

# Perceived Barriers to Open Science among Researchers in Mathematics, Natural Sciences, and Cognitive Sciences

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# Open Science



Open science is perceived positively across scientific disciplines

# Open Science definition

Open science is defined as an inclusive construct that combines various movements and practices aiming to **make multilingual scientific knowledge openly available, accessible and reusable for everyone [...]**

UNESCO (2021), <https://doi.org/10.54677/MNMFH8546>

# But what does Open Science mean?



Due to a variety of data acquisition, data analysis, and methodologic approaches across the scientific disciplines, there is a diversity of practices and perspectives on open science



Therefore, considerations and sensitivity to the characteristics of the individual research fields and research cultures are crucial to implementing such practices

Banks et al. (2018), <https://doi.org/10.1007/s10869-018-9547-8>

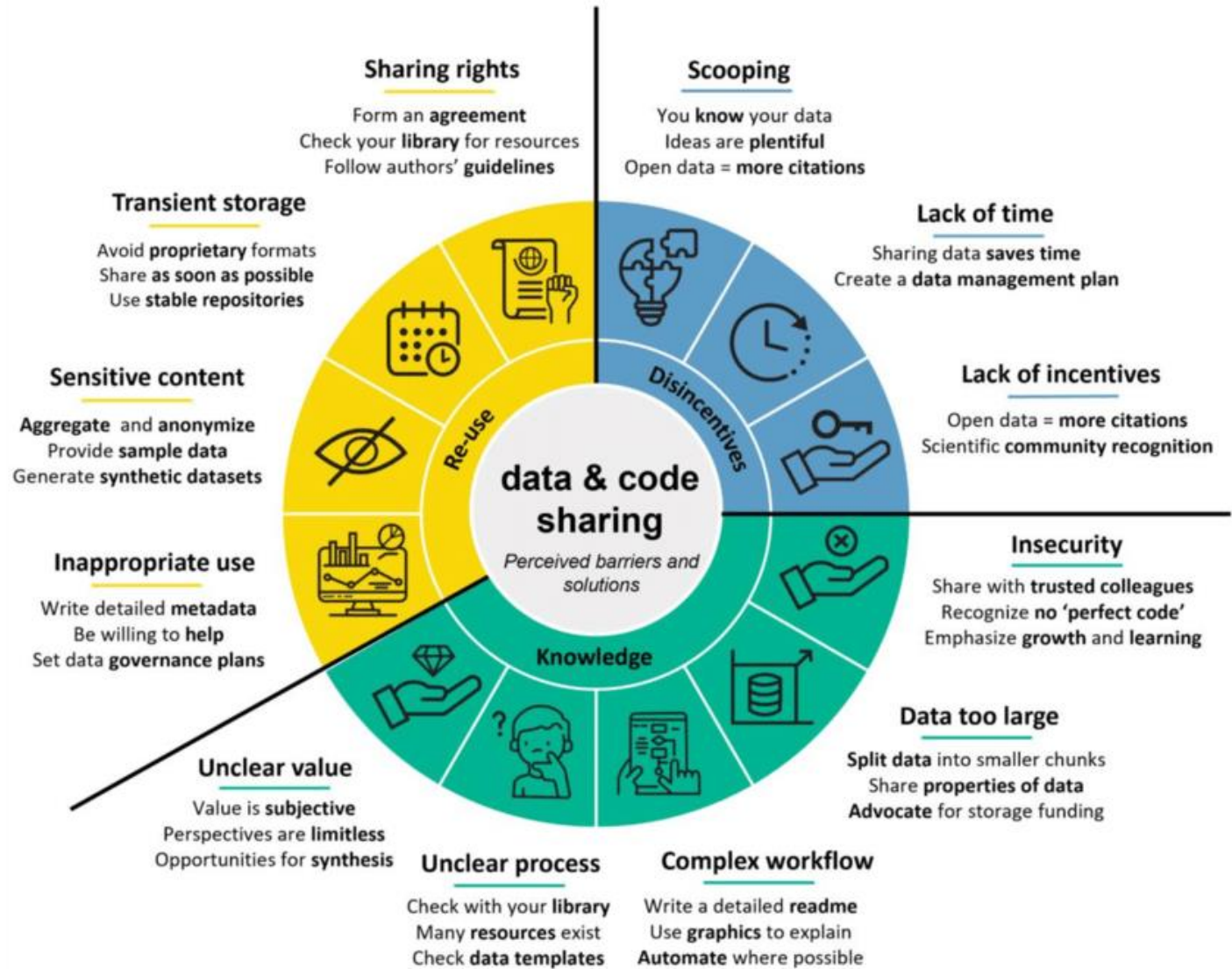
Feger et al. (2020), <https://doi.org/10.1145/3415212>

