NBIS
NATIONAL BIOINFORMATICS INFRASTRUCTURE SWEDEN

SciLifeLab

# Managing your data

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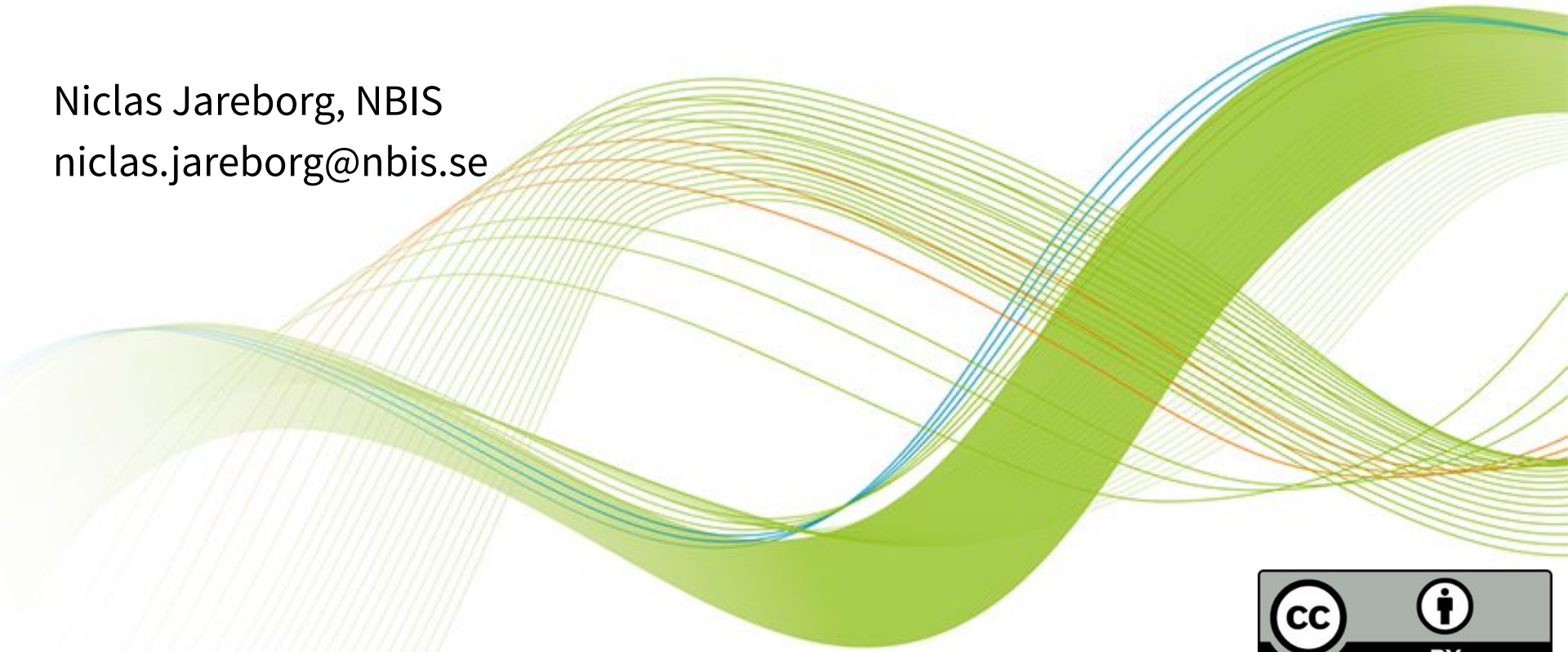


# How do you know how an old result was generated?

# The Research Data Life Cycle

# Study & Analysis


Uppsala Multidisciplinary Center for Advanced Computational Science UPPMAX

"rackham"

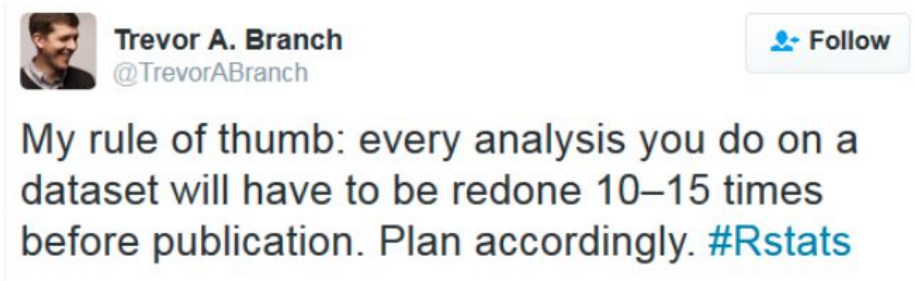
"bianca"

*Human derived data*

# Structuring data for analysis

- Guiding principle
  - *“Someone unfamiliar with your project should be able to look at your computer files and understand in detail what you did and why.”*

- Research reality
  - *"Everything you do, you will have to do over and over again”*
    – Murphy’s law

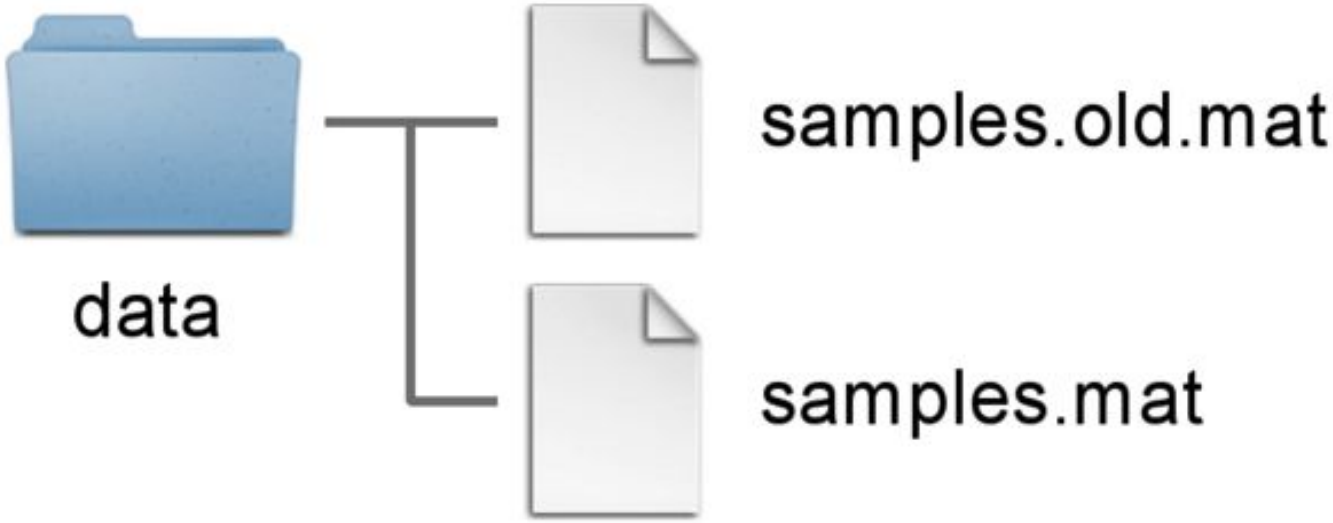



# Poor organizational choices lead to significantly slower research progress

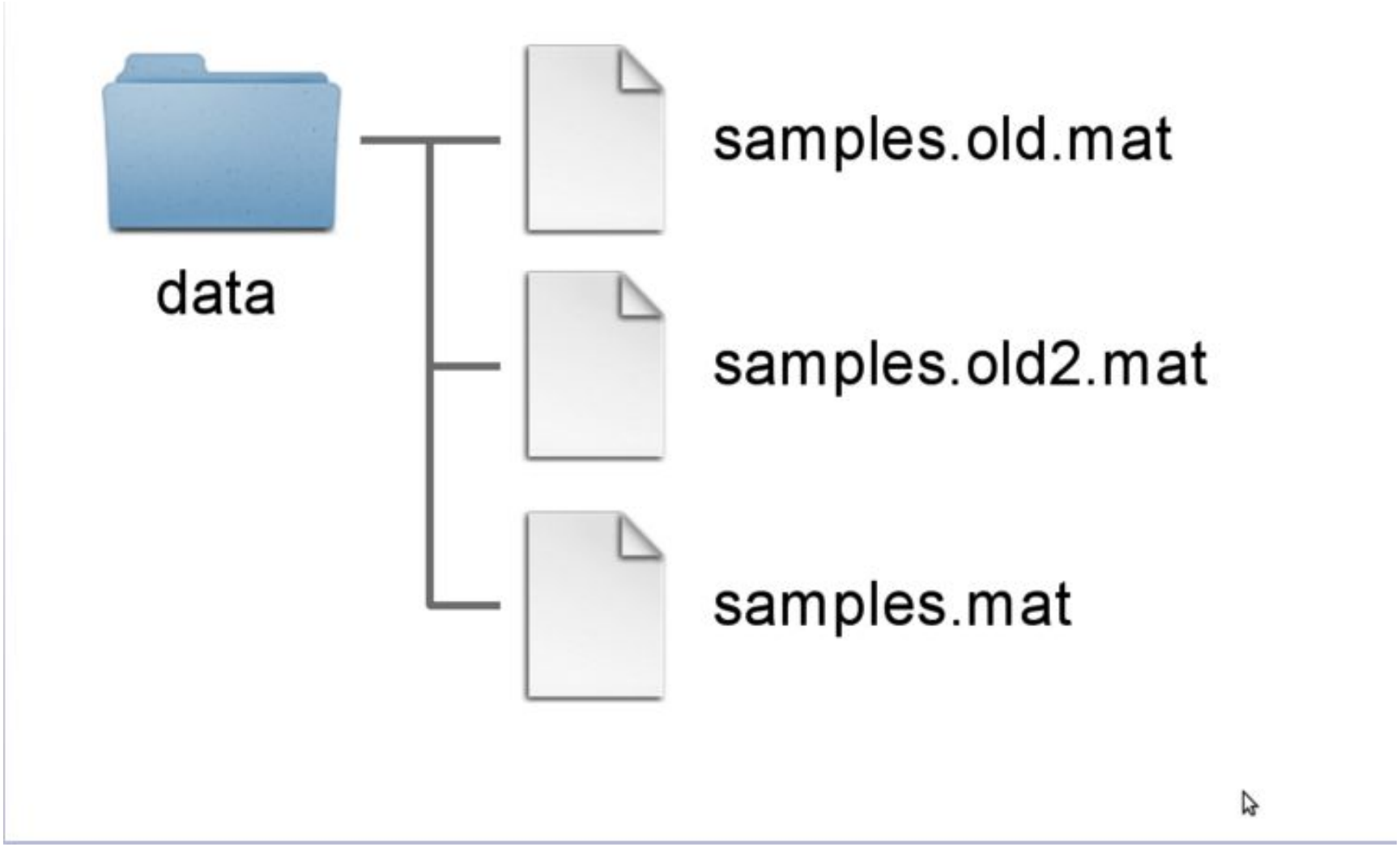
*"Your primary collaborator is yourself six months from now, and your past self doesn't answer e-mails."*

# First step - Organization

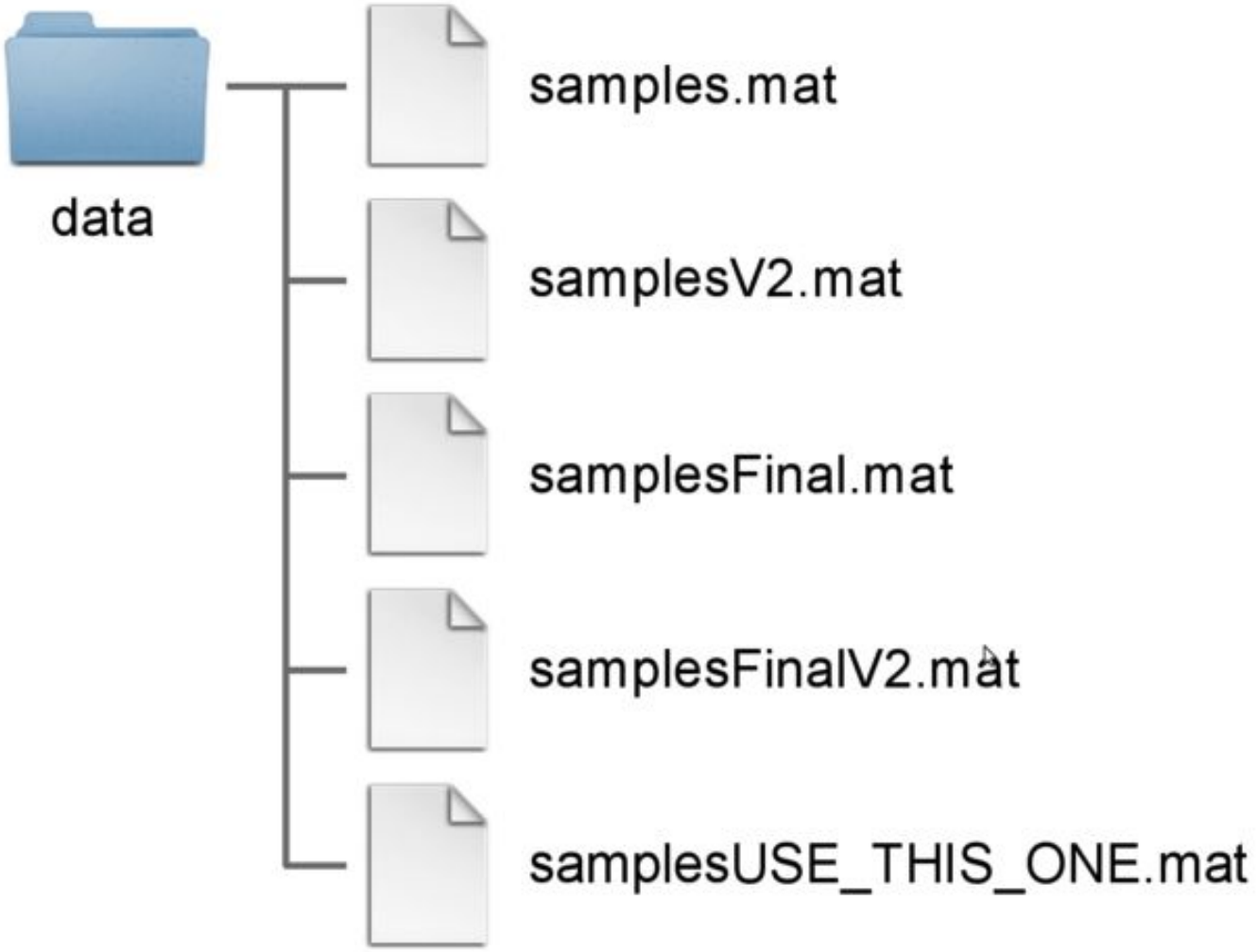
# Now what?



