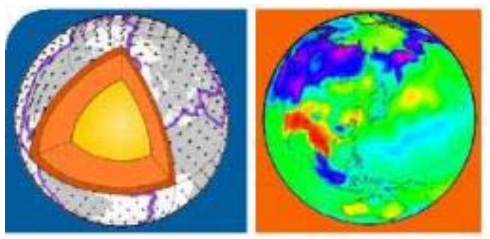
- **5 research areas (Departments) each with 4-6 sections and more than 1000 employees**
- **several research projects in every section**
- **more than 100 different data sources**

Example: Data sources at GFZ Potsdam

Many data sets come from …

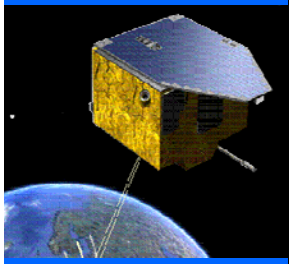
- **Earth monitoring systems**
  - generate large amounts of data
  - mostly homogeneous structures
  - partly automated workflows
- **Modelling systems**
  - produce medium-sized data amounts
  - partly homogeneous structures
  - but mostly no supporting workflows
- **Labs and field observations**
  - in most cases only small data amounts
  - but heterogeneous structures
  - data sets often generated in ad-hoc created structures





