





THE FAIR PRINCIPLES

The FAIR Guiding Principles for scientific data management and stewardship

Mark D. Wilkinson, Michel Dumontier, [...] Barend Mons [™]

Scientific Data 3, Article number: 160018 (2016) | Cite this article

194k Accesses | 2450 Citations | 1852 Altmetric | Metrics

A set of principles, to ensure that data are shared in a way that enables and enhances reuse by humans and machines

Findable

F1. (meta)data are assigned a globally unique and eternally persistent identifier.
F2. data are described with rich metadata.

F3. (meta)data are registered or indexed in a searchable resource.

F4. metadata specify the data identifier.

Accessible

A1 (meta)data are retrievable by their identifier using a standardized communications protocol.
A1.1 the protocol is open, free, and universally implementable.
A1.2 the protocol allows for an authentication and authorization procedure, where necessary.
A2 metadata are accessible, even when the data

are no longer available.

Interoperable

I1. (meta)data use a formal, accessible, shared, and broadly applicable language for knowledge representation.
I2. (meta)data use vocabularies that follow FAIR principles.

(meta)data include qualified references to other (meta)data.

Reusable

R1. (meta)data have a plurality of accurate and relevant attributes.

R1.1. (meta)data are released with a clear and accessible data usage license.
R1.2. (meta)data are associated with their

provenance.

R_{1.3}. (meta)data meet domain-relevant community standards.







FAIR FOR NON-DATA OBJECTS: SOME CONTEXT

- FAIR Principles, at a high level, are intended to **apply to all research objects**; both those used in research and those that are research outputs
- Text in principles often includes "(Meta)data ..."
 - Shorthand for "metadata and data ..."
- Principles applied via dataset creators and repositories, collectively responsible for creating, annotating, indexing, preserving, sharing the datasets and their metadata
- What about non-data objects?
 - While they can often be stored as data, they are not just data
- While high level goals (F, A, I, R) are mostly the same, the details and how they are implemented depend on
 - How objects are created and used
 - How/where the objects are stored and shared
 - How/where metadata is stored and indexed
- Work needed to define, then implement, then adopt principles

Slide adapted from the <u>presentation</u> of the <u>RDA FAIR4RS</u> steering group at the International Funders Workshop (Nov 2022), https://zenodo.org/doi/10.5281/zenodo.7350198





FAIR FOR NON-DATA OBJECTS: SOME EFFORTS

Ten simple rules for making training materials FAIR

Leyla Garcia, Bérénice Batut, Melissa L. Burke, Mateusz Kuzak, Fotis Psomopoulos, Ricardo Arcila, Teresa K. Attwood, Niall Beard, Denise Carvalho-Silva, Alexandros C. Dimopoulos, Victoria Dominguez del Angel, Michel Dumontier, Kim T. Gurwitz, [...], Patricia M. Palagi [view all]

Published: May 21, 2020 • https://doi.org/10.1371/journal.pcbi.1007854

Breakout 7 Data Infrastructures - Organisa... The FAIR Agenda WGs Getting started

WG FAIR for Virtual Research Environments: FAIR for VREs - The Path Forward
7:30 AM - 9:00 AM

Room E

DOI: 10.15497/RDA00065

Citation and download: Chue Hong, N. P., Katz, D. S., Barker, M., Lamprecht, A.-L., Martinez, C., Psomopoulos, F. E., Harrow, J., Castro, L. J., Gruenpeter, M., Martinez, P. A., Honeyman, T., et al. (2021). FAIR Principles for Research Software (FAIR4RS Principles). *Research Data Alliance*. DOI: 10.15497/RDA00065

FAIR for AI: An interdisciplinary and international community building perspective

E. A. Huerta , Ben Blaiszik, L. Catherine Brinson, Kristofer E. Bouchard, Daniel Diaz, Caterina Doglioni,
Javier M. Duarte, Murali Emani, Ian Foster, Geoffrey Fox, Philip Harris, Lukas Heinrich, Shantenu Jha, Daniel
S. Katz, Volodymyr Kindratenko, Christine R. Kirkpatrick, Kati Lassila-Perini, Ravi K. Madduri, Mark S.
Neubauer, Fotis E. Psomopoulos, Avik Roy, Oliver Rübel, Zhizhen Zhao & Ruike Zhu