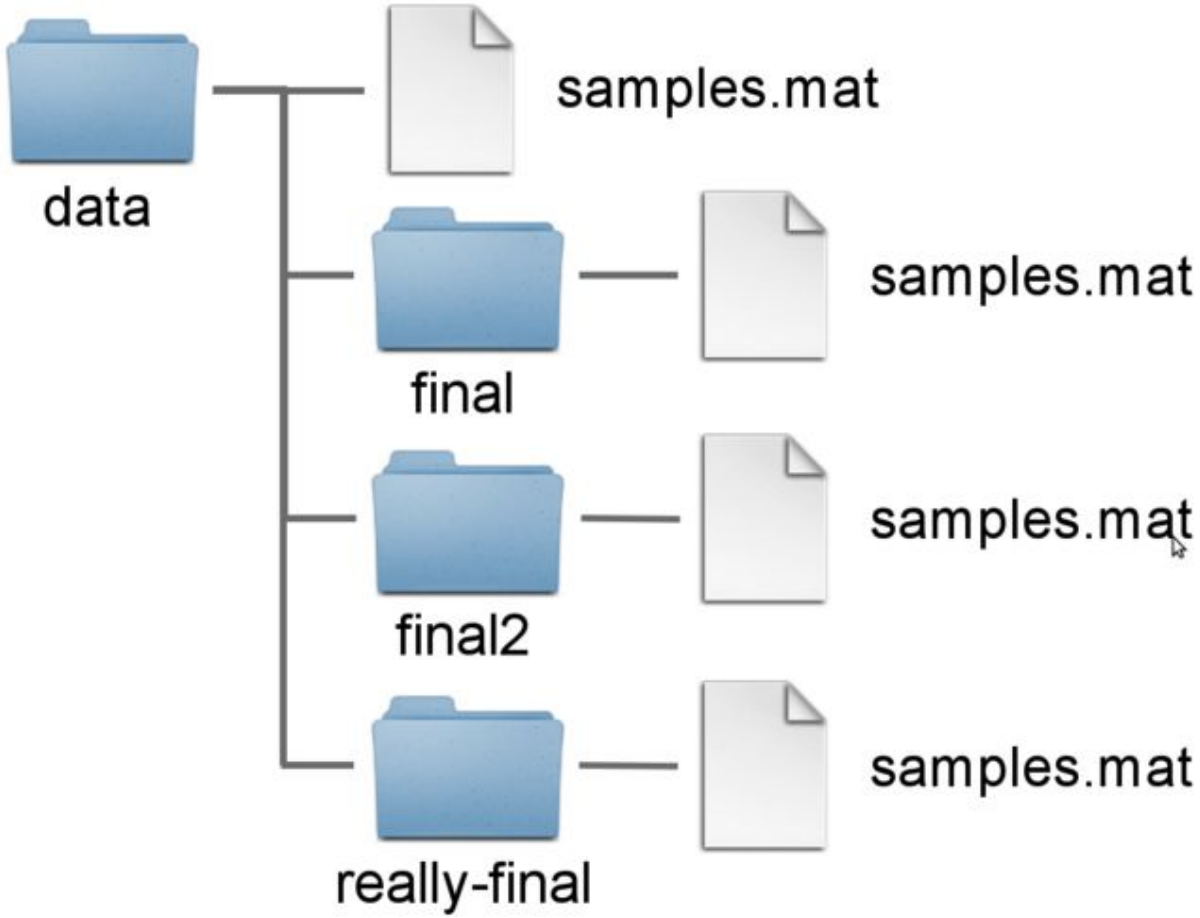
# I guess this is alright



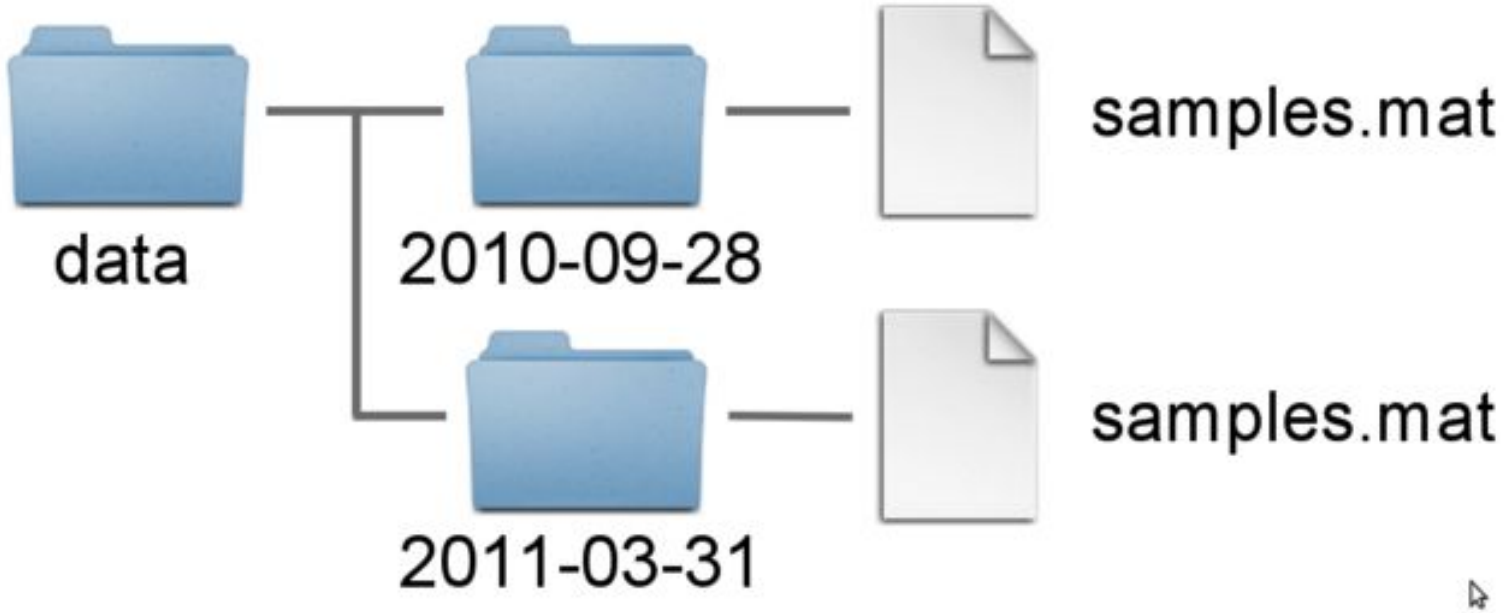
# Which one is the most recent?



# Another (bad) common approach



# A possible solution

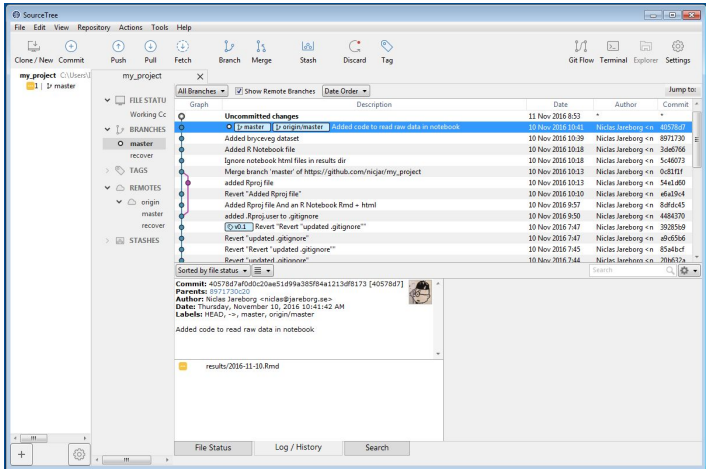


# Suggested best practices

- There is a **folder for the raw data**, which do not get altered, or intermixed with data that is the result of manual or programmatic manipulation. I.e., derived data is kept separate from raw data, and **raw data are not duplicated**.
- **Code is kept separate from data**.
- Use a **version control system** (at least for code) – e.g. **git**
- There is a **scratch directory for experimentation**. Everything in the scratch directory can be deleted at any time without negative impact.
- There should be a **README in every directory**, describing the purpose of the directory and its contents.
- Use **file naming schemes** that makes it easy to find files and understand what they are (for humans and machines)
- Use **non-proprietary formats** – *.csv* rather than *.xlsx*
- Etc…

# Version control

- What is it?
  - A system that keeps records of your changes
  - Allows for collaborative development
  - Allows you to know who made what changes and when
  - Allows you to revert any changes and go back to a previous state
- Several systems available
  - git, RCS, CVS, SVN, Perforce, Mercurial, Bazaar
  - **git**
    - Command line & GUIs
    - Remote repository hosting
      - GitHub, Bitbucket, etc



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# File naming

- Three principles
  1. Machine readable
  2. Human readable
  3. Plays well with default ordering

**NO**

myabstract.docx
Joe's Filenames Use Spaces and Punctuation.xlsx
figure 1.png
fig 2.png
JW7d^(2sl@deletethisandyourcareerisoverWx2*.txt

**YES**

2014-06-08_abstract-for-sla.docx
joes-filenames-are-getting-better.xlsx
fig01_scatterplot-talk-length-vs-interest.png
fig02_histogram-talk-attendance.png
1986-01-28_raw-data-from-challenger-o-rings.txt

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# Non-proprietary formats

- A text-based format is more future-safe, than a proprietary binary format by a commercial vendor
- ***Markdown*** is a nice way of getting nice output from text.
  - Simple & readable formating
  - Can be converted to lots of different outputs
    - HTML, pdf, MS Word, slides etc


- *Never, never, never use **Excel** for scientific **analysis**!*
  - Script your analysis – bash, python, R, …

# Tabular data / Spreadsheets

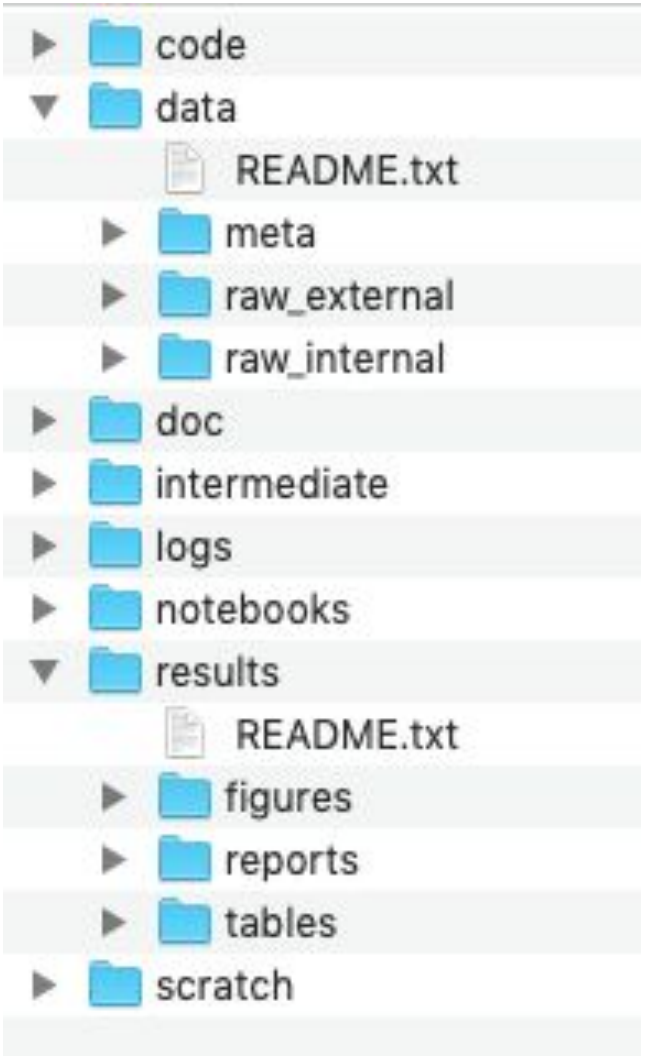
## DO

- Keep your raw data raw; calculations and analyses should be done in a copy of the file
- Put variables in columns and observations in rows
- Give each column a descriptive heading that does not include spaces, numbers, or special characters
- Differentiate between zero and null values
- Validate your data
- Keep a separate txt file with a title and a legend describing your dataset, and outlining any steps you take to tidy your data
- Use a version control system and back up your files
- Export each data file in an open non-proprietary format such as CSV or TAB, with a name that appropriately reflects the content of that file
- Check your data thoroughly. Your data should receive the same care as your publications

## DO NOT

- Put more than 1 piece of information in a cell
- Use colour coding, embedded charts, comments or tables – your spreadsheet is not a lab book
- Include special (i.e. non alphanumeric) characters within the spreadsheet, including commas
- Use merged or blank cells
- Create multiple worksheets within a spreadsheet

# Directory structure for a sample project




all code needed to go from input files to final results

raw and primary data, essentially all input files, **never** edit!

documentation for the study

output files from different analysis steps, *can be deleted*

logs from the different analysis steps

output from workflows and analyses

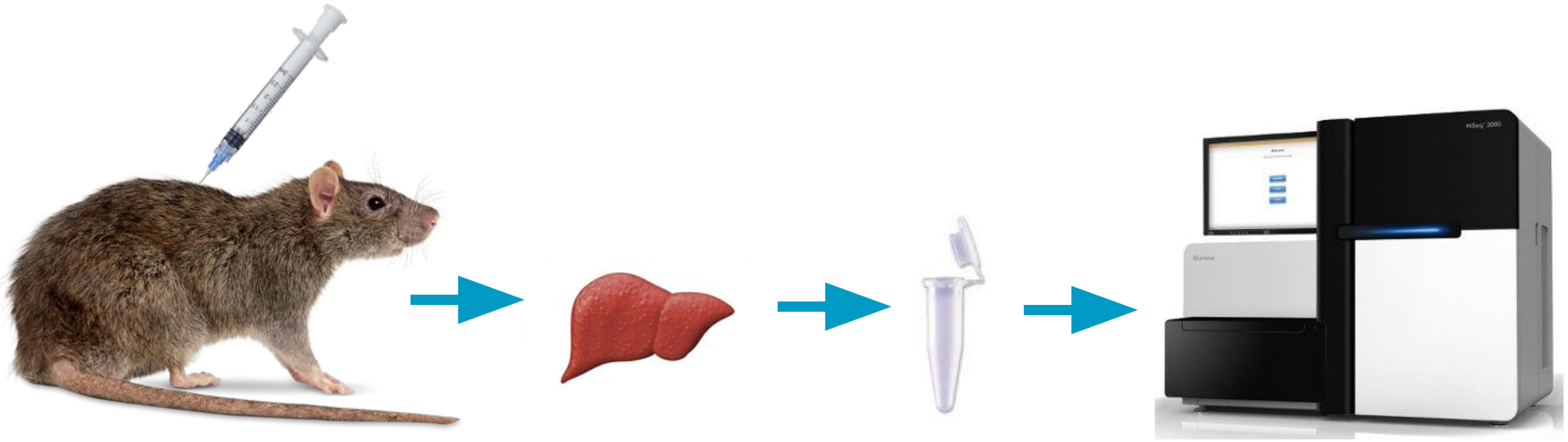
temporary files that can be safely *deleted or lost*

Noble WS (2009) A Quick Guide to Organizing Computational Biology Projects. PLoS Comput Biol 5(7): e1000424.
http://journals.plos.org/ploscompbiol/article?id=info:doi/10.1371/journal.pcbi.1000424