The Earth system: Monitoring, Analysis, Simulation

Hyperspectral mission

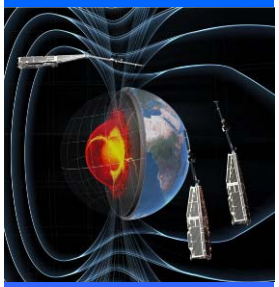


EnMAP

Terrestrial networks

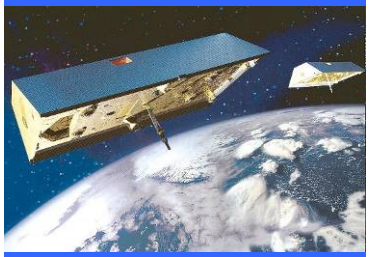
Tide gauge GPS SG Geophone

Magnetic field



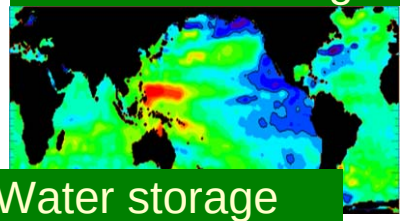
SWARM

Gravimetry missions

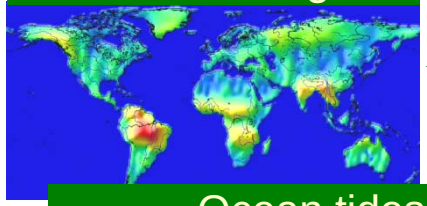


GRACE-FO

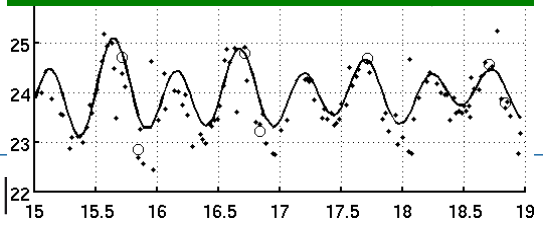
Sea level change



Water storage

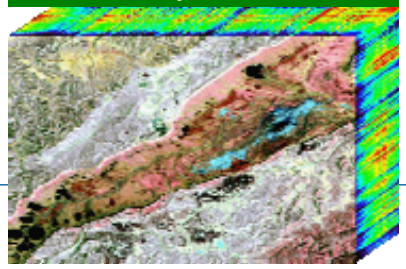


Ocean tides

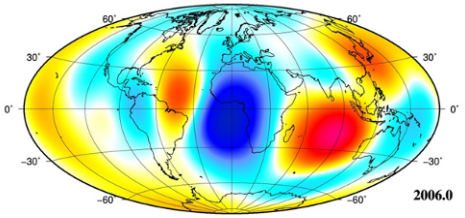


Integration, combination, assimilation

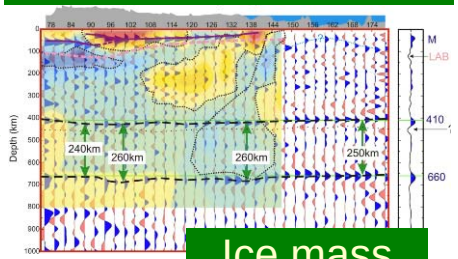
Surface parameters



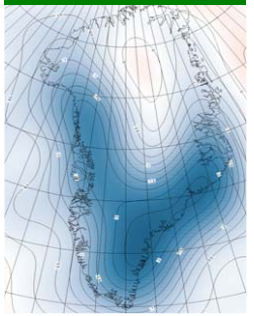
Magnetic field variations



Lithospheric structure



Ice mass change



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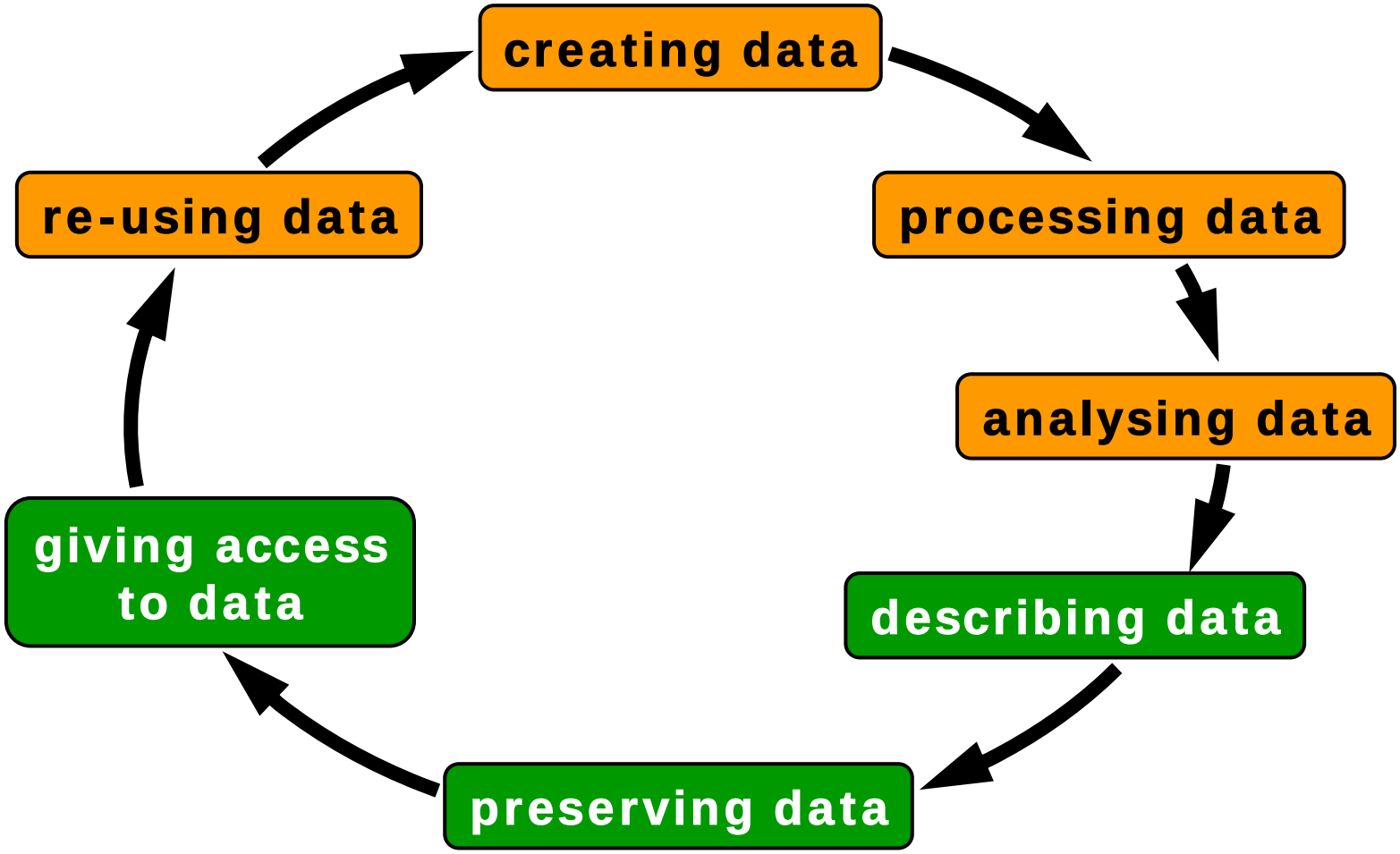
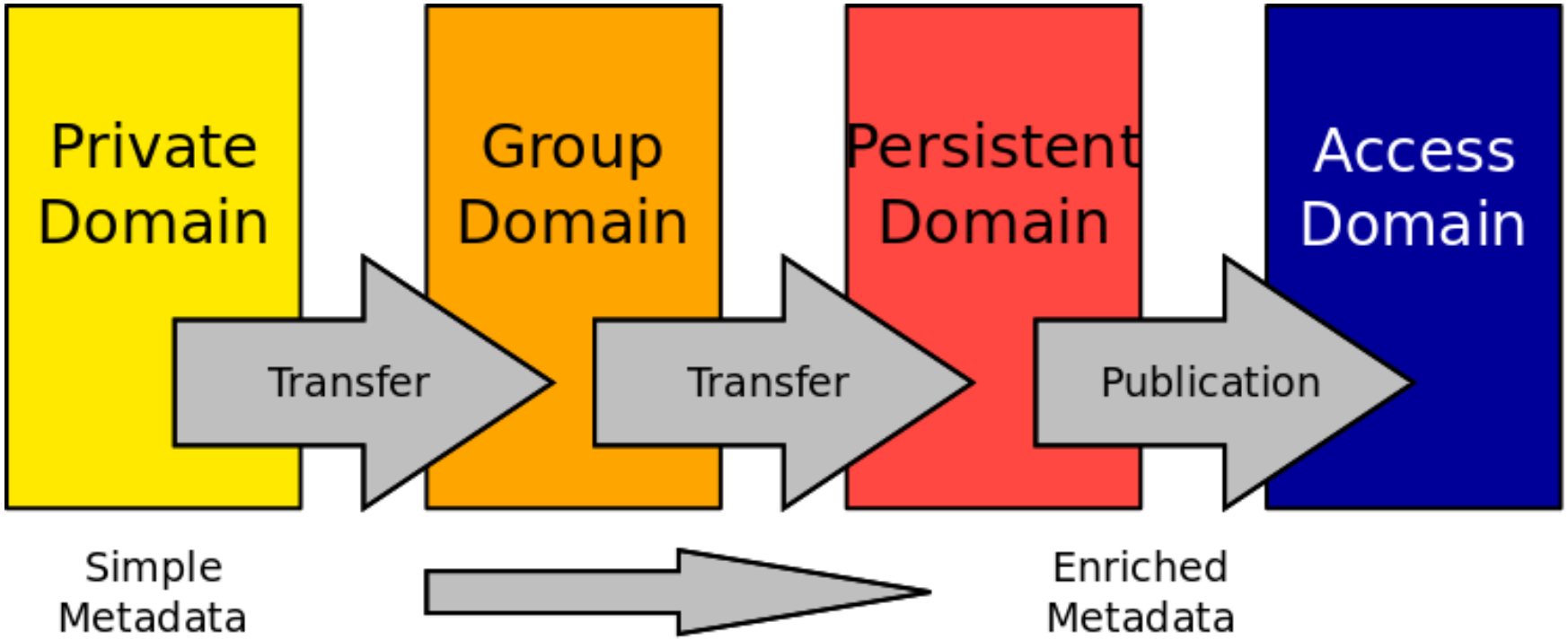
Data Life Cycle Model



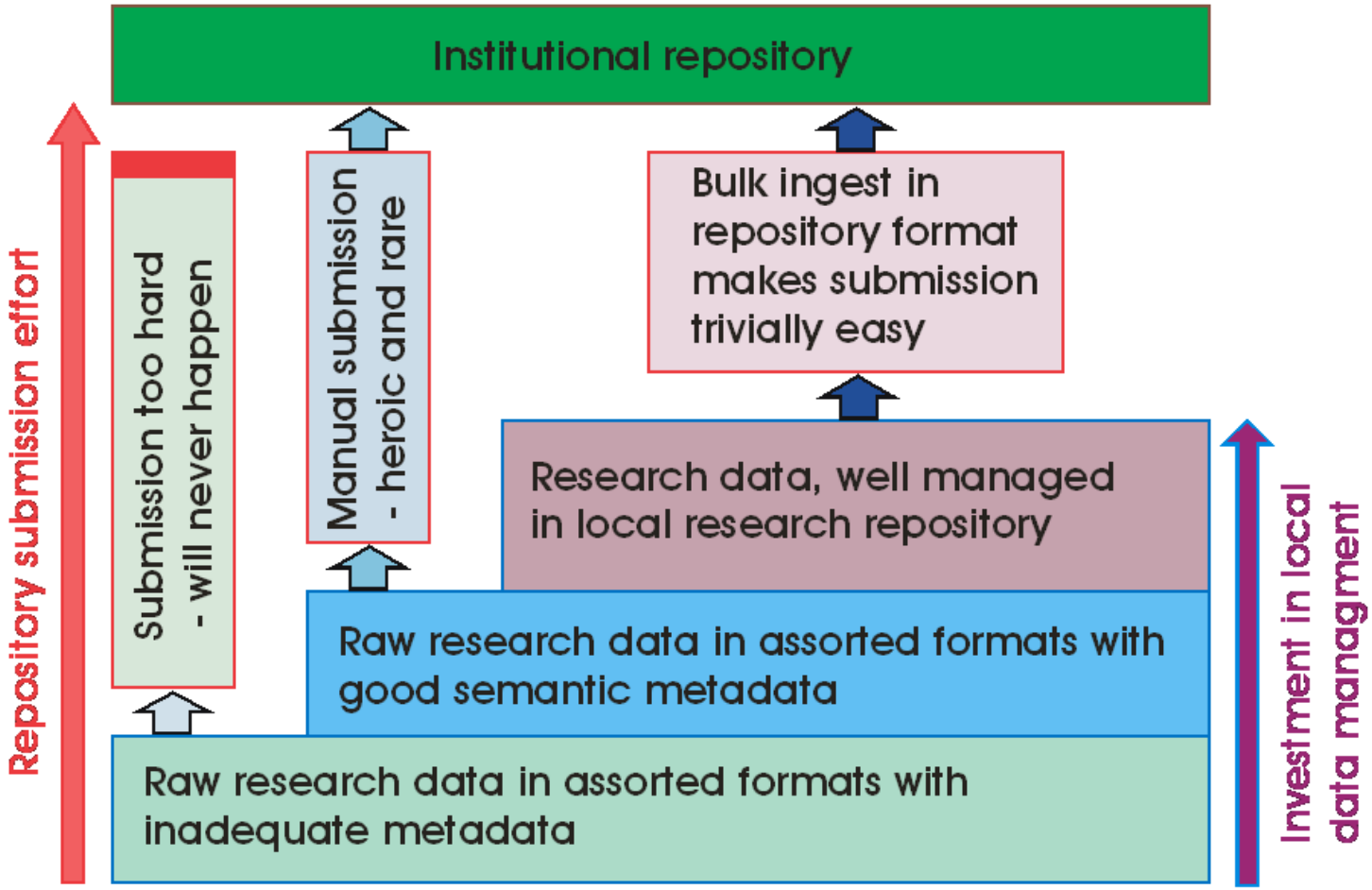
… defines demands on data management !

