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Opensciency - A core open science curriculum by and for the research community.

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Opensciency is core open science curriculum material, drafted to introduce those beginning their open science journey to important definitions, tools, and resources; and provide for participants at all levels recommended practices. The material is made available under a [CC-BY 4.0 International](#) license and is structured into five modules:

- Ethos of Open Science
- Open Tools and Resources
- Open Data
- Open Software
- Open Results

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Details

Opensciency is a result of the work of more than 40 open science experts and practitioners from across the world and from different disciplines. The first draft of the curriculum material was developed from [June 27 - July 1, 2022](#) as part of the Transform to Open Science (TOPS) [Open-Core](#) sprint. More information about the NASA TOPS initiative is available via their [website](#). After the TOPS Community Panel on [October 6, 2022](#), the original contributors created the Opensciency repository to allow all contributors to further engage with the curriculum and invite review on the initial draft material from the wider research community.

We encourage the wider community to reuse the material, and we are especially interested in creative approaches to displaying the material. An example we like is [Elements of AI](#).

Let us know if you have a creative approach to displaying and reusing the material by [submitting an issue](#). Please provide your contact details so we can add you to the contributors list.

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Ethos of Open Science

What is Open Science and what practices does it promote?

Introduction

This is the first lesson in the module on the Ethos of Open Science. We'll start explaining what we mean by the word, "ethos". Ethos is defined by Merriam-Webster as "the distinguishing character, sentiment, moral nature, or guiding beliefs of a person, group, or institution". So this lesson is about what makes Open Science, as an approach to knowledge-production, unique or distinguishable from other scientific methods.

Note that "ethos" is not exactly "ethics", but it is a broad enough term to include the moral attitudes held by the individuals or institutions practicing open science. To make it clear that there is a moral element to this discussion, we speak of "responsible Open Science" going forward.

The lesson introduces the concept of open science as a whole, by explaining the history underpinning open science, what open science is, and how it works. It then discusses different components of open science and the pillars that make them up. At the end of the lesson, students will have an understanding of the brief history of open science and its definition.

Open science goes beyond publishing— it is a redefinition of scientific collaboration and output. It is a culture intended to promote science and its social impact. Open science creates new opportunities for different stakeholders including researchers, decision makers, and public participants. Open science increases study transparency, repeatability, reproducibility, and confirmation. We expand what these terms mean and why they matter throughout this module and later OpenCore modules.

Context and Definition

Science evolves through collaborative development of theories and practices that are open for others to learn and build on. Throughout the ages - whilst in some cases, education and science was out of reach for the general populace and may have been kept for a privileged few, there have been other educational and scientific resources that were purposefully made available for others to re-use. Think of how dictionaries and encyclopedias have been around for centuries specifically to share standards of knowledge. ([The first](#) "dictionary" dates back over 3,000 years!) Libraries, in turn, have existed for millennia to serve as repositories of knowledge in diverse formats, from ancient tablets and scrolls, to the books we expect to see today. Public museums have also been around for some time and play the role of educating people, as well as maintaining archives for researchers to gain further insights from.

Institutions and practices throughout the ages have facilitated humanity's endless desire for knowledge. As far back as the Medieval era, we already find physicians being encouraged to

review one another's work to ensure it was carried out appropriately (Rogers, 2021). Today, we call this practice "peer review". And, during the Enlightenment, scientists formed networks with whom they shared their theories via hand-written letters, and the adoption of the printer allowed for the emergence of scientific institutes and journals (Green, 2017; see Kherroubi Garcia et al., 2022).

However, open science has only become a distinct set of practices in recent decades. We can see open science as both being encouraged by social and technological developments, and responding to problems in the scientific process. The emergence of the internet and other digital technologies have more recently allowed for science to be conducted even more collaboratively. In 1971, [Project Gutenberg](#) started making books in the public domain available online. In 1987, we saw [the first open access journal](#) being published. In 1991 the central storage platform arXiv was launched for the exchange of manuscripts in physics (though without peer review) (Ginsparg, 2021).

However, these endeavors do not amount to open science in the sense we discuss it today. In recent years, we have learned of various issues in the scientific process that necessitate specific responses. Two such issues are the replication crisis (Fidler & Gordon, 2013; Elsherif et al., 2021a) and publication bias (Joober, et al., 2012; Elsherif et al., 2021b). The replication crisis refers to scientific findings not being validated by other scientists' efforts to replicate them. The publication bias amounts to the greater ease to publish scientific findings that only "very clearly" confirm or disprove hypotheses.

Thus, open science captures both the spirit of making knowledge more accessible *and* responding to poor scientific practices. We will discuss more reasons why open science is important, both the personal benefits and as a public good, in Lesson 2 of this module, "Benefits and Challenges of responsible Open Science: Why does it matter?"

Definitions of Open Science and responsible Open Science

Formal definitions and governance mechanisms to ensure best practices in open science have emerged alongside the open science movement.

- In 1997, [COPE was established](#) and has since supported the fostering of responsible publishing culture.
- The [2001 Budapest Open Access Initiative](#) provided a clear working definition of *open access*, one of the components of open science (as we will see shortly).
- In 2012, the [Contributor Role Taxonomy](#) was developed so that more diverse collaborators in research can be adequately credited for their work.
- The [2013 Declaration on Research Assessment](#) then outlined best practices in the assessment of research.
- In 2014, the [Open and Collaborative Science in Development Network](#) was established to enable open science approaches to developmental research for the Global South.

- 2014 also saw the launch of the [Data Citation Principles](#), which advocate for – amongst other things – making data independently citable.
- The [2016 FAIR principles](#) emerged as a way to guide practices in open science, and enabled the implementation of the Data Citation Principles.
- The [2018 CARE principles](#) established data governance practices for indigenous data and practices.

In this complex context, we can draw on a few definitions of *Open Science*:

“Open Science is transparent and accessible knowledge that is shared and developed through collaborative networks” (Vicente-Saez & Martinez-Fuentes, [2018](#)).

“Open science is [...] an inclusive construct that combines various movements and practices aiming to make multilingual scientific knowledge openly available, accessible and reusable for everyone, to increase scientific collaborations and sharing of information for the benefits of science and society, and to open the processes of scientific knowledge creation, evaluation and communication to societal actors beyond the traditional scientific community. It comprises all scientific disciplines and aspects of scholarly practices, including basic and applied sciences, natural and social sciences and the humanities, and it builds on the following key pillars: open scientific knowledge, open science infrastructures, science communication, open engagement of societal actors and open dialogue with other knowledge systems (UNESCO, [2021](#)).

Globally, Open Science is being valued and given importance as it recognizes disparities and regional differences, providing a framework to handle challenges and contribute to minimize knowledge, technological and digital differences between countries. For instance, when different researchers from across the globe are invited to research collaboratively, trust and novelty increases and as a result it improves quality, efficacy and responsiveness in research as being the benefits of Open Science.

Here are some other definitions of Open Science. Are there any more you would add?

Open Science is a practice for increasing the accessibility and transparency of scientific research. The concept of Open Science is built around shared principles such as inclusion, fairness, equity, & sharing (Zee & Reich, [2018](#)).

An umbrella term reflecting the idea that scientific knowledge of all kinds, where appropriate, should be openly accessible, transparent, rigorous, reproducible, replicable, accumulative, and inclusive, all which are considered fundamental features of the scientific endeavor. Open science consists of principles and behaviors that promote transparent, credible, reproducible, and accessible science. Open science has six major aspects: open data, open methodology, open source, open access, open peer review, and open educational resources. (FORRT open science glossary, <https://forrt.org/glossary/open-science/>)

Open Science aspects

Open science has various components: open access, open access journals, open peer review, open research data, open source, open science policies, with use of open licensing, open software for reproducible research, among others (Open Science Basics, retrieved from <https://open-science-training-handbook.gitbook.io/book/open-science-basics>, 2022).

The below image from Robinson (2018) captures some of the components of open science, although the list differs depending on who you ask (see Pontika et al., 2015).



Figure 1: Open Sholarship umbrella contains Open Educational resources, EDI, community science, open data, open science, open access, open source.

(Image from Robinson, 2018; needs adapting in light of the list below)

Open Science is an umbrella term that captures eight components. The below list helps us reflect on the ambition that drives the open science movement. In short, open science is not limited to a discipline or a particular aspect of scholarly practice. Rather, open science seeps into every practice of scholarly work.

- **Open Access** refers to making research methods, data and outputs accessible by default, where advisable; this is touched on in lesson five below.
- **Open Data** relates to making data used in science accessible for others to study, reusable for other pertinent projects, and available for redistribution. More on this topic will be discussed in the module *Open Data*.

- **Open Software** is about making the source code of software transparent, allowing people to collaborate on its improvement; more will be said in the module *Open Software*.
- **Open Tools & Resources** are those that have been developed precisely to facilitate open science practices, from open hardware and online toolkits to behavioral guidelines; learn more in the module *Open Tools & Resources*.
- **Open Results** is a broad term capturing open access, open data and open software, as it is about making results from all stages of a research lifecycle open, including their evaluation, which should not be limited to traditional peer review; learn more in the module *Open Results*.
- **Open Educational Resources** are learning and teaching materials made available through [open licenses](#) that permit no-cost access, re-use, re-purpose, adaptation and re-distribution by others (see [UNESCO's explainer](#)); note that the present TOPS OpenCore is an example of such a resource!
- **Equity, Diversity, Accessibility and Inclusion** are crucial values for the growth and sustainability of open science practices, as they foster the wellbeing of open science practitioners and communities. Shared principles about responsible scientific outputs also shape the behaviors of open science communities, with codes of conduct as a mechanism to ensure inclusive practices (see the following component of open science).
- **Open Community Practices** refers to the fact that open science is conducted by communities of practitioners that foster collaborative working environments, beyond disciplinary boundaries and professions; this is touched on when discussing stakeholders in lesson three of this module.

There is no *one* ethos

It is important to note that there is no one unique way of practicing or conducting open science. The outlined categories show us the diversity of practices involved in open science. Research has also shown that there are at least five schools of thought in open science, each one holding different assumptions and striving for different goals (Fecher & Friesike, [2013](#)):

Diverse practices, assumptions and goals are just part of the complexity of open science. There are also divergent moral principles guiding open science communities. Such principles are captured in *codes of conduct*. A code of conduct is a community governance mechanism that outlines the principles and practices expected of a given research community's members, as well as the process for investigating and reprimanding those in violation of the code.

In a sense, a code of conduct constitutes the moral backbone of a research community. However, as with the numerous schools of thought, there are similarly many codes of conduct. In other words, there is no *one* set of universal principles that all open science practitioners abide by. For example, consider how [OLS](#), [INOSC](#), [allea](#), [AGU](#) and [Ethical Source](#) all have different codes of conducts and guiding principles.

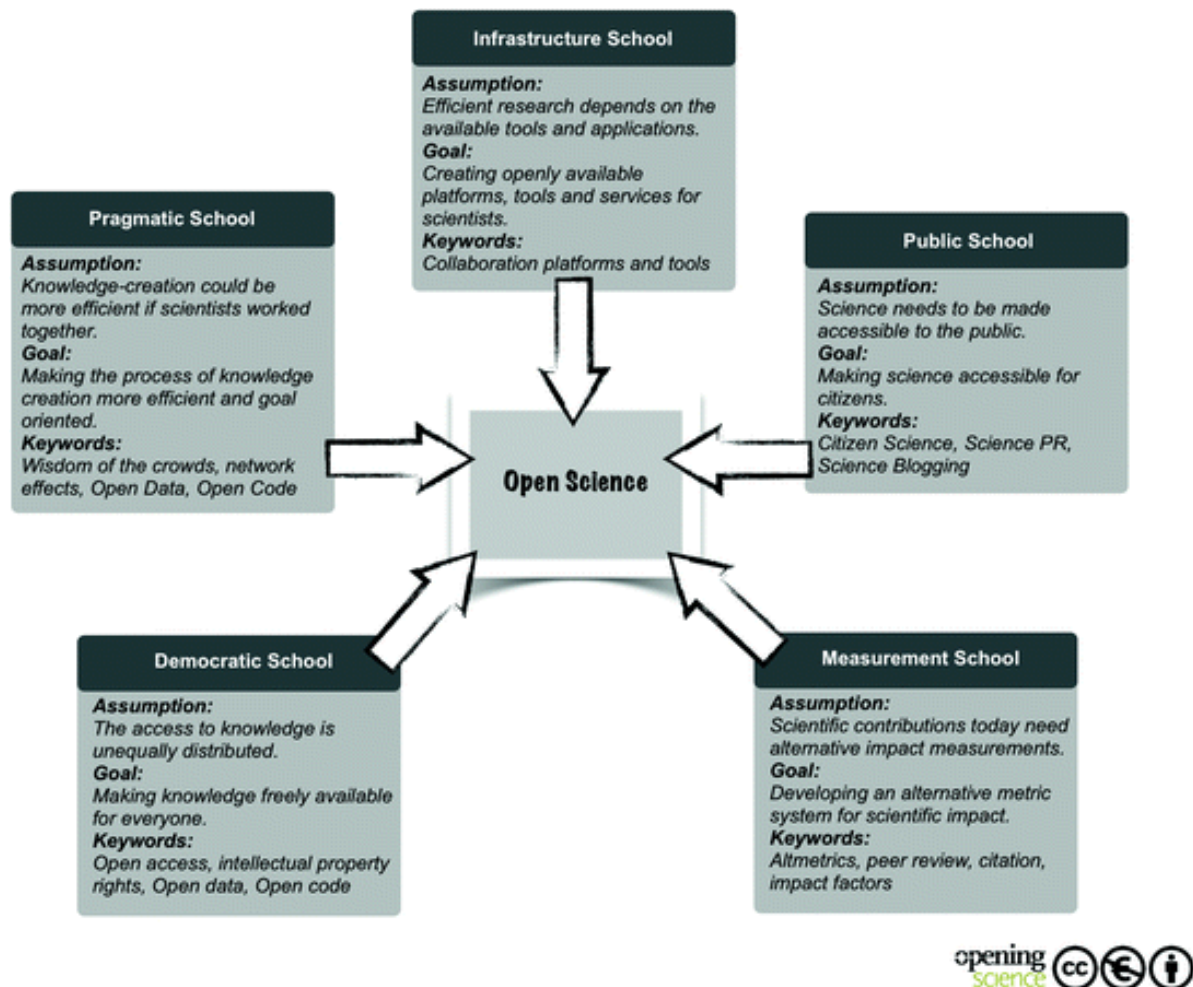


Figure 2: Five chools of open science: infrastructure, pragmatic, democrratic, measurement, public school. Visit the paper for more details on each.

This great diversity responds to the growing proliferation of open science initiatives and the great use we can make of open science approaches to knowledge.

One of the biggest driving forces is the effect of open science on the research performance. Indeed, some studies have even found that the best-performing universities are those that conduct science following open practices (see Huang et al., [2020](#)). More will be said in the following lesson about the benefits of open science and different stakeholders.

For now, consider some of the regional policies encouraging open science:

- The European Commission ([2017](#)) has outlined the skills and competencies researchers need to practise open science;
- The National Academies of Sciences, Engineering and Medicine ([2018](#)) promotes open science by design as a vision for 21st century research;
- UNESCO ([2021](#)) has developed a series of recommendations to ensure best open science practices, which are conducive to the United Nations' [Sustainable Development Goals](#);
- The European Open Science Cloud (EOSC) for finding and re-using data, and the Open Research Europe (ORE) publishing platform (European Commission, [2021](#)).

Ultimately, open science practices guide approaches to knowledge-creation that best help confront the challenges of our era. Through this module and the wider TOPS curriculum, you can become a part of this impactful movement.

Performing open science *responsibly*:

Responsible Open Science is a term we use through the rest of the module. We define it as: considering open science as the core of your science project and maximizing ethical actions for open science to minimize current challenges (e.g. data sharing, inclusion, and accessibility). In responsible Open Science, the best possible and practical practices should be explored at the early stage of your science project.

Here we share with you following rules of thumb:

- Using best practices where possible
- Being practical and realistic about resources available and pressures on open science practitioners
- Not sharing things that shouldn't be shared
- Being inclusive of all people

Summary

In this lesson, we have learned a brief history of open science, its definition, and the ethos of open science and definition of responsible Open Science. Open science practices provide significant advantages relative to more traditional closed practices. However, there are still problems that must be addressed, which many view as obstacles to open science. In the next lesson, we will talk about the benefits of open science and its challenges.

Further Reading:

Below are some further readings regarding this module:

1. [Open Science : One Term, Five Schools of Thought](#)
2. [How open science helps researchers succeed](#)
3. [Developing a Toolkit for Fostering Open Science Practices: Proceedings of a Workshop](#)

National Academies of Sciences, Engineering, and Medicine. 2021.

1. Reproducibility and Replicability in Science. Washington, DC: The National Academies Press. <https://doi.org/10.17226/25303> .
2. Open Science and Radical Solutions for Diversity, Equity and Quality in Research: A Literature Review of Different Research Schools, Philosophies and Frameworks and Their Potential Impact on Science and EducationGong, “Open Science.”<https://doi.org/10.1177/20966083221091867>
3. Book by Miedema, Open Science. <https://doi.org/10.1007/978-94-024-2115-6>

Further reading on terms and definitions:

1. Open Science glossary from the FORRT (Framework for Open and Reproducible Research Training) <https://forrt.org/glossary/open-science/>

Questions/Reflection:

* Questions for students of the course:

- How has research practice changed over the past few decades ?
- As a researcher how do different components of responsible Open Science transform knowledge contribution?
- We learned that there is “no one ethos” in this lesson. Can you explain what this means, and why?

Benefits and Challenges of responsible Open Science: Why does it matter?

Introduction

In the previous lesson, we learned about foundational concepts that define Open Science. In this lesson, we address some benefits and challenges of working in the open.

Here we aim to present a take on the development of science that's not only focused on scientific results but also on the process of creation, and the stakeholders that constitute the community.

Stakeholders can be individuals producing scientific knowledge (i.e., researchers themselves), individuals consuming, applying and regulating scientific research (i.e., practitioners, general public, policy-makers, organizations, communities, etc.), and the larger scientific ecosystem (i.e., scientific journals, repositories, archives, etc.). We discuss more about the people who perform and benefit from open science - and how to support them - in [Lesson 3](#).

In this lesson, we highlight the various benefits of open science across multiple stakeholders, providing some examples that can be explored further. Further, challenges in adopting open science practices are explored.

Benefits of Open Science

Quality of research

For researchers, a primary benefit of increased transparency and verifiability is that it allows readers and stakeholders to judge whether results presented are accurate (Chambers, & Tzavella, 2022) and, importantly, that the results are not produced by questionable research practices that lead to misleading or unreliable results (John et al., 2012). Open science practices assure that various statistical estimates of a study (e.g., p-values, effect sizes) can meaningfully be interpreted (Mayo, 2017; Cummings et al., 2016). And allows others to scrutinize the analytic decisions of the researchers, such as whether the analysis was planned before or after observing the data (Nosek et al., 2018). This allows others to check if they can arrive at the same conclusion as the original research team, and facilitates stronger public trust and support (UNESCO, 2021).

Real world implications of non-transparent science

The Free Software Foundation Europe (FSFE) provides a compelling position paper explaining why transparency is important for science. When computers are used to produce scientific research, the code is considered a “method”, much like in a lab research setting, a set of instructions for working with cells or agar plates might be a method. Peer-reviewed methods are an essential step in the scientific process. When these steps are not shared, no-one else can reproduce the work, or build upon it for future scientific endeavors. It also allows people to judge whether or not the methods are trustworthy.

In this case study, the FSFE reminds us of a time when closed methods were not trustworthy. Volkswagen revealed it intentionally programmed its diesel engines to cheat during laboratory emissions testing. This meant that people drove these cars thinking they were trustworthy and safer for the environment than they actually were. In this case, the real emissions from the engines were more than 40 times over the legal limit in the USA! Had the code for the diesel engines - the “scientific methods” - been open, it is possible that this untrustworthy behavior would have been picked up on much earlier. ([Gkotsopoulou et al., 2017](#)).

Quality and diversity of scholarly communications

Furthermore, open science improves the state of scientific literature. Scientific journals have traditionally faced the severe issue of publication bias, where journal articles overwhelmingly feature novel and positive results (Devito & Goldacre, 2018). This results in a state where scientific results in certain disciplines published scientific results may have a number of exaggerated effects, or even be “false positives” (wrongly claiming that an effect exists), making it difficult to evaluate the trustworthiness of published results (Simmons et al., 2011; Nissen et al., 2016). Open science practices such as registered reports mitigate publication bias, and improve the trustworthiness of the scientific literature. Registered reports are journal publication formats that peer-review and accept articles before data collection is undertaken, eliminating the pressure to distort results (Chambers, & Tzavella, 2022). Other open science practices, such as pre-registration also allows a partial look into projects that for various reasons (such as lack of funding, logistical issues or shifts in organizational priorities) have not been completed or disseminated (Evans et al., 2021) giving these projects a publicly available output that can help inform about the current state of the field.

Not everything should be pre-registered

Pre-registration is the practice of registering your scientific study/experiment plans before you start the study. This helps to ensure that the experiment isn’t changed part-way through if the results aren’t the conclusion the researchers had hoped for, and can help ensure publication of “null results” which otherwise might not be published.

Pre-registration is a good tool for hypothesis-driven science, when a researcher starts with a hypothesis, then proceeds to define steps (methods) to prove or disprove the hypothesis. Not all science is hypothesis-driven, though. Discovery driven science is more exploratory and doesn't usually start with a hypothesis. It may instead involve looking at existing data, or collecting more data, and trying to form conclusions based on the available evidence. Many domains perform discovery science, and generally these experiments and studies aren't suitable for pre-registration, since the exact direction of study may not be clear at the start of the research.

Open Science is also a valuable tool to be used in the public sector. Movements like Public Money Public Code were started by people who believe in the value of having open research and data freely available to the population. Remarkable advances on the way we exercise democracy are also being empowered by science made on the open, software like Polis which leverages the concepts of Computational Democracy, empowers scientists to run statistics and machine learning technologies on opinions of millions of citizens. In other words, open science facilitates citizen science .

Response to societal challenges

As science tackles consequential topics (climate change, pandemics and global health, democracy and misinformation), the transparency and verifiability of science is more important than ever. This is highlighted during the pandemic, where the creation of life-saving vaccines were spurred because the genomic sequence of SARS-CoV-2 was placed in GenBank, an open access database (Zastrow, 2020). Open science allows for rapid, global access and action especially for shared problems too difficult to solve by any one team alone.

Responsible Open Science is not only beneficial - it can also be characterized as an ethical imperative, especially for publicly funded projects. UNESCO (2021), for example, writes “so as to ensure the human right to share in scientific advancement and its benefits, member states should establish and facilitate mechanisms for collaborative open science and facilitate sharing of scientific knowledge while ensuring other rights are respected”

The recent years have shown the great momentum of open science, with a number of funders, regulatory organizations and governing bodies mandating open science practices across various disciplines across the globe (e.g. European Commission; UNESCO, 2021; National Academies of Sciences, Engineering, and Medicine. 2018), with more details about it in Lesson 4 . The practicing scientist of today and especially of the future needs to learn about open science and start applying it into everyday practice.

Less unnecessary repetition is better for study participants

Open science, in a way, also gives back to the communities that scientists hope to serve. Through open science practices, research waste can be avoided, such as unintentional and costly repetition of previous studies (Lusoli and Glenos 2020). In the human sciences, this also

reduces participant fatigue in the long term. By maximizing what is learned from publicly available data, one does not need to test repeatedly especially on already vulnerable communities. By “giving away” science, individuals, communities and organizations can more easily adopt research results to inform interventions for their own needs without the knowledge being gatekept by the original researchers and organizations involved. In this way, open science can facilitate strengthening the social and economic impacts of scientific results.

Personal/career benefits

Aside from accuracy, adhering to open science practices potentially offers personal career benefits to researchers themselves. Openly published research has a potential for greater visibility and impact by reaching larger audiences across the internet, leading to more citations and more like-minded collaborators and career/funding opportunities. (McKiernan et al., 2016).

Open science practices can also enable stronger collaborations, both within and between disciplines (Hormia-Poutanen, & Forsström, 2016). The ease of access to open data brings new agents to the landscape allowing for broader and more diverse participation. Through open science practices, such as pre-registration, one allows for a stronger research design because feedback from various collaborators and stakeholders can be solicited before data collection begins. Similarly, preprints allow for speedier feedback on conclusions drawn from the data once it is collected.

Case study of a successful collaboration:

Mozilla, an organization famous for the web browser Firefox, also runs a community-driven project called Common Voice. Common voice is an open crowd-sourced dataset of different voices and speech patterns, covering many different languages, accents, countries, and speech patterns. By making this data open and facilitating contributions from volunteers worldwide, speech recognition technology and text-to-speech technology is democratized and represents the members of the populace more equitably.

Practicing open science with transparency, collegiality, and research integrity do require development of a whole set of technical and transferable “soft” skills, which would be extremely useful for researchers in their careers both in academic or non-academic sector. Some examples include digital content creation; information, publication, data literacy; communication and collaboration skills - we will come back to it in the bonus section of lesson 5. Therefore, It is important to have the training and mentoring widely offered to the researchers.

Short on time? Make sure to read the top-ten reasons to do open science at the end of this lesson for a quick TL;DR summary.

Challenges in Open Science

However, open science also comes with its challenges. Doing open science requires some extra effort from researchers to start and maintain, but its long-term benefits include a great overall increase in research efficiency, integrity, and public trust. For example, putting your code in the open will probably mean that some adjustments must be made, and sharing it with a community will demand that you choose how your contributions can be used by others. Sharing data can imply extra work and planning; however this organization and widespread discovery can greatly improve science and confidence in it. We will see more details on code sharing and licensing in the “How” lesson 5.

In this lesson we focus on the challenges of your work, and the consequences of sharing - and in some cases, oversharing.

Not everything should be open - don't overshare without consent!

In order to practice responsible Open Science, careful attention should be given to how data is anonymized and how sensitive information is removed from it in order to safeguard people's identity and to prevent various harms stemming from breach of privacy. In recent history, we have seen many cases of how the misuse of data and illicit means to collect it is harmful to the population. Scandals like the Facebook–Cambridge Analytica, and outrageous services selling very personal parts of users' lives without their knowledge and full consent are far too common. Preparing documentation, using standards, and creating metadata takes time and effort

Additionally to treating users' data ethically, often further work is required to make research outputs not only publicly available but also understandable and accessible to various stakeholders. This means for example, that codes to be shared are understandable and properly documented. This might mean to have a testing system in place, make use of a distributed version control software and a CI/CD pipeline. If you're unfamiliar with any of these terms, don't worry! They will be covered in the “Open Software” module. (Maybe this last sentence is a cute little character with a balloon)

Besides caring about code, if the project utilizes data and that's being open sourced, it might be necessary to also have documentation that adequately describes the data set's contents, nature and layout. This type of “data about data” is known as “metadata”. It might also mean to tweak the formatting of the dataset to fit a specific pattern agreed by the broader community - this is known as using community-agreed data standards.

Open community members don't always agree with each other

Other than the more technical aspects of producing Open Science it's also important to keep in mind the societal aspects of the project. While interacting with the community can be one

of the most fulfilling things about Open Science, it might also be a source of disagreements about the direction of the project or how it should be used. That's where licenses and codes of conduct come into place. By explicitly setting out rules for the community interactions and use of resources, licenses and codes of conduct are useful to both protect the maintainers and their vision of what the original project and the forked projects should comply with.

Case scenarios in open communities

As you saw in the last lesson the story of Open Software (which builds the foundation for Open Science) is vast and at times different open values can conflict deeply. Two particularly relevant movements that helped to shape our ideas and actions in Open Science today are the Open Source and the Free Software movement.

The Open Source Initiative, an organization that advocates for Open Source, argues that Open Source code can't "discriminate against persons, groups, fields or endeavors", the Free Software movement affirms that "everyone should have the freedom to run the program as they wish, for any purpose". Even though these maxims might sound very encompassing and welcoming there are several critics of the carelessness that these movements have been treating both maintainers and users of Open Source, as well as their gullible negligence on how powerful a tool code is and how it can be used to do evil.

Speaking about doing evil, the Open Source Initiative addresses this problem with these exact words "Giving everyone freedom means giving evil people freedom, too". Recent movements like the Ethical Source and the First Do No Harm movements have been questioning the broadness of paradigms in which open resources are allowed to act, imposing ethical restrictions to the use of software through the use of licenses. There are also cases where the project maintainers took the lead and made their own licenses, such as for the data format JSON.

Examples of open science and open source that have been used for unintended purposes.

- ICE uses Chef-sugar (an open source project) [1] - an open source project being used by immigration enforcement authorities Illegal use of Elasticsearch branding by Amazon [2][3]
- All the "what's bad" essays on Stallman's website
- Data Sovereignty , indigenous rights, and parachute/helicopter research: when marginalized people share their data, sometimes privileged researchers re-use the data without fair credit or funding reaching the original data creators.

Further, science that is just "open" does not necessarily mean that it is of high quality. However, the transparency and verifiability that open science affords, makes readers and various stakeholders able to independently judge the trustworthiness of research products.

Cultural barriers: not everyone wants to change, and institutions often move slowly

A further challenge of adopting open science practices are institutional barriers to the researcher or practitioner. While one might be interested in adopting open science practices, they might lack support from their department or project supervisors and open science practices might not be given the budget, resources or time in a project cycle. Institutions might also not recognize open science practices in recruiting, training or promoting in the organization. These lack of incentives within organizations present difficult barriers to the adoption of open science.

While there are many challenges to the adoption of open science, we believe that its various benefits and its ethical imperative to the self and to the scientific communities, citizens and policy-makers outweighs the cost of barriers. In addition, recognising the barriers and places where caution needs to be taken provides a first step towards resolving them.

Summary

Open Science provides benefits not only to society but also to the individuals who perform it. Walking the line between responsible appropriate sharing and irresponsible oversharing requires diligence but the path and the results of science made in the open are very rewarding to all its stakeholders.

10 Reasons to practice open science responsibly:

responsible Open Science...

- ... (including availability of data, code, materials, and early results) accelerates research broadly and greatly.
- ... generates transparency and public trust and support
- ... fosters working across and engaging multiple disciplines, or “convergent” science.
- ... brings innovation through using big and aggregated data and information
- ... supports public and community uses of science: also known as community science, participatory science, or citizen science.
- ... helps fight misinformation and disinformation
- ... is intentionally and thoughtfully inclusive practice
- ... supports the key role of science in addressing major societal challenges in the 21st century (including climate change, sustainability)

- ... makes your research more efficient and impactful and provides credit broadly responsible Open Science is the new normal, and regulatory and governing bodies are reaching a consensus toward pushing it).

Questions/Reflection:

- Why are responsible Open Science practices important to a researcher's profile?
- How can a researcher benefit from responsible Open Science practices?
- How does society benefit from responsible Open Science?
- In this lesson, we learned that responsible Open Science often takes time and requires diligence and dedication of researchers. Can you explain how and why?

Stakeholders of Open Science: Who practices responsible Open Science and for whom?

Introduction

In previous lessons, we learned about the concept and motivation and aspiration of open science. Now let's think about "who" is practicing open science and for whom. In the first section of this lesson we dive deeper into understanding who the stakeholders for Open Science are. In the second part we cover essential topics about barriers to participation, and to include diverse stakeholders in open science communities and ways to overcome them.

In this module we offer you a person-centered approach to making open science happen. Our intention is to prevent harmful consequences of science's misuse (even unintentional misuses) and to increase the impact of science, by leveraging other researchers' works and improving society.

Who performs and benefits from open science? Stakeholders partaking in open science

As briefly discussed in previous lessons, Open science doesn't only concern researchers; many other stakeholders are affected by the outcomes of open science as well. Stakeholders are any individuals who can affect or be affected by open science projects. Although there are different ways to categorize stakeholders depending on your science projects, mainly there are three large groups; 1. Researchers, 2. Public, and 3. Policy-makers.

- Researchers
- Organizations
- Research Teams
- General public
- Decision Makers (regulatory, funding bodies, etc)
- Government

Researchers

Individuals engaged in creating new knowledge (e.g. researchers, students, faculty staff at universities, researcher centers, researchers at libraries). Responsible for creating an open science environment as well as open outputs and processes.

Public

Lay people who can drive/improve/conduct science (i.e. people who may not have an academic background or research experience). This may also be referred to as “citizen science”, but you do not have to be a citizen of any particular country in order to participate in science!

Policy-makers

Those with decision-making power (e.g. government, regulatory bodies)

How each group contributes to Open Science

Let’s take a look at these groups, how they can contribute to open science (input) and what benefits they experience from open science (this was also discussed in Lesson 2). Note that overlap among researchers, the general public, and policy-makers can happen.

Researchers’ contribution to open science manifests by sharing and communication their research via open access publications (more about it in the Lesson How and Module Open Results) As a result, community of researchers benefits from increased visibility and credit, reproducibility, access to more data and attraction of funding, reduced work’s duplication, conservation of resources and increased accessibility

The general public contributes to open science research by above mentioned “citizen science” projects, as e.g. as volunteers to collect or manage (e.g. categorize) some type of data.

As a result, individuals boost their understanding of science and feel empowered by having opportunities to exert influence. Disinformation in the public arena is decreased, and the routes of access to trustworthy sources of information are strengthened.

Policy-makers play important role in ensuring and facilitating open science by setting data management processes, open access legislation, developing ethical guidelines for experiments As a result, higher quality of research done with open science principles and efficient communication between stakeholders leads to better-informed decisions

This figure briefly shows how three groups of stakeholders interact with each other. Healthy interactions will foster respect and overcome power dynamics. Each group should focus on

empowering other groups and be aware that open science cannot exist without the others. Resources and tools for interactions are described in greater detail in Lesson 5, (and in the tools and results modules).

Case scenarios

Now let's take a look at examples of successful interactions around the world!

Case Scenario #1: Trend: Public —> Policy-makers

The public has many opportunities to join research projects and can play prominent roles in science. There are more than 30 ways to define Citizen Science (Haklay et al., 2021), and the principle is “active public involvement in scientific research” (Irwin, 2018). Citizen Science contributes to policy making at various stages of the policy cycle, including policy preparation, formulation, implementation, monitoring, and evaluation Scade et al (2021). That is to say, citizens are capable of setting trends and informing the directions in policy making.

In 2015, the United Nations adopted the 2030 Agenda for Sustainable Development for peace and prosperity for people and the planet, now and into the future (United Nations, 2021). This agenda has 17 specific goals that require a large amount of data. Citizens have been contributing by providing the water and air quality, marine litter, biodiversity, health, and gender issues data (Fritz et al, 2019), and Scade (2021) describe this as “a source of information for policy making.” This is a powerful example of citizens influencing global policy trends.

Case Scenario #2: Officialize: Policy-makers—> Researchers/Public

Policy-makers can implement new regulations for both researchers and the public. Bothwell and Smith (2017) reported that policy can shape knowledge. For example, policies such as dispersion of research funding (i.e. which science disciplines including Citizen Science receive the most funding), and data management plans for the public can impact the amount of knowledge produced.

Most importantly, policy-makers are mindful that researchers and citizen scientists conduct science projects safely and ethically. National Institute of Health (2022) states that policy sometimes sets the rules of the road for conducting research, helping ensure that scientific investigations are carried out safely, securely, adhering to the highest standards of research integrity, and in a way that addresses evolving ethical concerns. We can find these ethical policies, for example, NIH Guidelines for Human Stem Cell Research. Some countries have legislation requiring research to be published openly, such as Spain's open access legislature. Policies on open access for European countries are monitored and reported by corresponding OpenAire National Open Access Desks.

Case Scenario #3: Participate: Public —> Researchers

Currently, NASA has 28 Citizen Science projects that are open to people around the world (NASA, 2022). According to NASA Citizen Science policy, Citizen Science is defined as a form of open collaboration in which individuals or organizations participate voluntarily in the scientific process in various ways. The projects vary from Earth and planetary science to biological science such as researching meteorites, mosquitos, and the surface of Mars.

One of the evaluation criteria for NASA Citizen Science is; two-way communication between volunteers and NASA scientists and including diverse citizen scientists, with scientists giving feedback to and receiving feedback from the volunteers (NASA SMD Policy Document SPD-33, 2018). Also NASA creates opportunities for citizen scientists to be co-authors for publications and 191 NASA Citizen Scientists joined scientific publications since 2011 (NASA, 2022).

In addition to citizen science, there is an emerging concept called community science and co creation. Community science refers to science projects that honor community priorities. They can be initiated by a science practitioner or a community member, but they must become a collaborative endeavor (ASTC, 2021). Co-creation in science refers to the collaboration between a variety of actors (people from different societal roles) actively joining forces to tackle jointly defined challenges (Stier and Smit, 2021). We can also state that community science, which prioritizes community needs, succeeds through efforts of co-creation.

Charles et al (2020) introduced a successful case of community science. One example is protecting one of the remote islands in Canada that is facing the threats of sea level changes (e.g. salt water intrusion to groundwater and losing archaeological sites). As a result of community science and co-creation through public, universities, and policy makers, now climate-related mapping and visualization techniques for vulnerability assessments are available for use within the community. This provides opportunity for all the residents to explore adaptation options to ongoing sea level changes. The community was also able to work with archaeologists on preservation initiatives.

Case Scenario #4: Share: Researchers —>Policy-makers/Public

About 2,000 researchers work together to create a report for the Intergovernmental Panel on Climate Change on the current situation, which is a technical report that most people would have trouble understanding (Woolston, 2016). Some climate researchers break down their results to explain to policy-makers and citizens. Policy makers can utilize the results to officialize the restriction of CO2 emission level (e.g. Paris Agreement) and the public can be aware about what they can do in their daily lives to achieve the CO2 emission goal. This shows that each group is playing a significant role in addressing the climate science project, which is considered as one of the critical issues that our generation is facing.

How diverse stakeholders are included in open science:

Stakeholders are incredibly diverse in terms of culture, communication, and ability. To make science truly open, we must ensure that open science is accessible to everyone, so that we can all fully participate and benefit from the work. The best way to include stakeholders is to remove existing barriers and design for inclusion.

Creating a more inclusive environment will both increase the amount of people who feel welcomed to contribute back to your research and will broaden the scope of people that can comprehend and interact with the products of the research. Small actions towards conforming to accessibility and diversity guidelines will go a long way towards making your work truly open to all, maximize the visibility and impact of research..

Let's look at some factors and potential barriers for participation in the open science, with possible solutions:

Socioeconomic status:

Instabilities in the electric, electronic and internet access (e.g. load-shedding, internet speed, electronic device performance)

Possible solution(s): open science materials and communication channels should require less resources whenever possible

Neurodivergence:

Diversity of neural architecture leads to different learning and socialization styles

Possible solution(s): employ multimodal communication strategies using different visual and audio outputs, varied pace of events and conversations

Disability/impairments

- Sensory - e.g. colorblind, blind, deaf, auditory and/or visual processing conditions
- Physical - e.g. conditions that affect energy levels, neuromuscular coordination conditions
- Mental - conditions that affect mental health (e.g. depression, schizophrenia, etc)

Possible solution(s): employ multimodal strategies and universal design to provide proactive accommodations for as many as possible - captions, transcripts, colorblind-friendly palette, document formatting that are compatible with screen readers, normalizing flexible work schedules and rolling deadlines with collaboration

Intersecting Identities and intersectionality

Epistemic oppression - e.g. dominance of English as the international language for all science. Non-native English speakers are disadvantaged by default.

Possible solution(s): Proactive translation of open science results/communications in other languages

Microaggressions/macroaggressions:

Use of words with negative connotations towards individuals and groups and negative behavior/ostracization

Possible solution(s): Employ language and communication with neutral connotations that do not use pejorative terms or vilify a group (example) in biology, we use males to identify the parent with testes and the female with ovaries; should change language to “parents with testes/ovaries” etc) Gender Inclusive Biology for more detail

Caution: Full participation in open science requires respect of an individual’s identity, autonomy, and lived experiences. *Microaggressions, macroaggressions, and epistemic oppression are identity barriers to open science.*

This list is not comprehensive, but is meant to be a starting point in preparing your work in open science for diverse stakeholders.

Activity/exercise

Now let’s practice by looking at some typical case scenarios and solutions, reflecting on things you could do for inclusion:

Case Scenario #1: Accessible figures and writing

You have finished your project and are busy typing your paper to submit to an open science preprint journal. In your paper, you have several figures that use multiple colors at once - red, green, and blue. In addition, you have formatted your paper to use a serif font at size 10. **You want to make sure that your paper is easily readable for everyone. What are some things you can do?**

- Colorblind people have high difficulty with red, green, and blue colors. You can check your figures by running a color blind simulator, e.g. open source RGBind. Consider using colorblind-friendly palettes with colors such as green, magenta, and others. Avoid using color hues to convey information if at all possible.

- Legally blind and dyslexic people have difficulty with font size and font types. Consider using a font size of 12 or higher, and use a ‘Sans-Serif font’ such as Arial or Verdana to assist people with dyslexia in reading your manuscript.

Bonus question: What should you do if a journal insists on using a font size and font type that is inaccessible to some people?

Case Scenario #2: Organizing an inclusive physical event

You are the head organizer for an open source code hackathon for your organization. Your boss initially suggests using the large seminar room with one projector and screen that is hard to see from the back of the room. When starting the hackathon, you find out that a couple of attendees are deaf and a couple of other attendees have visual difficulties. What are some quick things you could do to help them fully participate in the hackathon?

- If doing a presentation on Powerpoint, you can turn on ‘Always use subtitles’ for live transcription. Check if your font size on your presentation is large enough to comfortably see at the back of the room.
- If possible, consider simulcasting the presentation on Zoom or other virtual platform with captions/transcription.
- Use text for communicating with deaf attendees.

Case Scenario #3: Organizing an inclusive virtual meeting and preparing in advance

You are organizing a virtual open science meeting with established and prospective members from different countries. You are unsure of what the prospective members need in order to participate fully, and no one emailed you to let you know about accommodations they need. What should you do?

