

Introduction to Data-driven Life Sciences

Wei Ouyang - August 29th 2023

What are the data types in life science?



The data types in life science?

16 Responses









Paradigm Shift: Compute Power & Al





9X More HPC Performance in 4 Years nvidia.com

Throughput for Top HPC Apps



Paradigm shift: Data-driven Life Science



- 1. Hypothesis
- 2. Experiment
- 3. Accept or reject



1.

3. Evaluate pattern

Acquire data

- 1. Acquire big data
- 2. Fit AI model
- 3. Evaluate AI model









About SciLifeLab

Our vision, mission and strategic objectives.



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What is SciLifeLab?

National hub **enabling** life science research that would otherwise not be possible

Government appointed mission as a **national research infrastructure**.

Started in 2010 by Karolinska Institutet, KTH Royal Institute of Technology, Stockholm University and Uppsala University.

Today, activities at **all major Swedish universities.**



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Vision and mission

Vision: for Sweden to be a world-leading nation in life science

Mission: Enable life science research that would otherwise not be possible

Three dimensions of SciLifeLab





Research environment Approx. 190 affiliated research groups

- Environment and climate change
- Farming and forestry
- Evolution and biodiversity
- · Gene editing
- · Biofuels and biomaterials
- Microbiology and microbiome
- Drugs and biomedicine
- · Healthcare and aging



Infrastructure Service to ~ 1400 Swedish researchers annually (2020)

- Bioinformatics
- Cellular and molecular imaging
- Clinical diagnostics
- Single cell biology
- Genomics
- Chemical biology and gene editing
- Drug development
- · Proteomics and metabolomics



Data-driven life science

3.1 billion SEK, 12-year-program

Putting Sweden at the forefront of data-driven life science research and fostering the next generation of life scientists

- Four strategic research areas
- Recruiting talent from across the globe
- Academic and industry PhD and postdoc programs
- Sparking collaborations, innovation and interdisciplinary team science
- Building a strong computational and data science base for open, real-time data



SciLifeLab and Wallenberg National Program for Data-Driven Life Science

Changing the way life science is carried out

SciLifeLab and Wallenberg National Program for Data-Driven Life Science







WALLENBERG CENTRES FOR MOLECULAR MEDICINE





SLU



Stockholms

universitet

Karolinska Institutet

UN RRS LA

UMEÅ UNIVERSITET









a UN

UPPSALA UNIVERSITET





Data has a central place in life science

- Practice of life science is more and more data-dependent
- The amount of data grows exponentially
- Data becomes more complex, continuous, and needs to be openly available accessible and reusable in real-time to all



Promoting a paradigm shift in life sciences

Launching a research program to:

- Foster the next generation of life scientists
- Enable every biologist to better analyze and interpret data patterns and integrate their own data seamlessly with the global life science data streams
- Create a strong computational and data science base
- Enable deep understanding of life through data

Advancing competence and innovation in **four research areas**

Using new **technologies** and **strategies** to utilize open and real-time data

Engaging in education, training, recruiting new talent, sparking collaborations, and in innovation activities

The data life cycle



Data Centre

Created to maximize impact of SciLifeLab generated data

Assists in communication between platforms, users, and research community

Acts as a point of contact for data management questions relating to SciLifeLab generated data

Assists platforms with data tracking and statistics

Facilitates providing SciLifeLab generated data with SciLifeLab funded bioinformatics and data management support

Assists with planning the handling of SciLifeLab generated data throughout projects



FAIR principles



Findable Accessible Interoperable Reusable **Findable** – assigning a globally unique and eternally persistent identifier (like a DOI or Handle), describing the data with rich metadata, and making sure it is findable through disciplinary discovery portals.

Accessible – data and metadata should be retrievable in a variety of formats that are sensible to humans and machines using persistent identifiers.

Interoperable – the description of metadata elements should follow community guidelines that use an open, well defined vocabulary.

Reusable – the data should maintain its initial richness. The description of essential, recommended, and optional metadata elements should be machine processable and verifiable, use should be easy and data should be citable to sustain data sharing and recognize the value of data.



Discuss in groups of 3-4 on the following questions (5 minutes):

- What are the obstacles for implementing FAIR in Life Science?
- How can the FAIR principles be implemented more effectively in life sciences?

Reminder: FAIR = Findable, Accessible, Interoperable, Reusable

SciLifeLab

Getting prepared for Data-Driven Life Science Efforts in the AlCell Lab



Human Cell Simulator



Data Generation

AI Models

Execution

Microscopy Imaging Farm



AI Cloud Infrastructure





REEF: Smart Microscopy Imaging Farm







For Massive Dataset Generation



REEF: Smart Microscopy Imaging Farm

Microscope+ Fluidics Microscope+ Fluidics Fluidics	Microscope+ Fluidics	Microscope+ Fluidics	pentrons Refrigera 4°C	ator
Microscope+ Fluidics Microscope+ Fluidics Fluidics Fluidics	Microscope+ Fluidics	Microscope+ Fluidics	pentrons Automati Incubatio Chamber	ic on r
Computer Computer		Squid Microscope	Fluidics System	

Squid Microscope Prototype



CODEX multiplex imaging



In collaboration with Manu Prakash group at Stanford and Heidstar CO., LTD

Computing platform

Supercharging interactivity and scalability in AI-powered life science





ImageJ.JS: ~1000 users / day



Ouyang* et. al, Nat Methods, 2019 (*co-corresponding)

Al4Life Consortium

- Making AI models more FAIR
- Supported by EU Horizon grant
- Since 2019, with EMBL, HT, Euro Bioimaging...



https://ai4life.eurobioimaging.eu/



Biolmage.IO

- Model description file standard
- Repository for sharing models
- Continuous integration for model testing
- Cloud-based model serving and test run





Preprint for the BioImage Model Zoo https://doi.org/10.1101/2022.06.07.495102

https://bioimage.io



BioEngine: Scalable Al Model Serving

🛞 kubernetes





About the course

More information at <u>https://ddls.aicell.io</u>

What do you do with ChatGPT?



Only the Menti presentation will be shown for this slide

What do you do with ChatGPT?

17 Responses

code faster

writing reportsknow things quicklylearn languagewriting my thesisexplaining difficult stufrewrite emailswrite thesishistoryto studywrite emailsmathcodingsummarize paperssummarize materialslearn abt hard topicsget instruction on method

Tips for using ChatGPT

- Set a role
- Be specific
- Provide examples
- Use formatting
- Avoid assumptions
- Check limitations
- Try different approaches
- Watch for repetition
- Limit prompt length
- Correct gently
- Give feedback



Assignments

See here: https://ddls.aicell.io/course/ddls-2023/module-1/#lecture-tuesday