



# Research Data Management

















### **Train for advanced research**

#### Welcome

MyScope was developed by Microscopy Australia to provide an online learning environment for those who want to learn about microscopy. The platform provides insights into the fundamental science behind different microscopes, explores what can and cannot be measured by different systems and provides a realistic operating experience on high end microscopes.

We sincerely hope you find our website: www.myscope.training an enjoyable environment. In there you can explore the microscopy space and leave ready to undertake your own exciting experiments.

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What is research data? Introduction to research data management Maintaining quality and integrity Benefits of managing research data **FAIR data principles** What is FAIR data The FAIR principles explained Persistent identifiers Are FAIR data and open data the same? **The CARE principles** What are CARE principles? What are indigenous data? The CARE principles explained **Be FAIR and CARE** FAIR and CARE are complementary **Global guidelines Research data management plans** Importance of file naming **File-naming conventions Directory structures Readme files** Credits

There is no authoritative definition of research data. It may depend on the discipline and the context, for example, the nature of the project and experiment. As a result, it is advised to check the definition of research data by the relevant institution.

**The Australian Research Data Commons (ARDC)** defines research data as: "Research data, from the point of view of the institution with a responsibility for managing the data includes:

- all data which is created by researchers in the course of their work, and for which the institution has a curatorial responsibility for at least as long as the [Australian Code for the Responsible Conduct of Research] and relevant archives/record keeping acts require, and
- third-party data which may have originated within the institution or come from elsewhere."

You could put this simply as, research data means information collected or recorded as part of a research project regardless of form or media.

It can include:

- all notebooks
- survey responses
- software and code
- measurements from laboratory or field equipment such as EDS spectra or hygrothermograph charts and 3D data points
- images such as digital photographs, films, scans, or autoradiograms
- audio recordings
- physical samples
- raw, analysed, processed or reduced data.

Some examples of the wide range of microscopy-related research data collected from around Microscopy Australia are below and on the following pages.









Research data management has become an integral part of research. Although it is required by many scholarly publishers and research funders, good data management is the best way to maximise impact and value of research through the reuse of research data. Suitable data storage and management protects your data from loss and ensures compliance with any rules and requirements associated with the research project (e.g. ethics and confidentiality).



Research data management covers the planning, collection, organisation, storage, preservation, documentation and sharing of research data over the lifetime of a research project, after the end of the project and beyond the publication of research findings. It thus encompasses both the everyday and long-term management of research data.



Research data management is essential for maintaining the quality and the integrity of research data, and ensuring compliance with legal, regulatory, ethical and funding requirements.



**USED IMAGEJ...** 

**This short video from NYU Health Sciences Library** shows that when research data aren't managed properly, reusing and sharing data is difficult or impossible.



Research data management has multiple and important benefits:

- It can enhance research impact and value by maximising the visibility of research data and promoting transparency in research;
- It improves accessibility to research data by ensuring that its quality and integrity are maintained during and beyond the lifetime of the research project. It is easier to replicate and validate findings from an experiment when research data are well managed;
- It saves times and optimises the use of storage resources;
- It ensures responsible use and reuse of research data, including by attributing appropriate authorship (citations) and by following confidentiality obligations, legislation, privacy or ethical principles or other guidelines associated with the research data;
- It can safeguard and future-proof research data by establishing appropriate, safe and secure short-term and long-term storage of data (including depositing data to a repository for long-term archiving and access);
- It minimises the risks of research data loss and corruption, as well as breaches in confidentiality or ethical requirements associated with the research data.

## WHERE ARE THE REST OF THE IMAGES FROM THAT EXPERIMENT?

WE DUMPED THEM BECAUSE THE SERVER WAS FULL. ANYWAY, WE'VE ALREADY PUBLISHED THE PAPER. Good data management has become increasingly recognised as integral to research. However, it is not a goal in itself. Instead, it favours the discovery, sharing and reuse of research data after the publication of the studies they underpin. The FAIR Guiding Principles for scientific data management and stewardship promote practices that enable research data to be Findable, Accessible, Interoperable and Reusable for both machines and humans. They enhance the exposure of research data thereby increasing the impact of research.

Machine-actionability is especially central to FAIR data. It means that computers can find, access, interoperate and reuse data with limited or no human intervention. In this era of big data, when the volumes, level of complexity and velocity of data keep increasing, machine-actionable data will play an important role.