Being proactive with small things you can do ahead of time by implementing some of the accommodations before the meeting (see possible solutions to barriers that we have just considered). While it is difficult to preconceive every possible accommodation that you might need to provide for your members, if you communicate your willingness to do everything you can to help members thrive, you are doing incredibly important work to not only recruit prospective members to open science, but also to retain them.

Bonus tip: Subtitles and closed captioning are not only for deaf/hard-of-hearing people, they are also very beneficial for non-native English speakers to understand the conversation fully. Consider using a third-party app such as otter.ai for accurate closed captioning and simultaneous transcripts that can be saved for members to read through.

What are some other accommodations that could be useful for everyone in general?

Summary

In the first part of this lesson, we learned about the types of stakeholders and how they can interact to empower each other. Successful examples were introduced, and you can reflect and analyze how to develop these interactions in your science projects. These arrangements may initially take time, but the outcome is essential to advance science, and is also rewarding.

In the second part of the lesson, we studied how diverse stakeholders can be included in open science with case scenarios designing for inclusion. Taking measures to maximize diversity, inclusion, and accessibility of your science project will enrich the project, boost its visibility and engagement of participants. Healthy interactions among stakeholders with diverse members creates the strength of science projects and rewarding results, and a diverse team drives innovation to success. Remember that what you learned here is not an optional choice but an integral part of responsible Open Science.

To learn more about joining, contributing to, and creating your own communities, consider visiting the Tools module.

Questions/Reflection:

- What steps can you take to make these open science resources more inclusive?
 - Written resources and images
 - Conferences - virtual, physical, or hybrid
- Communication with the general public and policy makers should not be something that researchers only do when they have spare time, after the research is done and published. It should be treated as a critical part of a science project, to certain extent at all stages of development. Explain multiple possible communication channels and strategies for researchers, and why each is important.

Impact of Open Science on academia, communities and society as a whole: Where open science happens.

Introduction

We have so far explored the fundamental parts of what Open Science is: why to pursue it and who the stakeholders of open research are. Where you are in the world when performing open science can have an impact on how you perform it, too. Laws across the world vary, and the advantage of open science means people from around the world can participate, co-create, and consume content together. This can affect your work from social and legal perspectives, and may present technical challenges as well.

Legal frameworks that affect responsible Open Science Open Science promises to make research work more accessible, all-encompassing, participatory, understandable and re-usable for wider audiences. Keep in mind, making the process open does not in itself result in wide participation unless it's partnered with sufficient financial resources, technological advancements, knowledge and skills. It's important that all these are available across regions, institutions and socio-demographics (review by Hellauer et al. 2022)

Data protection, privacy, and data sovereignty

Caution: To perform open science responsibly, it is important to consider not only what you should share, but also what not to share.

Individuals may have a right to privacy in their communications, for medical records, and for their physical locations. Similarly, certain countries, communities, and especially Indigenous peoples may historically have been exploited, and may wish to retain more rights over their knowledge to protect from further exploitation. Globally, there are laws around the world that may cover some of these issues, but not all countries and regions have equal levels of protection, and some have none at all.

We share some case studies:

European case: General Data Protection regulation

There are protective laws and legal frameworks in certain places around the globe that affect open science. European researchers have to abide by the General Data Protection regulation (GDPR) while making a data sharing statement stating the non-availability of data sharing. This hinders sharing particular data. Here, the scientific society should come forward to allow responsible Open Science data sharing possibilities in the global scientific space (Giske Ursin & Heidi Beate Bentzen, 2021)

South African case: Protection of Personal Information Act (POPI Act) and Open Science

The POPI Act No. 4 of 2013 is regulation by the government of South Africa to safeguard the personal information of South African citizens, like the General Data Protection Regulation (GDPR) in Europe. The regulation states that if one is obtaining personal information of South African citizens through phones, focus groups, interviews, containing identifiers such as names, contact information then you have to be POPI Act compliant.

In the research context, one needs to make sure that if the personal identifiers are collected then they must not be shared with third parties and stored securely in an access-controlled location to prevent a data breach. The act doesn't impede open data sharing, but personal identifiers should be removed from shared datasets. The POPI act affects the research process, in a way to make sure that storing of data of only de-identified datasets on cloud storage & onsite data storage is strictly controlled to specific designated individuals to ensure data safety (POPIA Code of Conduct for Research, 2021).

United States case:

In the United States, there is no federal-level legislation similar to POPI or GDPR, but there are some state-level laws, such as the California Privacy Rights Act, and the Virginia Consumer Data Protection Act.

Exercise: Check what laws, if any, apply in your state.

Summary: Working in a global society with varied data protection laws

Given the broad variation of data protection laws around the world, it may seem tricky to navigate. By practicing responsible Open Science, however, our response can get a little bit clearer. We can consider relevant legislation (if any) to be a bare minimum, and instead ensure that we are involving relevant stakeholders, as discussed in lesson 4, and listening to their needs respectfully, even if it means we are more cautious than local legislation may require.

Whose laws apply to my community?

Social, cultural, and legal norms will vary from country to country, and international communities. Avoiding culture clashes can be made more manageable by setting out explicit cultural norms for your community, such as may be specified in a code of conduct, which we discussed in lesson one of this module. Try to avoid assumptions that tie to a specific physical location or culture. Some examples why this is important:

- Laws are not uniform. If activity X is legal to do in one country, but not another, a code of conduct which says “obey the law” becomes impossible to interpret fairly or to enforce.
- Hosting a conference in a country that doesn’t have strong human rights records might result in someone breaking the law by being LGBTQIA+, or by not wearing religious garb.
- “We plan to release this in the summer” might be clear if you’re all in the same country, but if your collaboration is spread across the northern and southern hemisphere, is summer in the middle of the year or the end of the year? Consider using a month name instead - “we plan to release this by March” is unambiguous.

Equity and Open Science

Many countries in Asia, Africa and Latin America face many challenges, including lack of funding, inadequate access to literature and poor infrastructure. Across these regions, young scientists are working to build practices for open science from bottom-up. The aim is that scientific communities will incorporate these principles as they grow but these communities’ needs differ from those that are part of mature research systems.

The reasons for falling behind are lack of funding, poor infrastructure, inadequate access to research resources. There are government policies, which want greater productivity at the expense of quality. The open science collaborations can bridge the gap for developing countries by providing new ways and provide researchers access that might be currently out of reach (Onie, S. 2020).

Equitable terminology: what words should we use?

When talking about equity from a global perspective, it can be very hard to choose appropriate language, and historically many phrases have come and gone as we learn more equitable ways to communicate. Common phrases you may see include “Higher Income Country” and “Lower or Middle Income Country”. These are terms defined by the World Bank. Some people prefer to use “Global North” when referring to more privileged / high income countries, and “Global South” for lower income / more exploited and marginalized countries - but some “Global

South” countries are in the northern hemisphere, and vice versa! Other times, people use “minority” and “majority”, but again sometimes the phrase “minority” might be used for a populace that is not actually a minority! An older phrase is “first world country” or “third world country”. Many of these terms also have accidental or intentional negative connotations. For this module, we aim to use the phrases “marginalized” and “privileged” when referring to the inequitable distributions of resources and power amongst humanity.

The Global North have ascendancy over authorship and synergies in research networks, which margins out the Global South (Cash-Gibson L et al 2018).

In richer regions, a compulsion for the goal of excellence nurtures cumulative benefit in funding allocation for the highest funded institutions (Noble P et al) Across many countries, very few women have higher positions, senior positions are given at a later age, given less grant funding and few have high-impact publications (Gesiarz F et al 2020) (Brown JVE et al 2020) These are the impartialities, which are the societal imbalances (Zuckerman H. (1988). The above stated societal imbalances, which Open Science is focused to minimize in order to elevate the underrepresented societies, groups and create avenues for Global South countries to come forward & contribute to the global science community.

Prainsack & Lionello (2018) stated that open science is a political assignment greater than its technological part. The Open Science policy in Europe is shifting across nations, institutions & funding organizations. (Sveinsdottir T et al 2020). The emphasis on policies drive the incentive/reward structures and resource allocation and later helps in establishing strategies. Open Science started as a bottom-up approach by the researchers but has gone to the top-end level making it to the national and institutional policies setting wider goals like economic growth. The European Commission favors Open Science but in 2016 EU publication, the concern of Open Science perceived potential is being given that greater importance for fostering Europe’s competitive advantage in global markets (link to EU publication, 2016) Open Science positions to cover literature in languages other than English, supporting the value of bibliodiversity . We see a diverse set of communities in organizations working for Open Science data, software, tools, resources together as multilingual teams’ covering different languages of the world. Research indicates that there is a demand for regionally focused titles, in regional languages (Snijder 2022)..

A global perspective on open science

UNESCO on Open Science Infrastructure

UNESCO’s recommendation on Open Science states the potential of open science is in minimizing the present inequalities in Science, Technology and Innovation and pace towards SDGs 2030 implementation agenda, specifically in Africa, least developed countries, small island developing states and landlocked developing countries.

Open Science infrastructures are shared infrastructures (referred as virtual/physical, knowledge-based resources such as journals, collections, and open access publication platforms, archives, repositories, scientific data, present research informations systems, sets of instruments, open bibliometrics, scientometrics systems for assessing & analyzing scientific areas, open computational & data manipulation service infrastructures, multidisciplinary data analysis & digital infrastructures) where open science happens and serves the needs of diverse communities. Please see UNESCO Recommendation on Open Science

UNESCO on Open Science policies clearly recommends monitoring Open Science through combining qualitative and quantitative methods to assess the efficacy and efficiency of Open Science as per the member states' particular conditions, constitutional structures and constitutional provisions. Also, gathering & communicating progress, good practice, research work & innovation in open science and its outcomes with support of UNESCO and diverse stakeholders approach.

Organisation for Economic Co-operation and Development (OECD) and Open Science

The OECD's recommendation regarding research data from public funding helped gain collaboration and global sharing of data as a policy priority, with the objective of making the global science system more effective and seamless. There has been progression in a number of OECD member states and partner economics, with 58 countries successfully delineating their policies for open data & research publications. - For IT infrastructure, academic institutions and data repositories, international networks have been established in the form of repository networks such as OpenAIRE. - "Science clouds" - national and international computational resources - are being initiated such as European Open Science Cloud, the Australian cloud NECTAR, the National Research Data Infrastructure in Germany, the National Institute of Health Data Commons in the USA & Research Center for Open Science and Data Platform in Japan.

Questions/Reflection:

- What strengths do marginalized communities bring to open science? What challenges may they face compared to privileged communities?
- Name at least one data privacy law, and describe ways you can keep personal data safe. Do all countries have data privacy laws?
- Bonus: You're working on an open science consortium that gathers data in the Netherlands, Kenya, and India. You plan to use servers in the EU to store your data. What concerns should you take into account?

Not an afterthought

Previous lessons have shown the importance and benefits of open science, presented some key stakeholders involved, and discussed the barriers to participation and ways to overcome them. This lesson will guide you in how to start infusing responsible Open Science in your own work, which might be independent, or could be in a research group or lab.

Plan for open science into the design

Practicing responsible Open Science requires organizing your work and research, and your team, if you have one, around open science and planning for it from the inception, even designing the project with open science in mind. There are many resources and tools that make these easy, and indeed doing so will improve efficiency and the value and impact of your work, and help you focus on your research itself. We'll provide a brief overview in this lesson, but you may wish to explore the later modules in this course too, which cover Open Data, Open Results, Open Tools, and Open Software . Additional resources, and knowledge, may be available at your institution, including in your department or library or among your colleagues. An additional resource is a recent report from the U.S. National Academies “[Open Science by Design](#).”

It is important to discuss responsible Open Science with your research team, lab, group or partners regularly. Much of responsible Open Science may seem to be related to outputs – such as data, software, and publications – but preparing and organizing work for these in advance is critical. It would be hard or impossible to follow leading practices for these at the end of research, in the “afterthought” mode. Responsible Open Science is both a mindset and culture.

Planning for outputs in advance includes: speaking about it and organizing with your research team; deciding which tools to use; thinking about authorship and credit; engaging with relevant stakeholders and research partners, for example, industry, around open science; identifying repositories for software and data; highlighting these approaches in your grant; and much more.

Perks of digital and internet age for responsible Open Science:

The internet has made it very easy to share digital work. The popularization of Open Source computer code and the rise of Open Science has resulted in many outlets for public and free hosting of research and data. One key to open science, and why it is so empowering for 21st century science, is that we can now connect all the participants, stakeholders, and outputs of a research result together so that they are easy to discover.

Here we present a non-exhaustive list of digital platforms and tools used with for open science:

- Digital Persistent identifiers - for objects and researchers (such as doi and ORCID)
- [Open Journal System](#): open source software for managing & publishing scholarly journals
- Electronic notebooks such as [Jupyter](#) and [R Markdown](#)
- Data repositories: genetic sequence database [Genbank](#), protein data bank ([PDB](#)), Dataverse, figshare, Zenodo and for wide search use <https://www.re3data.org/> and/or <https://datacite.org/>
- Softwares/Codes: Zenodo used with Github / mybinder
- Materials: Addgene (for molecular biology)
- Reference management tools: Zotero, Mendeley
- Academic Social networks: Academia.edu, ResearchGate
- Peer Review: Publons, PreView
- Project management: Open Science framework
- Github as a platform for collaborative work on training materials etc

A variety of tools are emerging to help manage open science workflows, and to support global collaboration. These include spaces for project management, such as the Open Science Framework from [Center for Open Science](#), [electronic](#) notebooks which help projects organize data, software, and content together; online platforms for creating manuscripts, etc. More information about the open science collaboration and management tools are described in the Open Tools module .

Now let's move to the tools and procedures to ensure credit and attribution for our work, and allow its use and reuse in new, powerful ways, using the internet.

Digital persistent identifiers - for objects and researchers

A key to the interoperability is that each piece is assigned a “persistent identifier” and “metadata” that provides a secure path and basic information about it in such a way that they can be linked automatically (machine-readable).

How many times have you gone to an old link, only to find the page is no longer there? A persistent identifier is powerful because it is designed to point to the Web resource even if, or when, the URL or domain changes. One very common type of persistent identifier is a “digital object identifier” (DOI) that is usually assigned to a digital object (e.g. document) by publishers, preprint servers, data and software repositories. This has allowed automated linking of references across publications, including to citations after a publication.

Case scenarios:

1. A researcher writes a script in R that they use to analyze their results and produce a bar chart. They can upload their R code to a repository, and get a DOI for their script, so others can peer-review the code if they wish.
2. A member of the public attends a conference online and shares a digital poster and a short talk about their work as a citizen scientist. They deposit their poster and talk slides on to Zenodo, and can share the slides and poster using the DOI URL and receive credit for it.
3. A consortium member collaboratively authors a paper summarizing the results of a workshop they attended, alongside other workshop attendees. The journal they publish in automatically assigns a DOI to the paper.

ORCID: A permanent unique identifier for *you*, as a scientific author

Researchers and authors also have a digital identifier in this system: The Open Researcher and Contributor Identifier or ORCID. **Thus a first step to enabling responsible Open Science is to sign up for your identifier at [ORCID.org](<https://ORCID.org>).** This identifier will be included in your research outputs and work so that they can be linked uniquely to you (this can also happen automatically). You control your information on ORCID and what is public or private. Your ORCID can also be a way to get credit and recognition for reviews, awards, and more. Many funding agencies now integrate fully with ORCID, for example, for preparing grants and reference lists.

Other identifiers that are regularly used include those for funding agencies—which along with the grant ID provide a connecting link back to their repositories, institutions, samples, open reviews, and even annotations on web pages. Identifiers for research instruments, reagents, and materials are under development and implementation too.

In most cases, the identifiers will not be managed or assigned by you. Publishers and data repositories may ask you and your co-authors to link your ORCID and provide a grant ID (if you have one!) but they will then automatically provide the digital linking and create the metadata record. Often, you can sign on to repositories using your ORCID, so that this is automatically linked to any work you upload.

Having basic metadata - remember, metadata is documentation *about* your data - for each object with a persistent identifier helps *discoverability*. For publications and datasets, an identifier usually includes the title, authors (with their own identifiers), grants (with identifiers), journal (with its identifier as well, the ISSN), and publication date among other information. This allows search engines to discover and index the content. For data sets, leading repositories will also help ensure that information on standards, uncertainty, and calibration are included to allow appropriate reuse.

Collectively, this system allows widespread discovery and connection of the various pieces of research—even connecting open preprints and conference presentations to later versions and publications to data sets that underlie and support them.

Sharing data, and software, and getting cited: Repositories you can use

A key part of responsible Open Science, which is enabled by this system, is that research outputs – data sets, software, publications, conference reports, etc. – should go to the respective places that best manage, curate, and host that type of output. Previously, a data set may have been included as a supplement to a paper, usually a PDF file at a publisher’s site, or not included at all (“data not shown” or “data available upon request” statements were common even a few years ago but are thankfully waning).

Publishing a data set separately from your paper, at a repository that handles that type of data well (ideally a popular “domain repository”), allows others to cite your data separately, with its own metadata and authorship and expert curation of that data. Some also allow data that have appropriate restrictions on access (such as personal medical data) to be hosted in a secure way. This allows separate credit and authorship (if appropriate) for data or software products. A publication or research project may, and often will, have multiple data sets across several leading repositories.

In general, domain repositories are preferred over a general repository for data, because of the *expert curation* and better metadata they can provide, but not all disciplines or types of data have appropriate repositories. In this case general repositories can be used. In some cases, you may create and deposit data in a repository throughout a project; in other cases the natural time to “publish” the data is when a paper is submitted to a journal.

See the [Open data module](#) for more information on sharing your data appropriately.

Open software is usually developed in a collaborative workspace such as GitHub. Github works with a general repository, Zenodo, to enable software versions to be assigned an identifier and metadata.

Sharing data, codes and software is a key for ensuring reproducibility of findings, improvement of code and software, for enabling other researchers to easily re-use , extend and cite that work (Gorgolewski & Poldrack, 2016). Sharing the data & materials is also a signal of valuing transparency and trust in their own research, boosts authors' visibility and recognition (McKiernan et al., 2016).

See the Open source module for more information on sharing your code and software appropriately.

Collectively, this set of identifiers, metadata, and infrastructure helps enable content and especially research data to be “*findable, accessible, interoperable, and reusable*” or **FAIR**. This is a key concept and part of responsible Open Science. For researchers it means directing research outputs to their best open science home and planning for this throughout the research process. For all these reasons, it is best to think about how to share data and software supporting a publication *before submission*. More about FAIR principles can be found in the Open Data module , and now we will consider some foundational principles on licensing the content for reuse.

As a general rule, when you create something - a blog post, a scientific paper, a drawing, a data set, computer code, or any other ‘creative’ work - you automatically own the copyright for that work yourself. This means that others **aren’t allowed to re-use it** without your permission, even if it’s freely available on the internet. As an open scientist, you can use a **license** to grant others permission to re-use your work, and even specify conditions - perhaps you always want others to credit your work, or perhaps you don’t want your work to be used commercially.

Caution: When you perform work for someone else, as an employee, contractor, volunteer, or a student, your contract may stipulate that the copyright for that work belongs to the institute you are working for. Before assigning a license to your work, check that you have the right to do so. Your employer, institution’s intellectual property office, and your funder may all have specific expectations around how you share your work and what license you use.

Distributing content to the best open science home also allows each output to have the right license that allows access and reuse. Usually you would include the type of license in the metadata. When you publish a preprint or publication, or deposit a dataset, or software version at a repository, you will usually be asked about the correct license to assign to that content. Usually a repository will recommend or require a specific license to enable broad reuse. The basic standard is that leading licenses support reuse generally with attribution (citation) so that the creators of the content are recognized. Citations are supported by leading publishers.

Here we list a general guide on the best licenses for published content, data, and software:

Written content of any kind, papers, posters, slides, images, audio files, videos, other creative works

[Creative Commons licenses](#) are designed to allow re-use of these types of content. Authors can choose to require **credit** for their work (CC-BY attribution), allow or disallow **commercial use** and/or **derivative works**, and to require [reciprocal sharing](#) of works. Open Results Module for more information

Data

Including spreadsheets/csv/text files with experiment results, videos/audio files/images created from a study, databases of computationally processed data.

[Creative Commons Public Domain \(CC0\) licenses](#) are often best for data. Whilst you may be tempted to use a creative commons attribution license (CC-BY), this can make it difficult for people who wish to reuse or integrate different data sources in the future. Visit the Open Data Module for more information .

Computer code, such as scripts written in R, Python, Matlab, SPSS

[The Open Source Initiative](#) has a set of licenses designed specifically for code projects, that covers both open distribution of the code itself, as well as executable versions of the program that non-programmers can run. Visit the Open Source Module for more information

Other items: Whilst this is beyond the scope of the module, this list is not exhaustive. Other types of work may require different license types. For example, what license would you use for an open hardware drone design, a 3d-printed microscope kit, or a reagent used in a laboratory? These items may have different constraints and needs.

Caution: As a general rule, if an item does not include a license for reuse, it's illegal to reuse even if you can see the work online. Licenses are designed to take into account the legal ins-and-outs that each type of work can encounter. Try not to use one license type for a different output - Creative Commons specifically advises [not to use their licenses for computer code](#), for example.

Making your work useful to others:

Sharing and publishing your manuscript:

Public repository/Preprints

Sharing drafts of research as preprints can improve citations and help establish or provide credit and a reference months before formal publication in the journals (McKiernan et al 2016). A manuscript posted by author(s) to a repository for facilitating open sharing of early work without any limitations to access is *Preprint* (Puebla et al 2022). Basic screening is carried out to the manuscript, which is usually posted on the preprint server within a few days of submission without peer review and is freely accessible online. More than 1200 journals now allow posting of preprints, and some connect directly to deposit submitted manuscripts directly (see directory here: <https://v2.sherpa.ac.uk/romeo/>) or allow transfer from the server to the journal. Many funding agencies now allow citations of preprints in grants. Many preprint repositories are field-specific; see a directory here: <https://asapbio.org/preprint-servers>. Many also will link to the published version of the manuscript once it is available.

In addition, many institutions have open sharing repositories, and mandates to share author-versions of published manuscripts. Many funders also have repositories for sharing manuscripts after publication or a means to connect to manuscripts on publisher platforms. Check with your institution or funder, and journal publisher for requirements.

Some preprint repositories also accept conference presentations (e.g. posters and slide decks for talks).

Publishing Open Science and Open Access

When you publish your work in peer-reviewed journals, traditional publishing models may result in papers that are not openly available for anyone to access, and instead may require a subscription fee. Publishing in subscription journals is usually without cost, but where possible we recommend using “Open Access” publications. There are many routes to open access publishing, discussed further in the tools module, often with different trade-offs - for example, if a scientist has to pay to publish in an open access journal, fewer scientists can afford to share their work!

Discipline- and sector-specific nuances

The above information applies across nearly all scholarly disciplines. There are some discipline or research specific responsible Open Science steps that also apply or for which you should think and learn about:

For some fields of research, pre-registration of hypotheses –as a publication–is strongly encouraged and it helps avoid bias and supports the publication of negative results. These are becoming common in behavioral and social sciences and clinical trials. We explored this previously in lesson 2.

If your research involves partnering with *industry* where some outcomes may be restricted from publication, it is best to discuss and reach agreement in advance on responsible Open Science outputs to avoid complications or misunderstanding at the time of publication. If you have one, consult your institutional legal, knowledge transfer or intellectual support office resources should be consulted to be clear that publication of relevant data and software are acknowledged and supported and that data can be placed in appropriate repositories.

Working with physical samples and tools:

Some disciplines also require or encourage sharing of physical materials such as reagents, cell lines, animal models, and materials and have repositories for these, which will also provide appropriate licenses.

If you are collecting or analyzing physical or biological samples, a permanent identifier system has been developed to help you manage your work, support open science, and enable standard methods for identifying, citing, and locating physical samples and comparing analyses from different labs–the [IGSN](#) or International Generic Sample Number. Identifiers can be reserved in advance (before collecting). Additional information is in the Data module

Some disciplines in paleontology and anthropology also require and expect open archiving of precious samples in public museums and/or other means to provide open access (digital casts).

If your research involves field work and sample collection, appropriate permits should be obtained including engagement with local authorities and stakeholders—including them openly in your research has many benefits, as we discussed in Lesson 3.

Work on human data and samples, and other sensitive areas often requires initial external ethical review, e.g. in the US by an Institutional Review Board (IRB).

Authorship: recognizing the contributions and giving credit

Working in collaboration and with the teams of researchers oftentimes led to sour disputes on the order of the authors at the stage of the final publication. It is important to remember that practicing open science responsibly implies giving the credit to the contributors in an equitable, fair and ethical way. Here, open science calls upon the knowledge and practices of research integrity and ethics. Separate, unique citations for a variety of outputs (such as data sets, software), expanded and refined contributors roles, such as [CRediT taxonomy](#), are crucial in recognition and giving the credit. Importantly, all involved stakeholders such as

publishers, funders, regulatory bodies, and research institutions need to consider recognizing diverse contributions and especially [open data in their evaluation systems](#).

For in-depth, check the [COPE's materials](#) on authorship and contributorship.

Summary: think beforehand, design for open science, never as an afterthought.

It is important to think about, discuss, and plan for desired outcomes and processes when you begin your research. Learn about where the best repositories are for your data; discuss credit and authorship for each separate open science output, and start using open science tools to organize your work. Reach out to repositories in your discipline and institution (usually library) for help. Indeed this information in your grants and data management plans will make you more likely to receive funding.

Bonus section: Open Science Skills

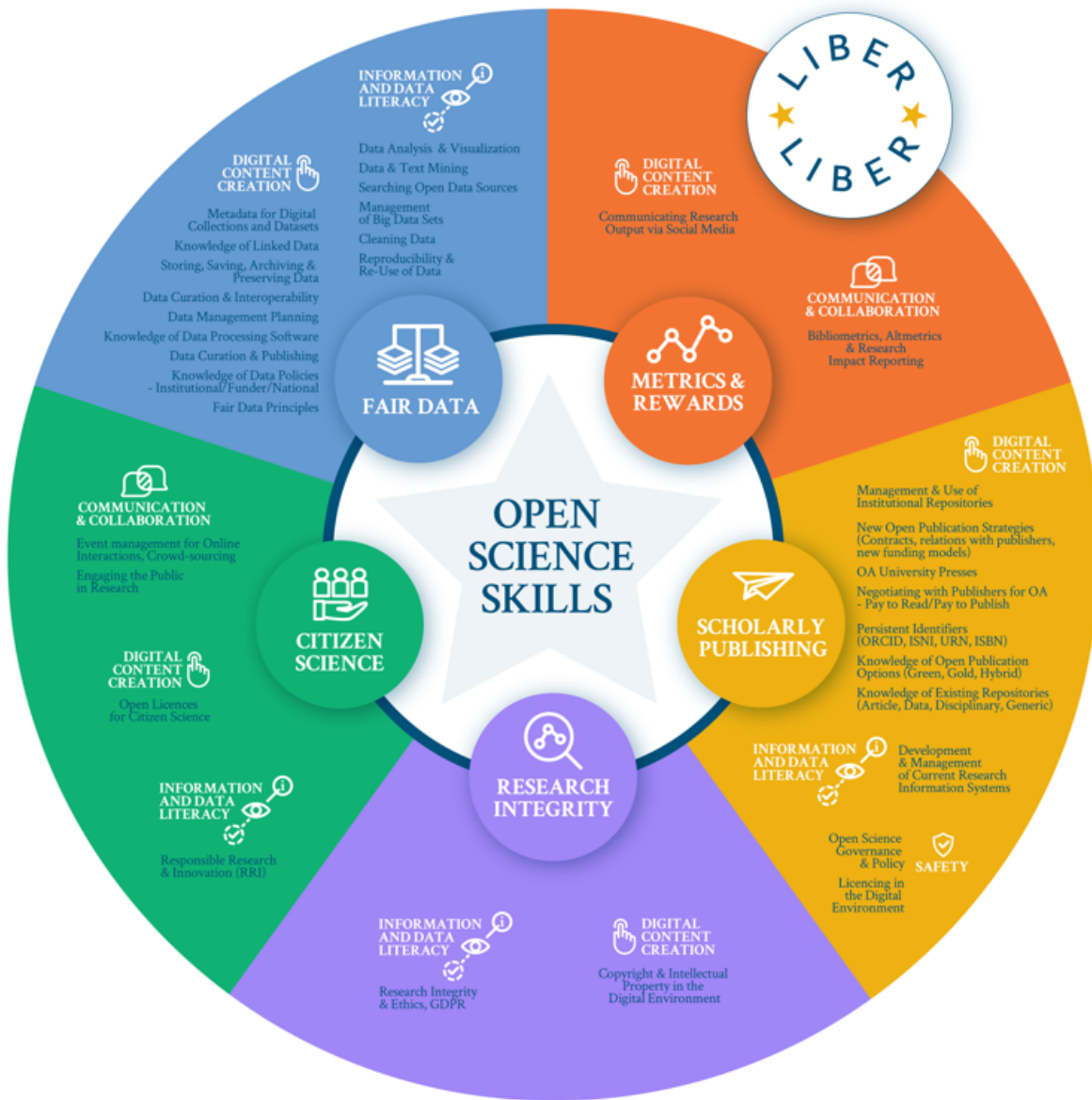
Open science can foster a range of skills, across many domains - the figure below touches on some of the topics you may have learned about so far.

You may have expected to see technical/data science skills, digital content creation, research management, library and information, and publication literacy. Research integrity and ethics are often less straightforward sets of skills, in which researchers may not be trained or aware of.

Each step of responsible Open Science involves considering and thinking about other people - collaborators, contributors, users and consumers of the research outputs. Therefore, communication and interpersonal skills, especially applied to virtual environments, are key.

In this module, we emphasized ethos and ethics of open science, inclusion and accessibility in participation of open science stakeholders, equity (or lack of it) in conducting and sharing science openly. responsible Open Science researchers of the 21st century need to develop their [reflective practice](#), just like practitioners, to enable them to become aware of their own stance towards science or assumptions regarding other stakeholders, aware of the values and worldviews, and provide means to adapt the responsible Open Science practice. ([Roedema et al 2022](#))

[Source](#) of the visual



- Discipline-specific skills needed to practice open science (does not include generic computer skills, wider librarianship skills and personal competencies)
 - Mapped to LIBER OS Roadmap 7 focus areas, Digcomp 2.0 framework and FOSTER learning resources
 - Produced by the LIBER Working Group on Digital Skills for Library Staff & Researchers with input from other LIBER Working Groups, 2020

Figure 1: Open science skills diagram - visit source below to see in detail

Summary of the module

This module provided a broad overview of the *ethos* of responsible Open Science, the imperative for scientific and societal challenges and opportunities in the 21st century, and an introduction to how you and your research team can begin to follow leading practices to enable open science. Part of the ethos is to help enable these practices within your team and with your colleagues—that is, you are encouraged and empowered to share what you have learned and help them learn about, be aware of, and practice Responsible Open Source.

In a larger context, many of you will be participating in scholarly efforts, including in peer review, in leadership positions as journals and societies, in organizing meeting sessions, and more. Enabling responsible Open Science is a broader cultural shift in scholarly practices worldwide. In many ways, the recognition, reward and award systems in science are not fully aligned yet with responsible Open Science as this cultural shift is ongoing. You are encouraged to leverage this learning in having conversations to develop this culture broadly.

Here are the six key guidelines to start practicing and supporting open science responsibly:

1. Plan for responsible Open Science from the beginning and begin discussions in your group, with colleagues, and your librarian.
2. Plan for making data and code open and available in leading repositories and citing it in publications in the reference section. Cite others' data and software that you make use of.
3. Learn about and adopt open science tools
4. Develop and foster inclusive workgroups, meeting sessions, and meetings.
5. Learn the routes to make your publications open and what your institution supports and funders require; preprints provide an easy and robust route
6. Support and inform your colleagues.

Questions/Reflection:

- How can a researcher publish in an open access journal ?
- Predatory journals are very harmful and some early career researchers may not be even familiar with these journals. Describe why they should be not included in responsible Open Science. Also discuss what are the possible ways to restrict and control these journals?
- Why are licenses an integral part of Open Science practice ?
- Can you briefly describe the differences between licence types? When conflicts arise among co-authors about which types of licenses they should choose, how do you discuss and resolve the issue using your knowledge learned from this lesson?
- What are two types of permanent identifiers, and why are they useful?

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Open Software

Have you ever marveled at mesmerizing scientific visualizations and wondered how they were generated and whether you can recreate them or even maybe tweak them to produce new results? These types of images have been created by researchers using **research software**. These software products and sometimes their **source codes** are freely available to the public. Reproducing such results and using them to advance the knowledge produced by these types of research software products are among the pillars of open science. For example, Figure 1, is generated using [E3SM](#), an Earth System model, the source code of which is available on [GitHub](#).

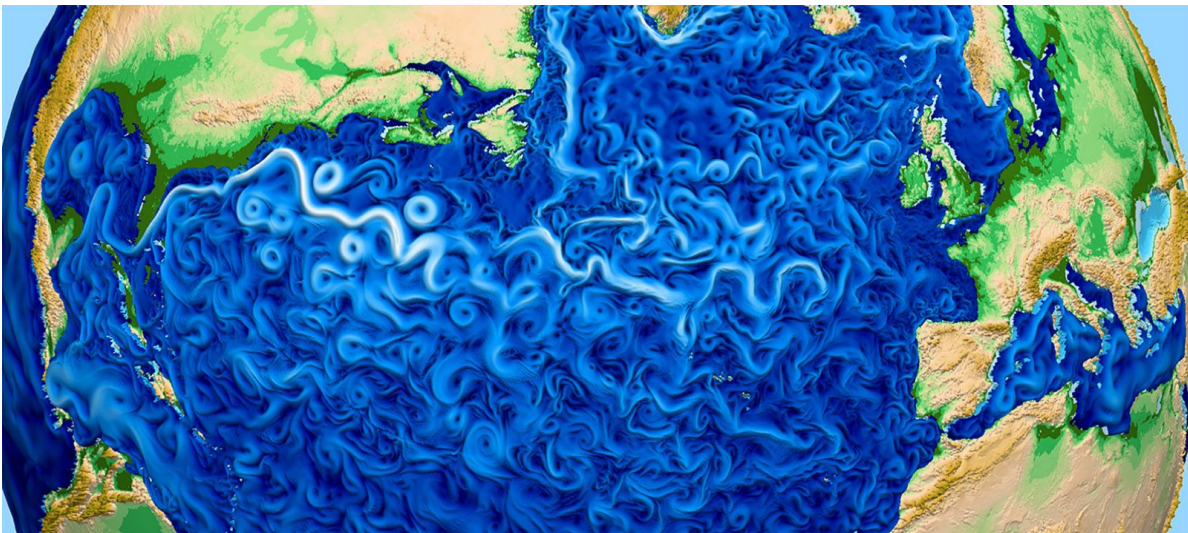


Figure 1. Global E3SM simulation showing eddy activity, credits M. Petersen, P. Wolfram and T. Ringler

Now, let's say that you are intrigued by the idea of recreating Figure 1 and tweaking the E3SM's source code. We should start with obtaining the source code. Someone might ask since this project already has a fancy website why is the source code on GitHub? Let's assume that we successfully got the source code and want to start recreating the figure. Naturally, the next question is how do we install it since there is no executable file in the source code? Maybe you are used to installing software packages using [installation wizards](#), or maybe you are comfortable working from [command line](#). Which one is possible or preferable for installing this software? The next step after installation is running the software and visualizing the results. So, the question is, for generating the desired outputs, how do we configure the software, what are the required input data, and how do we get them? Let's take it a step further and say that you have some brilliant new ideas and want to implement in the source code, analyze the outputs, publish the results, and make your code publicly available. Therefore, the questions become: How do we facilitate navigating this seemingly complicated source code? After making modifications, are we allowed to share and republish the modified source code,

and if so, how do go about it? How do we ensure that the republished code is findable and other researchers can reuse and build upon it?

The purpose of this module is to answer these questions, provide guidance for streamlining the workflow and ensuring that we give/get proper credits, and last but not least, draw your attention to and promote the importance of contributing and giving back to the Open Science community.

Open software in the context of Open Science

Learning objective:

- Understanding core principals of Open Software
- Learning Open Software terminologies

Introduction

The software that is created through/during research can be an important research product in and of itself. Open science principles like reproducibility, reusability, and replicability are especially important when it comes to research software. Within this module we will use the terms software and code interchangeably. We use these terms refer to any product written in a programming language, and can cover anything from a short script to a full software package with a full graphical interface.

Open Science Principles: How they relate to software/code

Reproducing findings of a published study is imperative for the scientific community. Therefore, results that are produced by a scientific software should be **reproducible**, *i.e.*, users should be able to obtain > “consistent results using the same input data; computational steps, methods, and code; and conditions of analysis” ¹.

If software/code used to make a figure or generate results is not shared along with the results/figures themselves, then it would take significant time, effort, and likely funding, for another researcher to reproduce those same results and determine they were correct.

We all aim to make significant contributions to our field and can do this by “standing on the shoulders of giants” (Isaac Newton). By sharing the code, trust in the work can increase, and

¹National Academies of Sciences, Engineering, and Medicine; Policy and Global Affairs; Committee on Science, Engineering, Medicine, and Public Policy; Board on Research Data and Information; Division on Engineering and Physical Sciences; Committee on Applied and Theoretical Statistics; Board on Mathematical Sciences and Analytics; Division on Earth and Life Studies; Nuclear and Radiation Studies Board; Division of Behavioral and Social Sciences and Education; Committee on National Statistics; Board on Behavioral, Cognitive, and Sensory Sciences; Committee on Reproducibility and Replicability in Science. Washington (DC): National Academies Press (US); 2019 May 7.

future work can build on it without duplicating effort. Therefore, it is important for a research software to be developed in such a way that it can be understood, modified, built upon, or incorporated into other software. This is called **reusability**.²

Another important aspect of scientific studies is **replicability**, *i.e.*, studies answering the same scientific questions - but using independent data and/or methods - should find consistent results^{3 4}.

Many communities already have a strong replication tradition, where trust in any scientific result is built when multiple codes achieve results that demonstrate consistent behaviors. By requiring multiple codes to achieve the same scientific finding, replication reduces the impact of individual code errors or numerical issues.⁵

Open Software and Open as a Spectrum

As we've said, sharing code can increase trust and lead to better science by allowing a more thorough review process. However, the degree to which a code is shared and when that code is shared can vary. Any sharing is a step on the spectrum of what we will refer to as **open software**, the most open of these equates to what is known in the computer science and software development industry as **open source software**. Open software can be a spectrum that can be anything from sharing an executable of a code with a description of how it was used to developing the software in a public repository from the start of the project. There are also a variety of license choices that can be made under the umbrella of open software which can allow the developer/researcher to retain various levels of ownership and rights to future commercialization.

Now, let's take a step back and give formal definitions for some of the terms that we just used.

²Chue Hong, Neil P., Katz, Daniel S., Barker, Michelle, Lamprecht, Anna-Lena, Martinez, Carlos, Psomopoulos, Fotis E., Harrow, Jen, Castro, Leyla Jael, Gruenpeter, Morane, Martinez, Paula Andrea, Honeyman, Tom, Struck, Alessandra, Lee, Allen, Loewe, Axel, van Werkhoven, Ben, Jones, Catherine, Garijo, Daniel, Plomp, Esther, Genova, Francoise, ... RDA FAIR4RS WG. (2022). FAIR Principles for Research Software (FAIR4RS Principles) (1.0). <https://doi.org/10.15497/RDA00068>

³National Academies of Sciences, Engineering, and Medicine; Policy and Global Affairs; Committee on Science, Engineering, Medicine, and Public Policy; Board on Research Data and Information; Division on Engineering and Physical Sciences; Committee on Applied and Theoretical Statistics; Board on Mathematical Sciences and Analytics; Division on Earth and Life Studies; Nuclear and Radiation Studies Board; Division of Behavioral and Social Sciences and Education; Committee on National Statistics; Board on Behavioral, Cognitive, and Sensory Sciences; Committee on Reproducibility and Replicability in Science. Washington (DC): National Academies Press (US); 2019 May 7.

⁴National Academies of Sciences, Engineering, and Medicine 2018. Open Source Software Policy Options for NASA Earth and Space Sciences. Washington, DC: The National Academies Press. <https://doi.org/10.17226/25217>.

⁵National Academies of Sciences, Engineering, and Medicine 2018. Open Source Software Policy Options for NASA Earth and Space Sciences. Washington, DC: The National Academies Press. <https://doi.org/10.17226/25217>.

Source Code Source code is a human-readable (vs. machine-readable) text written in a specific programming language. The goal of the source code is to set exact rules and specifications for the computer that can be translated into the machine's language. ⁶

Open Source Software (OSS) An Open Source Software is distributed with its source code without additional cost that makes it available for use, modification, and distribution with its original rights and permissions. ⁷

We should note that researchers are not always able to share their complete code, or software package (*e.g.*, due to national security concerns, data privacy, institutional policies). Again, open software doesn't necessarily mean open-source software and sharing to the level that is allowed by funding agencies, institutions, and security requirements is still a step in the right direction towards a world with more open science.

From [Openscapes](#):

Open is a spectrum – what you share, who you share it with, or how you share it. It's not all-or-nothing. What: slides, tweets, blogs, forums, wikis... then also code, data, protocols Who: your self, research group, project team, institution...then also public How: internal servers, Dropbox ... then also Google Drive, GitHub, data repos

We might also add here:

When: at the start of your project, when it reaches its first fully usable version, at the end during publication, etc.

Before jumping into the next lessons, let's have a brief overview of the core principals of open source software in general and, more importantly, in the context of research software.