# Metadata

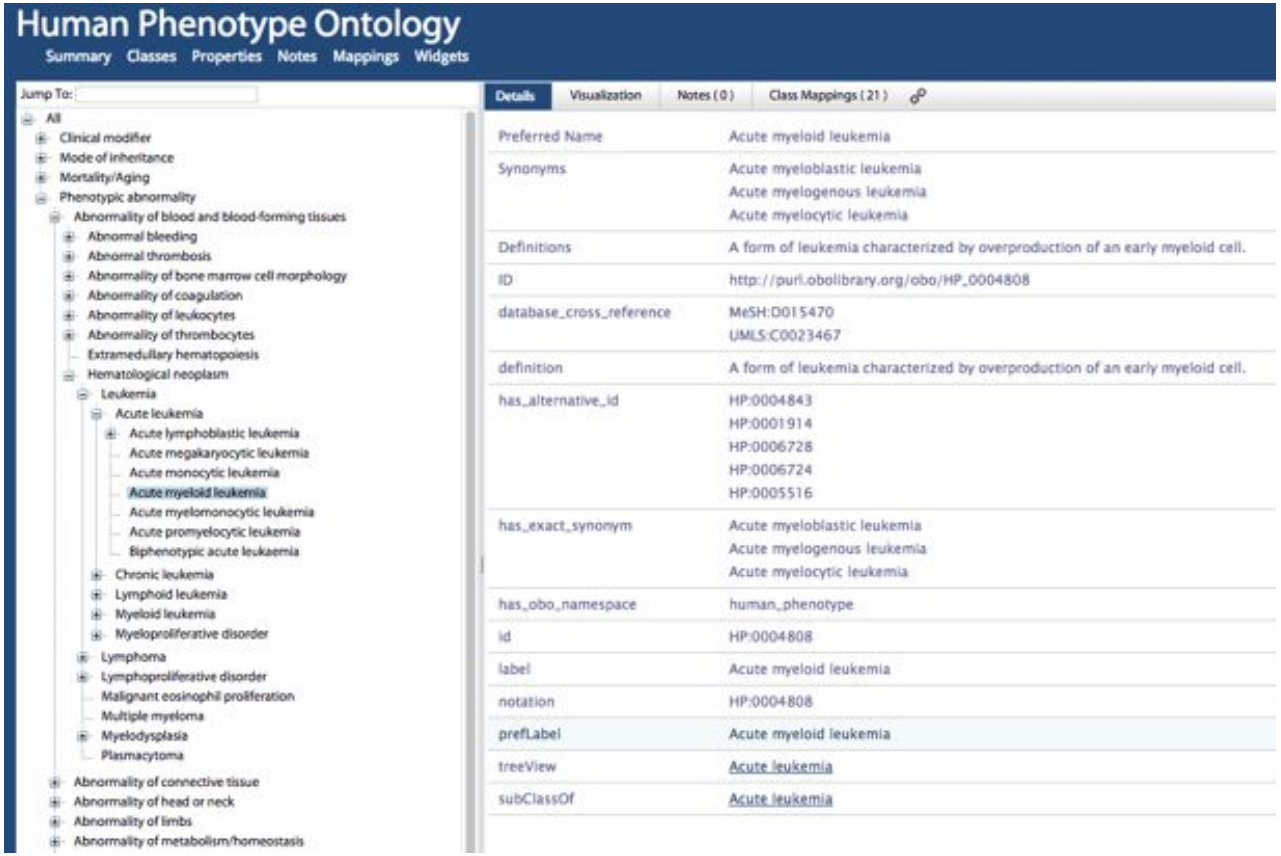
- "Data about the data"
  - From what was the data generated?
  - How do the samples differ?
  - What where the experimental conditions?
  - Etc

| SampleID | Species | Strain | Treatment | Dose | Organ | *etc…* |
|---|---|---|---|---|---|---|
| A9876 | rat | Balb/c | Paracetamol | 10 mg | liver | ... |
| A6543 | brown rat | Bagg Albino | . | 0 | liver | ... |
| ... | | | | | | |

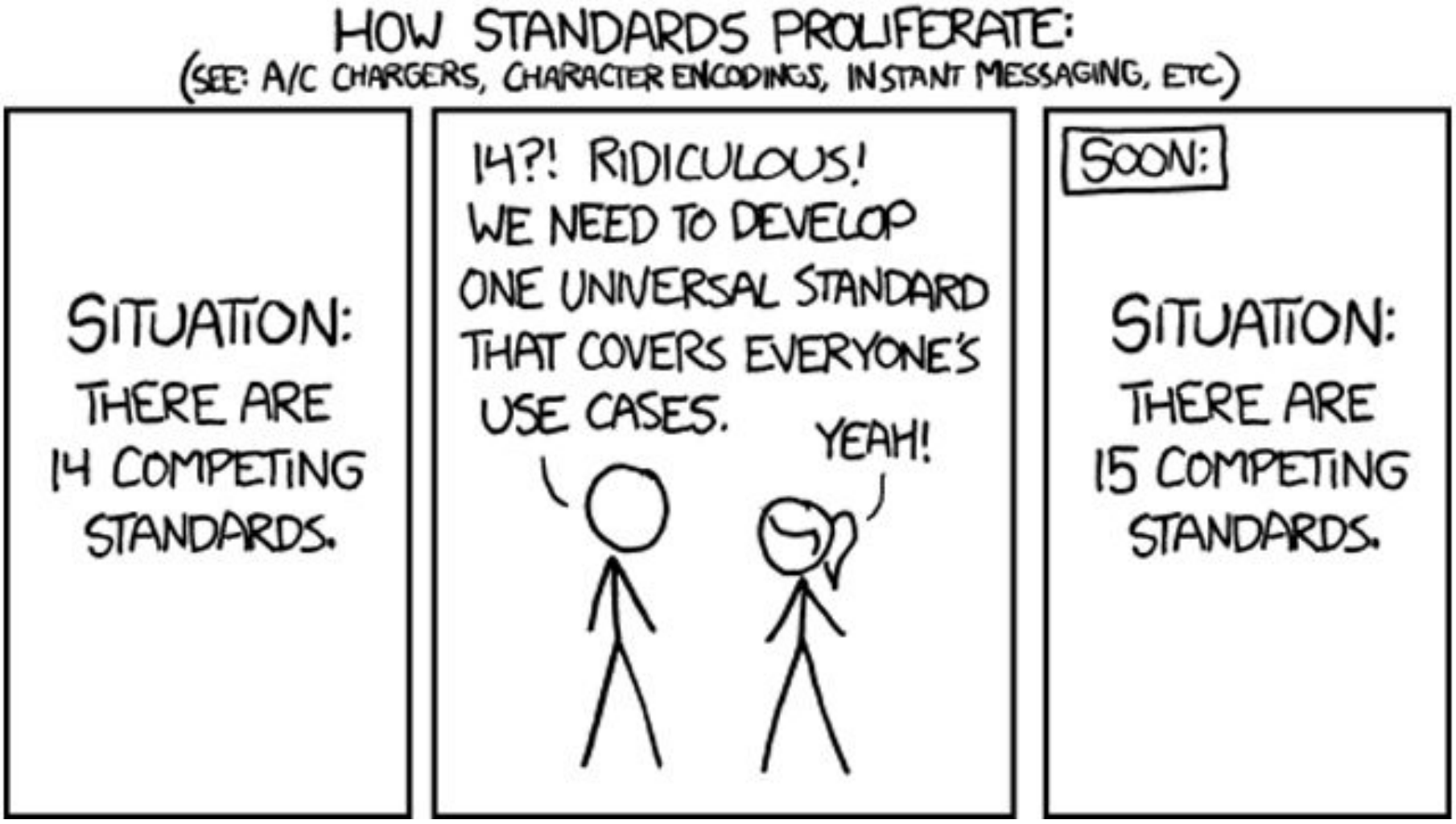


# Metadata standards

- Controlled vocabularies / taxonomies / Ontologies
  - Agreed terms for different phenomena

