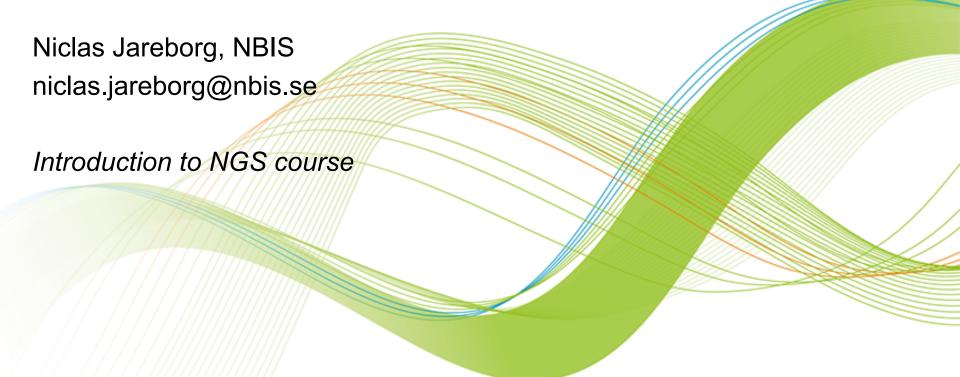




Research Data Management





Why manage research data?



- To make your research easier!
- To stop yourself drowning in irrelevant stuff
- In case you need the data later
- To avoid accusations of fraud or bad science
- To share your data for others to use and learn from
- To get credit for producing it
- Because funders or your organisation require it

Well-managed data opens up opportunities for re-use, integration and new science





Accusation of fraud



Science

Cite as: J. Berg., Science 10.1126/science.aan5763 (2017).

Editorial Retraction

Jeremy Berg

Editor-in-Chief

After an investigation, the Central Ethical Review Board in Sweden has recommended the retraction of the Report "Environmentally relevant concentrations of microplastic particles influence larval fish ecology," by Oona M. Lönnstedt and Peter Eklöv, published in Science on 3 June 2016 (I). Science ran an Editorial Expression of Concern regarding the Report on 1 December 2016 (2). The Review Board's report, dated 21 April 2017, cited the following reasons for their recommendation: (i) lack of ethical approval for the experiments; (ii) absence of original data for the experiments reported in the paper; (iii) widespread lack of clarity concerning how the experiments were conducted. Although the authors have told Science that they disagree with elements of the Board's report, and although Uppsala University has not yet concluded its own investigation, the weight of evidence is that the paper should now be retracted. In light of the Board's recommendation and a 28 April 2017 request from the authors to retract the paper, Science is retracting the paper in full.

REFERENCES

O. M. Lönnstedt, P. Eklöv, Science 352, 1213 (2016).
 J. Berg, Science 354, 1242 (2016); published online 1 December 2016.

Published online 3 May 2017 10.1126/science.aan 5763

- Be able to show that you have done what you say you have done
- Universities want to avoid bad press!



Why manage research data?



- To make your research easier!
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Well-managed data opens up opportunities for re-use, integration and new science





Open Access to research data



- The practice of providing on-line access to scientific information that is free of charge to the end-user and that is re-usable.
 - Does not necessarily mean unrestricted access, e.g. for sensitive personal 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]









Why 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



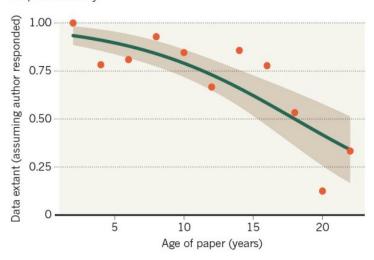


Data loss is real and significant, while data growth is staggering



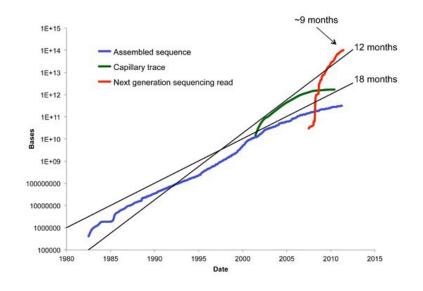
MISSING DATA

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



Nature news, 19 December 2013



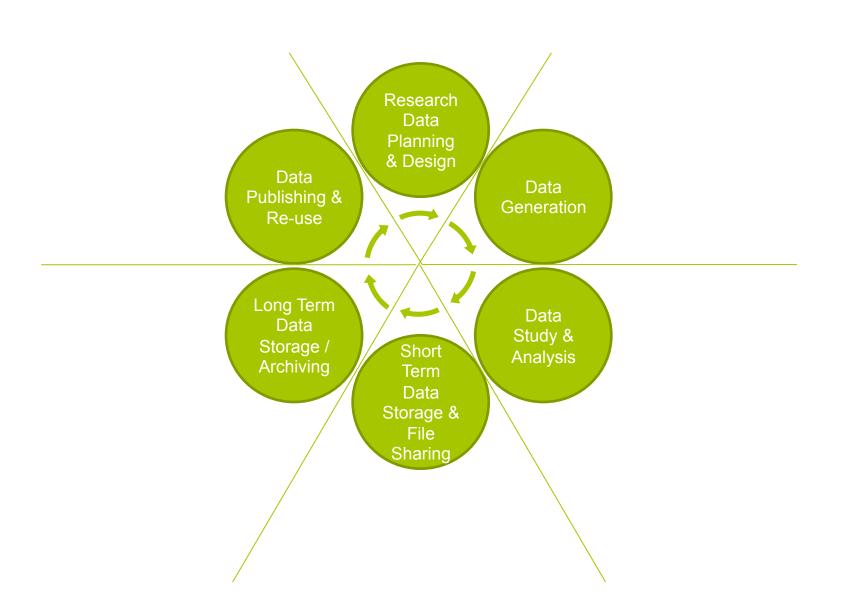


- DNA sequence data is doubling every
 6-8 months and looks to continue for this decade
- Projected to surpass astronomy data in the coming decade



The Research Data Life Cycle SciLifeLab

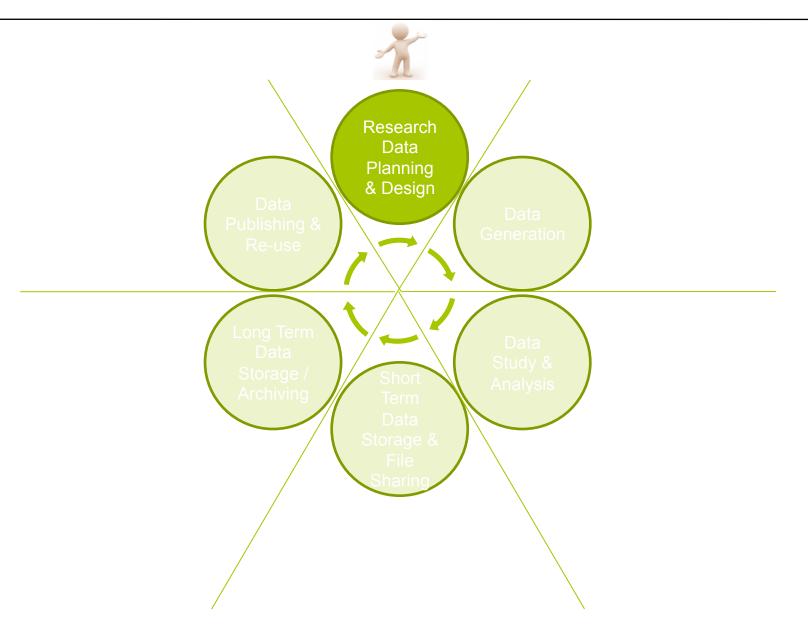






Planning & Design







Planning & Design



- Data Management planning
 - Data types
 - Sizes, were to store, etc
 - Metadata
 - Study, Samples, Experiments, etc
 - Use standards!

