

PTOS-open-data

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Invalid Date

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Welcome

This is a workshop on open data.



PUBLICATIONS AND DATA

CC-BY aukeherrema.nl

If you like the workshop...

0.0.0.1 and want to keep it forever, make it yours

For that...

1. Fork [the github repo](#) this Quarto book is based on
2. Go to settings of your new repo and go to the “pages” section. Then set the “Branch” option to `gh-pages` (leave the dropdown to the right of this at `/root`)
3. Wait a minute to let the website get deployed. You can check on the status in the “Actions” tab of your repo.
4. Back on the main repo site, click on “About” (top right). In the URL of the website, change “j-5chneider” to your username “[your github username].github.io/PTOS-open-data/” (you might need to activate GitHub Pages for that, by [creating a GitHub Pages repo](#))
5. open your new webpage by clicking on that link in the “About” section

0.0.0.2 give it a star in GitHub

So you get noticed if I update something on [the github repo](#).

And I get that sweet sweet dopamine. Hmm dopamine.

Part I

Reasons

1 Your reasons

What are your reasons?

Now that you are here, there seem to be some drivers for you to share data. What are these?

1. Go to [this whiteboard](#)
2. Take one or more of the blue rectangles from the right
3. Write your reason in the rectangle. One reason per shape.

Questions to be answered at the end?

Please [put them here!](#)

2 External Incentives I

2.1 Research funders

DFG

DFG Guidelines on the Handling of Research Data

“Assuming that the publication of research data [...] does not conflict with the rights of third parties (in particular data protection or copyright), research data should be *made available* as soon as possible [...] that allows it to be usefully *reused* by third parties” (DFG, 2015, p. 1)

Guidelines for Safeguarding Good Research Practice. Code of Conduct

“Where possible and reasonable, this includes making the research data [...] available” (DFG, 2019, p. 19)

BMBF

Typical section in calls for proposals

“Zuwendungsempfänger sollen, wann immer möglich, die im Rahmen des Projekts gewonnenen Daten [...] in nachnutzbarer Form einer geeigneten Einrichtung [...] zur Verfügung stellen, um [...] Replikationen und

gegebenenfalls Sekundärauswertungen [...] zu ermöglichen. Repositorien sollten aktuelle Standards für Datenveröffentlichungen (FAIR Data-Prinzipien) erfüllen”

z.B. in

Ausschreibung “Wissenschafts- und Hochschulforschung (WiHo)” 10.11.2023

[BMBF, 2023](#)

Ausschreibung zu “klinischen Studien mit hoher Relevanz für die Patientenversorgung” 09.11.2023 [BMBF, 2023](#)

ERC

Open Research Data and Data Management Plans Information for ERC grantees

“Grantees are required to deposit their research data in a repository and provide open access at least to those data” (ERC, 2022, p. 4)

Questions to be answered at the end?

Please [put them here!](#)

3 External Incentives II

3.1 Policies scientific societies

Examples of societies that have established a policy on open and FAIR data.



(DGfE et al., 2020; DGS, 2019; Gollwitzer et al., 2021)



Look up the scientific society most relevant to you and check if they have a **policy/recommendation/guideline** on open science or sharing data.

3.2 Journal policies

Some journals encourage, some journals mandate a **data availability statement** in the manuscript.

	Not Implemented	Level I	Level II	Level III
Data Transparency	Journal encourages data sharing, or says nothing.	Article states whether data are available, and, if so, where to access them.	Data must be posted to a trusted repository. Exceptions must be identified at article submission.	Data must be posted to a trusted repository, and reported analyses will be reproduced independently prior to publication.

E.g.,

- [Psychological Bulletin](#): Data transparency Level 2
- [Meta Psychology](#): Data transparency Level 3
- [Psychological Science](#): Data transparency Level 1

See the [TOP Factor website](#) to search for the data transparency rating of your favorite journal.

Keep in mind: “Data available upon reasonable request” -> only 6.8% of data sets were actually provided (Gabelica et al., 2022)

Questions to be answered at the end?

