

Four use cases of FAIR maturity indicators in the life sciences

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Abstract

Data reuse is crucial to enhance scientific progress and maximize return on science investments. Given the incremented availability, manual and automatic retrieval of data for new research questions can be challenging. Among the guidelines created to enhance data retrieval, the FAIR (findable, accessible, interoperable, reusable) principles are increasingly adopted at an institutional and funding level. Metrics to assess FAIRness of data repositories are under study and contributions are highly encouraged. In this work, we propose four real use-cases of researchers retrieving data from four different repositories (eNanoMapper, ChEBI, Gene Expression Omnibus, and Array Express) to answer their research questions. For each use case, we harvested data and metadata via application program interface (API) and we calculated FAIR metrics assigning “automatic pass”, “manual pass” and “not passed” scores. We found [...]. To conclude [...]

Introduction

Data sharing and data reuse are two complementary aspects of modern research activity that are growing substantially in a short time. Researchers share their data for a sense of community, to demonstrate data quality, and to enhance quality and reproducibility of research work [cit]. In addition, data sharing is supported by the emerging citation system for datasets [1], scientific journals requirements [3], and funding agencies that want to maximize their return on science investments [5]. At the same time, researchers are eager to reuse available data to integrate information coming from different fields [cit], to answer interdisciplinary research questions [4], and for low availability of funding [cit]. Although attitudes towards data sharing and reuse are increasingly favorable [12], data discovery and re-use remain difficult in practice [11]. Studies show that 40% of qualitative data sets were never downloaded, and about 25% of data is used just 1-10 times. In addition, data availability decreases 17% per year [10]

However, data harvesting as a discipline is relatively new. Although information retrieval (IR) has been extensively studied for over 60 years (Sanderson and Croft, 2012), data retrieval is a nascent field. Info from Sompel: The Open Archives Protocol for Metadata Harvesting (OAI-PMH) [Lagoze et al. 2002] has been widely adopted as an approach to allow harvesting of metadata. [...]. harvestable in an interoperable manner (interoperability is about harvesting, not about data format, but about metadata formats) Dunning: The FAIR guidelines are a fairly recent invention, published in 2014. Many of the data repositories have histories longer than that, and draw on discipline-based practices that have well established protocols for how data should be shared.s

In the life sciences (but not exclusively), the FAIR (findable, accessible, interoperable, and reusable) guidelines have gained remarkable popularity. They were established by the FORCE 11 group in 2016, and in a few years they have been adopted by funding agencies and at a political level (G7, the European DMN, NIH) [vd. paper by Mons].. Critics about the practical implementation of the FAIR principle: Dunning (p.187) The FAIR principles are one of the element for data management. focus on data harvest as they do not provide guidelines for other aspects of data management, such as openness, quality, standardization, ethics, and security [Mons]. The authors define these principles as aspirational [cit] and general so that they can be adaptable to specific requirements of different communities.

Paragraph about Metrics The main point here is about metrics. For this reason there are attempts to measure the possibility of reusing the data. One way is to use linked data, and this is measured using the 5 star system and Yummydata + King + Others tend to have already established internal and international guidelines (e.g. s) The promoters of the FAIR principle have recently written a guideline for metrics. They are in this paper, a website and this github repository. They do not provide any concrete example on how to implement them. We tried to implement them in the context of nanomaterial. Safety of nanomaterials is of particular interest because they can be very dangerous. The European community strongly supports data integration to merge information, and the European community tells these people to apply the FAIR principle for their databases. Therefore we tested some databases to see if they are compliant to the FAIR metrics. Bla bla on metrics with citation to their paper. Other works: Dunning et al analyzed 37 repositories (non solo repositories, but a bit of everything - see excel, CO-Frequencies and Proportions). Without explicitly mention the metrics (metrics paper out afterwards?), they tested the FAIR principle on repositories. We are going to do the same for nanosafety online databases Current tools to assess FAIRness of data are developed by the European xxx () and by the NIH group (<https://www.fairshake.cloud/>), which allow Tables

In this paper, we focus on calculating FAIR metrics databases for the scientific community working on nanomaterials. We specifically analyze online databases, not repositories We created four use-cases

where a researchers queries a database to find a dataset that can support the answer to her/his research question. We queried the database using API and we calculated the FAIR metrics using python in Jupyter notebook

Table 1: Data harvesting evaluators in the literature.