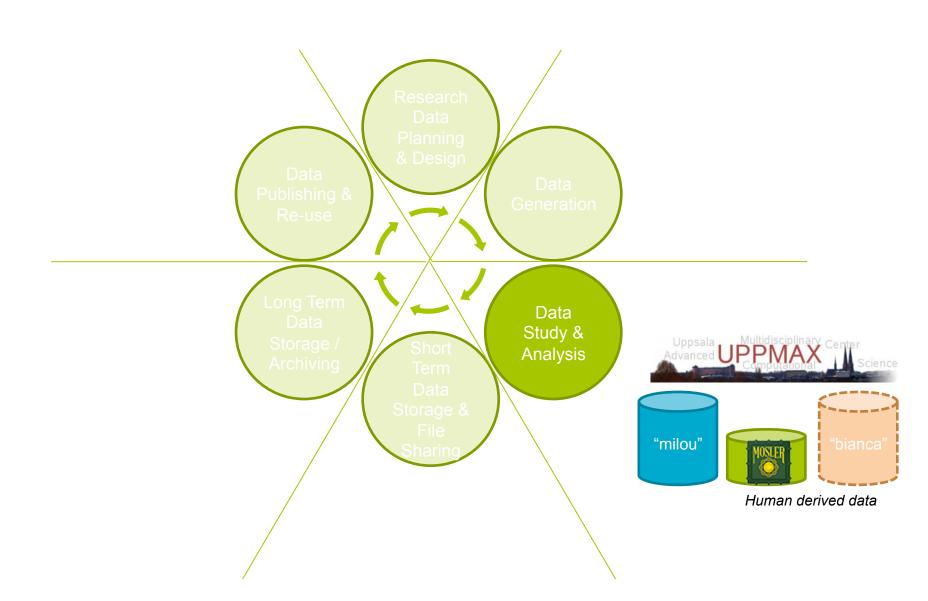


- Data Management Plans
 - Will become a standard part of the research funding application process
 - What will be collected?, Size?, Organized?, Documented?, Stored and preserved?, Disseminated?, Policies?, Budget?



Study & Analysis







Study & 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

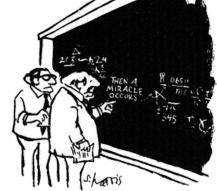


♣ Follow

My rule of thumb: every analysis you do on a dataset will have to be redone 10–15 times before publication. Plan accordingly. #Rstats

- Structuring data for analysis
 - Poor organizational choices lead to significantly slower research progress.
 - It is critical to make results reproducible.



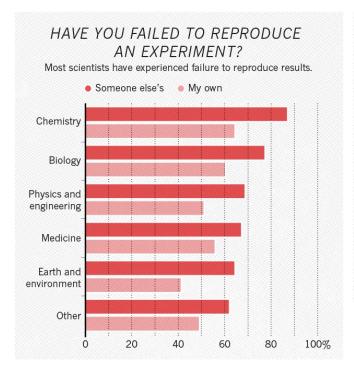


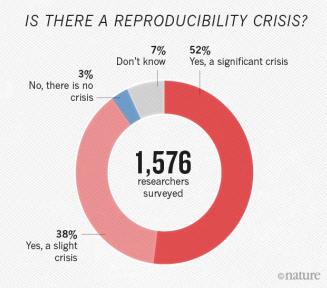
"I think you should be more explicit here in step two.



A reproducibility crisis







A recent survey in Nature revealed that irreproducible experiments are a problem across all domains of science¹.

Medicine is among the most affected research fields. A study in Nature found that 47 out of 53 medical research papers focused on cancer research were irreproducible².

Common features were failure to show all the data and inappropriate use of statistical tests.

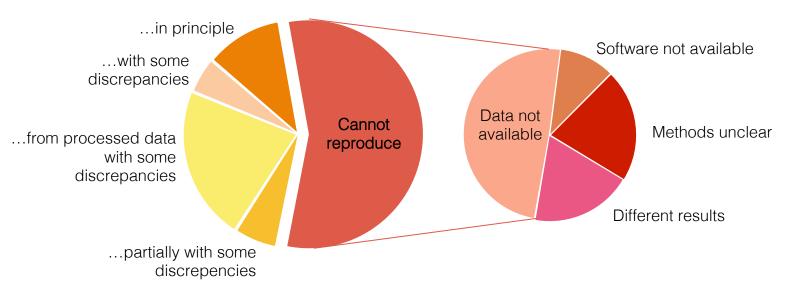


A reproducibility crisis



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

Can reproduce...



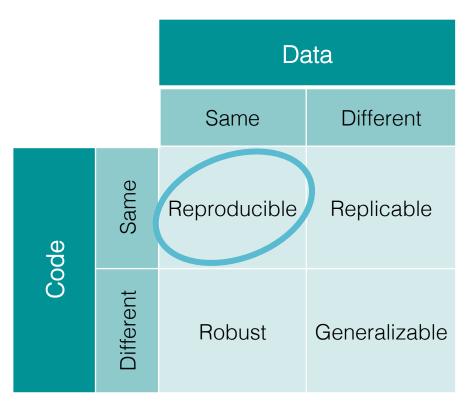
Summary of the efforts to replicate the published analyses.

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?



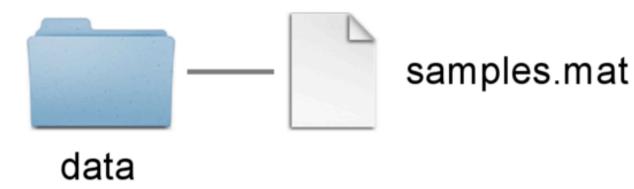


Is it really any point doing this?

- Primarily for ones own benefit!
 Organized, efficient, in control.
 Dynamic team members.
- Transparent what has been done
- Some will be interested in parts of the analysis. Make it easy to redo, then adapt to own data.



Organization - So far, so good...SciLifeLab





Now what?