The FAIR principles build on earlier efforts that have aimed to facilitate the reuse and sharing of data. A major change associated with the FAIR principles is that raw research data have become usable, analysable, citable scholarly outputs. This has had consequences for research organisations, publishers, funders and data repositories because for research data to be FAIR, metadata (information about data) and supporting infrastructure (for data storage and sharing) must also be compatible with the FAIR principles. So the FAIR principles require standardised approaches to how research data can be:

- described and identified using metadata (information about data) and persistent identifiers (PIDs) following community-agreed conventions;
- stored in a way that maximises discoverability, reuse and sharing of data while keeping a balance between sustainability of storage and permanence of data.



#### Findable data

Reusing data implies that, first, it is possible to find them. It should be easy for both humans and computers to find data and the associated metadata. Data and metadata are assigned a globally unique and persistent identifier that is machine-readable. Metadata and PIDs are essential for automatic and reliable discovery of datasets and services. Metadata must clearly and explicitly include the identifier of the data it describes. Using community-endorsed, general or discipline-specific, standards for data and metadata (such as data models and metadata schemas) is recommended.

#### Accessible data

Once data are found, it must be clear how to access them. This can include information on whether authentication or authorisation is required prior to accessing the data. Data and metadata should be retrievable by their identifier using a standardised and open communications protocol that is open, free and universally implementable. It must allow for authentication and authorisation procedures for restrictions if necessary. Importantly, metadata should remain available and accessible even when the data are no longer available.

#### Interoperable data

It should be possible to integrate the data with other data, applications and workflows. This means that the format of the data should be open and interpretable by various tools for analysis, storage and processing (which proprietary software appications generally don't allow). They must use a formal, accessible, shared and broadly applicaable language for knowledge representation Interoperability applies both at the data and metadata level. Common formats and standards for data and metadata, as well as controlled vocabularies (that follow FAIR principles), are essential to ensure data and metadata interoperability.

#### **Reusable data**

The ultimate goal of FAIR is to optimise data reuse. To achieve this, data and metadata should be well-described, have a clear and accessible license for data usage, include detailed information on provenance, and meet domain-relevant community standards. This will ensure repeatability of experiments and reproducibility of results.

#### Persistent identifiers

Persistent identifiers (PIDs) are long-lasting, globally unique, digital references to objects (for example, publications, datasets, funding grants), people or organisations. The figure below shows examples of established PIDs that can be used. Additional PIDs are being developed so their levels of awareness and adoption can vary across communities

PIDs are a critical piece of infrastructure that underpins the FAIR principles:

- They link entities in the research ecosystem (for example, researchers, organisations, articles, datasets, samples, software, funders, grants, instruments) so they can favour innovation and improve efficiency;
- They enable tracking of research impact by enabling research provenance and attribution;
- They support research integrity and reproducibility of results by identifying research inputs (for example, samples, instruments, software) and research outputs (for example, datasets, articles).

Overall, the integration of PIDs into information systems used in research organisations (for example, laboratory information management systems, image repositories, research management systems, finances, human resources) facilitates information exchange between those systems across and between organisations. It also eliminates the need to re-enter information manually multiple times into multiple systems, for example about a grant, a publication or a person (such as their publications), which can lead to significant cost savings for research organisations.

#### Researchers

## ORCID

ORCID ID (Open Researcher and Contributor ID) *orcid.org* 

Research organisations



ROR ID (Research Organization Registry) *ror.org* 

#### Objects (physical, digital, abstract)



DOI (Digital Object Identifier) *doi.org* 

#### **Research projects**



RAiD (Research Activity Identifier) *raid.org* 

#### **Physical samples**



IGSN ID (International Generic Sample Number Organisation) *igsn.org* 

Key resources in biomedical research



RRID (Research Resource Identifier) *rrids.org* 

Examples of PIDs used in research

#### Are FAIR data and open data the same?

No. FAIR data doesn't necessarily mean "open" or "free" data. Data can be restricted and still be FAIR. The important aspect of "accessible data" is that the conditions to access the data are explained in a way that both a human and machine can understand. FAIR data can be as open as possible and as closed as necessary. Both private and protected data can be FAIR. For example, accessible data can mean that users register an account to access a repository.