Please [put them here!](#)

4 Research I

4.1 Reproducibility & Replication

1. No data = not reproducible

Computational reproducibility :=

“a second investigator (including the original researcher in the future) can recreate the final reported results of the project, including key quantitative findings, tables, and figures, given only a set of files and written instructions” (Kitzes et al., 2018, p. xxii)

Same data + same analysis = same results

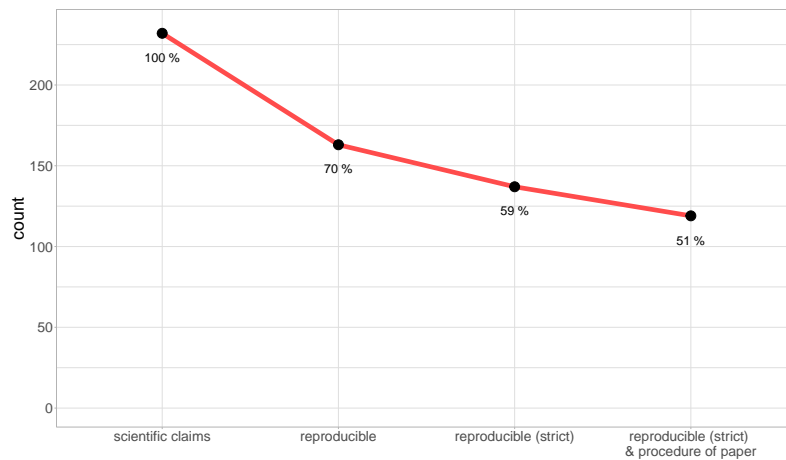
Allows independent researchers to assess the analytic choices, assumptions, and implementations that led to a set of scientific claims -> check for validity and generalizability
(Clyburne-Sherin et al., 2019; Obels et al., 2020)

2. No FAIR data = reproducibility tedious

But this isn't as easy as it sounds (Artner et al., 2021)

4.1.1 Design

- checked 232 primary statistical claims
- from 3 journals
- after data was provided and accessible (33%, 25%, 26%)



4.1.2 Results

4.1.3 Conclusions

Vagueness Makes Assessing Reproducibility a Nightmare

most successful reproductions are predominantly the result of tedious and time-consuming work information about the provided raw data was often difficult to understand, and information about the relevant variables, data manipulations, and the used statistical model was often vague or inaccurate (Artner et al., 2021, p. 12)

3. No data = barrier to replication

Missing data are a major barrier to replication: data were open for 4 of 193 experiments (Errington et al., 2021)

Questions to be answered at the end?

Please [put them here!](#)

5 Research II

5.1 Reuse

The reuse of research data can take many forms

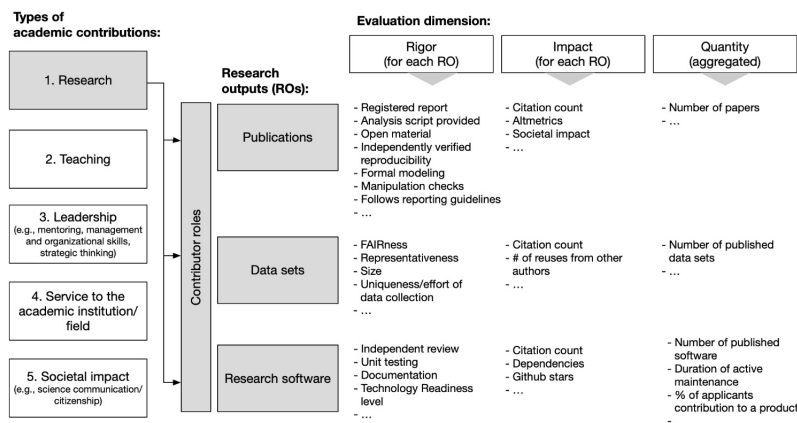
Purpose	Advantage	Needs
Answer new research questions	Saves resources	analysis potential of data, good documentation (Logan et al., 2021; Steinhardt et al., 2021)
Teaching / student theses	Real-life-oriented education	good documentation
Meta-analyses	Easier estimation of parameters	Strictly reproducible code (e.g., Burgard et al., 2022)
Historical perspective	Data as historical artifacts	Potential of data varies

Questions to be answered at the end?

Please [put them here!](#)

6 Researcher

- Open Data Survey:
 - 75% of researchers say there is too little credit for sharing data (Goodey et al., 2022)
 - main drivers: perceived higher citation (67%), increased perceived impact and visibility (61%)
- new metrics for evaluation evolving
 - **CoARA**: “Value outputs associated with openness (FAIR data sets, [...])” (CoARA, 2022, p. 21)
 - signatories: DGPs, ERC, European Commission, DFG, Leibniz Association, ...
 - Example: DGPs recommendations on hiring and promotion (Gärtner et al., 2022; Schönbrodt et al., 2022)



- Data for reuse: Additional publication (e.g., data note in F1000 Research)

- With higher transparency, researchers have higher trust in authors (Schneider et al., 2022)

Questions to be answered at the end?