Data Life Cycle Model: From private to public



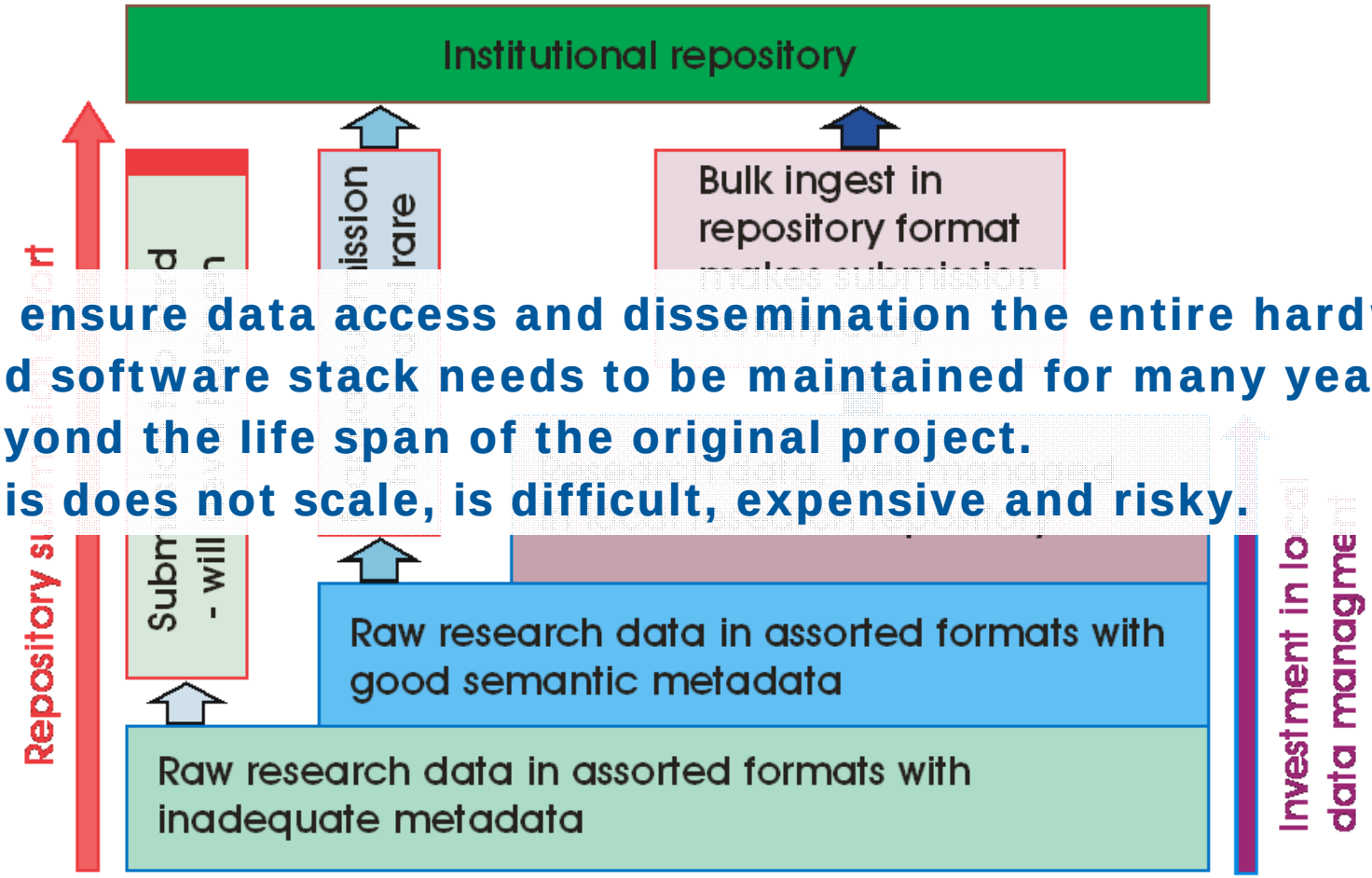
↳ **How to provide research data sets to external users ?**

Submitting data to repositories



D. Shotton, opencitations.wordpress.com

Submitting data to repositories



- To ensure data access and dissemination the entire hardware and software stack needs to be maintained for many years beyond the life span of the original project.
  This does not scale, is difficult, expensive and risky.

D. Shotton, opencitations.wordpress.com

An isolated application: Information System and Data Center (ISDC)

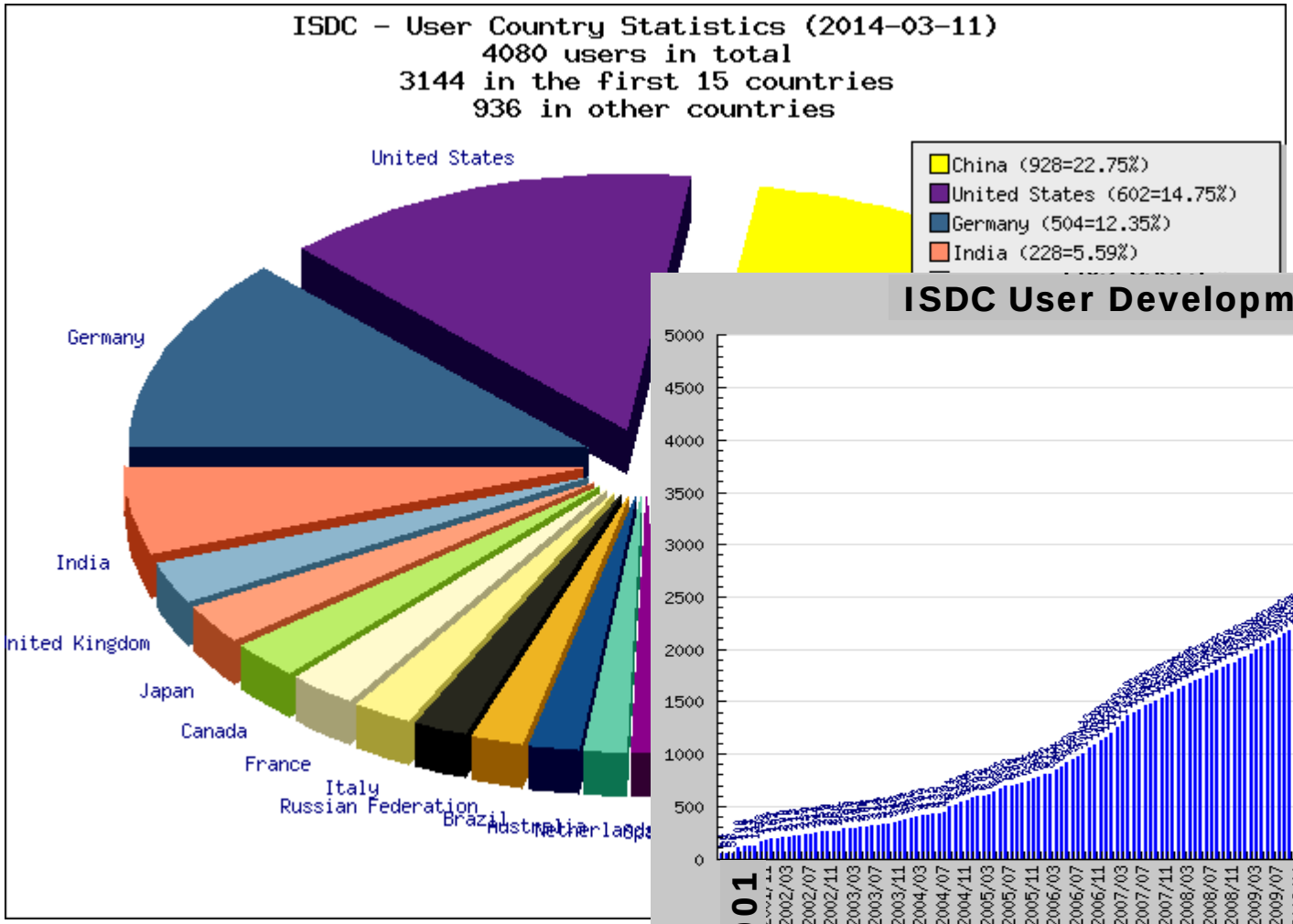
Gravity fields & models

terrestrial gravimetry

GPS data, Galileo-Service

- various data products with corresponding meta data, error estimates
  FAQs and user support, assigned to major research projects