Scientific Data 10, Article number: 487 (2023) Cite this article

January 01 2020

FAIR Computational Workflows

Carole Goble 🗹 👵 , Sarah Cohen-Boulakia, Stian Soiland-Reyes, Daniel Garijo, Yolanda Gil, Michael R. Crusoe, Kristian Peters, Daniel Schober

> Author and Article Information

Data Intelligence (2020) 2 (1-2): 108-121.

https://doi.org/10.1162/dint_a_00033

Introducing the FAIR Principles for research software

Michelle Barker [™], Neil P. Chue Hong, Daniel S. Katz, Anna-Lena Lamprecht, Carlos Martinez-Ortiz, Fotis

Psomopoulos, Jennifer Harrow, Leyla Jael Castro, Morane Gruenpeter, Paula Andrea Martinez & Tom

Honeyman

Scientific Data 9, Article number: 622 (2022) Cite this article







SOFTWARE IS NOT JUST ANOTHER TYPE OF DATA

- FAIR Principles, are intended to be applied to all digital objects (Wilkinson et al. 2016)
- Efforts to adapt and adopt the FAIR principles to research software (RDA FAIR4RS)

Recommendation $n^{\circ}5$:

Recognise that FAIR guidelines will require translation for other digital objects and support such efforts.

2020: 'Six Recommendations for Implementation of FAIR Practice'

(FAIR Practice Task Force EOSC, 2020)







WHAT IS RESEARCH SOFTWARE?

Research software - Source code files, algorithms, scripts, computational workflows and executables that were created in either of two categories:

- A. Within a research project as a by-product to do the research, or
- B. Through intentional development of a software product for general use in research by one or more projects.

Gruepenter et al. 2021 Defining Research Software: a controversial discussion (Version 1). Zenodo. https://doi.org/10.5281/zenodo.5504016

Not all software used in research is research software





FAIR4RS PRINCIPLES







- **Findable**: Software, and its associated metadata, is easy for both humans and machines to find.
- Accessible: Software, and its metadata, is retrievable via standardized protocols.
- Interoperable: Software interoperates with other software by exchanging data and/or metadata, and/or through interaction via application programming interfaces (APIs), described through standards.
- Reusable: Software is both usable (can be executed) and reusable (can be understood, modified, built upon, or incorporated into other software).

(key differences from FAIR data principles in italics)

Output of <u>the FAIR principles for research software</u> (FAIR4S) - joint Research Software Alliance (**ReSA**), Research Data Alliance (**RDA**), **FORCE11** Working Group/Task force

Slide adapted from the <u>presentation</u> of the <u>RDA FAIR4RS</u> steering group at the International Funders Workshop (Nov 2022), https://zenodo.org/doi/10.5281/zenodo.7350198







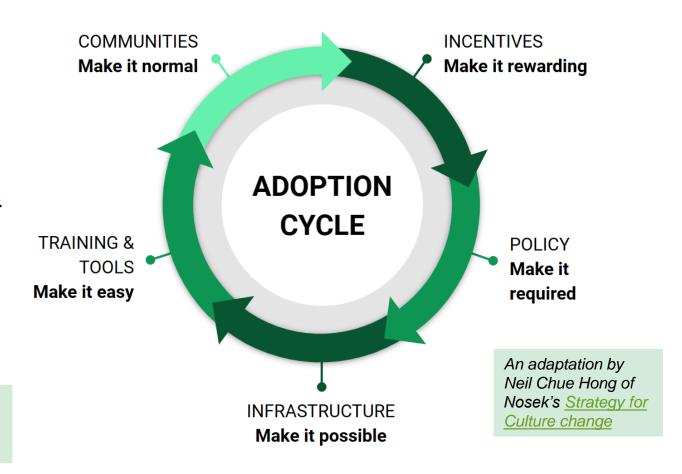
WHO IS RESPONSIBLE FOR FAIR SOFTWARE?