Please [put them here!](#)

Part II

Open and FAIR Data

7 Openness

- **anyone**
- can **readily access** the data
- at no more than a **reasonable reproduction cost** (i.e., internet connection)

(Open Knowledge Foundation, 2023)

💡 Openness is not a dichotomy

“As open as possible as closed as necessary”
(European Commission, 2023, p. 36)

Questions to be answered at the end?
Please [put them here!](#)

8 FAIRness

<https://www.go-fair.org/fair-principles/>

(Wilkinson et al., 2016)

i FAIRness vs. openness

“does not necessarily mean that data has to be “open” [...] even highly protected data can be FAIR data”
(Kraft, 2023)

(Kraft, 2023) (FAIR principles and the role of scientists)

8.0.1 Findable

The problem:

Just because we provide data online, doesn't mean that others will find it.

We could have the greatest data set to answer further research questions - if our colleagues don't know it exists or can't locate the data, openness will be of little value.

The solutions:

- Get a persistent identifier (e.g., DOI), where you provided your data
 - search for a research data center that fits your needs: re3data.org
 - recommended research data centers: [Verbund FDB](#) (education, Germany), [RDC at ZPID](#) (psychology, Germany), ...
 - recommended repositories: [Zenodo](#), psycharchives.org, osf.io, ...

- Mention DOI in publication that builds on this data (e.g., in the “data accessibility statement”)
- Describe your data as richly as possible (metadata). *Research data centers* offer form fields tailored to the discipline or data type. With *repositories* use alternative possibilities, such as keyword fields.
 - e.g., which variables does the quantitative data set contain?
 - e.g., which topics does your data cover?
 - e.g., which population did you draw your sample from?

8.0.2 Accessible

The problem:

Just because others find our data doesn’t mean the *access barriers* are as low as possible and doesn’t mean they know *in which way* they are allowed to access it. Examples:

- Providing a link to the data in the text of a paywalled journal article
- Unclear licensing / use conditions when providing data (e.g., are non-researchers allowed to access the data or is it only open for qualified researchers?)

The solutions:

- Make sure access is free of charge (or as cheap as possible)
 - e.g., by providing link to data in publicly accessible sections of journal articles that are not open access
 - e.g., by using repositories or research data centers that allow access free of charge
- Make sure users know if they can access and under which conditions
 - e.g., *research data centers* ensure that terms of use are clear (who may access under what conditions) and offer different levels of access restriction

- e.g., on *repositories* provide a readme-file and an open license (e.g., [CC0](#), [CC-BY](#), [CC-BY-SA](#)) with data sets for access cases

8.0.3 Interoperable

The problem:

Just because others downloaded our data doesn't mean they can open and manipulate it.

The solutions:

- Use file formats with open licenses
 - e.g., tabular data: CSV (with additional labelling script), RData
 - e.g., text data: PDF, HTML, ODT, RTF
- Make sure users know how different files are related to one another
 - e.g., define which file contains student data and which teacher data
 - e.g., define which file contains data from cohort 1 and which cohort 2, ...

8.0.4 Reusable

The problem:

Just because others opened our data doesn't mean they understand the data and its use-conditions. Examples:

- Others can't understand what the column names of the tabular data set mean: Which columns in the data set relate to which variables in the journal article?
- Can someone from sociology use the data set from psychology they found on [osf.io](#)?
- Does someone reusing a data set have to cite the authors?

The solutions:

- Adhere to standards in folder organization

- e.g., [PSYCH-DS](#) (see technical specification draft)
- Rich description/explanation of what user will find *in* the data set (meta descriptions about the data set *as a whole*, as for accessibility)
 - e.g., provide a codebook. How to semi-automatically create a codebook, see the R package [codebook](#)
- Provide a license for the use-cases
 - again, *research data centers* ensure that terms of use are clear (who may use under what conditions)
 - again, on *repositories* provide a readme-file and an open license (e.g., [CC0](#), [CC-BY](#), [CC-BY-SA](#)) with data sets for the use-cases

Questions to be answered at the end?
Please [put them here!](#)

9 Exercise

1. Go to [this repository](#)
2. Discuss for **which purposes** you consider this type of sharing to be **suitable** / **less suitable**
3. Discuss what you think makes this type of data sharing **FAIR** and what could be **improved**

Questions to be answered at the end?