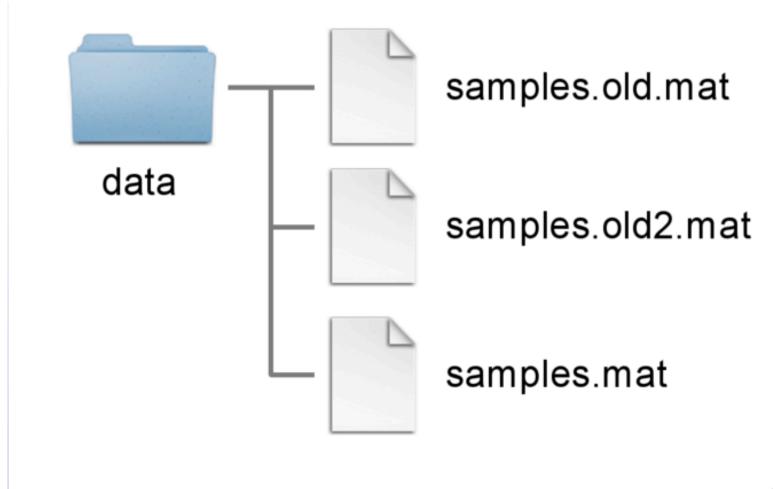
samples.old.mat

samples.mat



I guess this is alright

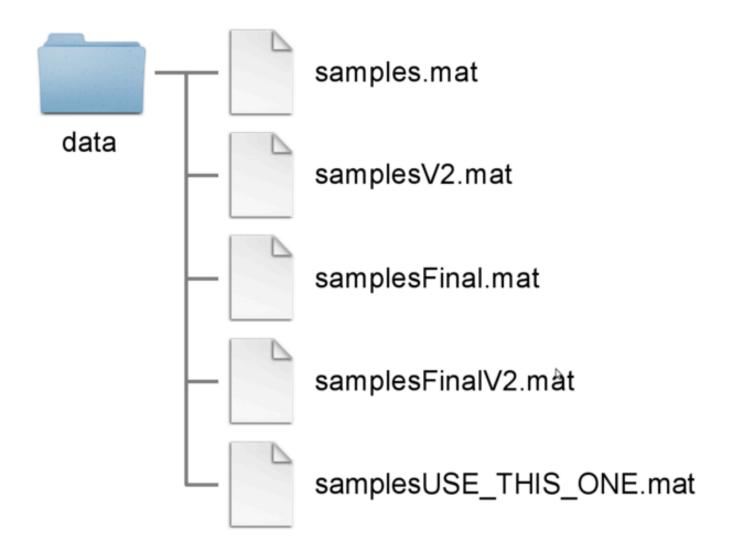






Which one is the most recent? SciLifeLab

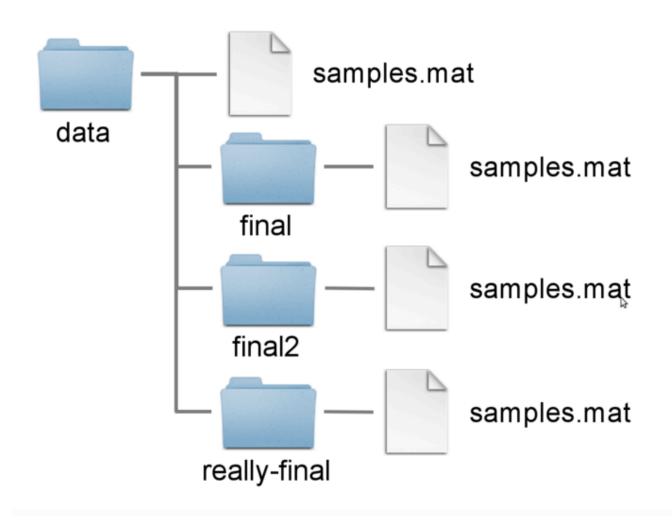






Another (bad) common approach SciLifeLab

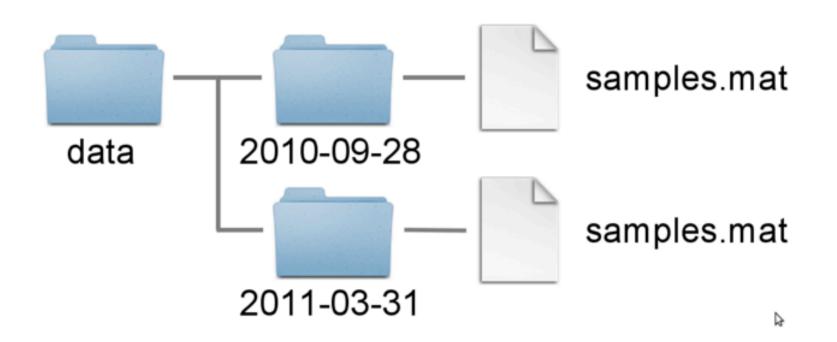






A possible solution







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 **non-proprietary formats** .csv rather than .xlsx
- Etc...



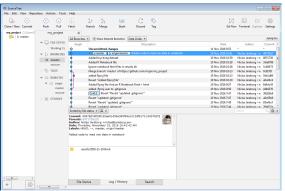
Version control



- 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









Suggested best practices



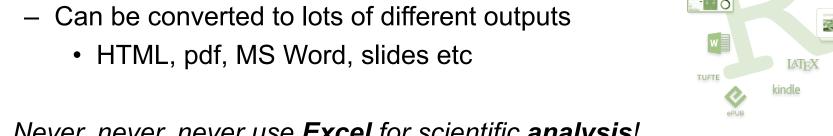
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- Etc...



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



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



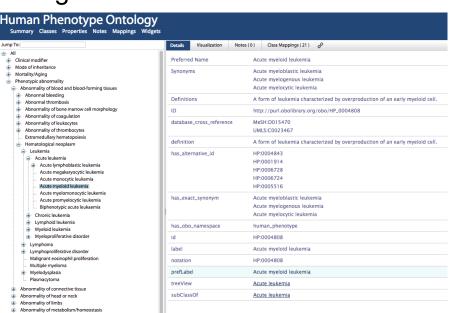
明神



Still missing something



- Need context → document metadata
 - How was the data generated?
 - From what was the data generated?
 - What where the experimental conditions?
 - Etc
- Use standards
 - Controlled vocabularies / Ontologies
 - Not straight-forward...