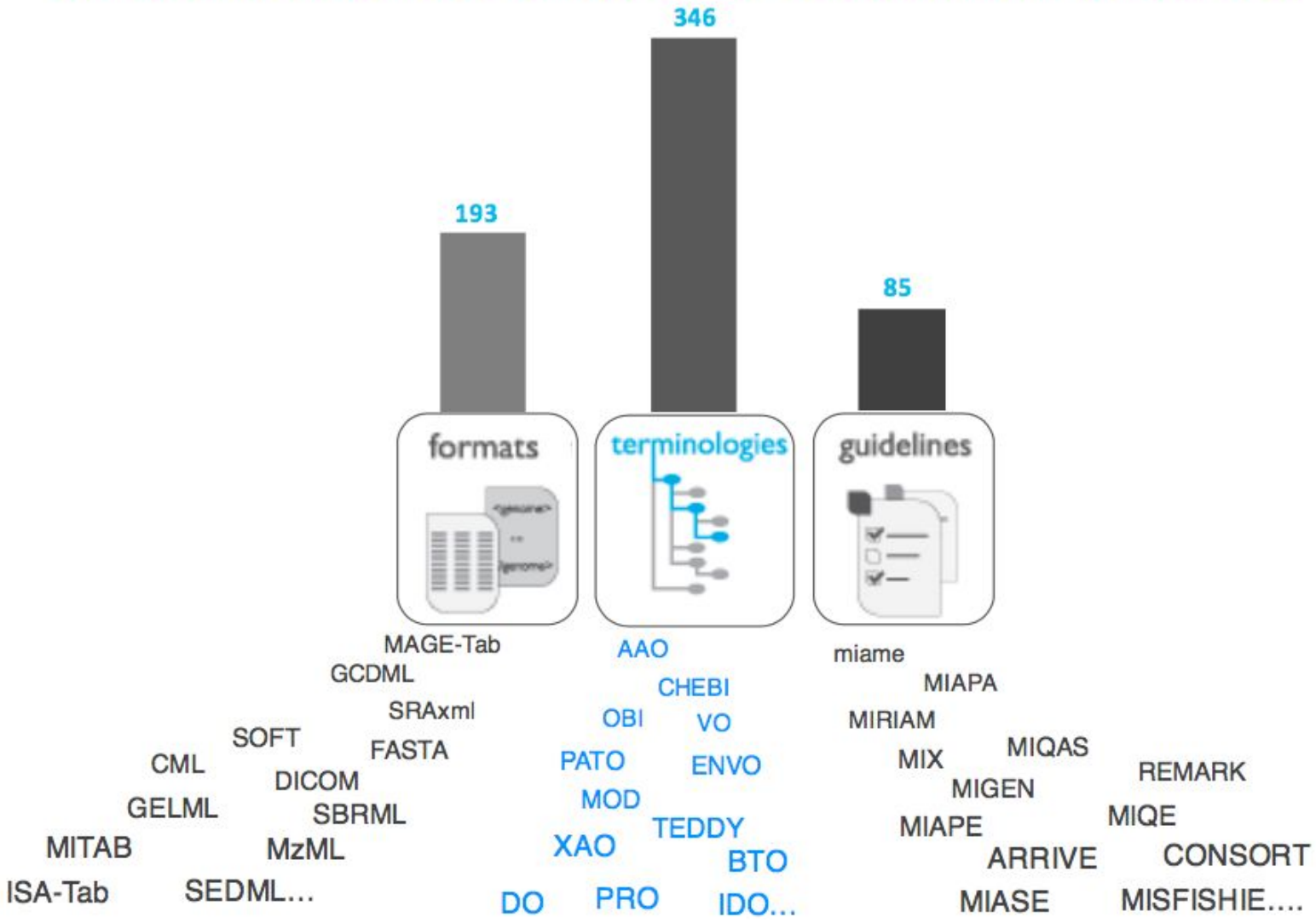


# Metadata standards



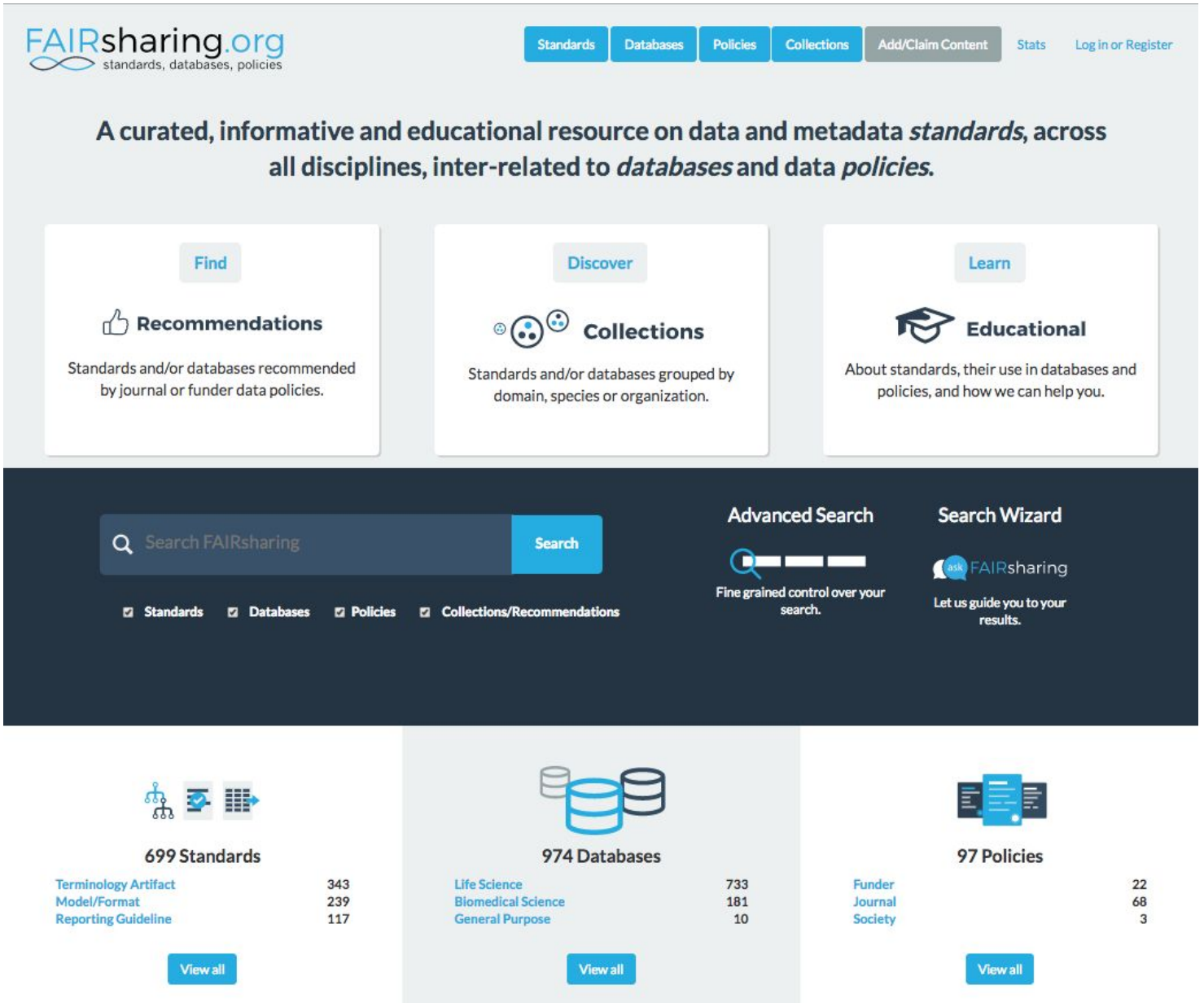
# Life science standards

## In the life sciences there are >600 *content standards*



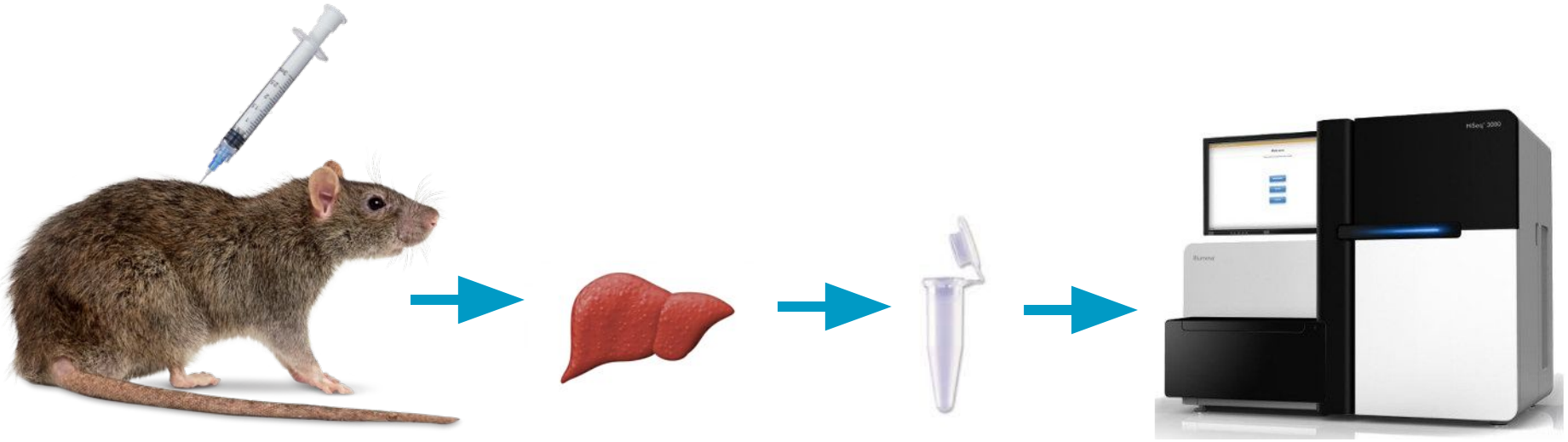
# FAIRsharing.org

*(was biosharing.org)*



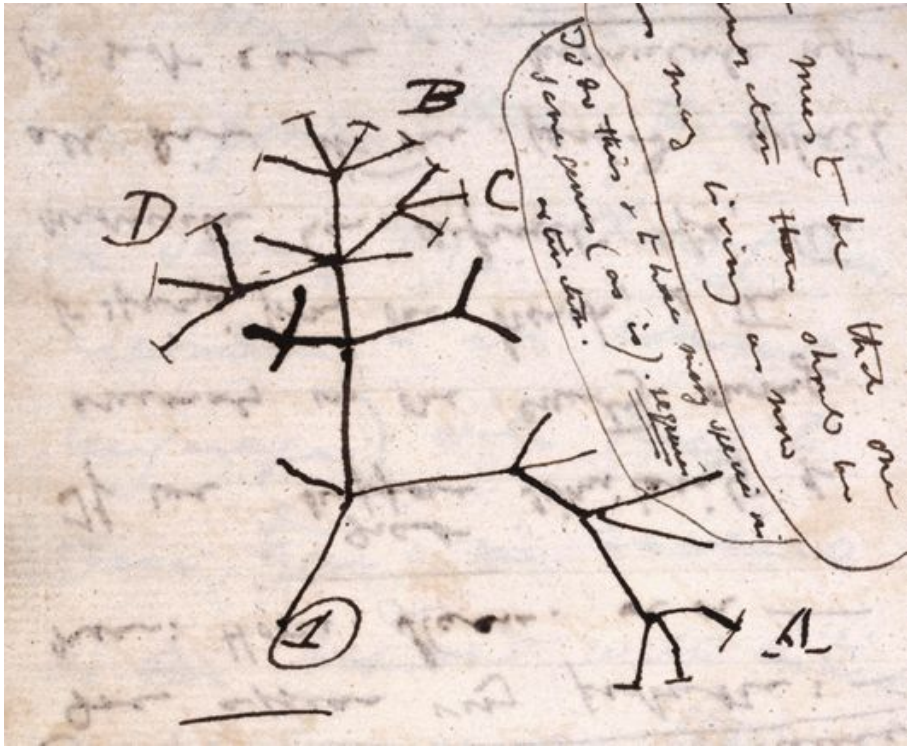
# Metadata

| SampleID | Species | | Strain | | Compound | | Dose |
|---|---|---|---|---|---|---|---|
| | *NCBI:txid* | *SciName* | *MGI_ID* | *name* | *ChEBI_ID* | *name* | *mg* |
| A9876 | 10116 | Rattus norvegicus | *MGI:2161072* | *BALB/c* | *CHEBI:46195* | paracetamol | 10 |
| A6543 | 10116 | *Rattus* norvegicus | *MGI:2161072* | *BALB/c* | null | null | null |
| ... | | | | | | | |



# Lab notebooks

- Why?
  - You have to understand what you have done
  - **Others should be able to reproduce what you have done**

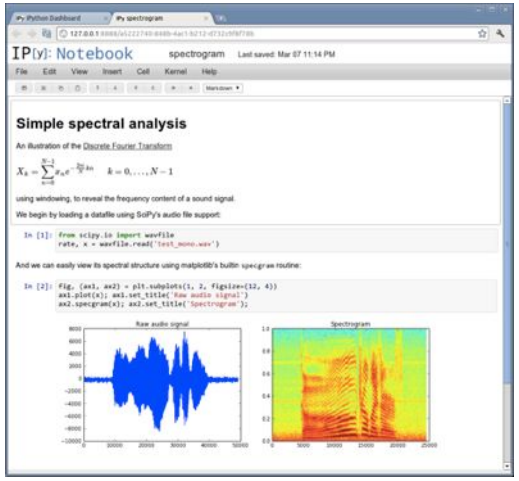


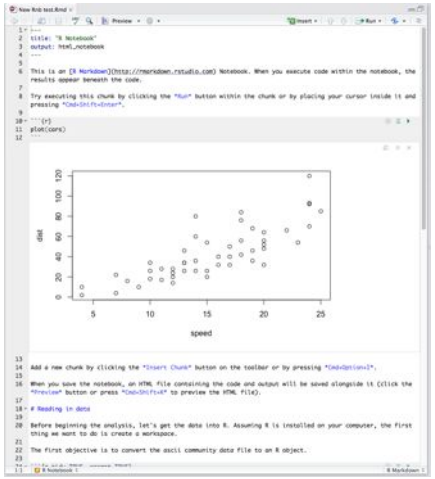
# Lab notes – useful practices

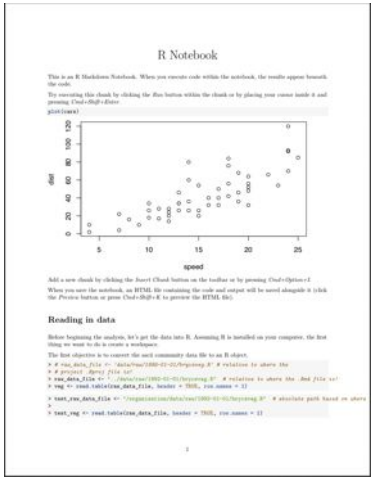
- Put in *separate* directory (e.g. *results*, *documentation*)
- *Dated* entries
- Entries relatively verbose
- Link to *data* and *code* (including versions)
    - Point to commands run and results generated
- Embedded images or tables showing results of analysis done
- Observations, Conclusions, and *ideas* for future work
- Also document analysis that *doesn't* work, so that it can be understood why you choose a particular way of doing the analysis in the end

# Where to take down notes

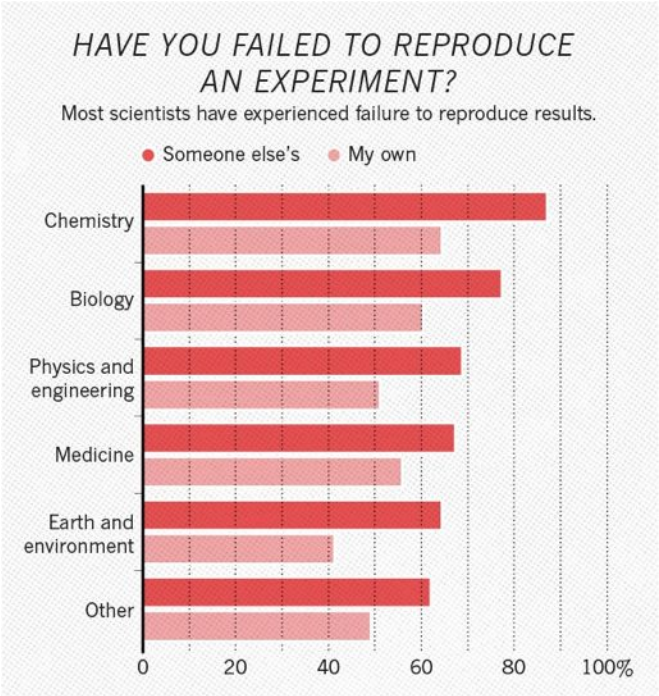
- Paper Notebook
- Word processor program / Text files
- Electronic Lab Notebooks Systems
- **Computational Notebooks**
  - e.g. jupyther, R Notebooks in RStudio
  - Plain text - work well with version control (Markdown)
  - Embed and execute code
  - Convert to other output formats
    - html, pdf, word

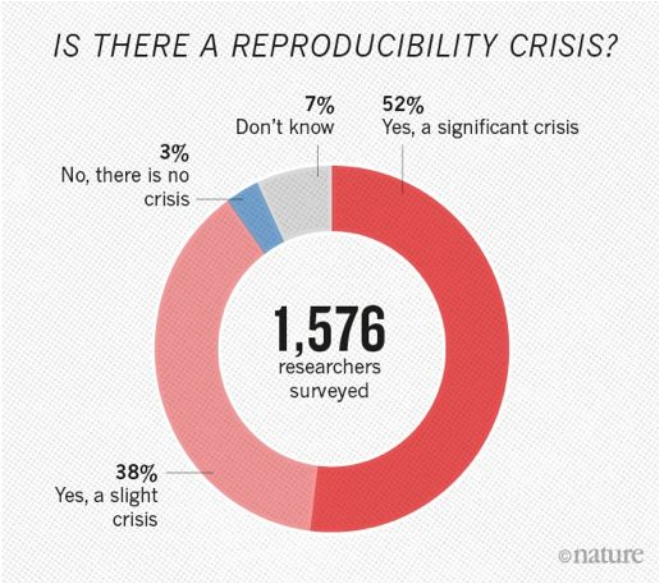




# A reproducibility crisis

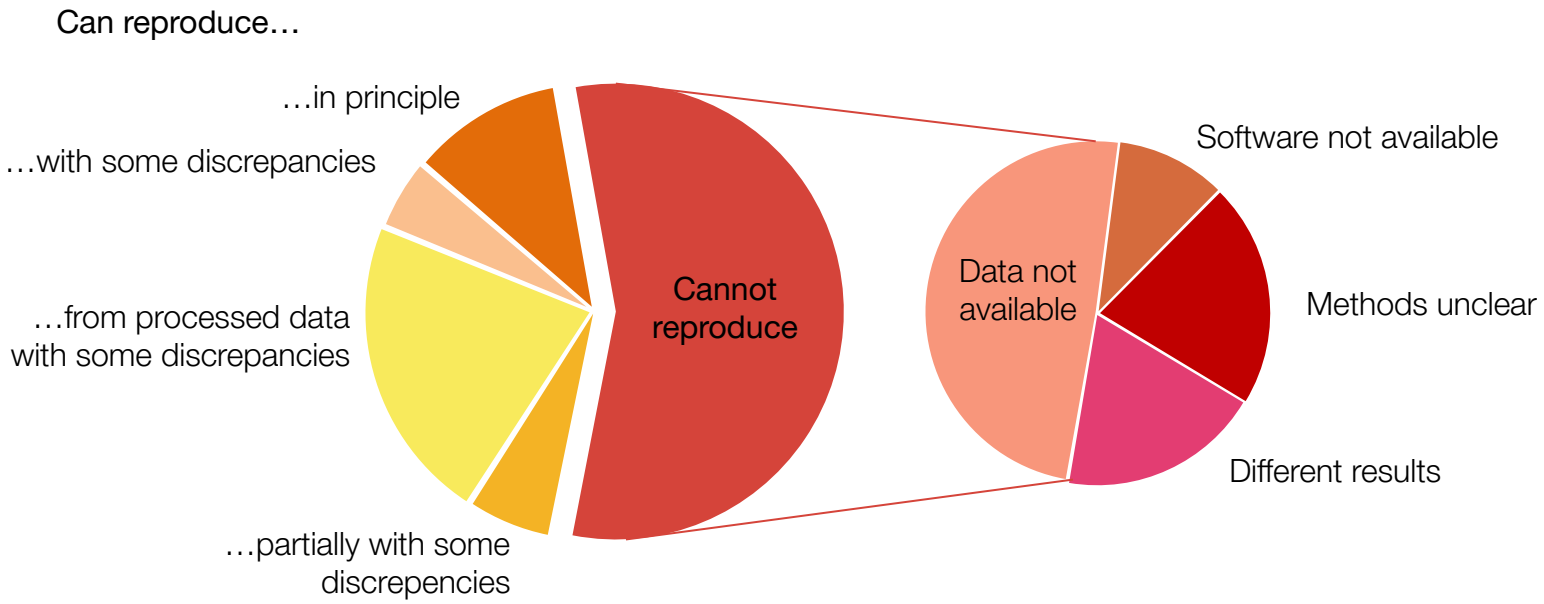



[1] "1,500 scientists lift the lid on reproducibility". Nature. 533: 452–454
[2] Begley, C. G.; Ellis, L. M. (2012). "Drug development: Raise standards for preclinical cancer research". Nature. 483 (7391): 531–533.

# A reproducibility crisis

Reproduction of data analyses in 18 articles on microarray-based gene expression profiling published in Nature Genetics in 2005–2006:

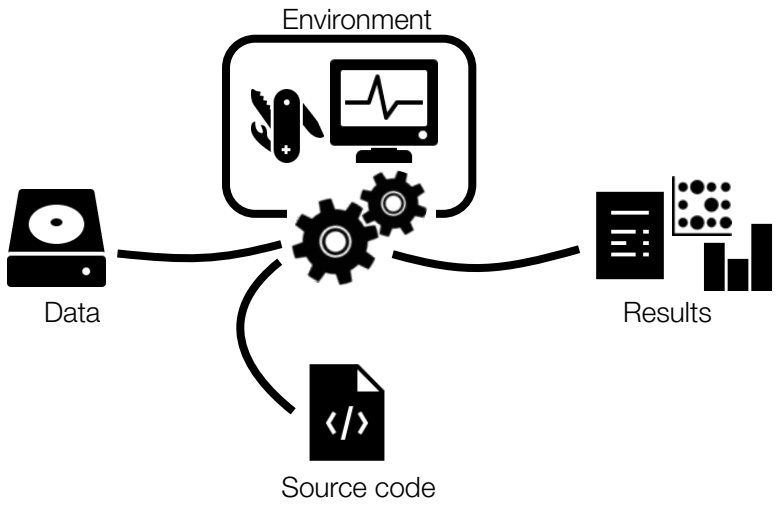



Summary of the efforts to replicate the published analyses.
Adopted from: Ioannidis et al. Repeatability of published microarray gene expression analyses. *Nature Genetics* 41 (2009) doi:10.1038/ng.295

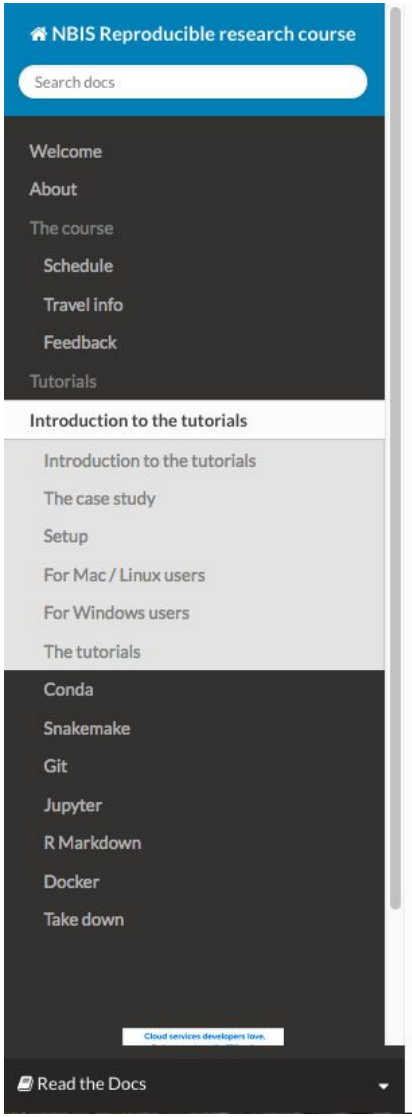
# What do we mean by reproducible research?