The "CARE Principles for Indigenous Data Governance" were defined following the FAIR principles. They focus on Collective benefit, Authority to control, Responsibility and Ethics in the use and sharing of Indigenous data and they highlight Indigenous Peoples' rights and interests in data governance [3]. FAIR and CARE principles are complementary.

Despite the current movement towards FAIR data, open data and open science to promote accessibility and sharing of research data, Indigenous Peoples' rights and interests over Indigenous data have often been overlooked or not explicitly considered. For example, the emphasis on greater data sharing needs to take into account the fact that Indigenous Peoples are also asserting greater control over the application and use of Indigenous data and Indigenous Knowledge for collective benefit.

The CARE Principles for Indigenous Data Governance (Collective benefit, Authority to control, Responsibility and Ethics) were developed by the Global Indigenous Data Alliance (GIDA) (Figure below). They provide guidance to data producers, users, managers and publishers on the inclusion of Indigenous Peoples in data governance and management for a better discovery, access, use, reuse and attribution of Indigenous data. The CARE principles build upon earlier work by the Te Mana Raraunga Māori Data Sovereignty Network in New Zealand, the US Indigenous Data Sovereignty Network, the Maiam nayri Wingara Aboriginal and Torres Strait Islander Data Sovereignty Collective in Australia and numerous other Indigenous Peoples, nations, and communities.



## CARE Principles for Indigenous Data Governance

Source: https://www.gida-global.org/care

#### What are Indigenous data?

They are data, information and knowledge, in any format, that impact or concern Indigenous Peoples, Nations and Communities at the collective or individual levels. They include data about:

- Individuals: for example, administrative, legal, health, social, commercial, corporate, services;
- Collectives: for example, traditional and cultural information, languages knowledge systems, ancestral and clan knowledges;
- Resources and environments: for example, land, water, geology, titles, air, soil, sacred ecosystems, territories, plants, animals.

#### **Collective benefit**

Data ecosystems should be designed and function in ways that enable Indigenous Peoples to derive benefit and equitable outcomes from Indigenous data (such as inclusive development, innovation, improved governance and citizen engagement) through Indigenous nation and community use and reuse of data, use of data for policy decisions and evaluation of services, and creation and use of data that reflect community values.

#### Authority to control

Indigenous Peoples' rights and interests in Indigenous data must be recognised and their authority to control such data be empowered. Indigenous data governance enables Indigenous Peoples and governing bodies to determine how Indigenous Peoples, as well as Indigenous lands, territories, resources, knowledges and geographical indicators, are represented and identified within data.

#### Responsibility

Those working with Indigenous data have a responsibility to build relationships based on respect, reciprocity, trust and mutual understanding, as defined by the Indigenous Peoples to whom those data relate. Information must be shared about how data are used to support Indigenous Peoples' self-determination and collective benefit through openly available and meaningful evidence. This includes enhancing data literacy and supporting the development of an Indigenous digital infrastructure, able to generate data grounded in the languages, worldviews and lived experiences of Indigenous Peoples.

#### Ethics

Indigenous Peoples' rights and well-being should be the primary concern at all stages of the data life cycle and across the data ecosystem. Representation and justice, as well as consideration of potential future use (or harm) should be incorporated. This includes acknowledging the provenance and limitations or obligations for secondary use, especially in issues of consent.

The CARE principles are complementary to the FAIR principles. While the FAIR principles focus on the characteristics of data, the CARE principles highlight and consider the people and purpose for which data exist and are used, reflecting the important role that data can play to advance Indigenous innovation and self-determination [6].

It is important to be aware that Indigenous data may be buried in larger datasets, or that samples or additional data used in experiments or projects may be associated with Indigenous data. They may be hard to find, mislabelled and controlled by others in a manner inconsistent with the FAIR and CARE principles. As a result, data could be subject to both CARE and FAIR. It is in general easier to apply the CARE principles to data that are already FAIR.



Source: https://www.gida-global.org/care

#### **Global guidelines**

An increasing number of funding bodies and scholarly publishers have required or encouraged appropriate management of research data in Australia and other parts of the world. In Australia, the **Australian Code for the Responsible Conduct of Research**, **2018** sets out basic requirements in research data management. Compliance with it is mandated by the Australian Research Council (ARC) and the National Health and Medical Research Council (NHMRC) to be eligible to their funding.