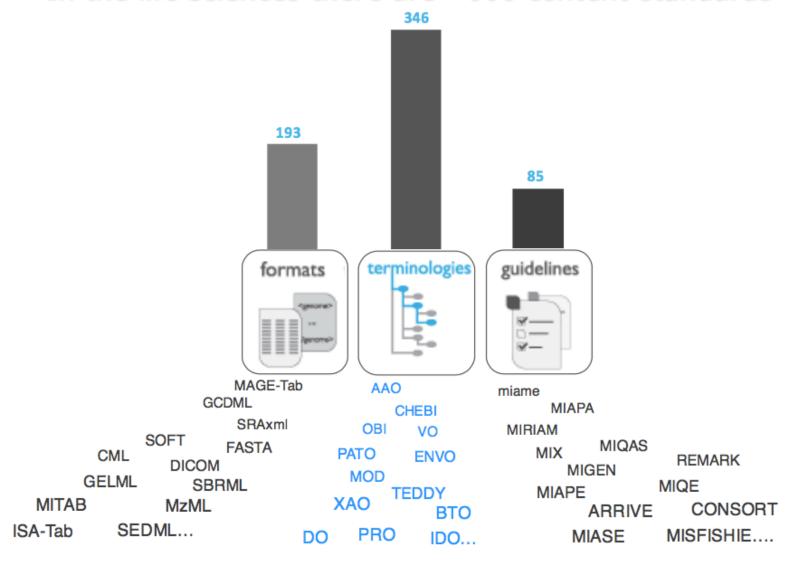




FAIRsharing.org



In the life sciences there are >600 content standards

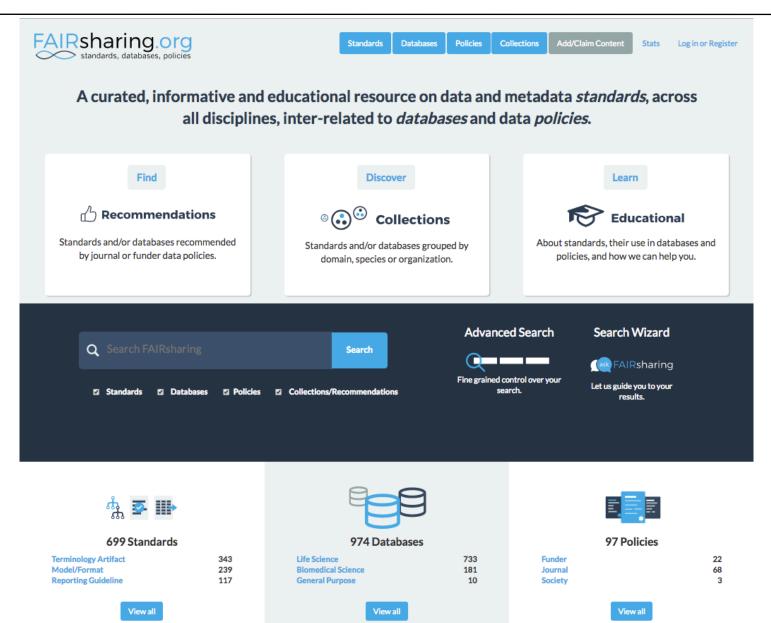




FAIRsharing.org





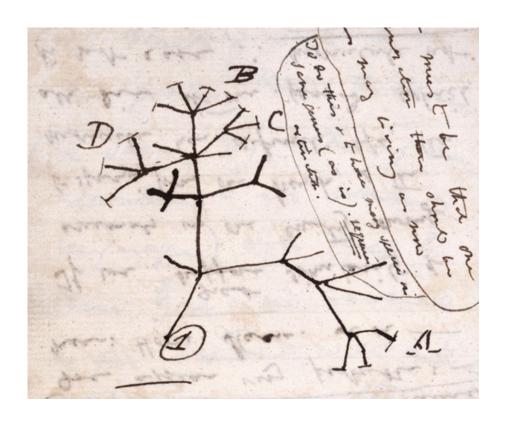




Lab notebooks



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





Lab notes – useful practices



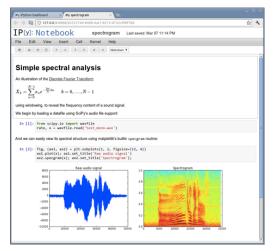
- Put in results directory
- Dated entries
- Entries relatively verbose
- Link to data and code (including versions)
 - Point to commands run and results generated
- Embedded images or tables showing results of analysis done
- Observations, Conclusions, and ideas for future work
- Also document analysis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

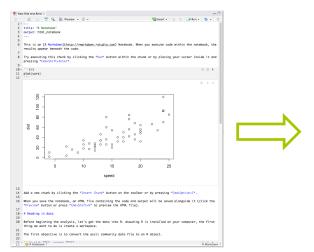


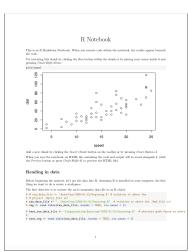
Where to keep notes



- Paper Notebook
- Word processor program / Text files
- Electronic Lab Notebooks
- 'Interactive' Electronic 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



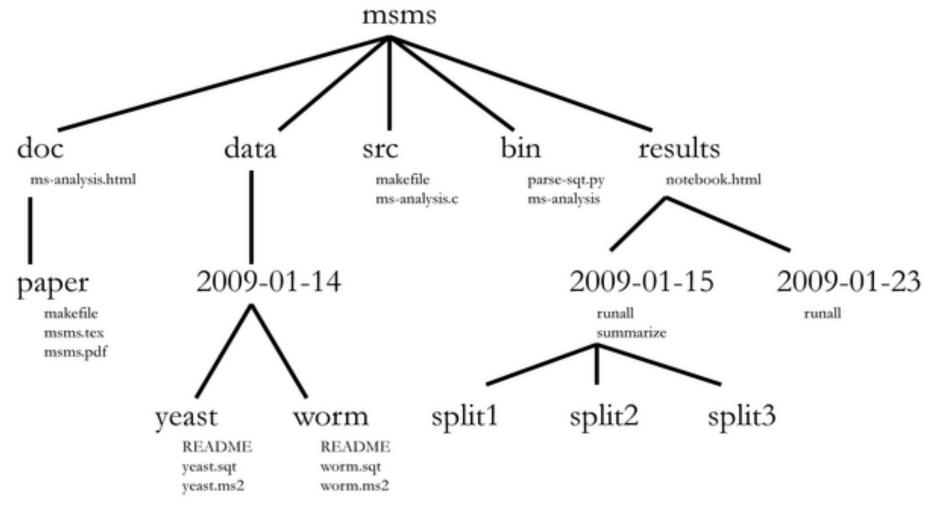






Directory structure for a sample project





Noble WS (2009) A Quick Guide to Organizing Computational Biology Projects. PLoS Comput Biol 5(7): e1000424. doi:10.1371/journal.pcbi.1000424

http://journals.plos.org/ploscompbiol/article?id=info:doi/10.1371/journal.pcbi.1000424





Or like this...



```
bin <----# Binary files and executables (jar files & proj-wide scripts etc)
conf <----# Project-wide configuration</pre>
doc <----# Any documents, such as manuscripts being written
experiments <----# The main experiments folder
 — 2000-01-01-exa <-# An example Experiment</p>
      - audit <----# Audit logs from workflow runs (higher level than normal logs)</pre>
      - bin <----# Experiment-specific executables and scripts</p>
      - conf <----# Experiment-specific config</pre>
      - data <----# Any data generated by workflows</p>
      log <----# Log files from workflow runs (lower level than audit logs)</pre>
      - raw <----# Raw-data to be used in the experiment (not to be changed)
       results <---# Results from workflow runs
            <----# All files rel. to running experiment: Workflows, run confs/scripts...
            <----# Any temporary files not supposed to be saved
raw <----# Project-wide raw data
results <----# Project-wide results
    <----# Project-wide source code (that needs to be compiled)</pre>
```

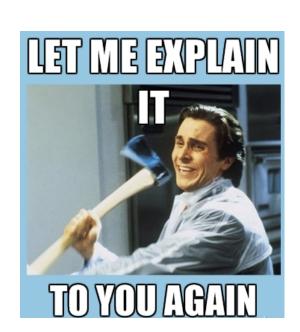
From Samuel Lampa's blog: http://bionics.it/posts/organizing-compbio-projects



Project organization



- There's no perfect set-up
 - Pick one! e.g.
 - https://github.com/chendaniely/computational-project-cookie-cutter
 - https://github.com/Reproducible-Science-Curriculum/rr-init
 - https://github.com/nylander/ptemplate
 - ...
- Communicate structure to collaborators
- Document as you go
- Done well it might reduce post-project explaining



Reproducible research for bioinformatics projects

Everything can be a project

Divide your work into distinct projects and keep all files needed to go from raw data to final results in a dedicated directory with relevant subdirectories (see example).

Many software support the "project way of working", e.g. Rstudio and the text editors Sublime Text and Atom.