Core Principals of Open Source Software: What research software can move towards

In the previous section, we provided a formal definition for open software and open source software. For better understanding, let's define what these concepts are juxtaposed against: **Closed Source Software**

Closed Source Software (CSS) Closed source software is a proprietary software that its source code is not distributed to the public. Therefore, only the original authors who created the code exclusively have rights to legally copy, modify, update, and edit the source code. Closed software imposes restrictions on what the end user can do with

⁶[Ionos](#)

⁷[Synopsis](#)

the application, preventing users from modifying, sharing, copying, or republishing the source code. ⁸

The major differences between CSS and OSS products are two-fold: End-users cannot modify CSS products and although, OSS products might have some restrictions on redistribution, CSS products usually are more restrictive on their terms of usage and redistribution. We can think of OSS as a form of thinking based on intellectual freedom that follows three core principles: transparency, participation, and collaboration. ⁹

Transparency Operating in such a way that it is easy for others to see what actions are performed and implies openness, communication, and accountability. ¹⁰

Participation Actively giving back and contributing to OSS through either committing time and lending skills, or monetary sponsorship. ¹¹

Collaboration Collective engagement toward making improvements and advancements through knowledge sharing and creating an inclusive environment. ¹²

The exchange of ideas and software developed by communities has driven creative, scientific, and technological advancement in nearly every aspect of our lives. Developers share insights, ideas, and code to create innovative software solutions both collectively and individually. Open source software operates with the underlying principles of peer production and mass collaboration, creating more sustainable software development for end users. ¹³

Not only users can make any kind of changes to the source code, but they can repurpose it into other new software and distribute their own software. However, there are some nuances on redistribution that we will cover in [Lesson 3](#).

Open source software is also sometimes conflated with the free software movement. Usually, “free software” is meant to emphasize freedom in the rights of end-users, but can sometimes be confused as meaning “free of cost”. In actuality, neither free software nor open source software denote anything about cost—both kinds of software can be legally sold or given away. Free software and open source software share common values, and the terms are sometimes combined in the popular phrase “free and open source software” (FOSS). ¹⁴

To support adapting OSS principals (transparency, participation, and collaboration), several new concepts have been introduced by the open source community. These are especially useful in the move to open science and has produced tools and methodologies that can be used to make research software more open:

⁸[IBM](#)

⁹[IBM](#)

¹⁰[Wiki-Branching \(version control\)](#)

¹¹[OpenSource](#)

¹²[OpenSource](#)

¹³[IBM](#)

¹⁴[RedHat](#)

- To facilitate sharing and community engagement a central file location storage is needed for source codes which is called a **Code Repository**. Some examples of such repositories are [GitHub](#), [GitLab](#), and [Bitbucket](#). Although, source code sharing and community engagement are their most basic capabilities, they go much beyond that and provide a wide range of tools for code *testing* and *version control*. Code testing in general refers to the process of evaluating and verifying that a software product does what it is supposed to do. The benefits of testing include preventing bugs, reducing development costs, and improving performance ¹⁵. There are various types of tests with different objectives that will be covered in more details in [Lesson 5](#). Version control is the practice of tracking and managing changes to source code over time. It keeps track of every modification to the code in a special kind of database. If a mistake is made, developers can turn back the clock and compare earlier versions of the code to help fix the mistake while minimizing disruption to all team members [Lesson 5](#). ¹⁶
- In addition to sharing the source code, software executables require a storage location to facilitate *software packaging* (for developers) and installation process (for end-users). These types of storage locations are called **Software Repositories**. These repositories are usually programming language dependent, for example, [PyPi](#) and [Conda](#) for Python-based software, [CRAN](#) for R-based software, and [Julia Packages](#) for Julia-based software. However, software packaging cannot always be done using automated services such as PyPi due to complexities of the source code structure itself (*e.g.*, intricacies of the software objectives, use of several programming languages, etc.) and/or its dependencies (other software packages that it depends on). In these situations, *containerization* is a viable option. [Docker](#) and [Apptainer](#) are example services for containerization.

Summary

Here we introduced the concept of open software, how it relates to the broader open science principles, and how sharing and openness can be a spectrum. At the most open end of this spectrum is what the computer science/software development community refers to as open source software. The core principles of open source software are introduced as a paradigm towards which research software can move towards. The tools and methodologies developed by the open source community are particularly helpful in opening research software. Next, we'll dive into the benefits and hurdles associated with having open software.

¹⁵[IBM software testing](#)

¹⁶[Atlassian](#)

The Pros and Cons of Open Software

Learning objective:

- Benefits of Open Software for developers and users
- Understanding the responsibilities of developers and user for a thriving Open Software culture

Introduction

This lesson addresses particular benefits of open-software, presenting how you as a researcher can benefit from it, and also how can it improve your research, moving yourself and your teams towards Open Science. We will also address some common challenges - and misconceptions - of adopting open software, and how to overcome them.

Benefits of open software

Open software offers a multitude of advantages to both developers and users. There are several benefits of open software are highlighted in this section.

As a developer/provider

- **High Visibility:** Publishing open software enables the repository to be more reachable and attainable. It can broaden the audience from a diverse group and draw more attention to the software repository.
- **Long-term Sustainability:** Subsequently, open software allows more people to access the repository and can cultivate more users to be involved in its development. It results in the long-term sustainability of the software. ¹ Since it is unlikely to have perfect software, having a larger user base is likely to have more collaboration or feature requests that can directly contribute to some improvements in the software. “Given enough eyeballs, all bugs are shallow.” ² Testing out software with a large base of users can easily detect

¹Forking: the Invisible Hand of Sustainability in Open Source Software

²Linus' Law

the issues in the software, and they can submit bug reports or submit proposed fixes directly.

- **Quality Improvement:** Besides bug fixes, the contributions can also be in feature enhancement, such as submitting additional features to the software repository or proposing modified codes that increase the effectiveness of the software. As a result, open software that comes with community support will tend to have continuous improvement, unlocking the potential to create new inventions, and produce better quality software versions. By ensuring the quality of the open software, it can gain users' trust to rely on it rather than redeveloping a software, therefore, minimizes the duplication of efforts, both within an organization and across organizations, by allowing for individual components to be shared.
- **Future Employability:** As a developer or maintainer of open source software, your skills and experience are an important asset to improve your chances of getting a job. ³ Experience in developing open software is a positive portrayal of the abilities as it helps in demonstrating technical abilities. In addition, it also demonstrates the personality and work ethic in software development. If someone has experience working on complex software development and maintenance, it can make the profile outstanding, especially to companies that will take into account the contributions of the candidate to open software. The hiring manager may also view the product or shared code. Hence, open source provides visibility into both how a candidate solves problems, and how they collaborate in a team.

As a user ⁴

- **Accessibility:** Shared code certainly increases the democratization of science, it promotes more diverse and inclusive community to use the open software without a cost-prohibitive barrier.
- **Flexibility:** Open software provides users a certain freedom to utilize the software for any purposes as they wish. It also allows users to make changes freely on the software and customize it according to their needs or even redistribute the software based on the license that has been applied.
- **Knowledge Sharing:** Open software is also a great learning opportunity for the community ⁵, it can help to achieve knowledge sharing through the community, which in turn, increases motivation for a continued practice.

³Categorizing the Content of GitHub README Files

⁴Open Source Software (OSS) Quality Assurance: A Survey Paper

⁵Synopsys

Are there any disadvantages of open software - and if so, how to mitigate them?

Making a software open source and valuable to the community requires additional efforts and considerations. In this section, we will discuss responsibilities that come with this decision and provide you with guidance for maximizing the impacts of your efforts.

As a user

Require a skill set

Open software comes in many forms and shapes. There are open-source codes that come as packages available in a repository for a programming language or environment (*e.g.* PyPi for Python, CRAN for R, Conda for a variety of languages). Others are code that require installation from scratch. Even for skilled programmers, this setup can incur in costs (time and financial).

So, if you are familiar with a programming language that offer repositories which are easy to download from within your environment (*e.g.*, R), you can start from there, and build up your confidence and skills.

To compile and generate an executable code from a repository from scratch, you will need to be able to check for the necessary computation environment, check and install dependencies, and compile the code. Programming language might be a barrier, as well as operating within a command line environment. The good news is that there are many resources to help you go through these stages. Widely used open software are usually well documented, with step-by-step instructions, and some even have a community which can offer support for installation and running their code. Sometimes, developers share alongside their open-source code an executable version for your operating system. *E.g.*, the repository of [Stock Synthesis](#)⁶, a software used for stock assessment of fisheries populations, offers both the source code and compiled versions for different operational systems. So these are good choices for a beginner.

Bear repeating that while learning these skills incur a cost, by doing so you might not only gain access to a useful research tool, but might also gain experience and skills that are useful for your career.

⁶Methot Jr, R. D., & Wetzel, C. R. (2013). Stock synthesis: a biological and statistical framework for fish stock assessment and fishery management. *Fisheries Research*, 142, 86-99. <https://doi.org/10.1016/j.fishres.2012.10.012>

Depreciation

Technology changes fast, and software - open and closed - becomes depreciated. If you rely on a certain open-source tool for your work, you run the risk of it becoming depreciated. It can happen to projects that are not maintained, or no longer maintained, for a number of reasons.

If this happens to you code you use, you can offer the developer to be a contributor to their open-software and update the code yourself. This will require programming skills, but it is a viable route. You can also team up with other users for a group effort.

If you are choosing a tool and are not interested to fix depreciation issues in the future, aim for widely-used community open software, which are maintained by numerous people and thus, less likely to be depreciated.

Security concerns

Open-software can be perceived as to present more vulnerabilities than proprietary software - when all software can present vulnerabilities. You should check if your institution has an open software security policy in place - if so, follow their guidelines to assure compliance and up-to-date security protocols ⁷. To minimize security risks, we also encourage you to download code/software from an authoritative source - such as the original project repository - rather than a third party site.

However, an important benefit of open source is that you can see exactly what the code is doing and know what are the dependencies, what is useful if any of them becomes vulnerable. You don't have the same level of transparency with a closed-source code. Open source codes also might have (some or many) eyes on them, which can result in better oversight. Widely-used open software will have a community of researchers and developers working on its code, looking closely at inputs, outputs and computer performance. But always, check with your institution about their requirements, guidelines and policies regarding open-source software.

As a developer/provider

Open Software can require extra work

Some extra work might be required to share code that is already written to improve readability (*e.g.*, comments, variable names, indentation) and documentation (*e.g.*, README and code of conduct files) of your work, so others can easily understand it and use it. However - and we cannot stress this enough - open software is a journey, not a destination. How much to change and add is totally up to you. The important part is to publicly share your code.

⁷[Linux foundation](#)

By writing code that is easily readable by humans, you can make it more usable even to yourself! It will save you time when you want to re-use it years later. Moreover, the more upfront effort you put into developing an accessible code, the more others will be able to use it - which might lead to more collaborations, better feedback, and career opportunities.

There is also a time commitment for basic steps of creating documentation, choosing a license, getting a DOI. Our module gives you an understanding of these terms, providing you a checklist with clear steps to sharing your code. We also point you to resources to make this process smooth and save you time in decision-making.

After sharing your code in a repository, you will have a reliable backup that won't depend on your own hard drive - and you have many free options to choose from! Added benefits are that by creating a license, you are allowing others to use your work on the terms you will choose. By having a [DOI](#), your code is a findable (by online search engines) and [citable](#) reference, and you thus, you will get credit for your work! You can also learn more about DOIs in the lesson about [Licenses](#).

Becoming a [maintainer](#)

Maintaining an open software (particularly open-source) long-term can bring its special sets of challenges - from the time commitment, to the procurement of funding, to navigating requests from users. Maintaining your code after sharing it is a personal choice, and you can step out of this role at any time you chose (more about this in [Lesson 5](#)).

Sustainability

Despite the importance of open-software for researchers, support and incentive for open-software development and maintenance are frequently inadequate ^{8 9 10 1112 13}. As reported by the Australian Data Commons (2022):

⁸ARDC Ltd. (2022). A National Agenda for Research Software. Viewed online at: <https://doi.org/10.5281/zenodo.6378082>

⁹NAA, N. A. of A. (2021). Current state assessment | naa.gov.au. September, 1–127. <https://www.naa.gov.au/information-management/building-interoperability/interoperability-development-phases/current-state-assessment>

¹⁰Akhmerov, A., Cruz, M., Drost, N., Hof, C. H. J., Knapen, T., Kuzak, M., Martinez-Ortiz, C., Turkyilmaz-van der Velden, Y., & van Werkhoven, B. (2020). Raising the profile of research software: Recommendations for funding agencies and research institutions in the Netherlands. Zenodo.

¹¹Katz, D.S., Druskat, S., Haines, R., Jay, C. and Struck, A., 2019. The State of Sustainable Research Software: Learning from the Workshop on Sustainable Software for Science: Practice and Experiences (WSSSPE5.1). Journal of Open Research Software, 7(1), p.11. DOI: <http://doi.org/10.5334/jors.242>

¹²National Academy of Sciences (2018). In Open Source Software Policy Options for NASA Earth and Space Sciences. <https://doi.org/10.17226/25217>

¹³Mangul S, Mosqueiro T, Abdill RJ, Duong D, Mitchell K, et al. (2019) Challenges and recommendations to improve the installability and archival stability of omics computational tools. PLOS Biology 17(6): e3000333. <https://doi.org/10.1371/journal.pbio.3000333>

Software is an often invisible part of research, produced quickly within a funding window, often struggling to be maintained beyond that. ¹⁴

Contributions to open software within traditional academia don't carry the same weight as publications - software is often seen as a by-product of research, and dedicated funding is unusual ^{15 16 17 18 19 20}. As reflected by reports and analyses from several countries, a shift in paradigms of funding and career advancement are required, along with an increase in software literacy, so open-software can be more sustainable.

While this is a larger, structural issue that cannot be easily overcome by an individual, we have strength in numbers. More researchers in the open-source community, will result in more visibility of these issues, both for our institutions and funding entities. As more researchers move towards an open, collaborative framework of science, it is expected that more changes will happen to the current paradigm, allowing a fruitful future for open-software.

Summary

In this module, you reviewed particular benefits of open software to improve: 1) visibility of your work, 2) Long-term Sustainability, 3) Quality of your software, and 4) your career prospectus. You also could explore how open-software furthers the open-science principles, increasing 1) accessibility, 2) freedom, and 3) democratization of science.

Despite its multiple benefits, adopting and creating open-software also brings challenges. In this module, we addressed some common challenges, with some tips to overcome - perceived and real - barriers to open-software.

Lastly, we want to emphasize that adopting open-software (as a user or as a developer) on your research is a journey. As with the practice of open-science, there is a spectrum, and you

¹⁴ARDC Ltd. (2022). A National Agenda for Research Software. Viewed online at: <https://doi.org/10.5281/zenodo.6378082>

¹⁵ARDC Ltd. (2022). A National Agenda for Research Software. Viewed online at: <https://doi.org/10.5281/zenodo.6378082>

¹⁶NAA, N. A. of A. (2021). Current state assessment | naa.gov.au. September, 1–127. <https://www.naa.gov.au/information-management/building-interoperability/interoperability-development-phases/current-state-assessment>

¹⁷Akhmerov, A., Cruz, M., Drost, N., Hof, C. H. J., Knapen, T., Kuzak, M., Martinez-Ortiz, C., Turkyilmaz-van der Velden, Y., & van Werkhoven, B. (2020). Raising the profile of research software: Recommendations for funding agencies and research institutions in the Netherlands. Zenodo.

¹⁸Katz, D.S., Druskat, S., Haines, R., Jay, C. and Struck, A., 2019. The State of Sustainable Research Software: Learning from the Workshop on Sustainable Software for Science: Practice and Experiences (WSSSPE5.1). Journal of Open Research Software, 7(1), p.11. DOI: <http://doi.org/10.5334/jors.242>

¹⁹National Academy of Sciences (2018). In Open Source Software Policy Options for NASA Earth and Space Sciences. <https://doi.org/10.17226/25217>

²⁰Mangul S, Mosqueiro T, Abdill RJ, Duong D, Mitchell K, et al. (2019) Challenges and recommendations to improve the installability and archival stability of omics computational tools. PLOS Biology 17(6): e3000333. <https://doi.org/10.1371/journal.pbio.3000333>

make your own choices of how, what and when you are able to share, given your personal skill set, institutional policies, time and funding limitations. The most important is to take the first steps, and continue this journey together with the open-source community.

References

Licensing, Ownership & DOIs

Learning objective:

- Understanding ethical and legal aspect of giving credits and attributions
- Learning about existing open source licenses and Digital Object Identifier

Disclaimer: the contents of this lesson are for educational purposes only. They do not constitute legal advice and should not be used as such.

Introduction

After deepening your understanding of the reasons to use open-software in the context of open-science, we here address the first considerations when using an open software tool in your research. First and foremost, if you are going to be building your own code on prior work you need to choose a software that is **open**, *i.e.*, that you are allowed to use, modify and redistribute. As a developer, you also need to ensure that you are sharing a product that is open - and thus, usable - to others. This is presented on the section **Licenses**.

Then, we present how you get credit for your work, and how you give credit to others' work. This is the content of the section *Attribution and citation*.

At the end of this lesson, you will be able to: *Choose and abide by appropriate usage and referencing standards of open-software.*

Licenses

A software license is a legal document that grants users particular rights to the use of a certain software. This license can take many forms, but in many cases they outline contractual obligations (if any exist) between the company/software developer behind the software and the end user, what the user can do with the software, who the user can distribute the software to (if any such distribution rights exist) and the length of time the user has the right to use the software.

A user cannot (technically and ethically), use a software without a license! A user can reach out to the developer/owner to ask for permission, and go ahead *if* the owner/developer furnishes

written permission. But, if you share your software without a license, no one can use it without your written permission!

Types of licenses

A license can fall under several categories. License types have general definitions of what can be done with the software. By picking a type of license, or by understanding what type the license of a software you're considering using is, you'll be able to navigate the license process more quickly than reading each license individually and interpreting the permissions. An overview of types of licenses is given in the table below¹:

Public domain license	Lesser general domain license	Permissive	Copyleft	Proprietary
Anyone can use or modify the software.	Can link to open source libraries and code can be licensed under any license type.	Has some requirements for distribution and modification.	Licensed code can be distributed or modified if all the code involved is licensed under the same license.	Software cannot be copied, modified or distributed

¹[licensetypes](#)

Table 1. Summary of select attributes of cited licenses types.

	Name	Latest Version	Copyright	Patent Grant ^a	Permits ^b Code Linking	Used by ^c
FOSS	BSD	2-Clause	No	No	Yes	Gabedit, Chemkit, Sci
	MIT	1.0	No	No	Yes	Weblogo, APBS
	ECL	2.0	No	Yes	Yes	RCrane, Sakai Project
	Apache	2.0	No	Yes	Yes	Imagemagick, Autod
	MPL	2.0	Partial	Yes	Yes	Firefox, Thunderbird
	LGPL	3.0	Weak	Yes	Yes	ClustalW/X, IMP, BioJ
	GPL	3.0	Strong	Yes	No	R Project, Perl, Coot,
Proprietary	Traditional “bespoke” ^d		No	Varies	Varies	Majority of scientist-c
	“Inspection only” ^e		No	Varies	Varies	Satisfies minimum pu
	Commercial		No	No	No	MS Windows, iTunes,
Hybrid	Any combination		Varies	Varies	Varies	Pymol, MySQL, BDB,

Note that the values assigned in the table are only a general summary of each license attribute and may not

^aLicense text explicitly describes the treatment of patents related to the software.

^bAllows the linking of computer code under different licenses.

^cSelect examples of popular software employing these licenses.

^dRefers to a range of custom-tailored licenses traditionally used by academic and research institutions.

^eTraditional “bespoke” license that also makes source code available for inspection purposes only.

doi:10.1371/journal.pcbi.1002598.t001

Summary of selected attributes of licenses types ²:

Some of the common licenses used in open software are:

1. MIT license
2. Apache License 2.0
3. Mozilla Public License 2.0
4. BSD 3-Clause “New” license
5. GNU General Public License (GPL)
6. Common Development and Distribution License

For more information on different types of licenses please refer to the ([Open Source Initiative OSI](#)).

How to choose a license

There are a number of steps that have to be made before choosing a particular license. Arguably one of the first decisions to be made is based upon whether you intend to use the code for commercial purposes or not, or at least foresee it as a possibility in the future. Some licenses are more favorable for commercial purposes than others, such as the *General Public License, version 2*.

The next decision that has to be made is relating to the issue of distribution. When using other software as a dependency, you should always be wary of their licenses. Some licenses enforce certain types of licenses upon redistribution. The GNU GPL, for instance, is incompatible with proprietary licenses, because it requires the combined work to be licensed under the

²Morin, A., Urban, J., & Sliz, P. (2012). A quick guide to software licensing for the scientist-programmer. PLoS Computational Biology, 8(7). <https://doi.org/10.1371/journal.pcbi.1002598>

GPL, with no additional restrictions allowed. Having a part of the work under a proprietary license is such an additional restriction, so you cannot distribute such a combination (unless the copyright owner of the GPL code gives special permission).³

For licensing open software, it is always good practice to consult with the [Open Source Initiative \(OSI\)](#) website. They provide a list of approved licenses that guides you through this process. Remember that a first step is always to consult with your institution (if applicable). You should ensure that you are complying with any applicable local laws and any policies set by your employer and/or funding entity.

Additional Resources

- [Choosing a License](#)
- [Turing Way on licensing](#)

Attribution and citation [Katz]⁴⁵

Both when choosing a license and publishing your software for future citation, a decision has to be made in relation to the issue of *attribution*, *i.e.*, crediting a person or group of people or other entity with a particular action in relation to the software. This can be thought of as the software/code equivalent to authorship on an academic paper. It is important to consider this to avoid accusations of plagiarism or copyright infringement. There is a short discussion in the final lesson in this module regarding ethical considerations on how contributions can be considered for authorship/attribution/ownership.

When deciding to cite a software or code that was used in your research you can start with the question: is this research software? Research software includes source code files, algorithms, scripts, computational workflows, and executables that were created during the research process or for a research purpose. Software components (e.g., operating systems, libraries, dependencies, packages, scripts, etc.) that are used for research but were not created during or with a clear research intent should not be cited (e.g. Microsoft Word, Linux, Python –the language itself; specific packages might be citable in this context). This differentiation may vary between disciplines. Some examples of research software that would be cited are [E3SM](#), [SciML](#), and [Stock Synthesis](#).

³[Turing Way Compatibility](#)

⁴Smith et al. (2016). Software Citation Principles. PeerJ Comput. Sci., DOI 10.7717/peerj-cs.86

⁵Chue Hong, Neil P., Katz, Daniel S., Barker, Michelle, Lamprecht, Anna-Lena, Martinez, Carlos, Psomopoulos, Fotis E., Harrow, Jen, Castro, Leyla Jael, Gruenpeter, Morane, Martinez, Paula Andrea, Honeyman, Tom, Struck, Alessandra, Lee, Allen, Loewe, Axel, van Werkhoven, Ben, Jones, Catherine, Garijo, Daniel, Plomp, Esther, Genova, Francoise, ... RDA FAIR4RS WG. (2022). FAIR Principles for Research Software (FAIR4RS Principles) (1.0). <https://doi.org/10.15497/RDA00068>

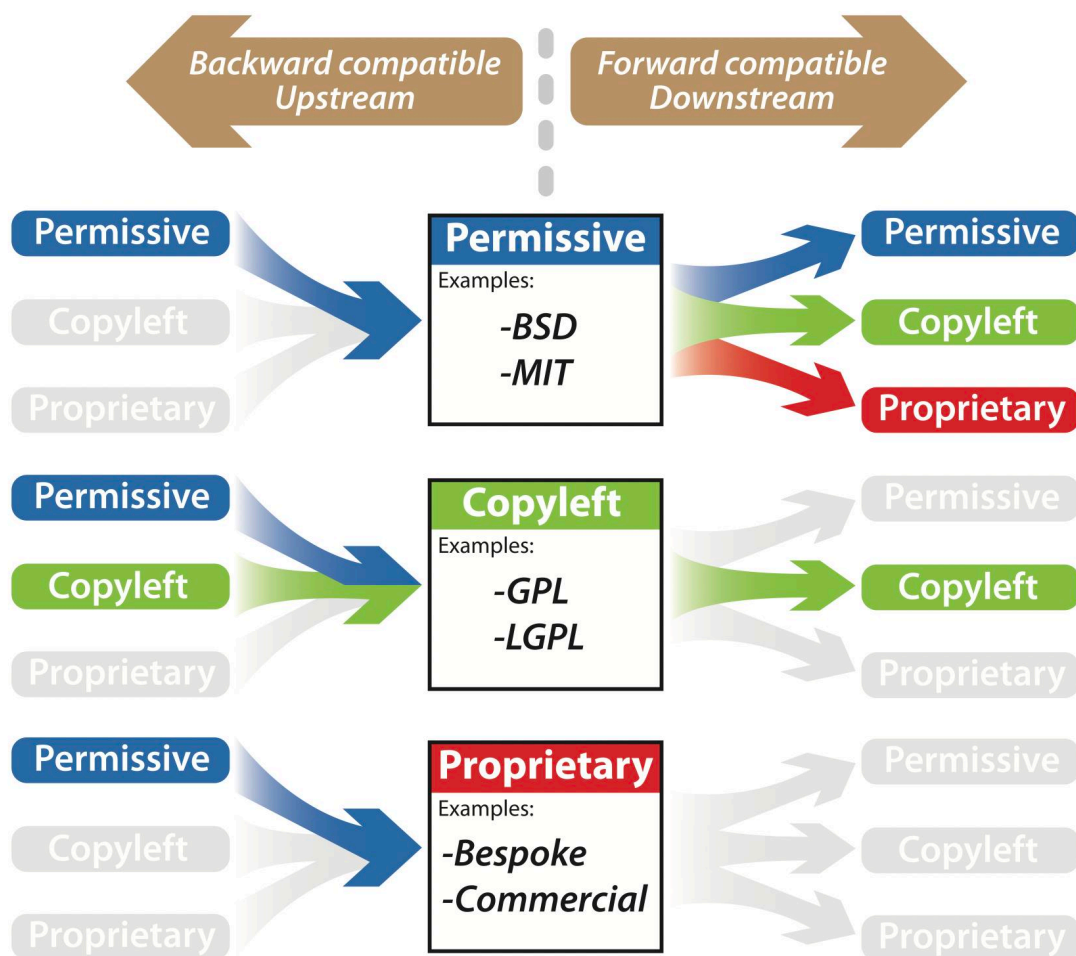


Figure 2. Schematic representation of license directionality. In general, permissively licensed code is forward compatible with any other license type. However, only permissive licenses, such as the BSD and MIT, can feed into other permissive licenses. Restrictive licenses like the GPL are backward compatible with themselves and permissive licenses, but must adopt the restrictive license from then on. Proprietary licenses can incorporate upstream permissively licensed code, but by definition are incompatible with any other downstream license. Grey represents actions that are not permitted without negotiating a separate license agreement with the rights owner.
doi:10.1371/journal.pcbi.1002598.g002

Figure 1: schematic

The majority of open source software licenses require some degree of attribution, and a small minority (such as the 0BSD) do not. The license will also dictate where the attribution must be displayed - some licenses will require the user to include attribution in a dedicated file such as the software license agreement.

Digital Object Identifier (DOI)

By having a persistent identifier, a software version can be cited. A digital object identifier (DOI) is a persistent identifier or handle used to uniquely identify various objects. It is provided and standardized by the International Organization for Standardization (ISO). In contrast to dynamic web addresses such as URLs, DOIs are static, *i.e.*, do not change over the life of a document, and point to the location of the document on the internet. You can get a DOI for your own software/code by adding it to a preservation repository.

Just as we publish scientific findings in writing in journals to ensure its preservation over time, its supplementary material, *e.g.*, source code and produced data, should also be stored in a permanent location. We call these preservation repositories. Some of these repositories are general-purpose such as [Zenodo](#) and [Figshare](#), and some are more research field-oriented such as [Hydroshare](#).

It is important to keep in mind that a DOI refers to a static version of your code and so, you'll need to get a new DOI for each version you release and want cited. By using the same repository each time you need a DOI for a new version, you can be sure that when a user looks for your DOI they are directed towards the most recent version.

Citing code without a DOI

As a user, if you'd like to cite a software that does not have a DOI you can use [Software Heritage](#) to create a SWH-ID which is also a citable persistent identifier, but can be created for codes that are not your own. This should only be done *after* ensuring a DOI is not available otherwise there can be multiple identifiers being used for the same piece of software.

Attribution for pieces/snippets of code

While DOI and SWH-ID allow citations of full pieces of software/code, there is also the case where a small code snippet or section might be copied into another code. It is common practice to take a few lines of code to solve a piece of a problem from websites such as [StackExchange](#) or [Code Ranch](#), but there should still be an attribution if no changes are being made. This can be done effectively with a comment that includes a link to the webpage from which the code was taken (most sites have an option to create a shortened shareable link that is more code friendly).

Publishing open software in peer-reviewed journals

It is also possible to publish open software or a research article detailing the inner workings of that software in peer review journals. A general example is the [Journal of Open Source Software](#); there are also more discipline specific journals such as [Astronomy and Computing](#) and [Environmental Modeling & Software](#). The peer review process for some of these journals may include a review of the code itself, some may be focused just on the describing journal article that accompanies it. These publications will also come with a permanent identifier as is customary with most journal articles.

External Requirements

There are various legal considerations to keep in mind with regard to software and code you write. For example, these may be considered intellectual property, and you may wonder who has ownership over it. Generally speaking much of this is dependent on your employment and funding situation at the time you did the work. Your institution may have claim to part or all of the work product, however it is highly variable, and your institutional offices should be contacted to understand this better.

There may also be other institutional, governmental, or other legal policies that may be dependent on your region. Please make sure you understand your locality's laws and regulations regarding the sharing of research software and follow your institution and funding agency's requirements (if any) on licensing and intellectual property.

Additional Resources

Code publication Computational Infrastructure for Geodynamics - example of a preservation repository that provides peer review *When to cite software* CiteAs: a resource for finding the correct attribution of a research product

Summary

Here, we reviewed that a software license is a legal binding document, made between the developer and the user of a software, which outlines how that software can be used and distributed. Open software will carry licenses that follow the [Open Software Initiative \(OSI\)](#) definitions: allowing the software to “to be freely used, modified, and shared.” ([Open Software Initiative OSI](#))

We have presented the major categories of licenses that might fall under the *open-source* definition, and what considerations to take when choosing a specific software and/or license

for your software, *i.e.* 1) what is the intended use of this software?, 2) how others can reuse it? And what are the policies of my institution and local laws regarding open-software use and dissemination?

We have also learned about proper attribution - how to get credit for your work (DOI, archival of code, publishing options), and how to cite others work.

References

Code management/Quality

Learning objective:

- Understanding some best practices for publishing Open Software
- Learning the basics of code management

Introduction

While we maintain that sharing software at all is a great initial first step regardless of it's state, the more the code is kept clean, maintained, and documented, the more others will be able to cite, use, and contribute to it.

What does it mean for software/code to be of good quality?

There are two perspectives that you can take when engaging with this lesson: a user of open software, or a developer/provider of open software. As a user, you will want to make sure that a code or software project you are considering using in your research/project is quality. As a developer/provider, you will want to make sure your project is of high enough quality that others will want to use and engage with it. When we say “quality” code, we are referring to precisely that, a software/code that a user can be confident in using.

Here we outline some baseline expectations for open software. While there are definitely good open software projects out there that do not include all of these items (and, unfortunately, plenty of projects out there that contain many of these items but still don't function well), this guide will assist in ensuring the software/code that you develop/use is quality.

Good documentation

Good documentation for code is possibly the most important item on this list for creating a quality code. This will help a user know what the software does and how it can be used, but also can be a real time saver for a developer when going back to look at code they haven't looked at in a while.

The README file

The first stop for a user when they approach a new project should be the README file. Aptly named, this file should contain orientation information that will help a user understand the project's purpose as well as shows examples of how it can be used, and lists most other important information the creator deems necessary. Note that there is no one agreed upon convention for the location of these documentation pieces, so we encourage exploration of the software you're interested in. Some information we describe as in a README file may be moved into its own file in some conventions, e.g. having installation instructions in an INSTALL file, but the README is still usually the best place to start. Keeping that in mind, if you are developing a code/software for use by others, they will expect a descriptive and useful README, without one using your code may be a nonstarter for many.

[Here is an example of a README file](#) from a NASA-funded project that shows many of the specifics we are going to discuss below including multiple installation options. As you read the suggested parts of documentation below feel free to reference this for an example.

Let's dive into the specifics of information you should include/find in a README file. First, a description of what the software does: it's purpose, the problem it's solving. You don't need to write a whole academic paper here, a sentence or two is fine. If you do happen to have a research paper written on the topic no one would be upset if you link it here, though do be careful that any linked papers are either (a) not behind a paywall or (b) if it *is* behind a paywall, that the important information a user would need to use and understand your software is reiterated separately within the code documentation.

A compatibility description is also necessary. Sometimes this is wrapped into the installation instructions and that is acceptable. Here the operating systems (e.g. Linux, Windows, macOS – and their versions) that the software/code works on with are listed. If the code runs in a browser which does it work with? There are many tools for testing the compatibility of code across operating systems and environments, we won't get into those here as they can be specific to the coding language you're working in.

If installation instructions are not in their own file, they'll live inside the README. These should be written with very little prior knowledge expected of the user. Most people are used to downloading a software package, double-clicking on the executable, and having a setup wizard walk them through any required steps. Setups such as this are achieved through packaging. Packaging bundles all the necessary pieces for a software to run, usually including dependencies, and distributes it to the user as one "package". Packaging software can make installation a lot simpler for users and allow it to be installed consistently that aids in reproducibility. Most open software won't be packaged to the double-click-with-setup-wizard level and some won't be packaged at all. They will require a bit more up front work for the user, but an advanced knowledge of installation practices shouldn't be assumed. For example, an exact command that can be copied and pasted into the command line is a lot more helpful than something like "clone the repo" or "install using git pip".

Usage examples are another important part of a README document. While how to run and use the software may be obvious to the developer, many times this is not the case for the user. Simple/small usage examples are great for the README file. If there are more complex examples that require input files or that are interactive for the user and the programming language you are using supports interactive environments, such as [Jupyter](#) (for R, Python, and Julia), [Pluto](#) (for Julia), [Quarto](#) (for R, Python, and Julia), and [RStudio](#) (for R), these can be used and included in a repository and pointed to in the README. If interactive environments are not an option for the language you are using and your usage examples are necessarily complex, consider writing a standalone script and including a pointer to this with instructions on how to use and run that example script in the README.

If relevant, the README is also one of the places you may find descriptions of the outputs of a software/code. Both what kind of objects these may be in terms of their type (e.g. string, integer, etc.) and in their general description (e.g. a list of names, the amount of rain the model calculated, etc.).

As the README is the first place a user will look, this is also where you can find other notes and caveats of using the software. This should include at least something on the state of the software: is it in active development (meaning it may have some bugs and may not always work as expected), consistently maintained (meaning the software is updated when necessary—like when a dependency is updated or a bug is reported), or here for posterity purposes only (meaning the author/developer/researcher will not be working to maintain or improve this code any further)? How can you contact the developer/researcher that created this software/code? How can issue/bugs be reported (if at all)? This would also be a good place to list any known bugs/issues, so you get repeat requests.

The README is also a great place to acknowledge team members that worked on the code/project as well as agencies and grant numbers that funded the work.

Dependencies

The dependencies – the other software on which the software/code relies – should be listed somewhere in the documentation, but are not always in the same place depending on the coding language. For example, in Python software, it is common to include a file titled something like `environment.yml` which will list dependencies and which can be used to install them quickly and easily. Other conventions may include listing them in the README file, a README can also be used to point to an additional file that lists dependencies (such as the `environment.yml` or `requirements.txt`)

License

A license file should be included with your documentation. This is expanded upon more in another lesson in this module, but without one, the code/software is technically and ethically

not allowed to be used at all by anyone other than the author/developer.

The CONTRIBUTING.md file

One of the great benefits of open software is that it enables contributions from the community. The CONTRIBUTING.md and CODE_OF_CONDUCT files in software can be referenced for information on how to do this. This is expanded upon more in a later lesson.

Documentation Checklist

- [] Description of the software and the problem it solves
- [] Compatibility description
- [] Dependencies
- [] Installation instructions
- [] Usage examples (perhaps including an interactive notebook)
- [] Development status of the software (under development, actively maintained, etc.)
- [] Contact information
- [] How to report issues/bugs (and a list of any known issues/limitations)
- [] Acknowledgments of team and funding
- [] License
- [] Contribution guidelines
- [] Code of conduct

Additionally, a GitHub template from NOAA for open software documentation can be found [here](#).

Clean/readable code

Code for software is very rarely written only for one individual. Code typically has to be read and evaluated by others. In private companies, this is usually because software is written by a group of programmers and so it is important that programmers are able to read and understand the code, both in order to improve it and to “debug” or fix it. Open software also operates similarly: there may be many programmers working and contributing to a particular project from different backgrounds and walks of life. With different programmers with different backgrounds collaborating together, it’s important that code is transparent and can be easily understood by others. This is sometimes referred to as “clean code”.

Clean code is code that is easily understood by others. Clean code has a number of advantages. One advantage is that it is easier to spot if or whether something is wrong with the code (known as “debugging”). Another advantage is that code that is “clean” is more likely to be shared than code that is not. This is fundamental to open software, which aims to be reproduced as widely as possible. There are a number of principles that should be adhered to when using clean code.

Code Comments

Arguably one of the most important is that code should be commented. Comments are annotations that help other programmers reading to understand what is going on. In many languages, they are designated by the sign `//` or `#` or `/* */`. As a rule, more comments are better than less but this should be prefaced with the warning that comments should not explain the obvious. For example, in the language JavaScript, the following would be an inappropriate comment

```
var a = 5; //I'm assigning the value of 5 to the variable a.
```

It is inappropriate because the code is self-explanatory.

Descriptive naming

Another point to bear in mind when it comes to clean code is that variables, functions, and similar entities should be given descriptive names as opposed to vague names. These are names that, when another programmer reads them, instantly gives an idea of what the variable or function is. For example, the variable name `colourOfCat` is a good name because it describes what it intends to do, which is to encompass the color of a cat. As a rule, the more descriptive a name for a variable, function, etc., is the better. Names for variables, functions, etc. should avoid using words that are likely to be keywords - names with reserved meanings in many languages - such as “while”, “for”, “override” and so on. Needless to say, names for variables, functions, etc. should similarly avoid giving offense and clean code should consider the sensitivities of those from different backgrounds.

It's frequently the case that code may point to external files; where possible, a programmer should ensure that the external file has a descriptive filename. In addition, clean code should also conform to programming conventions. For example, it's common in many programming languages to use camel case to describe variables, such as `colourForCat` rather than `COLOURFORCAT`, but one would do well to ascertain what a convention may be for a particular language.

Whitespace and indentation

Lastly, clean code should contain sufficient spaces between lines of code (also known as whitespace) and sufficient indentation so that they are easily discernible. Sometimes code that does not contain sufficient lines of code can go through a process known as *beautification* or *pretty-fying* that helps them become more readable. Ultimately, a key test for whether code can be considered “clean” is the following: if you left the code and came back to it 2 years from now, would you be able to easily understand it?

Summary

In this lesson we go over two main topics regarding markers of quality code: (1) good, descriptive documentation and (2) clean, readable code. As a user, documentation can be the difference between spending hours or days trying to understand a code and being able to use it right out of the box. As a developer/researcher, documentation improves the reproducibility and reusability of your code and lets others know what to expect both of your code and of you yourself as a maintainer. Next, we'll discuss maintaining quality code.

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Maintain good code quality

Learning objective:

- Understanding basics of Version Control
- Learning the basics of testing in code development
- Understanding the responsibilities of Open Software developers

Introduction

We've talked about markers of quality software in the prior lesson: good documentation and clean, readable code. The reality is that for most software, this is a journey, and it is going to continue to change and develop over some period of time. Here, we discuss version control, testing, and responsibilities after sharing. These topics are centered around the evolution of your code and ensuring the work you've done to make quality open software is able to endure.

Version control

Open source codes can change overtime. This brings several challenges to researchers developing and using an ever-changing software. We covered the importance of reproducibility for open-software - and open-science as a whole. Now, how can we achieve reproducibility with a changing code source? That is done by keeping track of changes to our source code, using version control.

Version control can be done with tools and systems designed to manage changes not only to source code, but also to documents, websites, and datasets. [Google Docs](#), for instance, has its own complex version control. This allows you and your collaborators to have access not only to the most updated google document you all are working on, but to the complete history of changes. So, if something goes wrong in a document: a child includes a thousand smiley faces in the text, a cat walks on the keyboard and deletes an entire section - you can just revert to the earlier, error-free version.

This is the same for coding. For instance, you - the developer - receive a notification from a user that your code has a bug. You know that this bug was not present in the last version, so you can easily work through your history to look what recent changes might have caused a

specific error, narrowing down your debugging work to specific parts of the code. So, version control allows a group of developers/users to know exactly what version of the code they are using, what changes were made and when - facilitating reproducibility. Version control also fosters collaboration, making it easier for people to work together at the same time and to merge changes from different users.

There are several version control systems (VCS) available. We won't get into detail here, but some of the most popular open-source systems include [git](#), [SVN](#), and [Mercurial](#). It is important to note that while some repositories have already a built-in version control, repositories and version control systems are different - *e.g.*, *git* is the *version control system*, while [Github](#) is a *hosting service* for [git](#) repositories.

In [lesson 6](#), we revisit version control, giving some concrete examples of how you can use it to contribute for new or existing open-source code.

Testing

In [Lesson 1](#), we introduced the concept of code testing and its importance in software development. There are many types of testing that range from testing the smallest testable parts of a code to verifying if a code works as whole under different scenarios. Since code testing in general can be a complicated and technically involved topic, we will not go into the details of each types of testing and refer you to external sources for further reading. Instead, we focus on benefits and difficulties of testing in general, how to measure test coverage, and what to expect from a “tested” code as an end-user.

We recall that reproducibility in research software plays a critical role. In the context of testing, we can think of reproducibility as a test objective of which is to reproduce a specific output, *i.e.*, results obtained from a specific version of the code that has been published in a journal. This test should include all the required inputs (configuration files, input data, etc.) so users can easily run and get the same published results.