ISDC: Statistics



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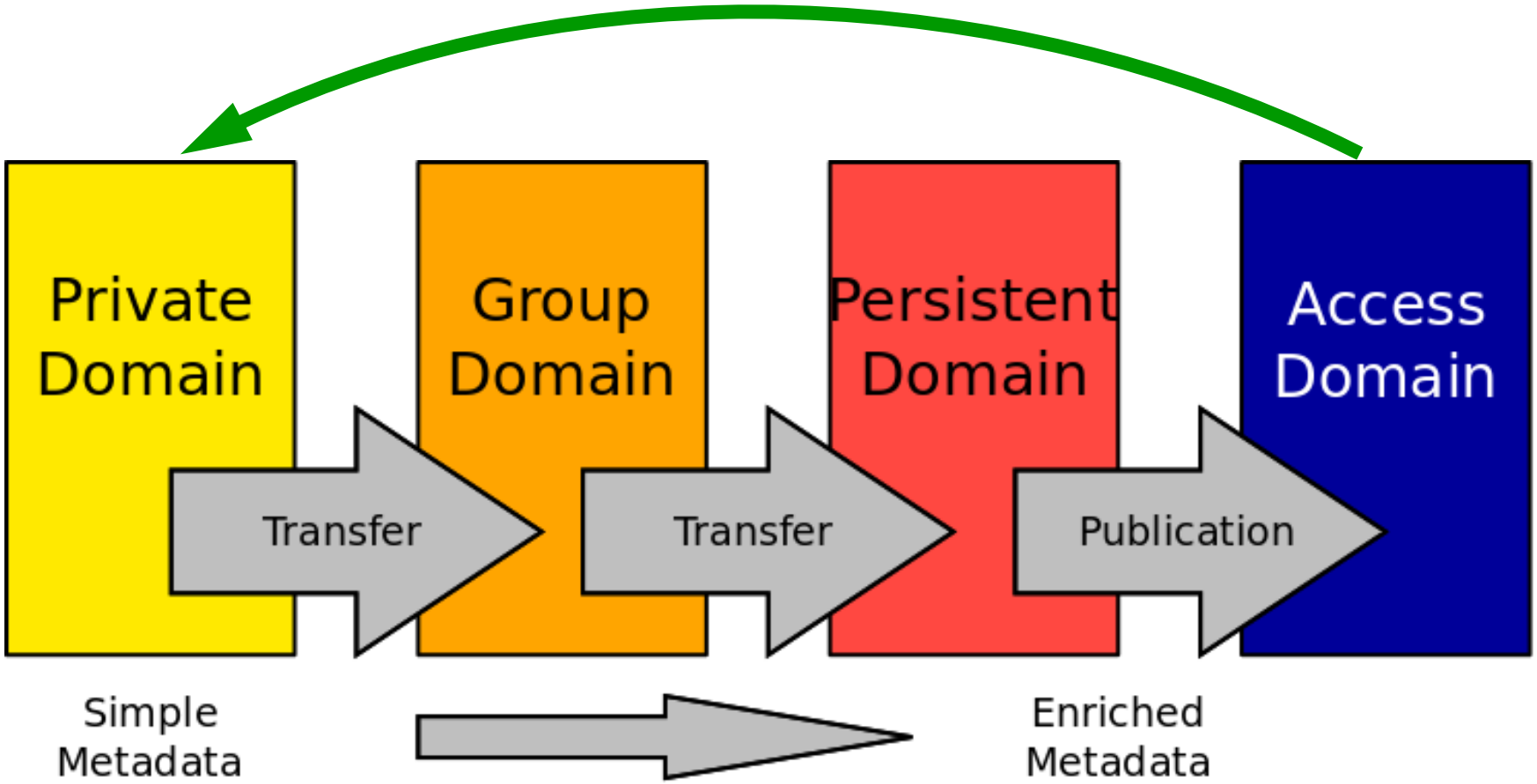
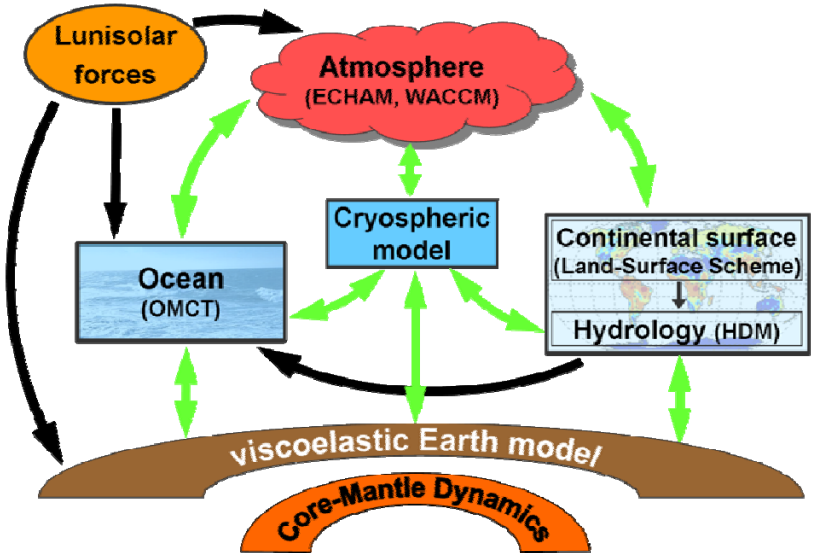
Data Life Cycle Model



➧ **What about the integration of data from neighboring research areas ?**

From multi- via inter- to transdisciplinarity

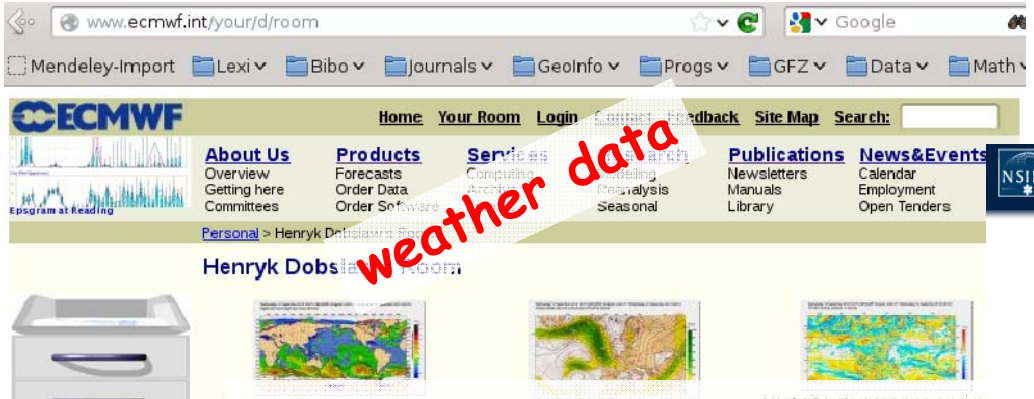
- **differentiation of disciplines** due to scientific specialization over the last decades
- significant **scientific progress** is presently expected from research across specialized disciplines
- recent **IT developments** principally provide excellent conditions **for rapprochement of scientific disciplines** (and research cultures ?)