Tip! Learn how to use git, a widely used system (both in academia and industry) for version controlling and collaborating on code.

https://git-scm.com/

data sample1.fastq sample2.fastq intermediate bam

counts.csv results fig1.pdf

fig2.pdf source

align.py

fig1.py fig2.py

sumcounts.py

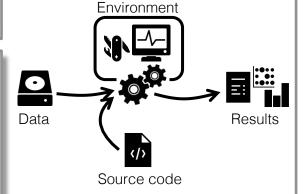
Treasure your data

- Consider your input data static. Keep it readonly!
- Don't make different versions. If you need to preprocess it in any way, script it so you can recreate the steps (see box below).
- Backup! Keep redundant copies in different physical locations.
- Strive towards uploading it to its final destination already at the beginning of a project (e.g. specific repositories such as ENA, or GeneExpress, or general repositories such as Dryad or Figshare).

Take control of your research by making it reproducible!

By moving towards a reproducible way of working you will quickly realize that you at the same time make your own life a lot easier! The added effort pays off by gain in control, organization and efficiency.

Below are all the components of a bioinformatics project that have to reproducible.



Organize your coding

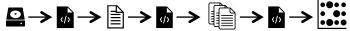
- Write scripts/functions/notebooks for specific tasks (connect raw data to final results)
- Keep parameters separate (e.g. top of file, or input arguments)



data script1



Avoid generating files interactively on the fly or doing things by hand (no way to track how they were made).



data script2A fileA script2B filesC,D,Escript2C Figure 2

For the advanced

As projects grow, it becomes increasingly difficult to keep track of all the parts and how they fit together. Snakemake is a workflow management system that keeps track of how your files tie together, from raw data and scripts to final figures. If anything changes (script code, parameters, software version, etc) it will know what parts to rerun in order to have up to date and reproducible results.



Leif Väremo (leif.varemo@scilifelab.se) Rasmus Ågren (rasmus.agren@scilifelab.se) Bioinformatics long-term support (WABI)

Connect your results with the code

Rmarkdown and Jupyter notebooks blur the boundaries between code and its output. They allow you to add non-code text (markdown) to your code. This generates a report containing custom formatted text, as well as figures and tables together with the code that generated _ jupyter them.

http://rmarkdown.rstudio.com/

http://jupyter.org/

Master your dependencies

- Full reproducibility requires the possibility to recreate the system that was originally used to generate the results.
- Conda is package, dependency, and env-ironment manager that makes it easy to install (most) software that you need for your project.
- Your environment can be exported in a simple text format and reinstalled by Conda on another system.



https://conda.io

For the advanced

- Conda cannot always completely recreate the system, which is required for proper repro-ducibility.
- A solution is to package your project in an isolated Docker container, together with all its dependencies and libraries.
- A vision is that every new bioinformatics publication is accompanied by a publically available Docker container!
- Singularity is an alternative to Docker which runs better on HPC clusters.





https://www.docker.com/

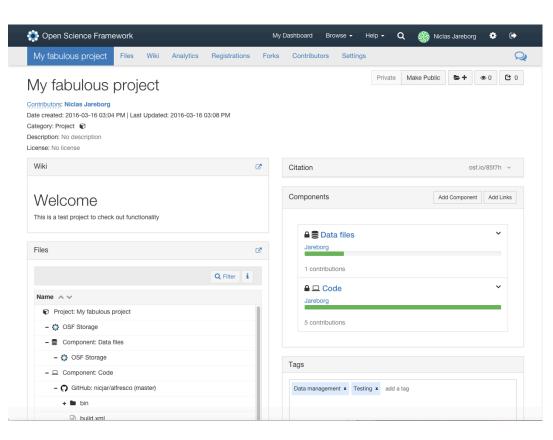
http://singularitv.lbl.gov/



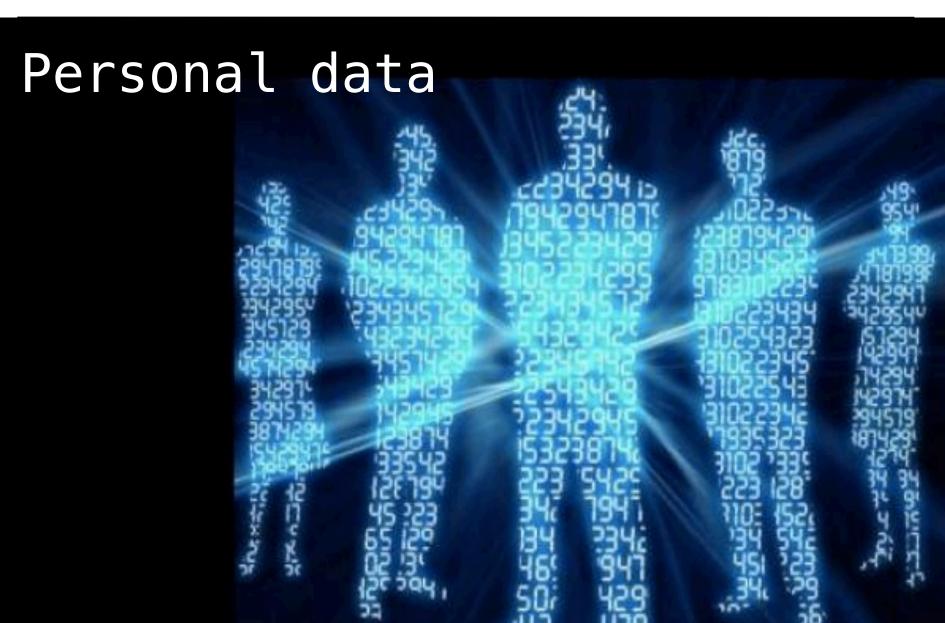
Project collaboration tools



- Open Science Framework http://osf.io
 - Organize research project documentation and outputs
 - Control access for collaboration
 - 3rd party integrations
 - Google Drive
 - Dropbox
 - GitHub
 - External links
 - Etc
 - Persistent identifiers
 - Publish article preprints









Personal data - Legislation







Personal Data Act



- All kinds of information that is directly or indirectly referable to a natural person who is alive constitute personal data
- Sensitive data
 - It is prohibited to process personal data that discloses ethnic origin, political opinions, religious or philosophical convictions, membership of trade unions, as well as personal data relating to health or sexual life.
 - Sensitive personal data can be handled for research purposes if person has given explicit consent
- The Data Inspection Board (Datainspektionen) is the supervisory authority under the Personal Data Act
- May 2018: General Data Protection Regulation (GDPR)
 - New Swedish laws
 - Data Protection Act (Dataskyddslag)
 - Research Data Act (Forskningsdatalag)
 - Several other



Personal Data Act – cont.



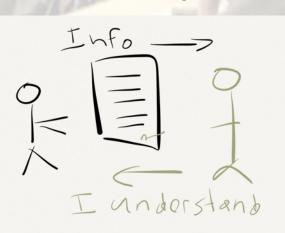
- The (legal) person that decides why and how personal data should be processed is called the controller of personal data (personuppgiftsansvarig)
 - e.g. the employing university
- The controller of personal data can delegate processing of personal data to a personal data assistant (personuppgiftsbiträde)
 - e.g. UPPMAX/Uppsala university
- A personal data representative (personuppgiftsombud) is a natural person who, on the assignment of the controller, shall ensure that personal data is processed in a lawful and proper manner
- Obligation to report handling of personal data to the Data Inspection Board
 - Or, notify the Board of the named representative



Act concerning the Ethical Review



- Research that concerns studies of biological material that has been taken from a living person and that can be traced back to that person may only be conducted if it has been approved subsequent to an ethical vetting
- Informed consent
 - The subject must be informed about the purpose or the research and the consequences and risks that the research might entail
 - The subject must consent





Genetic information



- The genetic information of an individual is personal data
 - Sensitive personal data (as it relates to health)
 - Explicitly defining in GDPR
 - Even if anonymized / pseudonymized
 - In principle, no difference between WGS, Exome, Transcriptome or GWAS data
- Theoretically possible to identify the individual person from which the sequence was derived from the sequence itself
 - The more associated metadata there is, the easier this gets
 - Gymrek et al. "Identifying Personal Genomes by Surname Inference". Science 339, 321 (2013); DOI:10.1126/science.1229566
- "The controller is liable to implement technical and organizational measures to protect the personal data. The measures shall attain a suitable level of security."



