

## Al in research software: Best practices

Research Data Unit:Dr. Georg Schwesinger and Dr. Sebastian ZangerleScientific Al group:Peter LippmannScientific Software Center: Dr. Inga Ulusoy

February 2025



## 1. Requirements of "ML-based science"



## What this course is not

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- An introduction to data science
- An introduction to machine learning
- A course about different ML algorithms
- A course about different ML training approaches and libraries



## What this course is

- A best practices guide to creating machine learning based research software (MLBRS)
- A recommendation on how to manage and prepare your data
- A recommendation on how to train your models
- An introduction to software engineering best practices for MLBRS
- A guideline on how to generate independently reproducible scientific results using data-based approaches
- A guideline on how to publish your data and your models



# What is special about research software based on data? ("ML-based science")

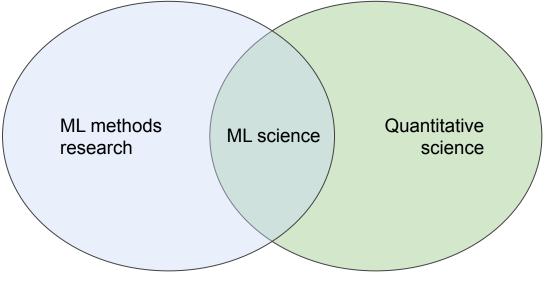


## ML science

- Scientific research that uses machine learning models to extend scientific knowledge
- Answers a scientific question by using ML
- No restriction on algorithm, method, library, domain

Contrary to:

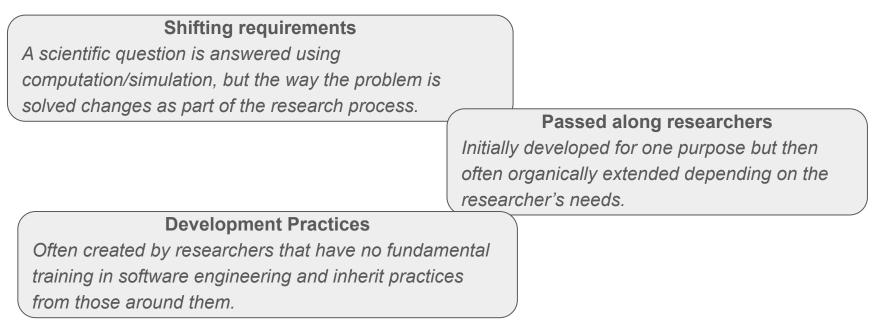
 ML methods research: Research on ML methods and algorithms with the goal to improve the field of ML





## Research software

"... software that is developed and used in the context of research..."





## ML-based research software

"... software that is developed and used in the context of research and predicts but la cut comes abased on initial data..." Massive data and complex underlying mathematical structure Data size data quality!  $\diamond$ Data *and* software quality provide limits to quality and software impact of research! quality! Mathematically complex problem, but simple data Trivial case: simple problem, Complexity of the problem few data points





## MLBRS: Data

#### Data is foundation for..

...model training, decision making and/or predictions.

### Different kinds of data

For example, numerical data, textual data, images, audio, video.

### Metadata

What is relevant metadata and should be included on the data card?

### Legal considerations

Where does the data come from? Is it licensed? Is it public or private data? In what form is the data stored and processed?

### **Ethical considerations**

Does the data exploit work of others? Does it break some sort of confidentiality? Will it impact in a possible harmful way or can it be misconstrued to do harm?

#### Bias

*Is there an inherent bias in the data itself, due to the data collection approach, or other reasons?* 

**Availability and licensing** Will the data be publicly available to the community? What license does/will the dataset have?



## **MLBRS:** Software

Purpose

Will the software be more widely used, be an in-house code, or one-person software?

**Software engineering best practices** Does the software follow software engineering best practices (version control, testing, documentation, ...)?

#### Usability and reproducibility

Does the software include documentation on how models can be trained, and keeps track of training parameters? Does the software help to generate model cards and provide models in transferable format?

#### Accuracy and reliability

Does the software create robust and consistent results, even though it is based on a non-deterministic process? Legal considerations Does the software incorporate third-party models and/or code?

### Legal considerations

What license is the software published under? What license are models published under?

#### Security

Is the software secure against data injection?



## **Reproducibility**

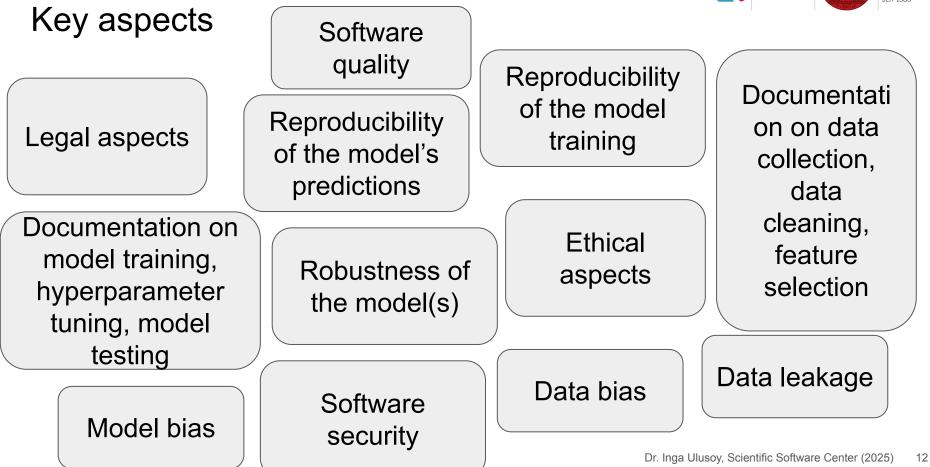
- Provide data to enable others to reproduce findings
- Provide code to enable others to reproduce findings
- Computational reproducibility (i)
  - Make sure your findings are true findings, and do not arise from problems with your data/code
- → Independent reproducibility (ii)

Research software engineering generally targets (i), but with MLBRS we target (ii)

### Why should you care?

Your research integrity, scientific best conduct (malpractice), can have long-lasting detrimental effect on science (impact on others and the field), affects society!







## 2. Research Data Management



## Al in research software Part X: Research Data Management







Dr. Sebastian Zangerle Heidelberg University Library sebastian.zangerle@ub.uniheidelberg.de

#### Dr. Georg Schwesinger Heidelberg University Library schwesinger@ub.uni-heidelberg.de





## **RESEARCH DATA UNIT**



## **Research Data Unit at Heidelberg University**



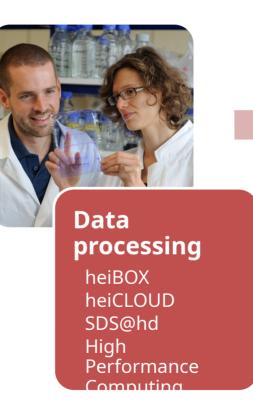
### http://data.uni-heidelberg.de/



### Project Planning

Data Management Plans Courses & workshops







### Data Archiving & Publication

heiDATA heidICON Archive – your data preserved heiARCHIVE

Pictures: © Universität Heidelberg, Kommunikation und Marketing





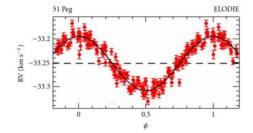
### Newsletter Research Data Unit (RDU)

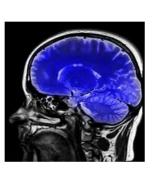
Update information regarding RDM at Heidelberg University:

- new services
- current workshops
- Training courses
- Please subscribe to the mailing list: DATA-NEWS@LISTSERV.UNI-HEIDELBERG.DE
- https://listserv.uni-heidelberg.de/cgi-bin/wa?A0=DATA-NEWS









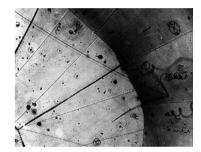


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## WHAT IS RDM ABOUT?



## Data Driven Research



Neelie Kroes The Topics ~ Print edition More Economist nt of the European Commission responsible for the la Regulating the internet giants Simon C. Lin The world's most value the new gold **Eric Yen** Editors resource is no longer Data Driven The data economy demands a new appr check Against Delivery e-Science Seulle texte prononcé fait fe ES BIIL das gesprochane Wort Use Cases and Successful Applications of Distributed Computing Infrastructures (ISGC 2010) 2 Springer opening reamarks, Press Conference on Open Data Strategy Brussels, 12th December 2011



### **Data Driven Research?**



"The underlying data researchers analyze to come to their published conclusions ... becomes less and less accessible to researchers over the years." (Vines et al, 2014; Dehnhard, Weichselgartner &

Krampen, 2013; Wicherts et al, 2006)

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(D) Predicted probability that the data were extant (either "shared" or "exist but unwilling to share") given that we received a useful response.

Slide: Dehnhard 2014



### **Data Driven Research?**



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published conclusions ... becomes less and less accessible to

researchers over the years." (Vines et al, 2014; Dehnhard, Weichselgartner & Why is that disastrous? Krampen, 2013; Wicherts et al, 2006)

" […] data is the currency of science, even if publications are still the currency of tenure. To be able to exchange data, communicate it, mine it, reuse it, and review it is essential to scientific productivity, collaboration, and to discovery itself." (Cold 2007)

15

age of paper (years)

20

Transparency and verifiability

- Reproducible vs. non-reproducible data
- Re-use

(either "shared" or "exist but unwilling to share") given that we received a useful response.



Slide: Dehnhard 2014

## What is research data management?

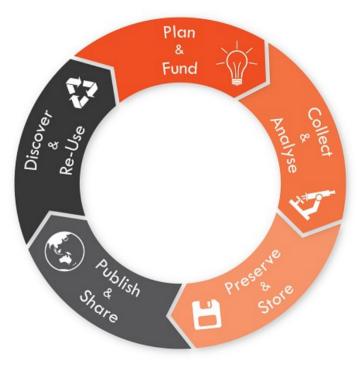


#### Research data management

"Research data management concerns the organisation of data, from its entry to the research cycle through to the dissemination and archiving of valuable results. It aims to ensure reliable verification of results, and permits new and innovative research built on existing information."

(Whyte, A., Tedds, J. (2011).

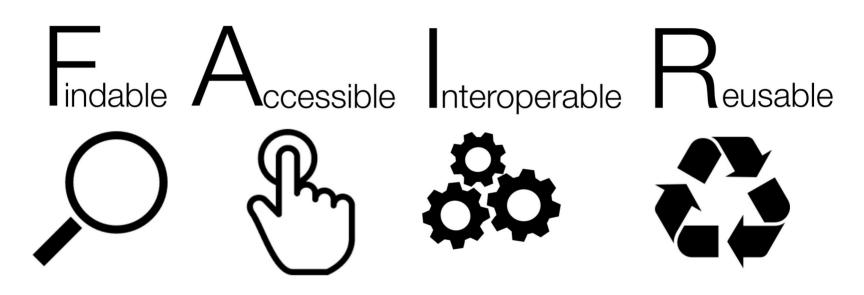
<u>'Making the Case for Research Data Management</u>'. DCC Briefing Papers. Edinburgh: Digital Curation Centre.)



https://library.sydney.edu.au/research/data-man agement/research-data-management.html



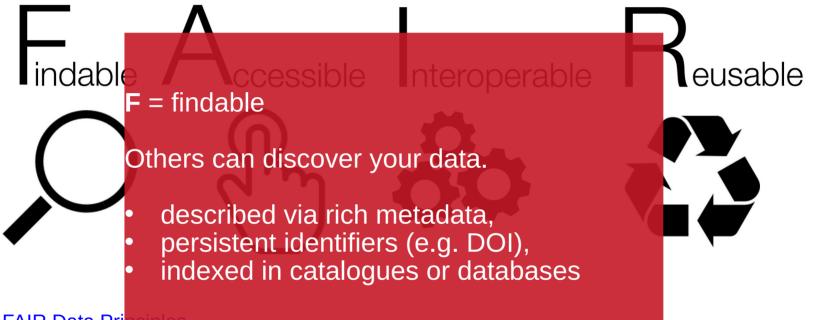




- FAIR Data Principles
- Wilkinson et al. (2016), The FAIR Guiding Principles for scientific data management and stewardship, Scientific Data 3, <u>doi:10.1038/sdata.2016.18</u>
- SNF: Explanation of the FAIR Data Principles



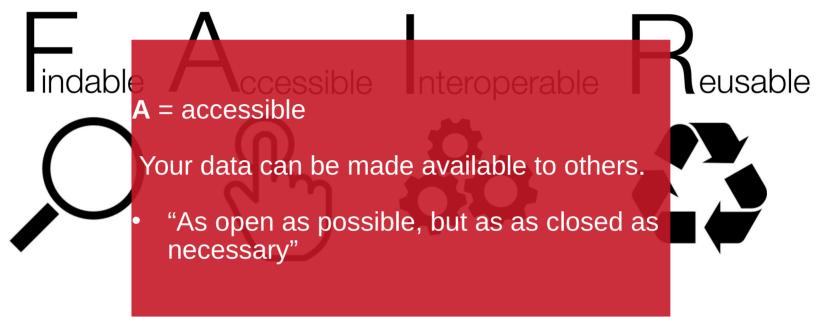




- <u>FAIR Data Principles</u>
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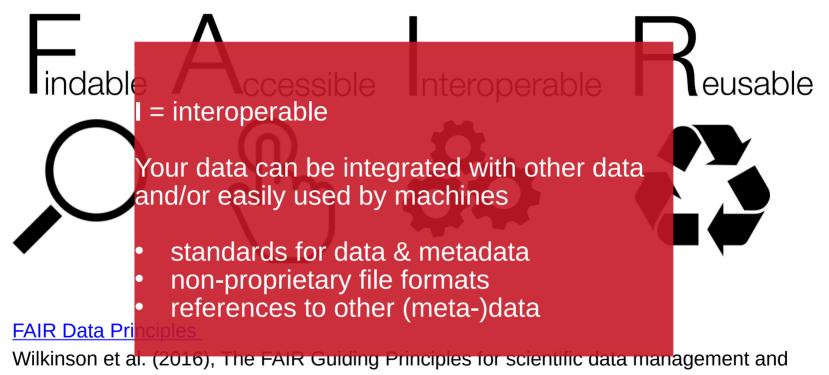




- FAIR Data Principles
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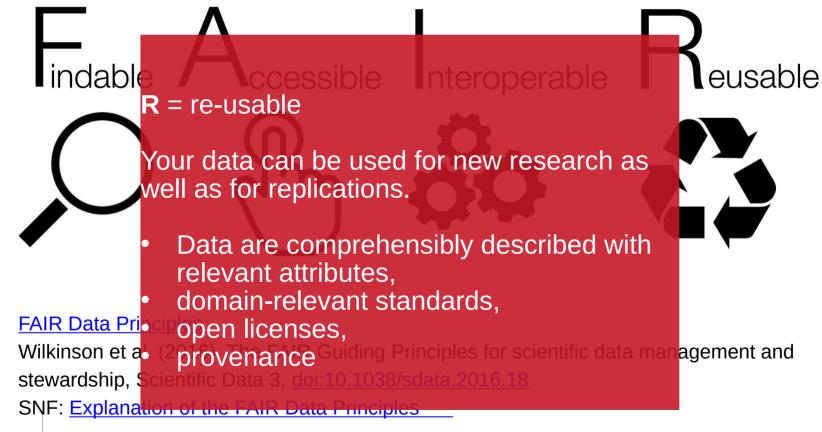


stewardship, Scientific Data 3, doi:10.1038/sdata.2016.18

SNF: Explanation of the FAIR Data Principles

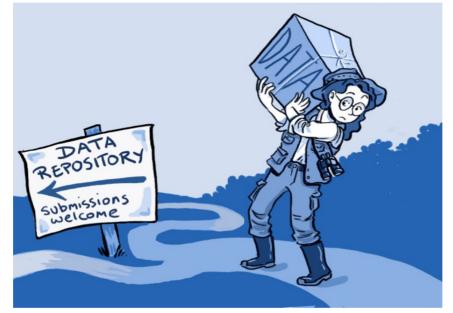












Picture: Ainsley Seago. doi:10.1371/journal.pbio.1001779.g001

## **OPEN RESEARCH DATA**



## NFDI





- National Research Data Infrastructure
- The National Research Data Infrastructure (NFDI) has the objective to systematically index, edit, interconnect and make available the valuable stock of data from science and research.
- Funding for subject- and/or methods-specific consortia
- Overall budget: 85 Mio € per year for 10 years
- 27 subject-specific consortia

## <u>https://www.nfdi.de/</u>



## NFDI



#### GHGA



The German Human Genome-**Phenome Archive** 

IMPACT

ABOUT US

We are building a secure national omics data infrastructure, enabling the use of human genome data for research purposes while preventing data misuse

GHGA Metadata Catalog A public frontend for the discovery of human omic study data from Ger research institutions

Our Mission

Consent Tools **GHGA Lecture Series** GHGA has developed "Advances in Data-Drive medicine" diving into different tools to held clinicians, researchers and omics data to GHGA ethical legal and social implications. Learn more

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fascinating world of datadriven medicine and their

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#### NFDI4Culture – Consortium for Research Data on Material and Immaterial Cultural Heritage

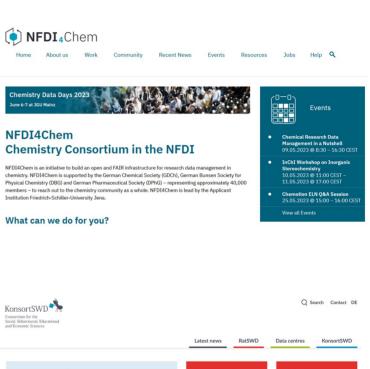
We establish a needs-based infrastructure for research data ranging from architecture, art history and musicology to theatre, dance, film and media studies.