Who is expected to apply FAIR?

- The application of the FAIR4RS

 Principles is the responsibility of the owners (who are often the creators) of the software, not the users.
- The FAIR4RS Principles are also relevant to, and require support from, the larger ecosystem and various stakeholders that support research software (e.g., repositories and registries).

Slide adapted from the <u>presentation</u> of the <u>RDA FAIR4RS</u> steering group at the International Funders Workshop (Nov 2022), https://zenodo.org/doi/10.5281/zenodo.7350198

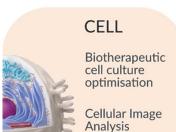




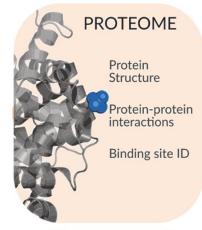


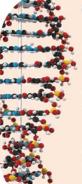
MOVING BEYOND SOFTWARE?

Machine learning is being used across many biological areas



Cell type Annotation





GENOME

ID gene coding regions

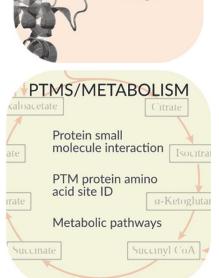
Pharmacogenomics

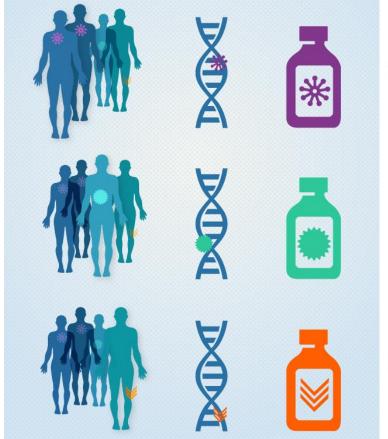
CRISPR Target sequence ID

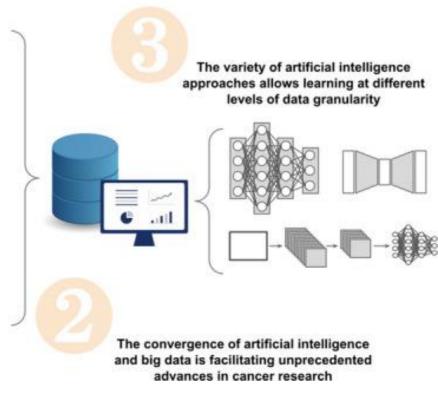
Methylation side ID

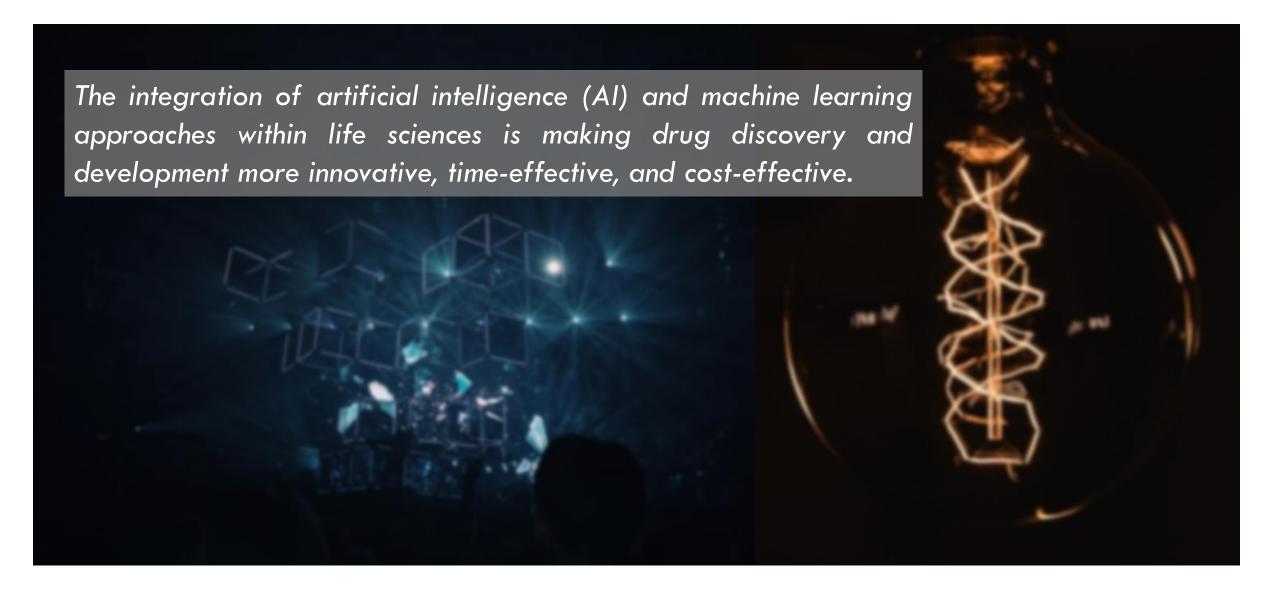
Gene expression (microarrays)