Bianca & Mosler



Bianca

- Swedish Research Council funded SNIC Sens project
- Implemented by SNIC/UPPMAX
- 3200 cores / 1 PB
- Opened april 2017

https://uppmax.uu.se/resources/systems/the-bianca-cluster/

Mosler

- e-Infrastructure for working with sensitive data for academic research
 - Developed & operated by NBIS
- Inspired by Norwegian solution (TSD)
- Designed to look like UPPMAX clusters
 - UPPMAX modules
 - UPPMAX can assist with installing custom tools
- Implementation project completed Nov 2015
- "Pilot-size system"
- 24 nodes, 270 TB



Provide users with a compute environment for sensitive data, with an appropriate level of security



Mosler



- High-performance computing in a virtualized environment (OpenStack)
 - Each project environment is isolated from all other projects
 - Separated private networks and file systems
 - No internet access
 - No root access
- Only accessible over remote Linux desktop (ThinLinc) via a web dashboard
- 2-factor authentication for login
- Restricted data transfer in/out
 - Via a file gateway
 - Project members can transfer IN / only PI allowed to transfer out
 - Not possible to copy/paste out







TRYGGVE2: NORDIC COLLABORATION ON E-INFRASTRUCTURE FOR SENSITIVE DATA

https://wiki.neic.no/tryggve

TRYGGVE2 VISION AND KEY TARGETS

Tryggve2 develops and facilitates access to secure e-infrastructure for conducting large-scale cross-border biomedical research studies utilizing sensitive data

Key targets

- Develop state-of-the-art infrastructure
- Support wide range of use cases
- Improve cross-border access and usability of services
- Leverage Nordic secure resources and expertise



TRYGGVE2 MAJOR DELIVERABLES

COLLABORATION FOR SENSITIVE BIOMEDICAL DATA

- I. Sensitive data archiving
- 2. Production quality processing services
- 3. Homogenized user experience
 - I. User mobility
 - 2. Workflow mobility
 - 3. Data mobility
- 4. Nordic use cases
 - I. Research
 - 2. Infrastructure development
- 5. ELIXIR AAI
- 6. IT Security
- 7. ELSI Topics

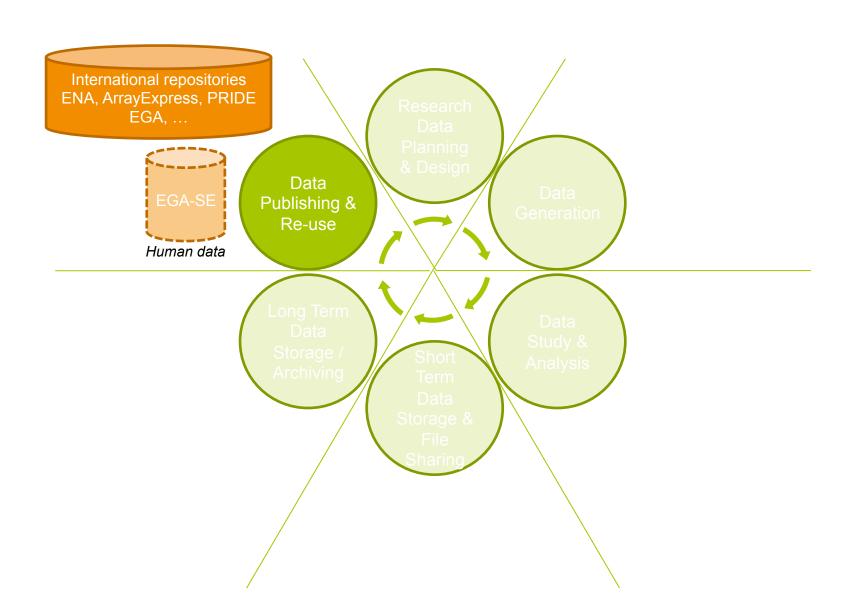






Data Publishing & Re-use







Data Publishing & Re-use



Research Data Publishing is a cornerstone of Open Access



- 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



Long tradition of data publication



- DNA sequence databases: Genbank and EMBL db 1982
- Protein structures: PDB 1969

Proc. Natl. Acad. Sci. USA Vol. 86, p. 408, January 1989 Data Submission

1989

Submission of data to GenBank

CHRISTIAN BURKS AND LAURIE J. TOMLINSON

Theoretical Biology and Biophysics Group T-10, MS K710, Los Alamos National Laboratory, Los Alamos, NM 87545

In response to both the ever-increasing rate of determining nucleotide sequences (1) and the growing trend among journals to allow articles to appear that describe the results of determining a sequence without explicitly presenting the sequence (1), GenBank* (2-5) and a number of the journals that publish nucleotide sequence data are working together to promote the direct, timely submission of nucleotide sequence data to GenBank. The policy being established by the PROCEEDINGS is described in the editorial on p. 407; here, we will provide a brief summary, in the context of this policy, of

Electronic file transfer. Files can be network to the network GenBank submit above. This address—in most cases with can be reached from various networks, ARPANET, USENET, JANET, JUNET, etc. It work or system expert how to send electrous for help. Floppy disks. We can read in or 51/4-in diskettes written on MS-DOS sethat the submitted data be written as flat to in a format specific to a given word

"The author will provide the accession number to the PROCEEDINGS [PNAS] office to be **included in a footnote to the published paper**."

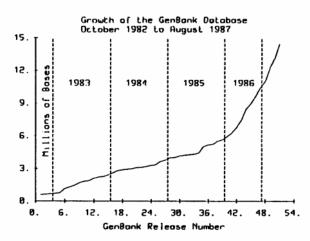


Figure 1.

Bilofsky & Burks (1988) Nucleic Acids Research v16 n5



Human genome project



Bermuda Principles for sharing DNA sequence data

- Automatic release of sequence assemblies larger than 1 kb (preferably within 24 hours).
- Immediate publication of finished annotated sequences.
- Aim to make the entire sequence freely available in the public domain





Data persistency issues

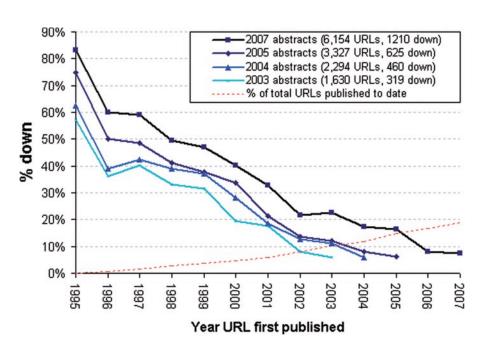


URL decay in MEDLINE—a 4-year follow-up study



+ Author Affiliations

Received January 22, 2008. Revision received March 11, 2008. Accepted April 6, 2008.