Stracke et al. (2020), [https://doi.org/10.1007/978-981-15-4276-3\\_2](https://doi.org/10.1007/978-981-15-4276-3_2)

# Structural and Individual Challenges

- **Individual** challenges (challenges that discourage researchers from adopting open science practices)
  - Additional effort required to prepare and share data
  - Lack of formal training in data and software management and open science practices
  - Concerns over intellectual property and the potential misuse of shared data
  - Research culture (e.g., only share data upon specific requests rather than proactively)
  - Concerns about data and software quality
  - Navigating legal and licensing issues
- **Structural** Challenges (imposed by specific research cultures and institutional frameworks)
  - Technical hindrances (e.g., interoperability issues)
  - The reliance on closed and proprietary tools
  - Challenges associated with large data volumes
  - Lacking institutional support structures and regulatory frameworks
  - Strategic concerns complicating the willingness of providers (e.g., privacy, security, legal issues)
  - Misaligned career incentives, where the current academic reward system does not sufficiently recognize or reward open science efforts

# Limitations to Data & Code Sharing



How do researchers from different fields perceive the barriers to open science?



# Objectives

## Questionnaire:

- Based on the Gomes et al. barriers, we created a questionnaire to investigate the respective limitations
- Shared in CRC 1294 – „Data Assimilation“ and intra-faculty unit 'Cognitive Science' at the University of Potsdam

## Objectives:

- How do researchers perceive the individual barriers?
- Which discipline-specific characteristics in perceived barriers and data/software use can we investigate?



# Questionnaire

## General - Questions

<b>Q1.</b> In my publications, the research teams often use code or data to obtain results.	Yes, No, Not sure
<b>Q2.</b> In my publications, I am often (at least partly) responsible for data generation, code, or software development	Yes, No, Does not apply
<b>Q3.</b> I conduct research in the following scientific field(s) (multiple selections possible):	Biology, Chemistry, Cognitive Sciences, Computer Science, Didactics, Geosciences, Health Sciences, Linguistics, Mathematics, Nutritional Science, Physics, Psychology, Sports Sciences, Other

## Barriers - Questions

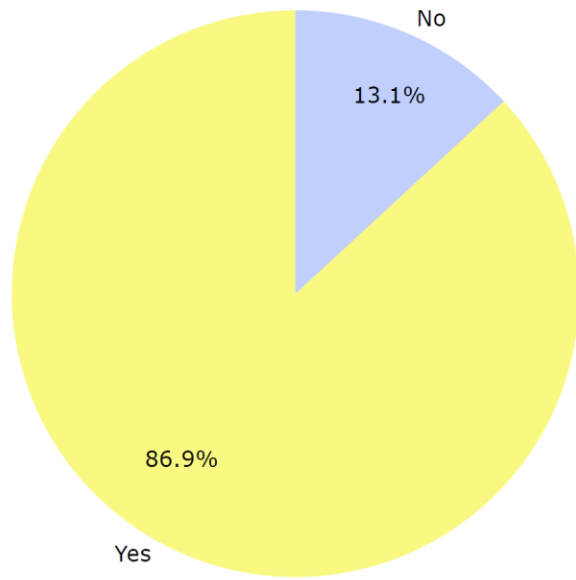
1-Strongly disagree, 2-Disagree, 3-Agree, 4-Strongly agree