Authors	Metrics		Platform	Automatic assessment			Manual assessment	Data / code repository
	FAIR	others		code / language	metadata format	protocol / library		
Data Archiving and Networked Services	x	-	FAIR enough? and FAIR data assessment tool	-	-	-	x	-
Australian Research Data Commons	x	-	FAIR self-assessment tool	-	-	-	x	-
Commonwealth Scientific and Industrial Research Organization	x	-	5 star data rating tool	-	-	-	x	
Shultes et al. (GO FAIR consortium)	x	-	FAIR ImplementationMatrix	-	-	-	x	Open Science Framework
Dunning et al.	x	-	-	-	-	-	x	Data repository
Wilkinsons et al.	x	-	FAIR evaluation services	Ruby on Rails	JSON, Microformat, JSSON-LD, RDFa	nanopublications	-	GitHub
Clark et al.	x	-	FAIRshake	Django and python	RDF	Extruct	?	GitHub
Weber et al.	x	-	-	python	DataCite	OAI-OMH	-	GitLab
Yamamoto et al.	-	Umaka Score	YummyData	Ruby on Rails	RDF, JSON, JSON-LD	SPARQL	-	GitHub
Our approach	x	-	-	Jupyter notebook and python	XML, JSON	request?	x(partially)	GitHub

Materials and methods

Use cases in the life sciences

We calculated FAIR maturity indicators for four use-cases. The first and the second use-cases are real research questions asked by two researchers in our department, whereas the third and fourth use-cases are plausible research questions in the life sciences. For each use-case, researchers harvested data from a different database. The four use cases are:

- What are the differentially expressed genes between normal subjects and subjects with Parkinson's diseases in the brain frontal lobe? The researcher looked for a dataset in Array Express, a repository containing xxx [ref]
- What is the function of mutation of WDR45 protein in the brain? The researcher looked for a dataset in Gene Expression Omnibus (GEO), a database that contains xxx [ref][Gwen-Freddie]
- What are the concentrations of titanium dioxide that kill cells? The database is eNanoMapper, a repository for xxx
- What are the biological roles of caffeine? The repository is ChEBI (Chemical Entities of Biological Interest) a repository created by the embo... [ref]

Array Express are repositories containing data in form of files that researchers download to compute their analysis and answer their questions. eNanoMapper and ChEBI are repositories where data are not downloadable in the form of files and the extracted information are the data themselves.

What is data and what is metadata?

The FAIR guidelines recursively use the terminology *data*, *metadata*, and *(meta)data* (Table [???]). For our computational implementation, we needed precise definitions of these terms. Accordingly to the Merriam-Webster online dictionary, *data* are "information in digital form that can be transmitted or processed" [2] whereas *metadata* are "data that provides information about other data" [3]. Following these definitions, we considered the answer to the research question as *data*, and the extra information provided in the database about *data* as *metadata*. In addition, we divided *metadata* (M) in subcategories according to the requirements of the FAIR guidelines (indicated with their enumeration):

- M(F2): Information that allows researchers to find the dataset s/he looks for. It coincides with the keywords used in the search;
- M(F3): Data identifier in the repository;
- M(I3): Reference to other metadata;
- M(R1): Further information about data content, other than the search keywords;
- M(R1.1): Data license;
- M(R1.2): Data provenance: author name, publication title, and one author's email address

Metadata corresponding to guidelines F2 and R1 change with the research question, while metadata corresponding to F3, I3, R1.1, R1.2 are independent from the research question. We do not define M(1.3) as it requires community consensus. In all cases, we assumed that *data* and *metadata* were hosted in the same repository.

Calculating FAIR maturity indicators

As the FAIR principles stress on the importance of *data* and *metadata* being "machine-readable", we collected information about datasets and repositories via application programming interface (API)

wherever possible.

We asked the researchers to show us how they searched for their dataset using repositories user interfaces in the browser, and we reproduced their search via API. For the two plausible we did the same. We investigated using three methods:

- API of the repository: We queried the repository to get the dataset. Once found the dataset of interest, we retrieved presence of keywords in metadata (F2), presence of identifier in metadata (F3), metadata are retrievable using a standard communication protocol (A1), reference to other metadata (I3), plurality of metadata (R1), and provenance (R1.2).
- API of re3data.org: We looked for information about persistent identifier (F1 - manual?), license (R1.1)
- API of [Google Dataset Search](https://www.google.com/datasetsearch/): To see if the dataset is indexed in searchable research (F4)
- Email to repository curators: We asked for information that we could not retrieve via API or to confirm information we found in the online documentation, i.e. policy describing metadata accessibility when data are no longer available (A2), and structure of metadata representation (I1).

Finally, we did not calculate maturity indicators concerning use of FAIR vocabularies (I2), and meet domain-relevant community standards (R1.3).

To calculate the maturity indicators, we built on the already available implementations. As a searchable resource (F4), we used re3data.org, a registry containing metadata of more than 2000 data repositories from various disciplines. re3data.org also provides information about licenses used if the repository provides unique and persistent identifiers (F1).