 Link rot – more 404 errors generated over time

 Reference rot* – link rot plus content drift i.e.
 webpages evolving and no longer reflecting original content cited

Jonathan D. Wren Bioinformatics 2008;24:1381-1385

^{*}To whom correspondence should be addressed.

^{*} Term coined by Hiberlink http://hiberlink.org



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



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:

- 11. (meta)data use a formal, accessible, shared, and broadly applicable language for knowledge representation.
- 12. (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

DOI: 10.1038/sdata.2016.18





G20 HANGZHOU SUMMIT

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

HANGZHOU, CHINA 4-5 SEPTE





European Commission



Ref. Ares(2017)3488418 - 11/07/201

- European Open Science Cloud EOSC
 - Enable trusted access to services, systems and the re-use of shared scientific data across disciplinary, social and geographical borders.
 - FAIR principles are a cornerstone of EOSC





Brussels, 10 July 2017

EOSC Declaration

RECOGNISING the challenges of data driven research in pursuing excellent science;

GRANTING that the vision of European Open Science is that of a research data commons, widely inclusive of all disciplines and Member States, sustainable in the long-term,

CONFIRMING that the implementation of the EOSC is a process, not a project, by its nature iterative and based on constant learning and mutual alignment;

UPHOLDING that the EOSC Summit marked the beginning and not the end of this process, one based on continuous engagement with scientific stakeholders, the European Commission,

<u>PROPOSES</u> that all EOSC stakeholders consider sharing the following intents and will actively support their implementation in the respective capacities:

Data culture and FAIR data

- [Data culture] European science must be grounded in a common culture of data stewardship, so that research data is recognised as a significant output of research and is appropriately curated throughout and after the period conducting the research. Only a considerable cultural change will enable long-term reuse for science and for innovation of data created by research activities: no disciplines, institutions or countries must be left behind.
- [Open access by-default] All researchers in Europe must enjoy access to an open-by-default, efficient and cross-disciplinary research data environment supported by FAIR data principles. Open access must be the default setting for all results of publicly funded research in Europe, allowing for proportionate limitations only in duly justified cases of personal data protection, confidentiality, IPR concerns, national security or similar (e.g. 'as open as possible and as closed as necessary').
- [Skills] The necessary skills and education in research data management, data stewardship and data science should be provided throughout the EU as part of higher education, the training system and on-the-job best practice in the industry. University associations, research organisations, research libraries and other educational brokers play an important role but they need substantial support from the European Commission and the Member States.
- [Data stewardship] Researchers need the support of adequately trained data stewards. The European Commission and Member States should invest in the education of data stewards via career programmes delivered by universities, research institutions and other trans-European agents.
- [Rewards and incentives] Rewarding research data sharing is essential. Researchers who make research data open and FAIR for reuse and/or reuse and reproduce data should be rewarded, both



International public repositories





























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



ELIXIR Deposition Database list



Deposition Database	Data type	International collaboration framework ¹	Deposition Database	Data type	International collaboration framework ¹
ArrayExpress	Functional genomics data. Stores data from high- throughput functional genomics experiments.		PDBe	Biological macromolecular structures.	wwPDB
BioModels	Computational models of biological processes.		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 © Support
EGA	Personally identifiable genetic and phenotypic data resulting from biomedical research projects.	European Bioinformatics Institute and the Centre for Genomic Regulation			
			Pending incorporation into a Node Service Delivery Plan (see How countries join):		
			BioSamples	BioSamples stores and supplies descriptions and metadata about biological samples used in research and development by academia and industry.	NCBI BioSamples database
ENA	Nucleotide sequence information, covering raw sequencing data, contextual data, sequence assembly information and functional and taxonomic annotation.	International Nucleotide Sequence Database Collaboration The International	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.	
IntAct	IntAct provides a freely available, open source database		EVA	The European Variation Archive covers genetic variation data from all species.	dbSNP and dbVAR
	Excha	Molecular Exchange Consortium	EMDB	The Electron Microscopy Data Bank is a public repository for electron microscopy density maps of macromolecular complexes and subcellular structures.	
MetaboLights	Metabolite structures and their reference spectra as well as their biological roles, locations and concentrations, and experimental data from metabolic experiments.				

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



Surprisingly few submit to international repositories



- NIH funded research
 - Only 12% of articles from NIH-funded research mention data deposited in international repositories
 - Estimated 200000+ "invisible" data sets / year

Read et al. "Sizing the Problem of Improving Discovery and Access to NIH-Funded

Data: A Preliminary Study" (2015)

PLoS ONE 10(7): e0132735. doi: 10.1371/journal.pone.0132735



What about sensitive data?



SEQUENCE

Raw unaligned reads

Aligned BAM read

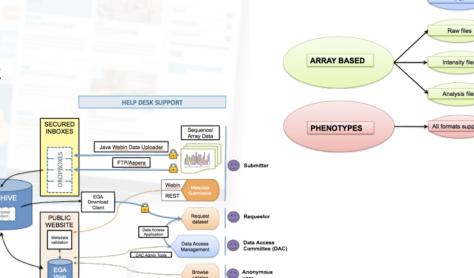
- EGA European Genome-phenome Archive
 - Repository that promotes the distribution and sharing of genetic and phenotypic data consented for specific approved uses but not fully open, public distribution.

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

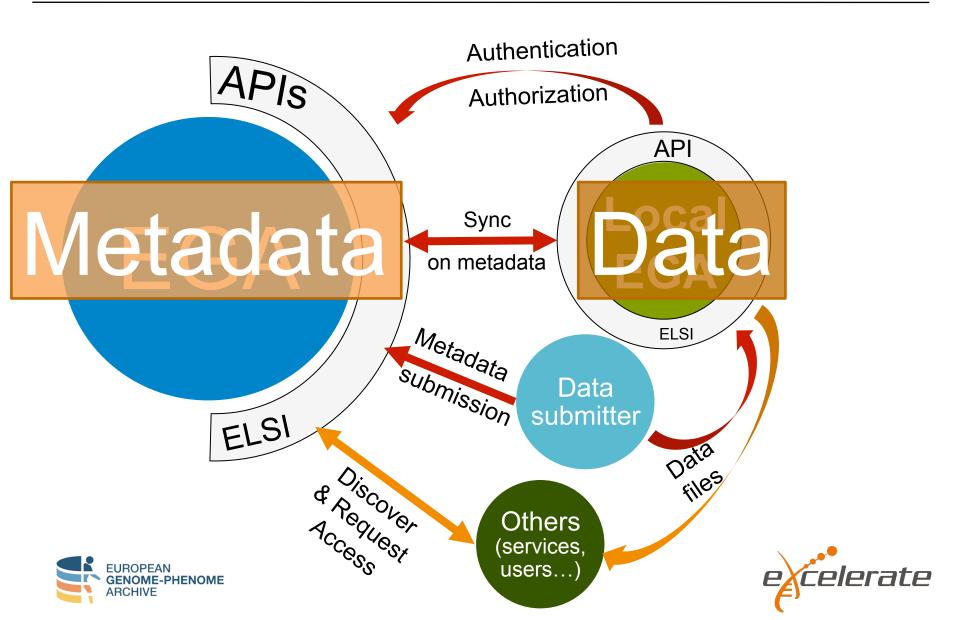
Decided by the data owner





Local EGA







"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
 - Not FAIR
- Sensitive data a potential issue
 - Figshare https://figshare.com/
 - EUDAT http://eudat.eu/
 - Data Dryad http://datadryad.org/
 - Zenodo http://www.zenodo.org/