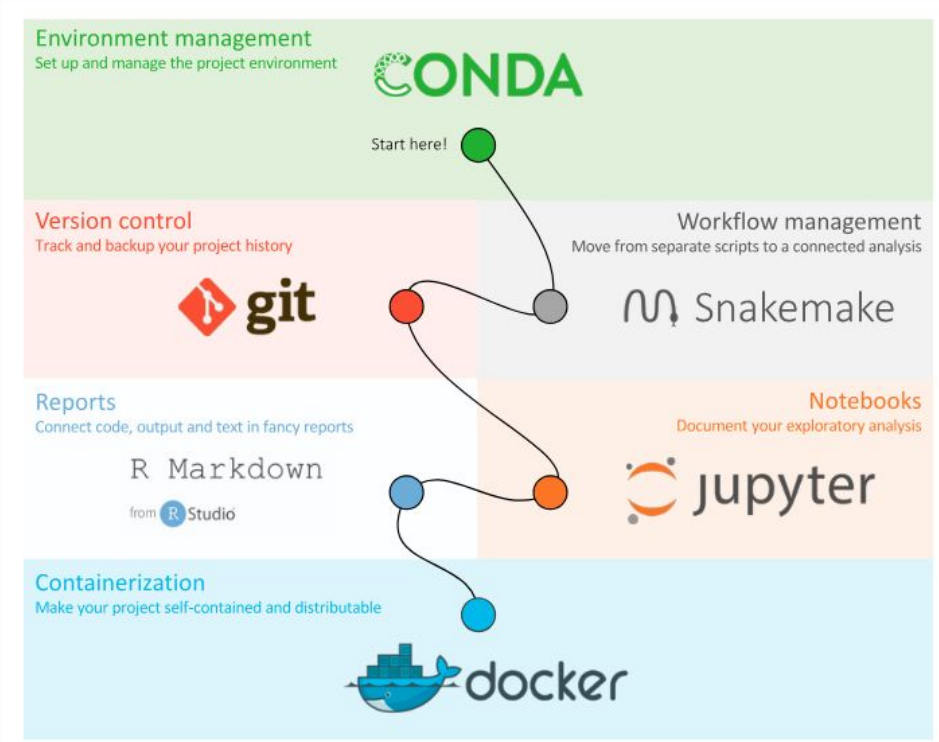
| | | Data: Same | Data: Different |
|---|---|---|---|
| Code | Same | Reproducible | Replicable |
| Code | Different | Robust | Generalizable |

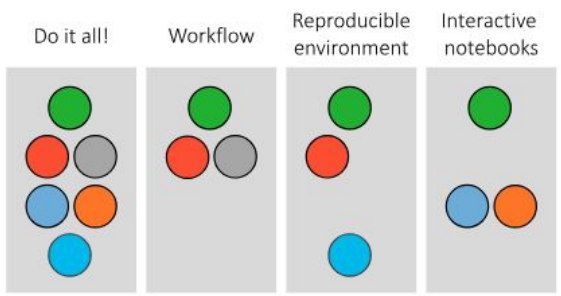
All parts of a bioinformatics analysis have to be reproducible:



# Reproducible Research tutorials

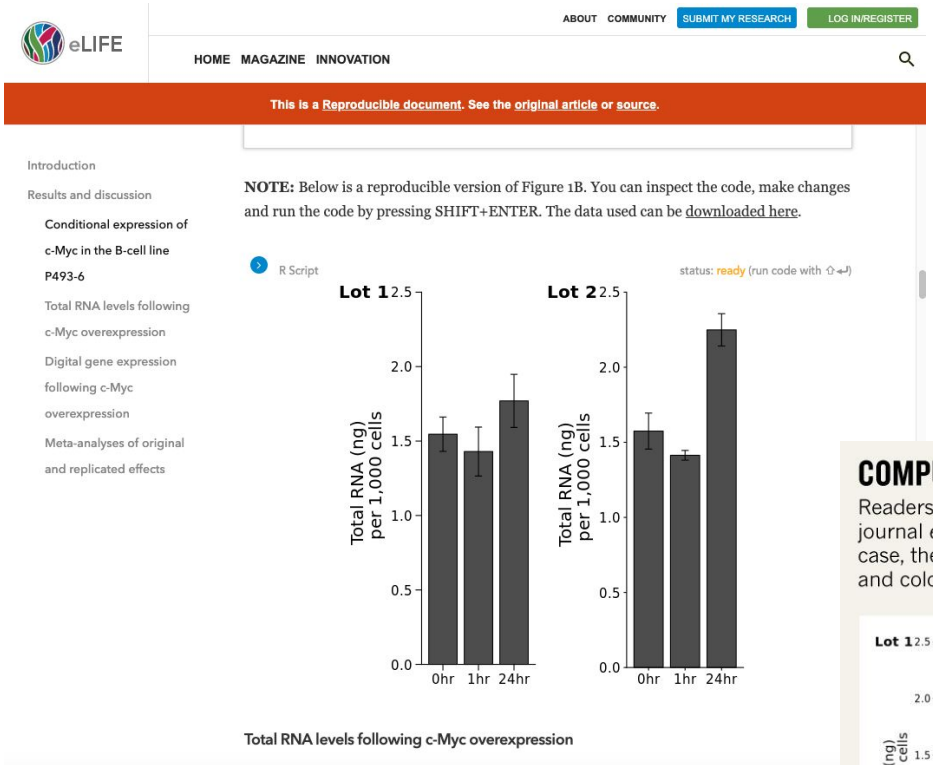



https://nbis-reproducible-research.readthedocs.io/en/latest/

# Reproducible publications

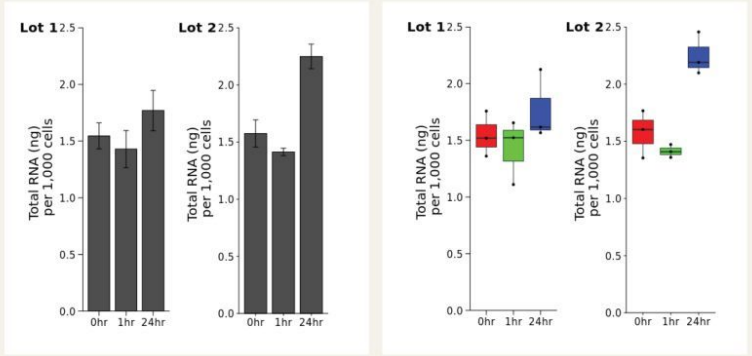



eLIFE

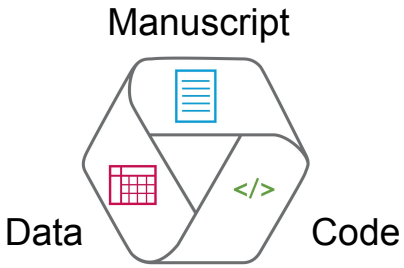
https://elifesciences.org

## COMPUTATIONAL REPRODUCIBILITY

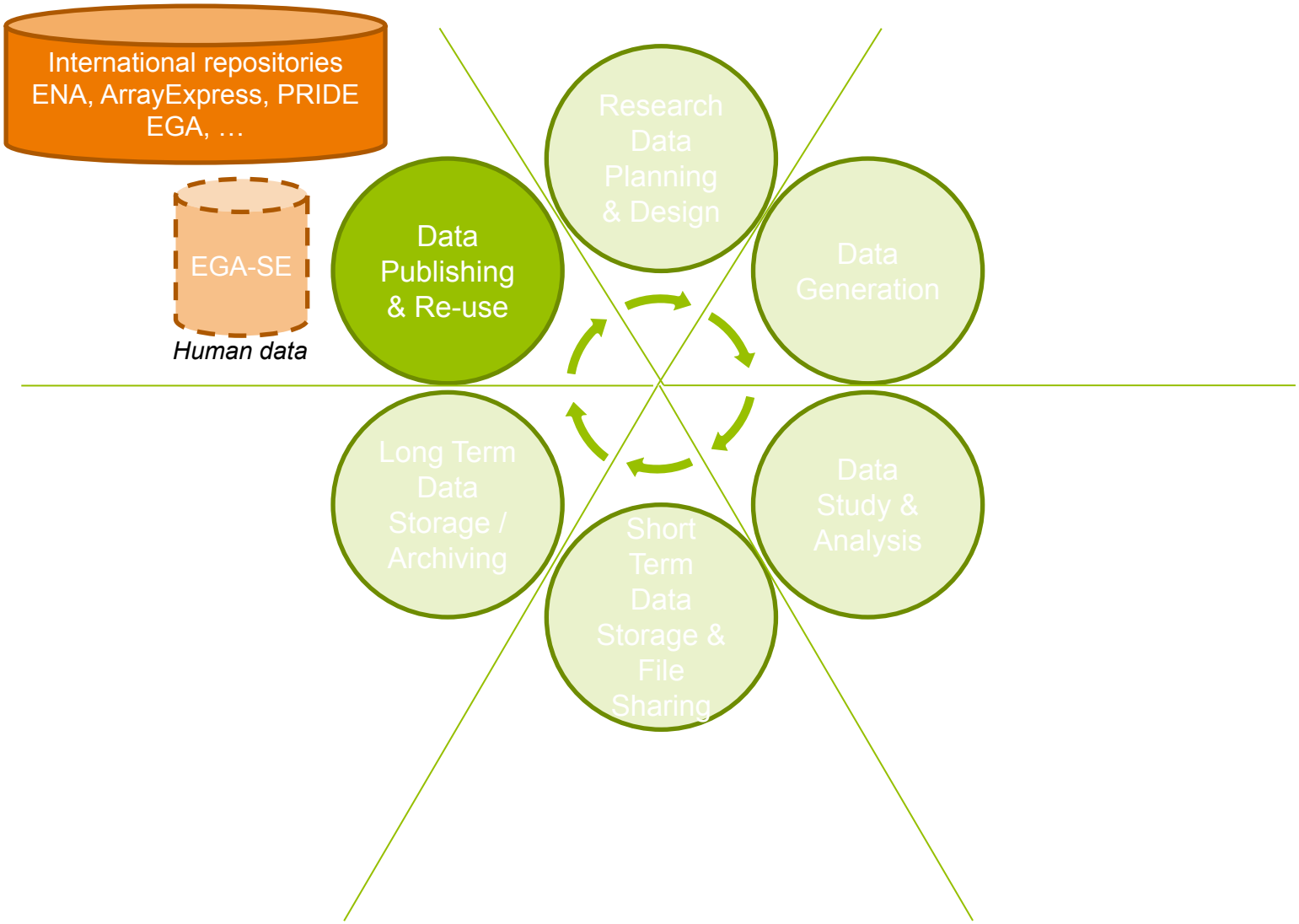
Readers of the first computationally reproducible article published by the journal *eLife* can tweak the underlying code to change the figures. In this case, the authors' original figure (left) was altered to change its chart type and coloration.




©nature



# Data Publishing & Re-use



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# Why should you make research data available for others?

# *Open Access*

*The practice of providing* ***on-line access*** *to scientific information that is* ***free of charge*** *to the end-user and that is* ***re-usable****.*

# *Open Access*

- Democracy and transparency
  - Publicly funded research data should be accessible to all
  - Published results and conclusions should be possible to check by others
- Research
  - Enables others to combine data, address new questions, and develop new analytical methods
  - Reduce duplication and waste
- Innovation and utilization outside research
  - Public authorities, companies, and private persons outside research can make use of the data
- Citation
  - Citation of data will be a merit for the researcher that produced it

# *Open Access* to research data

- Strong international movement towards Open Access (OA)
- European Commission recommended the member states to establish national guidelines for OA
  - Swedish Research Council (VR) submitted proposal to the government Jan 2015
- Research bill 2017–2020 – *28 Nov 2016*
  - "*The aim of the government is that all scientific publications that are the result of publicly funded research should be openly accessible as soon as they are published. Likewise,* ***research data*** *underlying scientific publications should be* ***openly accessible*** *at the time of publication.*" [my translation]
- 2018 – VR assigned by the government to coordinate national efforts to implement open access to research data


# Research Ethics

- Is it ethical to do bad/careless science?
  - Wasting resources
    - *… or even resulting in dangerous medical practices*
  - Contribute to the current research credibility crisis
  - harming the profession
  - harming the public trust

- But!
  - Careless science -> longer CV

My take of material by Rochelle Tractenberg "Unexpected Ethical Challenges in Bioinformatics and Genomics."

# What is needed for others to be able to re-use your data?

# Data Management Snafu


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

# FAIR

- To be useful for others data should be
  - ***FAIR*** - Findable, Accessible, Interoperable, and Reusable
    *… for both Machines and Humans*

Wilkinson, Mark et al. *"The FAIR Guiding Principles for scientific data management and stewardship"*. Scientific Data 3, Article number: 160018 (2016)
http://dx.doi.org/10.1038/sdata.2016.18

www.nature.com/scientificdata

SCIENTIFIC DATA

OPEN

SUBJECT CATEGORIES
» Research data
» Publication characteristics

Received: 10 December 2015
Accepted: 12 February 2016
Published: 15 March 2016

**Comment: The FAIR Guiding Principles for scientific data management and stewardship**

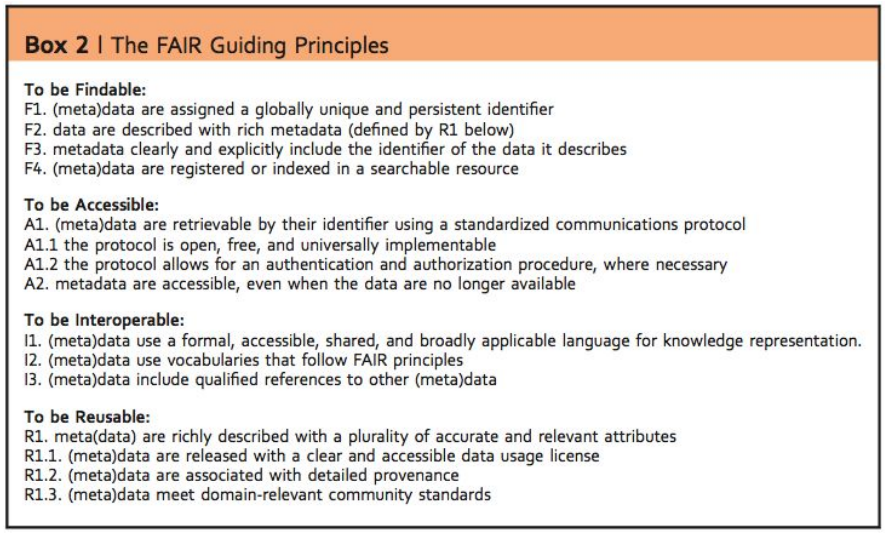
Mark D. Wilkinson *et al.*#

There is an urgent need to improve the infrastructure supporting the reuse of scholarly data. A diverse set of stakeholders—representing academia, industry, funding agencies, and scholarly publishers—have come together to design and jointly endorse a concise and measureable set of principles that we refer to as the FAIR Data Principles. The intent is that these may act as a guideline for those wishing to enhance the reusability of their data holdings. Distinct from peer initiatives that focus on the human scholar, the 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. This Comment is the first formal publication of the FAIR Principles, and includes the rationale behind them, and some exemplar implementations in the community.