Enter keywords.

Search

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#### **KonsortSWD** As part of the NFDI, KonsortSWD is expanding its services for research with data in the social, educational, behavioural and economic sciences.

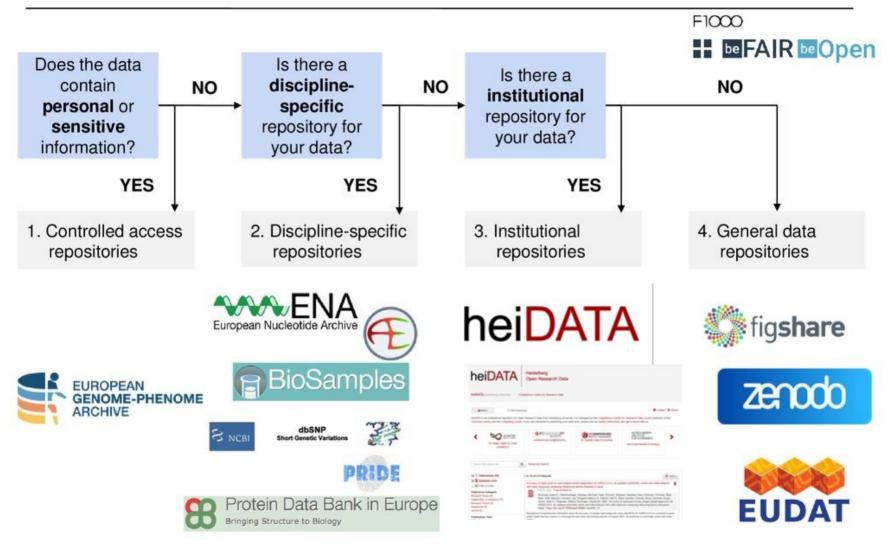
Services → Tasks → Engagement in the NFDI  $\rightarrow$ 





Repositories





Etc...

Slide adapted from: N. Jareborg (2019), "Data management and repositories", <u>https://player.slideplayer.com/105/17629367/</u>.

## Finding repositories







### https://www.re3data.org/

### https://fairsharing.org/

### Publisher Guidelines

- <u>https://www.nature.com/sdata/policies/repositories</u>
- <u>https://journals.plos.org/plosone/s/recommended-repositories</u>
- <u>https://www.springernature.com/gp/authors/research-data-policy/re</u>
   <u>commended-repositories</u>

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### **Data publication**



e3data.org		Search Browse - Suggest Resources - Contact 👤 DataCit
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Keywords ⊞	<b>.</b> . ,	lection of databases, domain theories, and data generators that are used by the machine learning community for the empirical analysis of
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Repository languages 🕀	Open Machine Learning	
Software 🕀	Subject(s)	Education Sciences Computer Science Social and Behavioural Sciences Humanities and Social Sciences
Repository types ⊞		Computer Science, Electrical and System Engineering Engineering Sciences
Versioning ⊞	Content type(s)	Standard office documents Structured graphics Plain text Software applications Source code Configuration data other Databases
	Country	Belgium European Union Netherlands

OpenML is an open ecosystem for machine learning. By organizing all resources and results online, research becomes more efficient, useful and fun. OpenML is a platform to share detailed experimental results with the community at large and organize them for future reuse. Moreover, it will be directly integrated in today's most popular data mining tools (for now: R, KNIME, RapidMiner and WEKA). Such an easy and free exchange of experiments has tremendous potential to speed up machine learning research, to engender larger, more detailed studies and to offer accurate advice to practitioners. Finally, it will also be a valuable resource for education in machine learning and data mining.



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#### Repository details

### **OpenML**

General	Institutions	Terms	Standards	
Name of repos	sitory	OpenN	IL.	
Additional nam	ne(s)	Open N	lachine Learning	
Repository UR	۲L	http://w	ww.openml.org/	
Subject(s)		Educatio	on Sciences Computer Science Social and Behavioural Sciences Humanities and Social Sciences	
		Comput	ter Science, Electrical and System Engineering Engineering Sciences	
Description		researc experim be direc WEKA). machine	L is an open ecosystem for machine learning. By organizing all resources and results online, h becomes more efficient, useful and fun. OpenML is a platform to share detailed ental results with the community at large and organize them for future reuse. Moreover, it will stly integrated in today's most popular data mining tools (for now: R, KNIME, RapidMiner and Such an easy and free exchange of experiments has tremendous potential to speed up e learning research, to engender larger, more detailed studies and to offer accurate advice to oners. Finally, it will also be a valuable resource for education in machine learning and data	
Contact		openma	achinelearning@gmail.com	
Content type(s)	Standar	d office documents Structured graphics Plain text Software applications Source code		
		Configu	ration data other Databases	
Keyword(s)		machine	e learning [ meta-learning ] experimental methodology ] datasets ] algorithms ] experiments ]	
Repository size	e	1700000 machine learning experiments on 19630 datasets and 3370 implementations		
Repository typ	e(s)	disciplinary		
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Repository details

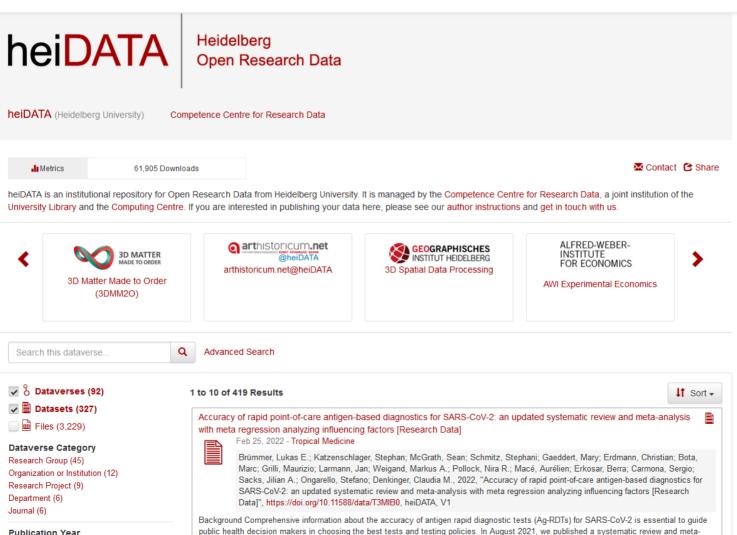
### heiDATA



General Institutions	Terms Standards	
Name of repository	heiDATA	
Additional name(s)	heiDATA Institutional Repository for Research Data of Heidelberg University	
Repository URL	https://heidata.uni-heidelberg.de	
Subject(s)	Humanities       Social and Behavioural Sciences       Economics       Jurisprudence       Biology       Medicine         Microbiology, Virology and Immunology       Agriculture, Forestry, Horticulture and Veterinary Medicine         Chemistry       Physics       Geosciences (including Geography)         Computer Science, Electrical and System Engineering       Humanities and Social Sciences       Life Sciences         Agriculture, Forestry, Horticulture and Veterinary Medicine       Natural Sciences       Engineering Sciences	
Description	heiDATA is Heidelberg University's research data repository. It is managed by the Competence Centre for Research Data, a joint institution of the University Library and the Computing Centre. All researchers affiliated with Heidelberg University can use this service for archiving and publishing their data.	
Contact	http://www.data.uni-heidelberg.de/contact.html	
Content type(s)	Standard office documents       Databases       Raw data       Structured text       Source code       other       Images         Structured graphics       Audiovisual data       Scientific and statistical data formats       Plain text       Structured text         Archived data       Archived data       Scientific and statistical data formats       Plain text       Structured text	
Keyword(s)	data processing       computer science       linguistics       economics       geograhy       history       mathematics         social science       chemistry       earth sciences       modern languages	

## heiDATA





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Findability - DOI's





#### Journal of Geophysical Research: Oceans

#### **RESEARCH ARTICLE**

10.1002/2017JC013678

#### **Key Points:**

- First bomb-<sup>14</sup>C peak reconstruction in the high-latitude NW Pacific made with a high-resolution analysis of long-lived bivalve shells
- Relatively high bomb-<sup>14</sup>C peak, though at high latitude (40°N), is due to water transport by Kuroshio Current
- Bomb-<sup>14</sup>C record provides a reliable tracer of water mixing



#### Bomb-<sup>14</sup>C Peak in the North Pacific Recorded in Long-Lived Bivalve Shells (*Mercenaria stimpsoni*)

## Kaoru Kubota<sup>1,2,3</sup> <sup>(D)</sup>, Kotaro Shirai<sup>2</sup>, Naoko Murakami-Sugihara<sup>2</sup>, Koji Seike<sup>2,4</sup> <sup>(D)</sup>, Masayo Minami<sup>5</sup>, Toshio Nakamura<sup>3</sup>, and Kazushige Tanabe<sup>5</sup>

<sup>1</sup>Kochi Institute for Core Sample Research, Japan Agency for Marine-Earth Science and Technology, Nankoku, Japan, <sup>2</sup>Atmosphere and Ocean Research Institute, University of Tokyo, Chiba, Japan, <sup>3</sup>Institute for Space-Earth Environmental Research, Nagoya University, Furo-cho, Nagoya, Japan, <sup>4</sup>Now at Geological Survey of Japan, National Institute of Advanced Industrial Science and Technology, Tsukuba, Japan, <sup>5</sup>Department of Earth and Planetary Science, University of Tokyo, Bunkyo, Japan



Findability - DOI's





#### Journal of Geophysical Research: Oceans

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#### **RESEARCH ARTICLE** Bomb-<sup>14</sup>C Peak in the North Pacific Recorded in Long-Lived

Proceedings of the National Academy of Sciences of the United States of America, 112, 9542–9545.

Guilderson, T. P., Schrag, D. P., Kashgarian, M., & Southon, J. (1998). Radiocarbon variability in the western equatorial Pacific inferred from a high-resolution coral record from Nauru Island. *Journal of Geophysical Research*. 103. 24641–24650.

Hammer, S., & Levin, I. (2017). Monthly mean atmospheric Δ<sup>14</sup>CO<sub>2</sub> at Jungfraujoch and Schauinsland from 1986 to 2016 (heiDATA Dataverse V2). Heidelberg: Heidelberg Univerity. https://doi.org/10.11588/data/10100

Hanawa, K. (1983). Sea surface temperature off Sanriku coast and east of Tsugaru Strait monitored by ferry Ishikari (I). Tohoku Geophysical Journal, 29, 129–149.

Hanawa, K., & Mitsudera, H. (1986). Variation of water system distribution in the Sanriku Coastal Area. Journal of the Oceanographic Society of Japan, 42, 435–446.

Hua, Q., Barbetti, M., & Rakowski, A. Z. (2013). Atmospheric radiocarbon for the period 1950-2010. Radiocarbon, 55, 2059-2072.

Ishizu, M., Itoh, S., & Tanaka, K. (2016). Influence of the Oyashio Current and Tsugaru Warm Current on the circulation and water properties





Title 😧	GECCA mapped
Subtitle 😌	Mapping Western Group Exhibitions of Contemporary Chinese Art after 1979
Author 😌	Franziska Koch (Heidelberg Centre for Transcultural Studies, Global Art History, Heidelberg University, Germany)
Contact 😯	Use email button above to contact.
	Franziska Koch (Heidelberg Centre for Transcultural Studies, Global Art History, Heidelberg University, Germany)
Description	GECCA mapped is a pilot project that visualizes and provides geo-referential metadata of sixty exhibition entries collected in the larger GECCA data base (more than 700 entries). The exhibition sample is limited to Western, i.e. Western European and Northerr American group exhibitions, and excludes bi-/ triennials. With the support of the HRA (Heidelberg Research Architecture), GECCA mapped allows the user to trace the exhibition sample implemented in Google Earth. The GECCA mapped logo indicates the place where a particular exhibition was staged and is scaled according to the number of participating artists. A click on the logo opens a pop-up window presenting more information on the exhibition. The Google Earth timeline enables the user to follow the exhibition development in any chosen geographical area in the period from 1982 (earliest exhibition entry) to 2009 (latest exhibition entry).
	Group Exhibitions of Contemporary Chinese Art (GECCA): The medium of (group and panoramic) exhibitions has played a fundamental role in creating a global context for Chinese art within and outside of the Poeple's Republic after the end of the "Great Proletarian Cultural Revolution" (1966-1976) and since the political reforms initiated by Deng Xiaoping in 1978/79. In economic, discursive, aesthetic and institutional terms, the Western reception of these shows was very influential for the establishment of a certain international canon of artworks, artists and curators. This particular canon in fact came to be considered representative of the whole of Chinese artistic production, although It actually tends to exclude large parts of the overall artistic activity such as "national ink painting" (guohua), conventional or conservative academic oil painting, as well as those works involving political or consumption oriented subject matter, including mass-produced decorative and popular artworks.
	With 60 exhibitions entries, the data that GECCA mapped visualizes is a comparatively small sample of the database GECCA - which contains more than 700 exhibition entries. The data was individually researched and includes information on the location, institution, dates, exhibition topic, participating artists and curators. The sources for the data stem from exhibition catalogues, museum websites, archival documentation of public art libraries and other archives.
	A typical use of the kmz-file that visualizes GECCA mapped is Google Earth.
Subject 😯	Arts and Humanities
Keyword	contemporary Chinese art group exhibitions North America (general region) (TGN) http://vocab.getty.edu/tgn/7029440 Europe (continent) (TGN) http://vocab.getty.edu/tgn/7000003 Australia (nation) (TGN) http://vocab.getty.edu/tgn/7000490 Art, Chinese20th centuryExhibitions (LCSH) http://id.loc.gov/authorities/subjects/sh2007101410 GECCA mapped Geographic information systems (LCSH) http://id.loc.gov/authorities/subjects/sh90001880 Dicital mancing.0 (ESH) http://id.loc.gov/authorities/Subjects/sh90001880
Deleted Dublication	Digital mapping (LCSH) http://id.loc.gov/authorities/subjects/sh85037980
Related Publication 😣	Koch, Franziska. 2016. "Die »chinesische Avantgarde« und das Dispositiv der Ausstellung: Konstruktionen chinesischer Gegenwartskunst im Spannungsfeid der Globaliserungr. Bielefeldt (transcript. isbn: 978-3-8376-2617-9 http://www.transcript- verlag.de/978-3-8376-2617-9/tdie-chinesische-avantgarde-und-das-dispositiv-der-ausstellung
	Chinoco: English

#### Findability -

DOI's Metadata -





Title 😣	GECCA mapped
Subtitle 😯	Mapping Western Group Exhibitions of Contemporary Chinese Art after 1979
Author 😧	Franziska Koch (Heidelberg Centre for Transcultural Studies, Global Art History, Heidelberg University, Germany)
Contact 😌	Use email button above to contact.
	Franziska Koch (Heidelberg Centre for Transcultural Studies, Global Art History, Heidelberg University, Germany)
Description	GECCA mapped is a pilot project that visualizes and provides geo-referential metadata of sixty exhibition entries collected in the larger GECCA data base (more than 700 entries). The exhibition sample is limited to Western, i.e. Western European and Northern American group exhibitions, and excludes bi-/ triennials. With the support of the HRA (Heidelberg Research Architecture), GECCA mapped allows the user to trace the exhibition sample implemented in Google Earth. The GECCA mapped logo indicates the place where a particular exhibition was staged and is scaled according to the number of participating artists. A click on the logo opens a pop-up window presenting more information on the exhibition. The Google Earth timeline enables the user to follow the exhibition development in any chosen geographical area in the period from 1982 (earliest exhibition entry) to 2009 (latest exhibition entry).

