

ljgarcia@zbmed.de





The FAIR Guiding Principles

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

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

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.

13. (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.

R1.3. (meta)data meet domain-relevant community standards.

Let's go FAIR



Why should I (researcher) go FAIR?

- I read about FAIR and it's cool, I want my data to be FAIR
- I got reviews on my submission and reviewers say my data is not FAIR
- My PI is telling me I have to make my data FAIR
- The Data Steward tells me that according to the RDM plan, my data is not ready because is not FAIR

Why should we (researchers) go FAIR?

- FAIR helps us reflect on our data (in combination with RDM plans)
- ► FAIR provides some guidelines for data sharing (so we can reuse other's data less work, more collaboration)
- FAIR provides some guidelines for data to interoperate (so we can connect our data to other data

 good approach to more complex problems)
- FAIR is a step forward to reproducibility (but on its own does not solve the crisis)

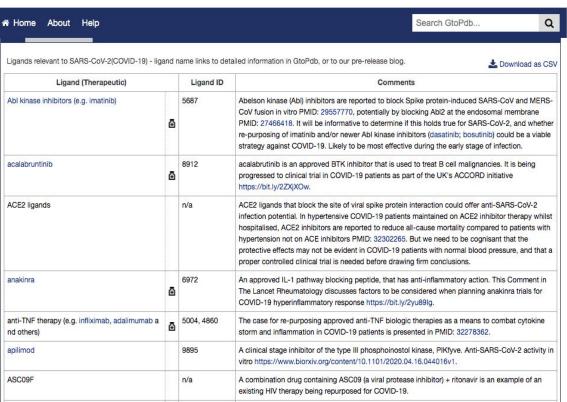
FAIR metadata for humans and machines alike





IUPHAR/BPS Guide to PHARMACOLOGY

Search Database



```
<!--- BioSchemas ---->
<script type="application/ld+json">[
 "@context": "https://schema.org",
 "@type": "Protein",
 "@id": "https://www.guidetopharmacology.org/GRAC/ObjectDisplayForward?objectId=1614#",
 "http://purl.org/dc/terms/conformsTo": "https://bioschemas.org/profiles/Protein/0.9-DRAFT-2019 08 20/",
 "identifier": "1614",
 "name": "ACE2",
 "associatedDisease": "COVID-19".
 "description": "Receptor on host cells that is exploited by some betacoronaviruses for viral entry. Engaged by
 SARS-CoV-2 spike protein as the first step towards infection of host cells",
 "url": "https://www.guidetopharmacology.org/GRAC/ObjectDisplayForward?objectId=1614"
 "@context": "https://schema.org",
 "@type": "MolecularEntity",
 "@id": "https://www.guidetopharmacology.org/GRAC/LigandDisplayForward?ligandId=8912#",
 "http://purl.org/dc/terms/conformsTo": "https://bioschemas.org/profiles/MolecularEntity/0.4-DRAFT-2019_11_11/",
 "identifier": "8912",
 "name": "acalabruntinib",
 "url": "https://www.guidetopharmacology.org/GRAC/LigandDisplayForward?ligandId=8912"
 "associatedDisease": "COVID-19",
 "description": "acalabrutinib is an approved BTK inhibitor that is
 progressed to clinical trial in COVID-19 patients as par
```

Images taken from a slide by Alasdair Gray from a presentation given at LEIXIR Data and Tools Workshop, November 2021

Ok, let's go FAIR but how?





Start with explicit and concrete steps coming directly from the FAIR Guiding Principles

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



R1.2. (meta)data are associated with their provenance.





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



F1. (meta)data are assigned a globally unique and eternally persistent identifier.







You are not alone, link to others





Focus on the linked aspect

I3. (meta)data include qualified references to other (meta)data.



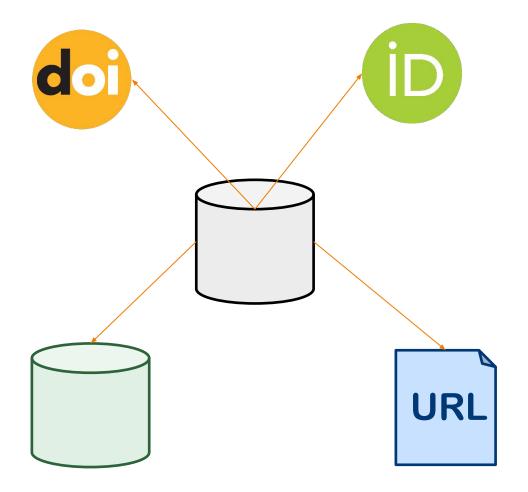


Image taken from http://sites.linkeddata.center/help/devop/training/introduction-to-linked-data