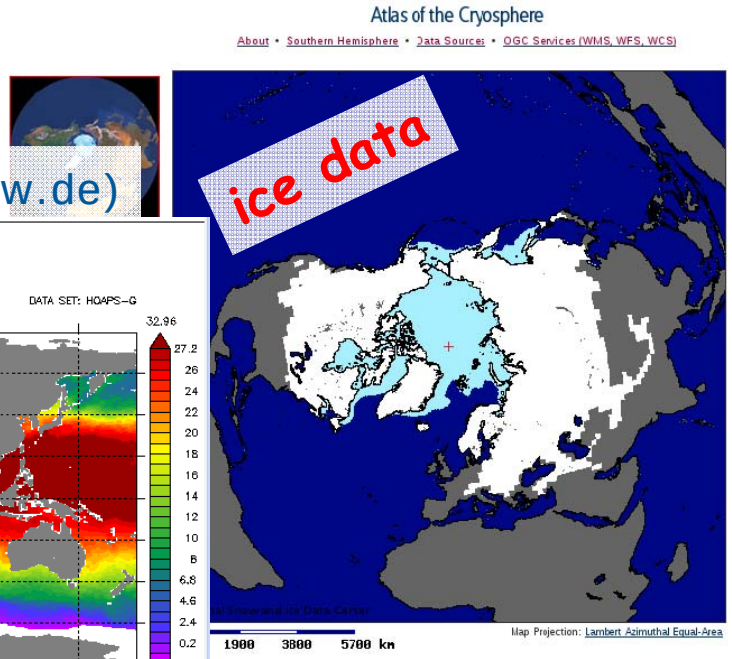
Example:
climate change and its socio-economic impacts

The geoscientific archipelago

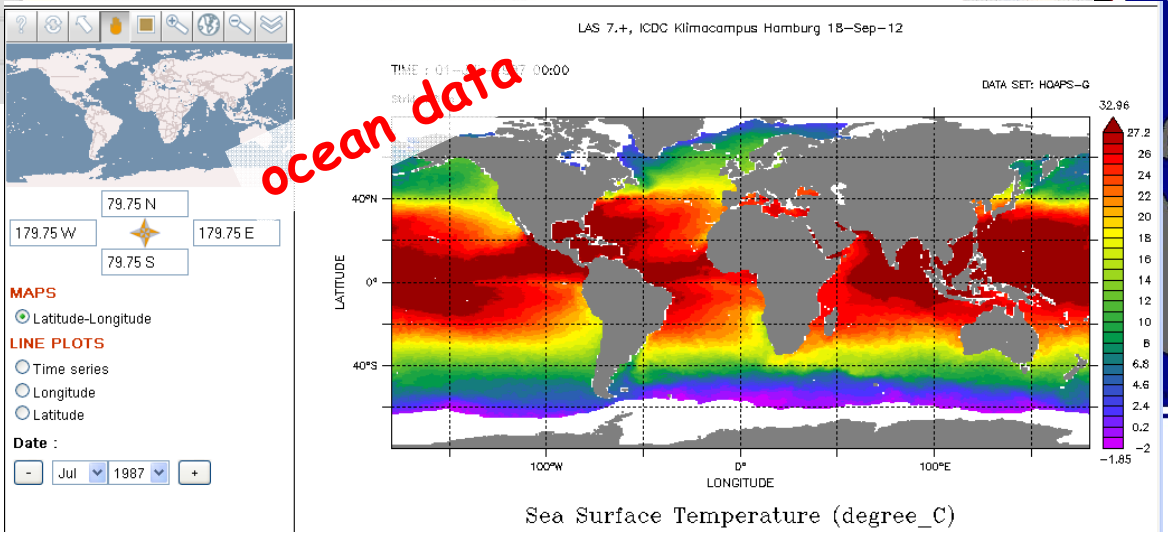
MARS @ ECMWF (www.ecmwf.int)



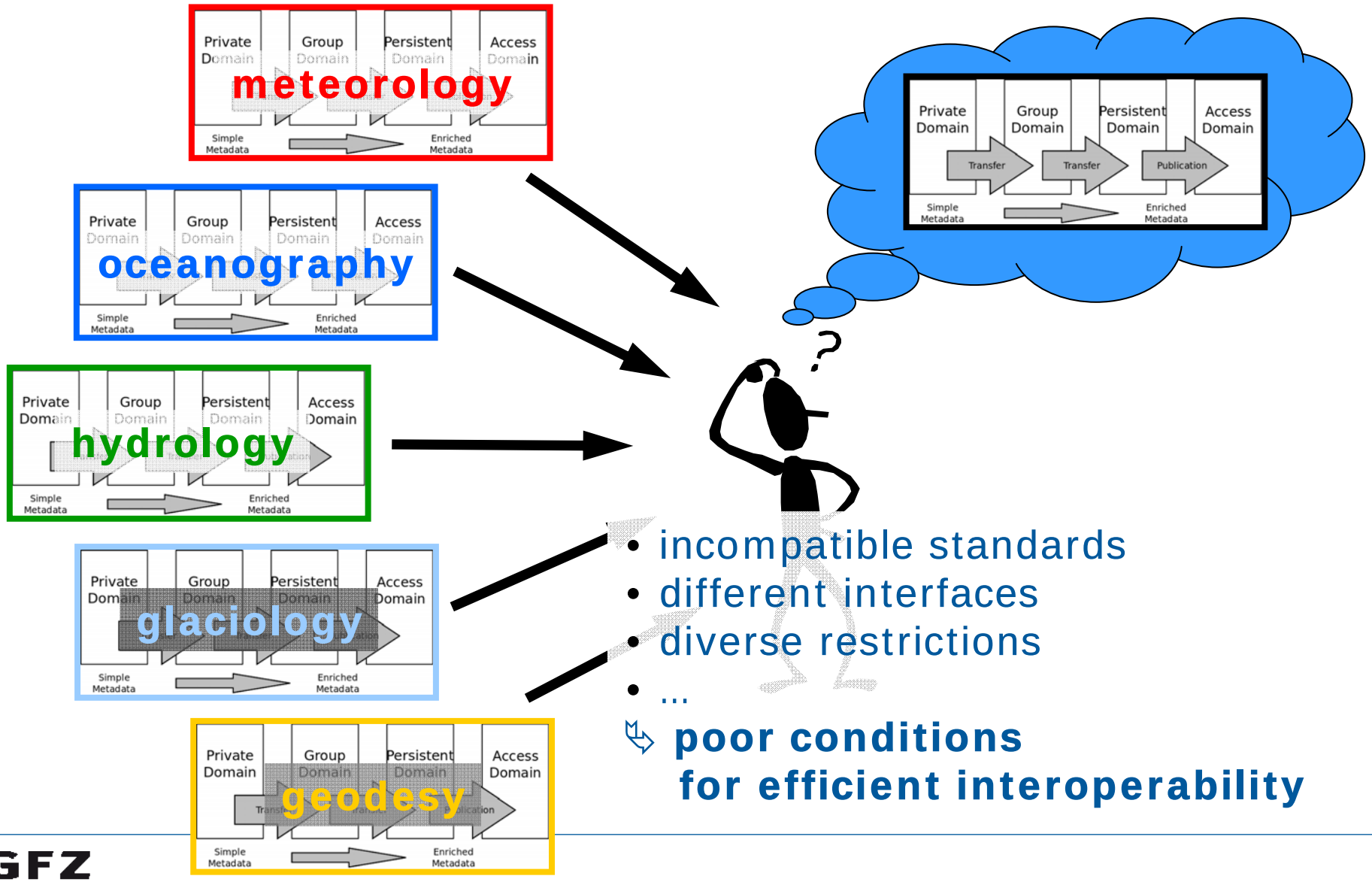
NSIDC (www.nsidc.org)



ICDC @ ZMAW Hamburg (icdc.zmaw.de)



Many communities and even more „solutions“ …



- incompatible standards
- different interfaces
- diverse restrictions
- …

↳ **poor conditions for efficient interoperability**

Expected Results?