Group Exhibitions of Contemporary Chinese Art (GECCA): The medium of (group and panoramic) exhibitions has played a fundamental role in creating a global context for Chinese art within and outside of the People's Republic after the end of the "Great Proletarian Cultural Revolution" (1966-1976) and since the political reforms initiated by Deng Xiaoping in 1978/79. In economic, discursive aesthetic and institutional terms, the Western reception of these shows was very influential for the establishment of a

certain international canon of artworks, artists and curators. This partithe whole of Chinese artistic production, although it actually tends to  $\epsilon$ "national ink painting" (guohua), conventional or conservative academ consumption oriented subject matter, including mass-produced decora

With 60 exhibitions entries, the data that GECCA mapped visualizes is which contains more than 700 exhibition entries. The data was individu institution, dates, exhibition topic, participating artists and curators. Th museum websites, archival documentation of public art libraries and of

A typical use of the kmz-file that visualizes GECCA mapped is Google

Subject 🕄 Arts and Humanities Keyword 😡 contemporary Chinese art group exhibitions North America (general region) (TGN) http://vocab.getty.edu/tgn/7029 Europe (continent) (TGN) http://vocab.getty.edu/tgn/1000003 Australia (nation) (TGN) http://vocab.getty.edu/tgn/7000490 Art. Chinese--20th century--Exhibitions (LCSH) http://id.loc.gov/author GECCA mapped

Geographic information systems (LCSH) http://id.loc.gov/authorities/subjects/sh90001880 Digital mapping (LCSH) http://id.loc.gov/authorities/subjects/sh85037980

Related Publication 🚱

Koch, Franziska. 2016. "Die »chinesische Avantgarde« und das Dispositiv der Ausstellung: Konstruktionen chinesischer Gegenwartskunst im Spannungsfeld der Globalisierung". Bielefeld: transcript. isbn: 978-3-8376-2617-9 http://www.transcriptverlag.de/978-3-8376-2617-9/die-chinesische-avantgarde-und-das-dispositiv-der-ausstellung



#### RESEARCH DATA UNIT

Chinoso: English

Design Type 🙆 Factor Organis Other O

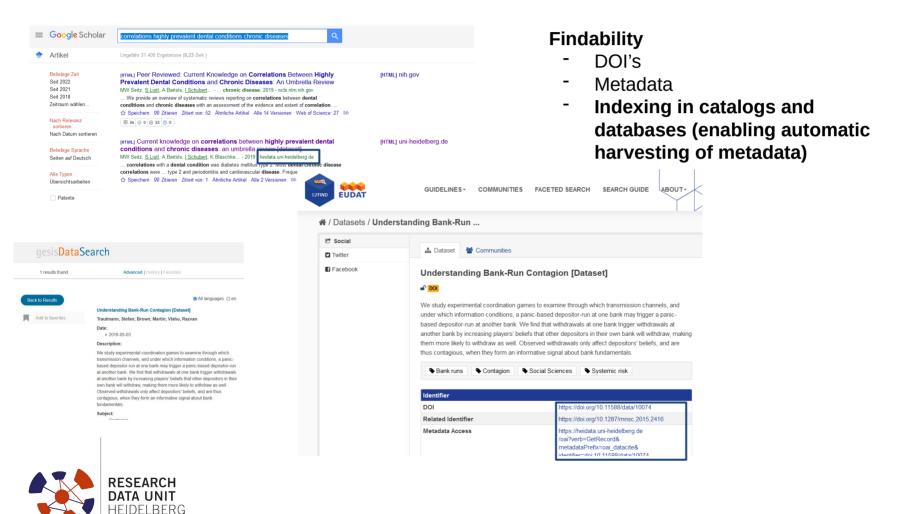
Life Sciences Metadata 🔺

Design Type 🚱	Not Specified
Factor Type 😧	Cell Type/Cell Line; Developmental Stage; Organism
Organism 😧	Homo sapiens; Mus musculus
Other Organism 😯	Monodelphis domestica
Measurement Type 😯	transcription profiling
Technology Type 😧	nucleotide sequencing
Other Technology Type 😣	single nucleus RNA-seq
Technology Platform 😣	Illumina
Other Technology Platform 😯	10x Chromium 3' protocol

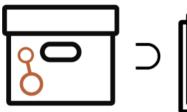
#### Findability

- DOI's -
- Metadata -





#### **Organization of a Dataverse Repository**



#### Dataverse

Collection of datasets Own administration Own branding (& can be embedded in your site)



#### dataset

Citation Metadata Versioning Terms/permissions Collection of Files



File

Citation Preview/Explore Metadata Versioning Permissions



- DOI's
- Metadata
- Indexing in catalogs and databases (enabling automatic harvesting of metadata)
- Dataverses: collection of datasets e.g. For research groups, projects,...





RESEARCH DATA UNIT HEIDELBERG

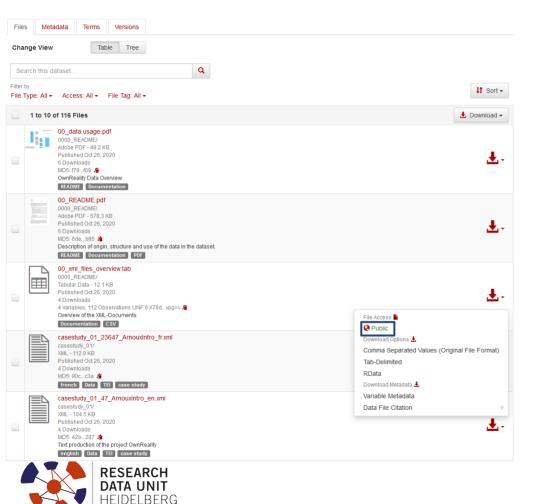


3D MATTER MADE TO ONDER	3D Matter Made to Order (3DMM2O) (Heidelberg University and Karlsruhe Institute of Technology	ogy (KIT))	
heiDATA >			
	🐱 Contact	C Share	🖌 Edit 🗸
Technology (KIT 3DMM2O establi	s from the Cluster of Excellence 3D Matter Made to Order (3DMM2O). The Cluster is a collaboration of Kar and Heidelberg University. It pursues an interdisciplinary approach through conjunction of natural, engine shes scalable digital 3D Additive Manufacturing transcending from the molecular to the macroscopic scale. D manufacturing and material processing.	ering, and so	ocial sciences

Search this dataverse		Advanced Search	🕇 Add Data 🗸		
✓ <sup>o</sup> → Dataverses (0)	1 to 10 of	15 Results	↓† Sort -		
<ul> <li>Datasets (15)</li> <li>Files (147)</li> </ul>	Printing	8			
Publication Year 2022 (8)		Goepfrich, Kerstin; Walther, Tobias, 2022, "Printing and Erasing of DNA-based Photore: Synthetic Cells [RESEARCH DATA]", https://doi.org/10.11588/data/MKOC9S, heiDATA			
2021 (7)	In the pursuit to produce functioning synthetic cells from the bottom up, DNA nanotechnolog powerful tool. However, the crowded yet highly organized arrangement in living cells, bridging				
Publication Status Published (14)	micron-scale, remains challenging to recreate with				
Draft (1)		netrization Strategy to Achieve Triptycene-Based 3,6-Dimethoxytriphenylenes via C hydrogenation [Data]	oxidative 🗎 🔗		
Author Name		Feb 4, 2022 - Institute of Organic Chemistry - AK Mastalerz			
Rominger, Frank (7) Mastalerz, Michael (5) Bunz, Uwe H. F. (2) Elbert, Sven Michael (2)	(5) (5) (2) Reinhard, Dennis; Rominger, Franl Triptycene-Based 3,6-Dimethoxytr /10.11588/data/OH6757, heiDATA,	Reinhard, Dennis; Rominger, Frank; Mastalerz, Michael, 2022, "Desymmetrization Stra Triptycene-Based 3,6-Dimethoxytriphenylenes via Oxidative Cyclodehydrogenation [Dat /10.11588/data/OH6757, heiDATA, V1			
Freudenberg, Jan (2)	triptycen	ve a highly symmetric triptycene based hexamethoxytriphenylene in high yield of 97 %, a i e (C3V) is necessary as precursor for cyclodehydrogenative Scholl-type oxidation, by takir widdy for a constraint or particle and the state of the			
More	regiosele	ctivity of the C-C bond formation controll			

#### Findability

- DOI's -
- Metadata -
- Indexing in catalogs and databases (enabling automatic harvesting of metadata **Dataverses: collection of** datasets e.g. For research groups, projects,...





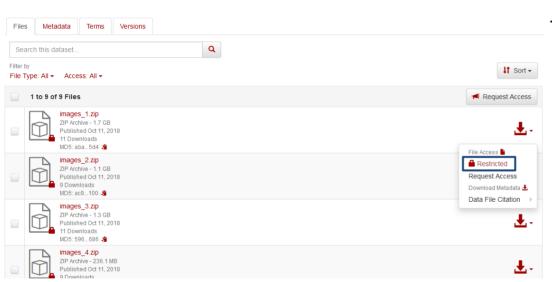
#### Accessibility

-

Download of public files via browser or via API

RESEARCH DATA UNIT HEIDELBERG





#### Accessibility

- Download of public files via browser or via API
- "As open as possible, but as closed as necessary"

**Unpublished Dataset Private URL** – Privately share this dataset before it is published: https://heidata.uni-heidelberg.de/privateurl.xhtml?token=ffb013cb-25ae-46f0-904d-381190a8ca13

#### **Test Dataset**

Draft Unpublished

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Apel, Jochen, 2022, "Test Dataset", https://doi.org/10.11588/data/CKSXU7, heiDATA, DRAFT VERSION

Cite Dataset - Learn about Data Citation Standards.



#### Accessibility

- Download of public files via browser or via API
- "As open as possible, but as closed as necessary"
- Private URLs for prepublication access (e.g. for reviewers)





#### Interoperability - Metadata standards

#### Metadata References

The Dataverse Project is committed to using standard-compliant metadata to ensure that a Dataverse installation's metadata can be mapped easily to standard metadata schemas and be exported into JSON format (XML for tabular file metadata) for preservation and interoperability.

Detailed below are what metadata schemas we support for Citation and Domain Specific Metadata in the Dataverse Project:

- Citation Metadata: compliant with DDI Lite, DDI 2.5 Codebook, DataCite 3.1, and Dublin Core's DCMI Metadata Terms (see .tsv version). Language field uses ISO 639-1 controlled vocabulary.
- Geospatial Metadata: compliant with DDI Lite, DDI 2.5 Codebook, DataCite, and Dublin Core (see .tsv version). Country / Nation field uses ISO 3166-1 controlled vocabulary.
- Social Science & Humanities Metadata: compliant with DDI Lite, DDI 2.5 Codebook, and Dublin Core (see .tsv version).
- Astronomy and Astrophysics Metadata : These metadata elements can be mapped/exported to the International Virtual Observatory Alliance's (IVOA) VOResource Schema format and is based on Virtual Observatory (VO) Discovery and Provenance Metadata (see .tsv version).
- Life Sciences Metadata: based on ISA-Tab Specification, along with controlled vocabulary from subsets of the OBI Ontology and the NCBI Taxonomy for Organisms (see .tsv version).
- Journal Metadata: based on the Journal Archiving and Interchange Tag Set, version 1.2 (see .tsv version).

See also the Dataverse Software 4.0 Metadata Crosswalk: DDI, DataCite, DC, DCTerms, VO, ISA-Tab document and the Metadata Customization section of the Admin Guide.