### Sharing Data and Software

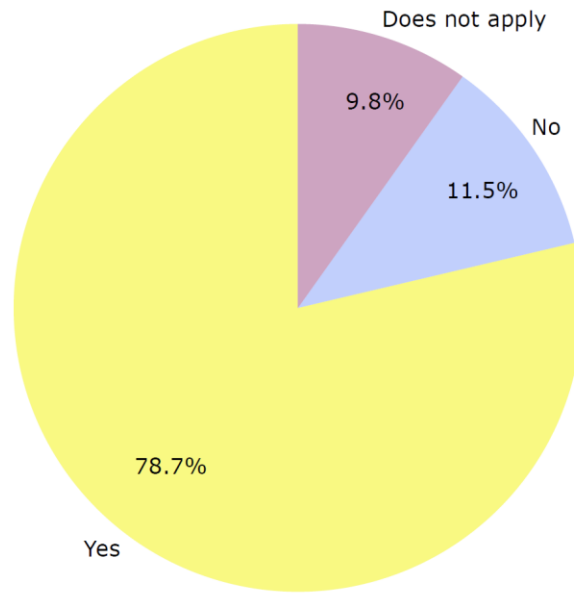
<b>Q4-Unclear Value:</b> I do not see a significant benefit in sharing data or code associated with my publications.	1, 2, 3, 4, Does not apply
<b>Q5-Unclear Process:</b> I am still determining where and how to upload my data or code for sharing.	1, 2, 3, 4, Does not apply
<b>Q6-Complex Workflows:</b> My workflows are usually very complex and can not be shared and applied easily.	1, 2, 3, 4, Does not apply
<b>Q7-Large Files:</b> My datasets are often too large to be shared.	1, 2, 3, 4, Does not apply
<b>Q8-Insecurity:</b> The quality of my generated data or code needs to improve to be appropriately reused by others.	1, 2, 3, 4, Does not apply
<b>Reuse Concerns</b>	
<b>Q9-Inappropriate Use:</b> The data or code I generate could be misinterpreted or misused.	1, 2, 3, 4, Does not apply
<b>Q10-Privacy Concerns:</b> The data or code I generate contains information that would raise privacy concerns when shared.	1, 2, 3, 4, Does not apply
<b>Q11-Sensitive Content:</b> The data or code I generate includes content that may not be in the best interest of science or society when shared.	1, 2, 3, 4, Does not apply
<b>Q12-Transient Storage:</b> There need to be more appropriate long-term publication platforms to publish my data or code	1, 2, 3, 4, Does not apply
<b>Q13-Sharing Rights:</b> I am concerned that the generated data or code ownership is not in my hands anymore when shared.	1, 2, 3, 4, Does not apply
<b>Disincentives</b>	
<b>Q14-Scooping:</b> Sharing the data or code limits my ability to generate further publications from the investigation.	1, 2, 3, 4, Does not apply
<b>Q15-Lack of Time:</b> The commitment to preparing and publishing data or code takes too much of my time.	1, 2, 3, 4, Does not apply
<b>Q16-Lack of Incentives:</b> Sharing my data or code does not benefit my academic career.	1, 2, 3, 4, Does not apply

# Results – General

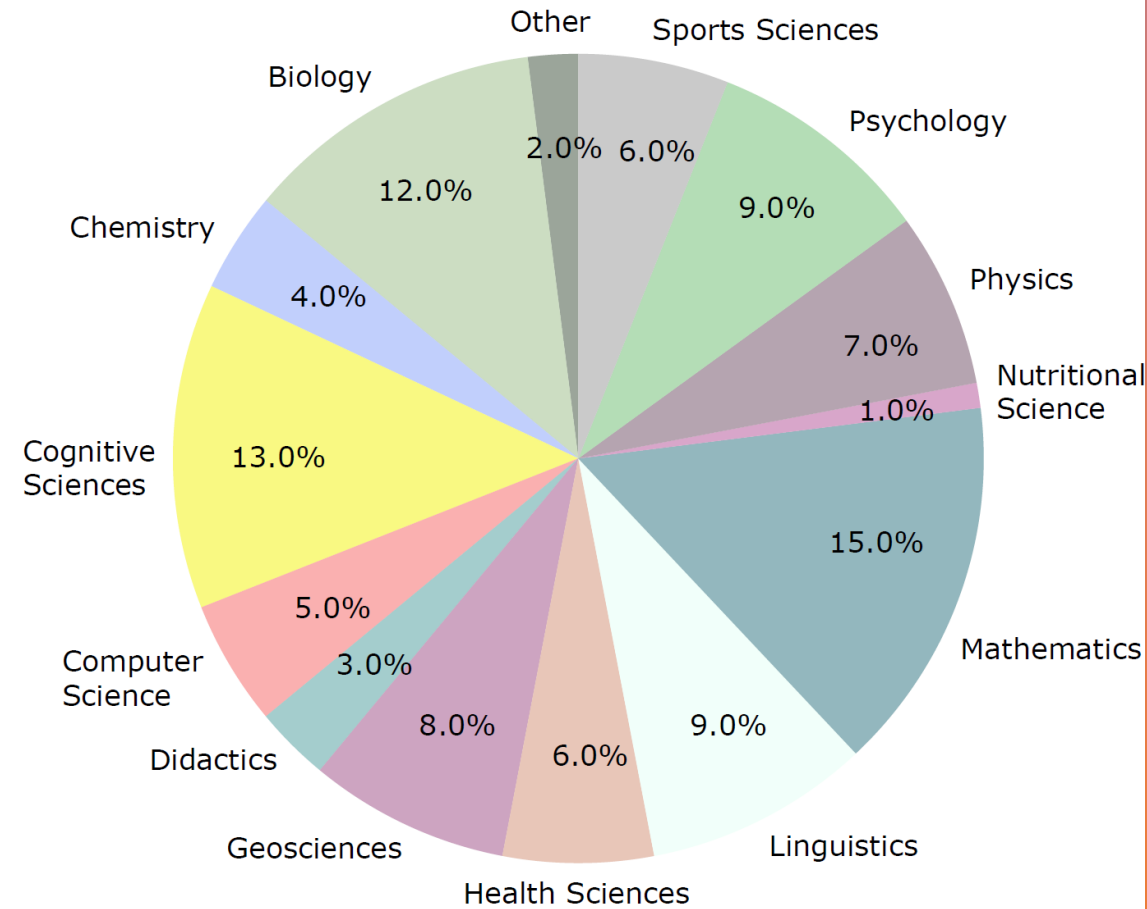
**In my publications, the research teams often use code or data to obtain results**