**Supporting discovery through good data management**

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. Unfortunately, the existing digital ecosystem surrounding scholarly data publication prevents us from extracting maximum benefit from our research investments (e.g., ref. 1). Partially in response to this, science funders, publishers and

DOI: 10.1038/sdata.2016.18



**Box 2** | The FAIR Guiding Principles

**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
I3. (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


**'We support appropriate efforts to promote open science and facilitate appropriate access to publicly funded research results on findable, accessible, interoperable and reusable (FAIR)'**

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# FAIR

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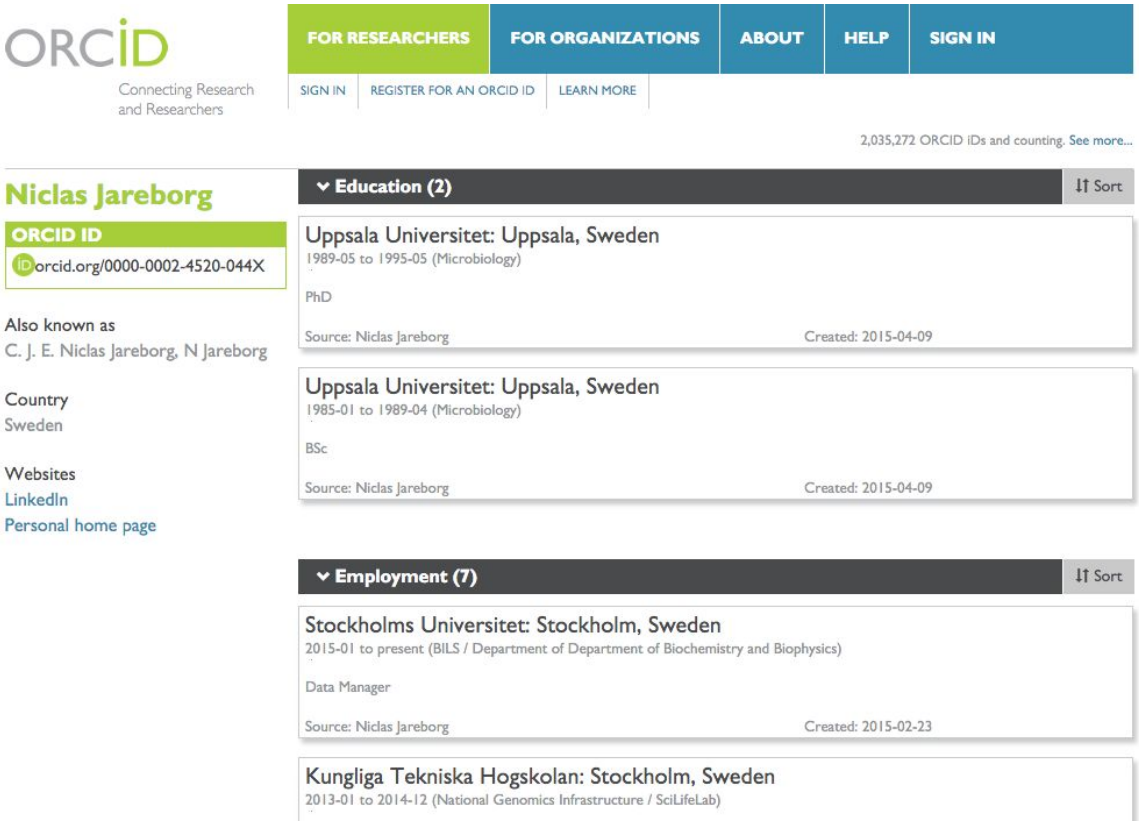
# Findable

- Long-term storage
  - Data should not disappear
- Persistent identifiers
  - Possibility to refer to a dataset over long periods of time
  - Unique
  - e.g. DOIs (Digital Object Identifiers)
- Discoverability
  - Expose dataset metadata through search functionalities

# *Persistent identifier for yourself*

- ORCID is an open, non-profit, community-driven effort to create and maintain a registry of unique researcher identifiers and a transparent method of linking research activities and outputs to these identifiers.
- http://orcid.org
- Persistent identifier for you as a researcher



# Repositories

Does the data contain **personal** or **sensitive** information?

**YES** → 1. Controlled access repositories

**NO** → Is there a **discipline-specific** repository for your data?

**YES** → 2. Discipline-specific repositories

**NO** → Is there a **institutional** repository for your data?

**YES** → 3. Institutional repositories

**NO** → 4. General data repositories

ENA European Nucleotide Archive

BioSamples

EUROPEAN GENOME-PHENOME ARCHIVE

NCBI

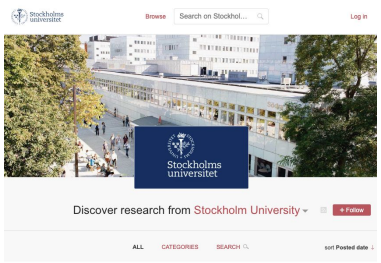
dbSNP Short Genetic Variations

PRIDE

Protein Data Bank in Europe
Bringing Structure to Biology

SND Svensk nationell datatjänst

figshare




zenodo

EUDAT

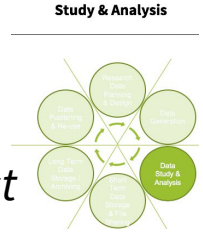
Etc...

# Accessible

## International public repositories

- Best way to make data **FAIR**
- Domain-specific metadata standards

💡 *Strive towards uploading data to its final destination already at the beginning of a project*



# Recommended repositories

## ELIXIR Deposition Database list

| Deposition Database | Data type | International collaboration framework [1] |
|---|---|---|
| ArrayExpress | Functional genomics data. Stores data from high-throughput functional genomics experiments. | |
| BioModels | Computational models of biological processes. | |
| BioSamples | BioSamples stores and supplies descriptions and metadata about biological samples used in research and development by academia and industry. | NCBI BioSamples database |
| BioStudies | Descriptions of biological studies, links to data from these studies in other databases, as well as data that do not fit in the structured archives. | |
| EGA | Personally identifiable genetic and phenotypic data resulting from biomedical research projects. | European Bioinformatics Institute and the Centre for Genomic Regulation |
| EMDB | The Electron Microscopy Data Bank is a public repository for electron microscopy density maps of macromolecular complexes and subcellular structures. | |
| ENA | Nucleotide sequence information, covering raw sequencing data, contextual data, sequence assembly information and functional and taxonomic annotation. | International Nucleotide Sequence Database Collaboration |
| EVA | The European Variation Archive covers genetic variation data from all species. | dbSNP and dbVAR |
| IntAct | IntAct provides a freely available, open source database system and analysis tools for molecular interaction data. | The International Molecular Exchange Consortium |
| MetaboLights | Metabolite structures and their reference spectra as well as their biological roles, locations and concentrations, and experimental data from metabolic experiments. | |
| PDBe | Biological macromolecular structures. | wwPDB |
| PRIDE | Mass spectrometry-based proteomics data, including peptide and protein expression information (identifications and quantification values) and the supporting mass spectra evidence. | The ProteomeXchange Consortium |

https://www.elixir-europe.org/platforms/data/elixir-deposition-databases

## Scientific Data Recommended Data Repositories

SCIENTIFIC DATA

### Biological sciences

#### Nucleic acid sequence

Sequence information should be deposited following the MIxS guidelines.

Simple genetic polymorphisms or structural variations should be submitted to dbSNP or dbVar (please note that these repositories cannot accept sensitive data derived from human subjects); the NCBI Trace Archive may be used for capillary electrophoresis data, while SRA accepts NGS data only.

| | |
|---|---|
| DNA DataBank of Japan (DDBJ) | view FAIRsharing entry |
| European Nucleotide Archive (ENA) | view FAIRsharing entry |
| GenBank | view FAIRsharing entry |
| dbSNP | view FAIRsharing entry |
| European Variation Archive (EVA) | view FAIRsharing entry |
| dbVar | view FAIRsharing entry |
| Database of Genomic Variants Archive (DGVa) | view FAIRsharing entry |
| EBI Metagenomics | view FAIRsharing entry |
| NCBI Trace Archive | view FAIRsharing entry |
| NCBI Sequence Read Archive (SRA) | view FAIRsharing entry |
| NCBI Assembly | |

#### Protein sequence

| | |
|---|---|
| UniProtKB | view FAIRsharing entry |

#### Molecular & supramolecular structure

These repositories accept structural data for small molecules (COD); peptides and proteins (all); and larger assemblies (EMDB).

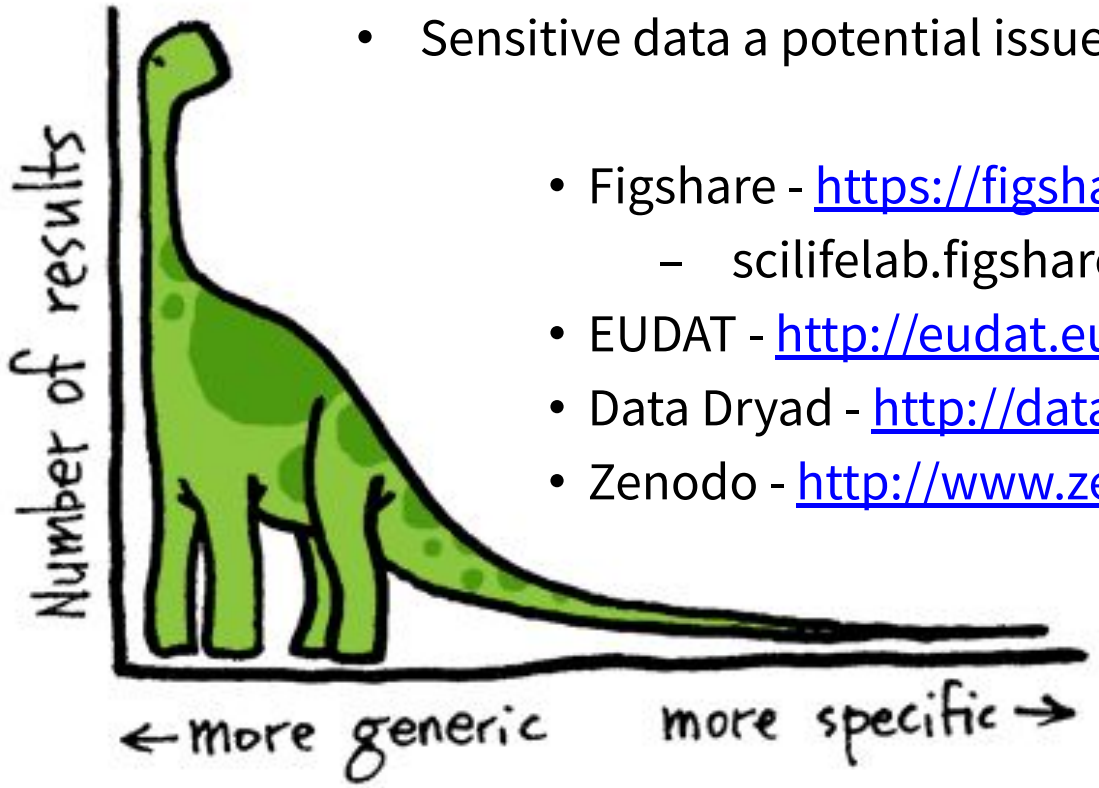
Small molecule crystallographic data should be uploaded to Dryad or figshare before manuscript submission, and should include a .cif file, a structural figure with probability ellipsoids, and structure factors for each structure. Both the structure factors and the structural output must have been checked using the IUCR's CheckCIF routine, and a copy of the output must be included at submission, together with a justification for any alerts reported.

| | |
|---|---|
| Protein Circular Dichroism Data Bank (PCDDB) | view FAIRsharing entry |

https://www.nature.com/sdata/policies/repositories#life

# “Long-tail data” repositories

- Research data that doesn’t fit in structured data repositories
- Data publication – persistent identifiers
- Metadata submission – not tailored to Life Science
  - *Affects discoverability*
  - *(Less) FAIR*
- Sensitive data a potential issue

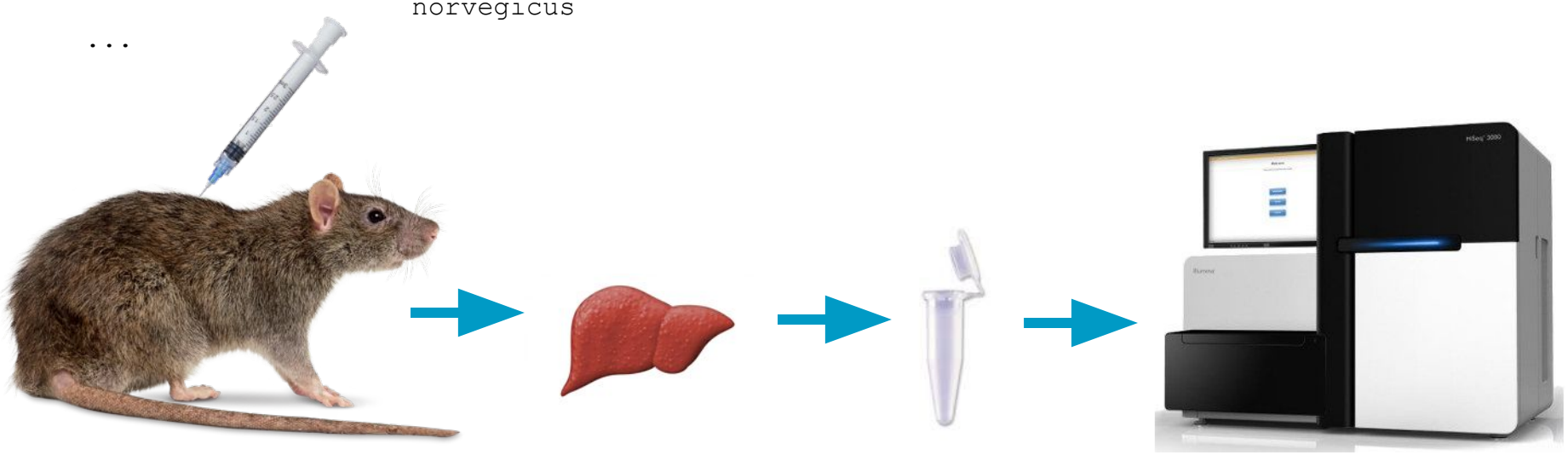



- Figshare - https://figshare.com/
  - scilifelab.figshare.com - **coming soon!**
- EUDAT - http://eudat.eu/
- Data Dryad - http://datadryad.org/
- Zenodo - http://www.zenodo.org/

# Interoperable & Re-useable *Metadata*

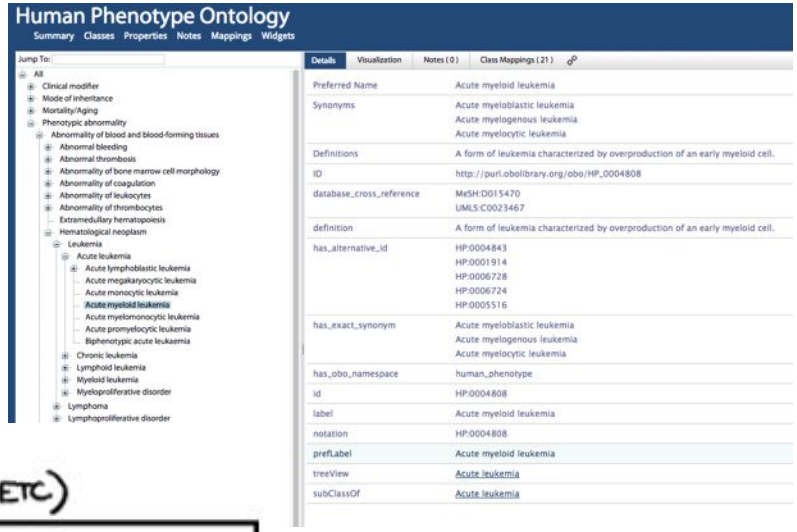
- Standards
  - Repositories have recommended standards
  - Controlled vocabularies / Ontologies

