# WGISS-54

FAIR Dataset Quality Information Guidelines

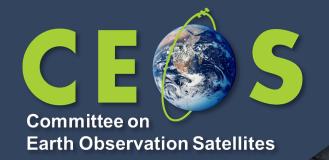


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WGISS-54

Tokyo, Japan (JAXA)

3-7 October 2022



#### **Executive Summary**



- FAIR DQI community guidelines provide specific advice on ensuring quality metadata compliant with the FAIR principles for the dataset
- FAIR DQI community guidelines are a living document developed by international community for international community
- Use-Cases on challenges with quality information are wanted!
- FAIR DQI guidelines support Priority 3: Support to CEOS Cal/Val Initiatives to increase CEOS Agency Cal/Val Collaboration

#### Why FAIR Quality Information?



- Increasingly the reuse of a dataset, particularly where multiple datasets are being merged, requires knowledge of the "quality" of the datasets to be merged.
- Particularly where datasets are repurposed for use cases beyond what the original creator intended: "quality" information becomes critical.
- With the rise of Artificial Intelligence (AI) and Machine Learning (ML), a new interpretation of FAIR is that it stands for "Fully AI Ready": knowing the "quality" of data to be used is essential to avoid erroneous conclusions

Cloudy, increasingly FAIR; revisiting the FAIR Data guiding principles for the European Open Science Cloud

June 22, 2017

https://content.iospress.com/articles/information-services-and-use/isu824

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#### Abstract

The FAIR Data Principles pr Interoperable, and Reusable behaviours that researchers principles should manifest i Principles has spread, so ha spread of interpretation, see the Principles, to clarify bot https://danielskatzblog.wordpress.com/2017/06/22/fair-is-not-fair-enough/

#### FAIR is not fair enough

The <u>FAIR data principles</u>, defined as "a set of guiding principles to make data Findable, Accessible, Interoperable, and Re-usable," came out of <u>a meeting</u> in Jan 2014 that "brought together 25 high level participants representing leading research infrastructures and policy institutes, publishers, semantic web specialists, innovators, computer scientists and experimental (e)Scientists."

The idea of FAIR seems to be catching on, and potentially being applied to other types of objects, such as software. For example, a recent paper, "Four simple recommendations to encourage best practices in research software" (of which I am one of many co-authors), says:

"While the FAIR principles were originally designed for data, they are sufficiently general that their high level concepts can be applied to any digital object including software. Though not all the recommendations from the FAIR data principles directly apply to software, there is good alignment between the OSS recommendations [the software recommendations in the paper] and the FAIR data principles"

eResearch'21 - BoF on FAIR Data Quality Information

# What actually is the 'quality' of a dataset? How does the user know which dataset to trust?



- **Few common quotes:** 
  - "We can't use that dataset because it is of poor quality";
  - "Don't trust data from sector, organisation or a person: it does not meet OUR quality requirements"
  - "Don't trust repository XXXX: their datasets are full of errors and of low quality"
- But when pressed, very few could provide concrete examples of:
  - Exactly what and where the supposed errors were in the dataset;
  - What they were benchmarking the supposedly "poor" quality dataset against
  - None could provide a "community-agreed" reference/best practice document that specified what their expectations on quality were.
- "community-agreed" guidelines on quality, preferably at an international level are urgently needed