Please [put them here!](#)

Part III

Limits

10 Limits

10.1 Not being open may be important

When does “as closed as necessary” apply?

The protection of individuals comes first and is more important than the potential reuse of data

10.1.1 Marginalized/vulnerable groups

- e.g., individuals traumatized by war or who experienced sexual abuse
- At the same time: Can sharing data help to protect these groups from being over-researched (possibly re-traumatization)?

10.1.2 “Closed doors”

- cases in which field access is obstructed or denied by the data provision (Prosser et al., 2022)
- cases in which sharing reduces the willingness to participate

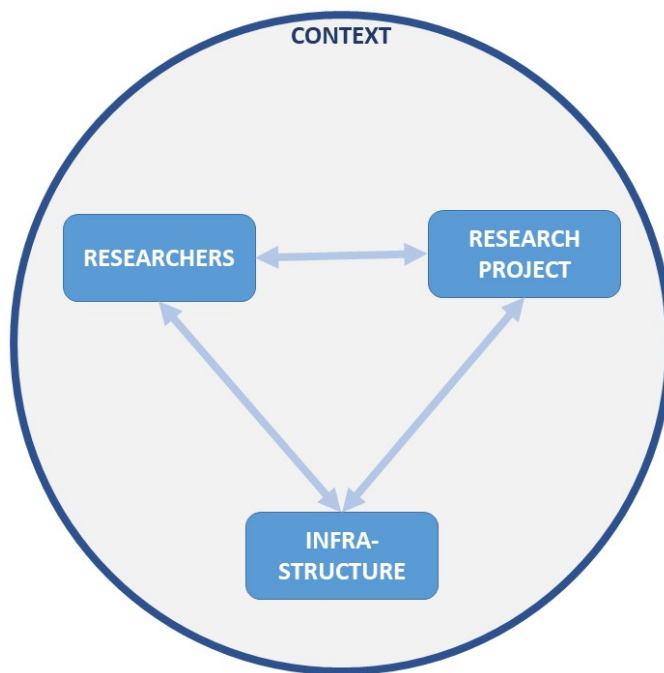
10.1.3 Costs > benefit

- e.g., low reuse potential -> publish for reproducibility
- e.g., when obtaining permission from the school authorities is extremely complicated
- e.g., epistemic problem: context of data collection is highly relevant and other researchers “haven’t been there” (Mauthner et al., 1998) -> publish for intersubjective comprehensibility

- e.g., can't anonymize data -> synthpop, create input-output-documents via RMarkdown/Quarto

10.2 It's not all your responsibility

- Responsibility of opening research is a **collective responsibility** in the “**research ecosystem**” (European Commission, 2018; RfII, 2019)
- Researchers are just one part of this



Infrastructure

- Does suitable infrastructure exist?
- Is it “easy to use” and cheap?
- Is it tailored to my needs and type of data?
- Does it allow the implementation of FAIR data?
- Are there resources to support data sharing?

Context

- Do scientific societies, journals, or research funders encourage sharing?
- Is it common practice (“culture”) in my field of research to share data? (Bishop, 2006)
- Are there standards established for data sharing?
- Do ethics committees request detailed reasoning for the intent to collect own data as opposed to re-using?