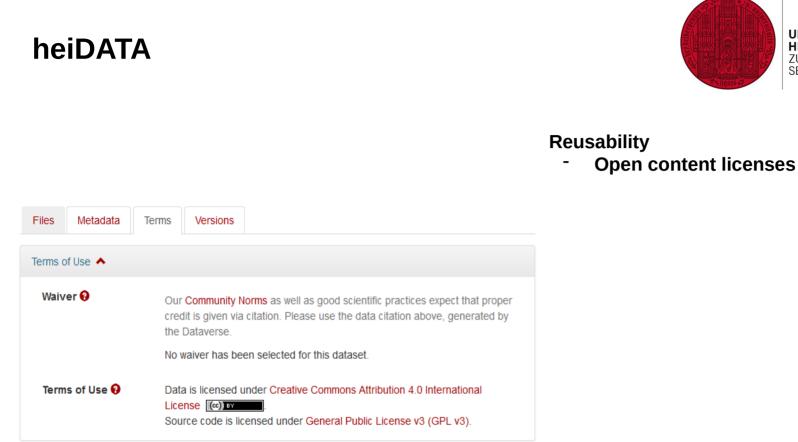
**RESEARCH DATA UNIT** HEIDELBERG

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Interoperability - Metadata standards

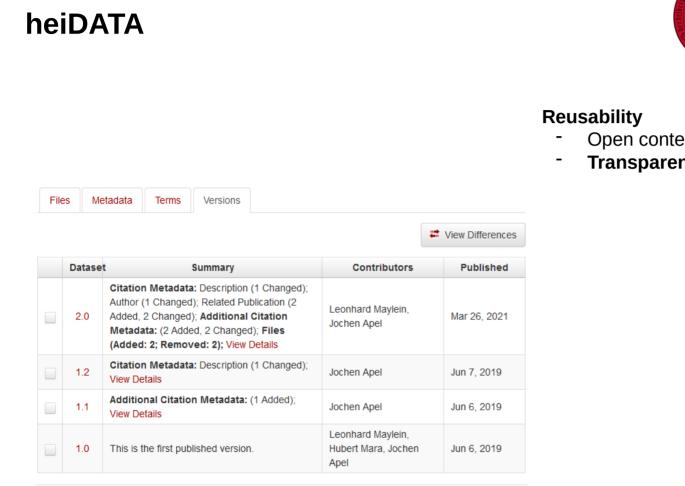
- Advice on suitable file formats, support with format conversion
- Technical validity checks



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- Open content licenses
- **Transparent versioning**



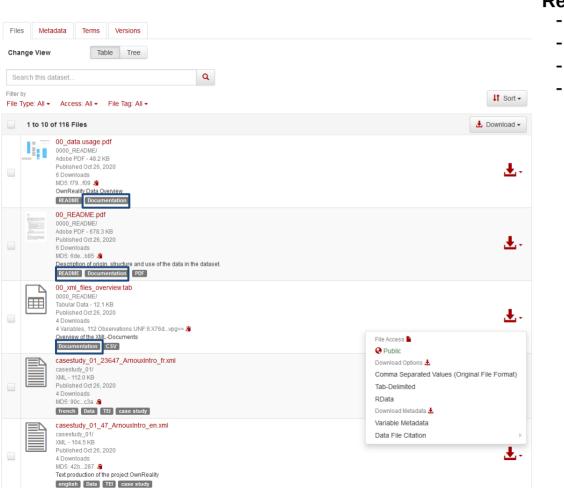


	-
Producer 9	Hubert Mara (IWR, Heidelberg University) (HMara) https://orcid.org /0000-0002-2004-4153 Bartosz Bogacz (IWR, Heidelberg University) (BBogacz) https://orcid.org /0000-0002-2004-4153
Production Date 😣	2019-03-11
Production Place 😯	Heidelberg, Germany
Contributor 😌	Project Member : Bayer, Paul Victor
Deposit Date 😯	2019-02-25
Date of Collection 😯	Start: 2018-07-24 ; End: 2018-08-22 Start: 2019-03-01 ; End: 2019-03-11
Kind of Data 😯	Cuneiform tablets; 3D Measurement data
Software 😯	GigaMesh Software Framework, Version: 181100 to 190300
Related Datasets 😯	Heidelberg Cuneiform 3D Database (HeiCu3Da) for the Hilprecht Collection: https://doi.org/10.11588/heidicon.hilprecht
Origin of Sources 😣	Hilprecht Sammlung, Jena, Germany, https://hilprecht.mpiwg-berlin.mpg.de/
	Cuneiform Digital Library Initiative (CDLI) https://cdli.ucla.edu/

#### Reusability

- Open content licenses
- Transparent versioning
- Provenance information





#### Reusability

Open content licenses

UNIVERSITÄT

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- transparent versioning
- Provenance information
- Documentation files



#### heiDATA > Synthetic Quantum Systems (SynQS) >

## Stochastic dynamics of a few sodium atoms in presence of a cold potassium cloud [data]

Version 2.0

No March Larbur	rí,	
	J U	

heiDATA, V2, UNF:6:JJrxDHuluVKxO7FoMyqIAw== [fileUNF]
Cite Dataset - Learn about Data Citation Standards.

Bhatt, Rohit Prasad; Kilinc, Jan; Höcker, Lilo; Jendrzejewski, Fred, 2021, "Stochastic dynamics of a few sodium atoms in presence of a cold potassium cloud [data]", https://doi.org/10.11588/data/HRCX1P,

Link Dataset
Contact Owner Share
Dataset Metrics 9

Edit Dataset -

178 Downloads 🚱



- Open content licenses
- transparent versioning
- Provenance information
- Documentation files
- Integration with external services, e.g. binder ( <u>https://mybinder.org/</u>)

 Description •
 We provide the data and our jupyler notebooks used to generate the figures of our publication.

 Abstract: Single particle resolution is a requirement for numerous experimental protocols that emulate the dynamics of small systems in a bath. Here, we accurately resolve through atom counting the stochastic dynamics of a few sodium atoms in presence of a cold potassium cloud. This capability enables us to rule out the effect of inter-species interaction on sodium atom numer dynamics, at very low atomic densities present in these experiments. We study the noise sources for sodium and potassium in a common framework. Thereby, we assign the detection limits to 4.3 atoms for potassium and 0.2 atoms (corresponding to 9% fidelity) for sodium. This opens possibilities for future experiments with a few atoms immersed in a quantum degenerate gas.

 Subject •
 Physics

Keyword 😯	Ultracold mixture, Stochastic dynamics
Related Publication 😣	Bhatt, R., Kilinc, J., Höcker, L., Jendrzejewski, F. Stochastic dynamics of a few sodium atoms in presence of a cold potassium cloud. Sci. Rep. doi: 10.1038/s41598-022-05778-8
Notes 😧	Run jupyter notebooks with binder: https://mybinder.org/

Files Atstadata Tarma Marsiana





## Solution binder

## Turn a Git repo into a collection of interactive notebooks

Have a repository full of Jupyter notebooks? With Binder, open those notebooks in an executable environment, making your code immediately reproducible by anyone, anywhere.

New to Binder? Get started with a Zero-to-Binder tutorial in Julia, Python or R.

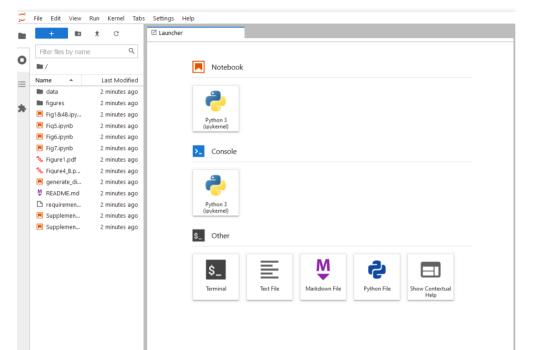
ataverse DOI - 10.11588/data/H	RCX1P		
ref (branch, tag, or commit)	Path to a notebook file (optional)		
EAD	Path to a notebook file (optional)	File 👻	
opy the URL below and share your E	Ninder with others: 2/10.11588/data/HRCX1P/		ŕ



#### Reusability

- Open content licenses
- transparent versioning
- Provenance information
- Documentation files
- Integration with external services, e.g. binder ( <u>https://mybinder.org/</u>)





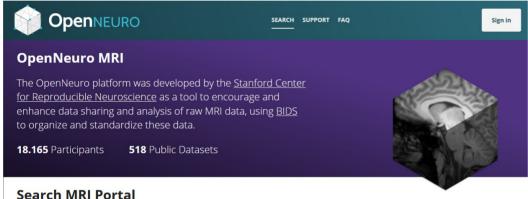


#### Reusability

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- Documentation files
- Integration with external services, e.g. binder ( <u>https://mybinder.org/</u>)





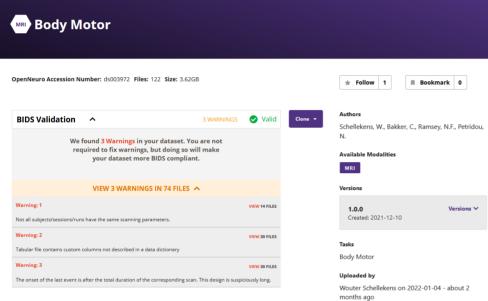


**Domain-specific metadata** ٠ enable specific functionalites and more effective retrieval.

Keywords ?	Loading Results CLEAR ALL
Enter Keyword(s) to Search	MODALITY: AGE: MRI X 20-30 X
Choose Modality	SORT BY: Relevance 1
Age of Participants	Increasing stimulus similarity drives nonmonotonic
20 30	Uploaded by: Compmem Lab on 2022-01-20 - about   Updated: 2022-01-11 - about 2 1 month ago months ago
Number of Participants	MODALITY:
Diagnosis ^	MRI
Healthy / Control	TASKS:
Schizophrenia	AB Pairs Random Ordered
ADD/ADHD	OPENNEURO ACCESSION NUMBER: ds004006         SESSIONS: 1         PARTICIPANTS: 41           PARTICIPANTS' AGES: 18 - 35         SIZE: 110.45GB         FILES: 721

Alzheimers





🛓 Download	🐱 Metadata
🖨 Body Motor \land	
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log dataset_description.json	7
(a) participants.json	7
participants.tsv	¥ @
README	7 3

Last Updated 2021-12-10 - 3 months ago Session

Participants

8

Dataset DOI

doi:10.18112/openneuro.ds003972.v1.0.0



- Domain-specific metadata enable • specific functionalites and more effective retrieval.
- Data standards are • implemented and data are validated against these standards.



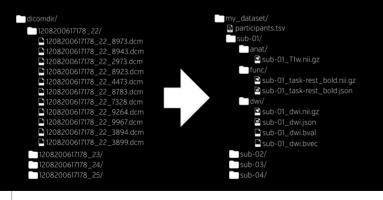
#### **Brain Imaging Data Structure**

A simple and intuitive way to organize and describe your neuroimaging and behavioral data.

ABOUT NEWS BENEFITS SPECIFICATION GET STARTED GET INVOLVED GOVERNANCE ACKNOWLEDGMENTS

#### About **BIDS**

Neuroimaging experiments result in complicated data that can be arranged in many different ways. So far there is no consensus how to organize and share data obtained in neuroimaging experiments. Even two researchers working in the same lab can opt to arrange their data in a different way. Lack of consensus (or a standard) leads to misunderstandings and time wasted on rearranging data or rewriting scripts expecting certain structure. With the Brain Imaging Data Structure (BIDS), we describe a simple and easy to adopt way of organizing neuroimaging and behavioral data.







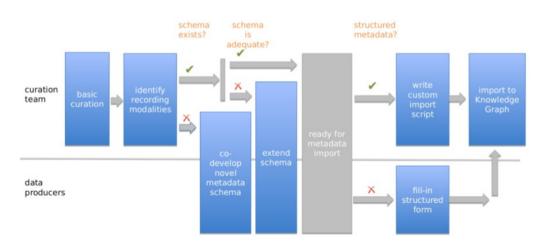
- Domain-specific metadata enable specific functionalites and more effective retrieval.
- Data standards are implemented and data are validated against these standards.



EBRAINS			Share data About Login
<b>Q</b>  Search (e.g. brain or r	neuroscie	ence)	(i) SEARCH
ATEGORIES		Viewing 1-20 of 1215 results	Sort by Relevance -
Project	124		
Dataset	1215		
Aodel	107	1000BRAINS study, connectivity data (v1.1)	
Software	163	The human brain shows considerable interindividual vari	ability, particularly during the course of aging, which
Contributor	1763	is influenced by genetic and environmental factors. To cl	haracterize this variability across a wide range o
Sont ibutor	1703	Keywords: • imaging	
ILTERS	Reset	Methods : © diffusion-weighted magnetic resonance imaging (DV	VI)
IODALITY		fiber tract reconstructions	
microscopy	828	<ul> <li>Spatial atlas registration</li> </ul>	
neuroimaging	754		
histological approach	596		
anatomical approach	586		
anatomy	232	Preprocessed data from the Individual Brain Cha	arting (IBC) project
neural connectivity	210	We present the preprocessed version of the Individual B	
histology	208	multi-task, functional Magnetic Resonance Imaging data	set, intended to support the investigation on the
expression characterization	100	functio	
electrophysiology	99	Keywords:	
multimodal research	90		
iew more		• audition	
		behaviour assay	
PECIES		Methods:	
Homo sapiens	852	<ul> <li>functional magnetic resonance imaging (fMRI)</li> </ul>	
Mus musculus	155	nonlinear image registration	
Rattus norvegicus	115	<ul> <li>nonlinear transformation</li> </ul>	
Macaca fascicularis	33	• rigid motion correction	
Macaca mulatta	14	<ul> <li>anatomical segmentation technique</li> </ul>	
Mustela putorius	4		
Chlorocobus acthions sabaous	2		

Danio rerio

- Domain-specific metadata enable specific functionalites and more effective retrieval.
- Data standards are implemented and data are validated against these standards.
- May be limited with regard to data types (e.g. OpenNeuro only accepts human-derived data)



https://wiki.ebrains.eu/bin/view/Collabs/tier-3-data-curation/Data%20Curator%27s%20Handbook/



- Domain-specific metadata enable more effective retrieval or specific functionalites.
- Data standards are implemented and data are validated against these standards.
- May be limited with regard to data types (e.g. OpenNeuro only accepts human-derived data)
- Optional, depending on the repository: Data curators with specific expertise supervise data publication and help preparing data for deposit.



## Finding data journals



### https://www.forschungsdaten.org/index.php/Data\_Journals





scientific <b>data</b>	<u>View all journals</u>	Search Q Log in
xplore content Y About the journal Y Publish with us Y	<u>Sign up for al</u>	erts 🗘 RSS feed
ature > scientific data		
Dynamic World, Near real- time global 10m land use land cover mapping		



## **Data publication**







## **Data publication**





#### Contra

- My data are neither useful nor interesting for others.
- I want to publish my results first, before someone else uses my data.
- There is no time and no money for data processing and curation.
- My data contain personal data personal rights, difficult search for test persons, anonymising impossible.
- My data include copyrighted material.
- My funder has no interest in making the data publicly accessible.
- My data will not be understood or will be misunderstood. People will bother me with emails.
- There is no incentive. Why should I do all the work?



## **Data publication**



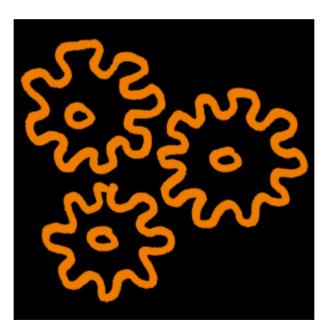
#### Pro

- Visibility and the accompanying scientific reputation
- Transparency and verifiability of research results
- Possibility of reuse of data in new contexts, for different problems in combination with other data and in interdisciplinary contexts
- New research possibilities through the "Data Web"
- Increased visibility of publications: Papers accompanied by research data are cited more often. See Piwowar, Day & Fridsma (2007), Piwowar & Vision (2013), Belter (2014), Henneken & Accomazzi (2011)
- Avoid duplicate work by reusing research data from third parties.
- Availability of negative results
- Fulfil requirements concerning the accessibility of research material as demanded by funders like DFG and EU, as well as scientific journals.
   Faster and more efficient circulation of knowledge
- Right of access to publicly funded results





## Data Handling Storage & Archiving – some practical issues....





#### Piled Higher and Deeper by Jorge Cham

#### www.phdcomics.com

A STORY TOLD IN FILE NAMES	:		
Location: 😂 C:\user\research\data			~
Filename 🔺	Date Modified	Size	Туре
😝 data_2010.05.28_test.dat	3:37 PM 5/28/2010	420 KB	DAT file
🚦 data_2010.05.28_re-test.dat	4:29 PM 5/28/2010	421 KB	DAT file
🚦 data_2010.05.28_re-re-test.dat	5:43 PM 5/28/2010	420 KB	DAT file
ata_2010.05.28_calibrate.dat	7:17 PM 5/28/2010	1,256 KB	DAT file
👸 data_2010.05.28_huh??.dat	7:20 PM 5/28/2010	30 KB	DAT file
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🛿 data_2010.05.29_crap.dat	3:22 AM 5/29/2010	437 KB	DAT file
data_2010.05.29_notbad.dat	4:16 AM 5/29/2010	670 KB	DAT file
data_2010.05.29_woohoo!!.dat	4:47 AM 5/29/2010	1,349 KB	DAT file
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DUNK	2:45 PM 5/29/2010		Folder
👪 data_2010.05.30_startingover.dat	8:37 AM 5/30/2010	420 KB	DAT file
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Type: Ph.D Thesis Modified: too many times	Copyright: Jorge Cham	www.phdo	omics.com

## **Data Handling Storage & Archiving**



### File handling

- Data best practices (file naming, formats, versioning,...): <u>https://guides.library.stanford.edu/data-best-practices/</u>
- Make different versions of data distinguishable. Conventions for file naming for you and in your research group.
- File names should deliver context. Distinguish a file from similar but different datasets and from different versions of the same dataset.
- Files may leave their folders. File names should be unique and descriptive without a directory structure.
- Never delete your raw data!
- But delete versions of processed data you do not need any longer.