Rich, plurality, community



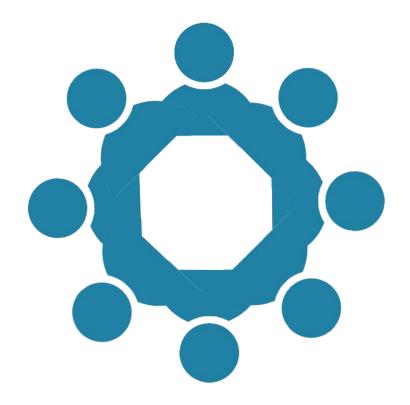


Start with (good enough) metadata

F2. data are described with rich **metadata**.

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

R1.3. (meta) data meet domain-relevant community standards.



What about the rest of the principles?

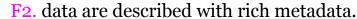


Findable



F1. (meta)data are assigned a globally unique and eternally persistent identifier.







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



F4. metadata specify the data identifier.

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.



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









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.

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.



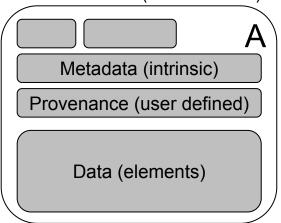
R1.2. (meta)data are associated with their provenance.



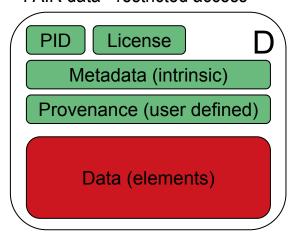
R1.3. (meta)data meet domain-relevant community standards.

Note: diagram has been modified, in particular, it adds a explicit mention to the license

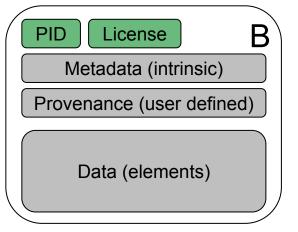
Re-useless data (common case)



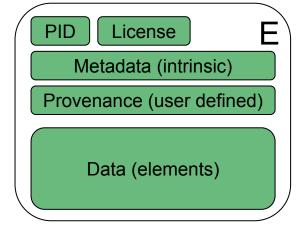
FAIR data - restricted access



Findable and reusable



FAIR data - open access



Authors: Mons, Barend^{a; b; c; * | Neylon, Cameron^d | Velterop, Jan^e | Dumontier, Michel^f | da Silva Santos,} Luiz Olavo Boninob; g | Wilkinson, Mark D.h

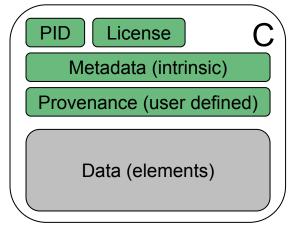
Keywords: FAIR Data, Open Science, interoperability, data integration, standards

Article type: Research Article

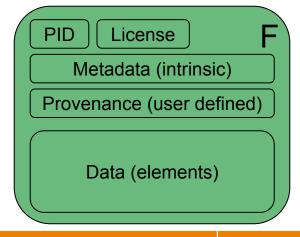
Journal: Information Services & Use, vol. 37, no. 1, pp. 49-56, 2017

Published: 7 March 2017

FAIR metadata



FAIR data - open access and functionally linked



I think my data is FAIR, how can I know for sure?



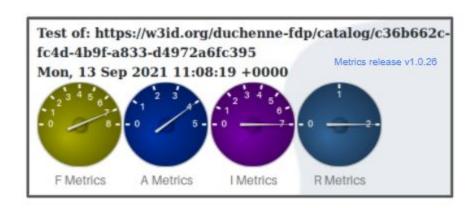
FAIRassist.org

Resource	Execution Type
5 Star Data Rating Tool	Manual - questionnaire
<u>AutoFAIR</u>	Semi-automated
<u>Data Stewardship Wizard</u>	Predictive; based on a manually filled questionnaire
<u>F-UJI</u>	Automated
FAIR Data Self-Assessment Tool	Manual - questionnaire
FAIR Evaluator	Automated
FAIR enough?	Manual - checklist
<u>FAIR-Aware (BETA)</u>	Manual - questionnaire
<u>FAIR-Checker</u>	Automated
<u>FAIRdat</u>	Manual - questionnaire
FAIRness self-assessment grids	Manual - checklist
<u>FAIRshake</u>	Manual - questionnaire, Semi-manual
GARDIAN FAIR Metrics	Manual - checklist
RDA Maturity Model	Manual - checklist

How different can they be?