EGS Abstract for
Nice, 2000

accepted !

A NEW WAY OF PRODUCING AND ANALYZING INCO[illegible] TENT, IRRELEVANT AND INCOMPREHENSIBLE DATA

[redacted] WWU Münster, D 48149 Münster, Germany.
[redacted]i-muenster.de/Fax: +49 251 83 39083

The ways by which data are produced are sometimes not fully understood. However, data may sometimes bear no significance at all, especially with respect to coffee making or egg frying ([redacted] *et al.*, 1974). In this paper, we present a new method to produce and analyze data that are inconsistent, irrelevant and incomprehensible.

To produce a set of irrelevant data we sampled the distribution of frozen rabbit droppings in front of our institute as cometary analogues. However, due to the climatic peculiarities of the sampling site, none of the samples collected during the period of July 2nd through August 3rd contained the necessary quantities of water ice. The droppings distribution was then analyzed as a function of space and time by means of the Hazelnut transform, with no understanding of the production process whatsoever. These data may have been inconsistent with the findings of other researchers we have probably never heard of. As a method for analyzing these irrelevant data points, [redacted] (1999) recently proposed a method applying the standard deviation of curvatures in clustered data points as a measure for quantifying an arbitrary line. We cunningly extended this method by inventing an adaptation technique for a semi–empirical line–drawing law for almost any distribution of data points on a double logarithmic scale.

Abstracts to be submitted on or before December 15, 1999 to

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http://www.copernicus.org/EGS/EGS.html

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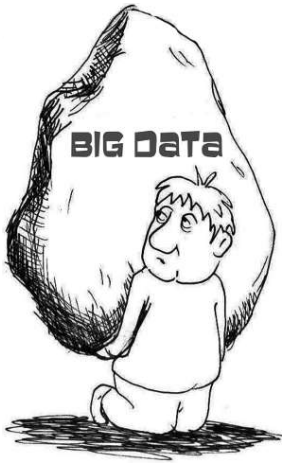
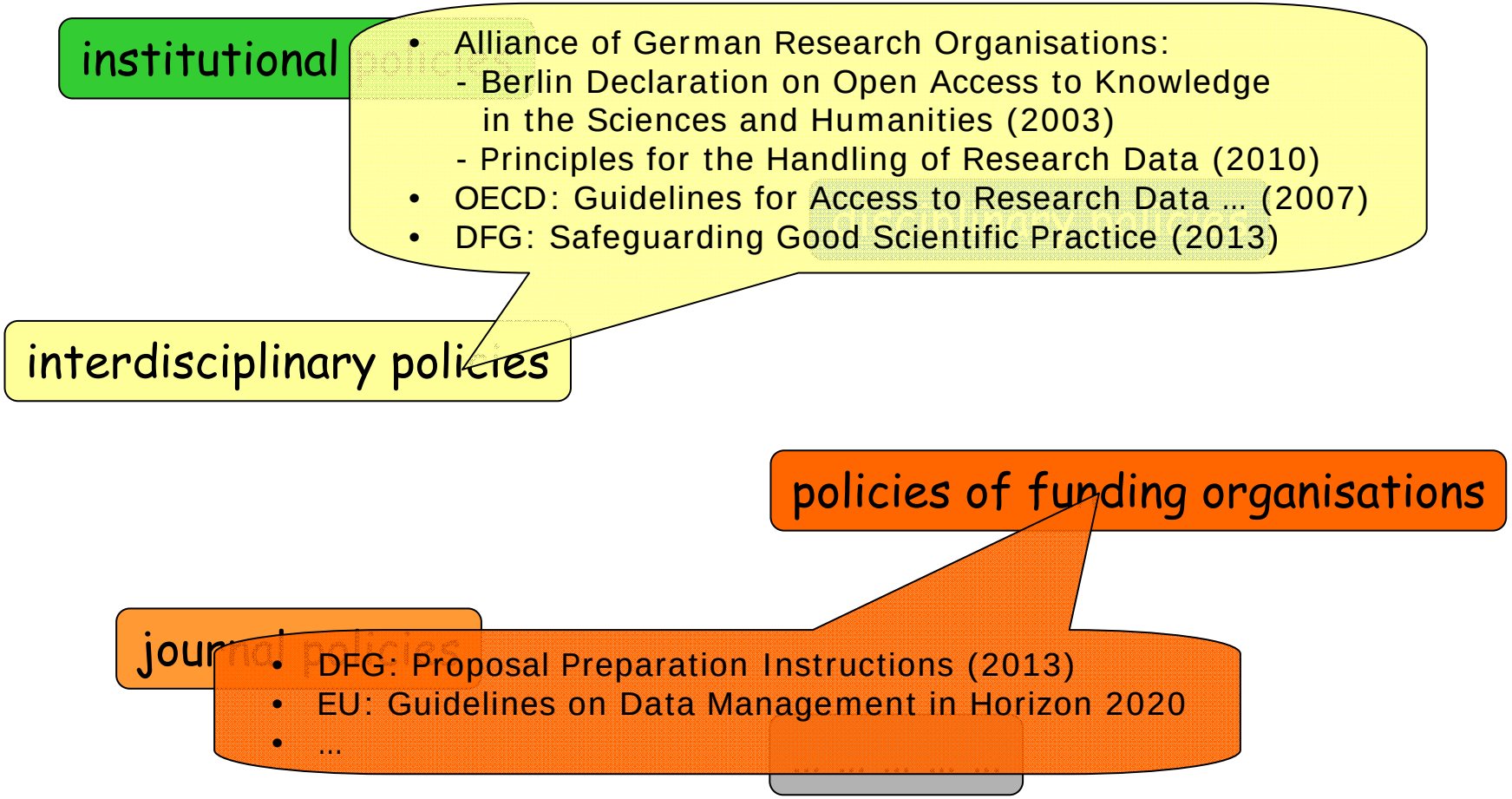
Data Science: From Big to Small

- **Usually:** Rules for data handling focus on large data sets, i.e., *„Big Data Science“*.
  ↳ Existing data management due to regulations of funding bodies.
- „**Big Data Science**“: large data amounts, homogeneous structures
- **But:** Vast investments of time (and brain work !) to generate an immense number of small data sets,
  i.e., *„Small Science Data“*.
  ↳ Systematic data management missing !
- „**Small Data Science**“: small data amounts, heterogeneous structures



(P. Vettiyattil, 2012)

Policies


institutional
• Alliance of German Research Organisations:
- Berlin Declaration on Open Access to Knowledge in the Sciences and Humanities (2003)
- Principles for the Handling of Research Data (2010)
• OECD: Guidelines for Access to Research Data … (2007)
• DFG: Safeguarding Good Scientific Practice (2013)
interdisciplinary policies
policies of funding organisations
• DFG: Proposal Preparation Instructions (2013)
• EU: Guidelines on Data Management in Horizon 2020
• …

Example: Policy at GFZ

- DFG-Rules of Good Scientific Practice (GSP) are standing instructions.
- Helmholtz-Association supports Open Access initiative (1st signee of Berlin Declaration!).
- GFZ's publication policy includes GSP and Berlin Declaration on Open Access and is part of the employment contract.