RNA structure











SCIENCE FORUM

NEW SET OF CHALLENGES

COMPUTER SCIENCE

By Matthew Hutson

Artificial intelligence faces reproducibility crisis

Unpublished code and sensitivity to make many claims hard to verify

machine intelligence

Ten common statistical mistakes to watch out for when writing or reviewing a manuscript

Abstract Inspired by broader efforts to make the conclusions of scientific research more robust, we have compiled a list of some of the most common statistical mistakes that appear in the scientific literature. The mistakes have their origins in ineffective experimental designs, inappropriate analyses and/or flawed reasoning. We provide advice on how authors, reviewers and readers can identify and resolve these mistakes and, we hope, avoid them in the future.

TAMAR R MAKIN* AND JEAN-JACQUES ORBAN DE XIVRY

Makin and Orban de Xivry. eLife 2019;8:e48175. DOI: https://doi.org/10.7554/eLife.48175

OXFORD

Briefings in Bioinformatics, 17(5), 2016, 831-840

Advance Access Publication Date: 26 September 2015

Correct machine learning on protein sequences: a peer-reviewing perspective

Ian Walsh, Gianluca Pollastri and Silvio C. E. Tosatto

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Training data Training set

Generalization check

Self-consistency check

SCIENCE sciencemag.org 16 F Validity of machine learning in biology and medicine increased through collaborations across fields of expertise

Code break

summary of an algorithm.

In a survey of 400 artificial intelligence papers pre-

sented at major conferences, just 6% included code for the papers' algorithms. Some 30% included test data, whereas 54% included pseudocode, a limited

Test data Pseudocode

Maria Littmann ^{1,27*}, Katharina Selig ^{1,27*}, Liel Cohen-Lavi^{3,4}, Yotam Frank⁵, Peter Hönigschmid ^{1,6} Evans Kataka⁶, Anja Mösch^{©6}, Kun Qian^{©7,8}, Avihai Ron^{©9,10}, Sebastian Schmid^{©11}, Adam Sorbie^{©12}, Liran Szlak¹³, Ayana Dagan-Wiener¹⁴, Nir Ben-Tal¹⁵, Masha Y. Niv^{14,16}, Daniel Razansky¹⁵, Daniel Razansky¹⁵, Masha Y. Niv^{14,16}, Masha Y. Niv¹ Björn W. Schuller 21, Donna Ankerst 2, Tomer Hertz 3,22,23 and Burkhard Rost 1,24,25,26

Setting the standards for machine learning in biology

David T. Jones 1,2

Machine learning is a branch of artificial intelligence (AI) involving computer programs that are able to improve their own performance through experience (training). The diverse applications of new 'deep learning' approaches with neural networks are now expanding into the field of biology. But these applications to biological data require more scrutiny and caution to increase the standards of publishing and allow the AI revolution in biology to take off.

https://doi.org/10.1038 s41580-019-0176-5

NATURE REVIEWS | MOLECULAR CELL BIOLOGY

30 November 2023





FAIR IN MACHINE LEARNING (MODELS)?