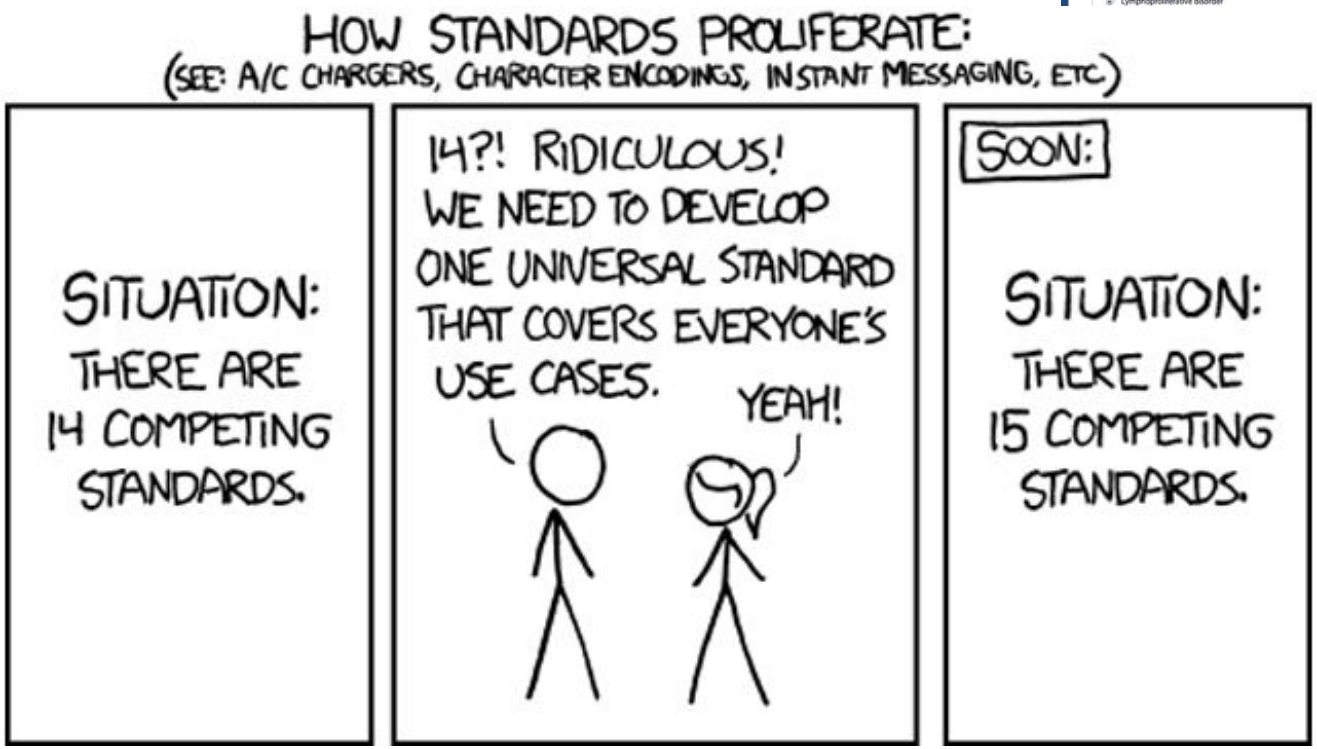
| SampleID | Species | | Strain | | Compound | | Dose |
|---|---|---|---|---|---|---|---|
| | *NCBI:txid* | *SciName* | *MGI_ID* | *name* | *ChEBI_ID* | *name* | *mg* |
| A9876 | 10116 | Rattus norvegicus | *MGI:2161072* | *BALB/c* | *CHEBI:46195* | paracetamol | 10 |
| A6543 | 10116 | *Rattus* norvegicus | *MGI:2161072* | *BALB/c* | null | null | null |
| ... | | | | | | | |



# Interoperable & Re-useable *Metadata*

- Standards
  - Controlled vocabularies / Ontologies
    - Agreed terms for different phenomena



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# FAIRsharing.org

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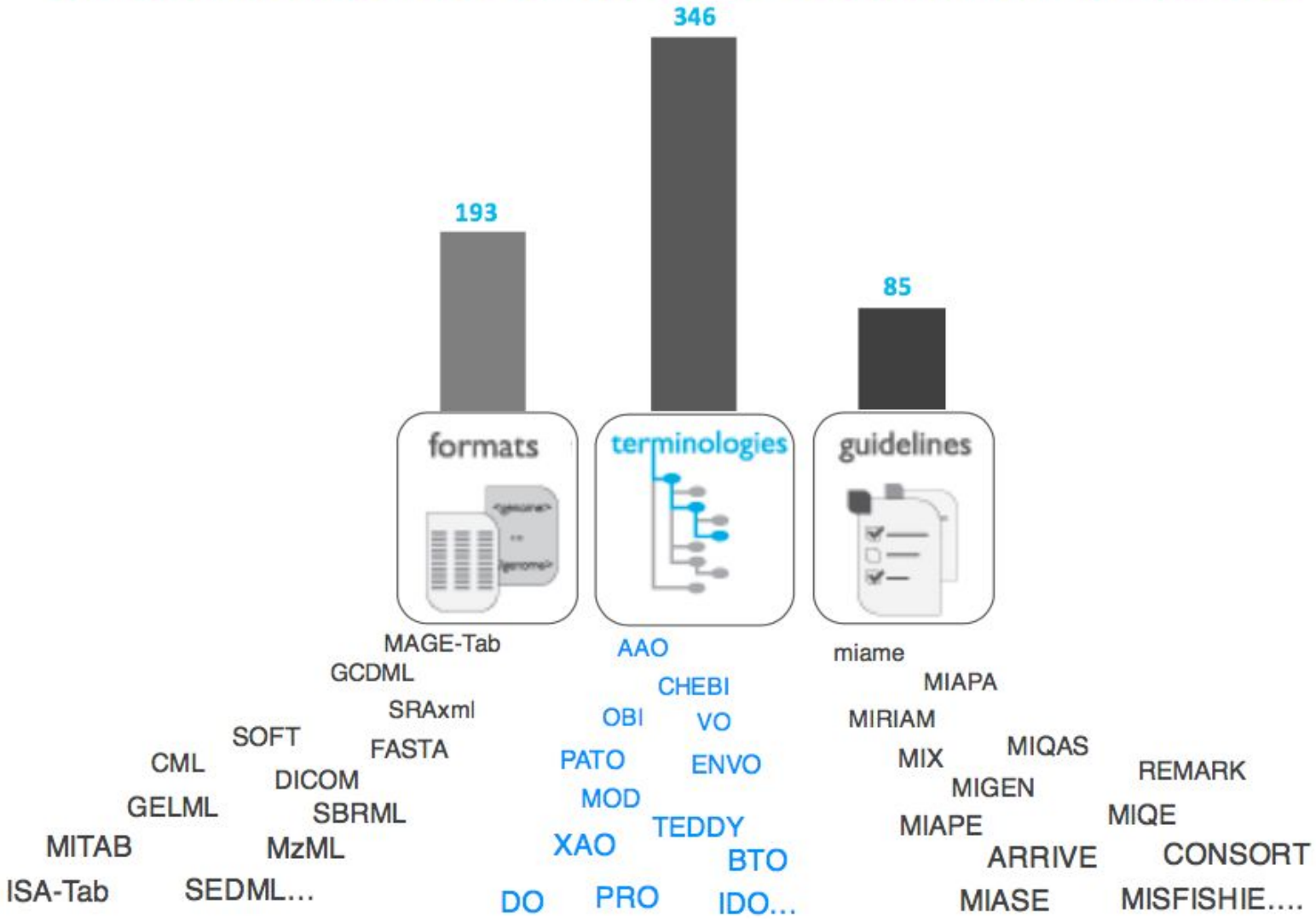
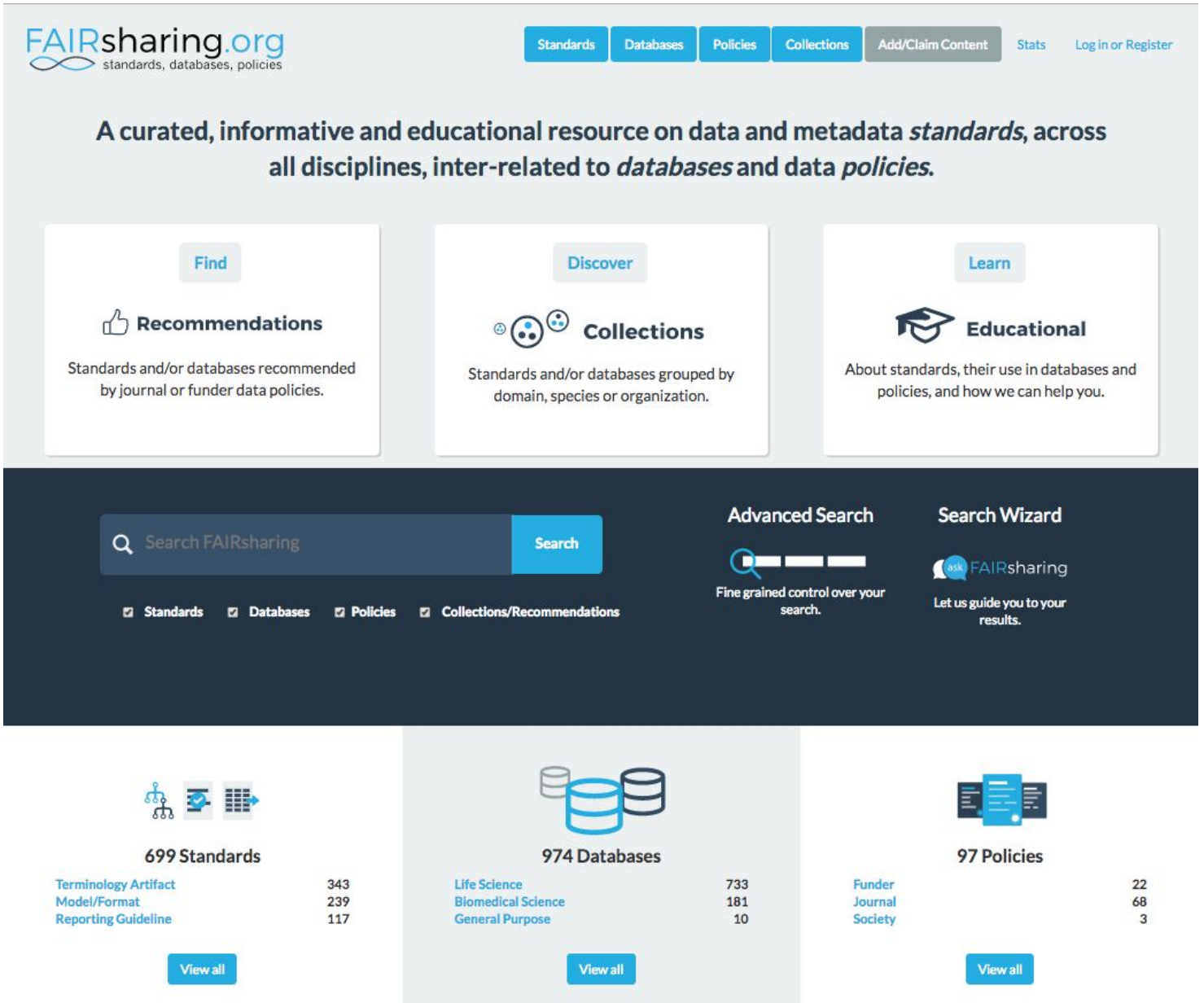
## In the life sciences there are >600 *content standards*



# FAIRsharing.org

*(was biosharing.org)*



# Can you share all types of data publicly?

# If not, what would be the reasons?

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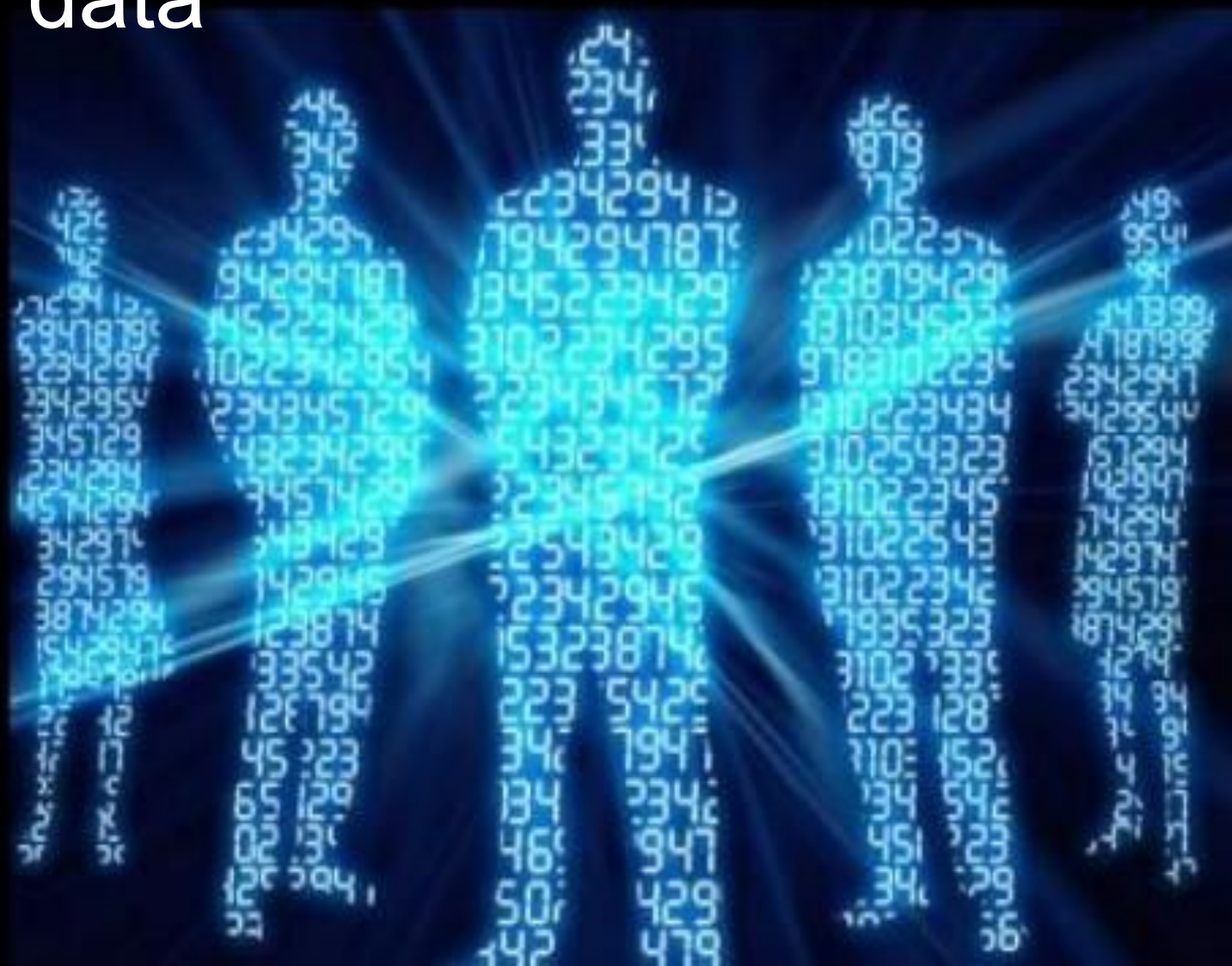
# Personal data