More broadly, the main objective of code testing is to evaluate if a code is doing what it is supposed to do. It is important to recognize that testing a code comprehensively can be very difficult since not only we should test the code for generating expected outputs but also for failing when it should. For example, when an unacceptable input is passed, *e.g.*, wrong type, out of range, edge cases, etc., or when if implemented the algorithm doesn't converge for the given set of inputs. Taking into account all these scenarios can be extremely difficult and in some cases impossible. Therefore, we should manage our expectations when taking the tests as a measure of code's quality both as a developer (*e.g.*, realizing that the end-users might apply the code to scenarios that we don't anticipate) and an end-user (*e.g.*, realizing that the difficulties associated with testing and, if possible, evaluate the accuracy of outputs independently).

From a developer perspective, there are also secondary benefits for testing. Whenever you make a change to a part of your code, for example to improve its performance, having tests for that portion of the code, ensures that the modified code does not change the output. Another scenario could be related to dependencies. For example, research software often depends on other software, therefore, if those dependencies release new versions, the tests help us evaluate if those new versions make any changes to outputs of our code.

On the other hand, as an end-user, using a code that includes tests, gives us more confidence in the state of the code. Users can check the status of tests (pass/fail) when the developers make changes, or the code has been tested for the use-case of our interest.

Now that we have a better understanding of the testing, we can discuss measuring its effectiveness. One of the ways that we can measure the testing is through percentage coverage. There are two levels of coverage: *test coverage* and *code coverage*. *Test coverage* refers to the coverage of different scenarios that the code would be used in while *code coverage* is the percentage of lines of code that tests cover. As we discussed previously, enumerating all the different scenarios the code could be used in can be very difficult, thus, it can be difficult to quantify *test coverage* both from a developer and end-user perspective. However, *code coverage* is just a simple percentage value: how many lines of code do the tests activate vs. not. It is important to note that a high *code coverage* does not necessarily mean that a code has good *test coverage* since testing different usage scenarios can not directly be translated to lines of code.

Additional Resource

- [IBM on Testing](#)
- [Software Testing](#)
- Martin, R. C. (2008). Clean code: A handbook of agile software craftsmanship. Prentice Hall.

Responsibilities after Sharing

After sharing software, there are certain steps that need to be taken in regard to maintenance of that code/software.

First, you should know it is not a requirement for you to be a permanent maintainer forever, but it is your responsibility to let users know if you do or don't intend to maintain the software/code. You can do this in your documentation where you discuss the development status of the project. This helps a user know if it will continue to be supported in the future, and make choices about if they should base ongoing work off your project. You don't want someone to spend a huge amount of time using your work as a dependency and then have their project become unusable in the future.

The reality is that a developer/researcher may not have the time or continued funding to keep up with a project. In this case, perhaps consider handing ownership of the software to another researcher/developer, involved user, or entity invested in its continued use. You can either approach potential parties you think may be interested in this; or you can make your license permissive enough to allow others to create their own copies and continue your work (see more on choosing a license in this module). Depending on the license you choose, the use of your project, and if you have significant interest, you may be able to commercialize your software/code to provide funding for continued maintenance and feature requests. There is also the potential to apply for continued funding from agencies both governmental and private if your open software is widely used. If you're a user of a software that is no longer maintained, consider contacting the owner/developer and volunteering either as a maintainer, or to take over ownership of the project (you'll be more likely to get a positive response if you leave that choice up to the current owner).

If you receive requests for features and fixes, and you have indicated you intend to maintain the code, these should be responded to. Either tell the users that (a) you intend to perform their requested action or (b) you think that's out of scope of your project. Additionally, you can invite the requester to (a) contribute to the project and add that feature/fix themselves (which you can then approve and add into your project) or (b) fork (make a copy of) the project and create the feature/fix, notifying that you will not merge changes into your (main/original) copy.

Summary

Here we discuss how version control and testing can both be used to increase the reproducibility and trust a user can place in open software. These are tools that can be used whether your software is shared or not. We go over what responsibilities a developer/researcher has after sharing their code: namely to inform your potential users if you will be maintaining the software and if so, respond to requests for feature additions and bug fixes. We discuss options for allowing your code to undergo continued development even if you don't have time/motivation/funding to continue iteration and encourage users of code that is no longer maintained to explore these options themselves by reaching out to the original developers. Furthermore, we discuss how users can become involved in existing projects in our next lesson.

Contributing to existing open software

Learning objective:

- Understanding the importance of contributing to Open Software
- Learning different ways of contributing to open source projects
- Learning best practices for useful contributions to open source project
- Learning about ethical aspect of contributions

Introduction

In [previous lesson](#), we have discussed the importance of using version control and testing to maintain good quality of code. Community contributions are the primary driving force behind open software initiatives. Open software contribution not only benefits the contributor, but also help to maintain the software's long-term viability. In this lesson, we will cover the various types of contributions that can be made, which are not limited to coding contributions; non-coders can also make significant contributions to open source software. In addition, we will cover how to use version control in open-source project contributions; some good contribution practices will be discussed in this lesson as well.

Benefits of contributing to an open software

Contributing to open software provides many valuable advantages and opens doors to a number of highly lucrative and rewarding opportunities, and there are not too many other industries that can boast the massive number of global contributions like the open-source community can.

A first advantage of contributing to open software is that it will require you to write clean, documented, structured code. In combination with the feedback you will obtain from leading developers in the field, this can help to improve your coding and communication skills.

Secondly, contributions that you have made to open software constitute a documented and publicly available record of your work (git commits, for example, get indexed within google search). This allows you to reference to your contributions as part of a software portfolio or resume, providing a direct evidence of your work and skills.

Finally, contribution to software by members of the community creates a unique constellation in which the contributors to the software are also its main users. Often, contributions to open software stem from users who wish to improve or change the software for their own use and adapt the software problem constellations in the software's field of use. This direct feedback loop between user and developer allows for a fast development cycle and makes open software more flexible to changes in needs and requirements than software products that are maintained by a company.

Types of contribution to an open software ¹

There are several types of contributing to open software. Not all of them require writing actual code.

Add new features. The most obvious case for contributing to open software is enhancing its usability by adding new features. Make sure to open a new issue first.

Fix bugs/issues. Alternatively, you can reply to an already opened issue by fixing it. Make sure to reference the issue when creating a pull request/ request for reviewing your fix.

Report issues/ suggestions about improving code. Reporting an issue is a valuable contribution even if you don't know how to fix it. For example, you might be using a different browser in which the software has not been tested yet, have discovered a particularly uninformative error message, be colorblind or be otherwise able to feed a valuable user experience back to the developers that can help to improve the overall usability of the software.

Improving and contributing to documentation. Contributing to documentation constitutes a great starting point to contributing to open source software and is often overlooked in its importance. Writing documentations allows you to familiarize yourself with the use of the software, while helping to teach others.

Create tutorials, use cases or visuals. Another way to contribute is to make your experience and use of the software publicly available. For example, you could create a tutorial based on your use of the software, summarize a use case or provide a summary of your use in a graphic. This part of contribution is particularly appealing as it does not create much extra work to just publish what you have used the software for.

Improve layout, automatization, structure of code. Apart from creating new code, a good way to contribute to open source software can also be to improve, restructure or automatize existing code. This is called *refactoring* and helps to make the software project more effective and stable.

Organize/attend a meetup/community building. Another way to contribute to open source software is via community building. Many software products and toolboxes have a lively community of users that meet on a regular basis in person and online to discuss and improve

¹[opensource (<https://opensource.com/life/16/1/8-ways-contribute-open-source-without-writing-code>)

the software and its use. Participating or even organizing such a meetup can be a good way to improve your knowledge of the software, get to know its community and contribute to open source projects

Code review. Pull requests or other requests to integrate new contributions into the main code base usually require a review of the contribution by at least one other user. In the git version control system, code review entails writing a short summary about the quality of the code, making suggestions about improvements and then approve or reject the request.

How to contribute? ²

Before you contribute to an open source project, there are several resources that you can check in order to get a feel for the community, the general environment the software lives in and the contribution and maintenance process. Below some examples of essential files ³ that you might find in a repository and that might be worth looking at.

- The **README.md** file gives first information/summary about the project. Here you might also find installation instructions, software and operating system requirements or a reference to published papers on the software.
- The **CONTRIBUTING.md** file gives information about how to contribute to the project. It explains in more detail how the contribution process works and what type of contributions are needed. While not every project has a **CONTRIBUTING.md** file, the existence of one is a clear indicator that contributions are welcomed.
- The **LICENSE** file contains the legal aspects and boundaries of contributions. It specifies in which ways the code can be altered and how to proceed with altered code. While alterations to code just for your private use are usually always possible, the **license** file comes into play in case you intend to publish or commercialize and alteration to the software.
- The **CODE_OF_CONDUCT** file: The code of conduct sets ground rules for participants' behavior associated and helps to facilitate a friendly, welcoming environment. While not every project has a **CODE_OF_CONDUCT** file, its presence signals that this is a welcoming project to contribute to.

Contributing via a version control system

Congratulations! You have decided to contribute to an open source repository. However, to protect the code in the original repository, you usually don't have rights to commit directly into that repository.

²[freecodecamp](#)

³[Categorizing the Content of GitHub README Files](#)

Hence, as a user, the next step on your way to a contribution is to create a **fork** (a copy of the original repository into your own account). In contrast to the original repository, you will be *owner* of the fork, and thus you will have writing rights.

You can also **clone** this fork onto your local machine. Then there will be three copies of the repository: The original **upstream** repository, the fork in your (online) account, called **origin** in git, and the local clone.

Alternatively, as a developer, you can also create a new git repository from scratch (use `git init` here). This will make you the owner of the repository and give you writing rights directly.

You can now make changes to your local clone, your local initiated repository or to your online repository, each of them also being called your respective **working directory**. Changes to the working directory will be tracked in a **staging area**, from which you can and commit them using the command `git commit -m message`. If you committed to you local clone or initiated local repository, you need to push them to the origin repository (your online fork) first, if you want to make use of them online.

From there, you can create a **pull request** to an upstream repository. The owner of upstream repository will then review your changes and approve them or request changes.

Simple version control workflow

We have again summarized those steps in a checklist for you. We present here a simple definition of the workflow with common terms you will encounter, and offer some suggestions for a more in-depth lesson. [Software Carpentry](#) can be a great place to start!

[] Create Repository

- Developer: creates a new repository from scratch. Our tip: just go for it. You can create your repository with one file, or an entire existing open software.
- User: will create a copy (*clone* or *fork*) of an existing repository.

[] Make changes

- You can make any changes you want to your copy, but no one will see your changes until you *commit* (*i.e.*, submit them).

[] Publish your changes

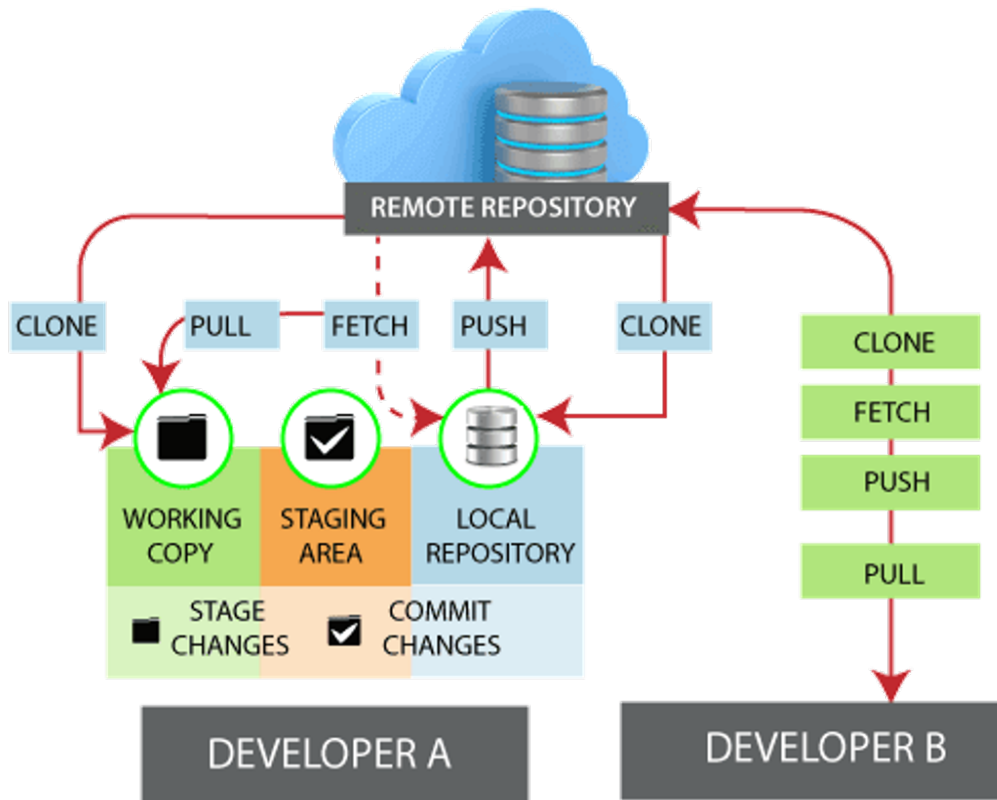
- If you are like your changes and additions, *commit*. This will update your local repository.
- So far, only your local repository has changed. To update your remote repository, *push* your modifications.

[] Get changes from others

- While you were working on your copy, other users might have changed the remote repository. To keep your local repository updated, you need to retrieve, or *pull* the latest changes.

[] Keep track of changes

- To check what is different in your copy since the last commit, you can check the *status* of your repository.



<https://www.javatpoint.com/git-remote>

Figure 1: remote

As a last note, version control is a good practice for coding, so use it even if you are not sharing it immediately. You can use version control with your codes privately on your computer, or use the private mode on hosting services (*e.g.*, GitHub and GitLab). And, once you are ready, you are one step ahead to share your code.

Further Resources

- [Software Carpentry Version Control with Git](#)
- [The Turing Way, Version Control](#)
- [FAIR Use a publicly accessible repository with version control](#)

Types of Commits

A sustainable open software usually depends on active contribution from the community through commits access to the repository. In software version control, a commit is an operation which sends the latest changes of the source code to the repository ⁴. In general, commit operation can be classified into 3 categories ⁵ : Core, External, and Mutant.

- **Core Commit** refer to any commits that directly associated with the main repository. The Core Committer usually refers to the individual who has write access to the repository of software, and responsible for reviewing pull requests
- **External Commit** is the contributions that go back into the upstream repository through patches or pull requests, and it need permission from Core Committer.
- **Mutant Commit** is a modification to the code-base of a project which is not incorporated back into the upstream repository. This situation happen due to the changes request rejected by the Core Committer or the committer intend to personal use only.

Branching and Merging

In software version control or software configuration management, branching is the process of object duplication from the original work under version control. ⁶ In this context, the duplicated objects are known as branch. A branch is a version of the repository that deviate from the main working project, and it is independent line of the development process.

Branching allows parallel development works including bug fixes, feature addition, and safely experiment on the same software while retaining the original source code. The subsections from the main project allow development teams working on the branch independently and free to make any changes without impacting on each other.

Every repository has a default branch, which is the first main branch, and it sometimes called parent branch or upstream branch, whereas the child branches are the branches from a parent. We can create many branches from the existing branch. A branch also acts a pointer to one of the commits in the repository. The HEAD is a special pointer that simply points to the latest checked out branch or commit. For example, the default branch named as `master`, and this

⁴[Wikipedia Commit \(version control\)](#)

⁵[ACM](#)

⁶[Wiki-Branching \(version control\)](#)

master points to the most recent commit called **bug-fix**, remember that pointer is movable when there is new commit.

Upon completion, the branches can be reassembled to the mainline and become new version of the software release. The process of integrate changes into the upstream repository is called merging. If you have no permission to commit directly into the upstream repository, create a pull request from the branch into main is necessary. It is a good practice of software development etiquette by ensure the branch is stable before merging them into main branch. Once the merged completed, the local branch can be safely deleted. On the other hand, a branch that not intended to be merged is known as a fork.

In summary, branching and merging is typical process that allows development team to work on shared codebase and manage the software effectively.

Merge conflicts

Definition

The merge conflicts occur when the version control systems unable to automatically resolve the differences in codes between two commits. It requires manual changes and decision to incorporate in the final merge.

Here is an example to explain the scenario, both developer *A* and *B* make changes on code file in different branch, they make changes on the same line of codes. During process of merging these two branches, it will cause merge conflict as it has competing or ambiguity changes.

How to resolve

In order to resolve the merge conflict, we must find out where is the conflict occur, identify the affected code file and specific lines that causing error, make necessary correction and then make a new commit before merging these branches again. Make sure latest changes are made on the file that we want to keep.

How to avoid

There are few ways to avoid merge conflict, the simplest way is make sure changes are made on different lines, or different files, to ensure not introduce any ambiguity lines. Secondly, make sure the local branch or the branch that currently working on is updated before make any changes.

Recommended Practices

Here are some recommended practices [[^deepsources](#)]⁷ for version control.

1. Adhere to templates when opening an issue

It is a good practice for version control to check on the documentation in open software repository if the repository consists of **CONTRIBUTING.md** file. This file usually is in the root directory which describing how others can contribute to the project.

2. Make clean, single-purpose commits

It is better to commit the changes with single purpose instead of commit combined changes at single time. For example, we prefer push the changes for bug fixing and feature adding in different commits.

3. Write meaningful commit messages

It is always best and easy practice to commit the changes with descriptive commit messages. A good commit message gives reviewer a clear and insightful description about what has been changed.

4. Commit early, commit often

Other than single-purpose commits, commit early is also one of the good practice. Commit the work more often and in small chunk will help the repository keep updating and avoid conflicts.

5. Don't alter published history

It is strongly not recommend altering the published history. As some version control tools allows to rewrite branch history, but it might cause unnecessarily confusing.

6. Don't commit generated files

Only commit the files that have been generated manually is also a good practice. The files that can be re-generated usually do not work with line-based difference tracking.

7. Refer to issue when creating a pull request

If the intention of your pull request is to fix an issue in the software, it is highly recommended using a supported keyword in the pull request's description or in a commit message.⁸ Linking a pull request to an issue is certainly helpful for showing the status of fixing is in progress.

⁷[Perforce](#)

⁸[Github Docs](#)

8. Assign reviewers

Assign reviewers to validate the commit before merging definitely is a good practice in contribution as it help to avoid unnecessary conflict and quality assurance.

Naming Etiquette

Deprecated terms. ⁹ The computer industry’s use of the terms *master* and *slave* caught everyone’s attention in the summer of 2020. Amid the many protests and the growing social unrest, these harmful and antiquated terms were no longer considered appropriate.

“Both Conservancy and the Git project are aware that the initial branch name, ‘master,’ is offensive to some people, and we empathize with those hurt by the use of that term,” said the Software Freedom Conservancy.

The name master for a new repository is outdated and has been replaced by *main*.

Ambiguous terms While git is the most common version control system, terms may vary between git and its alternatives.

Ethical considerations

The ability of anyone from the public to contribute to an open source software project creates an interesting ethical and legal situation regarding the software’s ownership. It should be clear that contributions such as fixing a typo in the documentation does not create the right to claim for (partial) ownership of the software, but the lines for more substantial contributions tend to blur fast and are often up for discussion. In general, the answer to when a contribution has altered a software enough to justify a partial transfer of ownership has to be determined on a case to case basis. Often, this process requires considering the license and contributing agreements. For example, while many repositories state for example in their contributing file that contributing includes a loss of ownership and rights of the code to the owner of the main repository, other repositories acknowledge already minor contributions, for example by assigning rights to the repository or adding the contributor’s name to a an acknowledgment file.

Summary

To summarize, contributing to open software delivers multiple benefits to the community while also assisting in the product’s maintenance. Contributing to an open source project can help you enhance your technical abilities and get a better reputation. If you are a coder, you

⁹[theserverside](#)

may typically contribute by reporting issues, resolving bugs, and creating new features. Aside from that, you can help by producing documentation, increasing the repository's visibility, or refactoring it. Both coders and non-coders have an equal opportunity to contribute to open source software. We explained how to use the basic version control workflow, which begins with creating a repository, then making changes, publishing changes, and pulling modifications if any exist, and finally keeping track of the changes. In general, commit operations are classified as Core, External, or Mutant. Branching and merging are both important steps in version control, and merge conflicts should be avoided. We also talked about creating a list of best practices to which you can refer before contributing. Nonetheless, there are some ethical issues to keep in mind while contributing to open source software.

OpenSciency Open Software: Authors

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Open Data

Learning objectives:

- Describe key characteristics of open data
- Categorize types of open data

Introduction

As mentioned earlier, data is a major part of scientific research, and why wouldn't it be? It is evident that data permeates many aspects of our daily life with significant consequences.

For instance, it has become all the more common to see news articles discussing how data and efforts like [Open Street Map](#) are critical in supporting the disaster emergency responses all over the world [1](#). This is only an example among many others demonstrating the value of data, particularly open and public data, in our daily life and for public good.

Similarly, data shared openly in scientific research brings tremendous value which is not limited to the scientific community but extends to communities at large from indigenous communities to urban populations! Before we look further into this, let's first look at what is data in scientific fields? What is open data? What are the key characteristics of open data?

What is Data?

Data is any type of information, recordable or observable facts. **Research data** is thus data collected in order to answer the research questions of a project.

Research data can be numbers, texts, measurements, images, model output, and more. Research data is collected in different ways, and formats. Generally speaking it can be grouped in different ways, for example:

- Qualitative data describing information in words
- Quantitative data with defined numerical information, e.g. from measurements
- Grouped versus ungrouped data; grouped data being data that has been put into classes for next steps such as interpretation and ungrouped data referring to the raw data [2](#)
- Structured versus unstructured data: unstructured data referring to data in its raw format, and structured data referring to data formatted for storage in files or records such as relational databases [3](#)

In the next section we will introduce the different types of data most commonly generated/found in research. This will provide an understanding of how different data should be handled and what considerations to keep in mind when sharing it openly. More on this can be found in lessons 3 (Responsible Open Data) & 4 (CARE and FAIR Principles). We will also highlight one particular type of data (metadata, described below) and its role in supporting the development of Open Data and Open Science initiatives.

Primary (raw) data

Primary data refers to data that is directly collected or created by researchers. Examples include surveys, questionnaires, interviews, physical samples, specimens, output from models, remote sensing data (spectral/photons), etc. Research questions guide the collection of the data. Typically, a researcher will formulate a question, develop a methodology and start collecting the data. Some examples of primary data include:

Responses to Interviews, questionnaires, and surveys.

Typically, interviews generate data, in the form of recorded audio files, transcripts, and notes and other observational data. Interviews can address a broad range of quantitative or qualitative oriented research questions, and may be used on their own or as part of a mixed methods approach. In behavioral and social sciences, these data collection methods are often used to collect self-reported data. A researcher designs these questions to collect data from participants that are necessary for the research. In most cases, this type of data can be openly shared under certain conditions or considerations e.g. if they are de-identified, and if the participants consent about the data being shared, more on that in lesson 2 and 3.

Data acquired from recorded measurements, including remote sensing data.

In many cases, the raw primary measurements from an instrument are processed in various ways such that what is typically stored and reported is based on a variety of calibration, normalization, and even compression steps that are ideally well defined and described by a discipline or measurement protocol. For example, satellite captured imagery is often used in online map services and navigation (e.g., Google Maps). These satellite imagery is often captured by various sensors onboard space-borne satellites (e.g., [NASA Earth Observing System](#)) or airborne measuring platforms (e.g., uncrewed drones or planes) [4]. These data often need careful calibration and correction including the values recorded by the sensors and the geospatial location of the imagery before it can be used appropriately for research and application. After the calibration and corrections, remote sensing imagery are often used to create products that can help us understand the environment that we live in and useful for societal benefits (e.g., studying air quality impact on community health).

Data acquired from physical samples and specimens form the base of many studies.

Tests and analyses are conducted on these resources, such as biological specimens, rocks and minerals, soils and sediments, plants and seeds, water samples, archaeological artifacts, or DNA and human tissue samples ([Research Data Alliance Interest Group on Physical Samples](#)) [5]. While it may be more difficult to share these types of physical resources, information about the samples and as well as data derived from using them, can be shared via thorough description, such as in the case of BioBanks, [IGSN](#) (a persistent identifier specifically for physical samples), and [iSamples](#) [6] [7]. For more information about how to manage physical samples, check out the 23 Things Physical Samples [8].

Data generated from models and simulations.

Not all primary data are observations collected by people or instruments. We also build mathematical models either based on physical laws or empirical relationships to understand a subject or system. The model can produce a suite of simulations driven by various input scenarios or initial conditions. For example, Coupled Model Intercomparison Project ([CMIP](#)) is an international project with participation from more than 20 earth system modeling centers to generate the historical simulation and future projections of the earth system under different greenhouse gas emission scenarios [\[9\]](#). The Open Data generated in the most recent phase of CMIP (CMIP6) allows researchers to understand the impact of the changing climate to the ecosystem and our society. The understanding derived from CMIP6 data can then be used to inform climate policies and adaptation strategies. Model data are valuable assets because it can provide data that can be hard or sometimes impossible to collect in real world [\[10\]](#).

Processed data

Processed data typically refers to data that is created or collected by someone else and used by others.

Examples include data from literature, academic publications, generated statistics such as government statistics, transcripts of recordings, and a variety of streaming environmental and biological data that are deposited and made available in databases and repositories.

This type of data is oftentimes used for a different purpose than originally intended, for example, from a previous experiment or from another research project or a different discipline.

It is very common in the era of digital scientific research to see new primary datasets created or produced by collecting and repurposing secondary data and/or mixing them with new primary data. This kind of research practice is made possible by the promotion of Open Data. Data sharing provides opportunities for all researchers, even the novice and/or unfunded, therefore leveling the playing field [\[11\]](#). More on the benefits of Open Data in lesson 2.

Metadata

Metadata is a special type of data that describes other data or objects (e.g. samples). It is often used to provide a standard set of information about a dataset to enable easy use and interpretation of the data.

It can facilitate assessment of dataset quality, by answering key questions, such as including key information on:

- How the data was collected (e.g. which equipment/instruments were used)
- Which variables/parameters are included in this dataset
- Who or which organization created or collected the data
- When the data was collected and deposited
- Where to find the data (e.g. DOI) and how to cite it

- Which geographic region the dataset covers
- Who can use the data and how
- Which version is the dataset
- What is the format of the data
- Whether the dataset follows any community or international standards or guidelines

In addition, metadata allows cataloging and data discovery specially, if it follows leading practices allowing them to be indexed by search engines ([The Turing Way, 2019](#)) [12]. In other words, it enhances searchability and findability of the data by allowing other machines to read and interpret datasets (see the concept Findable in lesson 4 - CARE and FAIR principles)

In later lessons, we will describe how to create metadata and where to find help (e.g. repositories)

There are different types/categories of metadata addressing different purposes:

- **Descriptive metadata** can contain information about the context and content of your data; such as, abstract, title, subject keywords.
- **Structural metadata** is used to describe the structure of the data (e.g., file format, the dataset hierarchy).
- **Administrative metadata** explaining the information that is used to manage the data (e.g., when and how it was created, which software and the version of the software used in data creation).

1.2 What is Open Data?

The term “Open Data” is relatively new with the first appearance in 1995 in an [article in Paris Tech Review](#) describing the need of sharing Earth and environmental data because “our atmosphere, oceans and biosphere form an integrated whole that transcends borders.” [13]

In this lesson, we are adopting the definition of “Open Data” as defined in the [Open Data Handbook](#) from the [Open Knowledge Foundation](#) [14] [15].

“Open data is data that can be freely used, re-used and redistributed by anyone - subject only, at most, to the requirement to attribute and share alike.” Open Data is defined by a set of key attributes, but keep in mind that not all aspects can or will be present at all times as there might be some considerations or restrictions to take into account. This could be partly due to the fact that Open Data does not necessarily mean Open for ALL, but rather Open for the specific individuals and/or communities, more on this in lessons 3 (Responsible Open Data) & 4 (CARE & FAIR Principles) .

Availability and accessibility

Open data is characterized by being available and accessible, meaning that it is published on a publicly available platform and accessible to download over the internet allowing others to find it and use it.

Ideally, scientific research outputs including; research data, metadata, manuscripts, open educational resources, software, hardware and source code are published and made available in both human and machine readable.

Generally, Open Data does not require a payment. Nevertheless, in some cases infrastructure costs might be required which can be covered by societies, institutes, organizations etc. More on this in lesson 6 (Sharing Open Data).

Reusability

For Open Data to be useful, it has to be reusable. Without the capacity to reuse it, we are creating “data tombs” where the data is hosted (lives) but is of no value to others [16].

There are essential factors that need to be addressed in order for the data to be in good enough shape for others to use it. For instance, researchers and data reusers are mostly looking for data which is “comprehensive, easy to obtain, easy to manipulate, and believable”. For these criteria to be fulfilled the data needs to:

- Sufficiently described with appropriate metadata, which greatly affects open data reusability. There is no one size fits all for metadata as its collection is guided by your data. More on metadata generation in lesson 5 (Planning for Open Data)
- Has appropriate license, copyright and citation information. See lesson 6 (Sharing Open Data)
- Has appropriate access information. Learn more in lesson 6 (Sharing Open Data)
- Findable in an accredited or trustworthy resource. Learn more in lesson 6 (Sharing Open Data)
- Maintained on a regular basis, addressing feedback from different levels of users
- Accompanied with history of changes and versioning (see definitions) allowing users to point to the exact state of data when they decided to use it.
- For processed data it is also important to include details of all processing steps. For instance in the case for large data, e.g., it is not common practice for raw genomic reads to be preserved. In most cases processed genomic data are reported, and archived. Hence, some case sensitive data should contain information about processing steps (e.g., calibration, normalization) depending on varying instances.

Inclusivity

Data inclusivity refers to making the data truly open and available for all independent of nationality, location, race, age, gender, sexual orientation, religion, income, socio-economic circumstances, career stage, discipline, language and culture, ability and disability, political ideology, ethnicity or immigration status or any other grounds.

Open Data is free from all types of organizational, cultural, political and/or commercial restrictions. A good example of open data transcend national boundary is the [world bank poverty headcount data](#), which is anonymously developed from primary household survey data of different government statistical agencies and World Bank country departments [17].

The inclusiveness of open data brings many benefits (Lesson 2 - Benefits of Open Data). However, there are situations when it is not ethical or appropriate to share certain data, or there are considerations to implement additional steps and policies to ensure the proper use of open data (Lesson 3 - Responsible Open Data).

Summary

In this lesson, you learned about different types of data, the definition and key characteristics of open data. You might already have some great examples about open data that you are interested in learning more. In the next lesson, you will be exposed to various benefits and challenges of open data.

Assessment

Self Assessment: Do you know the difference between different types of data?

A researcher needs data on trauma for an ongoing project.

Scenario 1: They visit a trauma center and ask questions to patients in the trauma center. What type of data was collected?

Senario 2: In the trauma center, they are referred to a database where they can find responses from patients. What type of data was collected? Footer

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Benefits of Open Data

Learning Objectives

- Communicate the benefits and challenges of Open data and its effects on science

Introduction

In this lesson, we'll discuss the benefits of open data and in particular its direct effect in advancing Open Science. We will also discuss details of how Open Data can impact the response of science in global emergencies, and how Open Data facilitates multidisciplinary work.

Open Data for the greater good

As we mentioned earlier, data plays a significant role in our day-to-day lives. Open Data, in particular, has played a key role. If you pause and think about it, you may realize that Open Data is not only common in our society, but you might have benefited from it and used it yourself.

Here, are some notable examples of Open Data that has positively impacted society at large:

Each country or territory often provides open access to a variety of socioeconomic information about the population, community, and business in its jurisdiction. These data are often called census survey data which may include the aggregated statistics of gender, race, ethnicity, education, income, and health data of a community. These data are often used to understand the composition of a local neighborhood and are critical to inform decisions on resource allocation to ensure the quality of life for the community.

The changing climate poses a significant risk to our daily lives and has been responsible for intensifying drought, increasing flooding, and devastating fire incidents worldwide. Open data is therefore critical in providing life-saving information to adapt to the changing climate and help assess the climate risks of the place where we live. Government agencies (e.g., National Oceanic Atmospheric Administration in the U.S., UK Met Office, European Centre for Medium-Range Weather Forecasts) have been providing public access to long-term weather and climate

information for decades. A more recent initiative stems from organizations developing value-added open data products to advise society on the risk of changing climate. One recent example is the flood and fire risk in the United States developed by a non-profit organization First Street Foundation

Open Data for better Open Science

Scientific discovery and innovation stand to gain a tremendous amount from Open Data. This impact stems directly from the multiple inputs and methods developed for investigating problems. Specifically, three core components of Open Data drive this diverse scientific innovation and provide enormous societal and scientific benefits:

Validation:

Open Data that is easily accessible by other researchers allows for scrutiny, which helps discover mistakes more quickly and ingrains confidence that the research was conducted with sound and ethical principles and methods. Evidence-based progress is important in providing confidence in the scientific results and is important for the insights drawn to inform future research.

Data that has been reviewed, maintained and scrutinized by many, as well as informed by diverse consultation, drives robust and thorough scientific pursuits.

This validation process is a key component of reproducibility, which is important in building on prior research. Reproducibility is the cornerstone of pushing science forward, as it is the very baseline to check results and expand upon them by introducing new experiments and questions.

Transparency:

Building on the idea of validation and scrutinization, transparency facilitates this process. It allows for early engagement with the data and ensures the data was collected with sound and ethical principles (these will be elaborated upon in lessons 3 (Responsible Open Data) & 4 (The CARE and FAIR principles)).

This transparency allows for early intervention if there are unexpected harms. This is where the idea of multiple perspectives becomes important again. Collaboration:

Open datasets are made available to all (see section Inclusivity in lesson 1) - which means new, robust insights are gathered at a faster pace as mistakes can be caught more easily, expensive data collection doesn't need to be repeated, and researchers build upon the work of their peers. For example, the first image of a black hole; Scientists recently produced the first image of a black hole in our galaxy. This achievement was only possible through open collaboration

and sharing of telescope data by different observatories distributed across different parts of the world [1].

The data isn't limited to those within a specific field nor exclusive to those with institutional access. Importantly, this means the data can be shared with non-traditional academic researchers such as nurses, social workers, agronomists, journalists and other communities. This allows for researchers to also derive insights from varying perspectives.

The scope of research can be easily expanded to derive more holistic insights. For example, the Coupled Model Intercomparison Project (CMIP) that started in 1995 paved the way to understand how climate change was impacting our daily lives by investigating factors such as malaria distribution in Africa, infrastructure and urban design as well the implications of climate change on the risk of epilepsy [2, 3].

Collating similar data sets and performing meta-analyses on those data sets can provide a substantially improved signal that would not be possible in any one of these data sets. Additionally, this facilitates convergence across scientific disciplines, increasing the value of the research.

Open Data to support policy change

Open data can lead to policy change which directly impacts the lives of communities, such as those destined to suffer first from the slow changes to the Arctic. A study, taking advantage of the OpenStreetMap data [4], helped map projected changes in the Arctic. These mappings in turn helped emphasize the need for adaptation-based policies at community and regional levels to avoid stagnation of change in the light of a sudden and dramatically worse situation fueled by climate change.

Open Data in face of global emergencies

The COVID-19 pandemic demonstrated to the world, in real-time, how the collective movement of researchers sharing their data (such as sharing of coronavirus genome data [5]) can lead to an unprecedented number of discoveries in a relatively short amount of time. This directly impacted radical vaccine development efforts and the timely control of the COVID-19 infection [6]. These insights will continue to pay off, with this research spurring future developments.

Data sharing has many benefits and can aid access to knowledge. However, it is also important to bear in mind where the data has come from, who should have a say in its interpretation and use, and how the data can be shared responsibly, more on that in lessons 3 & 4.

Open Data and public engagement (citizen science)

A citizen scientist is a citizen or amateur scientist that will collaborate with professional researchers to help gather data on a broader spatial and temporal scale than the researchers might be able to achieve on their own [7, 8]. This outsourcing of responsibility helps members of the public engage in scientific pursuits that ultimately benefit them and allow research to be conducted on a grander scale than that might be possible with only professional researchers. Citizen science is gaining popularity, with increasing recognition as a valuable contribution to scientific advancements [9].

For example, volunteer citizen scientists in Beirut were recruited from 50 villages to help test water quality [10]. These volunteers were trained to be able to conduct the tests and in turn, not only was the data collected to inform the scientific advancements, the citizen scientists had the opportunity to learn to better manage their water resources and were able to improve conditions, creating a mutually beneficial interaction.

Open Data and decolonisation of knowledge

Free distribution of knowledge gives rise to increased participation in science. Open Data is central to fostering science that is inclusive and diverse, with direct and relevant benefits to impacted individuals and communities. This fostering is particularly important in the mission towards the decolonisation of knowledge [11].

In a world where knowledge can be a commodity, with currency in the form of published papers and hoarded datasets, exclusion from research can limit progress and negatively impact a community's progress in a world driven by a knowledge-based economy.

Open Data, and its positive side effect of decolonisation of knowledge, promotes and benefits from diverse perspectives through purposeful inclusion of African, Latin American and other underrepresented Low and Middle Income Countries. This inclusion allows a dramatic change in who has access to work with and reuse data.

It can also become a powerful tool in the fight for visibility and credit. By fostering a global research culture of transparency and validation, where the work of underrepresented groups is celebrated and compensated, such as giving credit or much needed vaccines in exchange for the world-class genome sequencing in Africa, we will create a sustainable model that ensures under-represented countries are able to keep contributing towards a global revolution for example against infectious disease. It also gives marginalized groups such as women, under-represented communities, indigenous scholars, non-Anglophone scholars, as well as scholars from less-advantaged countries a voice in how the global and nuanced narrative of science is developed. This broad scale participation and inclusion shows respect to the involved people and communities and helps raise the profile of the research through considerate inclusion.

Having said that, Open Data has been demonstrated to further marginalize or exploit small-scale and community driven initiatives, such as in the case of African researchers neither receiving due credit nor compensation for their genome sequencing during the COVID-19 pandemic [12]. This is further explored in the next section as we introduce ways of mitigating harms that could happen via unthoughtful and irresponsible sharing of data.

Summary

Open Data which is purposefully inclusive and open to scrutiny, benefits scientific innovation by allowing for a more diverse and robust scientific process that draws on multiple perspectives. This also allows for the early identification of mistaken insights as well as early intervention for unforeseen harms to impacted communities.

Open Data allows non-traditional researchers to contribute to scientific development and bring their unique insights to the table. With these benefits in mind, we should always bear in mind that Open Data requires careful consideration of the possible downsides of making data open without due credit and consultation with potentially vulnerable and/or marginalized communities. The next lesson discusses important considerations for the responsible management, collection and use of open data by all stakeholders.

Assessment

Can you think of any examples where opening data might help you answer a question, or a question that will impact your community?

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Responsible Open Data

Learning Objectives

- Recognize open data that is created responsibly
- Appreciate how to use data responsibly

Introduction

Data is a precious resource that should be shared whenever possible. As demonstrated in the previous lesson, dramatic improvements can arise from Open Data and the decolonisation of knowledge by ensuring sure data is open and available to all.

While Open Data benefits science in wonderful ways and already provides enormous benefits to society, the misuse and inconsiderate sharing of data can have far-reaching harmful effects. There may be also cases where the research data should not be collected nor shared publicly out of respect for the legal frameworks and communities needs. Understanding these potential harms requires reflection on the part of the research team and consultation with people and communities impacted by the research.

In this lesson, we introduce the concept of Responsible Open Data. These are points for consideration when thinking about making data open and managing it once it is open, as well as elaborating on ways for providing impacted communities the opportunity to drive the scientific narrative and the direct impact on their lives. In the next lesson, we will discuss a framework for actively engaging in and actioning these considerations in your research (CARE principles in lesson 4 - CARE and FAIR principles).

Empowering Individuals and Communities through Open Data

The needs of marginalized and underrepresented communities can and have been ignored with respect to Open Data. Communities that are the participants, or the main drivers of some types of data collection tend to be invisible when it comes to publishing as credit is taken by the bigger academic or institutional researchers.

Some of the notable factors that contribute to the exploitation of marginalized and underrepresented communities, oftentimes leading to disastrous outcomes including inappropriate use and sharing of data, include:

Lack of protective frameworks:

There are instances where it might not be appropriate to share data openly. For example, there are legal frameworks on a regional, national and international level to take into account; however, these might not always be sufficient to protect contributors and communities from exploitation. It is also important to note that there may be instances where no such frameworks exist, and people as contributors to the content of the data might be open for exploitation. In any case, whether a framework exists or not, careful, frequent, and ongoing communication and direct involvement of communities/contributors in any data decisions is needed, or a blanket ban should be assumed where consultation is not feasible.

Lack of proper informed consent:

Informed consent is an essential step in ethical research practices and is a responsibility for researchers to fulfill before the research takes place. Informed consent allows participants to participate fully, with a complete understanding of the research, without coercion or undue influence. This consent can be withdrawn at any time, without consequence [1]. While an exceptionally important component of science and open science in general, the exact requirements for obtaining informed consent are highly discipline specific and understanding these nuances are beyond the scope of this work.

With this in mind, it is important to understand that even if one has obtained true informed consent, it is not a once-off action. It requires consultation and education. This is important in the context of data being put online for use and reuse - especially seeing as research and its impact changes over time, and as such, communities could be opened up to unexpected harms in the future. Therefore measures need to be in place so that this consent can be withdrawn or altered without consequence to the communities at risk. This understanding needs to be ensured, as a lack of understanding can be demonstrated in the open data 1000 Human Genomes consortium's consent form [2]: the consent form has a passage most don't catch, but open themselves to biocolonialism by agreeing to have their blood samples used for an unlimited supply of DNA.

Lack of equitable participation:

Open Data that is shared with due consideration and consultation allows impacted communities to take charge and guide research in a way that best suits their narrative, values and

needs. It allows more autonomy in these communities to further their scientific development and to contribute to the larger field of open science.