- What does FAIR apply to?
 - Are they data?
 - E.g., a set of parameters and options for a particular framework
 - Are they software?
 - E.g., an executable object that takes input and provides output
 - Are they something else?
- How does FAIR apply?
 - Searched and shared via repositories?
 - Searched and shared via executable platforms?
 - Searched and shared via something else? (e.g., DLHub, OpenML, HuggingFace...)
 - Models and training data are linked should they be shared together?





Slide adapted from various presentations of the <u>RDA FAIR4ML</u> interest group during Plenary events









NEED FOR COMMUNITY-LED STANDARDS (1/2)











DOME adopted as part of the submission system for GigaScience

(see example here: http://gigadb.org/dataset/102404)

Online registry of annotated papers: https://registry.dome-ml.org



the DOME Recommendations

DOME: recommendations for supervised machine learning validation in biology

Ian Walsh, Dmytro Fishman, Dario Garcia-Gasulla, Tiina Titma, Gianluca Pollastri, ELIXIR Machine Learning Focus Group, Jennifer Harrow ≅, Fotis E. Psomopoulos ≅ & Silvio C. E. Tosatto ≅

Nature Methods (2021) | Cite this article

4927 Accesses | 73 Altmetric | Metrics



DOME Registry

A database of annotations for published papers describing machine learning methods in biology.

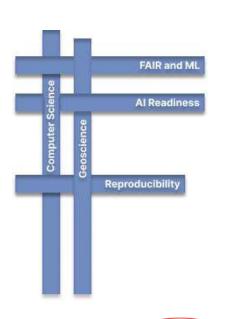






NEED FOR COMMUNITY-LED STANDARDS (2/2)

FARR: FAIR in ML, AI Readiness, & Reproducibility Research Coordination Network

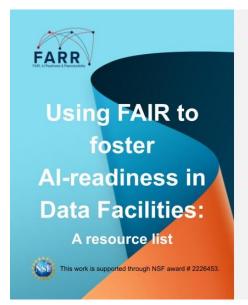




EDI

Ways to Get Involved

- Input on community needs, gaps & roadmap
- Suggest use cases and let us promote your project's use of Al and FARR-related practices
- Let us feature you in a science story



What is FAIR?

- A refresher on FAIR: More than an acronym, it stands for 15 principles for making research objects more Findable, Accessible, Interoperable, Reusable https://www.qo-fair.org/fair-principles/
- Suggestions on how to implement FAIR: https://bit.ly/implementFAIR

Data repositories supporting AI with FAIR practices

- The geosciences: https://www.hydroshare.org/
- High energy physics: https://bit.lv/Al-readvHEP
- · Materials science: https://bit.ly/MLinMS

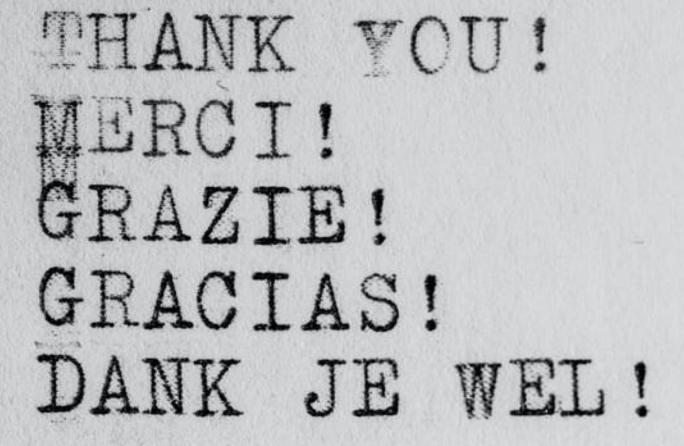
Contact:

https://www.farr-rcn.org/
community@farr-rcn.org

This work is supported through the NSF award #2226453.











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