Research project

See reasons above on

- Marginalized/vulnerable groups
- “Closed doors”
- Costs > benefit

Researchers

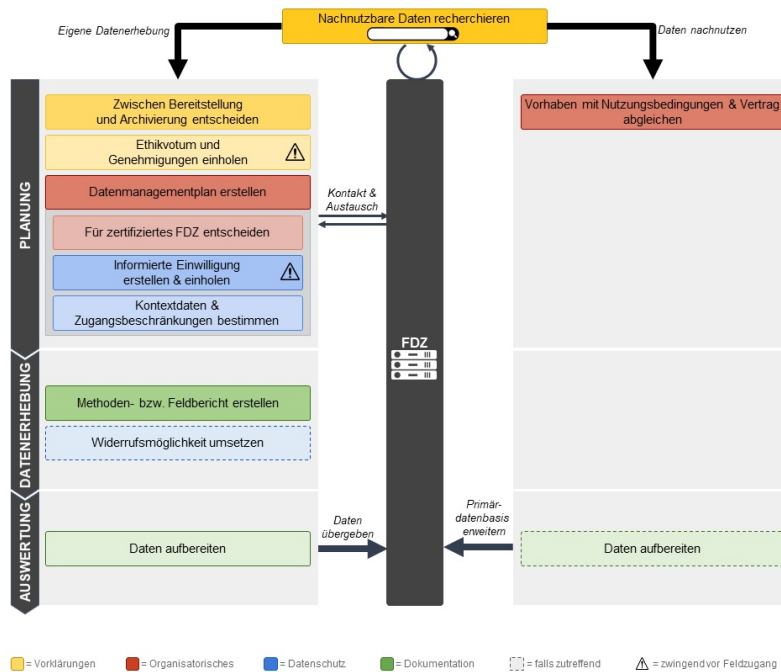
- Willingness and concerns toward sharing data (Mozersky et al., 2021)
- Knowledge, experiences and skills with relevant processes

Questions to be answered at the end?
Please [put them here!](#)

Part IV

Workflow

11 Overview



💡 Table: Steps & resources

What	Resources
Search for reusable data	Research data centers have searchable databases re3data.org , Verbund FDB, RDC at ZPID

Sharing for reuse or reproducibility	Costs: How much effort is required? Consent for reuse available? Benefits: Analysis potential, Quality of data
Data management plan	Templates/ToolsStamp (VerbundFDB)DMP template (ERC)Online tool (OpenAire)Online tool “DataWiz” (ZPID)StandardD-Psy-FAIR (ZPID)
Decide for a repository or RDC	Research data center:Search: re3data.orgVerbundFDB (education, Germany)RDC at ZPID (psychology, Germany)RepositoriesZenodopsycharchives.orgosf.io

Informed consent	Checklist , GermanTemplate , German standard languageTemplate , German plain languageTemplate (qualitative data), GermanExplanations + template (DGPs), GermanOverview + links (ZPID), EnglishExplanations + definitions (Michigan Tech), US Examples (VerbundFDB) Examples (DGPs) There are alternatives to restricting access!
Access restrictions	
Create codebook	R Package codebook codebook from DataWiz

Questions to be answered at the end?
Please [put them here!](#)

12 Workflow 1

12.1 Search for reusable data

Why?

For resource intensive data collections this could save you a lot of time and money

Resources

Research data centers have searchable databases

- re3data.org (Database to search for databases)
- [Verbund FDB](#) (Education)
- [RDC at ZPID](#) (Psychology)

12.2 Decide between sharing for reuse or reproducibility

Why?

- Providing data at a research data center costs time and money for you and the data center
- Typically,
 - sharing for reuse purposes is suited for research data centers
 - sharing for reproducibility purposes is suited for repositories

Resources

I am not aware of any standards to make this decision. Here are a couple of guidelines to decide, if your data is fit for reuse

- Costs:
 - How much effort is required for well-documented data sharing (e.g., Does a codebook exist? What steps are necessary for data cleaning/editing?)
 - Is there consent for reuse available or would it have to be obtained retrospectively?
- Benefits:
 - Analysis potential (e.g., not fully analyzed, type of data, connected with other data sources)
 - Quality of data (e.g., representative, size, special features of sample)

Questions to be answered at the end?

Please [put them here!](#)

13 Workflow 2

13.1 Write a data management plan

Why?

* supports researchers in the process of generating FAIR research data
* ensures good scientific practice

Resources

Templates and online tools for specific applications

- Template [Standardized data management plan for educational research “Stamp”](#) (FDZ Bildung): Detailed form with specific instructions and assistance
- For rapid documentation: [DMP template](#) (European Research Council) with four open questions
- [Online tool for machine-readable DMPs](#) (Open Aire)
- [Online tool supporting the creation of a DMP](#) (ZPID)

Standards

- D-Psy-FAIR (Blask et al., 2022)
 - [Manual](#)
 - [Online Tutorial](#)
 - [Slides](#)

13.2 Decide for a repository (or research data center)

Why?