-more generic

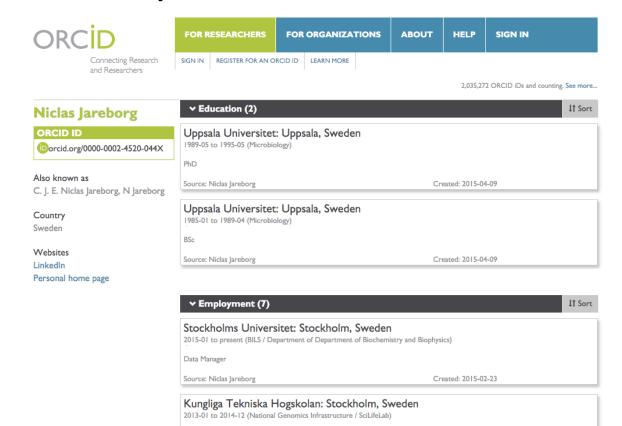
more specific >



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





NBIS Data Management support

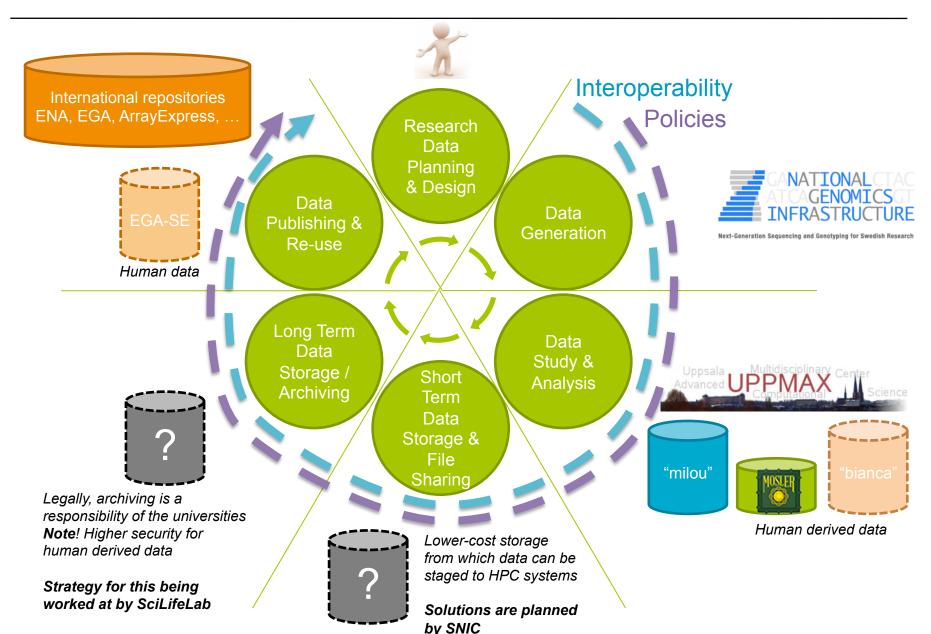


- Project planning
 - Metadata
 - File formats
 - Licensing
 - Data Management Plans
- Data analysis
- Data publication and submission
 - Automate submissions to public repositories
 - Metadata
 - Licensing



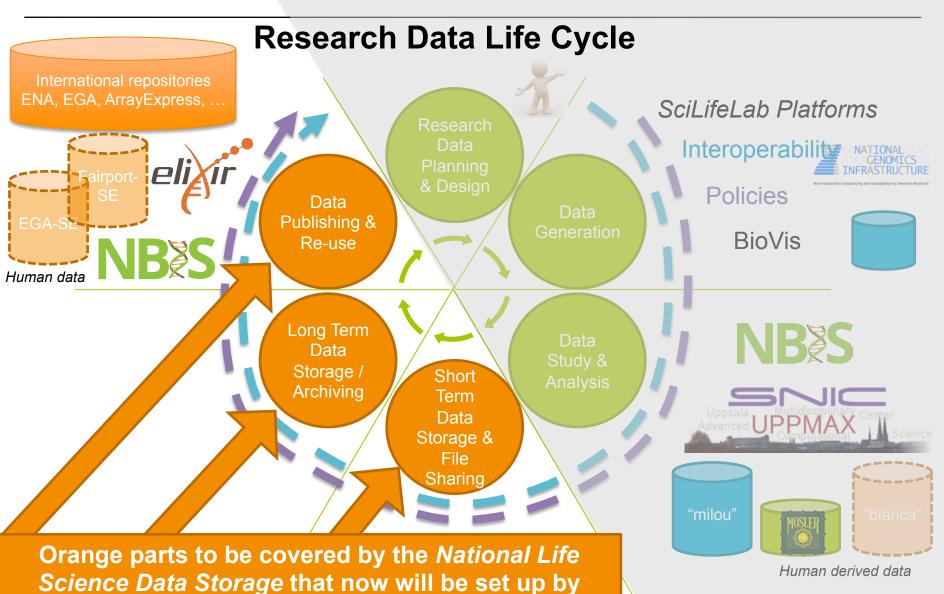
Gaps in the NGS Data Life Cycle











SUNET in collaboration with SciLifeLab

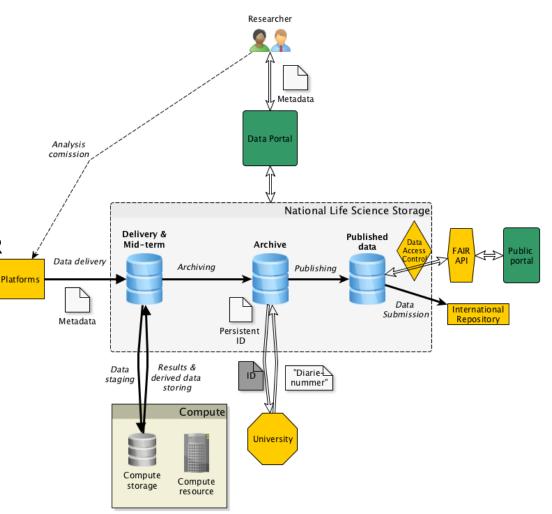


National (Life Science) Data Storage



Components

- "Active Data" storage
- Data staging to HPC resources
- Archiving
 - Offer a solution to the universities' legal obligation (possible funding stream)
- Data publication
 - Making SciLifeLab data FAIR
- User-friendly interface to manage the data life cycle process





Source Acknowledgements



- Research Data Management, EUDAT -http://hdl.handle.net/11304/79db27e2-c12a-11e5-9bb4-2b0aad496318
- Barend Mons FAIR Data
- Antti Pursula Tryggve https://wiki.neic.no/wiki/Tryggve
- Noble WS (2009)
 <u>A Quick Guide to Organizing Computational Biology Projects. PLoS Comput Biol 5(7): e1000424. doi:10.1371/journal.pcbi.1000424</u>
- Samuel Lampa http://bionics.it/posts/organizing-compbio-projects
- Reproducible Science Curriculum <u>https://github.com/Reproducible-Science-Curriculum/rr-init</u>
- Leif Väremo -<u>https://bitbucket.org/scilifelab-lts/reproducible_research_example/src</u>