We assigned 1 when the principle was completely satisfied, 0 when it failed and 0.5 when we entered manual information. We assigned decimal number for the principle F2, where we divided the number of found keywords in the metadata over the number of keywords used by the researcher.

See table [??]

Table 2: FAIR maturity indicators. In our approach, manual procedures are labelled with *.
Acronyms: GUID = Globally Unique Identifier.

FAIR principle	Wilkinson	Dunning	Weber	Our approach
F1: (meta)data are assigned a globally unique and persistent identifier	The GUID matches a scheme that is globally unique and persistent in FAIRsharing	Persistent identifier is DOI or similar	Pass	Information from re3data *
F2: data are described with rich metadata (defined by R1 below)	Metadata contains "structured" elements (micrograph, JSON) or linked data (JSON-LD, RDFa)	Title, creator, date, contributors, keywords, temporal and spatial coverage	Q_{geo} , Q_{chrono}	Search keywords are in metadata
F3: metadata clearly and explicitly include the identifier of the data it describes	Metadata contains both its own GUID and the data GUID	DOI of data is in metadata	Pass	Metadata contains dataset ID found with search
F4: (meta)data are registered or indexed in a searchable resource	The digital resource can be found using web-based search engines	Dataset title searched in google.com or duckduckgo.com	Pass	Dataset title searched in Google Dataset Search
A.1 (meta)data are retrievable by their identifier using a standardized communications protocol	N/A	If HTTP, pass	Q_{ret}	HTTP request returns 200
A1.1 the protocol is open, free, and universally implementable	The resolution protocol is universally implementable with an open protocol	If HTTP, pass	Q_{ret}	Included in A.1
A1.2 the protocol allows for an authentication and authorization procedure, where necessary	The resolution protocol supports authentication and authorization for access to restricted content	If HTTP, pass	Q_{ret}	Included in A.1

FAIR principle	Wilkinson	Dunning	Weber	Our approach
A2. metadata are accessible, even when the data are no longer available	There is a policy for metadata	Repository has a clear policy statement	N/A	Manual check or on re3data
I1. (meta)data use a formal, accessible, shared, and broadly applicable language for knowledge representation	If hash-style metadata (e.g. JSON) or Linked Data are found, pass	Metadata is structured (e.g. Dublin Core)	Pass	Metadata are in a defined structure *
I2. (meta)data use vocabularies that follow FAIR principles	(meta)data uses vocabularies that are, themselves, FAIR	N/A	N/A	N/A
I3. (meta)data include qualified references to other (meta)data	Metadata contain links that are not from the same source (domain/host)	Links to publications and terms definitions	N/A	Metadata includes reference to other dataset IDs
R1. meta(data) are richly described with a plurality of accurate and relevant attributes	N/A	Metadata provide information on how to reuse a dataset	Q _{geo} , Q _{chrono}	TBD
R1.1. (meta)data are released with a clear and accessible data usage license	Metadata contains a pointer to the data license	Metadata license is present	Q _{lic}	Check on re3data.org
R1.2. (meta)data are associated with detailed provenance	N/A	Documentation on how data was created	N/A	Author name, email address, publication
R1.3. (meta)data meet domain-relevant community standards	N/A	N/A	N/A	N/A

Results

Table 1

- What passed/failed
- Comparison of outcomes in Figure 1. Data are still work in progress and are not final