Managing Research Data responsibly

Many research disciplines work with personal data that can be used to identify an individual (see [3]). This type of data cannot be shared easily, as data should be anonymized before doing so, and this is increasingly difficult in the current rapid state of development. New technical progressions may make it easier to recombine datasets and re-identify individuals. Some individuals or communities are more susceptible to exploitation, as described earlier.

The accidental detrimental effects of Open Data may extend beyond individuals and affect others; i.e., endangered species or natural resources that should be protected [4], for example; the local extinction of *Goniurosaurus luii* (Chinese cave geckos) in Vietnam was attributed to poaching activities which occurred shortly after data related to their discovery was published, this, in turn promoted a call for scrutinizing Open Data sharing practices in the field of biodiversity [5].

Additionally, research can be carried out in collaboration with industry, generating commercially sensitive data, which may place restrictions on what can be shared. Research can be used for harmful purposes (see Ethos, lesson 2) or pose a risk to (inter)national security.

There are several tools available that will help making decisions about what you can share publicly:

- CARE and FAIR principles (lesson 4)
- (inter)national laws that apply to data sharing (lesson 6 - Sharing Open Data)
- Guidelines/policies set up by your discipline or research institute (lesson 6 - Sharing Open Data)
- License restrictions (lesson 6 - Sharing Open Data)

Summary

In summary, you may not always be able to share the research data openly and there may be other responsibilities that are associated with managing the data if it has been made open. In such instances, the focus is placed on controlled and limited access with reuse in mind.

The CARE principles, presented in the next lesson provide a framework for responsibly collecting data with all stakeholders in mind. The FAIR (Findable, Accessible, Interoperable, Reusable) principles, also described in the next lesson, provide guidelines for this and allow you to share part of the data without necessarily disclosing all the data.

Assessment

- Can you think of a specific example in which releasing data could lead to harm? Which people and/or communities might you consult to determine this and discuss remedies?
- Example of how one can re-identify a person from shared data?

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CARE & FAIR Principles

Learning Objectives

- Recognise the relationship between FAIR, CARE and Open Data

Introduction

In the previous lesson on Responsible Open Data, we acknowledged that you may not always be able to share the research data openly. This lesson will introduce you to two sets of principles that provide a framework for responsible open data. The CARE principles may help you to responsibly collect and share data. If you are able to make (part of) the data openly available, it is helpful to do this in a manner that facilitates reuse by yourself and others. The FAIR principles provide guidelines for this, and allow you to share part of the data without necessarily disclosing all the data. After this lesson, you'll be able to understand the relationship between FAIR, CARE and Open Data.

CARE Principles of Indigenous Data Sovereignty

The CARE Principles of Indigenous Data Sovereignty apply whenever you're collecting data with or that belong to a particular community. The CARE principles are people- and purpose-oriented, and are originally set up to use data in a way that advances data governance and self-determination among Indigenous Peoples [1]. The principles are applicable to any research that involves communities or local stakeholders and cover:

- **Collective Benefit:** data must facilitate collective benefit to achieve inclusive development and innovation, improve governance and citizen engagement, and realize equitable outcomes.
- **Authority to control:** Recognition of the rights of (Indigenous) communities to govern data
- **Responsibility:** nurture respectful relationships with the communities from whom the data originate

- **Ethics** requires representation and participation of Indigenous Peoples, who must be the ones to assess benefits, harms, and potential future uses based on community values and ethics.

The [Global Indigenous Data Alliance](#) has made further resources available and translated the CARE principles in other languages [2]. The genomic research community has also worked on a framework for enhancing ethical genomic research with Indigenous communities [3].

Indigenous scientists have already written extensively of the harms visited upon indigenous communities through promises of medical benefits that have never materialized and sharing of genomic data without tribal consent [4, 5, 6]. Whenever you are handling data that belongs to an indigenous or other under-served community, the CARE principles are more important than the benefits of Open Data. Developments are currently underway to provide practical guidelines or ways to assess whether the CARE principles have been followed throughout the research process.

The CARE principles are complementary to the FAIR principles which were developed to facilitate data sharing practices.

FAIR (Findable, Accessible, Interoperable, Reusable)

The FAIR principles for scientific data management and stewardship are guidelines to improve the Findability, Accessibility, Interoperability and Reusability of digital assets [7]. A dataset that is FAIR is not necessarily Open. The phrase “as open as possible, as closed as necessary” [8] is often used to describe the interaction between the principles. Thus a dataset describing fishery locations might not be open (due to the harm caused by illegal fishing), but could be FAIR with a rich metadata record available and an identifying persistent ID. Datasets can be FAIR, but closed, because of personal data or because they fall under other ethical precepts that would mean opening them would be harmful (Lesson 3 - Responsible Data).

The [FAIR Data Principles](#) emphasize both human and machine readability and machine-actionability for data as research becomes more dependent on computation and automation [9]. For example a PDF version of a spreadsheet is human readable, but it is not easily used by machines. A better format for both humans and machines would be a structured data format like CSV or XML.

FAIR principles explained

- **Findable:** It is important that data is not only open but also Findable, by you and others in your field. If people from your community of practice can not find it, it will not be used frequently and its value will decline over time. Depositing your data in repositories will preserve it over time (see Lesson 6, Sharing Open Data for more on

repositories) and assign datasets with a persistent identifier (PID). Sharing data using a data repository will ensure that data are uniquely identifiable, and searchable. Another aspect that helps with searchability is having robust documentation (sometimes called data dictionaries/codebooks, metadata or a README file). Images, large files and binary data are examples of data that can not be searched by machines or humans. Providing metadata that is searchable is particularly important in these cases [10].

- **Accessible:** Once someone has found your data, they should be able to access the data using standardized mechanisms (e.g. https). Your data should be accessible (both retrievable and understandable) for both humans and machines. In other words, specify what the users need to do to access this data, and ideally, a machine can automatically translate those requirements and act on it (such as two factor authentication or request access from the author). Accessible does not equate to open. If the full content can not be made openly available, the metadata can be made openly available [10].
- **Interoperable:** During reuse, data may need to be integrated with other data, allowing machines and humans to interpret and use the data in different settings. Metadata must be detailed enough for data to be understood, especially by those who do not own or create the data in the first place. Keep in mind that people can have a hard time interpreting another person - some words can be different in spoken and formal languages; things get lost in translation, and many different terms can describe the same object. The same word can even have different meanings across various disciplines. The use of controlled terminologies, vocabularies, and ontologies for interoperability helps ameliorate otherwise substantial barriers to interoperability [10].
- **Reusable:** To be reusable, data and collections should have a clear usage license and provide accurate information on provenance. Provenance metadata provides context and details on the history of the source and its authenticity. Credit attribution (citation) is another important aspect to consider with regard to (re)usability and “paying it forward” to the researcher who released their data [10], more on that in lesson 6 (Sharing Open Data).

Summary

FAIR in short: Make your data as FAIR as possible by:**

- Depositing your data in a repository that can:
- Assign a PID
- Make sure the metadata will always be available even if the data isn't
- Using a standard data format for your domain
- Assign an appropriate license to your dataset
- Describe your data as richly as possible
- FAIR is not FAIR without due CARE

It is easier to adhere to the CARE and FAIR principles when you plan for this at the start of your research, the topic of the next lesson.

Assesment

- Consider a dataset that you contributed to. Have you followed the CARE/FAIR principles? Which of the principles can you incorporate in your workflow?
- When you reviewed datasets generated and shared by other researchers, were they following the CARE/FAIR principles? What did they do well and where could they improve?

Want to do a more extensive assessment on your knowledge of the FAIR principles? Beginners can use [FAIR-Aware](#), and if you're already more familiar you can try the [ARDC self assessment tool](#).

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Planning for Open Data

Learning Objectives

- Understand what the data life cycle is and how that affects the outlook on research.
- Understand what a Data Management Plan (DMP) and metadata are.
- Have an initial grounding on what communities to contact for support in this area.

Introduction

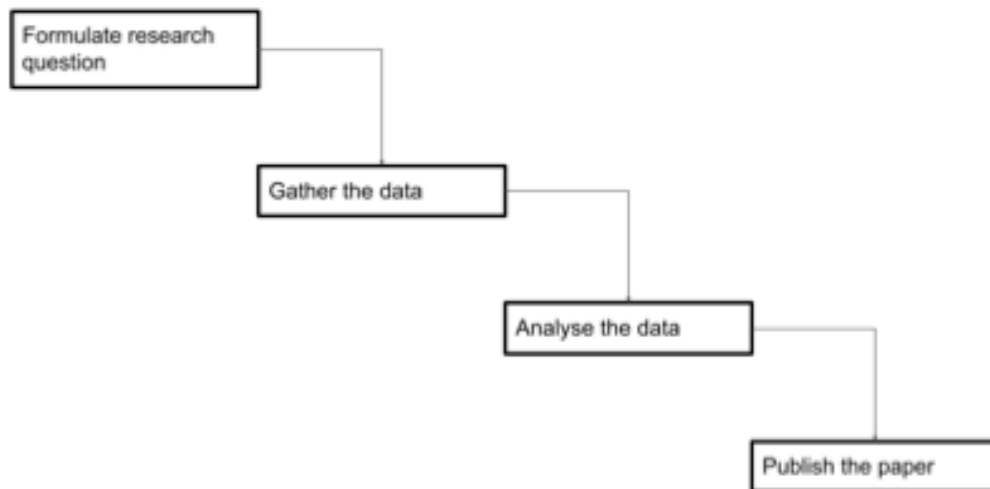
In the previous lessons it has been shown that effective open data needs to be managed. As we have seen this is not trivial and requires work and preparation. Correspondingly, there can be cost implications for your institutions to do this. Rather than facing these issues on an ad hoc basis, one should plan and prepare what you will need to do before you generate the data. With this in mind, we will

- discuss the data life cycle which places a focus on the reuse of data as it is generated.
- Introduce the concept of a data management plan, where one documents the steps that will be carried out to ensure that your data can be shared in an appropriate fashion.
- Introduce the concept of metadata, namely documenting your data which is essential if another researcher is to make use of your data.
- Finally, who to contact in terms of advice and support.

Planning

The data life cycle

With a focus on generating papers, a researcher implicitly ended up with the following research workflow model in mind of how they worked with their data. It's important to note here that because the focus is on the paper, there's no thought to how the data changes at different stages of the process, or thought to how the data should be managed after a paper is published. Usually the data were included as part of the paper as a supplementary file.



This figures illustrate a Linear workflow model for data life cycle.

On the other hand, if one thinks of open data that can be FAIR (and thus reused) then this model emerges. In particular that Data needs to be available beyond the publication of a paper. Data no longer has to be associated with one paper. Data can be reanalysed. More data, from different sources or the same lab, can be added in at any time, including later. Instead of the process being a linear progression, with a start and a finish, the process for data becomes more complex and there is cycle. These ideas were put together in the [DCC Curation Lifecycle model](#) [1]. The original life cycle is complicated but a summary of the life-cycle is listed below

The DataOne Data life cycle, as listed in the data management plan by the [University of Rhide Island Library](#), described these steps in the Data Life-Cycle: Plan, Collect, Assure, Describe, Preserve, Discover, Integrate, Analyze.

Here the focus is very much moved away from the idea of research -> publication and instead is on the data itself as a first class research output.

Let's look at these individual steps:

- **Plan:** a description of the data that will be compiled, how the data will be managed and made accessible throughout its lifetime.
- **Collect:** this corresponds to the data gathering step (illustrated [here](#)). It can include both primary (raw) and processed data.
- **Assure:** the quality of the data is assured through checks and inspections.

- **Describe:** data is accurately and thoroughly described through documentation (e.g. metadata).
- **Preserve:** these are the steps necessary to make sure that the data will be accessible going forward so in particular ensuring that the data is stored in a fashion that others can use it (in particular storing at a data repository). Ideally this should be done in a fashion that matches the CARE and FAIR principles (lesson 4). This may also include the step of removing data that may not be of use to future researchers. For example, high resolution images may no longer be themselves useful if in the analysis step one has extracted the features of interest from them. Not storing the high resolution image and simply storing the feature data would provide a considerable saving of storage.
- **Discover:** here other researchers can extract either the entirety or some subset of the data for their own purposes.
- **Integrate:** data from disparate sources are combined to form one homogeneous set of data that can be readily analyzed (this could include this one data set being analyzed).
- **Analyze:** corresponds to the data analysis step as illustrated in [here](#). There are a variety of different interpretations of the data life-cycle (see the reading list for this lesson) with varying degrees of complexity. It's also important to note that this is an idealization of what goes in general. Nonetheless, it is important to think of all these steps as an ongoing, interactive process that requires thorough planning and continued consideration and to recognize that they are non-trivial to do.

Data Management Plans (DMP)

Seeing as the above steps are not trivial before one begins to gather, collate or generate a data set it is useful to plan out what you will do with the data. This is referred to as a Data Management Plan or DMP for short.

A DMP means that you can think ahead of any particular issues that might crop up in terms of handling the data, such as the potential cost of storage, whether data needs to be anonymised and so on.

A detailed description of what one should put into a DMP is described [here](#) [3]. As outlined in this [document from the UKRI](#) [4], the central funder for the UK, these can include answering questions such as

- What type of data will be generated or preserved? This could include data formats, rough estimates of the amount of data to be stored during a research project and similarly what will be preserved beyond the lifetime of the project?
- What type of metadata will be used and preserved. It is worth noting that one of the more detailed aspects of the FAIR principles is to keep the metadata of the data set available even if the original data set no longer exists.

- Where should the data be preserved? i.e. what repository will be used (repositories are discussed in the next lesson). How long should it be stored? (five years? ten years?) More concretely, data regulations can require that certain data be kept in certain ways for at least a certain amount of time. This will vary depending on the type of data (e.g. medical records, population statistics). It is advised that these expiration dates are explored in the literature, and/or policy guidelines.
- How will any private data be stored so that it is kept securely?

DMPs are not meant to be exhaustive documents! Typically they are 1-2 pages of A4 and often are less than a few thousand words. The important point is that they sketch out what a researcher or research team plans to do with their data well before they are gathered and can identify any steps that need to be taken rather than facing a major challenge now.

DMPs are [increasingly used by funders](#) and their institutions as a means to have researchers map out what they will do with their data in a research proposal. Research proposals often require DMPs, and hence DMPs are often the ‘sharp end of the stick’ for researchers with respect to Open Science [5]. A good DMP is a criterion for assessment in grant applications and hence doing a good DMP will help your grant be funded.

Documenting your Data (Metadata)

As discussed in the previous lessons, the FAIR principles emphasize the importance of meta-data, namely documenting your data. Metadata is described in more detail [here](#) [6].

A perennial question is what type of metadata and description of the data should be provided for a data set. If you are dealing with electronic data should one provide metadata for a whole set of files, an individual file ... each individual bit?

The simplest rule of thumb is if there aren’t any guidelines for your type of data or domain repositories, then try and provide enough documentation about your data that you would ask for if you were downloading this data yourself.

For example if this was data taken from a field trip where location is important then you might want to include longitudinal and latitudinal coordinates. If it’s data from a wet lab then it might include parameters you normally include in the materials and methods section of a paper. If it’s data from purely computational work you may want to list the software run and the parameters used.

Data repositories will be discussed in the next lesson. Domain specific repositories will often give more precise requirements on metadata (another reason to use them).

If there are no guidelines then a simple README file attached with the data is a start (for an example see [here](#)) - though it’s important to note that ideally one should use metadata schema

which is described in much more detail [here](#) as FAIR data should be machine-actionable [7] [8].

Help

Much of the ins and outs of dealing with Open Data, or more particularly Open Data that follows good practice such as the FAIR principles, can be technical and lies beyond the domain of knowledge of researchers. How does one navigate this landscape?

This can be summarized in the following diagram -

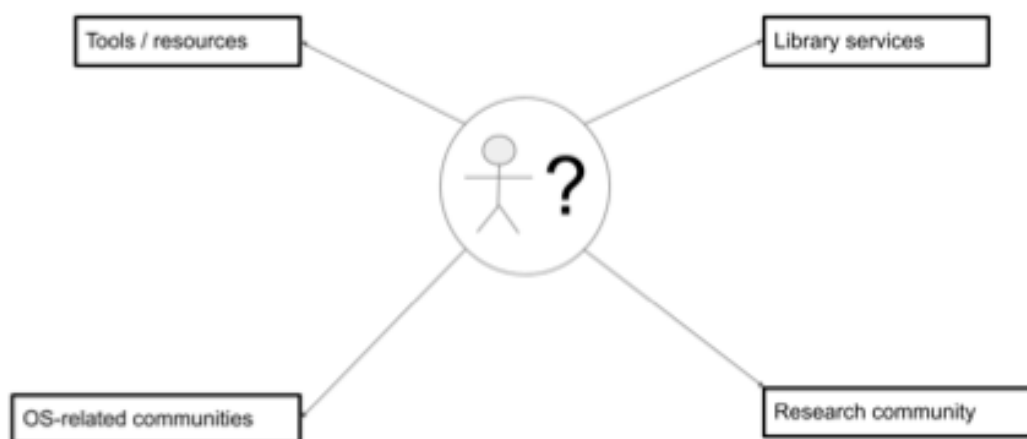


Figure 1: Figure 5.3 Diagram pointing to four possible sources of information a researcher can approach.

“Figure 5.3 Sources of information and support on Open Data that a researcher could access.”

Figure 5.3 Sources of information and support on Open Data that a researcher could access.

Research communities (international and national)

Individual research disciplines may already have put together materials and have advice on how to implement Open Science in their discipline. For example [FAIRsharing](#) is a educational and information resource on data and metadata standards [9]. The [Research Data Alliance](#)

have a variety of different [interest and working groups](#) in data sharing in specific disciplines. Scientific Societies and Publishers can also provide advice [10] [11].

Open Science related communities

There are a number of communities that are focussed on Open Science activities. [ReproducibiliTea](#) is a grass-roots journal club initiative that is based in over 100 institutions and is a forum to discuss reproducibility, closely allied to Open Science [12]. The [FAIRdata forum](#) allows you to browse materials and raise questions that are related to FAIR [13]. Correspondingly the [PID forum](#) allows you to ask questions on PIDs in general [14]. A list of Open Science communities is provided in the next module (Open Tools).

Tools and resources

Finally, there are a range of different tools to help you. For example, [DMPtool](#) and [DMPonline](#) allow you to build your own DMPs [15] [16]. See the module Open Tools for more details. There are a variety of different catalogs out there one can use to search for materials in this area. [Shanahan, Hoebelheinrich and Whyte](#) (2021) have a table of catalogs to search for materials [17].

Local library or IT services

The long term vision is that Higher Education Institutions (HEIs) or Research Performing Organisations (RPOs) [will employ data professionals to advise and support researchers](#) [18]. These individuals have a variety of possible job titles such as Data Librarian, Data Steward, Data Curator and so on. These individuals would advise on aspects on how to make your data adhere to the CARE and FAIR principles, providing appropriate metadata and so on. Some HEIs/RPOs have already made Open Science (or Open Research) policy statements and may not yet have an infrastructure to help but will be interested in supporting you. In some countries there has been progress in this area but it is very early days. Nonetheless, it is worth contacting your University library as they may be able to advise you even on relatively small questions or requests.

Summary

Making data open is not trivial. It is not simply a matter of placing a data set onto a cloud drive. Nonetheless, if it is done correctly then the open data is available for reuse. Reuse can be a completely different research team or it could be the same research team that need to carry after a member of the team responsible for the data has moved on. This means one

has to think of the data as part of life-cycle and that it is important to make plans (a Data Management Plan) prior to creating the data to ensure that it is stored appropriately. Part of making your data FAIR is provide metadata that describes the data that you are depositing. Finally, do not feel that you have to do all this from scratch. There are a variety of different avenues that you can approach, either on an online basis or sometimes on your own campus.

Assessment

Think about the data sets that were described in lesson 1 as examples of good data.

- Can you identify what were the above steps with that data?

Think now about a data set in your own discipline.

- What would be the steps that you would need to take with that data to match up with the data life cycle?

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Appendix: Finding Open Data

The reusability of openly shared data relies on the prospects of it being found in the first place, therefore data findability is a key step in accessing and utilizing data. There are three major ways to find Open Data that are shared by researchers – repository, web search, and literature search.

Repositories

Ideally, Open Data should be available in repositories where the datasets are properly indexed and assigned a unique persistent identifier (as discussed in **Lesson 6 – Sharing Open Data**) thereby ensuring the data is unambiguously identifiable, searchable, discoverable along with associated metadata and documentations.

Therefore, the first step in finding Open Data related to your field is to identify discipline specific repositories (if there are any) and search for datasets there (see **Lesson 6.4 – Repositories and Other Sharing Methods**).

Find repositories in your field:

- *[Re3data.org](#) is a global registry of research data repositories that covers research data repositories from different academic disciplines.*
- *[FAIRsharing](#) is a curated, informative, and educational resource on data and metadata standards, inter-related to databases and data policies.*
- *Recommended repositories by publishers (e.g., Recommended Data Repositories suggested by [Scientific Data](#) and [PLOS One](#))*
- *[World Data System](#) represents a network of repositories.*

Examples of generic repositories:

- [Zenodo](#)
- [Mendeley Data](#)
- [Figshare](#)
- [Dryad](#)

The [Generalist Repository Comparison Chart](#) is a tool you can use to decide where to store and share their FAIR data outside of their institutional repositories. Dataverse has also published a [comparative review of eight data repositories](#).

Web-searches

To explore a wide variety of datasets from projects or popular topics, the use of a more general search engine can be helpful. Some disciplines or large institutions such as NASA and the National Institute of Health's National Center for Biotechnology Information (NCBI) offer their own portal where you can search for their datasets, related publications and oftentimes tools for analysis (e.g., EMBL's European Bioinformatics Institute <https://www.ebi.ac.uk/>). There are also an increasing number of international and national data portals to enable data discoveries.

Generic data search portals:

- Google <https://datasetsearch.research.google.com/>
- Kaggle <https://www.kaggle.com/datasets>
- Wikidata https://www.wikidata.org/wiki/Wikidata:Main_Page
- Open Data Network <https://www.opendatanetwork.com/>
- Awesome Public Datasets <https://github.com/awesomedata/awesome-public-datasets#readme>

Examples of Discipline specific:

- NASA Earth <https://www.earthdata.nasa.gov/>
- Cern <https://opendata.cern.ch/>
- NCBI National Center for Biotechnology Information <https://www.ncbi.nlm.nih.gov/>
- EMBL's European Bioinformatics Institute <https://www.ebi.ac.uk/>
- ISPCR <https://www.icpsr.umich.edu/web/pages/>
- International Monetary Fund <https://www.imf.org/en/Data>
- NOAA Climate Data Online <https://www.ncdc.noaa.gov/cdo-web/datasets>
- Federal Reserve Economic Research <https://fred.stlouisfed.org/>
- USGS EarthExplorer <https://earthexplorer.usgs.gov/>
- Open Science Data Cloud (OSDC) <https://www.opensciencedatacloud.org/>
- NASA Planetary Data System <https://pds.nasa.gov/>

Examples of National or international data portal

- US Federal data <https://data.gov/>
- EU Data Portal <https://data.europa.eu/en>
- WHO <https://apps.who.int/gho/data/node.home>
- THE WORLD BANK <https://data.worldbank.org/>

- DATA.GOV.UK <https://www.data.gov.uk/>
- UNICEF <https://data.unicef.org/>

Literature search

While not ideal, datasets are often attached to scholarly publications in the form of supplementary material, or referenced in text where to find them e.g. GitHub repository or personal/institutional websites. In addition, there are emerging journals and special collections/issues focused on describing and publishing data (e.g. Nucleic Acids Research database issues <https://doi.org/10.1093/nar/gkab1195>, Scientific Data, Earth System Science Data, etc.). In other words, while the datasets are openly available in these media, they are not properly indexed and therefore not very findable nor machine readable.

Finding academic publications can be a challenge in itself depending on the discipline and field of study. For instance, in life science and biomedical research, there are a number of repositories and search engines (e.g. PubMed, EuropePMC) indexing research outputs (e.g. publications, abstracts, references and communications) from various journals.

However in other disciplines (e.g. arts and humanities), search is often carried out with general search engines or research databases such as Google Scholar and JSTOR. In that case, it is advisable to reach out to library personnel and community members for further advice on where to find related literature and data, see lesson 5.4 Help section.

Generic:

- Google Scholar <https://scholar.google.com>
- Open knowledge map: A visual interface allowing the exploration of interconnected topics with relevant documents and concepts. <https://openknowledgemaps.org/>
- JSTOR a wide range of scholarly content <https://www.jstor.org/>
- ResearchGate <https://www.researchgate.net/search>

Discipline specific:

- EuropePMC Life sciences <https://europepmc.org/>
- Pubmed biomedical literature <https://pubmed.ncbi.nlm.nih.gov/>
- arXiv is a free distribution service and an open-access archive for scholarly pre-prints in the fields of physics, mathematics, computer science, quantitative biology, quantitative finance, statistics, electrical engineering and systems science, and economics <https://arxiv.org/>
- Biorxiv Preprint server for biology <https://www.biorxiv.org/>
- EarthArXiv (<https://eartharxiv.org>) and Earth and Space Science Open Archive (<https://essoar.org>)
- ASAPbio provides a catalog of preprint servers <https://asapbio.org/preprint-servers>

Open Results

Welcome to the Open Results Module!

Recap: In Open Ethos, we learned about the ethics and principles underlying responsible open science practices. In Open Software, we explored and identified the right tools and methods that allow us to ensure reproducibility through version control, code testing, workflow, and a virtual research environment. In Open Data we developed a data management plan that can ensure the Findability, Accessibility, Interoperability and Reusability (FAIR) of our data throughout the research process, and not just at the end when the final report from the project is released.

In this module, we will explore the different stages of the research process—including identifying the different types of Research Objects in a study and the various ways in which they can be shared and disseminated as open results. We will define a Research Object and provide an overview of how they relate to the research lifecycle (Lesson 1). Specifically, we will discuss the different stages of the research process, from ideation and planning all the way through and beyond dissemination. Then, we will consider how these Research Objects can be shared (Lessons 2-3). By the end of the module, we will have looked at the important concepts and practices for publishing and sharing research components before, during and after the project. Lastly, we address ethical contributorship, – making sure collaboration is fair and inclusive, and that credit is assigned transparently and equitably (Lesson 4).

Objectives:

1. Identify research stages and elements of research objects that can be considered results
2. Identify the guiding practices and principles related to open results and the advantages of implementing them across stages of a research process
3. Identify paths for publicly communicating results
4. Create open results contributor guidelines and opportunities for open and equitable collaborations
5. Give credit to contributors in open results
6. Contribute and provide constructive feedback to others' results
7. Apply open result principles to new and ongoing research projects

Overview and key messages

This module addresses different questions discussed systematically across the following four lessons:

Lesson 1: The Research Process and Its Results

1. What are the different stages of the research process?
2. What are “Research Objects”?

Lesson 2: Results in the Context of Open Science

1. What are the advantages of making results open throughout the research process?
2. What resources are available to help make results open?
3. What are the guiding principles to turn a research result into an open result?

Lesson 3: Applying Open Result Framework to your Research

1. How can you apply an open framework across different research objects?
2. How can you share your results, and select **tools** that support open science?
3. Using a checklist to achieve open results

Lesson 4: Providing Equitable Opportunities and Credit for Contributors to Results

1. How can you define contributors to each digital research object and determine their suitable form of recognition?
2. How can you create contributor guidelines that ensure equity, access, inclusion, and diversity?
3. How can you ensure your open results are properly attributed and cited by others?

The Research Process and Its Results

Introduction

With the overarching goal of maintaining research integrity and ethical practices from the start, we need to consider reproducibility methods, collaborative approaches and transparent reporting for the research teams to ensure that all results can be replicated, validated, and built upon by other independent researchers. As researchers, this means: 1) broadening our perspectives regarding what shareable research outputs are produced throughout the research process, 2) providing sufficient documentation that describes the research workflow and the decision-making process, and 3) publishing all research outputs that would eventually enable others to validate the research findings.

Before we can begin to do that, we need to define what we mean by the research process, and what we consider research outputs at various stages of our research. Accordingly, this lesson will enable you to answer two questions:

1. What are the different stages of the research process?
2. What research objects can be considered a result?

What is a research object?

A **Research Object (RO)** is a method for the identification, aggregation and exchange of scholarly information on the Web [[Garcia-Silva et al. 2019](#)]. RO can be composed of both research data and digital research objects that are defined as follows by Organisation for Economic Co-operation and Development ([OECD Legal Instruments](#)).

Research data consists of “*factual records (such as numerical scores, textual records, images, and sounds) resulting from research that is partially or fully funded by public funds, used as primary sources for scientific research, and that are commonly accepted in the scientific community as necessary to validate research findings.*”

A **research-relevant “digital” research object** consists of any “*metadata, algorithms, workflows, models, and software (including code) resulting from research that is partially or fully funded by public funds, which are used in a research and development context.*”

Research Objects are often given an identifier. In this way, there is a mechanism to trace back related resources about a scientific investigation. The most important aspects to consider about ROs:

- They are not only associated with the end products as publications and final reports but also encompass research outputs created, revised and shared throughout the research lifecycle that help validate findings claimed in scholarly publications. More simply, ROs apply to any “single information unit” or research material that can be **shared and cited** with other scientists within and outside the project.
- Motivation behind RO is the need to identify and share all components such as data, source code, tools, and method documentation, as well as communication materials such as presentations, videos, blogs and other tangible outcomes.
- ROs facilitate reproducibility and reuse of the scientific methods and results through access to resources, context and metadata
- ROs help us to understand the entire research lifecycle through research outcomes including publications shared progressively. They also allow us to track the versioning and development of the entire project.

Ultimately, there are three guiding principles for ROs [\[reference\]](#):

1. Digital identity - Using unique identifiers, such as DOIs (link to data) for tangible outcomes such as publications or data, and ORCID ids for researchers (explained in detail in the next lesson). This enables others to cite and use individual components of your work.
2. Data aggregation - Using a method to aggregate all outcomes so that they are discoverable and hence allow anyone to investigate and reproduce the research.
3. Annotation - Use rich machine-readable metadata (discussed in open data) that help ensure the findability and accessibility of all scientific work.

Figure 1: *Research Objects allow working open by design and share during the research process and not only the research outputs at the end. The Turing Way project illustration by Scriberia. Used under a CC-BY 4.0 licence. DOI:10.5281/zenodo.3332807.*

Following from these we can now build a definition for an **Open Result**.

An **Open Result** is all the research outcomes, including successful products, reports on potential risks, experiments that worked as well as failed, or any other information such as experimental protocols, standards as well as all the individuals who contributed to the research can be recorded in the RO and shared as open results.

What are the different stages of the research process?

In previous modules, we have learned the fundamentals and practical concepts for planning our research for open science. Specifically, in the Ethos of Open Science [\[addlink-ethos\]](#) module,

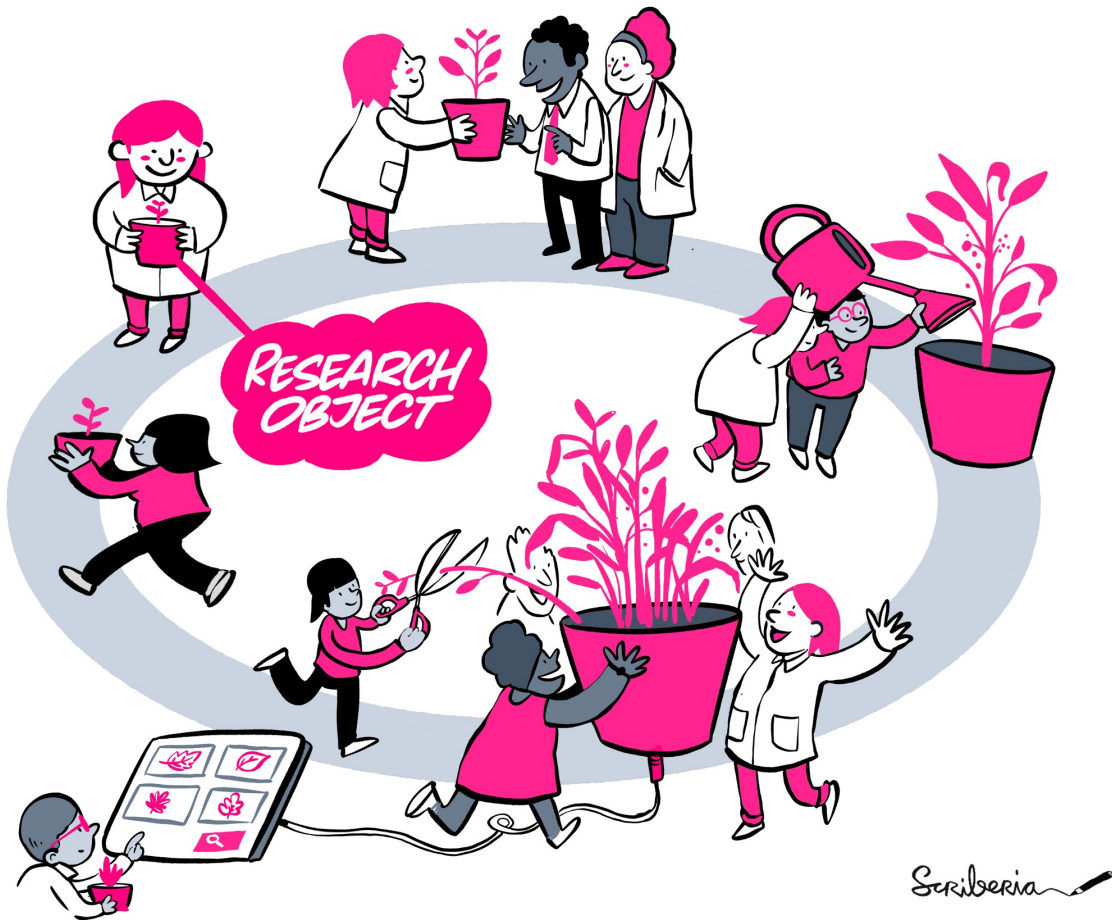


Figure 1: This image shows how research objects evolve and grow in content during the collaboration process and how new research objects can be derived from existing ones.

we learned that open science should be considered throughout the research process, and not just at the time of publication. With this understanding, when considering shareable research outputs, it is important to think about the entire research life cycle – different tasks carried out during the life cycle of a research project.

Many of us might be very familiar with the research life cycle but may not have considered what results could be shared openly throughout the process.

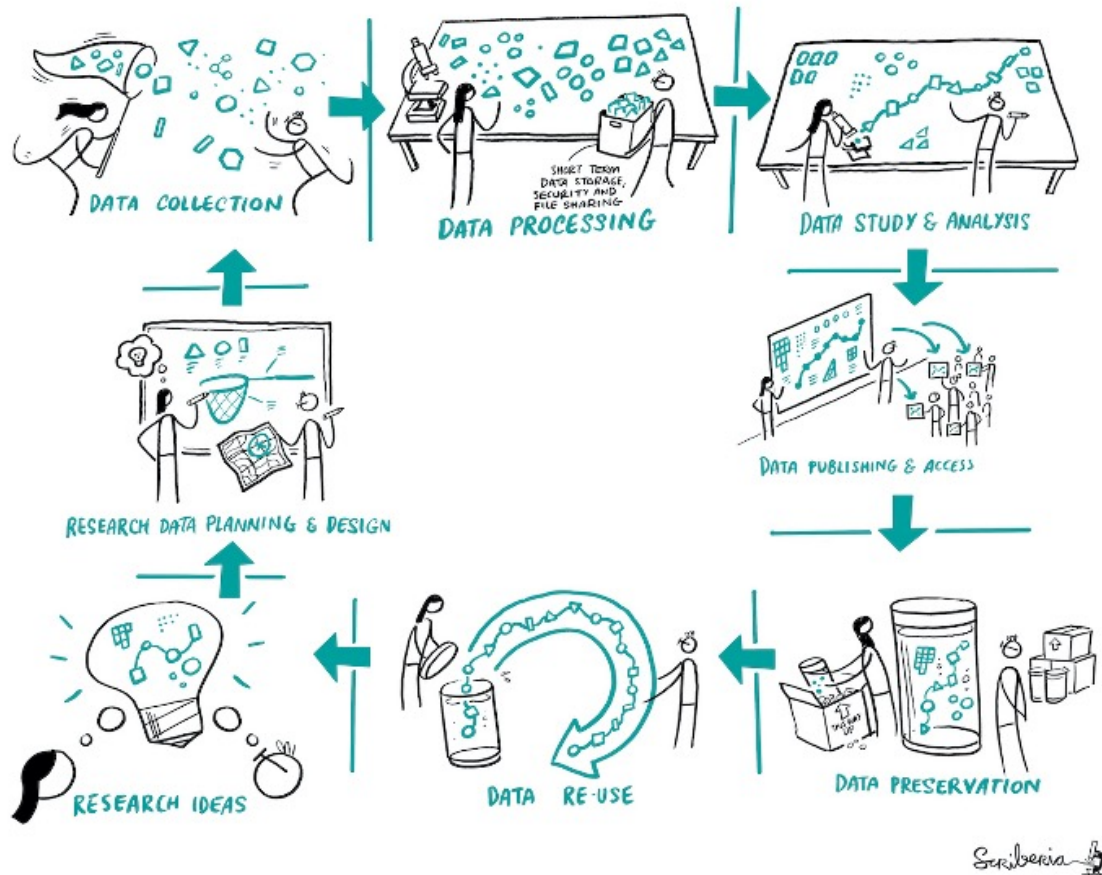


Figure 2: The research process is represented as a perpetual cycle of generating research ideas, performing data planning and design, data collection, and data processing and analysis, publishing, preserving and hence, allowing re-use of data.

Figure 2: *The Turing Way* project illustration by Scriberia. Used under a CC-BY 4.0 licence. DOI:[10.5281/zenodo.3332807](https://doi.org/10.5281/zenodo.3332807).

There are many ways to describe a research life cycle, but in this lesson, we define it in *nine* distinct phases based on Figure 1 *The Turing Way* that builds on various published examples.

Conceptualization/Ideation

In this stage, we focus on outlining and describing the research idea to different collaborators, students and/or postdocs. This could also encompass proposal writing, obtaining ethics approval documents, and/or securing funding.

Planning

In this stage, we are thinking about project management and workflows. Who is needed for the research project to be successful? During the planning phase, collaborations are often extended beyond close collaborators. Methods of collaboration are often defined, including team member roles and responsibilities.

Project Design

In this stage, we are concerned with describing the research protocols. For example, what the research hypothesis is, what protocols will be used to conduct the study, how will the data be collected, and processed, where will it be stored, and more.

Data Collection

In this stage data collection (from publicly available databases or resources) or data generation (through experiments or quantitative/qualitative studies) commence. See Open Data [addlink-data].

Data Wrangling and Processing

At this stage, we use existing software or write custom code to process the data that has been collected. See Open Software [addlink-software].

Data Exploration and Statistical Analysis

At this stage, we combine the workflows from Stages 4 and 5 and begin using our tools, code or software to analyse the data that has been obtained.

Reporting

Here we report on our findings, in other words, we share them with the research community. This can be done in the form of a research manuscript first published on a preprint server and in a peer-reviewed journal. However, reporting now far exceeds publication alone. Reporting also encompasses presentation materials (such as posters, and slide decks), lab websites or blogs, outreach materials for social media, podcasts or press releases, and many more.

Preservation and Reuse

In this stage, we consider archiving all outcomes for long-term preservation. This ensures that our research is accessible, and reusable, meaning that someone else can go through this whole process of reproducing or building upon our work.

Scientific Engagement, Training, and Feedback (cross-cutting)

In this stage, we conduct effective collaboration through active engagement, skill development and peer-review processes for both direct and indirect stakeholders of our research.

Important note: Although we describe these stages in sequential order, these stages may not always be linear. For instance, scientific engagement and data management efforts will be applied at all stages of research. Data exploration, analysis and reporting will be an iterative process, and reporting will happen at different points of the research lifecycle. Even before the study begins, research questions, hypotheses, and planned approaches may be openly reported or preregistered [Nosek et al. 2018]. Preregistration differentiates research outcomes which are the results of predictions, which occur before data collection, from predictions, which occur once the results of the data are obtained.

To build high-quality research outcomes, it is essential that everyone (1) can work together efficiently at all stages of the project, (2) has a shared understanding of how results from their work will be shared with each other, and more broadly beyond the project, and (3) gets fairly recognized for all their contributions.

What research objects are commonly associated with research stages?

Now that we understand the different stages of the research lifecycle, Research Objects and open results, we can expand on how they operate in the context of the research lifecycle. The most important outcome to consider is that these ROs can be produced throughout the research lifecycle and should be published throughout, rather than at the end of the research process.

Research stages and open result table

Research Stages	Possible research objects as open results
Conceptualization and planning	Proposal, ethics approval document, budget/funding plan, contributor and partnership plans (see lesson 4 [addlink]), preregistration reports, research materials, research protocol
Project design	Versioning system, shared project repository, project planning document (project goals, roadmap, ways of working, roles and responsibilities, communication), hypothesis and pre-registration, collaboration plan, Equity, Diversity, Inclusion and Accessibility (EDIA) guidelines, data management plan, metadata standards, governance plan, data safety and security guide
Data collection	File formats and data types, parameters/dimension, test data, metadata, data access plan/details, raw data
Data wrangling and processing	Statistical methods, tools, workflow and analysis pipeline, processed data, code for data exploration, statistical results
Data exploration, statistical analysis	Notebooks, figures, code, software package (R package, python library), code documentation, models, technical reports on scope and limitation of data, configuration and virtual research environment
Engagement, training, and feedback from peers (communications and collaboration)	Contribution guideline (feedback documents, process for inviting feedback), review sprint plan and outcomes, departmental and conference talks, user testing information, tutorials, executable notebooks, videos
Preservation and reuse (Research Data Management)	Data management plan with the versioning system, metadata standards, data governance and archiving plans, data sharing and archiving information, code packages, virtual research environments, hardware (if produced), physical samples

Research Stages	Possible research objects as open results
Reporting, publication	Posters/figures, talks/slides, preprints, journal/book publications, layman summary, lab website/blogs, outreach materials for social media, podcast/press release, containers for testing (Docker, Binder), documentation and manuals, research compendia, configuration files (for reproducibility), software release information, hardware plan and associated documentation

Contributions that are not Research Objects but should be considered as results and recorded openly

Research, like most technical professions, involves different kinds of contributions that do not always result in tangible outcomes and hence, can't always be defined by RO. For example, responsibilities associated with maintenance of RO, community management, data stewardship, library and archiving work, "Equity, Diversity, Inclusion and Accessibility" (EDIA) efforts, as well as tasks associated with funding, project management, scientific event organization, training activities and more. Outcomes from these roles cannot always be accurately captured besides documenting their processes, methods and impact, often recorded by some people involved in those roles. In Lesson 4, we discuss how to properly acknowledge the contributors to your results.