Deciding on a specific repository or research data center early on helps to

- meet its requirements
- identify whether there are templates available
- identify whether the employees can support the sharing process

Resources

* search for a research data center that fits your needs: re3data.org * recommended research data centers: [Verbund FDB](#) (education, Germany), [RDC at ZPID](#) (psychology, Germany), ... * recommended repositories: [Zenodo](#), [psy-charchives.org](#), [osf.io](#), ...

Questions to be answered at the end?

Please [put them here!](#)

14 Workflow 3

14.1 Informed consent

Why?

- Personal data is subject to General Data Protection Regulation (GDPR)
- Informed consent must therefore fulfill a number of requirements
 - purpose of data collection (includes *sharing the data* and *future use*)-> therefore often “broad consent”
 - participation is voluntary and without disadvantages
 - revocation is possible at any time (until anonymized)

Resources

- [Checklist of legally compliant consent forms, German](#) (VerbundFDB, 2019)
- [Template for informed consent, German standard language](#) (VerbundFDB, 2018)
- [Template for informed consent, German plain language](#) (VerbundFDB, 2018)
- [Template for informed consent, German](#) (Qualiservice)
- [Explanations including template \(DGPs\), German](#) (not specific for reuse)
- [Overview and links concerning informed consent \(ZPID\), English](#)
- [Explanations and definitions around informed consent \(Michigan Tech\), US](#) (not specific for reuse)

14.2 Decide for access restrictions

Why?

- Some data cannot or should not be anonymized (e.g., losing their reuse potential)
- Therefore access needs to be restricted to certain groups (as defined in consent form)

With repositories...

restriction levels are usually limited to

- public (everybody sees everything)
- private (only you and your collaborators see everything)

With research data centers...

there are different restriction levels possible for different files (*in the same project*). Restriction levels depend on what the research data center offers.

Level	Prerequisite	For what
Public Use-file		anonymized data, codebooks, transcription rules
Student Use-file	Short application states use purpose	non-anonymized data with right to use for teaching
Scientific Use-file	Longer application states use purpose, handling of data, and data analyses; identification via PostIdent	non-anonymized data with right to use for research

Remote...	+ access only via virtual machine	non-anonymized sensible data with right to use for research
Access		
Safe room	... + access only in person at research institute	non-anonymized very sensible data with right to use for research

An example: [Project DESI](#), where

- codebooks are publicly accessible (files on the right side)
- video data are restricted for scientific use (files on the bottom of page)

Resources

- [Examples of restriction levels \(VerbundFDB\)](#) (Meyermann & Porzelt, 2019, p. 30f)
- [Examples of restriction levels \(DGPs\)](#) (DGPs, 2021, p. 141ff)

Alternatives

- *Embargo period*
 - Specify a time period, before data go public
 - Possible with research data centers and some repositories
- *Exclude certain research questions* from reuse
 - Specify these research questions in the terms of use
 - Usually only possible with research data centers, except you are writing a very good license yourself
- Create *synthetic data* (e.g., with R package [synthpop](#))
 - Mimics the properties of your data

- Then possible to share this synthetic data set

Questions to be answered at the
end?

Please [put them here!](#)

15 Workflow 4

15.1 Create codebook

Why?

Remember?

Vagueness Makes Assessing Reproducibility a Nightmare

most successful reproductions are predominantly the result of tedious and time-consuming work
information about the provided raw data was often difficult to understand, and information about the relevant variables, data manipulations, and the used statistical model was often vague or inaccurate (Artner et al., 2021, p. 12)

Resources

- R Package [codebook](#)
 - semi-automated creation of a codebook (depending on how well prepared/labelled your data set is)
 - in combination with the [formr](#) survey framework, this package saves you a ton of time
 - still has some minor bugs, be prepared to mingle with it
- Codebook as a result of using [DataWiz](#)

Questions to be answered at the end?

Please [put them here!](#)

Part V

Reflection

16 Barriers

Let's assume: Sharing data is possible. In one way or the other.

Why isn't everybody sharing all data "as open as possible as closed as necessary"?

1. Check out the [flow chart](#) again
2. Individually [3min]: Reflect on
 - What is a barrier/challenge to you?
 - What might be a powerful barrier/challenge for others?
3. In the breakout rooms [12min]: Discuss
 - What are the biggest barriers for you/others?
 - What would be different in an "ideal world" that would lead to you/others overcoming these barriers?
 - Document the barriers & your needs in [this sheet](#)

Questions to be answered at the end?
Please [put them here!](#)

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