# Personal data - Legislation


- GDPR – General Data Protection Regulation (*Dataskyddsförordningen*) + others
- Act concerning the Ethical Review of Research Involving Humans (*Lag om etikprövning av forskning som avser människor*)

# GDPR

- All kinds of information that is directly or indirectly referable to a natural person who is alive constitute personal data
- To process personal data:
  - *All processing of personal data must fulfil the **fundamental principles** defined in the Regulation, among them are:*
    - Decide a **purpose** and stick to it
    - Identify the **legal basis** for data processing before it starts

- *Have you defined the **purpose** and **legal basis** for handling personal data in your project?*

# GDPR – Sensitive data

- Special categories (*Sensitive data*)
  - … **racial or ethnic origin**, […] **genetic data**, […], data concerning **health** … Art. 9 (1)
  - Processing is **prohibited** unless…
    - **explicit consent** is given Art. 9 (2)a
    - processing is necessary for **scientific research** in accordance with Article 89(1) based on Union or *Member State law* which shall be proportionate to the aim pursued, respect the essence of the right to data protection and provide for suitable and specific measures to safeguard the fundamental rights and the interests of the data subject. Art. 9 (2)j
    - Member State specific conditions and *limitations possible* for processing of health & genetic data Art. 9 (4)
    - **Sweden**
      - Consent?
      - Public interest → Ethical review necessary (often includes consent)

# GDPR – Help

- A **Data Protection Officer** (*dataskyddssombud*)
  - The natural person that is responsible for ensuring that the organization/company adheres to the GDPR
  - Educate
  - Audit
  - Contact point between organization and Data Protection Agency

GU

https://medarbetarportalen.gu.se/projekt-process/aktuella-projekt/dataskyddsforordning

KI

https://ki.se/medarbetare/gdpr-pa-karolinska-institutet

KTH

https://intra.kth.se/anstallning/anstallningsvillkor/att-vara-statligt-an/behandling-av-person/dataskyddsforordningen-gdpr-1.800623

LiU

https://insidan.liu.se/dataskyddsforordningen/anmalan-av-personuppgiftsbehandling?l=sv

LU

https://personuppgifter.blogg.lu.se

SU

https://www.su.se/medarbetare/organisation-styrning/juridik/personuppgifter/dataskyddsf%C3%B6rordningen

UmU

https://www.aurora.umu.se/regler-och-riktlinjer/juridik/personuppgifter/

UU

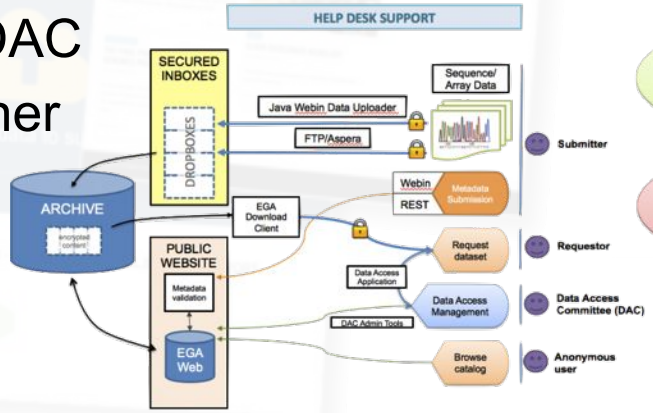
https://mp.uu.se/web/info/stod/dataskyddsforordningen

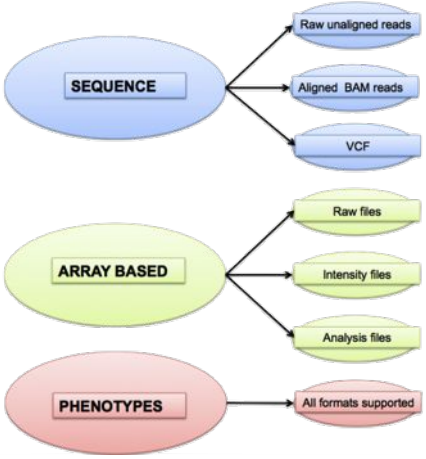
# What about publishing sensitive data?

*"As open as possible, as closed as necessary"*

- **EGA** – European Genome-phenome Archive
  - Repository that promotes the distribution and sharing of **genetic and phenotypic data** consented for specific approved uses but **not fully open, public distribution**.
  - All types of sequence and genotype experiments, including case-control, population, and family studies.

- Data Access Agreement
  - Defined by the data owner
- Data Access Committee – DAC
  - Decided by the data owner



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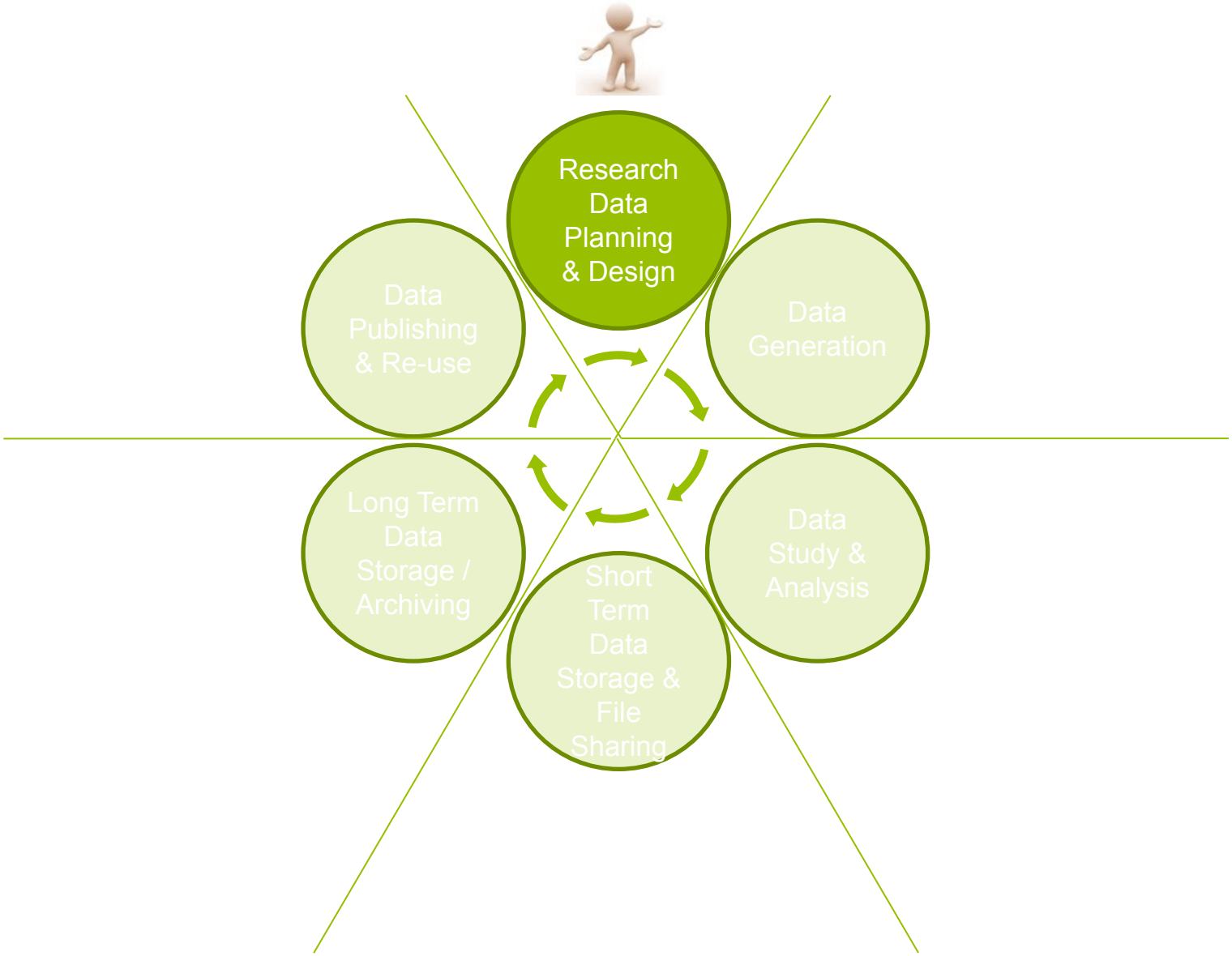
# When should you start planning for how to manage you data?

# Planning & Design



# Data Management Plans

*Will become a standard part of the research funding application process*

- **Data collection** - data types and volumes, analysis code
- **Data organization** - folder and file structure, and naming
- **Data documentation** - data and analysis, metadata standards
- **Data storage** - storage/backup/protection & time lines
- **Data policies** - conditions/licences for using data & legal/ethical issues
- **Data sharing** - *When* and *How* will *What* data (and code) be shared
- **Roles and responsibilities** - who's responsible for what & is competence available
- **Budget** - People & Hardware/Software

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EDITORIAL · 13 MARCH 2018

Everyone needs a data-management plan

*They sound dull, but data-management plans are essential, and funders must explain why.*

**By 2019, all who receive grants from us must have a data management plan**

As from spring 2019, if you are awarded a grant from the Swedish Research Council you must have a plan for how the research data generated within your project shall be managed.

You must not send in your data management plan to us when you apply for a grant, but your administrating organisation will be responsible for ensuring that a data management plan is in place when you start your project or corresponding, and that the plan is maintained.

# Data Management Plans


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# National guidelines

- VR & SUHF (Association of Swedish Higher Education Institutions)
  - *Work in progress*

- **Central parts of a data management plan**
  - Based on Science Europe's "Core Requirements for Data Management Plans"

1. Description of data – reuse of existing data and/or production of new data
2. Documentation and data quality
3. Storage and backup
4. Legal and ethical aspects
5. Accessibility and long-term storage
6. Responsibility and resources

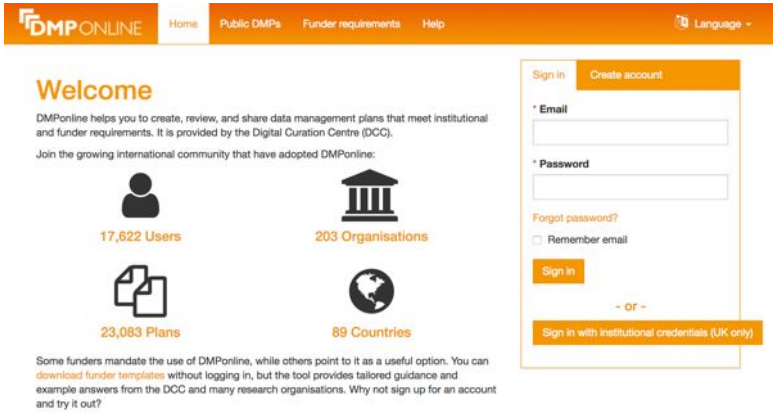
https://www.vr.se/english/calls-and-decisions/grant-terms-and-conditions/data-management-plan.html

# Planning ahead

*Consider structuring metadata in the format needed by the repository already at planning stage*

# DMP tools

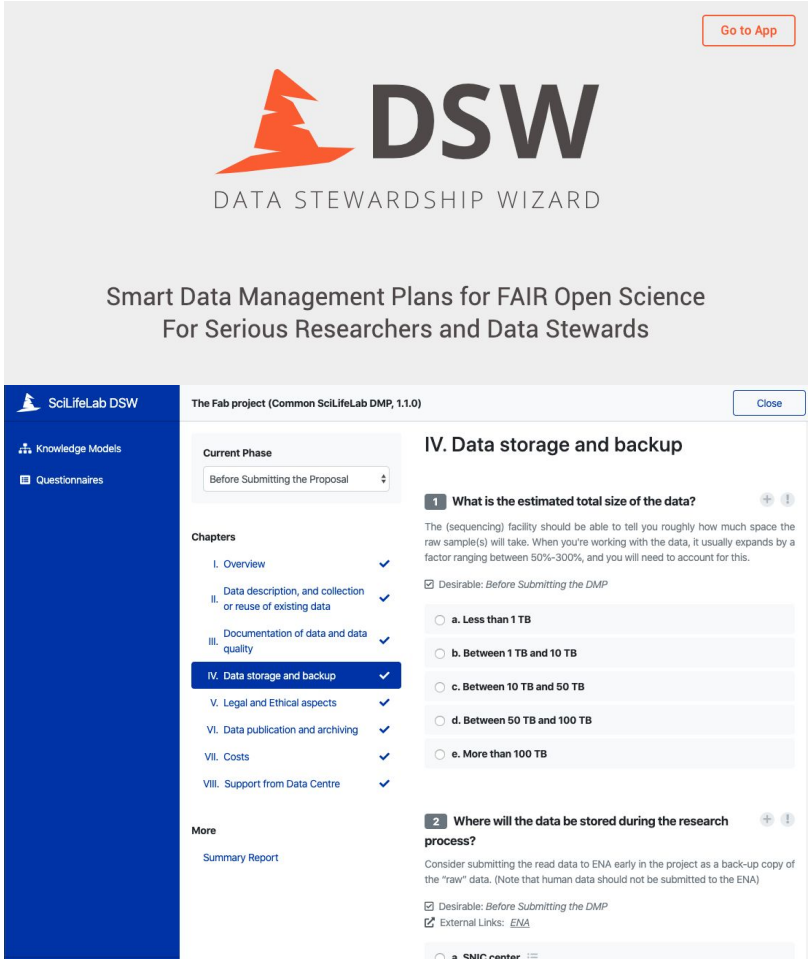
## DMPonline





https://dmponline.dcc.ac.uk/

## ELIXIR Data Stewardship Wizard




https://ds-wizard.org/

https://dsw.scilifelab.se

# NBIS Data Management support

- Project planning
  - Metadata
  - File formats
  - Licensing
  - *Data Management Plans*
- Data analysis
- Data publication and submission
  - Support submissions to public repositories
  - Metadata
  - DOIs to dataset (if needed)

# Take home messages

- Consider doing a Data Management Plan for your project
  - How do you ensure that your research output is FAIR?
- Plan for submitting ”raw data” to public repositories as early as possible
- Organize project metadata from the start
  - In ways that makes it easy to submit to public repositories
  - Use available standards
- Pick a thought-through file and folder structure organization for your computational analyses
- Strive for reproducibility
  - Data & Code
- Be aware that there are legal aspects to processing human data
- *Ask for help if you need it!*

# Source Acknowledgements

- Research Data Management, EUDAT - http://hdl.handle.net/11304/79db27e2-c12a-11e5-9bb4-2b0aad496318
- Noble WS (2009) A Quick Guide to Organizing Computational Biology Projects. PLoS Comput Biol 5(7): e1000424. doi:10.1371/journal.pcbi.1000424
- Reproducible research
  - Reproducible Science Curriculum – https://github.com/Reproducible-Science-Curriculum/rr-init
  - Leif Väremo & Rasmus Ågren
    - https://bitbucket.org/scilifelab-lts/reproducible_research_example/src
    - https://nbis-reproducible-research.readthedocs.io/en/latest/
- GDPR
  - Datainspektionen – https://www.datainspektionen.se/lagar--regler/dataskyddsforordningen/
- … and probably others I have forgotten