#### An International Effort Came Together...

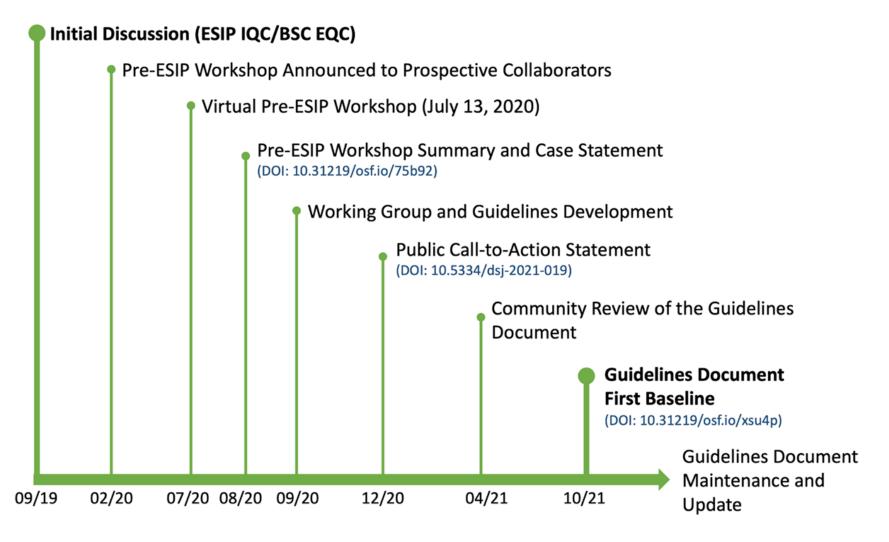


- Co-organized by:
  - ESIP Information Quality Cluster (IQC),
  - Barcelona Supercomputer Centre Evaluation and Quality Control Team (EQC),
  - ARDC-supported AU/NZ Data Quality Interest Group (DQIG)
- 22 International Interdisciplinary Domain Experts:
  - Data producers (in situ, satellite, model),
  - Stewards (data/science/technology),
  - Services providers (data/information/infrastructure),
  - Data publishers and users
- from 7 countries (USA, Spain, AU, NZ, Germany, UK, France),
  - with 22+ affiliations (government, academic, private sectors):
    - O Data, science, and service centres, institutional repositories
  - with expert knowledge from data acquisition or production, data and information management, data publishing, services, and applications.



#### Timelines and current status







#### Four key outputs so far

Hampapuram Ramapriyan, Ivana Ivánová, David Moroni, Yaxing Wei,

Irina Bastrakova, Anette Ganske, Lucy Bastin, Siri Jodha S. Khalsa,

Mingfang Wu, Chung-Lin Shie, Nancy Ritchey, Dave Jones, Ted Habermann

Christina Lief, Iolanda Maggio, Mirko Albani, Shelley Stall, Lihang Zhou,

Knowledge about the quality of data and metadata is important to support informed

decisions on the (re)use of individual datasets and is an essential part of the ecosystem that

supports open science. Quality assessments reflect the reliability and usability of data. They

need to be consistently curated, fully traceable, and adequately documented, as these are crucial for sound decision- and policy-making efforts that rely on data. Quality

systems and tools to allow for improved sharing of information on quality at the dataset

evaluation framework and presentation of resultant quality information to end users may not have been comprehensively addressed within and across disciplines. Global

interdisciplinary domain experts have come together to systematically explore needs,

challenges and impacts of consistently curating and representing quality information

through the entire lifecycle of a dataset. This paper describes the findings of that effort, argues the importance of sharing dataset quality information, calls for community action

to develop practical guidelines, and outlines community recommendations for developing

level for individual quality attribute or dimension. Although the need for assessing the

quality of data and associated information is well recognized, methodologies for an

assessments also need to be consistently represented and readily integrated across

Marie Drévillon, Sarah Champion, C. Sophie Hou, Francisco Doblas-Reves,

Gilles Larnicol, Lesley Wyborn, Mitch Goldberg, Jörg Schulz,

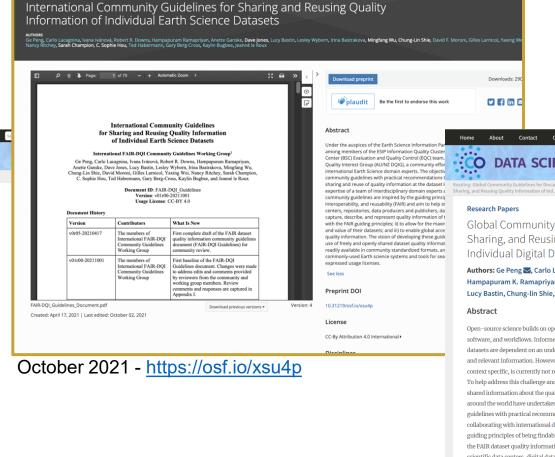
Kerstin Lehnert, Erin Robinson, Kaylin Bugbee