Globally, the Organisation for Economic Co-operation and Development (OECD) encourages good practices in data management to ensure long-term sustainability of research data from public funding (**OECD Principles and Guidelines for Access to Research Data from Public Funding**).

In the European Union, the European Commission encourages or requires data management planning be included in projects participating in Horizon 2020, the funding programme for research and innovation run by the EU.

In the USA, many funding agencies such as the National Science Foundation (NSF) and the National Institutes of Health (NIH) require appropriate data management planning with every funding request.



Effective and successful research data management requires planning. A research data management plan records the formal plan for the storage, retention and disposal of, and access to, research data that are created or generated over the course of a research project.



There are a range of issues to consider when planning, including:

- the nature of the research data and the description of the file format(s) (e.g. will new data be generated or collected? Will some existing data be reused and/or transformed? Will the data be in proprietary/non-proprietary formats, converted to other formats and archived in shareable formats?). Whenever possible, open, shareable formats are preferred;
- the description and documentation of the data, in particular the metadata and persistent identifiers (PIDs) collected, the metadata/PIDs standard(s) and format(s) used and the tools and software (including versions) used to capture, read and use the data and the metadata/PIDs;
- the policies and procedures for the organisation, storage, archiving (including backups), deletion and accessibility of research data throughout and after the research project, including whether the data will be deposited to a public repository;
- the institutional, legal, regulatory, ethical, funding and contractual policies, arrangements and requirements applicable to the research data, including privacy, confidentiality and security issues;
- the roles and responsibilities for data management amongst the members of the research team, in particular in relation to data ownership, stewardship and/or custodianship and upon the departure of a team member from the research project or the institution.

Institutions often provide online tools, templates and various other resources to create research data management plans. They often offer training sessions for students and researchers to help them manage their data.

Proper and consistent management of data files through consistent file-naming and directory-structure conventions are integral to good research data management.



Image credit: xkcd, 'Documents' (shared under CC-BY-NC License).

This video series from Adelaide Microscopy (the Microscopy Australia facility at the University of Adelaide, Australia) is part of their training program in best practice in research data management. Other facilities and institutions across the world have developed similar content and resources.

## Why you should manage your microscopy image data https://www.youtube.com/watch?v=ZHF7Y2e-pTw



File-naming conventions are a framework for naming files in a way that describes what a file contains and how it is related to other files. Consistently and logically naming research data, including microscopy images, makes it easier to organise, locate and quickly identify the information contained within these files.

A consistent file-naming convention facilitates sharing and navigating research data throughout the lifespan of a research project. It can help identify the most important attributes of a specimen or sample quickly. Importantly, it also helps navigate data in a shared or collaborative file-sharing setting.

A file-naming convention should be established at the start of a research project before data are collected or created in order to prevent an accumulation of files with improper names, which could lead to data loss.

The following videos set out some excellent strategies for developing file naming conventions for a project and can be used in conjunction with **this downloadable worksheet**. This is a great tool to help you or your users.

#### File-naming conventions for microscopy data https://www.youtube.com/watch?v=epU2DujHF1Q



File-naming convention in 7 steps – Worked example for biological data https://www.youtube.com/watch?v=dTGof1we3b4



#### Check your understanding

Consider a study using additive manufacturing (3D printing) which is a fabrication process used to create complex structures.

The experimental approach investigates the effects of printing parameters, including the nozzle diameter, layer thickness and the infill density on structural properties of 17-4 PH stainless steel specimens fabricated via additive manufacturing. Question:

#### Which set of parameters will be most useful to encode in the image file names?

• Option 1: Details of the nozzle diameter, the layer thickness, the date and the material type should be encoded in the file names, or

• Option 2: Details of the nozzle diameter, the layer thickness, the infill density and a project identifier should be encoded in the file names.

#### Answer:

Option 2 is the best strategy for a file-naming convention. Filenames should include sufficient information that make it easy for the user to distinguish one specimen from another. In the example study, two specimens may differ only by the infill density when the same nozzle diameter and layer thickness is used. Including a project identifier may be useful to easily link files back to the principal investigator and documentation related to the project.

In the case of option 1, the material type (17-4 PH stainless steel) is the same over all specimens. Meaningful filenames should include parameters that change from one specimen to another. Thus, the infill density in addition to the nozzle diameter and layer thickness should be included. The date may be encoded in the file metadata, so is not necessarily essential in the filename. Further, it is not clear that time (i.e. the date of manufacture) affects the specimen properties.