Assessment #1: Identify the research objects in your project or a case study

Invite project ideas from the learners and the broader open science community before delivering the training.

Self-assessment #2: Identify the research objects to be shared as open results of a project you are/were involved in

Provide an empty version of the "research stages and open result table" table to be filled by the learners.

Conclusion

The research consists of many different stages, each with several important tasks. In the early stages, we deal with Conceptualization and Planning. This can include a number of different things - depending on the project - but typically involves the development of a study protocol, research questions, and other study materials. Next, comes Project Design. In this stage, we often focus on developing a study timeline (or roadmap), assigning different roles to project team members, creating data and metadata management plans, and planning for data collection, management, and security. Next, is the active responsibility for Data Collection. Taking a step back from a project can help us establish an understanding of this multifaceted process and give us an appreciation of all the important elements (and people) involved in bringing a project or study from conceptualization through to completion and dissemination. In the next lesson, we will consider the advantages - for ourselves and the broader scientific community - of making our results open and transparent. In doing so, we will explore best practices for transforming our work from closed to open.

References

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Results in the Context of Open Science

Introduction

In the previous lesson, you learned that the results of a research project encompass much more than just a published paper. In this lesson, we will demonstrate the benefits and challenges of making your research results open.

You will learn that making results open entails making them findable, accessible, interoperable, and reusable (FAIR) while caring for both people and purpose. To this end, we will discuss available guiding principles to enhance the usefulness of your research results for you as a researcher, for your research team, for your collaborators and for society in general. Applying these principles requires key changes in the practice and culture of research and the implementation and normalisation of certain technologies and practices that will be covered in the next two lessons as well.

Some research projects produce sensitive research objects that cannot be shared due to ethical, legal, technical or institutional reasons. We will discuss how your research project can be reproducible and collaborative without necessarily having them all open.

What are the advantages of making results open throughout the research process?

In the Ethos of Open Science module, we discussed the general benefits of Responsible Open Science [addlink-ethos]. In this section, we will link how these advantages pertain to each Research Object (RO) learned in Lesson 1 [addlink-results1]. In order to simplify the discussion we will merge the possible ROs in four big categories:

- **Preparation documents.** This category includes all outcome ROs from the research project planning phases, for example, ideation & conceptualization, planning and project design
- **Datasets.** This includes raw or processed datasets from the following research stages: data collection, data wrangling and processing, and preservation and reuse. *Additional information about the advantages of making data open can be found in the Open Data module [addlink-data].*

- **Software.** This refers to all the software created and used in all research stages, in particular: data collection, data wrangling and processing, data exploration & analysis and Preservation and reuse. *Additional information about the advantages of making software open can be found in the Open Software module [addlink-software].*
- **Reports.** This category includes all ROs associated with communicating results within the research group or/and outside, e.g Communication, reporting and publications

The main identified advantages of making results open are the following:

1. **Avoids duplicating efforts.** *This is important for all types of ROs.*

For example, a single dataset can be analysed in multiple ways. Another example is that the same implementation of the data processing, exploration and analysis stages (including but not limited to analysis pipeline, statistical methods, tools, and software) can be reused for another phase of the same project or for a new project without the need of reimplementation by each researcher.

1. **Saves time and increases efficiency.** *This is important for all types of ROs.*

If the research project is open from the start, it can help you to be more efficient and save a considerable amount of time (see the Ethos for Open Science module for “planning for open science” [addlink-ethos]). First, having the preparation documents open will guarantee that all members of the team have at hand the information about the project design and planning big picture. Second, you save time when you are required to share your dataset, methods and software with funders and publishers. Third, an open workflow creates efficient pipelines from the start. Fourth, open ROs from the Engagement, feedback and reporting stages can also significantly improve the review process by validating the results available at each research stage within or outside your team. This improves replicability, as independent researchers can replicate and confirm the results at each step. Good and open documentation of data, codes and scripts, protocols and intermediate results will speed up writing your final papers/publications.

1. **Facilitates collaboration and onboarding of new members.** *This is also important for all ROs.*

Collaboration will be much easier when preparation documents, datasets, methods and software are open and well-documented. Having user testing, tutorials, executable notebooks and videos from the “Engagement, training, and feedback stage” will be additionally important for **onboarding** new members of your team or external collaborators. Your research project will be easier to be continued by you (even after some changes in the composition of your group) or by a different research group.

1. **Allows collaborators to receive credit, and provides incentives for others to contribute.** *This is important for all kinds of ROs.*

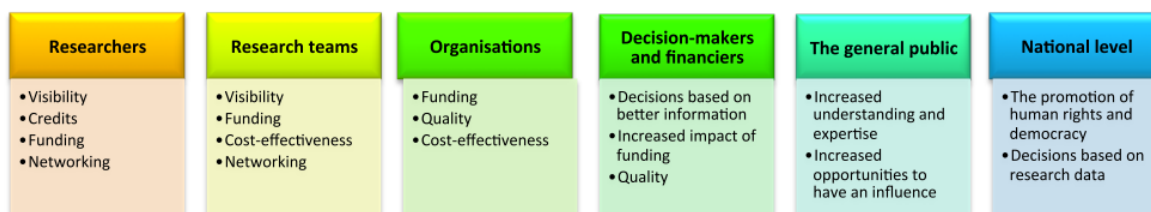
Making your results open also opens you up to clearer ways of **receiving credit** and can also reduce the risk of scooping (each result can be individually referenced as soon as made available). Applying reproducibility practices separately on different parts of the project such as **Preparation documents, datasets, software and reporting** allows other researchers to test and reuse your work in their research, and your research will be more cited thus bringing fair recognition for your work. Collaborators can get more **motivated to contribute** because they can easily get recognition in terms of authorship for their contributions made for each one of the ROs generated.

1. **Furtheres the reach and audience of our results.** *This is particularly related to the Communication and collaboration, reporting and publishing steps*

Open posters/figures, talks/slides, preprints, webpages, and journal/book publications will allow more members of your academic community to access your research, which in turn can turn into more collaboration and recognition and a greater impact of your research results. But the impact can be extended outside the academic community as well by also making available public summaries, lab websites/blogs, social media, podcast/press releases, and citizen science projects among others which can strengthen the link with the local community and enrich your research.

1. **Funding.** Since more and more funding agencies are paying attention to open science and requiring applying guiding open science principles to the research project they fund, open practices will make you **eligible for more funding opportunities**.

The picture below summarises the most significant advantages for all actors in the research ecosystem of making your results open.



(reference: @factorsoecd)

What are potential obstacles and what resources are available to help overcome them?

Overall Potential Obstacles

Along with all the clear benefits of open results, there is a suite of real and perceived challenges. These obstacles were described in the Ethos module [addlink-ethos] and can be divided into two categories:

- External obstacles: **cultural barriers** (lack of support and recognition from your institution), **disagreement** between collaborators involved in the research on what results to make open, **legal** and **security** considerations;
- Internal obstacles: investment of **time** and **effort** needed upfront to make your results open including the need to learn additional **skills**, afraid of scooping, the lack of **funding** for instance for open-access publications and curating research results.

These obstacles apply to any Research Objects (created by you or your collaborators), for example, data, software and documents, reports and publications.

Let's focus on results that are not software or data. You can check the Data [addlink-data] and Software [addlink-software] modules for any questions specifically related to open data and software.

One of the major steps in research is the communication of your ideas and insights from your work into a clear, open, and accessible format that can help key stakeholders make meaningful decisions. Mastering different ways to convey your insights responsibly is challenging and can be seen as a diversion to your research work: blogging, writing public summaries, podcasting, presenting posters and talks at conferences outside of your discipline, and using social media requires communication skills that are not usually taught to students, and can be extremely time-consuming while the impact is very difficult to measure.

Obstacles and Recommendations for Open Access Reporting

The most common way to communicate research results is by far their publication in journals. However, the choice of a journal or a publishing platform may affect the availability and accessibility of the research results.

Open Access publications allow you to make available articles and/or books accessible online, free of charge to the public without any restriction (no mandatory registration or log in to specific platforms required). Publishing your results in Open access journal is a good way to increase your research impact and allows everyone, including society as a whole, to use your research results.

However, publishing open access can incur additional costs that may not be covered by your research grant or institution. Before publishing to a journal, check if there are institutional open access agreements in place: the costs are usually significantly reduced and sometimes you may be able to publish Open Access with no additional cost. Several publishers also offer waivers and discounts to researchers living in low- and middle-income countries. Other cost-offsetting programs may be available too: for instance, some publishers have a fee support program to ensure that accepted articles can benefit from open access.

In many disciplines, there are also Open Access Journals where the content is open for everyone, with no need to pay or be a member of a subscribing institution. The number of these journals is still increasing rapidly and you can search open access journals and articles in the Directory of Open Access Journals [DOAJ](#).

You may choose to self-archive your research results to make them more discoverable and/or after you've published them in a subscription journal to ensure there is an open version of your paper. Preprint servers are also increasingly popular: you can deposit documents that have not been peer-reviewed in a traditional journal-led process but are considered a complete scientific publication in the first stage. Some of the preprint servers include open peer review services and the availability to post new versions of the initial paper once reviewed by peers. Preprints can be used to share a paper before it is submitted to a journal and can reduce scooping. Many publishers accept this as a standard practice but some may reject papers which have been shared in preprint form. It's therefore important you read carefully the publisher's policies when considering submitting a paper.

Finally, in most cases, you can also self-archive your publications in repositories, including [Zenodo](#). It is however recommended to check if the journal has any specific self-archiving policy. Your institution may also have an institutional repository. Check the [Registry of Open Access Repositories](#) to search and find an up-to-date list of available repositories.

More information about open access for other research objects like data and software have been discussed in Open Data and Open Software modules respectively.

Obstacles with being open when reusing closed Research Objects by others

We have detailed the most common challenges you may face when making your results open. Additional challenges may arise if you reused closed Research Objects (ROs) but want to make your results open. Below are examples and possible solutions to overcome these challenges:

Potential obstacle 1

you do not get consent from some of your collaborators for opening some datasets you have used in your analysis.

Recommended solution: you can create and share metadata (description of the content, data format, link to sample data files) instead. However, your results may not be reproducible. Therefore when sharing your workflows and software, it would be useful to provide sample datasets to demonstrate the reusability of your work. We recommend you agree early on (at the planning stage) on what datasets you will be using and whether they are open or will be opened (embargo).

Potential obstacle 2

your research work involves the usage of sensitive datasets that cannot be shared when publishing.

Recommended solution: make sure to detail clearly the protocol used to collect the dataset and the condition of access. Ensure you have metadata to increase the FAIRness of your work. You may also want to provide sample datasets (for example, anonymized) to ease reuse and support the creation of derivative work.

Potential obstacle 3

The software you have used for your analysis is not open. This will of course limit the reproducibility of your results to those who can access the software you used.

Recommended solution: if it is commercial software, you can add metadata information such as the software name, version, and prerequisites (see software module [addlink-software]) to help others to identify and possibly buy the very same software. The more costly the software is, the less likely your research results will be reproduced and reused. When the software is closed and/or available at a cost, it is recommended to add sufficient information on the algorithm used to make it more accessible. If possible, provide containerized executable versions or offer an online service to run the software. Whenever you start a project, you should assess the tools you would need and evaluate the impact of using close or commercial software.

What are the guiding principles to turn a research result into an open result?

Following the FAIR principles (see Ethos module [addlink-ethos]) can help ensure research results are Findable, Accessible, Interoperable and Reusable. This is one of the prerequisites to making research results open and available to everyone. The CARE principles (Collective benefit, Authority to control, Responsibility and Ethics) are also detailed in the Ethos module and complement the FAIR principles: they are people and purpose-oriented and aim at advancing the Indigenous peoples' rights and self-determination.

For Data and/or Software Research Objects, you can read the Data and Software Modules, respectively. Other types of Research Objects such as your planning and research results documents (for example, data management plan, project proposals, blogs, and videos) and publications need to follow the FAIR principles to allow others to understand your work and eventually derive new creative work.

Before learning how to ensure your research results are FAIR, let's clarify the concept of FAIR and highlight the differences between FAIR and Open.

FAIR for Closed Research Objects

Ideally, Research Objects should be FAIR and Open. However, it is not always possible. For instance, whenever there is sensitive data or data that cannot be distributed (for example, it could harm or target specific people or identify the location of endangered species or animals), sufficient metadata can make the RO FAIR while keeping the data itself closed. Therefore, Research Objects can be **FAIR but not open**.

FAIR for Open Research Objects

Openness is a necessary but not sufficient condition for maximum reuse. When the content of Research Objects can be made open without harming anyone and with the consent of all the contributors, ensuring their FAIRness increases reproducibility and reuse.

Transforming an “unFAIR” to “FAIR” result

Here we will explore two scenarios:

1. The Research Object is not yours
2. The Research Object has been created by you and your research team.

Turning someone else Research Object into a FAIR result

- Check the license of the Research Object: if you cannot share the content, you can still create metadata.
- Add metadata (authors information, size of the Research Object, contact information, title, description, and location such as a persistent identifier or Digital Object Identifier) with detailed information about the Research Object itself. Readme tools such as <https://readme.so/> can help guide building informative metadata [addlink-tools]
- Deposit the Research Object (if it can be re-distributed) in a repository where you can add metadata and get a persistent identifier such as Zenodo or a community-specific repository. If the Research Object itself cannot be redistributed, creating a record in a repository such as Zenodo where you can add as much metadata as necessary for others to understand and potentially request the Research Object itself.

Turning your Research Object into a FAIR result

- Check the colours of your figures and tables and change them to make them colourblind-friendly (see for instance <https://www.color-blindness.com/coblis-color-blindness-simulator/>).
- Check that making available your research results will not potentially harm anyone. In doubt, do not open the particular research result.
- Tidy your project structure to use descriptive file names and logical folder structure. See this resource for a good summary of what makes for good file names and folders: <https://datamanagement.hms.harvard.edu/collect/file-naming-conventions>
- Add a README file to your folders to explain what they contain. This tool can help easily build a readme: <https://readme.so/>
- Move data saved in proprietary formats to open standards (for instance, move files saved in DOCX format to RTF or HTML).
- Add code and code documentation, with descriptions of what each function does, and what their inputs and outputs are, and include examples. See open software for documentation standards [addlink-software].
- When publishing data, add an example of how to read and analyse the data.
- Upload reports in open archives.
- Choose Open-Access platforms that give free and online availability of research outputs.
- Publish a blog or a video abstract in simple language for the layperson.
- Link all the contents of your research outputs as an aggregated Research Object (for example, ensure data, code, and metadata can be found in the same archive).
- Add any other relevant metadata (add effective title/names, description and keywords) to each of your research outputs.
- Deposit the aggregated Research Object into a repository that can deliver persistent identifiers such as Digital Object Identifier.

The continuum from closed to open

All research results lie on a scale between closed and open because there are variances in how information is shared and the reasons to share. Your research results can be:

1. **Closed.** It is only available to certain individuals within an organization. It is patented or proprietary.
2. **Mediated.** It is semi-restricted to certain groups or it is open to the public through a licence fee or other pre-requisite. As we have discussed in previous sections, there are legitimate reasons to restrict access to data and when data is mediated possible users must request access. For example, health-related information collected by a hospital or insurance carrier

3. **Embargoed.** The result will be open in the future. For example, some groups might release their data following an appropriate latency period to allow a thorough understanding of the data as well as to allow time for the scientific exploitation of the data by the research team.
4. **Open.** It is accessible in a readable format and licensed as open source.

And there are many setups in between these four categories!

Aggregating your Research Objects

To work Open you may have created different Research Objects such as data, software and workflows. The Data and Software modules explained how to deal with these research results and obtained for instance Digital Object Identifiers for each of them. When publishing, additional material can be added (for example, software, data, workflows) but you usually limit the Research Objects to what is discussed in the paper. Failures, dead-ends and other trials and errors are part of the research process and usually do not have their place in scientific publication. To ease re-use and facilitate the creation of derivative work, you can aggregate all your research objects to create bundles that represent the entire research process and not only the selected positive results.

Assessment: Case study analysis

1. Building on self-assessment #2 in Lesson 1. Which of those elements were guided by FAIR principles?
2. Flag the research objects you think could benefit from FAIR principles.
3. Rank order those objects from “would benefit most from FAIR principles” to least
4. Rank order those objects from “would require most resources” to least
5. Identify a few research objects that strike a balance between high priority and resources required

Applying Open Result Framework to your Research

Introduction

After the previous section, you're probably raring to go to make your research objects as findable, accessible, interoperable and reusable as you can. But how can you go about actually doing so? In this section, we will delve deeper into the practical issues of open results and introduce some specific tools and services that will get you 80% of the way there.

Bear in mind that no tool is optimal in every context. All recommendations made in this lesson are based on what is generally useful but might not be ideal for your particular domain of research, institutional context, culture or legal framework. When in doubt, you can ask your relevant community (for example, the relevant people in your institution, your colleagues and your peers) what tools are available, validated and recommended (See Tools for in-depth discussion on Open Communities [addlink-tools]).

Also of note, these tools are not neutral. All of them are developed and maintained by people in the English-speaking developed world, which charges them with biases and assumptions that might not be relevant to your own situation.

How to apply an open framework across different research objects

An open result is the aggregation of all the research objects introduced in the last lesson (software, data, workflows, reporting, documents). Ideally, to open your research results you would need to open each Research Object that you can legally and ethically open and aggregate them into your final Research result. The approach you need to follow to open an individual Research Object is independent of the type of Research Object (RO) even though the tools may be very different. Below we introduce the main concepts that are necessary to open your Research Objects. Later, we will go through each type of research result (document-RO, data-RO, executable-RO, reporting-RO) and learn the most popular tools you can use.

Unique identifiers

Perhaps the single most important step to make your results open is to assign them a globally unique and persistent identifier. This will give you a single code, URL or number that you can use to uniquely refer to the research object unambiguously. Any derived research object can use this identifier to link to it and create a traceable and rich history of use and development. Crucially, this identifier can be used by others to cite and credit your work.

The identifier must also be persistent. This guarantees that the identifier points to the same research object for a long time. What counts as “persistent” is, of course, a matter of degree since even the most stable identifier probably won’t survive the Sun engulfing the Earth in a few billion years. In this context, “persistent” implies that it is registered in a database managed by an organisation or system that is committed to maintaining it stable and backwards compatible for the foreseeable future.

For example, URLs (for example, a personal website, GitHub repository, or cloud storage) are notoriously not persistent since they can change their contents frequently or become invalid without maintenance. On the other hand, Journal publications have a Digital Object Identifier, whose persistence is guaranteed by the International DOI Foundation.

As well as uniquely identifying each research object, it is important to be able to uniquely identify and cite all the authors and contributors. For this, it is recommended to get the permanent digital ID of each of the authors and contributors. [ORCID](#) (Open Researcher and Contributor ID) is an online service where you can get a permanent digital identifier.

Exercise:

(multiple choice) Select which of the following are globally unique and persistent identifiers:

- Digital Object Identifier 10.1371/journal.pone.0230416
 - The Digital Object Identifier is provided by the International DOI Foundation, which ensures that each ID is unique and ensures that a DOI link always links to the correct object.
- <https://github.com/alan-turing-institute/the-turing-way>
 - This is the URL of a GitHub repository. The contents of the repository can drastically change over time and the owner can delete it completely.
- ISBN-13: 978-0735619678
 - This is an International Standard Book Number, which has to be purchased by publishers by the International ISBN Agency.
- <https://web.archive.org/web/20220121051903/https://www.go-fair.org/>
 - The Internet Archive captures snapshots of websites and their links are really stable. Even if not ideal, it’s a handy tool for creating identifiers of websites easily.

Metadata

The second step to make your research objects open is to produce textual information *about* the research object (metadata) and link to it. This metadata serves both humans and machines. For humans, having metadata is imperative to ease understanding. For example, it can contain variable names contained in a dataset, physical units of a variable of a dataset, the software used to generate and/or read the dataset, the training method of a machine learning model, and the sampling method used for a particular dataset. For machines, metadata is useful for indexing and searching, as well as programmatically interacting with digital research objects. To be “understood” by machines, metadata must follow established conventions and/or standards that are often domain specific. To make your data, software, and workflow interoperable, mapping metadata standards from different disciplines and/or creating cross-disciplinary standards is often necessary but a very complex procedure.

In general, try to think about what information you would need to have in order to know if that research object is relevant to your needs. However, some metadata information that applies to almost any research object is:

- Title: A short but descriptive sentence that introduces the research object.
- Description: A longer text with a more thorough description of the research object. This might include descriptions of the process that created it, important caveats or limitations, and anything that you think would be useful to contextualise it.
- Authors: A list of people responsible for creating the research object and who should be credited if it is used.
- Contributors: A list of people who contributed to populate the content of the Research Object and/or the original authors when you create derivative work from another existing Research Object.
- Date of creation/publication: Try to use an unambiguous date format like the ISO 8601 year-month-day format.
- Version: a number or other sort of ID that helps disambiguate between different versions of the research object, in case it is updated (for instance, if you found an error after publishing it).

As mentioned earlier, many domains have adopted formal metadata standards. To facilitate interoperability between domains the Research Data Alliance (RDA) develops and maintains [the RDA Metadata Standards Catalog](#), a collaborative, open directory of metadata standards applicable to research data.

These guidelines we give for each type of Research Object are not domain specific and should be considered as the minimum required for making your research results open. In any case, metadata should always be open even though you cannot share the associated content (for instance for sensitive datasets and/or closed software).

Exercise:

(multiple choice) Select which pieces of information would be included in the metadata of a dataset of species, sex, body mass, height, flipper length, and bill length measured at three Antarctic Islands

- Date of the data collection.
 - When the data were collected can be important for ecological/longitudinal studies.
- Geographical coordinates of each island.
 - The location of the islands can be used for spatial analysis and also for indexing.
- Average height of all penguins.
 - This can be computed from the data itself.
- Make and model the scale used to collect weight measurements.
 - Instrument details are important to assess the quality of the measurements.
- Filename and extension of the files.
 - Descriptive filenames are very useful for humans to understand the contents of a file and can contain important information, such as dates or locations. The file extension can be used as a good heuristic to know how to read its contents.
- Software name and version.
 - Descriptive information about the software you used for producing and/or analysing data is crucial for reuse. See “Software module” [\[addlink-software\]](#) for more comprehensive information about Software release, documentation, and testing.

Licences/Rules for reuse

Another very important element to include with your research objects is clear rules for reuse (as is and for creating derivative work), which are often and most easily codified by the use of licences.

Without a licence, all rights are with the author of the research result, and that means nobody else can use, copy, distribute, or modify the work without consent. A licence gives this consent. If you do not have a licence for each of the research objects that constitute your research result, it is effectively unusable by the whole research community.

Choosing a licence is not always straightforward, especially since your institution might have legal requirements. If you are using other people’s work, you also need to pay attention to their licences and choose one that is compatible. Different types of licences can be used and the choice also depends on the type of Research Object: licences for software (executable research object) are very different than for documents. We recommend checking the Data module, and

software module to get a better understanding of the licences you can use for each type of Research object. In this lesson, we will recommend the most common approach for each type of RO.

To guide you in your choice, you can use Choose a licenced website: <https://choosealicense.com/>

For instance, if your Research Object is not software, attaching a Creative Commons Attribution 4.0 International gives permission to anyone to share and modify your research object as long as they credit you.

In the context of Research Results, we also recommend being consistent in the usage of the licences for all the different Research Objects you aggregate into your final Research results. For instance, if you choose a permissive licence for your dataset but a closed licence for the software needed to read the data, you significantly reduce the usage of your dataset.

How to share your results, and select tools that support open science?

Here we will go through each stage of the research cycle defined in the categories of Lesson 2 and discuss how you can share each of the components. First, it is important to understand: what is a repository, and why it is important to register research objects in a searchable resource:

Repositories

All the above needs services that can assign unique identifiers and link them to the research objects and associated metadata, including the licences. Repositories are services that cover all those bases.

Zenodo is a very popular repository (Yeston 2021) in which you can register metadata and obtain a Digital Object Identifier, as well as host digital objects such as data, code and publications.

[RoHub](#) is a Research Object registry where you can create Research Objects and aggregate Research results stored/deposited in different repositories.

Registering in a searchable resource.

If you use a service such as Zenodo and/or RoHub, your Research results will be automatically searchable, for instance in [EOSC Explore](#).

Being able to find a research object and understand its contents through its metadata is a great step. But it can be lost if the person who found the data is not able to access it.

For humans, providing detailed information on where to start, and what to look at in the aggregated Research results as well as in each Research Object, is key. Then, as mentioned earlier, pay attention to the font, colours (colourblind friendly palettes) and overall use of simple sentences that can be understood by non-english natives are a few of the recommendations you can follow. When a Research Object has private content (such as sensitive data), it is important to provide as many details as necessary to let other researchers know how to request data (clearance procedures, instructions on how to register and authenticate to servers hosting the data). For machines, standardised APIs (Application Programming Interfaces) are necessary to be able to access the metadata and data programmatically.

As before, while we will point you to solutions that can work a lot of the time, we encourage you to check with your institution, which might already have the infrastructure set up. Also, check which repository is mostly used in your community.

As you know from Lesson 1, the scope and variety of research objects are extremely large, so it's impossible to give guidelines (even brief ones) for all of them. Below we focus on four broad types of research objects.

Documents

Sharing all the documentation related to a project helps other researchers to understand the objectives and can bring further collaborations. Try to make open everything needed for your research project proposal, planning and during execution: proposal, ethics approval, preregistration, project planning, and data management plans.

Recommended tools:

- Upload to Zenodo with CC-BY 4.0 license for archiving and long-term preservation.
- Use Google docs or Overleaf for collaboration.

Data

Sharing data, especially large data, is not a solved problem (see the Open Data section for more in-depth guidance [[addlink-data](#)]). But if your datasets are small enough and don't carry privacy issues, it's relatively straightforward to upload them to a repository. Zenodo ([zenodo.org](#)) allows you to upload datasets of up to 50Gb (larger datasets can be hosted but

you need to ask permission) and it provides you with a unique identifier as well as a whole set of metadata.

Choose a format that is simple to use and read. Make sure to use a data format that can be read with free software and prefer open standards to closed formats (for example, plain CSV files are better than excel). If there is a trade-off between efficient storage and ease of use, prioritise accessibility, since storage is generally cheap. Some research communities have developed or embraced particular formats as their standards so your data will be much more accessible to your intended audience if you adapt to those.

Recommended tools:

- Upload to Zenodo. Check if both the data and metadata can be shared and open whenever you can use the CC-BY-4.0 license. The repository allows for datasets as large as 50Gb. Larger datasets can be hosted if you ask.
- Check if your domain has some standard and a domain-specific repository.

Software

If your analysis is code-centric, one of the best steps you can take to make your code more open is to develop it in a repository with a version control system. This will not only add transparency to the process but make collaboration much easier (after the initial investment in learning the new tool).

GitHub (github.org) is one popular remote repository system for open source projects. You can create repositories for your projects that can even be private and with special permissions for internal collaborators.

A GitHub repository is not an archival service nor does it provide a unique and persistent identifier. To release your code, you need to create a stable snapshot. To do this, you can connect Zenodo to your GitHub account to create DOIs of specific snapshots.

Besides where to host the code, an important aspect is documentation. The single most helpful piece of documentation is to include a README file that explains what the code does, how it can be installed and how it's used. To encourage collaboration from outside sources, you can also include contribution guidelines.

Recommended tools:

- Host your code on GitHub for development and collaboration.
- Connect your GitHub repository to Zenodo and create software releases (snapshots) to get their own DOI for release.

Exercise:

Think of a specific research object from a project you are/were involved in and use <https://readme.so/editor> to create a README template that applies to it.

Reports

As a scientist, you are probably trained to write and publish papers. However, traditional publishing outlets are not open, since they require hefty subscription fees or per-article payments.

Publishing your articles in an Open Access journal might be the easiest option to make documents open, but most Open Access journals charge article processing fees that can be prohibitively high. A free alternative is to upload your manuscript to a preprint server, where you can upload manuscripts before acceptance to a journal.

A very popular and long-running preprint server is ArXiv (<http://arxiv.org/>). ArXiv is mainly used in physics and computer science, so you might want to search for a more specialised one for your community. For biology, there's bioRxiv (biorxiv.org/) and for Earth sciences, there's EarthArXiv (<https://eartharxiv.org/>). Some journals provide one-click pre-print services upon submission.

If you or your team have a website, consider uploading your report there. Although simple, the main disadvantage of this is that an unstructured website doesn't provide unique identifiers and stable links like a preprint server do.

Something important to consider is what are you allowed to do with a manuscript that is published in a journal. Some journals don't allow you to make the final copy-edited version public or even the version with changes based on peer review.

Beyond publications, you probably want to communicate your research work to a larger audience. Writing blogs, developing tutorials and/or making short videos are becoming more and more popular, and an integral part of the research work.

Recommended tools:

- Upload to a preprint server such as ArXiv. Ask around in your community for a more specialised server.
- Upload the report, videos and/or blogs to your personal or institutional website.

Putting everything together

Each individual Research Object is now FAIR or as FAIR as you can, and now it is time to create one aggregated Research Object that constitutes your final research result (final being here used as complete).

The creation of this aggregated Research Object could be as simple as a single text or markdown file with all the links to each individual research result. You can upload that file on your personal or team website.

However, a more structured alternative is to use a registry of Research Objects such as [RoHub](https://reliance.rohub.org/) (<https://reliance.rohub.org/>). There you can add links to all the individual Research Objects that constitute your research result. The type of Research Object depends on the main constituents of your final Research Result. We usually recommend creating an executable Research Object for aggregating all your research results. Each Research Object has a persistent identifier. Once created and ready to be published, you can make snapshots and ultimately archive your Research Object to get a Digital Object Identifier. When you create a Research Object in RoHub, it is harvested in OpenAire and your research result is automatically searchable in [EOSC Explore](#).

Examples

- Executable Research Object “[Cosmos-UK soil moisture \(Jupyter Notebook\) published in the Environmental Data Science book](#)”
- Data-centric Research Object “[Mean ground velocities from ALOS-2 data at Changbais-han volcano \(China/North Korea\) during 2018-2020](#)”
- Bibliography-centric Research Object “[The effects of the Covid-19 pandemic seen through the lens of the Italian university teachers and the comparison with school teachers’ perspective](#)”

Recommended tools:

- Create a Research Object in RoHub (<https://reliance.rohub.org/>)

As open as possible as restricted as necessary

Reproducibility, and therefore FAIR, should be considered as a guiding principle in all stages of your research process. But reproducibility does always mean open. We share the idea that research should be as open as possible and as closed as necessary (Turning FAIR into reality, EC, 2018). Open principles should be applied when you can and never for private, confidential or sensitive results.

This does not contradict all that you have learned so far in this module because FAIR does not require your research objects to be open but it requires open metadata and open standards for interoperability.

Using a checklist to achieve open results

The first step to making your research results open is to register to [ORCID](#) to get a permanent digital ID for yourself. We also strongly encourage you to ask all your collaborators to do the same.

The table below summarises some initial steps that correspond to the [Minimum Viable Solution] (see lesson 2) to make your Research result open. You need to apply these recommendations for each Research Object that is part of your Research results.

MVS	F	A	I	R
Documents	Choose an explicit title, write an abstract and add keywords.	Deposit your document (project proposal, ethics approval, preregistration, project planning document Data management plans, others) in a repository such as Zenodo where a DOI is assigned	Avoid proprietary format and write your document in Plain text (markdown, LaTeX). For collaboration, you can use HackMD, overleaf or Google Docs.	Use an Open Licence such as CC-BY-4
Data	Add explicit information (metadata) along with your data. Use descriptive filenames. Use standards (if they exist) for naming the variables, and standard physical units for variables.	Deposit your data in a repository such as Zenodo where a DOI is assigned Make an example of how to use your data (for instance a Jupyter notebook to read data)	Avoid using data formats that require the usage of closed or commercial software. Use data standards that are long-lasting.	Use an Open licence such as CC-BY-4. See Data Module [addlink-data]

MVS	F	A	I	R
Software	Add information about dependencies, and computational environment necessary for running the software.	Use a code repository such as Github or software that is open source. Write tutorial, README, training material, and contribution guidelines. Write workflows with all the steps of your analysis.	Use Open source programming languages, write portable code and share your workflows.	Use an Open Licence such as an MIT licence. See Software Module [addlink-software]. Make internal/external reviews, and write documentation.
Reports	Choose an explicit title, write an abstract and add keywords.	Write publications, blogs, and press releases, and create accessible graphs (colourblind friendly).	Writing and collaboration: overleaf, google docs, among others. Avoid proprietary formats for storing your report.	Use Open Access.

Assessment: case study analysis

1. From Lesson 3, consider the three highest-priority research objects that could benefit from openness: 1. Identify possible platforms where these research objects could be hosted 2. Identify any modifications to this research object that would enable it to abide by principles of openness

Providing Equitable Opportunities and Credit for Contributors to Results

Introduction

If I have seen further it is by standing on the shoulders of Giants.

Turnbull, H. W. ed., 1959. The Correspondence of Isaac Newton: 1661–1675, Volume 1, London, UK: Published for the Royal Society at the University Press. p. 416

If you are a researcher, regardless of your career stage, chances are you are not working alone. And even if you are working alone on any given project, your work likely builds on the work of others. And just like that, others after you will build on your work, advancing our understanding of the world and beyond.

In the previous lessons of this module, we defined open results and talked about ways you can frame your research so that all your outputs are open. We also spent some time explaining why sharing your results openly avoids “reinventing the wheel” by reusing existing work, saves time and increases efficiency, and facilitates collaboration and onboarding of new members.

In this lesson, we will talk about authorship and contributorship, and dive deeper into why open results matter, specifically talking about how openly communicating your results can open up doors to unforeseen opportunities for collaborations. We will also provide you with guidelines on how to ensure contributions to your current or future work happen equitably, maximizing the chances of fair and successful collaborations. Finally, we will briefly go over how you can contribute to others’ open results in a way that helps your colleagues improve their work and work towards shared research goals.

How do we define contributors to each research object and determine their suitable form of credit?

Defining authors and contributors to your project

Too often conversations about contribution and authorship take place towards the end of a project or right when a scientific publication is drafted. However, as we learned in the previous

lessons, research outputs are generated throughout the lifetime of a research project. To share them as open results in different stages of research, we should build an agreement for how authorship and contributorship in the project will be managed. This requires collaboratively defining what is considered authorship in your project, who among the current contributors is going to get authorship, who will get acknowledged as a contributor, who goes first and last in the list of a scientific publication, and who makes these decisions.

First and foremost, we need to remember that *anyone* who has contributed to the research project must have their contributions recognized. With that shared understanding, in this lesson, we will explore what those recognitions as contributors or authors in your research project might look like.

First, let's define what a contributor is:

A **contributor** of research output is an individual who has contributed to any activity that made it possible for the open result to be published or shared. This includes the person(s) who first conceptualized the idea and designed the work, the project lead, external advisors, general mentors, the students, researchers, research assistants who conducted or helped conduct the experiments, the people who set up the tools essential for conducting the research, data stewards, the software developers, support staff, the project management team, the colleague(s) who provided feedback to the open results, as well as any collaborator. [\[reference\]](#)

Depending on the type of contributions, some of these people should be recognized as authors of the open result, while others are appropriately acknowledged as contributors.

Now, let's define authorship:

According to the definition provided by the International Committee of Medical Journal Editors (ICMJE) which is widely accepted in biomedical disciplines:

An **author** of an open result is a contributor who has given a substantial contribution to the conception or design of the work or the acquisition, analysis, or interpretation of the data for the work. Additionally, an author is a contributor who has contributed to drafting or revising critically the work providing important intellectual content. An author is also someone who has approved the final version of the open result and agrees to be accountable for all aspects of the work and for the integrity of all other co-authors. [\[reference\]](#)

There are several other definitions of authorship which vary across disciplines and describe how they relate to different research outputs. For instance, the [COPE Authorship Discussion Document](#) indicates that minimum requirements for authorship are 1) substantial contribution to carrying out the work and 2) accountability for the work conducted and shared in a publication. Authoring a research manuscript that is published in a peer-reviewed journal, for example, is widely considered one of the most valuable currencies for career advancement, promotion, funding opportunities, and overall chance of being recognized by the research community.

Given the weight traditionally placed on authorship in scientific publication and the fuzziness of the definitions (that often contain relative terms such as “substantial” or “extensive” leaving too much room for interpretation), it is not surprising that determining who amongst the contributors gets to be an author can lead to biased or unfair decisions, disputes between contributors, or at the very least leave someone resentful and feeling unappreciated.

To avoid these challenges, in the next section, we are going to provide some tips on how to determine who amongst the contributors is recognized as an author, as well as how to ensure all contributions to the open result are recognized fairly beyond authoring papers.

How to fairly determine authorship contributions

We established that all contributors to an open result should be acknowledged for their contributions. That said, the research team should first decide who amongst the contributors gets to be recognized as an author and how all the other “non-author” contributors are properly acknowledged in the open result (next section).

The National Institute of Health (NIH) provides a useful schematic representation to help with the first decision (Learn more: Colbert, M. C., Nussenblatt, R. B., & Gottesman, M. M. (2018). Integrity in Research: Principles for the Conduct of Research. Principles and Practice of Clinical Research (Fourth Edition). Academic Press. doi: [10.1016/B978-0-12-849905-4.00003-4](https://doi.org/10.1016/B978-0-12-849905-4.00003-4)):

Figure 1: Example authorship guidelines from NIH: https://oir.nih.gov/system/files/media/file/2021-08/guidelines-authorship_contributions.pdf (Colbert et al, 2018) DOI: [10.1016/B978-0-12-849905-4.00003-4](https://doi.org/10.1016/B978-0-12-849905-4.00003-4)

Even if guidelines like this one can help establish authorship and contributorship (see others cited below), it is rarely an easy “yes-no” (or for the image: “purple and green” decision). The power imbalance between project leads and students, for example, can often mean that members treated unfairly are the contributors with the least power.

Power dynamics amongst the team of contributors need to be acknowledged and discussed openly. Hierarchies and power imbalances can be due to many factors. The most obvious and somewhat accepted in academia are levels of seniority: those with more experience, and those who have been around longer tend to hold the most power. But other, sneakier factors are the legacy of a whole set of systems and structures that have oppressed groups of people for longer than we can remember. Academia and science are no exception. These systems of oppression include sexism, racism, white supremacy, heterosexism, ableism, and many more. It is important to recognize the implications of living in a society where people with less power and privileges continue to get disadvantaged by those systems and to become aware of the resulting biases that may consciously or unconsciously affect our choices.

General Guidelines for Authorship Contributions

Contributions	Authorship? (■ yes; ■ no)	Comments
Design & interpretation of results	original idea, planning & input ■	An idea alone may not warrant authorship, unless highly original & unique
	other intellectual contribution ■	Yes, but assuming active involvement
Supervisory role	supervision of the project ■	Yes, but assuming active involvement
	training, education ■	
	mentoring of 1st author ■	No, unless substantive contribution made to study
Administrative & technical support	resources: \$ ■	Acknowledgements yes, authorship no
	resources: animals, reagents ■	No if already published; yes if novel
	resources: patients ■	Maybe, depending on circumstances
Data acquisition	original experimental work ■	
	technical experimental work ■	No if routine; yes if novel methods added, or specific role, e.g., statistics, imaging etc.
	data analysis (assays) ■	Yes, unless only very basic
	data analysis (statistics) ■	Yes, unless only very basic (t-tests e.g.)
Writing & other	drafting of manuscript ■	Warrants first authorship
	reading/ commenting on manuscript ■	Substantial feedback can be acknowledged
	none ■	Includes honorary authorship for lab chiefs, celebrities etc.

Figure 1: General guide for authorship contributions should be given based on contribution, originality, and active involvement. Authorship and acknowledgement should be decided on case by case basis for all stages of research cycle.

We can even go further and try our best to correct those biases by intentionally deciding to make up for some of the inevitable shortcomings of which we may be unaware or may not have the lived experience.

That is bringing an equity lens to our work.

Equity is another word for fairness or levelling the playing field. **Equity** is an approach that recognizes that the magnitude of systemic barriers posed to a particular person will vary based on their gender identity, race, geographic location, class, age, ability, sexual orientation and other factors. Equity recognizes that different people will need different amounts of resources or support to succeed and overcome these barriers.

If you decide to bring an equity lens to this discussion, consider how you, with your powers and privileges, may be able to help others get a seat at the table. Let's say, for example, that you are a postdoc and the leading author of a research project. A rotating student spends 4 months in the lab helping you set up and perfect the experimental protocol that you will then use to carry out the experiments needed to answer your research question. They may even help you collect some preliminary data, but then they leave and later decide to join another lab. It may be tempting to not include them as authors in the final work and not even acknowledge them as contributors—which would be unethical. However, if you think that they have provided significant help and contributed to the success of your experiment, you should consider giving them authorship, perhaps contacting them to help write the methods section of the manuscript. You would give this student a huge opportunity to be cited and seen as a professional researcher.

Another aspect that can lead to unfair and unethical authorship recognition is the position of the author in the author list. Usually, the first author slot is reserved for the main contributor who has provided the largest contribution to the open result, someone who has been responsible for the work ideation, implementation, and completion carrying it all the way to publication. The last author is generally the group leader or principal investigator who has overseen the project from ideation to completion, providing mentorship and substantial contribution to the open result composition. The authors in the middle tend to be grouped as all the other contributors who have passed the “authorship test”, while disregarding the specific contribution each of them made to the project. Once again it seems that even if a contributor is recognized as an author, every time an open result is set to be published or shared, there is an opportunity for the genuine mistake, misunderstandings, and even plain exploitation and unethical behaviour [Fleming, N., 2021].