↳ Since GSP and Open Access are components of employment contracts, **data should be accessible.**
But …

The spirit is willing …

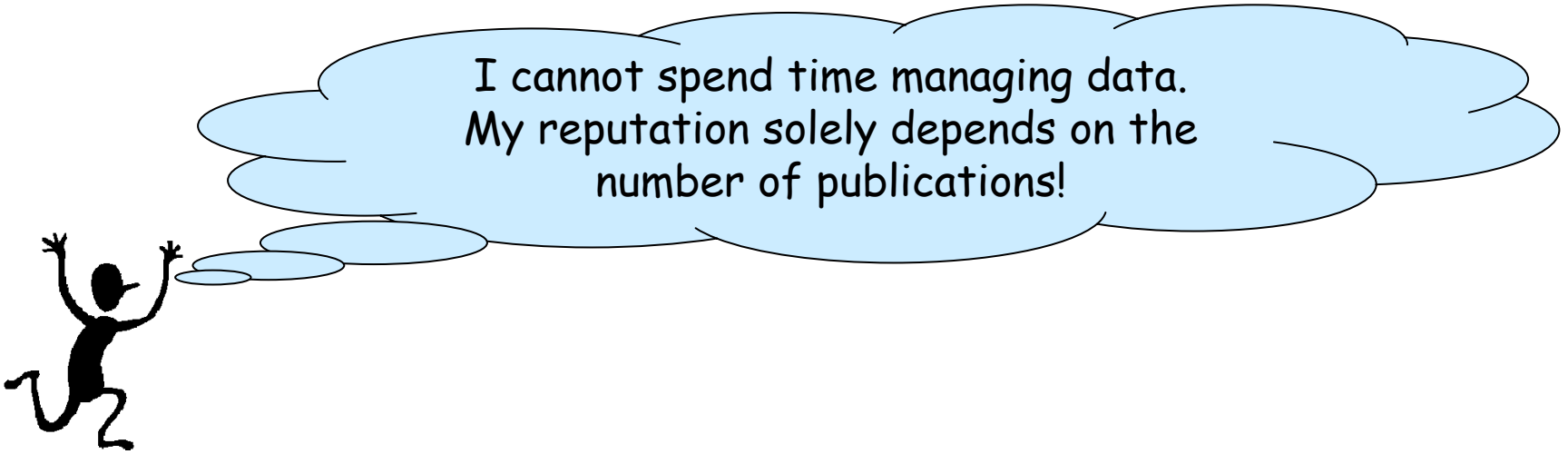
How can responsible data handling be established ?

- Imposition of sanctions fail in this context.
  ↳ Incentives are indispensable for realization !

- Implementation of policies requires:
  - Organization of scientific workflows
  - Tools allowing integration into scientific workflows

- Currently, data handling is not a primary task of scientists.
  For the individual scientist responsible **data handling is more trouble than it‘s worth !**

Trusting the cat to keep the cream …



On the other hand:

Every day several thousand eBay users add metadata to objects - and they do it correctly.

↳ **Are the tasks of players in data life cycles correctly assigned?**

Research data handling ...

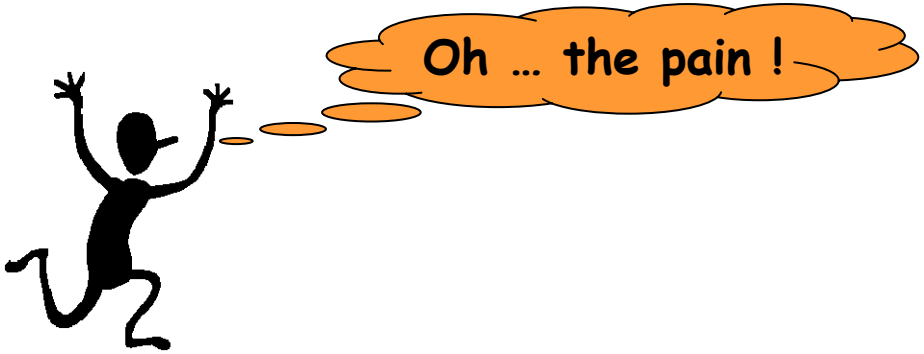
... is like driving a flock of cats over plains.

Example GFZ:

- ca. 150 simultaneous research projects
- annual fluctuation of ...
  ... research projects: ≈ 30
  ... employees: ≈ 200

↳ **We need generic tools !**

Describing data



- Content useless without description
- How can I get metadata without sprawling „data bureaucracy“ ?

- Collect data descriptions (semi-)automatically as part of the workflow

Why does it work in other areas?

- Ebay:
  - In order to sell goods people spend much time on describing metadata.
  - motivation: money

- Facebook:
  - People are willing to make vast amounts of personal data public.
  - motivation: self-expression, networking

- Wikipedia:
  - People write and edit articles on specific topics without being acknowledged as authors.
  - motivation: prospect of becoming a Wikipedia administrator

… sometimes it also works in the scientific world:

Institutional Workflows

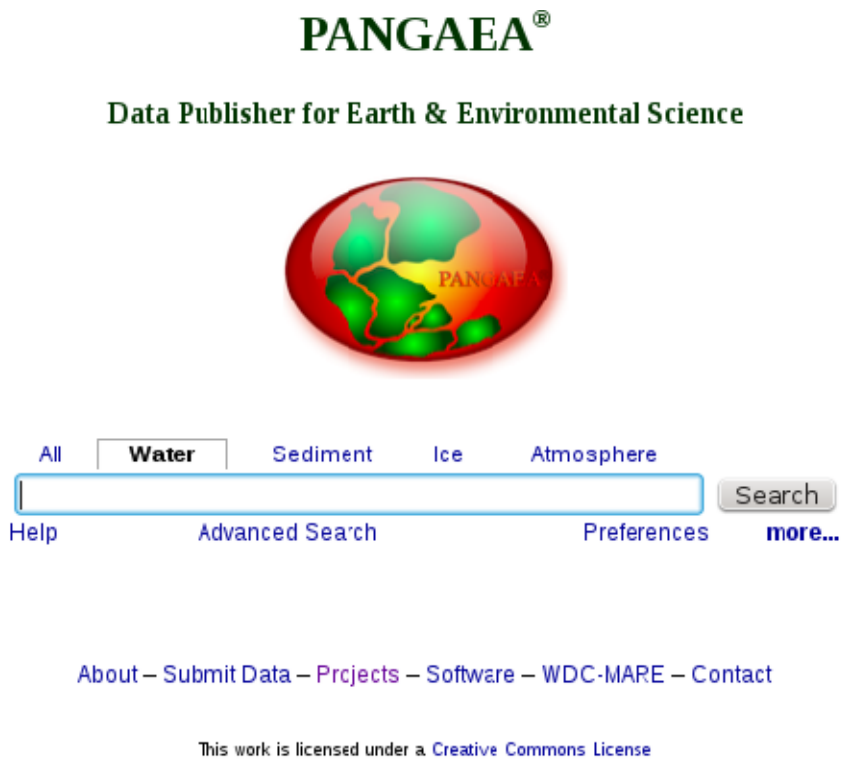
Examples:

ZALF (Leibniz Centre for Agricultural Landscape Research)
- Requests for lab performances are fed online.
  ↳ Metadata of experiments are already recorded.

Rutherford Appleton Laboratory
- Applications for using instruments are made online.
  ↳ Metadata of experiments are automatically recorded.

University of Bremen
- Graduates in geology have to store their research data in the database PANGAEA before the certificate is delivered.