How research files are organised is as equally important as how they are named. Having a well-organised and logical directory structure of folders and subfolders will make it easy for yourself and others to locate files.

Again, the attributes of specimens and samples can inform the design of the directory structure. Different directory structures may be used depending on whether data were collected at the microscope or organised for longer-term storage or analysis.

#### Directory structures for microscopy data

#### https://www.youtube.com/watch?v=GGibzrACkJI



#### Check your understanding

Consider an experiment in which one microscope image is acquired per specimen. The study contains 10 specimens; therefore 10 images will be created in total. Question:

#### Which type of directory structure is preferable for this amount of data?

• Option 1: A 'deep' directory structure where data is saved in subfolders nested in subfolders that are separated by study attributes, or

• Option 2: A 'shallow' or 'flat' directory structure where data is saved directly in the raw data folder.

Answer:

Therefore, option 2 is the preferred directory structure because ten files can be easily viewed at a glance. Combined with a meaningful and consistent file-naming convention, it will be easy to separate the files of one study group from another using the default sort tool available in the operating system of the computer used. It is however ultimately up to the researcher to decide when additional subfolders will benefit how they use and share data.

Consistent and logical file names and directory structures are useful for navigating research data. However, interpretation of these conventions may be challenging for other researchers, or even for the researcher who created them when accessing data in the future.

README files are a method for documenting all the information that any researcher would require in order to understand different attributes within a filename and/or for locating and future use of important research files. It is best practice to create README files as research data are collected or modified, and to update it throughout the project where changes need to be documented. Importantly, a README file is also important to the researcher who created the files or organised them.

#### README file for microscopy data https://www.youtube.com/watch?v=UPSRn\_CQUY0



#### Tips for what to include in a README file

The template README file can be downloaded **here**. The template includes the elements as described in the above video.

Below are examples of README files for microscopy data from:

#### **Biological sciences**

```
O README_example_biological.txt
Title of dataset:
Raw image data from 1276 microCT for ExperimentA, Anon Lab Group
Principal Investigator: Prof Anonymous, Research Institute, prof.anonymous@institute.edu
Data Creator: John Smith, Research Institute, j.smith@institute.edu
Project objectives:
To quantify changes in lower limb bone mass, between 14 and 24 weeks old, in two common mouse strains
with or without ovariectomy.
                                                                            --DATA FILES-
File naming convention:
strain-treatment-specimenNum_age_version_imageNumber.fileFormat
Attributes:
  ttributes:
    strain = indicates whether the mouse was C57BL/6 or BALB/c
    treatment = indicates whether the mouse was ovariectomized or non-surgical control
    specimenNum = a 2-digit number assigned to each mouse in the treatment groups (between 01-06)
        age = age of the mouse, in weeks, at the time of the microCT scan
    version = indicating whether image is raw data, reconstructed or processed data file
    imageNumber = 4-digit image slice number in the microCT z-stack (e.g. 00xx)
    fileFormat = 8-bit bitmap (.bmp) image files as generated by the microscope
Acronyms and Codes:
        nyms and Codes:
    strain = {C57BL/6: "BL6", BALB/c: "BaC"}
    treatment = {ovariectomized: "OVX", non-operated control: "CTL"}
    specimenNum = {"01", "02", "03", "04", "05", "06"}
    age (weeks) = {"wil4", "wil6", "w18", "w20", "w22", "w24"}
    version = {raw: "raw", reconstructed: "rec", processed: "proc"}
Examples:
The raw image file of the 3rd C57BL/6 mouse from the ovariectomized group, microCT scanned at 18-weeks-old

    BL6-OVX-03_w16_raw_0001.bmp
    The raw image file of the 1st BALB/c mouse from the non-surgical control group, microCT scanned at 22-weeks-old

              – BaC–CTL–03_w22_raw_0001.bmp
                                                                              -DATA LOCATION-
The raw image data is located in the following folder in the laboratory's central storage allocation:
.../AnonLabGroup/ExperimentA/RawData/
Raw data folders are organised by strain then by treatment as shown below:
RawData
           -BI 6
                       -0VX
                СТЬ
             BaC
                      _ovx
                 Г
                       -CTL
```