One of the best ways to avoid conflicts and unfair authorship assignments is to *be intentional about it and plan ahead* by creating an authorship and contributorship document or guideline for your project!

Below are some tips to guide you in implementing your version of a more ethical and just authorship assignment approach by **establishing authorship and contributor guidelines for your research group** [\[reference\]](#) .

- **Search for existing guidelines** (some are linked in this lesson) and use them as a starting place to create your own set of guidelines. In doing so, seek advice from open science colleagues where you are (starting with librarians).
- **The guidelines should include language to help guide the discussion around explicit recognition of power dynamics.** These are not easy conversations and having language in the guidelines that acknowledge the need for having a conversation around power dynamics helps make it happen as part of the shared norms of the group. This can include prompts to assess the position of the contributor within the team (such as the principal investigator (PI) whose name is attached to the grant funding, the student who just joined the group, and the staff scientist who has worked in the lab for 4 years) and their role and responsibilities in the context of the project implementation (such as the student is the one who wrote the first draft of the research proposal and is going to carry out the experiments, the PI co-wrote and submitted the research proposal and will supervise the whole project, the staff scientist is the one who is going to carry on the statistical analysis and the postdoc is going to help mentor the new student and teach them the technique).
- **Make sure all members of your research group have the opportunity to contribute to the guidelines.** You can draft the initial document, but then ask for constructive and honest input from the other members of your team. If new members join, make sure they are properly onboarded and have a chance to comment on the existing guidelines, especially if they are brought in as contributors to an ongoing project.
- **Re-evaluate and seek explicit agreement over the guidelines at the beginning of every new research project.** If someone does not agree to the guidelines, try to mediate an open conversation about why they don't agree, trying to find a common ground amongst the contributors. If the disagreement remains, you can consider having your team vote and follow what the majority chooses.
- **The guidelines should include instructions for contributors on how to report unethical deviations from the policy** to someone other than the group leader (this could be the Chair of the department, a dedicated office at the research institution, or the funder of the project).
- **At the time of publication of the open result check for any existing policy associated with the platform used for publication (such as provided by a journal).** If the policy does not align with yours, present your reasoning and negotiate with the platform. Also make sure that if the criteria you use to determine authorship deviate, you have a space to clearly state the change in the open result.
- **Make the guidelines publicly available.** If you have a group's website you can post it there, and/or you can decide to create a version of the record and point to a permanent identifier such that the link never breaks by publishing the guidelines on a public repository (such as GitLab/GitHub or [Zenodo](#)).

Speaking of power imbalances, one thing you may be wondering if you are not the group leader is, *how on earth am I going to bring this up to my research group is not at all on board with open science practices or simply has never thought of having explicit authorship and contributorship*

guidelines? Well, there is no one right way to do this, but one suggestion we can give you is to learn about it and then present an outline of the guidelines to your next group meeting. Even if the work should not be on one person, oftentimes the main barrier to having something done is to initiate discussion and create that initial draft to which others can contribute. So, if you are up for it and are committed to implementing open results practices, we recommend that you take that first step and then try to persuade others to join in. In most cases, your colleagues will be grateful and hopefully contribute to composing the guidelines. Check out the lesson on why and incentives for additional resources around the benefits of adopting open science practices (as discussed in the Ethos of Open Science module [addlink-ethos]).

Resources for additional contexts

- The Contributor Roles Taxonomy or CRediT (<https://credit.niso.org/>) is a high-level taxonomy that is increasingly being used to attribute different kinds of contributions made to scientific scholarly output. These include conceptualization, data curation, formal analysis, funding acquisition, investigation, methodology, project administration, resources, software, supervision, validation, visualization, writing of original draft, reviewing, and editing. In practice, the success of this authorship approach relies on all authors openly acknowledging the importance of everyone’s contributions (see an example by [Living with Machine team](#)).
- The Committee on Publication Ethics (COPE) (<https://publicationethics.org/authorship>) offers guidelines to understand ethical authorship.
- The Declaration on Research Assessment (DORA) (<https://sfedora.org/>) is also a good resource to understand what researchers, institutions, funders and publishers can do to improve how researchers and the outputs of scholarly research are evaluated.
- The [Authorship and Contributions on Academic Articles](#) in The Turing Way offers content to learn more about academic authorship practices, misconducts, discipline-specific authorship traditions, large and equitable authorships, as well as “Tips on How to Get Authorship Right.”

In the next section, we provide information on how to create **contributor guidelines** as a way to: a) acknowledge non-author contributors, and b) invite others who are not currently part of the research team to contribute to your project.

How to create contributor guidelines that ensure equity, access, inclusion, diversity

Figure 2: The process of acknowledging contributors in The Turing Way. The Turing Way project illustration by Scriberia. Used under a CC-BY 4.0 licence. DOI: [10.5281/zenodo.3332807](https://doi.org/10.5281/zenodo.3332807).

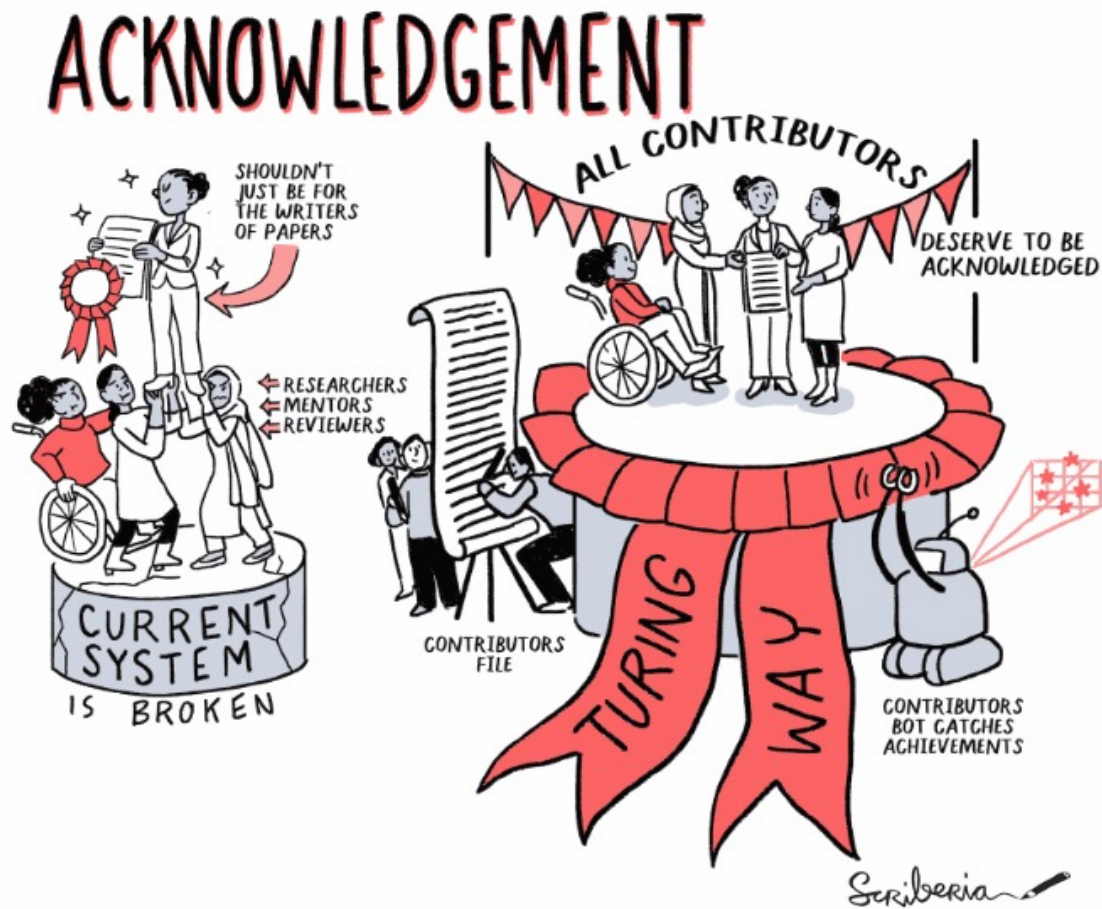


Figure 2: A hand drawn illustration shows that the traditional acknowledgement system is broken then it shows how we try to acknowledge them fairly. We have a contributors bot that catches all the contributors information and stores them in contributors record

Contributor Guidelines

In addition to providing guidelines on how to assign authorship, you also want to have a system in place to fairly recognize the contributions of non-authoring contributors. Additionally, you may want to consider guiding contributions to your results after they are made open. Let's take a look at how you can go about this.

Crediting non-authors project contributors

The CRediT taxonomy captures some of those “non-author contributors” roles such as funding acquisition and project administration, but there is more to open results than scholarly outputs. This may include work associated with maintenance, community management, data stewardship, library and archiving, equity, diversity and inclusion efforts, funding, project management, scientific event organisation, training activities and more. We must ensure that there are processes in place to acknowledge, value and reward this hidden labour [D'Ignazio, C., & Klein, L., 2020].

You can consider adding a section to your authorship guidelines that specifically talks about how you will acknowledge non-authors contributors to different components of your open results.

Inviting others to provide feedback and contribute

Often in academia, external feedback is not sought out until we submit our manuscripts for publication to a journal or a conference in which case 2-3 reviewers, presumably experts in the field, are recruited by an editor to anonymously review your work. The process of peer review is a much-dreaded one as the expectation tends to be the one that feedback will set us back, keeping us from publishing our work, or attending a conference.

But it does not have to be that way. Together, we can build a future in which feedback from outside of those who are directly involved in the project is sought out sooner and is welcomed with a much more positive attitude. Of course, what you don't want is to get showered with non-constructive opinions or personal attacks, which is why it is important that you set the rules for contributions.

Contributor guidelines are documents that are often used in open source projects to guide potential project contributors in providing constructive feedback. They contain information about the project itself, links to various parts of the project, and, most importantly, a detailed description of *how* to report errors (often called “bugs” in code jargon), suggest changes, and even request integration of large parts of code improving or adding features.

In the context of scientific research, the practice of having a document that accompanies results and explicitly states how and what type of feedback should be provided is not common.

However, when you begin to make more and more of your results open at different stages of your research cycle, you may want to invite feedback to aspects of your work that you think need it the most. You also may benefit from guiding the way that feedback is received so that a) other people feel like it is okay to contribute, and b) your corresponding author's email doesn't get filled with non-constructive, not actionable and unclear advice.

Let's say, for example, that you and your team are drafting a research manuscript and are getting pretty close to having all the information in place for it to be shared as a preprint. This does not necessarily mean that it is in a stage that you would consider "final"—if such a stage even exists!—but it contains enough information so that others who are in the same or similar fields of research can understand what the work is about and correctly interpret your results. This may be a good time to solicit feedback from colleagues or the broader community, and maybe even guide it towards aspects of your work that you think would most benefit from review or contribution.

Maybe you want to drive the attention of your contributors to the Methods section where a statistical analysis that is uncommonly used for the kind of experiment you conducted is described. Or maybe you would like some feedback from people with data visualization expertise so that you can best present your data.

Where should you put your contributor guidelines? You may have one place on your website where you can write general contributor guidelines to any open result you and your team share and link to the guidelines in your open result themselves. The contributor guidelines document can be part of the authorship and contributorship document you prepared following the tips in the previous section. You may even want to publish a version of the record that has a permanent identifier so that the links would never "break". Repositories such as GitLab/Github and [Zenodo](#) would allow you to post versions of the original documents that would be linked to one another.

General contributor guidelines that also include the authorship guidelines we talked about in the previous section may look something like the sample template provided in 'Assessment 1' at the end of this lesson

Additional resources for reference

- For additional tips on how to acknowledge contributors, check out [Acknowledging Contributors The Turing Way](#).
- If working with online repositories such as GitHub, an app like '[all-contributors](#)' bot is a great way to automate capturing all kinds of contributions, from fixing bugs to organizing events to improving accessibility in the project.
- More systematic work is being undertaken by [hidden REF](#) who constructed a broad set of categories (<https://hidden-ref.org/categories>) that can be used for celebrating everyone who contributes to the research.

- There are several [research infrastructure roles](#) like community managers, data stewards, product managers, ethicists and science communicators, who are also being recognised as valued members in research projects with an intention to provide leadership paths for technical and subject matter experts, even when their contributions can't be assessed in tangible or traditional outputs [[reference](#)].
- The Declaration on Research Assessment (DORA) (<https://sfdora.org/>) is also a good resource to understand what researchers, institutions, funders and publishers can do to improve the ways in which researchers and the outputs of scholarly research are evaluated.

How to ensure your open results are properly attributed and cited by others

A citation is a reference to a source, which can include any of the research objects described previously in this module, used in your underlying research work. Citation (and other forms of acknowledgement) have been the primary means by which researchers and scholars receive credit for their work. Put differently, citations have largely been the “reputation currency” of science. In this section, we will discuss various ways that you can ensure that your work is citable and that when it is cited, you are properly credited.

Persistent Identifiers (PIDs)

Persistent identifiers or PIDs are an invaluable part of the citation process. They are long-lasting references to digital resources which allow you to reliably find and verify resources leveraging underlying metadata which is associated with the identifiers. More importantly, by using PIDs, you can take advantage of scholarly systems that can help you be more productive and efficient in sharing research outputs while further supporting the acknowledgement of your research. Also, by using PIDs you will be making your research findable, accessible, interoperable, and reusable or FAIR, essentially machine-readable, and responding to International and national efforts to open up science. The best way to understand the value of PIDs though is to look at two examples, ORCID iDs and DOIs (Digital Object Identifiers).

ORCID iDs

ORCID iDs or ORCID iDs are long-lasting unique identifiers for researchers. [ORCID](#) is an acronym for ‘Open Researcher and Contributor ID’. A freely available service for researchers, ORCID is used globally. It is used to authenticate via a wide array of research systems but it is also increasingly recommended and even required by manuscript systems for preprints and journals ([introduction video](#) by ORCID).

ORCID addresses the challenges around disambiguating author names and distinguishing their works. So for instance, authors that share a common name do not have to worry about their research being mixed up and not associated with themselves. They are also useful towards supporting life changes where your name can change but where your work can still be associated via your ORCID. Of course, they also keep your data intact even if you change your legal name at any time.

While you can use ORCID for your profile, it also facilitates interoperability where entering your information once can save time on entering it into other research information systems. ORCID supports an array of research activities you are associated with from your roles and grants to peer reviews you have done and data you have created. By entering your ORCID and/or authenticating with other systems, they can reuse this information so you do not have to enter it again, but you can also have greater control over your record as well.

One way to demonstrate the benefit of interoperability via ORCID is by enabling auto-updating with [Crossref](#) (digital identifier service for primarily preprints and journals) and [DataCite](#) (Digital identifier service for data, software and other research objects). ORCID provides instructions on how to enable auto-updates with these services so you do not have to re-enter information from works and resources you have authored [[reference](#)]. By enabling this feature, you will get notifications when new works are connected with your profile and ready to be made public.

A good practice is to consistently use your ORCID and to provide ORCIDs upfront in research projects, for instance, creating a contributors resource (see [this example by The Turing Way](#)) where this information can be easily accessed when resources are created over the life of a project (versus always having to dig up this information in various places).

Digital Object Identifiers (DOIs)

Digital Object Identifiers or DOIs are unique, persistent identifiers provided for objects. The services mentioned above, Crossref and DataCite, both use DOIs, where Crossref primarily is for works such as journal articles, preprints, and book chapters, to name a few, while DataCite has been mainly used for datasets, software, and presentations. The two overlap in the types of research objects they cover, but DataCite is used more often by repositories to provide DOIs and Crossref is more widely used by publishers. Through these services, publishers and repositories are committed at least to the persistence of the overarching metadata about these objects and that links to them do not break or rot. URLs or links used on the web are known to break, move, or not be available while DOIs are intended to ensure that information about the objects that they are linked to is not lost.

While one of the advantages of using DOIs is maintaining persistent links to research objects that might normally be lost on project websites, another advantage is that you can use the underlying metadata associated with DOIs to help with certain tasks such as citing and reporting your activities. One such service that demonstrates this value is [CrossCite](#). Enter a DOI,

choose a citation style and then format. What normally takes time re-entering information and formatting took very little time. More and more, scholarly services are leveraging DOIs to automate workflows and help researchers reduce the amount of time it takes to do certain tasks like the one described above, but of course, a system like this is dependent on quality metadata and researchers using it.

Another advantage of using DOIs is that you can publish more aspects of your research, beyond just a paper, from data and software to presentations, and if done from the start, can be used to support the transparency and replicability of your research. Not only does this practice help towards recording how your research progressed, it can be used to respond to open science recommendations and requirements, but it is also just good scientific practice.

How can you use DOIs? One way to start and test how you can use DOIs is via a [sandbox](#) service such as the one Zenodo provides. Anything you do in the Zenodo sandbox is not permanent and will be removed periodically. Once you are comfortable, you can move to use the actual public [Zenodo](#) service or some of the other repositories listed at [re3data](#).

All the PIDs

There are a number of persistent identifiers beyond the two examples above. A brief list can be found via [Project Thor](#) but absent are other identifiers such as [RAiDs](#) (Research Activity Identifiers), [RRID](#) - Research Resource Identification, [ROR ids](#) (Research Organization Registry) and [IGSNs](#) (International Generic Sample Number Organization). Determining what identifiers to use can be challenging but thankfully [national PID strategies](#) are highlighting identifiers to focus on, at least to start. ORCIDs for people. RORs for organisations. Crossref/DataCite DOIs for research works/objects. RAiDs for research projects. Grant ID (Crossref) for research funding.

Limitations

Not all the systems are ready for identifiers. For instance, identifiers can be lost in research publications and it is best to provide context/annotate them. In the case of journal articles, you can include a bracketed description such as [Data set] or [Software](#) with the citation and DOI in your references section along with describing the identifiers used in data/software availability or sharing sections. Also, it is important to maintain consistency when using identifiers and citations so that it is clear what other researchers should cite and credit. For example, use the same citation/DOI on GitHub as compared to your website and in your publications.

Self-assessment #1: Develop a contributor guideline for a future project you are considering

Determining authorship and acknowledging project contributors

Many examples and templates for creating contribution guidelines and acknowledging contributors are available online. If you are starting a project from scratch, read this chapter for [_setting up your project_](#). You can directly clone/fork to start your repo with this [template repository](#) by making changes appropriate for your project.

Following the guidelines, the specific contributions of all author and non-author contributors to a scientific work should always be outlined in a project output or open results. Below we provide one way to determine and attribute authorship. If your research team has previously developed contribution guidelines, we recommend reusing them as your team might be already familiar with them.

TEMPLATE FOR ADDING YOUR AUTHORSHIP GUIDELINES

[HEADER]: Authorship guideline for [ADD PROJECT NAME]

Anyone who has contributed to the open results and is not already an author will be acknowledged as a contributor in the acknowledgements section of the open results.

How to contribute to our open results

Our team is committed to openly sharing our results with the research community. We welcome contributions to our work in the form of general feedback or specific suggestions on particular aspects of our work.

Open results for which we are seeking feedback are linked and listed below:

- New research proposal - Link to pre-registered proposal
- Preprint - Link/persistent identifier (PID), like Digital Object Identifier (DOI) - particularly the methods section
- Protocols - Link/PID to a protocols repository such as [protocols.io](#)
- Data - Link/PID
- Source code - Link/PID

We welcome feedback that is positive or negative as long as it is provided constructively. We would love to hear suggestions on how you would address any issue you may identify with the work, with as much clarity and examples as you may be able to provide.

Here are three ways you can provide us feedback:

1. You can send us an email at [provide a team email]. In the body of your email please specify what result you are providing us feedback on, and be as specific as you can in referring to the part that you are talking about so that we are most likely to understand the feedback. We also welcome questions. We will not respond to your email if the content is an attack on the work or our team. Without your explicit consent, we will not share the content of your email associated with your name outside of our team.
2. You can publicly review our preprint by leaving a comment on the preprint server (if this option is available)___
3. You can review our preprint on [add a link from the preprint server like__ [PREreview.org](https://www.premreview.org) or [ArXiv](https://arxiv.org)], where your review can be opened and attributed with an ORCID iD.

We thank you in advance for your time and willingness to help us improve our work.

What to expect after you contribute

If your feedback came via email and is not an attack on our work or team, we will make sure to reply to you in a timely manner—please give our team up to 2 weeks to get back to you.

If you wrote a review on PREreview or on the comment section of the preprint server, we would appreciate a quick email from you to let us know of your contribution. All reasonable feedback and comments will be addressed openly, or, if you prefer, we can reply via email.

If you give us consent either in the email or by posting your review under your real name, we will acknowledge you and your feedback in the next version of the open result.

If your feedback results in a substantial intellectual contribution to the work, we will contact you to discuss opportunities for authorship in the next version of the open result.

Assessment #2: A case study

- Identify the contributors to each research object, at each stage in the research process.
- Were these contributors credited? If so, was their contribution credited fairly with best practices from open science? If not, what could have been done differently?
- Identify contributors that could have added value to different stages of your project and attribute their work fairly

Assessment #3: Give citations for the Research Objects reused in your work

Generate a citation from a DOI associated with a dataset that you preserved in a repository, like, Zenodo, using the crosscite tool. Don't have a dataset DOI, note this for your future research. In the meantime, find a DOI for another one of your research objects, like a paper, and generate a citation using CrossCite.

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Open Science tools

- Definition: What do we mean by “Open Science tools”?
- What’s the difference between ‘open’ and ‘closed’ tools? Why use Open Science tools?
- How do Open Science tools fit into the research lifecycle?
- How do Open Science tools address responsible practices?

Introduction to Open Science tools.

(What are Open Science tools? Why use Open Science tools? How do Open Science tools fit into the research lifecycle?)

This lesson is the first of OpenCore Module 5: Open Science Tools and Resources. This Module provides a collection of tools that are available to increase the visibility and discoverability of your project. It complements the previous OpenCore Modules (Ethos of Open Science, Open Data, Open Software, and Open Results) by enhancing the practical implementation of the Open Science concepts explained previously. While earlier modules focused on the concepts, advantages, and disadvantages of responsible Open Science practices, this module will focus more on the practical applications of responsible Open Science practices. We focus on a few key tools, and highlight how they fit across the research lifecycle.

In this first lesson, you will be introduced to the *What* and the *Why* of Open Science tools. First, we provide a definition of Open Science tools. Second, we discuss the differences between ‘open’ and ‘closed’ tools and highlight the advantages of using open tools. Third, we elaborate on the research lifecycle, and show how Open Science tools fit into a researcher’s project workflow.

What do we mean by “Open Science tools”?

We use the word “tools” to cover any type of resource or instrument that can be used to support your research. In this sense, tools can be a collection of useful resources that you might consult during your research, a software that you could use to create and manage your data, or even a human infrastructure, such as a community network that you could join to get more guidance and support on specific matters.

In this context, Open Science tools are any tools that enable and facilitate openness in research, and support responsible Open Science practices. It is important to note that Open Science tools are very often open source and/or free, but not necessarily.

What's the difference between 'open' tools and 'closed' tools?

Why use Open Science tools?

One can intuitively grasp the difference between open and closed in relation to the “tools”, thinking of openness in terms of exchange with the environment. One should bear in mind that it is not a black and white separation, but rather a spectrum of options.

When speaking of useful resources that you can *re-use* - such as text, visuals, audio, video - it is important to pay attention to the license on the possibilities and conditions for re-use. Lack of indication of a license leads to impossibility to re-use the material. As indicated in Module 1 Ethos of Open Science, Lesson 5, [Creative Commons licenses](#) is one of the most common set of open licenses given to written content of any kind, allowing re-use and requiring attribution, with a spectrum of openness, from least to most open (or CC0, equivalent to public domain).

Software can be proprietary (“closed”) or open source. It is called open source when the original source code is made freely available and may be redistributed and modified. Generally, software has a separate set of licenses designed specifically for code projects that covers both the open distribution of the code itself as well as executable versions of the program which non-programmers can run. More information and details on open software can be found in the Open Software Module.

Human infrastructure refers to a network of relationships between stakeholders interested in the conduct and outcomes of responsible Open Science (more on those stakeholders can be found in Module 1, Lesson 3). Communities – or groups of people who share a geographical location, affiliation, common interest, or practice – play a key role in the human infrastructure aspect of open science. As everything else, communities can vary in their degree of openness. A community can take the form of a mailing list, conference, meet-up or messaging app as a way to stay in touch. In that case, being open would imply that anyone could join the community and be welcomed to speak, decisions would be made transparent, and communications are largely public. On the other hand, a closed community implies that membership is restricted by invitation and/or a fee, resources and communications are not public, and decision processes are not necessarily transparent. More ideas on how to increase participation of stakeholders and how to build and lead inclusive communities can be found in Module 1, Lesson 3 and this module, Lesson 4.

Activity/exercise

Now let's practice by looking at some typical case studies and solutions, reflecting on the benefits and obstacles of open and closed tools.

Case study #1: Closed vs open resources

Case study #2: Closed vs open software

You are a researcher who has been using a proprietary MATLAB platform to analyze data and create models. You are getting a new job, at a different institution. Unfortunately, the new workplace does not have a license for MATLAB, therefore you cannot access your own code and data, stored in the proprietary file formats, and moreover, cannot continue your routine workflow with analysis. What are your options now?

- You can purchase individual license for this proprietary software, or persuade the institute to purchase a group or campus-wide license
- You could consider using open source alternatives for programming and numerical computing, such as GNU Octave, Sage, or even Python programming language and its scientific packages. It would not only save you money now, but provide the continuity of the tool - if you move again, to a different institution.

Case study #3: Closed vs open communities

- **Example:**

Open science tools provide numerous benefits, many of which have been discussed in the previous modules. For example, they can help you collaborate openly and share easily; organize and manage your work; track how your work is treated and shared; and follow leading responsible Open Science practices.

Open Science practices enable easier access to existing tools and resources that promote collaboration between professionals with similar interests and research objects. For example, someone in Asia wanting to study Central African rainforest species could visit an online species database made available by other scientists. Despite their physical distance, many reasons lead to inequality in access to scientific resources, from institutional barriers to paid content.

There are efficient and coordinated ways to share resources in general. One of them is using version control , which is a system to keep track of any changes made to one or more files over time. That also serves as a backup for your work. You might have already done that – for example, if you ever used Google Docs. It stores a version of your work as you type it, and you can invite other users to work collaboratively in the same document, keeping record of all changes made by all users.

One broadly used tool for version control is Git. It enables version control either online or on the user's machine [see <https://git-scm.com/>]. Related services include GitHub, Gitlab, and Bitbucket. Information is stored in online repositories where people can clone, edit, and review each other's content.

Another way to share your work is by using standardized workflows . A standardized workflow is typically a sequence of steps commonly used for a given purpose, such as accessing and manipulating genomic data. A good open science practice, then, is to share those workflows in platforms such as <https://galaxyproject.org/> – which allows any user to replay those steps right there for free, quickly and easily. That and other similar services enable you to show a

step-by-step overview of what other researchers did, build on their work, and share your new ideas.

Including metadata, the data that describes your data, can significantly enhance the findability of your research object. Some examples of metadata are the keywords associated with a publication, the time range and instrument name of a given observational data set, and the ORCID number for a given person. Metadata is a tool that search interfaces use to more quickly find a resource. In fact, Google uses a metadata language called ‘Schema.org’ to build its search algorithm (see <https://schema.org/> for more information).

Many research fields have their own metadata standards (e.g. SPASE for space physics: <https://spase-group.org/data/>), but remember that each website you use has something similar behind that magnifying glass button. Taking the extra time to include some basic descriptors for your research object can make your contribution to your research field much more findable. The same way finding someone else’s work on the Internet might help you, making your own work more discoverable is a great contribution to Open Science!

Next, we’ll highlight how open science tools and resources fit in the research lifecycle.

How do Open Science tools fit into the research lifecycle?

The complex nature of research in the modern scientific community – involving multiple stages, steps, contributors, and stakeholders in the process – benefits from certain frameworks and definitions to structure, organize, and somewhat standardize the research process for the sake of responsible and reproducible practices.

The Open Results module introduced you to the definitions and nine stages of the research lifecycle and workflow. Let’s define these terms again.

- Research framework
- Research workflow
- Research lifecycle

There is quite some theory behind the models for research frameworks, lifecycles, and workflows (REF), including linear, circular, multi-loop, and multi-step flows. For the sake of clarity and pragmatism of mapping the Open Science tools used within the research lifecycle, we will consider a concise 6-stage spiraling model for the research workflow, covering **discovery**, **analysis**, and **writing** as well as **publication**, **outreach**, and **assessment** (see Fig.)

Reference: Bosman, J., & Kramer, B. (2016). Of Shapes and Style: visualising innovations in scholarly communication. figshare. doi: [10.6084/m9.figshare.3468641.v1](https://doi.org/10.6084/m9.figshare.3468641.v1)

Most steps of the research workflow are supported by online applications (Kramer and Bosman, 2016). These digital (Open Science) tools have actually influenced the way in which we perform and share research, opening it up to a global audience.

Open Science tools can be used for:

- **Discovery:** Tools for finding content to use in your research
- **Analysis:** Tools to process your research output, e.g. tools for data analysis and visualization
- **Writing:** Tools to produce content, such as Data Management Plans, presentations, and pre-prints
- **Publications:** Tools to use for sharing and/or archiving research
- **Outreach:** Tools to promote your research

The usage of such tools by researchers across different disciplines has been surveyed and reviewed in several efforts (Kramer and Bosman, 2016, Bezuidenhout and Havemann, 2021). Numerous digital tools have been mapped on the “discovery, analysis and writing, publication, outreach, and assessment” stages of the research lifecycle (see Fig). As we saw in the previous section, all tools have varying degrees of openness. Purposefully choosing tools to use at each stage to increase transparency, findability, and reproducibility, you are able to construct and define your research workflow in alignment with responsible Open Science practices. As was discussed in Module 1, Ethos of Open Science, open should not be a thoughtless default or afterthought, but included into the design and inception of the research project. Your choice of Open Science tools can be individual, but most often it would benefit from group discussions within your research team, institution, and communities of practice.

Note: the concepts of workflow and lifecycle are widely used and applied to parts of the research, e.g. data. Data workflow, data lifecycle are discussed in depth in Lesson X of the Module Open Data .

How do Open Science tools address responsible practices?

The Open Data and Open Results Modules introduced the concept of FAIR principles and discussed how their application according to best practices can increase the visibility and uptake of our research.

Let’s refresh the terms:

- **FAIR Data Principles** - Findable, Accessible, Interoperable, & Reusable. [Wilkinson et al. \(2016\)](#) provided FAIR Guiding Principles for scientific data management and stewardship; [Hong et al. \(2022\)](#) establish FAIR principles for research software.
- **CARE Principles** - Collective Benefit, Authority to Control, Responsibility, & Ethics. [Carroll et al. \(2020\)](#) established the CARE Principles for Indigenous Data Governance, complementing the FAIR data principles.

Best practices to implement these principles include describing data using metadata standards and controlled vocabularies, assigning licenses, and uploading data to repositories that allow for creation of “persistent identifiers”. Examples of useful Open Science tools include:

- Data Management Plan (DMP) tool, which allows you to create and share your data management plans to meet funder requirements and as a best practice for managing your data (link to website, to Lessons)
- Data Repositories, which assign persistent identifiers to your data (example or link)
- Tools for integration research management with DMPtool and repositories (example or link)
- Communities - national and international, discipline-specific, or open science-centered - can be of incredible value in curating resources and building communities of practice for researchers and other stakeholders in adopting FAIR principles. Examples include the FAIR Data Forum <https://fairdataforum.org/> and the Research Data Alliance (RDA) <https://www.rd-alliance.org/>

Working within the ethos of the FAIR and CARE principles can help to ensure that research is accessible, inclusive, ethical, and responsible. More about FAIR principles and practical steps to make your data FAIR can be found here: <https://www.go-fair.org/fair-principles/>

Self-Assessment: Questions for reflection:

1. Assessment of your (open science) tools and resources

Most probably you are already using some tools and resources, even if you are new to open science practices. Here we invite you make a preliminary revision of them:

- Think of all the tools and resources you use in your study/research/work and rely on - resources (content with text/media), software and communities. Think of all stages of your research - discovery, analysis, writing, publication, outreach and assessment.
- Tools have varying degrees of openness, dictated by various factors. Imagine (or draw) the scale from 0 to 10, where 0 stands for completely closed and 10 for completely open.
- For which of the tools (from categories of resources, software and communities) place it on the scale on a number that reflects the degree of openness.
- How many tools do fall towards the lower part of the scale (0 to 4)? Take a moment to reflect if these tools are in line with your actual preference, goals and necessities in the long-term run.
- Perform a quick search using search engine or this open dataset of Open Science tools (<https://kumu.io/access2perspectives/dost#dataset>) for more open alternatives (e.g. free, open source) and jot them down “for your information”.

In the next lessons we will introduce you to various tools, which you may not have heard yet. Stay tuned!

Open Science tools across the research lifecycle

- Open Science tools for protocols
- Open Science tools for data
 - Tools for Data Management Plans
 - Sharing data with your (research) team
 - Data repositories
- Open Science tools for code
 - Collaborative development tools
 - Code repositories
- Open Science tools for results
- Open Science tools for authoring
 - Collaborative writing tools
 - Reference management tools
 - Publishing Open Science and Open Access

Open Science Tools across the Research Lifecycle

In the first lesson, we briefly defined Open Science tools, distinguished open from closed tools, and highlighted the advantages of Open Science tools. We also gave a brief introduction to the Research Lifecycle, and discussed how open tools fit in this workflow. In this second lesson, we'll highlight a few key tools for each aspect of the research lifecycle.

In this module, we'll focus on the following elements of the project workflow rather than distinct research stages, because many tools support more than one stage. We will cover tools specifically for protocols; data; code; results; and authoring. We'll only highlight a few tools; more tools and resources are currently available than we could possibly list (see Figure below).

Ref: <http://46eybw2v1nh52oe80d3bi91u-wpengine.netdna-ssl.com/wp-content/uploads/2021/12/Data-and-AI-Landscape-2021-v3-small.jpg>

Open Science tools for protocols

In the last decades, we have seen an avalanche of development of the tools for management of research projects and laboratories, which address the ever-increasing need for speed, innovation, and transparency. Such tools are developed to support collaboration, ensure data integrity, automate processes, create workflows and increase productivity.

Some research groups have been adapting commonly used project management tools for their own team needs, such as Trello, a cloud-based online tool. Such software facilitates sharing materials within the group and managing projects and tasks, while allowing space for some customization.

Platforms and tools, which are finely tuned to meet researchers' needs (and frustrations), have appeared as well, often founded by scientists - for scientists. To give you a few examples, let's turn to experimental science. A commonly used term and research output is protocol .

Protocol can be defined as "A predefined written procedural method in the design and implementation of experiments. Protocols are written whenever it is desirable to standardize a laboratory method to ensure successful replication of results by others in the same laboratory or by other laboratories." (REF According to the University of Delaware (USA) Research Guide for Biological Sciences)

In a broader sense, protocol also comprises documented computational workflows, operational procedures with step-by-step instructions, or even safety checklists.

Protocols.io (<https://www.protocols.io/>) is an online and secure platform for scientists affiliated with academia, industry and non-profit organizations and agencies. It allows them to create, manage, exchange, improve, and share research methods and protocols across different disciplines. This resource is useful for improving collaboration and recordkeeping, increasing team productivity, and even facilitating teaching, especially in the life sciences. In its free version, protocols.io supports publicly shared protocols, while paid plans enable private sharing, e.g. for industry.

Some of the tools are specifically designed for open science with an open by design idea straight from the beginning, and aim to support the research lifecycle at all stages, and allow for integration with other open science tools.

Most prominent one includes **Open Science Framework (OSF)**, developed by Center for Open Science. OSF is a free and open source project management tool that supports researchers throughout their entire project lifecycle through open, centralized workflows. It captures different aspects and products of the research lifecycle, including developing a research idea, designing a study, storing and analyzing collected data, and writing and publishing reports or papers."

OSF is designed to be a collaborative platform where users can share research objects from several phases of a project. It serves as support for a broad and diverse audience, including researchers that might not have been able to access so many resources due to historic socioeconomic disadvantages. OSF also contains other tools in its own platform:

"While there are many features built into the OSF, the platform also allows third-party add-

(maybe a note on preregistration offered by OSF, which can be powerful)

Open Science tools for data

"Research data means any information, facts or observations that have been collected, recorded or used during the research process for the purpose of substantiating research findings. Research data may exist in digital, analogue or combined forms and such data may be numerical, descriptive or visual, raw or processed, analyzed or unanalyzed, experimental, observational or machine generated. Examples of research data include: documents, spreadsheets, audio and video recordings, transcripts, databases, images, field notebooks, diaries, process journals, artworks, compositions, laboratory notebooks, algorithms, scripts, survey responses and questionnaires." Ref: <https://policy.unimelb.edu.au/MPF1242#section-5>

Data is the one type of research object that is universal. Sharing your datasets publicly allows other researchers (and you!) direct access to the data to allow further study.

Tools for Data Management Plans

Every major research foundation and federal government agency now requires scientists to file a data management plan (DMP) along with their proposed research plan. Data as research in its whole, and as other elements (code, publication) have their own lifecycle and workflow, which needs to be in the plan. DMPs are a critical aspect of Open Science and they help keep other researchers informed and on track throughout the data management lifecycle. DMPs that are successful typically include a clear terminology about FAIR and CARE and how they will and are applied.

The data management lifecycle is typically circular. Research data are valuable and reusable long after the project's financial support ends. Data reuse can extend beyond our own lifetimes. Therefore, when designing a project or supporting an existing corpus of data, we need to remain cognizant of what happens to the data after our own research interaction ends.

There are a few Open Science resources available to get you started and to keep you on track. The *DMPTool* <https://dmptool.org/> in the US helps researchers by using a template which lists each funder's requirements for specific directorate requests for proposals (RFP). The DMPTool

also publishes other open DMP from funded projects which can be used for improving your own DMP. The Research Data Management Organizer (RDMO) enables German institutions as well as researchers to plan and carry out their management of research data. ARGOS is used to plan Research Data Management activities of European and nationally funded projects (e.g. Horizon Europe, CHIST-ERA, the Portuguese Foundation for Science and Technology - FCT). ARGOS produces and publishes FAIR and machine actionable DMPs that contain links to other outputs, e.g. publications-data-software, and minimizes the effort to create DMPs from scratch by introducing automations in the writing process. OpenAIRE provides a guide on how to create DMP.

Sharing data with your (research) team

Data repositories

Originally data repositories appeared in different disciplines of research around the needs of research communities and dataset types, such as *Protein Data Bank* (PDB) <https://www.rcsb.org/> for 3D structures of proteins and nucleic acids, or *Genbank* - NIH genetic sequence database, containing annotated publicly available nucleic acid sequences. Another example is a public repository of microscopy bio-image datasets from published studies, *The Image Data Resource* (IDR) (ref). _The Electron Microscopy Public Image Archive (_EMPIAR) <https://www.ebi.ac.uk/empair/>, is a public resource for raw cryo-EM images. *OpenNeuro* <https://openneuro.org/> is a open platform for validating and sharing brain imaging data. These tools enable easy access, search, and analysis of these annotated datasets.

As noted in Lesson 2, open science tools such as data repositories should ensure the guidelines for FAIR data, mainly attribution of persistent identifies (e.g. DOI), metadata annotation, machine-readability.

Data repositories that include FAIR principles and work across borders and disciplines include *Zenodo* (<https://zenodo.org/>), funded by the European OpenAire project and hosted by CERN. It is probably one of the most known and widely used, as it has an easy interface, support of community curation, and allows depositing diverse types of research outputs - from datasets and reports to publications, software, multimedia content.

The main drawback for this choice is that Zenodo is relatively lacking in documentation and metadata; a dataset stored on this site is not as easily findable or visible to the community compared to storing the data at a domain-specific repository (e.g. EarthData: <https://www.earthdata.nasa.gov/>, BCO-DMO for marine ecosystem research data, or Environmental Data Initiative for environmental or ecological data), or a cross-domain repository (e.g. DataOne: <https://www.dataone.org/>).

Noted exceptions to this rule include communities hosted on Zenodo that curate their materials to enhance findability (e.g. Open Science Community Saudi Arabia (OSCSA): <https://zenodo.org/communities/oscsa/>).

<https://zenodo.org/communities/1231231664/?page=1&size=20>, Turing Way community: <https://zenodo.org/communities/the-turing-way/?page=1&size=20>). More on the role and power of communities will be covered in Lesson X (communities).

Another example of a non-profit data repository is *Dataverse* <https://dataverse.org/>, hosted by Harvard University. The Dataverse Project is an open source online application to share, preserve, cite, explore, and analyze research data, available to researchers of all disciplines worldwide for free.

The Dryad Digital Repository <https://datadryad.org/> is a curated online resource that makes research data discoverable, freely reusable, and citable. Unlike previously mentioned tools, it operates on a membership scheme for organizations such as research institutions and publishers.

Datacite <https://datacite.org/> is another global non-profit organization that provides DOIs for research data and other research outputs, on a membership basis.

Data services and resources for supporting research require robust infrastructure which relies on collaboration. Some examples of initiatives on the infrastructures of data services include The EUDAT Collaborative Data Infrastructure (or EUDAT CDI) <https://www.eudat.eu/>, sustained a network of more than 20 European research organizations,

Private companies as well host and maintain online tools for sharing research data and files. *Figshare* <https://figshare.com/> is one of the examples of a free and open access service, giving a DOI for all types of files and recently developing a restricted publishing model to accommodate intellectual property (IP) rights requirements. It allows sharing the outputs only within a customized Figshare group (could be your research team) or with users in a specific IP range. Additional advances include integration with code repositories, such as GitHub, GitLab, and Bitbucket.