Comparison of The Evaluator with F-UJI, on the same URI

(a Catalog record in the Duchenne Muscular Dystrophy FAIR Data Point)



20/22 Tests Pass



2/24 Tests Pass

Image corresponds to a slide by Mark D. Wilkinson from a presentation given at FAIReScience Workshop @ IEEE eScience Conference 2021

My data is FAIR, is that FAIR enough?



January 01 2020

FAIR Computational Workflows 3

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



FAIR principles for Machine Learning models

Daniel S. Katz, University of Illinois Urbana-Champaign, d.katz@ieee.org, USA Tom Pollard, MIT Institute for Medical Engineering and Science, tpollard@mit.edu, USA Fotis Psomopoulos, Institute of Applied Biosciences, Centre for Research and Technology Hellas, fpsom@certh.gr, Greece

Eliu Huerta, University of Illinois Urbana-Champaign, elihu@illinois.edu, USA Chris Erdmann, University of North Carolina at Chapel Hill, Renaissance Computing Institute (RENCI), erdmannc@renci.org, USA

Ben Blaiszik, University of Chicago and Argonne National Laboratory, blaiszik@uchicago.edu, USA

- Developed in the context of scientific data management and stewardship in 2014 [1]; turned into specific principles in 2016 [2].
- Generalized in concept to apply to both data and other digital scholarly

in practice, what works for data does not directly work for all other digita

E.g., given differences between data and software, fundamental Interoperability principle cannot have the same meaning

Previous [3] and ongoing [4] work show many FAIR guiding FAIR principles need to either be re-written or reinterpreted for software

- · Machine Learning (ML) models have characteristics of both data and
- ✓ ML models are <u>trained on data</u>, and can be <u>represented by data</u>, but they are not just data
 - They are usually the key component of a software solution (for
- prediction, evaluation, etc.) ✓ May also include the pre- and post-processing logic needed to use
- It's difficult to share and exchange models effectively, even with the
- emergence of new services such as DLHub.org and OpenML.org
- This is nartly due to the fact that there is no established standard for FAIR. ML models (though there is some guidance in particular areas [5] [6])

Our proposal



- We need to investigate how the FAIR principles can be interpreted for ML models
 - ☐ This requires a study of relevant characteristics of data, software, and ML models ☐ Align with relevant community efforts (Pistoia Alliance, ELIXIR, FAIR4HEP)
- ☐ End goal; have a consensus for the principles, move on to adoption
- Short-term goal: Lay the groundwork for a BoF at RDA P17 that might lead to an IG or WG



3] Lamprecht, A-L. et al. Towards FAIR Principles for Research Software. Data Science, 3(1):37-59, 2020. https://doi.org/10.3233/DS-190026 [4] RDA, FORCE11, ReSA FAIR 4 Research Software (FAIR4RS) WG. https://www.rd-alliance.org/groups/fair-4-research-software-fair4rs-wg

[5] The Machine Learning Reproducibility Checklist, v2.0, Apr. 7 2020. https://www.cs.mcgill.ca/~ipineau/ReproducibilityChecklist.v2.0.p [6] Ian Walsh et al. DOME: Recommendations for supervised machine learning validation in biology, arXiv 2020, https://arxiv.org/pdf/20

Ten simple rules for making training materials FAIR

Levla 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



FAIR Principles for Research Software (FAIR4RS Principles)

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

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

WG FAIR for Virtual Research Environments: FAIR for VREs - The Path Forward

7:30 AM - 9:00 AM

Room E

FAIR for research digital objects



Recommendation n°5:

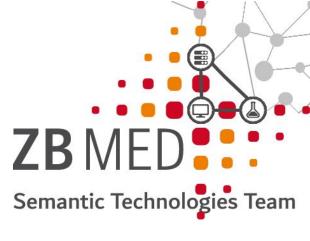
Recognise that FAIR guidelines will require translation for other digital objects and support such efforts.

2020: 'Six Recommendations for Implementation of FAIR Practice'

(FAIR Practice Task Force EOSC, 2020)

Thanks! Danke!

www.zbmed.de



Semantic Retrieval Team – Knowledge Management Group Dr. Leyla Jael Castro

ljgarcia@zbmed.de