README_example_geological.txt
Title of dataset: ICP-MS microcharacterisation data from Agilent 7900x for experimentA, Anon Lab Group
Principal Investigator: Prof Anon, Research Institute, prof.anonymous@institute.edu Data Creator: John Smith, Research Institute, j.smith@institute.edu
Project Objective: Use LA-ICP-MS for U-Pb dating of zircon in rock samples acquired from three locations (Flinders Ranges, Willouran Ranges and Mount Lofty)
Attributes collectionDate = date the rock sample was collected locationIDsampleNum = location code for sample with appending 3-digit site number sampleNum = a 2-digit number assigned to each sample for a given site samplePrep = indicates whether sample was thin-sectioned for SEM imaging with appended 2-digit image number fileFormat = .tiff image file or .csv as generated by the microscope or ICP-MS
Code collectionDate = YYYYMMDD locationID = {Flinders Ranges:"FR", Willouran Ranges: "WR", Mount Lofty: "ML"} samplePrep = {thin-sectioned: "TS"}
Examples: ICP-MS data for the 8th site and 2nd rock sample acquired during field trip between dates 21-26 April 2021 to Flinders Ranges. Sample was thin-sectioned and SEM imaged – 20210421FR008_02_TS01.tiff (thin section image) – 20210421FR008_02 – 1.csv (ICP-MS data, "- 1" is appended by analysis software")
DATA LOCATIONDATA LOCATION
The raw data is located in the following folder in the laboratory's central storage allocation: /AnonLabGroup/ExperimentA/RawData/
Raw data folders are organised by date as shown in the example below:
RawData 

#### **Plant sciences**

O README\_example\_plant.txt Title of dataset: Optical microscopy image data from Olympus BX51 for experimentA, Anon Lab Group Principal Investigator: Prof Anonymous, Research Institute, prof.anonymous@institute.edu Data Creator: John Smith, Research Institute, j.smith@institute.edu Project objectives: Investigate the effect of staining method on automatic cell identification in different regions of potato tubers -DATA FILES-File naming convention: sizeSampleNum\_zone\_coreRegion\_sectionNum\_stain\_imgNum.fileFormat Attributes: Attributes: size = indicates potato tuber size according to mass (small, medium or large) SampleNum = a 2-digit number assigned to each tuber zone = one of three subregions of each tuber (bud, middle or stem) coreRegion = origin of the tissue core, either from the inner core (pith) or outer core (perimedullary) sectionNum = a 2-digit number assigned to sections from each core sample (between 01-05) stain = indicates one of four agents used for staining the tissue samples imgNum = a 2-digit number at the end of each image file fileFormat = 24-bit .tiff image files as generated by the microscope Acronyms and Codes: Acronyms and Codes: size = {small: "S", medium: "M", large: "L"} zone = {bud: "Z1", middle: "Z2", stem: "Z3"} coreRegion = {inner core: "IC", outer core: "0C"} stain = {None = blank, Safranin-0: "safo", Toluidine Blue-0: "tolu", Lugol's iodine: "lugol"} Examples: The first image of the 2nd thin section taken from the inner core of the bud zone of the 1st medium potato tuber, stained with safranin-0 - M01\_Z1\_IC\_02\_safo\_01.tiff -----DATA LOCATION-----DATA LOCATION------DATA LOCATION------DATA LOCATION------DATA LOCATION: .../AnonLabGroup/ExperimentA/RawData/ Raw data folders are organised by strain then by treatment as shown below: RawData Small Unner Unter -Medium ----Inner Outer

#### Check your understanding

Question:

#### Using the excerpt from the README file below, is the following statement true or false?

```
README.txt
File naming convention:
treatment-specimenNumber_pixelSize_imageNumber.fileFormat
Attributes:
...
pixelSize = a 3-digit code indicating the pixel size, in
µm, used for imaging the specimen.
...
Acronyms and Codes:
...
pixelSize = {1µm: "01u"; 4µm: "04u"}
...
```

'Considering the following filename SHM-01\_01u\_0001.bmp, this specimen was imaged with 1 µm pixel size.'

Answer:

This is true. From the README file we can see that "01u" indicates that the specimen was imaged with a 1 µm pixel size. The README file helps understand the file-naming convention and codes, acronyms and abbreviations used.

#### **RDM module Credits**

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