GitHub <https://github.com/>, owned by Microsoft, is often the default data repository for coders. It allows collaborative work, version control, project management, and is widely used by researchers for uploading datasets, files, notes, hosting simple static webpages to showcase their achievements. Github does not give you a DOI, but allows you to state the license for re-use and ways to cite your work.

Much more research data repositories could be found in the publicly open Registry of Research Data Repositories <https://www.re3data.org/>. OpenAire-hosted search engine <https://explore.openaire.eu/search/find/dataproviders> provides a powerful search function of data and repositories, with country, type, thematic and others filters, and enables downloading of the data.

Caution: Amount of data, repositories and different policies can be overwhelming. When in doubt, which repository is for you, make sure you consult librarians, data managers and/or data stewards in your institution, or check within your discipline-specific or other community of practice.

Open Science tools for code

If your project involves coding, such as custom analysis code, you can share it or collaborate using tools such as Jupyter Notebooks. These notebooks can be shared with a variety of permissions on JupyterLab, Google Colab, and similar websites. For a more permanent solution, you can use containerized environments to share the entire analysis environment, which includes the installed software packages, the data used, all custom analysis and plotting routines, and even the publication draft. A few examples of containerized environment services are DeepNote and Binder (DeepNote: <https://deepnote.com/>, Binder: <https://mybinder.org/>).

Collaborative development tools

Code repositories

- Github
- GitLab
- BitBucket
- SourceForge

Open Science tools for results

- Visual tools for graphs, dataviz, sharing

Open Science tools for authoring

Collaborative writing tools

One of the commonly used processes in research is creation and editing of documents, such as meeting notes, conference abstracts, manuscripts, checklists etc.

Collaborative editing process has become really easy with online tools like Google Docs, Bit AI and others, because of their easy interface and version history. However, these tools are proprietary, so not fully open.

Open-source, web-based collaborative tools for editing include tools such as Etherpad <https://etherpad.org/>, HackMD <https://hackmd.io/> and HedgeDoc <https://hedgedoc.org/> (formerly known as CodiMD). These editors use a Markdown language, lightweight markup language, for creating formatted text for the web. It has a simple syntax, and therefore allows more users to be engaged and focus on content, including graphics, tables, lists. Moreover, Markdown is useful when creating documentation in GitHub, as we discussed in the previous sections, commonly used data and code repository and collaboration space.

LaTeX / TeX markup language provides a steeper learning curve, but allows much more nuanced features for scientific and technical documentation, such as formatting of books, articles, mathematical formulas etc. Collaborative online tool utilizing LaTeX is called Overleaf <https://overleaf.com/>, and it is widely used in the research community to share and edit LaTeX files.

Reference management tools

At the *Discovery* and *Publication* stages of the research lifecycle reference management tools are particularly useful to search for publications, collect and organize them, annotate, cite, and share. Such tools should facilitate your research workflow by easy addition/import of references, bibliography construction, adaptation to various citation styles requested by different journals/publishing houses.

EndNote is a citation manager tool owned by Clarivate Analytics. However, it is proprietary software and not free for researchers (closed tool), so it is beyond our interest.

Mendeley <https://www.mendeley.com/> - now owned by publisher Elsevier, is a free software with very similar functionality.

Zotero <https://www.zotero.org/> is an open-source and independent organization-hosted online tool.

Both Zotero and Mendeley tools allow easy addition of the publication from the browser or file upload, offer compatibility with major editing tools (like Microsoft Word, OpenOffice, LaTeX but not fully with Markdown-based online tools). Important feature of reference management tools is groups and collections of articles (libraries), which can be shared and therefore, provide capabilities of social networking and communication among researchers (community of practice).

Publishing Open Science and Open Access

Open Access is a set of principles and practices that make research publications freely available to anyone. Here we will focus on open access implementations both in the peer-reviewed journal publications and preprints uploaded on repositories.

When the data, workflows, or any results of your investigation are ready to be shared as publications, they can be uploaded to certain open websites. Many scientific journals and websites require payment for accessing materials, but a growing number now offer open access publications where the author is charged an additional fee (e.g. AGU publications: <https://www.agu.org/Publish-with-AGU/Publish/Open-Access>).

We discourage publishing in a journal that is not open access because it prevents researchers from marginalized groups from participating in knowledge sharing. In the case of open science

platforms, one can usually share research objects for free (e.g. Zenodo: <https://zenodo.org/> and FigShare: <https://figshare.com/>). Example research objects include executable notebooks, software packages, pre-prints, figures, presentations, and datasets.

Journals usually provide peer review for submitted manuscripts, and after acceptance and publication, there are few options to ensure an open access to the article. It is important to carefully choose the journals with suitable open access publishing models.

Here we list different types of Open Access (OA) publishing models, how to find out which type of Open Access model journals use and where publishing costs are associated.

- **Closed Access/Subscription Journal:** This is a traditional publication, where the reader (or their institution's library) pays a subscription fee for a year's access to the journal contents. The Subscription can be physical and/or digital. Many journals have reduced the print copies; some are digital only and some can be print and digital, both. Subscription can also be pay-per-article instead of complete journal contents subscription.
- **Gold OA:** This form of Open Access requires Article Processing Charge (APC), which may be paid by author(s) or a funding body. The final published version or record is immediately freely available & accessible in the journal by the publisher. The article is freely accessible under a Creative Commons license.
- **Green OA:** There is an embargo period set by the journal's publisher such as 6, 12 or 24 months. The version of the manuscript is freely available in a repository. No charges are paid.
- **Delayed Open Access:** In the subscription journals, the publisher provides free access to online articles at the expiry of a set embargo period.
- **Hybrid:** In the subscription journals, author(s) have an option to make their article Open Access but it has significantly higher open access publication fee in comparison to ***GOLD OA journals***; other articles remain toll access (articles behind paywall).
- **Gratis OA:** Publisher(s) optionally offering articles free to read at no charge to the author. This form of OA may be temporary and may be done for promotional purposes.
- **Libre OA:** Publisher(s) offering articles free to read and permission to re-use, share under Creative Commons licenses.
- **Diamond OA:** The journals/publishers charge no fee/Article Processing Charge (APC) by author(s) to publish. The readers are also free to access and read the articles. Hence, publishers charging no fee are normally funded by external sources like learned societies, funding associations, government grants, academic institutions.

Caution: There are also **predatory journals and publishers**, who advertise open access but are but are not part of responsible open science.

- Open access doesn't guarantee journal quality
- Open access doesn't imply that author(s) can pay to publish without any editorial and/or scientific review.

- Open access does not always require payment from author(s).

Please see COPE discussion document on Predatory Publishing and refer to leading **indexing databases** such as [Clarivate Journal master list](#), [Scopus Journal search](#), [DOAJ](#), [Sherpa Romeo](#).

[Directory of Open Access Books](#) provides access to scholarly peer reviewed open access books.

Many journals with Closed Access/Subscription model provide you permission to publish manuscripts on repositories, even before submitting to the journal. Such manuscripts without peer review are called preprints. Journals usually state the policies on their websites in regards to preprints.

Speaking of open science tools, *Sherpa Romeo* platform <https://v2.sherpa.ac.uk/romeo/> is a valuable online resource that aggregates publisher open access policies from around the world and provides summaries of publisher copyright and open access archiving policies in one place.

ArXiv is one of the oldest preprint repositories (since 1991), used by physicists and mathematicians. Nowadays, there are numerous preprint repositories, each for every discipline and community. Non-exhaustive list include servers of *ChemRxiv* – a preprint repository for papers in chemistry, *BioRxiv* – for preprints of research in biology and life sciences, *MedRxiv* – in health sciences, *PsyArXiv* – in psychology, *SocArXiv* – in social sciences, *engrXiv* – in engineering.

Local open access knowledge and dissemination is maintained and enhanced by communities servers like *AfricArXiv*, a community-led digital archive for African research and - the most recent - *Jxiv*, Japan-specific preprint repository.

Many of country- and discipline-specific smaller “Rxivs” are run by volunteers around the world, but the servers are hosted online by the non-profit Center for Open Science. Substantial costs pose the question of sustainability of maintaining the repository, and some of the repositories like *IndiaRxiv* closed down but were able to relaunch.

Preprints concept and infrastructure allow researchers to disseminate their results months to years ahead of final traditional journal publication. This definitely accelerates progress of science, which is crucial during societal challenges like e.g. COVID-2019 pandemics. However, lack of peer review is reducing the impact of the publication in terms of its rigor and credibility.

Here we will cover some of the key tools that use community/crowd to evaluate and curate the preprints by providing transparent feedback and peer review.

- *F1000Research* <https://f1000research.com/> has been the first open research publishing platform allowing for rapid publication of research articles and other outputs with transparent peer review, without editorial bias.

- *PREReview* <https://prereview.org/> is a platform encouraging early career researchers to provide peer review to preprints, with a mission to increase equity and transparency in scholarly communications.
- *ASAPbio* <https://asapbio.org/> stands for Accelerating Science and Publication in biology. It is a major crowd-sourced peer review by scientists in the life science discipline.
- *The PubPeer* <https://pubpeer.com/> is an online platform for post-publication peer review, “online journal club”, as the founders name themselves.
- *Sciety* <https://sciety.org/> is an online platform for public evaluation of preprints, and allows self-organization of peer review groups.

Case study: *SciPost* <https://scipost.org/> is a scientific publication portal managed by the SciPost Foundation, in the hands of the academic community, by scientists. It is 100% online, offers global, open access and free research publications. As of 2022, it hosts around 10 journals in disciplines of Physics, Chemistry, Astronomy and some others. Submissions can be made directly or via preprint from well established preprint repository arXiv. The peer review is provided by professional scientists (=with PhD and beyond) - anyone could register and serve, the reviews and author responses are published as well. Unlike most publishing houses, it is entirely not-for-profit, not charging any subscription fees to its readers, not charging any publication fees to its authors. The business model is based on the sponsorship from research institutions and foundations, and all agreements and subsidy amounts are openly shared on the website. Does it seem too idealistic?

Question for reflection:

- What are the limiting factors to developing and maintaining Open Science tools?
- What are the advantages and disadvantages for working with Open Science tools?
- What are your next 3 simple steps you could take to increase the openness of the research tools in your practice?
- What is the future of scholarly communications that embraces responsible Open Science practices? Check the Ethos Module, if necessary.
- How does the publication workflow should look to provide the robust, rapid and transparent communication of research results - to the peers, wide scientific community, public, policymakers?

Open Science tools for reproducibility

- What is reproducibility?
- Computational notebooks
 - Jupyter
 - R Markdown
 - Quarto

Open Science tools for reproducibility

This lesson is the third of the OpenCore Open Science Tools and Resources Modules. In this lesson, we take a deep dive into a few available tools for (computational) reproducibility.

What is reproducibility?

Reproducibility - the [National Academies Report 2019](#) defined reproducibility as:

- **Reproducibility** means computational reproducibility—obtaining consistent computational results using the same input data, computational steps, methods, code, and conditions of analysis
- **Replicability** means obtaining consistent results across studies aimed at answering the same scientific question, each of which has obtained its own data.

In practice, reproducibility is taken further by an additional step. The goal of reproducibility is not only reproducing the same result given by using the same steps, such as re-executing a notebook in a containerized environment, but also allowing a given user to copy the environment and build upon the new technology and result by editing the environment to apply to a similar problem (e.g., a shareable, copyable executable paper). This small additional step gives others the ability to directly build upon previous work and get more science out of the same amount of funding.

Check out resources for:

- [Computational notebooks](#)
- [Jupyter Notebooks](#)
- [R Markdown](#)
- [Binder](#)
- [Quarto](#)

Note: As you might have noticed, a lot of Open science tools require intermediate to advanced skills in data and information literacy and coding, especially if handling coding - intensive research projects. One of the best ways to learn these skills is through engaging with the respective communities, which often provide training and mentoring.

Self Assessment Questions: Reproducibility

Scenario 1: You stumble upon a research paper published a few years ago which used LANDSAT data and techniques similar to a project idea you want to apply for another area of interest. When you read the methods section of the paper, you find they published their derived data set in an international data repository (Dryad), but their algorithm code to generate the processed data from LANDSAT Real-Time (raw) data are not provided, only the description of the technique which they used is given in their Methods section and the mathematical equations for calculating their new index are in the Supplementary Materials.

Question S1-1: From the hypothetical Scenario above, when there is access to the raw data, results data, and some written methods are provided, does the research paper meet the definition of being “reproducible”?

Answer S1-1: No, the paper fails to provide a necessary level of detail to allow a different team, with a different experimental setup to obtain the same results exactly. The paper may support some aspects of “Replicability”, but only if someone is able to write their own code using the provided methods. With the same raw data product you could test your code and compare your results data to their results data. This would not be easy and is prohibitive.

Practicing open science in a team

- Team Open Science Practices
 - Build Team and Align Trust, Expectations, and Conduct
 - Work As Collaboratively, Transparently, and Openly as Possible
 - Establish Team Tasks and Responsibilities
 - Review Ethical Concerns
 - Establish Team Communications
- Resources and Team Guidelines Checklist
 - Establish Common Team Resources
 - Use Reminders and Milestones to Manage and Track Data and Digital Objects
 - Improve Practices Through Use and Feedback
- Team Results Preservation Checklist
 - Plan to Preserve and Share for the Long-term
 - Preserve the Research/Project Components as open and FAIR as possible
 - Manage a Project Registry (or Directory) for the Outputs

Practicing Open Science in a team

This lesson is focused on how you can practice Open Science in a team. First, we go through team open science practices, where we empower you to develop and use open science practices for your lab or research team. Second, we provide a resources and team guidelines checklist, where we help you ensure your team is working openly across all members and has access to common resources and guidelines that support collaboration, transparency, and openness. Third, we provide a preservation checklist for the research outputs and results generated by your team, to help you ensure that all outputs (e.g., for a mid-term report or project completion) are fully documented, preserved for the long-term, and made accessible to your team.

Team Open Science Practices

Develop and use Open Science practices for your lab or research team. Use this checklist to improve your team's data and software management practices supporting Open Science. Codify them in your team's Code of Conduct.

Note: The checklist is generalized and will need to be adjusted based on your institution, lab, research team, and/or funder requirements.

Build Team and Align Trust, Expectations, and Conduct

- **Co-build the team composition.** It is not just skill sets and needed disciplinary expertise, but the attributes and qualities of members that can make a team successful, such as the proportion of women, bridge-builders, record-keepers, and leaders.
- **Give the team time to converge and align** on an agreed goal and periodically revisit that as things may change (adaptive management).
- **Ensure team members do not discriminate against others** in the course of their work
- **Ensure team members comply with the team practices and guidelines** for conducting research, managing digital objects (e.g., data, software), authorship and publications, preservation of digital objects, and communication.
- **Ensure team members adhere to the appropriate community, national, and international standards** for reporting the results of their scientific activities including respecting the intellectual property rights of others consistent with the European Code of Conduct for Research Integrity (2011) downloadable from: <http://www.esf.org/coordinating-research/mo-fora/research-integrity.html>.

Work As Collaboratively, Transparently, and Openly as Possible

- **Work collaboratively:** Make an initial priority to establish trust and good communication between the team members and clear roles and responsibilities.
 - Establish a common purpose with the leadership and members of the team.
 - Co-design and co-own the project goals.
 - Establish a realistic understanding of the progress to be made and estimated timelines.
 - Create bridges between members from different disciplines.
- **Work transparently** (as possible): Share status, information, digital objects using the common project resources

- Team meeting notes, progress updates, presentations, recordings, shared folders, data/software.
- **Work openly** (as possible): Provide a way for all team members to participate and be included in the various aspects of the project work
 - Openness builds on transparency by providing the understanding needed to use and contribute to the work of another team member. This is an excellent way to support early career researchers and members from other disciplines in the objectives of the project.
 - Teams that are working openly have access to all the project research products, the training and support to understand and use the research products, and an expectation to contribute based on their roles and project protocols.

Establish Team Tasks and Responsibilities

Team tasks emerge from shared common goals, and the pathway to achieving them. Each project might require different tasks and team members should work together to define their responsibility.

- **Ensure digital output management tasks have responsible team members**
 - Develop the Data and Digital Output Management Plan (e.g, DMP or DDOMP)
 - Communicate tasks and responsibilities
 - Management of data and/or software
 - Quality check of the data and/or software
 - Management of archives and preservation for the project (and long-term preservation)
- **Review tasks and assignments periodically.** Especially when:
 - Improvements need to be made.
 - Team members change
 - To ensure there is a backup person - no single point of failure

Review Ethical Concerns

Consider **what ethical concerns apply** based on the nature of the research and the data. Ensure use of Institutional Review Board (IRB) or your local ethical committee. Areas to consider:

- Survey data, geo-coded data [UK Statistics Authority Ethical Considerations](#)
- Personal identification information US PII, EU GDPR
- Health information US HIPAA, EU GDPR

- Protected Species
- Indigenous data sovereignty [CARE Principles for Indigenous Data Governance](#), [Global Indigenous Data Alliance](#), [OCAP® \(Ownership Control Access and Possession\)](#) English, French.
- General Data Protection Regulation (GDPR)
- Artificial intelligence/machine learning [Assessment List Trustworthy AI](#) from the European AI Alliance

For more information (training):

[Ethics and Data Access \(General Information with BioMedical and Life Sciences Data\)](#) developed by Innovative Medicine Initiative (IMI) in collaboration with small and medium enterprises (SME) and pharmaceutical industry led by academics. Includes a legal and [ethical checklist](#) for researchers.

[Need a resource for geo-coded, protected species, indigenous data, AI/ML]

Establish Team Communications

Establish shared communication practices that facilitate the creation of continuities within a group/team

- Establish a regular set of contact points and times for meetings and discussions. For example, recurring meetings for leadership and work package tasks.
- Use password-protected modes of file sharing and note taking, such as Google Drive.
- If the group is multilingual, conduct meetings using both discussion and text to ease translation efforts.
- Allow sufficient time for continuities to develop. Good team approaches take time to build, and may need refreshment as new members join and others leave.
- Ensure the team has ample time to develop personal relationships, preferably in-person, to establish team cohesion, trust, and long-term collaboration. For example, projects that last more than one year, conduct a yearly in-person workshop. For international teams, these workshops should alternate locations between countries.

Resources and Team Guidelines Checklist

Ensure your team has access to common resources and guidelines that support collaboration, transparency, and openness. [Ensure the team is working openly across all members.]

Establish Common Team Resources

- Before or near the start of the project, make decisions on what resources the team will use to:**
 - Communicate and disseminate information, such as via Slack channel, email.
 - Develop and manage documents during the project. e.g., Google Drive
 - Store datasets during the project, considering size and access/controls, such as via [Open Science Framework](#) (OSF), [GitHub.com](#), institutional repository
 - Preserve datasets, images, and associated digital objects (except for software, workflow and training/workshop materials, such as via FAIR-aligned repository
 - Develop software, scripts, and/or workflows, such as via GitHub: establish a team repository
 - Preserve software, scripts, and/or workflows, such as via Zenodo: establish a community
 - Preserve conference, training or workshop materials, such as via Zenodo: establish a community
- Develop digital object management tracking tools (such as a spreadsheet, or database) for datasets, software, conference presentations, posters, preprints, and publications, such as via Sheets in Google Drive.
- Once determined provide each team member with a “summary list” of the team resources. Ensure each team member has access and provided with any needed overview/training. See [PARSEC example](#), section “PARSEC Team Resources”.

Use Reminders and Milestones to Manage and Track Data and Digital Objects

- Once for each team member: Automatically connect your peer-reviewed papers and registered digital research objects to the digital research ecosystem.
 - Activate the automatic updates of your ORCID profile. Your ORCID ID identifies you uniquely and provides a hub to connect your scholarly work in one place. To complete the actions necessary, [in the ORCID support page](#) find instructions for both **Crossref** (English language scholarly publications) and **DataCite** (primarily datasets and software, as well as other objects). For more information on establishing your ORCID ID and profile, review the Ethos Module.
 - Twice a month for team members: Ensure datasets and software are tracked. This supports efficiency especially when working with many digital objects.
 - Review the datasets and other digital material you are exploring. If you find them to be relevant, track them in the team resource defined above. Include descriptive information.

- Store datasets created by the team in the team resource defined above and tracked along with other datasets/digital objects you are exploring.
- Develop software in the team resource defined above. Ensure good version control.
- Slides/Video - clarify primary/secondary dataset definitions.
- Monthly for team members: Ensure all materials and presentations are preserved
 - Include posters, oral presentations, training, workshops, and any other disseminated materials. Provide information on the event such as the name of the conference and session, dates, website links, funder acknowledgement. Track this in the defined team resource.
 - Every three months for each team member (individual action): Ensure your digital profile reflects your current work.
 - Review your ORCID profile, and any other online profile (e.g., LinkedIn, Scopus, Researcher ID) and ensure that it is current and complete. Link all profiles to your ORCID account.
 - Ensure your CV is current, available digitally, and linked to your ORCID account

Improve Practices Through Use and Feedback

- Establish periodic team meetings to review effectiveness of resources and team guidelines. Review with the team their experiences and challenges using the resources and guidelines. Adjust as necessary working towards better support of Open Science objectives for the team.

Team Results Preservation Checklist

Ensuring all research/project team outputs (for a mid-term report or project completion) are fully documented, preserved for the long-term, and made accessible to the team. “As open as possible, as closed as necessary.”

Plan to Preserve and Share for the Long-term

- **Determine what needs to be preserved. Research project components should include:** project description, README files, datasets, software, physical samples, posters, oral presentations, workshop reports, training materials, and any other digital materials.
- Determine which components should remain open just to the team, and which should be made openly accessible to others.
 - Reference your data management plan for what is required for the project, or the lab.

- Reference your community best practices.
 - Reference country, funder, publisher, and institutional requirements for further consideration.
 - Comply with the licenses (e.g. data created by others).
 - Comply with any data request agreements (e.g., sensitive data).
 - For data created for the project, or derived data products, ensure that the full set of data are preserved. Note that most publishers will only require the data that supports the publication to be available in a trusted repository. The full set of data can be cited with description in the availability statement as to which data were used. This approach allows all the data to be preserved together and improves interoperability and reuse.
- **Determine where to preserve the research/project outputs.** Consult the team’s Resources Summary Checklist that was created in 2B [add link to the bullet point] (see [PARSEC example](#), section “PARSEC Team Resources”). If your team has not yet determined a preservation repository for the project components see “Resources and Lab/Team Guidelines Checklist”. Ensure all the links and persistent identifiers are included in the project registry [bookmark to below].
 - Reminder: ensure the repository selected has the necessary protections (access/controls) for the project components.
 - Ensure the repository selected is community-accepted and trusted. [Link to repository selection document.]

Preserve the Research/Project Components as open and FAIR as possible

You have already learned by FAIR in data module, we are specifically highlighting their relevance in different research contents.

- **Datasets:** Your data may require cleaning, reorganization, or documentation to make it understandable. If there is a version that you routinely use for sharing within your group, this is likely to be the version you will archive. It is important that a data file can be read by a computer program without error, i.e., that it does not require human interpretation or proprietary software. Reference for information.
- **Software, code, scripts, algorithms:** Your software may require documentation and reorganization to make it understandable. Ask a teammate to review it for understandability and future use. It is important that you document any relevant configuration information for using your software.
- **Images and associated digital objects:** Consult the team’s Resources Summary Checklist for the preservation locations. Review repository guidance for depositing these objects to ensure they are well-documented and in the best possible format for preservation**.

- **Conference, training, workshop reports and materials:** Consult the team's Resources Summary Checklist for the preservation locations. Review repository guidance for depositing these objects to ensure they are well-documented and in the best possible format for preservation.

Manage a Project Registry (or Directory) for the Outputs

It is common for different types of outputs to be preserved in different places to optimize discovery and reuse. An up-to-date Project Registry provides a quick overview of all the outputs.

- Create and update a Project Registry in conjunction with preserving outputs as described above in the form of a spreadsheet, or other type of list. This can be one registry for the entire project that is updated, or a new registry for each milestone.
- Include in each registry entry a description of the object, preferred citation, and the persistent identifier (e.g., DOI), and any other useful information supporting the project. For outputs that do not have a persistent identifier, provide a URL and description.
- Preserve the Project Registry as a project component. Many funders require in their yearly reports a list of both peer-reviewed publications and all project outputs. The Project Registry can be provided to the funder during the reporting process, or used as a tracking tool to assist with completing the report.

Open Science communities

- Why engage with Open Science communities?
- What is a Community of Practice (CoP)?
 - Communities list
- How to engage with Open Science communities
 - Pathways for contribution
 - Pathways for collaboration
 - Pathways for engagement
 - * Case Study: FORRT
- How to build and lead a community
 - Guidelines for building communities
 - Mountain of engagement

Open Science communities

Where to find (sustainable) support and help for identifying and using OS tools and resources?

This lesson is the fifth of OpenCore Module 5: Open Science Tools and Resources. It provides a curated list of communities supporting the dissemination of open principles and practices in research and beyond. The lesson complements the previous modules of the OpenCore Course by providing supportive environments fostering the gradual integration of the Open Science concepts explained in them.

The transition to open science requires a profound cultural change in academia and research, and communities are at the heart of a comprehensive change strategy. Hence, it is often extremely helpful to gain the support and help of communities and initiatives that help implement, contextualize, and sustain your open science work. These communities and initiatives can often turn out to be reservoirs of knowledge that could help sustain your open science project in the long run. In this lesson, you will be introduced to a number of communities that you could participate in and engage with to enhance your open science project experience.

Fostering a culture of open scholarship practices through communities can bring unique benefits to learners, practitioners, and trainers. Even if different communities have different missions and scope, all are working towards integrating open scholarship principles into research

and education and positively contributing to the advancement of research transparency, reproducibility, rigor, and ethics.

Why engage with Open Science Communities?

- Communities offer a low-entry point into improved research and pedagogical practices. As pedagogical communities welcome scholars from all levels, including early career researchers, they are an accessible space for all wishing to learn and practice open scholarship. By cutting across career stages, these communities, then, become essential to instilling the new and improved values and norms of open scholarship.
- Communities facilitate the co-creation of open scholarship training materials which are crucial in facilitating the integration of open scholarship into research projects.
- Communities also offer a much-needed environment wherein scholars share individual experiences, identify common hurdles, and iteratively enhance their knowledge and advance addressing the unique challenges ensuing from members' needs.
- Through peer-to-peer exchanges, communities help create a culture of open scholarship, benefiting those within the community, and those that interact with it.

What is a community of practice?

Communities of practice are social learning spaces, where individuals come together to learn a new skill, exchange knowledge and experiences, gain new skills, and then apply what they've learned in the contexts of their day-to-day work from the community.

Well-designed and managed communities of practice can support behavioral changes in individuals by connecting them and providing a safe environment where members can exchange ideas and best practices. They can also empower members with the freedom to set and accomplish goals that they are unable to attain on their own.

How to engage with Open Science communities

There are various ways through which you can start engaging with a community. Usually the websites of most of the communities provide information on where a new member can join a community platform and get involved. If there is a **newsletter** available, you can subscribe to it to know more about the activities taking place within the community. Communities may also have a presence on **platforms like Twitter, Facebook, and LinkedIn** where they might make announcements about their upcoming initiatives. **Community co-working platforms** are excellent places to get to know more and interact with current members. Some of the

communities also provide onboarding calls that provide a chance of joining the community in a more formal way.

For individuals who prefer written interactions and discussions, **GitHub discussions**, **Discourse**, **StackOverFlow**, and **Slack spaces** could be excellent to start with. Such written platforms tend to have lots of past knowledge and interactions available that give newcomers an idea of the discussions that take place within a particular community. While all these are excellent places to start with and ask questions, one should be mindful of the fact that most communities are volunteer-run and located across various time zones, hence sometimes it might take longer than usual to receive a response. Always try to be kind, patient, and appreciative.

Pathways for contribution

It is no surprise that newcomers in a community often go on to become future contributors if they find the right pathway. These pathways are explained using personas in the [Contributor Pathways](#) subchapter of The Turing Way book. This subchapter defines the different phases of community membership, as below:

1. Discovery - How an individual first hears about the project or group or community
2. First Contact - How they first engage with the project or group or community, their initial interaction.
3. Participation - How they first participate or contribute.
4. Sustained Participation - How their contribution or involvement can continue.
5. Networked Participation - How they may network within the community.
6. Leadership - How they may take on some additional responsibility on the project, or begin to lead.

Top Tip: Many communities and open source projects participate in [Google Summer of Code](#) and [Outreachy](#). There are many contributors who had their first contact with open science through [Google Summer of Code](#) and [Outreachy](#), and then developed into core contributors with leadership positions.

Pathways for collaboration

There are different ways to collaborate with a community or open project. Contributions spread over many pathways which include sharing resources, reviewing and updating other contributions, fixing typos, improving documentations, mentoring other contributions, or helping in localising the project and the resources within the project to different languages to support and satisfy the needs of multiple locales. **Many of the communities have a low-entry point and don't require expertise in open science or its digital tools.**

The image below shows some pathways of collaboration in the Turing Way, which is an open-source, community-led guide to reproducible, ethical and inclusive data science. Other communities of practices have similar pathways that allow you to interact with their community members without little know-how in open science.

Pathways for engagement

Communities of Practice are designed to offer plural and creative ways to engage with its members. Perhaps the easiest way a member can interact with the community is to introduce yourself on the community's platform. Another low-stake engagement is to share a relevant resource with the community in appropriate channels. Asking a question, or raising a point of discussion on the community platform, is not only welcomed but potentially instructive and beneficial to other members and the community.

Frequently, communities provide opportunities to give feedback—positive or negative, anonymous or not—which can be very useful to community managers and organizers. Communities of Practice often hold regular meetings—and some also hold seminars featuring pertinent content—and attending these meetings is another form to engage with OSCs.

Some communities offer ways for members to submit resources they know to a database so that others can find it, enriching the community. Reading and learning from a Community of Practice's own resources and approach is certainly one of the best ways to engage with it. Members can also engage with the community of practice by spreading the word or taking part in ambassadorship programs, which aim to (briefly) train members on the main issues a community is trying to tackle or improve.

As open communities tend to produce resources themselves, and most do so in one language (at least at first), translation efforts are fairly commonplace. These are extremely advantageous to those who would otherwise be disenfranchised and help foster an inclusive and accessible community atmosphere. A mutually beneficial pathway is to contribute to a community's existing projects and resources which often require constant review and update of its substantive content.

Folks with technical skills can volunteer their expert skills to maintain and improve the community's internal documentation, resources, *modus operandi*, databases, code of conduct, and website. Some communities offer mentored contributions on a community-supervised project - for example, in the context of STEM & Data Science - while others offer different types of mentorships such as helping with the supervision of Bachelor/Undergraduate or Master/Graduate theses.

Research- and education-oriented communities of practice often tackle projects collaboratively where members can take part in the process of science-making. Members can join these projects and contribute to them, and be acknowledged for their efforts and work. Some communities

extend further on this open-collaboration ethos to allow its members to propose new ideas for research and educational projects.

Case study: FORRT

FORRT stands for the Framework for Open and Reproducible Research Training. It is an interdisciplinary Community of Practice of almost 500 early-career scholars aiming to integrate open scholarship principles into higher education and to advance research transparency, reproducibility, rigor, and ethics through pedagogical reform and meta-scientific research. Anyone interested in engaging with FORRT can visit its website (forrt.org) and find an explanation of the initiative's mission, its projects, its open educational resources, and its publications.

Interested individuals can find ways to get involved in several places, with specific attention to FORRT. Once a member enters the community, they are given access to three kinds of channels: welcome/introductions, where anyone can introduce themselves and be welcomed by our community members; general, where anyone can share resources, links, and projects, ask a question, publicize other relevant communities of practice, start discussions, etc.; and community/events/opportunities, where organizers post about onboarding, projects, people, etc. After joining the Slack, a bot sends a DM to users with onboarding information and the 'Getting Started with FORRT' document, containing important links, a code of conduct, a description of FORRT's collaborative projects (their teams, leads, and Slack channels), how FORRT is structured organizationally, and a description of FORRT's contributorship model and guidelines. Folks can [submit resources](#) to a database of curated open science resources, give [\(anonymous\) feedback](#), subscribe to [mentorship programs](#), and learn how to [contribute to FORRT's research & educational projects](#) (including inclusion, reviewing, and translations efforts— e.g., project [Reversals](#), [Glossary](#) and [Summary](#)). Lastly, members can propose research and educational projects in the [#team-ideas](#) channel.

How to build and lead a community

As individuals, we look for opportunities to apply our knowledge to address problems. The most recent example of this is how the research community has reacted to the pandemic by organizing an unexpectedly large number of hackathons, data modeling initiatives, task forces, and working groups. While joining existing communities can provide rich learning experiences, at times we might realize the need to build a new community. Such communities might come into existence when we discover a lack of a community of our interest close to our geographical region, when we meet like-minded individuals closer to our existing time zones, or when we learn how other communities are developing in their local regions.

A key aspect in building community is to design and build projects that empower others to collaborate within inclusive spaces. Openness shouldn't be a thoughtless default, but something

that is consciously designed into what you and your team are doing, while carefully thinking about the ethics and implications at every step.

Guidelines for building communities

In this section, we have assembled suggestions from [The Turing Way](#), which are derived from the experiences of community and technical specialists to assist researchers in addressing this challenge, particularly when [launching a community or a team-oriented project](#).

- Choose a Communication Platform
- Provide a Project Summary File
- Select a Code of Conduct
- Provide Contribution Guidelines and Interaction Pathways
- Create a Basic Management/Leadership Structure
- Provide Contact Details Wherever Useful
- Identify Failed Approaches, and Stop Them
- Have Documentation and Dissemination Plans for Your Project

You can find more details about these guidelines and related within [The Turing Way Guide for Collaboration](#) and contribute to refining them further.

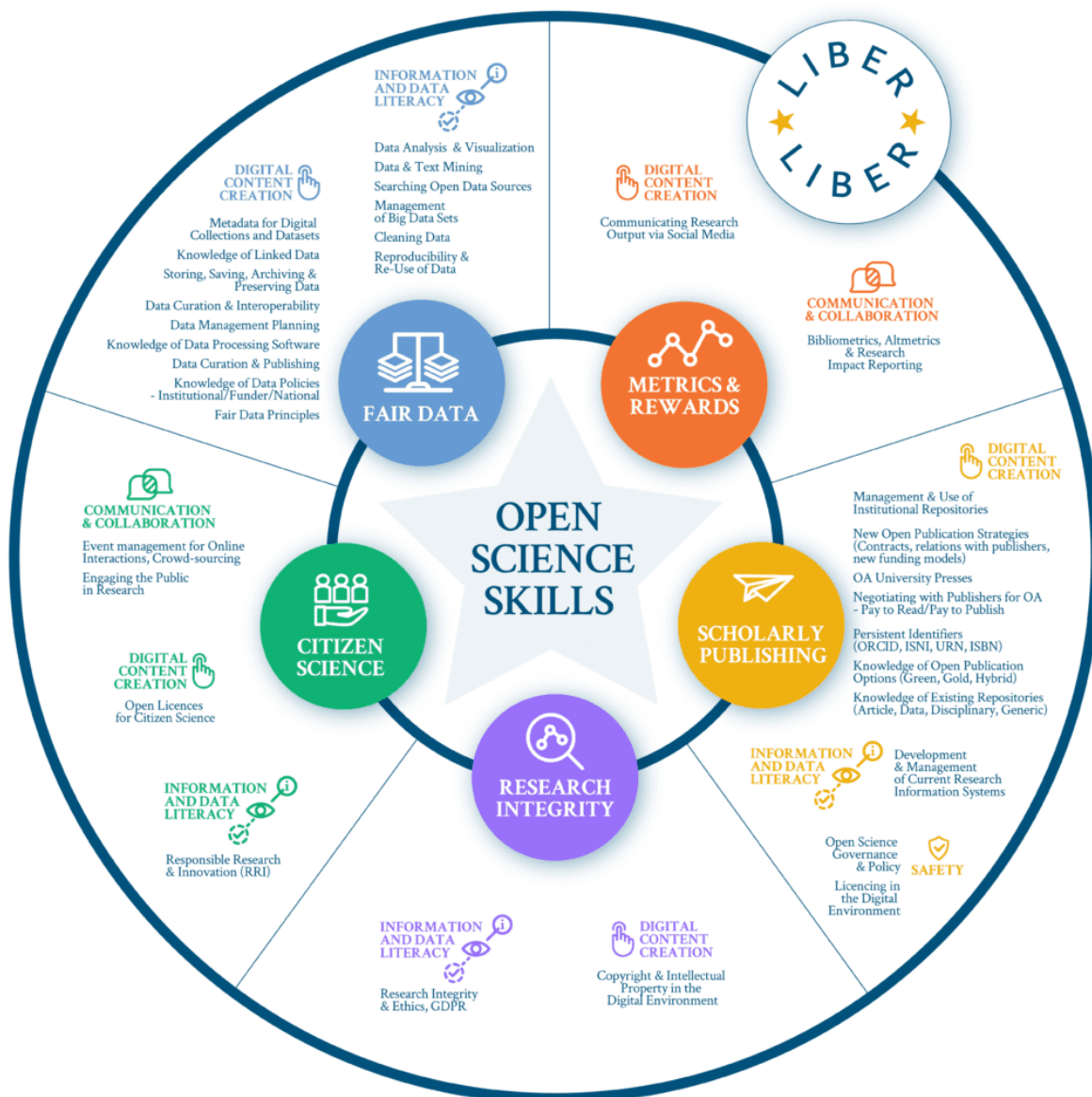
Mountain of engagement

1. **Leading:** A high-touch relationship; we maintain relationships and co-branded events and trainings with alumni and allies to increase the impact, prestige, and reach of both parties' work.
2. **Collaborating:** A high-touch relationship; we offer professional development through our own events in return for co-creation, localization, and spread.
3. **Participating:** A high-touch relationship; we offer community management and professional development through our own trainings and events in return for soliciting ideas & learning through use.
4. **Endorsing:** A low-touch relationship; we share information with people who gain social capital by spreading it and networking with others who share common interests.
5. **Learning:** A low-touch relationship; we gift resources like open curriculum and get back aggregate data (like downloads, registrations, and views) showing people use our resources and pay attention to us.

Model describes four modes of member engagement that can occur within a community – CONVEY/CONSUME, CONTRIBUTE, COLLABORATE, and CO-CREATE. The concept was included in the later cohorts of [Mozilla Open Leadership programme](#), co-designed and curated by Abigail Cabunoc Mayes (Abby) and Chad Sansing. More can be read in the article written by Abby on [Creating Pathways That Invest in New Maintainers](#). Similar ideas have

been previously developed and shared in citizen science, for instance in the post by Muki Hakaye, [How many citizen scientists in the world?](#) in 2018 and paper later published by the author with several researchers: *Albert, A., Islam, S., Haklay, M., & McEachan, R. R. C. (2023). Nothing about us without us: A co-production strategy for communities, researchers and stakeholders to identify ways of improving health and reducing inequalities. Health Expectations, 26(2), 836–846. doi: 10.1111/hex.13709*

Open Science Skills with the Communities; learning and practicing



· Discipline-specific skills needed to practice open science (does not include generic computer skills, wider librarianship skills and personal competencies)
 · Mapped to LIBER OS Roadmap 7 focus areas, Digcomp 2.0 framework and FOSTER learning resources
 · Produced by the LIBER Working Group on Digital Skills for Library Staff & Researchers with input from other LIBER Working Groups, 2020

Open Science Skills Visualisation - Visualisation des compétences en science ouverte. Zenodo.
doi: [10.5281/zenodo.472759](https://doi.org/10.5281/zenodo.472759)

The array of knowledge, skills and competencies needed to practice Open Science

(OS) effectively can be daunting for many librarians and researchers, particularly those who are new to OS concepts and practices. Identifying which skills are needed is the first step for anyone wishing to upskill themselves or others in OS.

- McCaffrey, C., Meyer, T., Riera Quintero, C., Swiatek, C., Marcerou-Ramel, N., Gillén, C., ...Egerton, F. (2020), Zenodo. doi: [10.5281/zenodo.472759](https://doi.org/10.5281/zenodo.472759).

Skills for Open Science are vast, as shown in the image above. Many of are covered in the previous modules introducing a number of digital, data and information skills. Here are a list of resources that can be utilised for learning more about them, while also gaining hands-on deeper experience of integrating open science in your work.

Communities of practice list

Below, we have provided a few recommendations for the communities based on the following resources:

- [NSF COSGN/Network of Networks Proposal - 2020](#)
- [Open-Science-Community-Saudi-Arabia / CoP list](#)
- [Open-Communities-database \(promoting/impacted by open science practices\)](#)

These are selective, and non-exhaustive list of communities:

- **Software Communities:** PyData, SPEC, rOpenSci, pyOpenSci, PyHC, Research Software Engineering, NumFOCUS
- **Data Communities:** OpenAIRE, SPDF, CCMC, RDA
- **Communities with the specific goals to advance gender diversity:** R-Ladies, PyLadies, Julia Gender Inclusive Community, Women of Color Code, Women who code
- **Research domain-based Communities:** UKRN (and other national networks), PSA, SIPS, CREP, OpenMOOC, IGDORÉ, Centre for HelioAnalytics, Masakhane (A grass-roots NLP community for Africa, by Africans), SisonkeBiotik - Lowering barriers in participatory research for machine learning and health across Africa, Bioinformatics Hub of Kenya Initiative
- **Pedagogical & Education Communities:** The Carpentries, FORRT, ReproducibiliTea, ProjectTIER, SIOS, CREP, NowhereLab, RIOT, ReplicationWiki, Open Education Group, Open Education Network, NASA HEAT, ABRIR, Open Hardware Community, Swedish Youth Astronomical Society
- **Community of communities:** CSCCE, Open Life Science, The Turing Way, Reproducibility Networks, Deep Learning Indaba (collective African ML community), Deep Learning IndabaX chapters - different countries in Africa

- *Idea: envision questions on reflection in relation to communities, are you a part of any community? What is the value that you take from it? What do you bring to it? Does the balance seem right? What next 3 simple steps could be done to change it, to improve?*

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