## Data Storage & Archiving – some practical issues....



### **Backup**

- **3**...**2**...**1**... Backup!
  - At least 3 copies per file...
  - ...on at least 2 different media...
  - and 1 at a different spatial location.









# EXTERNAL REQUIREMENTS & POLICIES



Picture: https://www.flickr.com/photos/raummaschine/9214045295/

### **Policies & external requirements**



#### SATZUNG ZUR SICHERUNG GUTER WISSENSCHAFTLICHER PRAXIS UND ZUM UMGANG MIT WISSENSCHAFTLICHEM FEHLVERHALTEN

in der Fassung vom 28.09.2021

#### Präambel

Zur Wahrnehmung ihrer Verantwortung in den drei Handlungsfeldern Forschung, Studium und Lehre sowie Wissenstransfer trifft die Universität Heidelberg im gesetzlichen Rahmen Vorkehrungen zur Verankerung einer Kultur der guten wissenschaftlichen Praxis. Der Senat hat deshalb in seiner Sitzung vom 28.09.2021 gemäß § 3 Abs. 5 S. 4 LHG i.V.m. § 19 Abs. 1 S. 2 Nr. 10 LHG die folgenden Regelungen beschlossen, durch die die Leitlinien zur Sicherung guter wissenschaftlicher Praxis der Deutschen Forschungsgemeinschaft (DFG) vom August 2019 rechtsverbindlich umgesetzt werden:



## **Policies & external requirements**



Heidelberg University Rules for safeguarding good academic practice and handling academic misconduct

#### § 10 Documentation

(1) Researchers must document all information relevant to the establishment of a research result with the degree of transparency that is required and appropriate in the respective field. The same applies to individual results that do not support the research hypothesis. There must be no selection of results in such cases. Where research software is developed, the source code must be documented.

(2) The information required to understand the research, in particular research data and methodological, evaluation and analysis steps, is recorded. Third parties are to be given access to this information where this is possible.



## **Policies & external requirements**



Heidelberg University Rules for safeguarding good academic practice and handling academic misconduct

#### § 11 Public access to research findings

"Researchers decide on their own responsibility whether, how and where to make their research findings publicly available. If they decide to publish their results, the data and principal materials upon which the published work is based must be stored in recognised archives and repositories where this is possible. The provisions of § 14 must be observed."



## **Policies & external requirements**



Heidelberg University Rules for safeguarding good academic practice and handling academic misconduct

#### § 16 Archiving

"(1) Once they have been made publicly available, research data and findings, and particularly the materials on which they are based, as well as the instruments and, where applicable, the research software used, must be backed up by adequate means according to the standards of the respective field and stored for the legally required time period (usually ten years). A shortening of this storage period must be justified. The storage period begins when the materials are first made publicly available.

(2) The materials are archived a) in the researchers' home institution or b) in repositories serving several locations. In case a) the university will provide the necessary infrastructure for archiving. The selected publication medium must make reference to the archiving location in an appropriate



# Policies & external requirements RESEARCH DATA POLICY



#### **RICHTLINIEN FÜR DAS MANAGEMENT VON FORSCHUNGSDATEN**

Die Verfügbarkeit von Forschungsdaten ist die Gewähr für die Nachvollziehbarkeit und Überprüfbarkeit sowie die weitergehende Nutzung nach der Veröffentlichung. Sie ist ein zentraler Aspekt guter wissenschaftlicher Praxis der Universität. Ihr Management nach höchsten Standards baut auf diesem Prinzip auf und ist Teil der Exzellenzstrategie.

- 1. Die Verantwortlichkeit für den Lebenszyklus(\*) von Forschungsdaten, insbesondere die Sicherstellung und Bereitstellung der Forschungsdaten zur langfristigen Archivierung liegt primär beim Projektverantwortlichen (PI).
- 2. Teil jedes Forschungsprojektes ist ein Plan für das Datenmanagement, der explizit adressiert, wie die Akkuratheit, Vollständigkeit, Authentizität, Integrität, Vertraulichkeit, Veröffentlichung und der offene Zugang von Daten gehandhabt werden. Dabei werden fachspezifische Besonderheiten berücksichtigt.
- 3. Die Universität unterstützt nach bestem Vermögen die PIs durch ein Kompetenzzentrum Forschungsdaten. Es bietet Beratung und Unterstützung bei der Entwicklung von Konzepten für ihr Datenmanagement an. Dafür ist eine frühzeitige Kontaktaufnahme vor oder zu Projektbeginn erforderlich.
- 4. Der Plan für das Management von Forschungsdaten stellt den Zugriff und die Nutzung unter Einhaltung von ethischen und Open Access-Prinzipien unter geeigneten Sicherheitsmaßnahmen sicher. Der Open-Access-Policy der Universität folgend ermuntert die Universitätsleitung Wissenschaftler ausdrücklich, Forschungsdaten gemäß der Grundsätze von Open Access, wie sie in der "Berliner Erklärung über offenen Zugang zu wissenschaftlichem Wissen" von 2003 beschrieben sind, zugänglich zu machen, solange keine entgegenstehenden rechtlichen Verpflichtungen bestehen (insb. Verträge mit Verlagen). Für Daten, die Grundlage von schutzfähigem, geistigem Eigentum sind, gilt grundsätzlich die Verpflichtung zur Einreichung einer Erfindungsmeldung gemäß Arbeitnehmererfindungsgesetz (§§ 5, 42 Nr. 2) und die IP-Policy der Universität Heidelberg vorrangig.
- 5. Die persönlichen Daten von Probanden, Patienten und andere von Datenerhebungen betroffenen Personen werden gemäß den Datenschutzrichtlinien geschützt.
- 6. Daten, die außerhalb der Universität als Teil des Datenmanagementplans bereitgehalten werden, sollten beim Kompetenzzentrum Forschungsdaten registriert werden. Das Kompetenzzentrum Forschungsdaten bietet eine Datenregistrierung an, die Datensätze sowohl aus universitären als auch externen Repositorien nachweist.
- 7. Alle Rechte an Daten, insbesondere das Recht, die Daten weitergehend zu nutzen oder zu publizieren, sollten den PIs vorbehalten sein und nicht an Dritte vergeben werden.



# Policies & external requirements RESEARCH DATA POLICY



RICHTLINIEN FÜR DAS MANAGEMENT VON FORSCHUNGSDATEN

#### Seven paragraphs

Universität. Ihr Management nach höchsten Standards haut auf diesem Prinzip auf und ist Teil der

- 1) PI's are responsible for the whole research data lifecycle.
- 2) Every research project should develop a data management plan.
- 3) University offers support via the Research Data Unit.
- 4) University encourages researcher to publish open access if possible.
- 5) Importance of data privacy.
- Data published outside of the university's webspace should be registered at the RDU.
- PI's shall keep their right on data use and publication and shall not transfer it to third parties.
- Research Data Policy Universität Heidelberg (uni-heidelberg.de)

Alle Rechte an Daten, insbesondere das Recht, die Daten weitergenend zu nutzen oder zu publizieren, soliten den Pis vorbenalten sein und nicht an

vergeben werden



#### Funders are pushing RDM & Open Data



#### **DFG** Deutsche Forschungsgemeinschaft

#### DFG Guidelines on the Handling of Research Data

"[...] For this reason, the handling of research data and the objects on which the data is based have to be carefully planned, documented and described. Wherever possible it is important to enable subsequent use of the research data and potentially also the objects by other users.

[...] For this reason, the DFG expects research projects to include a description of how research data is handled. The description should be based on the checklist for handling research data [...]

Costs incurred for the project-specific handling of research data should be requested in connection with the project.[...]"





#### **Checklist Regarding the Handling of Research Data**

1. Data description

How does your project generate new data? Is existing data reused? Which data types (in terms of data formats like image data, text data or measurement data) arise in your project and in what way are they further processed? To what extent do these arise or what is the anticipated data volume?

2. Documentation and data quality

What approaches are being taken to describe the data in a comprehensible manner (such as the use of available metadata, documentation standards or ontologies)? What measures are being adopted to ensure high data quality? Are quality controls in place and if so, how do they operate? Which digital methods and tools (e.g. software) are required to use the data?

Storage and technical archiving the project
 How is the data to be stored and archived throughout the project duration? What is in



Funders are pushing RDM & Open Data





#### Horizon 2020 & Horizon Europe: FAIR Data Management

- Participating projects will be required to develop a Data Management Plan (DMP)
- Participating projects are required to deposit research data, preferably into a research data repository
- "[...]as far as possible, projects must then **take measures to enable for third parties to access**, mine, exploit, reproduce and disseminate (free of charge for any user) this research data."
- <u>http://www.dfg.de/foerderung/antrag\_gutachter\_gremien/antragstellende/nachnutzung\_forschungsdaten/</u>
- Guidelines on Open Access to Scientific Publications and Research Data in Horizon 2020 Guidelines on Data Management in Horizon 2020





#### 1. Data Summary

What is the purpose of the data collection/generation and its relation to the objectives of the project?

What types and formats of data will the project generate/collect?

Will you re-use any existing data and how?

What is the origin of the data?

What is the expected size of the data?

To whom might it be useful ('data utility')?

#### 2. FAIR data

#### 2. 1. Making data findable, including provisions for metadata

Are the data produced and/or used in the project discoverable with metadata, identifiable and locatable by means of a standard identification mechanism (e.g. persistent and unique identifiers such as Digital Object Identifiers)?

What naming conventions do you follow?

Will search keywords be provided that optimize possibilities for re-use?



## Journals: Nature

UNIVERSITÄT HEIDELBERG ZUKUNFT SEIT 1386

An inherent principle of publication is that others should be able to replicate and build upon the authors' published claims. A condition of publication in a Nature Portfolio journal is that **authors are required to make materials, data, code, and associated protocols promptly available to readers without undue qualifications.** 

[...]Providing large datasets in supplementary information is strongly discouraged and the preferred approach is to make data available in repositories.

https://www.nature.com/nature-portfolio/editorial-policies/reporting-st andards#availability-of-data

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



Journals: PLOS

#### Data Availability

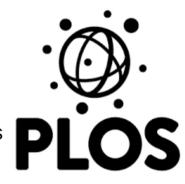
PLOS journals require authors to make all data necessary to replicate their study's findings publicly available without restriction at the time of publication. When specific legal or ethical restrictions prohibit public sharing of a data set, authors must indicate how others may obtain access to the data.

#### [...]

Publication is conditional on compliance with this policy. If restrictions on access to data come to light after publication, we reserve the right to post a Correction, an Editorial Expression of Concern, contact the authors' institutions and funders, or, in extreme cases, retract the publication. [...]

https://journals.plos.org/plosone/s/data-availability









# **LEGAL ISSUES**



## Legal issues



#### Research data and copyright

- Textual data typically are protected by copyright
- Copy right holder can grant simple or exclusive usage rights
- For publications in subscription journals: typically unlimited and irrevocable transfer of rights to the publishers
- Research data? Facts like measurements generally do not reach the threshold of originality, even though the data collection can be very sophisticated.
- Therefore: According to German copyright law, research data are in many cases not copyrighted.
- But many data are in databases and there is some kind of protection for these (EU directive 96/09/EG, UrhG §§ 87a-e). Virtually all data are useless without documentation. This documentation might very well be protected by copyright.



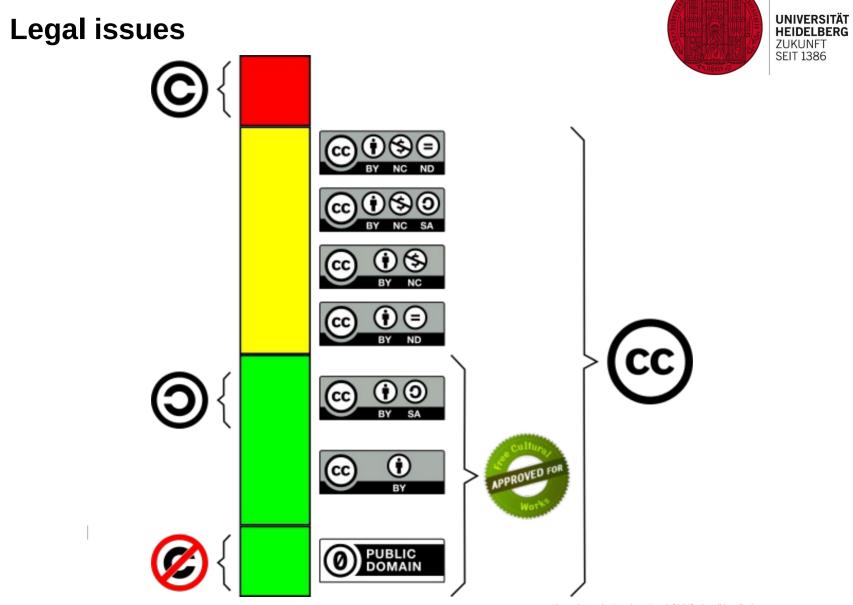
## Legal issues



#### **Creative Commons Licences**

- Standard licences that determine the scope of use of a work
- Combination of layperson-friendly formulation and a legally proper license text adapted to the relevant national law.
- Modular structure with differing "degrees of freedom"
- There are also alternatives, e.g. the Open Data Commons licenses.
- For Software there are specific software licenses





#### Legal issues



#### Data publication and data protection

• Informed consent: data sharing not excluded; information on whether and how data are disseminated

#### Beispiel UK Data Archive "Managing and Sharing Data": SAMPLE CONSENT FORM FOR INTERVIEWS

CONSENT FORM FOR [NAME OF PROJECT]		
Please tick the appropriate boxes	Yes	No
Use of the information I provide for this project only I understand my personal details such as phone number and address will not be revealed to people outside the project.		
I understand that my words may be quoted in publications, reports, web pages, and other research outputs.		
Use of the information I provide beyond this project I agree for the data I provide to be archived at the UK Data Archive. <sup>b</sup>		
I understand that other genuine researchers will have access to this data only if they agree to preserve the confidentiality of the information as requested in this form.		
I understand that other genuine researchers may use my words in publications, reports, web pages, and other research outputs, only if they agree to preserve the confidentiality of the information as requested in this form.		



# Thank you very much!

Dr. Georg Schwesinger <u>schwesinger@ub.uni-heidelberg.de</u>

Dr. Sebastian Zangerle sebastian.zangerle@ub.uni-heidelberg.de Research Data Unit <u>https://data.uni-heidelberg.de/</u>

General Information on RDM https://www.forschungsdaten.info/

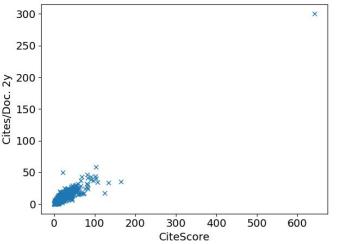




# 3. Research Data Quality



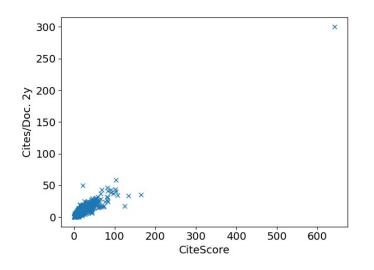
- Data must contain all ranges of the condition that is to be sampled
  - For example: To predict the impact of temperature on reactivity, all temperatures that are of interest need to be sampled (predictions only interpolate between data points but cannot extrapolate).
  - For example: CiteScore (Scopus citation index) vs. citations over all documents from last 2 years, for scientific journals.