Abstract





August 2020 - https://osf.io/75b92/



**SFPREPRINTS** ▼

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Global Community Guidelines for Documenting, Sharing, and Reusing Quality Information of Individual Digital Datasets

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**DATA SCIENCE JOURNAL** 

Authors: Ge Peng █, Carlo Lacagnina, Robert R. Downs, Anette Ganske, Hampapuram K. Ramapriyan, Ivana Ivánová, Lesley Wyborn, Dave Jones, Lucy Bastin, Chung-lin Shie, David F. Moroni

#### Abstract

Open-source science builds on open and free resources that include data, metadata, software, and workflows. Informed decisions on whether and how to (re)use digital datasets are dependent on an understanding about the quality of the underpinning data and relevant information. However, quality information, being difficult to curate and often context specific, is currently not readily available for sharing within and across disciplines To help address this challenge and promote the creation and (re)use of freely and openly shared information about the quality of individual datasets, members of several groups around the world have undertaken an effort to develop international community guidelines with practical recommendations for the Earth science community. collaborating with international domain experts. The guidelines were inspired by the guiding principles of being findable, accessible, interoperable, and reusable (FAIR). Use of the FAIR dataset quality information guidelines is intended to help stakeholders, such as scientific data centers, digital data repositories, and producers, publishers, stewards and managers of data, to: i) capture, describe, and represent quality information of their datasets in a manner that is consistent with the FAIR Guiding Principles; ii) allow for the maximum discovery trust sharing and reuse of their datasets; and iii) enable

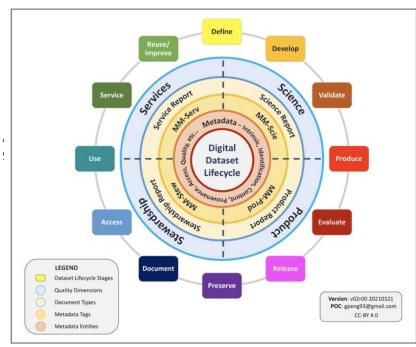
May 2021 - http://doi.org/10.5334/dsj-2021-019

March 2022 - http://doi.org/10.5334/dsj-2022-008

### Guidelines development principles



- Adapting the FAIR guiding principles (Wilkinson et al. 2016);
- Taking a whole dataset-lifecycle approach;
- Being quality-attribute and assessment-type agnostic:
- Common terminology is essential for enabling interoperability;
- Developing for the community by the community:
  - Through an iterative process, with continuous engagement with all stakeholders,
  - Leveraging the experiences and expertise of a team of interdisciplinary domain experts and community best practices and standards.



### Framework defined by 4 dimensions





## Basic workflow for curating and reporting DQI



Monitoring &

**Improvement** 

Quality Specification  Define and describe the scope of the assessment and associate quality attribute(s) or dimension(s)

**Evaluation Specification** 

Identify and describe the assessment method and framework

**Evaluation Execution** 

 Perform the assessment and capture the results in a structured, human- and machine-readable, standard-based format

Quality Dissemination

 Make the assessment results readily available and usable to stakeholders and collect feedback for improvement