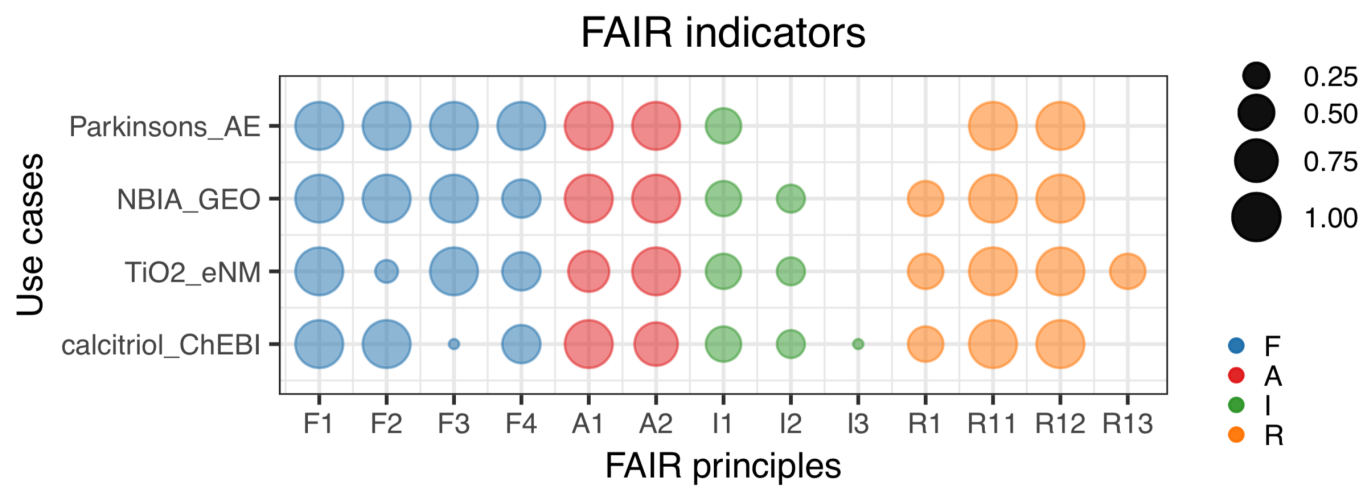


Figure 1: Outcome comparison

Table 3: Use cases.

Use case	Parkinson's disease	NBIA	Titanium Dioxide	Caffeine
Repository	Array Express (https://www.ebi.ac.uk/arrayexpress/)	(Gene Expression Omnibus(https://www.ncbi.nlm.nih.gov/geo/) [https://www.ncbi.nlm.nih.gov/geo/])	eNanoMapper	ChEBI
Search output on browser	experiments/E-MTAB-1194/	/query/acc.cgi?acc=GSE70433		
API				
API type	REST	REST		
API documentation	help/programmatic_access.html	/info/geo_paccess.html	http://enanomapper.github.io/API/	https://www.ebi.ac.uk/chebi/webServices.do
Output format	XML	XML	RDF	XML (?)
FAIR principles				
F1 (Persistent identifier)	No	No		
F2 (Findable metadata)	Parkinson's disease, normal, homo sapiens, transcription profiling by array, raw data, frontal lobe, male, female	nbia, homo sapiens, expression profiling by array	Publication and protocol	
F3 (Unique identifier)	219251	200070433		
F4 (In Google Dataset search)	?	?		
A1 (Communication protocol)	HTTP	HTTP		
A2 (Metadata always accessible)	?	?		
I1 (Language representation)	help/programmatic_access.html#Format_XML_results			

Use case	Parkinson's disease	NBIA	Titanium Dioxide	Caffeine
I2 (FAIR vocabularies)	Not evaluated	Not evaluated		
I3 (Reference to other metadata)	?	?		
R1 (Reusable metadata)	?	?		
R1.1 (License)	help/data_availability.html	http://www.ncbi.nlm.nih.gov/geo/info/disclaimer.html		
R1.2 (Provenance)	Garcia-Esparcia P et al.			
R1.3 (Community standards)	Not evaluated			

Discussion

Main points to discuss (still in random order):

- We had to adapt the code based on API type and response schema. Our implementation requires specific knowledge of the database structure and thus it is difficult to directly generalize it to various databases
- Comparison to Wilkinson: Our metrics do not start with metadata GUID (general user identifier) (see gen2) but with the researcher's question. Using GUID implies that the researcher has already found the dataset of interest
- Compare to Weber and Dunning
- Database APIs do not allow to retrieve all the information that the user interface allows (example 1: ChEBI does not allow to retrieve information about reactions; example 2: Array Express has some metadata in tables that must be downloaded locally before being queried)
- We considered use cases where all queries provided one final dataset. In real practice, researchers often need to compare subset of retrieved datasets manually because there are not enough information to discriminate them computationally (the information is present, but not machine-readable)
- Comments on the findings
- We chose to use Jupyter notebooks for reproducibility of our results. However, databases change but they do not provide versions. Therefore, we can just declare the time stamps when our query was done. In addition, Jupyter notebooks are both machine and human readable, and easier to export to other domains that do not use specifically programming languages designed for the web
- We chose to plot our results instead of providing a final score to avoid negative connotations (see FAIR metrics vs. maturity indicators). However, we wanted to be able to compare our results, so we used balloon plots, usually used for categorical data visualization and comparison. (FAIR shake uses visualizations too but they are not comparable)

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References

1. Credit data generators for data reuse

Heather H. Pierce, Anurupa Dev, Emily Statham, Barbara E. Bierer

Nature (2019-06) <https://doi.org/gf3j9t>

DOI: [10.1038/d41586-019-01715-4](https://doi.org/10.1038/d41586-019-01715-4) · PMID: [31164773](https://pubmed.ncbi.nlm.nih.gov/31164773/)

2. Definition of DATA <https://www.merriam-webster.com/dictionary/data>

3. Definition of METADATA <https://www.merriam-webster.com/dictionary/metadata>

4. manubot/rootstock GitHub repository

Daniel Himmelstein

GitHub (2019) <https://github.com/manubot/rootstock>