Dataset: journal ranking dataset https://www.kaggle.com/datasets/xabirhasan/journal-ranking-dataset

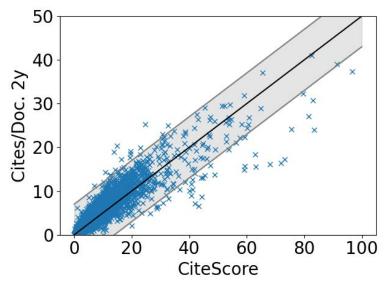


- Data must be homogeneous throughout feature space
  - For example: If temperature and pressure are both sampled, all combinations of features must be recorded for a homogeneous distribution of data points.
  - For example: CiteScore (Scopus citation index) vs. citations over all documents from last 2 years, for scientific journals.





- Data must be of good quality
  - Whether it is real or synthetic data, the model can only make accurate predictions if the data itself is accurate.
  - For example: CiteScore (Scopus citation index) vs. citations over all documents from last 2 years, for scientific journals

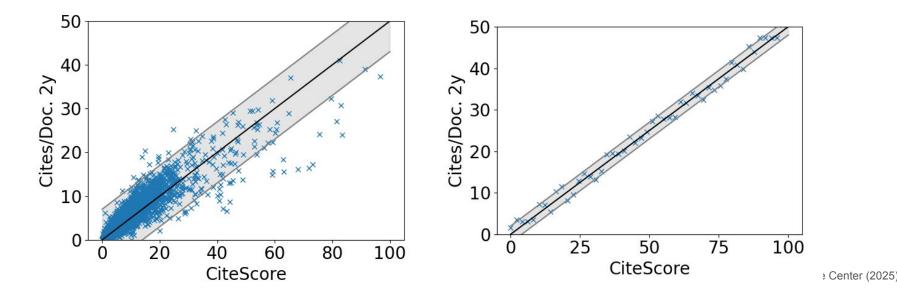




19

# Collecting data

- Data volume must be sufficient
  - Only with enough data can a model be trained to make accurate predictions.
  - For example: Complex data more data points required; simpler data fewer data points required





• Depending on the type of learning, data must be labeled and labeled correctly • Incorrect labelling interferes with the learning process.



Photo by nishizuka: https://www.pexels.com/photo/brown-chihuahua-485294/



Photo by Maksim Goncharenok: https://www.pexels.com/photo/a-chocolate-muffin-on-blue-surfac e-5994864/



- Make sure data is clean.
  - Correct typos, misidentified data types

#### Chihuahuah →Chihuahua



Photo by nishizuka: https://www.pexels.com/photo/brown-chihuahua-485294/

#### "26-04-24" →2024-04-26

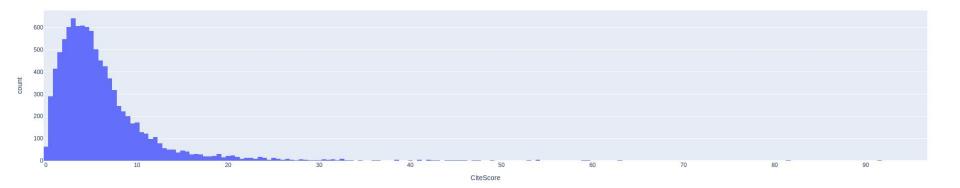


- Make sure data is homogeneous.
  - Visualize the data and use clustering analysis to identify outliers.
  - Use df.describe() and plotly.express to better understand your data





- Remove duplicates.
  - Duplicates introduce bias.
  - Use df.drop\_duplicates()



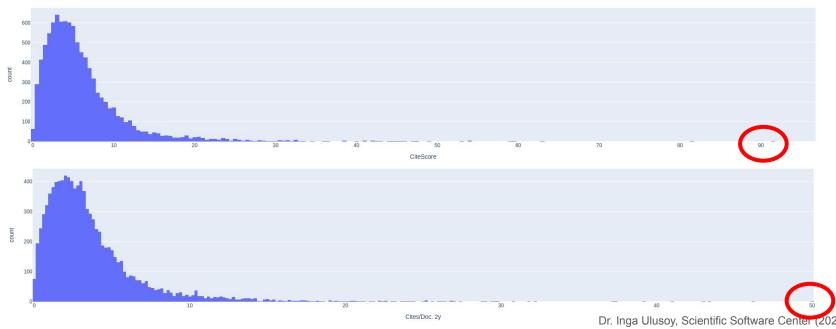


- Feature Engineering: Select influential features, remove unnecessary ones.
  - Unimportant features increase the complexity and reduce robustness.
  - For example: only choose features that are clearly correlated

	Rank	OA	SJR- index	CiteScore	H-index	Best Subject Rank	Total Docs.	Total Docs. 3y	Total Refs.	Total Cites 3y	Citable Docs. 3y	Cites/Doc. 2y
Rank	1.000000	0.111300	-0.503617	-0.485568	-0.625403	0.558208	-0.192069	-0.196795	-0.196338	-0.243070	-0.185484	-0.560625
OA	0.111300	1.000000	-0.069304	-0.056997	-0.178146	0.114037	0.061870					-0.045120
SJR-index	-0.503617	-0.069304	1.000000	0.878000	0.565015	-0.281225	0.091092	0.102424	0.094227	0.270083		0.828618
CiteScore	-0.485568	-0.056997	0.878000	1.000000	0.527957	-0.279983	0.112000	0.127705	0.122350	0.285965	0.110357	0.943584
H-index	-0.625403	-0.178146	0.565015	0.527957	1.000000	-0.362788	0.331053	0.393130	0.313698	0.505095	0.362266	0.512423
Best Subject Rank	0.558208	0.114037	-0.281225	-0.279983	-0.362788	1.000000	-0.114754	-0.117089	-0.132615	-0.150247	-0.118463	-0.334142
Total Docs.	-0.192069	0.061870	0.091092	0.112000	0.331053	-0.114754	1.000000	0.934468	0.968011	0.806830	0.932626	0.150987
Total Docs. 3y	-0.196795	0.046403	0.102424	0.127705	0.393130	-0.117089	0.934468	1.000000	0.887417	0.854647	0.995085	0.148272
Total Refs.	-0.196338	0.058683	0.094227	0.122350	0.313698	-0.132615	0.968011	0.887417	1.000000	0.802696	0.893789	0.173401
Total Cites 3y	-0.243070	0.024084	0.270083	0.285965	0.505095	-0.150247	0.806830	0.854647	0.802696	1.000000	0.844114	0.308644
Citable Docs. 3y	-0.185484		0.081086	0.110357	0.362266	-0.118463	0.932626	0.995085	0.893789	0.844114	1.000000	0.139525
Cites/Doc. 2y	-0.560625	-0.045120	0.828618	0.943584	0.512423	-0.334142	0.150987	0.148272	0.173401	0.308644	0.139525	1.000000
Refs./Doc.	-0.390894	-0.064572	0.267383	0.306913	0.247259	-0.299281	0.032381	0.019822	0.109949	0.076826	0.030626	0.382891
Life Sciences	-0.166150	0.073645	0.071380	0.125088	0.210414	-0.183625	0.044939	0.050866	0.068018	0.051387	0.051948	0.114416

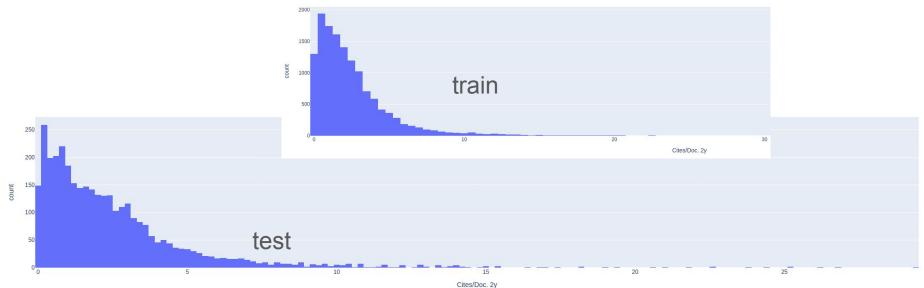


- Feature Engineering: Normalize features.
  - Features should have similar data ranges for the weights to be in similar ranges, and improved model robustness and faster training.



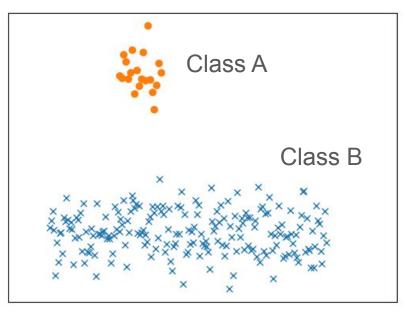


- Make sure to randomize your data.
  - Otherwise, your train and test data could contain more/less data of a certain kind (inhomogeneous data)



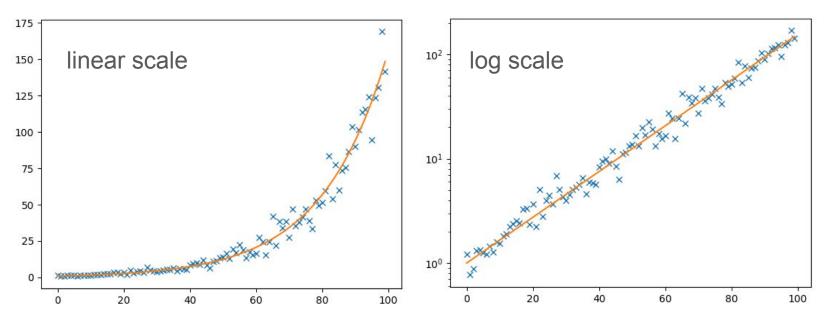


- Feature engineering: Make sure your dataset is balanced.
  - For classification tasks, all classes should have comparable sizes (similar numbers of examples).





- Feature engineering: Pick the right scale.
  - Visualize your data to see if you need to transform ie. onto a log scale.





# 4. Modeling of Research Data

#### Deep learning – Tools and Tricks

I wish I had known earlier

Peter Lippmann

13.02.2025

Scientific AI Lab, IWR, Heidelberg University

$$\min_{\theta} \mathbb{E}_{(x,y)\sim \mathcal{D}} L(y, f_{\theta}(x))$$

 $\mathcal{D}$ : data distribution, typically approximated by a set of finite set of input-target pairs  $\{x_i, y_i\}$  $f_{\theta}$ : neural network with learnable parameters  $\theta$ 

L: differentiable loss function, typically minimal if  $y = f_{\theta}(x)$ 

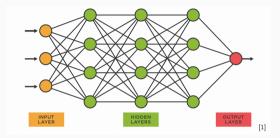
In practice:

$$\min_{\theta} \sum_{\substack{\text{sample } (x_i, y_i) \\ \text{in training set}}} L(y_i, f_{\theta}(x_i))$$

Optimization via (mini batch) gradient descent:

$$\theta^{(t+1)} = \theta^{(t)} - \alpha \sum_{\text{sample } (x_i, y_i)} \nabla_{\theta} L(y_i, f_{\theta}(x_i)) \Big)$$

#### The neural network



- Network types: CNNs for images, GNNs for graphs, sequence models for language, recurrent NNs for time series data, ... (often clear what to choose)
- BUT many representatives, e.g. many different CNN architectures (less clear)
- For given architecture, several hyperparameters (educated guess + trying out)



- heavily depends on your task, e.g.:
  - classification  $\leftrightarrow$  Cross Entropy loss,
  - regression  $\leftrightarrow$  MSE (a.k.a. L2-loss)  $||y_i f_{\theta}(x_i)||^2$
- Tip: for regression L1-loss  $|y_i f_{\theta}(x_i)|$  can be more robust to outliers
- sometimes a combination of losses is used (weighting them can be tricky)

#### The data

- more data is better, higher quality data is better
- visualize your data before: PCA, UMAP, t-SNE
- check your data is balanced (e.g. in instances per class)
- split your data into train, validation and test set
- standardize your data (both input and output)
- use data augmentation if possible



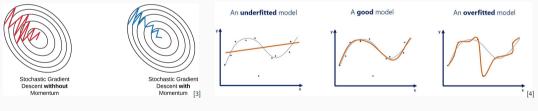
[2]

Data augmentation

#### The optimizer

Use AdamW (Adaptive Moment Estimation + weight decay):

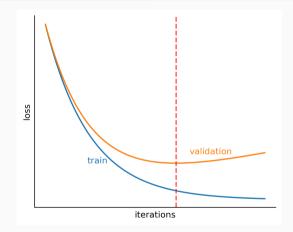
- adaptive learning rate helps against too small or too large gradients
- momentum stabalizes the gradient descent
- weight decay helps against overfitting



(a) Effect of momentum

(b) Overfitting model

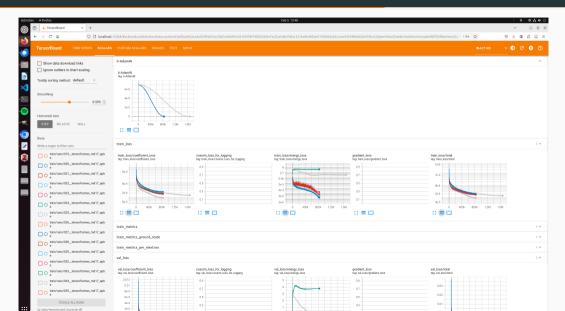
#### More tricks against overfitting



- Dropout: randomly drop/ignore neurons during training
- Save checkpoints of your model: last (if training crashes) & k-many best

- try to overfit on a single sample to debug your pipeline
- set a seed during training for reproducibility
- use enough CPU workers in dataloader to properly use GPU
- in dataloader use shuffle=True and drop\_last=True
- try gradient clipping in the optimizer against instable training
- use a learning rate scheduler (e.g. cosine schedule)
- try learning rate warmup
- definitely try normalization layers: helps to standardize activations

#### Logging via Tensorboard



8

#### **Pytorch Lightning**

 $Py torch \ lightning \ module \ combines \ py torch \ model \ + \ optimizer \ + \ logging$ 

- abstracts away to("cuda"), loss.backward(), model.eval() and much more
- makes complicated things which many people use easy, e.g. multi GPU support

```
import lightning as L
import torch
from lightning:pytorch.demos import Transformer
class lightningTransformer(L.LightningModule):
    def __init__(self, vocab_size):
        self.model = Transformer(vocab_size=vocab_size)
    def forward(self, inputs, target):
        return self.model(inputs, target)
    def training_step(self, batch, batch,idx):
        inputs, target = batch
        output = self(inputs, target)
        loss = torch.nn.functional.nll_loss(output, target.view(-1))
        return torch.optim.SED(self.model.parameters(), lr=0.1)
```

Many great tutorials at https://lightning.ai/docs/pytorch/stable/starter/introduction.html

#### Config management with Hydra

• save on boilerplate by "programming" in configs (customize models in config not in code)

```
edge mlp:
  target : mldft.ml.models.components.mlp.MLP
 hidden channels: [768, 32]
  activation laver:
  target : mldft.ml.models.components.mlp.MLP
 in channels: 768
 hidden channels: [768, 1]
  activation laver:
    target : hydra.utils.get class
 disable dropout last laver: True
 disable activation last laver: True
 disable norm last laver: True
```

- easy to add new models, datasets, tasks and experiments
- uses OmegaConf for configuration management

Great Hydra + Lightning template at

https://github.com/ashleve/lightning-hydra-template

### Thank You!

Any Questions?