Based on: Peng et al. (2022). DOI: 10.5334/dsj-2022-008

### FAIR DQI guidelines at a glance

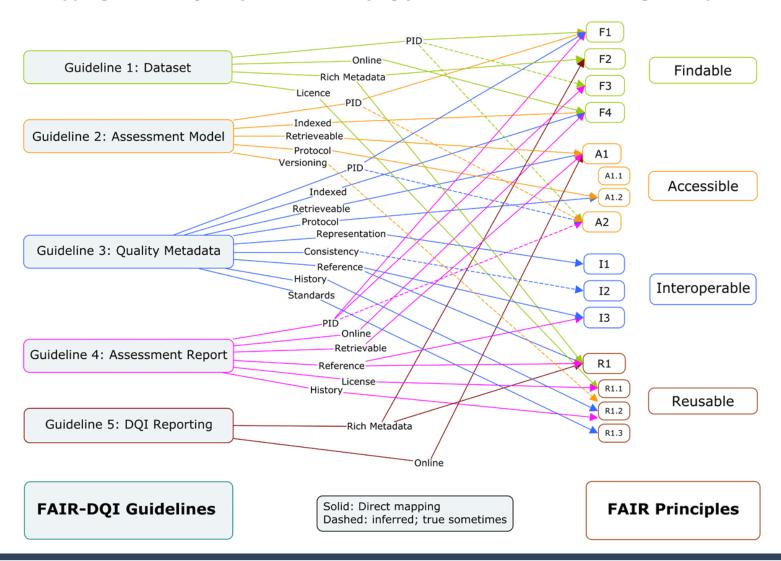


- Guideline 1: Describing Dataset (e.g. version, producer)
  - Ensure the dataset is findable and accessible.
- Guideline 2: Utilizing a quality assessment model
  - Ensure the assessment model is findable and accessible.
- Guideline 3: Capturing the assessment method and results
  - Ensure the quality information is interoperable and reusable (machine endusers).
- Guideline 4: Describing the assessment method, workflow and results
  - Ensure the quality information is findable, accessible, citable and reusable (human end-users)
- Guideline 5: Reporting the dataset quality information
  - Ensure the information is FAIR

## FAIR DQI guidelines are really FAIR CE



#### Mapping Dataset Quality Information (DQI) Guidelines to FAIR Guiding Principles



Peng et al. (2022). DOI: 10.5334/dsj-2022-008

#### FAIR DQI use cases



- ❖ At the moment: we are collecting use cases to:
  - Ensure that the guidelines are in line with the user communities and their applications;
  - Justify the need for best practices in describing quality information to ensure and proper use data;
  - Collect examples from multiple application domains on the use of FAIR quality information;
  - Provide the community with implementation examples of the guidelines;
  - Develop the guidelines for the community by the community;

The template is developed to collect data quality use cases to ensure that the guidelines developed by the international FAIR dataset quality information community gudelines working group (2021) are in line with the user communities and their applications. Please contact Ivana Ivanova at ivana.ivanova@curtin.edu.au for questions regarding the use cases collection effort or Ge Peng at ge.peng@uah.edu for issues with accessing the template.										
ID	Who (Name/Organization)	In what capacity (e.g., data producer; data custodian, funder,)	Use-Case Description	Typical object type (e.g., dataset, collection, observation, algorithm, instrument)	Current Data Quality best practice	What quality info is needed in addition to current practice (e.g. license info, provenance info)	What quality indicators make you decide to not use a dataset?	If there is no quality information, what happens?	Additional notes	Contact who can develop the use-case in detail
1	Hazard Consortium (500+ orgs)	GIS officer for disaster aftermath recovery	The damage done by Superstorm Sandy in October 2012 was unprecedented in its size and scope. In the aftermath of Sandy, Edison Electric Institute (EE) members also recognized the need to enhance and formalize the mutual assistance program for national events. In September 2013, EE's Board of Directors programs of the personal content of the	Spatial datasets	none	Information about trust and reliability of the resource	Data source and producer unknown	The recovery process will be lengthy and therefore more costly (e.g., operational expenses, properties and lives)		Dave Jones (dave@stormcenter.com)

Please contribute <u>here</u>.

## FAIR DQI guidelines: path forward



- Continue promotion through regular presence at: <u>ESIP</u>, <u>OGC</u>, <u>RDA</u>, <u>SciDataCon</u>, <u>eResearch Australasia</u>...
- FAIR DQI guidelines is a living document expected to evolve over time based on user feedback and emerging community best practice
- FAIR DQI guidelines are not only for Earth Science datasets we are expanding the discipline diversity



Thank you! ivana.ivanova@curtin.edu.au

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