Example: PANGAEA (www.pangaea.de)



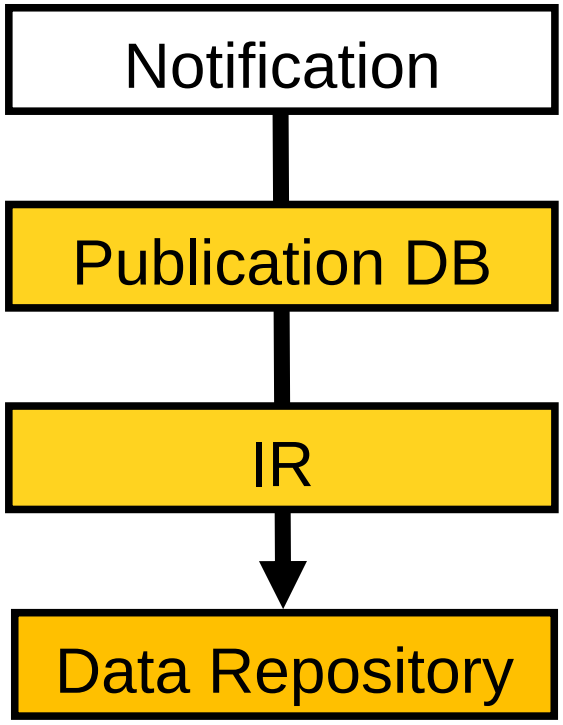
- Data sets are not sorted, but searched for.
- At the time of incorporation a DOI is added to the data sets.
- No incorporation of operationally increasing data, so far.

PANGAEA:
Cross-linking papers with research data

Basic institutional measures

- Implement infrastructure for systematic publication of research data by means of Digital Object Identifiers (DOI).
- Install virtual research environments to integrate data management into scientific work flows.
- Install adequate data repositories (centralized vs. decentralized; internal vs. external).

Once again: Institutional workflows



- In general, required policy components already exist at institutions.
- Bibliometric indicators are crucial factors for scientific reputation !
- Institutional support of workflows (library, administration, data center, ...)
- Platform for data publication (e.g., provided by library)

Vision

- **Vision** of research organisations:

  „Research data should be openly accessible for the long-term and distributed without restrictions.“

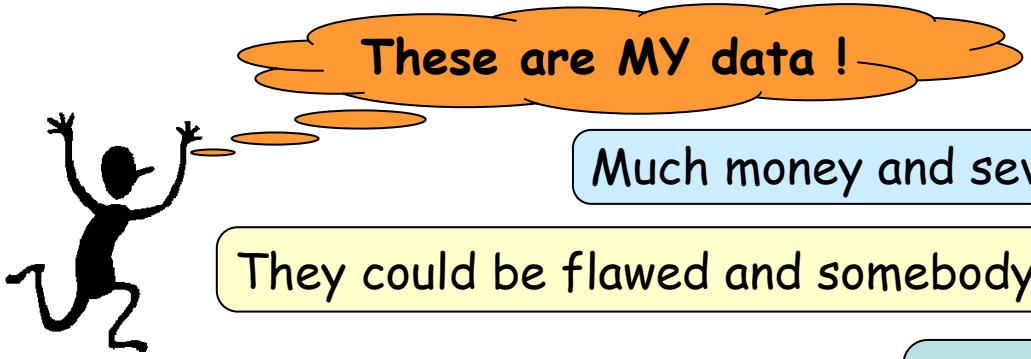
- **Intended purpose:** Open and long-term access …
  - promotes new findings and research approaches,
  - allows new interpretations,
  - is of strategic importance for the competetiveness of science locations
  - enables syntheses of research results of different disciplines,

  ↳ accelerates the process of making scientific findings !

- **But …**

… and practice

There are good reasons not to share data, that is „give away“:



Much money and several years of work were put into it.

They could be flawed and somebody else could find it.

They could easily be misinterpreted.

Someone could find something interesting in it.

Somebody could publish them before me.

How and where should I submit my data?

It needs too much time to put the data into a new context and a new format.

Data management generates costs (at the expense of research itself?).

I won't have any control on my data.

Data preparation for repositories often is too complicated.

… and practice

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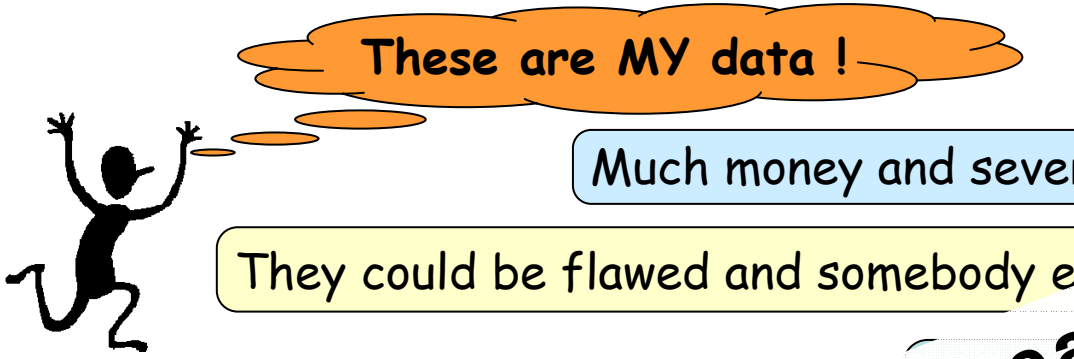

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⇨ Researchers need reasons to share !

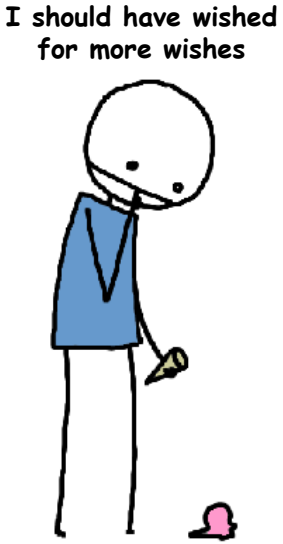
Challenges

123RF Limited

- "Temporary scientists generate the major share of research data." (and are gone …)
- Integration of data management into work flows
- Conflicting areas of „Big Data Science" vs. „Small Data Science"
- Interoperability of discipline specific, international and across discipline structures
- Data management as obligatory component of training ?
- Establishment of incentive schemes
- Cultural change: availability of research data as a matter of course

Desires

- Visualization of mutual „*benefits*“
- Adapting data management, tools and services into typical work flows in disciplines
- Generic tools that are easily transferable (e.g. within temporary projects)
- Flexible interfaces for adaptation to needs and previous knowledge of users
- Transparency: *What happens with my data ?*
- Balance between information science and service



sam brown, explodingdog

Incentives:
On honour, fame and fortune

- **Reputation**:
  - Establishment of a culture of data publication
  - Integration in reputation systems (h-index, …)
- **New professions**:
  - Bridging the discontinuity between science and information management
- **Provision of science-related tools**:
  - generic (and comprehensible) interfaces, virtual research environments, …
- **Service-oriented infrastructures**

↳ Recognition of the data provision for further use as a natural component in science

Imagine ...

... that backup and archiving, sharing and reuse of research data would be possible and is reality !

imageshostingfree.com

- ✓ **We would have more data per discipline.**
- ✓ **We could compare many differing data.**
- ✓ **Doubling of studies could be avoided.**
- ✓ **Quality control would be guaranteed.**
- ✓ **Other data sets could support own findings.**
- ✓ **Own results could be supported by others.**
- ✓ **Unique, non-reproducible results would be documented.**
- ✓ **New interpretations would be possible.**

↳ Make your data usable, not just accessible !


Thank you for your attention !