- https://www.marktechpost.com/wp-content/uploads/2022/09/Screen-Shot-2022-09-23-at-10.46. 58-PM.png
- [2] https://media.datacamp.com/legacy/image/upload/v1669203370/Data\_Augmentation\_Header\_ f42227f2cb.png
- [3] https://i.sstatic.net/epW89.jpg
- [4] https:

//static.wixstatic.com/media/0ed3e8\_a9b7d6d3dc6b4d5cbcb30c8b2fd4782b~mv2.jpg/v1/fill/w\_
1000,h\_449,al\_c,q\_90,usm\_0.66\_1.00\_0.01/0ed3e8\_a9b7d6d3dc6b4d5cbcb30c8b2fd4782b~mv2.jpg



# Model Metadata



# Model card

- Model details
  - Architecture, parameters, citation information, license information
- Intended use
  - Use cases within the model's scope
- Performance metrics
  - Intended performance on given data
- Training data
  - Description of training data and data distribution
- Quantitative analysis
  - Potential biases and limitations
- Ethical consideration
  - Privacy and fairness concerns, impact on society
- https://huggingface.co/spaces/huggingface/Model\_Cards\_Writing\_Tool
- https://github.com/openai/gpt-3/blob/master/model-card.md



# Where to share/publish/deploy your models



# Model sharing platforms

You can make models available for others on model sharing platforms like

- Hugging face,
- OpenML,
- Kaggle.

*Advantages:* Public platform with version control and model cards, you can link the data into the repo, allows others to use your model for production or fine-tuning.



# How to test your software that is based on ML models



# Testing of non-deterministic processes

Try to make processes deterministic

For example: Use specified random seed.

Separate deterministic and non-deterministic processes and test separately

For example: Input processing can be tested separately from model prediction.

predictions.

Test for output parameters and properties that remain constant For example: Number of predictions, feature length, etc.

Include multiple valid outputs in your tests

For example: Three most probable classifications.

Robustness: A robust model is more likely to behave like a deterministic system

Make sure your model output is stable under a range of conditions.

Accuracy: The model accuracy will affect the testing strategy A higher accuracy leads to more consistent

Distribution: You can also test for the distribution of results rather than the results themselves



# Deploying machine-learning models and software



# Model deployment

In addition to making models and software available for others to use in their own code, you can also directly deploy the model - together with your code - directly so that it can be used.

#### Examples:

- Diffusers: google colab <u>https://colab.research.google.com/github/huggingface/notebooks/blob/main/di</u> <u>ffusers/stable\_diffusion.ipynb</u>
- <u>https://lightning.ai/</u> for paid service and deployable models
- See

<u>https://www.freecodecamp.org/news/deploy-your-machine-learning-models-fo</u> <u>r-free/</u> for tutorials and services



# 5. Software Engineering best practices



# Version Control: git

#### What is this?

A tool to allow you to track and revert changes, and collaborate with others (change management).

- Allow versioning of the code and continuing functionality.
- Allow simultaneous changes to the same files.
- Fundamental for reproducing historic states in the line of development.
- Development follows a story and allows other users to build confidence in your work.



# Version Control in practice: git

- Create a repo on GitHub
- Clone the repo to your local machine
- Checkout a branch and make changes
- git add, git commit and git push: IDEs such as VSCode make it easy for you!
- Observe how your repo changes on GitHub
- GitHub offers great learning labs: GitHub skills https://skills.github.com/
- roadmap.sh offers a git roadmap: <u>https://roadmap.sh/git-github</u>





# Development workflows: GitHub-flow

#### What is this?

GitHub-flow is a lightweight workflow with creating branches, making changes, opening Pull Requests, running Continuous Integration, and requesting code review, together with Issues and Kanban project boards.

- Manage how changes are incorporated in the software.
- Track progress in your project and highlight bottlenecks.
- Adhere to development guidelines, ensuring the implementation follows defined rules to ensure software quality.

# Development workflows in practice: GitHub-flow

- After you made changes in your git branch, open a Pull Request on GitHub
- Observe how the PR highlights the changes in the line of development
- You can link issues, comment on the PR and run automated checks
- See GitHub skills https://skills.github.com/
- See roadmap.sh https://roadmap.sh/git-github

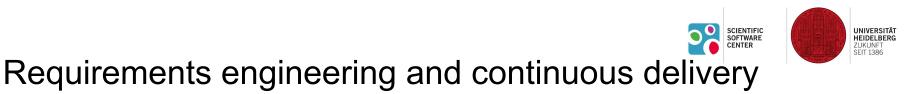


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#### What is this?

Requirements engineering is the process of translating stakeholder requirements on the research software into defined tasks. Early delivery and iteration over it allows refinement of the requirements and tasks.

- Ensure that the software fulfills its purpose. In research software, requirements engineering is closely intertwined with the research process and subject to frequent changes.
- Understand the problem the software should solve and map this onto an efficient technically feasible solution considering all constraints.
- Allows prioritization of requirements/tasks and decision-making. Decisions should be documented together with the requirements.



# Requirements engineering in practice

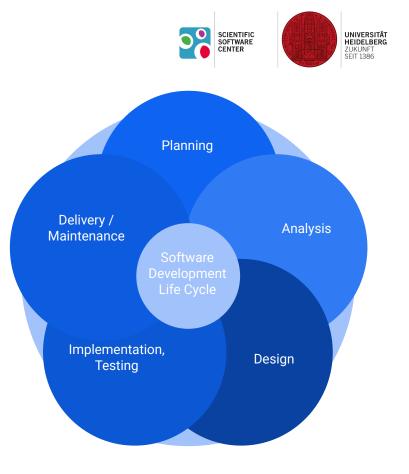
- Functional requirements (what the software should do: features)
- Non-functional requirements (how the software performs a task: ie performance, security)
- Domain requirements (specific to the domain: ie. Healthcare)
- Use tools such as draft.io or miro to gather requirements

As a <type of user>, I want <some goal> so that <some reason>.

```
As a <researcher>, I want
<to obtain the
probability of a class>
so that <I can classify
incoming images>.
```

# Continuous delivery in practice

- Deliver early to find out if your software is fulfilling its purpose/moving in the right direction
- Use quality control to allow early delivery through the main branch in your GitHub-flow (*keep your main branch operational at all times*)
- Following agile / lean principles
- Use git to allow consistent usability of your software





# Project management: Kanban boards

#### What is this?

Project management is used to track progress, identify intertwined or dependent processes, and allows visual access to the project's status.

- Ensure that the software development moves in the right direction.
- Increase the flow of ongoing work.
- Allow prioritization of tasks and understand interdependencies and bottlenecks in the development.



# Project management in practice

- Use a Kanban board to organize tasks
- Separate tasks into backlog/todo, in progress, done
- Prioritize tasks and assign contributors/identify necessities
- For example, GitHub projects

O No Status ③ (Estimate: 0) ···· Entry point for all new initiatives	<b>Backlog</b> 9) (Estimate: 0) ···· Initiatives waiting assignation to SSC members	In Progress (3) (Estimate: 26) ···· Initiatives actively being worked by the SSC.	Done () (Estimate: 0) ···· Initiatives successfully completed, awaiting closing ceremony (feedback)	Here is when the project is archived once has	Ince has     Estimate: 0       Initiatives cancelled or stopped for any administrative reason.
() Draft MatlabWebApp	Oraft predicTCR Funded project Medical 2024	sc-projects #29 Establishing a knowledge graph community in biomedical science		© ssc-projects #27 VASP: Voxel Attributes and Statistics for Point Clouds	
ssc-projects #32     Onboarding SSC     PROJECT	Draft Control Dr	P0       21       XL       Funded project         Bioinformatics       2024       PROJECT         O       ssc-projects #28       S         Project management tools for SSC       P2       5       M       internal SSC       2024       PROJECT         O       ssc-projects #54       S       S       ParzivAl       S         Small-scale       2024       2024       S       S       S		Image: Solution and Solution       Image: Solution and So	
Oraft 🕀 🕀 Project-w Small-scale 2024	Oraft Sci-rocket			Hamming distance project           Small-scale         Maths         2020           Ssc-projects #34         Image: Comparison of the state of the	
	Small-scale Bioinformatics 2024			Neuroscience data processing and analysis Open call Biology 2021 PROJECT	



# Project planning: Architecture and design

#### What is this?

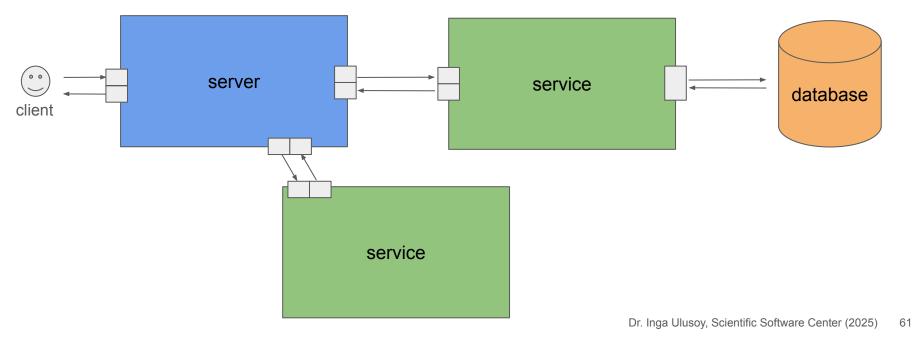
Software architecture describes how the system is composed of different pieces, and the interplay of the components. Design refers to the actual implementation of the requirements in the system as a whole and the different components.

- Makes the software efficient and allow re-use of functionalities.
- Allows extensions and additions of features at a later stage without major refactoring.
- Makes the software maintainable.



# Architecture in practice

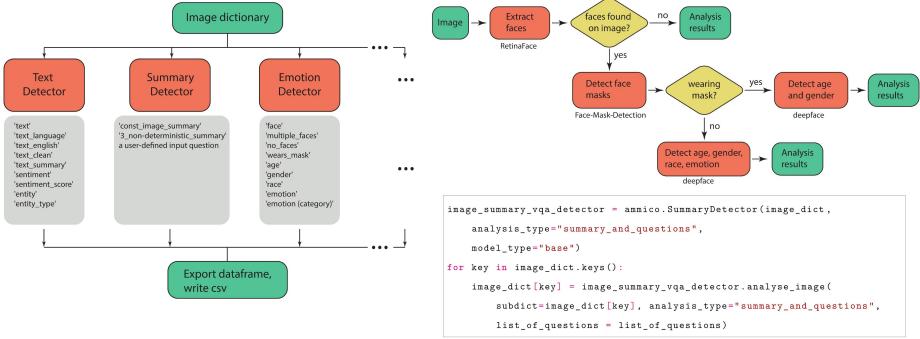
- Use (black/white) box diagrams to identify components and their interactions
- <u>https://roadmap.sh/software-design-architecture</u> / miro / draft.io / draw.io





# Design in practice

- Map the input/output, data formats, transformations / logic / processes
- <u>https://roadmap.sh/software-design-architecture</u> / miro / draft.io / draw.io





# Quality management: Testing and continuous Integration

#### What is this?

GitHub-flow is a lightweight workflow with creating branches, making changes, opening Pull Requests, running Continuous Integration, and requesting code review, together with Issues and Kanban project boards.

- Manage how changes are incorporated in the software.
- Track progress in your project and highlight bottlenecks.
- Adhere to development guidelines, ensuring the implementation follows defined rules to ensure software quality.



# Testing in practice

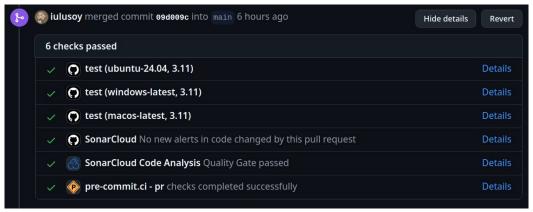
- Use testing frameworks such as pytest
- Write tests in a tests/ folder: unit tests, integration tests, system tests, compatibility tests, ...
- To learn how to use pytest: https://docs.pytest.org/en/stable/,

	26 27	@pytest.fixture			
∽ 💽 ammico	28	<pre>28 def accepted(monkeypatch):</pre>			
∽ 💽 test	29	<pre>monkeypatch.setenv("OTHER VAR", "True")</pre>			
> O test_races.py		<pre>tt.TextDetector({}, accept_privacy="OTHER_VAR")</pre>			
> 💽 test_multimodal_search.py	31	return "OTHER_VAR"			
> o test_summary.py	32				
✓ ⊘ test_text.py	33				
<ul> <li>test_privacy_statement</li> </ul>	▷ 34	<pre>def test_privacy_statement(monkeypatch):</pre>			
o test TextDetector	35	<pre># test pre-set variables: privacy</pre>			
		<pre>monkeypatch.delattr("builtins.input", raising=False)</pre>			
o test_run_spacy	37	<pre>monkeypatch.setenv("OTHER VAR", "something")</pre>			
<ul> <li>test_clean_text</li> </ul>	38	with pytest.raises(ValueError):			
<ul> <li>test_init_revision_numbers</li> </ul>		<pre>tt.TextDetector({}, accept_privacy="OTHER_VAR")</pre>			
⊘ test_check_add_space_aft		<pre>monkeypatch.setenv("OTHER_VAR", "False")</pre>			
⊘ test_truncate_text	41	with pytest.raises(ValueError):			
o test analyse image	42	<pre>tt.TextDetector({}, accept_privacy="OTHER_VAR")</pre>			



# Continuous integration in practice

- Set up your tests to be automatically run by GitHub actions
- Include code linter and quality control in your actions
- These should be set up to run automatically when you open a Pull Request
- GH actions, codecov, sonarcloud, snyk, pre-commit, code formatting (black), GitHub Guardian, dependabot





# Software Management Plans

#### What is this?

Software Management Plans (SMPs) help to identify goals and the means required to pursue the goals in practice.

- Identify criticality and required maturity of your software.
- Identify which measures are needed to ensure compliance of your software with the intended goals.
- Quantify milestones and tools for the intended purpose.

# SMPs in practice

#### RDMO for MPG Back to project

Software Project Schedule

When does the software project start?

When does the software project end?

#### Software Project Management

Which software development process is defined? How will process roles be assigned? How do you track the different tasks and use cases? Will there be a specification document (briefly) outlining the most important requirements?

Software Development Requirements

Are there institutional requirements for software development? Are there requirements regarding the software development form other parties?

#### Technical

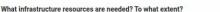
#### Code

Which programming language(s) do you plan to use? Which technology or process is used for versioning?

#### Third Party Components and Libraries

Which external software components will be used? What dependencies on software libraries do exist? How do you document this? What licences are on the third-party software components? What is the process to keep track of the external software components? Can critical dependencies be eliminated or mitigated? Do you plan to use third party web services? Does the software refer to other software projects or objects?

#### Infrastructure





- Use the SMPs provided by the Max Planck digital library
- Helps you with your requirements and project management
- https://rdmo.mpdl.mpg.de/



# Documentation

#### What is this?

Documentation can be comments, docstrings, readme's, tutorials, demonstration notebooks, and contains technical and domain-specific / application-specific descriptions of the software.

- Document what the software can and cannot do, and parameter ranges.
- Allow others to install and use your software (or yourself, at a later time).
- Allow others to contribute to your software.

## Documentation in practice

#### text module

class text.PostprocessText(mydict: dict | None = None, use\_csv: bool = False, csv\_path: str | None = None, analyze\_text: str = 'text\_english')

Bases: object

analyse\_topic(return\_topics: int = 3)→ tuple

Performs topic analysis using BERTopic.

 Parameters:
 return\_topics (int, optional) - Number of topics to return. Defaults to 3.

 Returns:
 tuple - A tuple containing the topic model, topic dataframe, and most frequent topics.

 $get\_text\_df(analyze\_text: str) \rightarrow list$ 

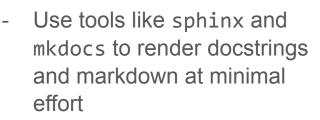
Extracts text from the provided dataframe.

Parameters:analyze\_text (str) - Column name for the text field to analyze.Returns:list - A list of text extracted from the dataframe.

#### get\_text\_dict(analyze\_text: str)→ list

Extracts text from the provided dictionary.

Parameters:analyze\_text (str) - Key for the text field to analyze.Returns:list - A list of text extracted from the dictionary.



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- Include jupyter notebooks that showcase use of your software - these can be run on google colab
- Document dependencies in a requirements file and provide installation instructions



## Deployment: Runtime environment / containerisation

#### What is this?

Deployment information such as runtime environments or containers allow easy adaption as they provide direct access to running the software without installation and dependency conflicts.

#### Why is this important?

- A big step towards reproducibility and transferability of your approach.
- The software ecosystem changes quickly, and this allows to preserve a snapshot that can be shared and run easily.



## Containerisation in practice

- Use docker to provide build instructions for containers, and possibly deploy the containers on Dockerhub for anyone to download and use
- Docker roadmap <u>https://roadmap.sh/docker</u>, official tutorial <u>https://docs.docker.com/</u>

1	FROM jupyter/base-notebook	
2		
3	# Install system dependencies for computer vision packages	
4	USER root	
5	RUN apt update && apt install -y build-essential libgl1 libglib2.0-0 libsm6 libxrender1 libxext0	5
6	USER \$NB_USER	
7		
8	# Copy the repository into the container	
9	COPYchown=\${NB_UID} . /opt/ammico	
10		
11	# Install the Python package	
12	RUN python -m pip install /opt/ammico	
13		
14	# Make JupyterLab the default for this application	
15	ENV JUPYTER_ENABLE_LAB=yes	
16		
17	# Export where the data is located	
18	ENV XDG_DATA_HOME=/opt/ammico/data	[



## Software Licensing

#### What is this?

A software license states the terms of use, re-use and distribution, among others, without violating copyrights, and defines responsibilities.

#### Why is this important?

- So that others may use your code, and to prevent misuse.
- So that others may contribute to your code.
- So that the responsibilities for how the software is used are clear.
- Establishes the rights of all parties involved with the software.



## Software licensing in practice

- Use the provided templates from GitHub: Either at repository creation or when adding a new file called LICENSE
- Permissive open-source license: BSD 2-Clause, MIT, Apache License 2.0
- Copyleft open-source license: GNU version 3, LGPL
- Proprietary licenses: Do not only keep your project close-source and potentially less visible, but also carry responsibilities for contract fulfillment
- <u>https://opensource.org/licenses</u>, <u>https://choosealicense.com/</u>
- When using/incorporating third-party software: ie. use of open-source libraries is the third-party code distributed with your software? If so, compatibility needs to be confirmed!



# 6. Making your work public: **Considerations of more** general use and prominent failures

Dr. Inga Ulusoy, Scientific Software Center (2025)



## **REFORMS** checklist

To ensure reproducibility, include a checklist such as the REFORMS checklist into the publication of your results.

There are a couple of domain-specific checklists, especially in health and life sciences (STARD, CLAIM, see the EQUATOR<sup>\*</sup> network), that may be more appropriate - REFORMS aims to be most general.

REFORMS checklist: DOI 10.48550/arXiv.2308.07832, <u>https://reforms.cs.princeton.edu/</u> \*EQUATOR Network, <u>https://www.equator-network.org/reporting-guidelines/</u>



- 1. Study goals
- 1a) Population or distribution about which the scientific claim is made

The group to which the claim will be generalized



1. Study goals

#### 1b) Motivation for choosing this population or distribution

Choice of a particular population of interest - pure scientific interest, need for applied knowledge, ...



1. Study goals

#### 1c) Motivation for the use of ML methods in the study

Why focus is on building a model that reliably maps input data to output data, instead of using traditional statistical methods?



2. Computational reproducibility

#### 2a) Dataset

*Cite datasets with permanent links to clarify which version; if contains sensitive data: release synthetic datasets (synthpop library)* 



2. Computational reproducibility

2b) Code

Cite exact version of code (DOI, version number, commit tag)



2. Computational reproducibility

#### **2c) Computing environment**

Details about the hardware (CPU, RAM, disk space), software (operating system, programming language, version number for each package used), and computing resources (time taken to generate the results)



2. Computational reproducibility

#### 2d) Documentation

Document installation, requirements, running the code, usage examples, expected results



2. Computational reproducibility

#### 2e) Reproduction script

Reproduction scripts that prepare the environment, install the code, download datasets, and run code to reproduce the results in the paper



- 3. Data quality
- 3a) Data source(s)

When, where, how data were collected, how ground-truth annotations were performed



- 3. Data quality
- 3b) Sampling frame

List of people or units from which a sample is drawn



3. Data quality

**3c) Justification for why the dataset is useful for the modeling task** 

Why the dataset is a good representative to model the question/answer



3. Data quality

#### 3d) Outcome variable

How outcome variable is defined - this is usually not a perfect match for what it should represent



3. Data quality

#### **3e) Number of samples in the dataset**

Total sample size, number of samples in each class, number of individual data in the dataset



- 3. Data quality
- **3f) Missingness**

Missing data



3. Data quality

#### 3g) Dataset for evaluation is representative

Justify why data is representative for target selected in 1a)



- 4. Data preprocessing
- 4a) Excluded data and rationale

Justify why particular subset of data was chosen



4. Data preprocessing

#### 4b) How impossible or corrupt samples are dealt with

How erroneous or impossible data points are identified and amended



4. Data preprocessing

#### 4c) Data transformations

Normalizing, augmenting, imputing missing data, oversampling - the latter two must be done separately on each fold of the dataset!



- 5. Modeling
- 5a) Model description

Input, output, type of model, loss function, algorithm



5. Modeling

#### **5b) Justification for the choice of model types implemented**

Model needs to be interpretable, types of models considered



5. Modeling

#### 5c) Model evaluation method

Model needs to be evaluated on test data different from training data



5. Modeling

#### 5d) Model selection method

How final model(s) and hyperparameters were selected



5. Modeling

#### 5e) Hyperparameter selection

How hyperparameters were optimized



- 5. Modeling
- 5f) Appropriate baselines

How baseline models were trained and optimized



6. Data leakage

#### 6a) Train-test separation is maintained

Use a hold-out test set that is also not used in synthetic data generation



6. Data leakage

#### **6b) Dependencies or duplicates between datasets**

Could be more than one sample from same origin (patient); or time series is split randomly



- 6. Data leakage
- **6c) Feature legitimacy**

Any feature in the training that somewhat identifies predicted variable



- 7. Metrics and uncertainty quantification
- 7a) Performance metrics used
- Proper choice of metric depending on application

See Leist et al., Sci. Adv. 8, eabk1942 (2022) Table 4 for appropriate metrics depending on the data and algorithm used



- 7. Metrics and uncertainty quantification
- 7b) Uncertainty estimates

Randomness in training or evaluation data, or training process



#### **REFORMS: Example**

- 7. Metrics and uncertainty quantification
- 7c) Appropriate statistical tests

Statistical testing of ML models

For an overview, see Sebastian Raschka, Model evaluation, model selection, and algorithm selection in machine learning, arXiv:1811.12808



#### **REFORMS: Example**

8. Generalizability and limitations

8a) Evidence of external validity

Report how claim is generalizable to target population.



#### **REFORMS: Example**

8. Generalizability and limitations

#### 8b) Contexts in which the study's findings will not hold

Clear expectations and unjustified hype.



# Software security



### Security

- Threat modelling: who, what, how
- Data-oriented attack: Access training data, poison data, inject trojan data
- Model-oriented attack: Modify training process (pre-trained malicious models), manipulate the deployed model (model patches, privacy information leakage, model inversion attacks)
- System-oriented attack: specialised hardware accelerators for ML software (SOC, trojan in GPU/TPU, for model corruption, backdoor insertion, model extraction, spoofing, information extraction, sybil attack)
- Possibility to carry out *pentests*



#### Security: best practices

Phases	Vulnerabilities Causes
Analysis phase	No risk analysis/ No security policy
	Biased risk analysis
	Unanticipated risks
Design phase	Relying on non-secure abstractions
	Security/Convenience tradeoff
	No logging
	Design does not capture all risks
Implementation phase	Insufficiently defensive input checking
	Non-atomic check and use
	Access validation errors
	Incorrect crypto primitive implementation
	Insecure handling of exceptional conditions
	Bugs in security logic
Deployment phase	Reuse in more hostile environments
	Complex or unnecessary configuration
	Insecure defaults
Maintenance phase	Feature interaction
	Insecure fallback

Chen, Barbar, Security for Machine Learning-based Software Systems, DOI 10.1145/3638531



## Legal aspects



#### Legal aspects

- If you reuse data / models / code: Make sure the license terms allow this and that your license(s) is (are) compatible
- Make sure you do not violate the DSGVO / GDPR / European Data Act / Copyright / European AI Act
- Once a model is made available, it is impossible to restrict its use!
- Examples:
  - Models can put out near-exact copies of images/text in training data, ie Dall-E/Stable Diffusion generating images with Shutterstock/Getty Images watermarks, or reproducing artist's work
  - Code-generating tools such as GitHub copilot allow recreation of code, that is already contained in other software, regardless of the license terms of that software



## **Ethical aspects**



#### **Ethical aspects**

- Be aware of people's tendency of overreliance!
  - ELIZA effect: the tendency of users to project human traits onto interactive software
  - Trusting into predictions above one's own assessment
- Misuse of AI by bad actors
  - Face recognition used to detect Uyghur population (China), Clearview used to track people's movement and employment status by police officers (even though use was prohibited)
  - Facial analysis used to track people's attention on billboards, change advertisement based on their demographic
- Source of truth during training
  - Data that foundation models are trained on does also contain false statements, ie law professor incorrectly accused of sexual harassment
  - Data in foundation models contains toxicity of the internet; human labor is used to label / annotate toxic text and images ie subcontracting workers in Kenya (OpenAI)



### Infamous AI mistakes



### AI in general

- Automatization at amazon: Experimental hiring tool, developed by a team of five, used artificial intelligence to give job candidates scores ranging from one to five stars
- Tool was trained on all resumes of the last 10 years
- Tool preferably suggested male candidates, penalyzing resumes that contained the word "women's" or graduates from all-female colleges
- Why?

https://www.reuters.com/article/us-amazon-com-jobs-automation-insight/amazon-scraps-secret-ai-recruiting-tool-that-showed-bias-against-women-id USKCN1MK08G/ (2018) Dr. Inga Ulusoy, Scientific Software Center (2025)



#### Automated resume selection at amazon

- The dataset consisted of the resumes of the last 10 years: Predominantly male applicants
- Women are underrepresented in tech:

https://fingfx.thomsonreuters.com/gfx/rngs/AMAZON.COM-JOBS-AUTOMATION/010080Q91F6/index.html

- Thus, the model learnt it was more often correct if it suggested male candidates: Unbalanced representation of the dataset



#### Al in general

- U.S. healthcare system uses commercial algorithms to guide health decisions
- Algorithm (Optum's Impact Pro) to target patients for "high-risk care management" programs
- Identify patients who benefit the most: <u>https://www.science.org/doi/10.1126/science.aax2342</u>
- Model is trained on healthcare spendings to determine the healthcare need
- Algorithm was much more likely to recommend white patients for these programs than black patients, even though the black patients were evidently sicker
- Why?



#### Bias in healthcare need estimation

- Training based on healthcare spendings: But people of color are more likely to have lower incomes making them less likely to access medical care even if they are insured
- Also, they may experience higher barriers to accessing health care (geography, transportation, work/childcare constraints), in addition to direct doctor-patient bias
- Data shows that race is correlated with substantial differences in health-care spendings: This results in a bias of the trained model



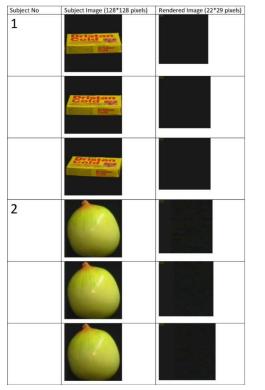
#### AI mistakes in research software

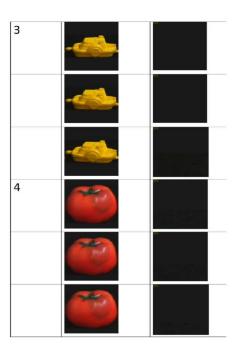
- Applying machine learning methods to COVID-19 radiological imaging for improving the accuracy of diagnosis
- Distinguish patients with COVID-19 from patients without COVID-19 but also bacterial pneumonia
- Tools were trained on public datasets with CT and CXR images
- Predictive tools failed practical tests: <u>https://www.nature.com/articles/s42256-021-00307-0</u>
- Why?



### Learning from the image background

- Image dataset was collected under controlled conditions:
   Does not represent the target distribution of interest
- CNN learnt to distinguish the image background rather than the image content
- Actually quite prevalent problem in computer vision





COIL-100 dataset <u>https://www1.cs.columbia.edu/CAVE/publications/pdfs/Nene\_TR96\_2.pdf</u> https://doi.org/10.1016/j.visinf.2021.10.001



#### AI mistakes in research software

- Predict whether a country is likely to slide into civil war based on GDP, poverty rates, type of government structure, etc.
- Complex models using Random Forests and Adaboost outperform more standard statistical approaches like logistic regression by far
- Missing values in the dataset were constructed using imputation on the complete dataset
- Models proved to be over-optimistic and erroneous https://doi.org/10.1016/j.patter.2023.100804
- Why?



#### Civil war predictions: Data leakage

- Data leakage: The data was imputed for missing values using the whole dataset
- Thus, the training dataset contained information about the test dataset
- This leads to an inflated estimate of the model performance

https://doi.org/10.1016/j.patter.2023.100804 supplemental material



## Classification of failures/errors



#### Data leakage

Spurious relationship between independent variables and target variable

Artifact of collection, sampling, pre-processing

Leads to inflated estimates of model performance

Lack of clean separation training/test

- no test set
- pre-processing on training and test set (over/under sampling, imputation)
- feature selection on entire dataset
- duplicates in dataset



#### Data leakage

#### Model uses features that are not legitimate

for example, use of a certain drug when predicting illness (hypertensive drug, antibiotics)

#### Test set is not drawn from distribution of scientific interest

- temporal leakage (test set must not contain data from before the training set)
- non-independence between training and test samples (same people/units in both sets use block crossvalidation)
- sampling bias in test distribution (spatial bias, age, image settings)



#### Resources

- Good practices in machine learning (Mathieu Bauchy) (<u>https://www.youtube.com/watch?v=WScUQnU-ozQ&t=3213s</u>
- Roadmaps <u>https://roadmap.sh/roadmaps</u>
- Kaggle https://www.kaggle.com/learn
- Hugging Face: <u>https://huggingface.co/learn</u>
- REFORMS checklist https://reforms.cs.princeton.edu/
- More resources <u>https://www.cs.princeton.edu/~arvindn/,</u> <u>https://www.aisnakeoil.com/p/introducing-the-ai-snake-oil-book</u>
- Scikit-learn resources <a href="https://scikit-learn.org/stable/common\_pitfalls.html">https://scikit-learn.org/stable/common\_pitfalls.html</a>
- Testing of non-deterministic software <u>https://bssw.io/blog\_posts/testing-non-deterministic-research-software</u>