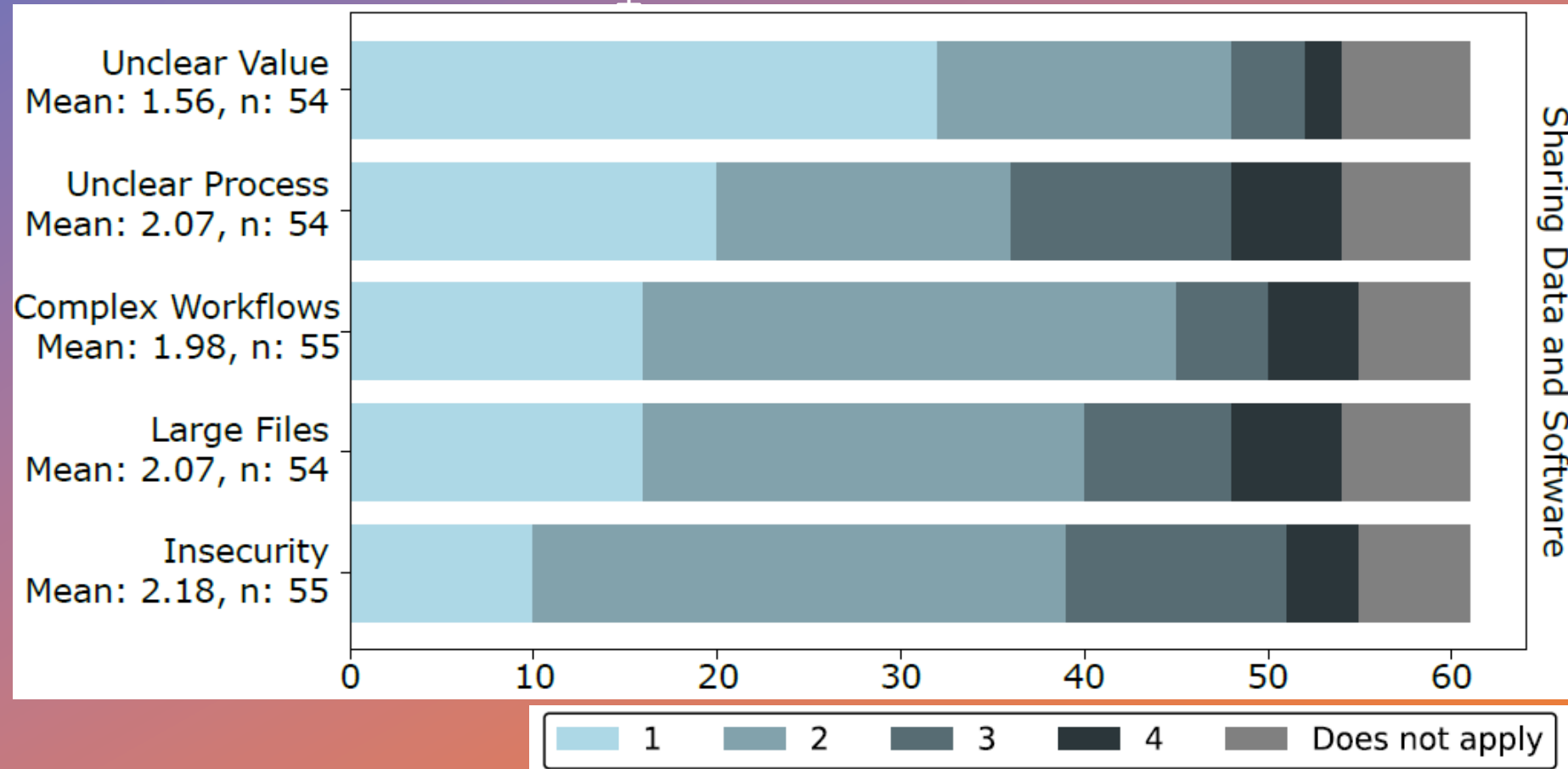
**I am often (at least partly) responsible for data generation, code, or software development**



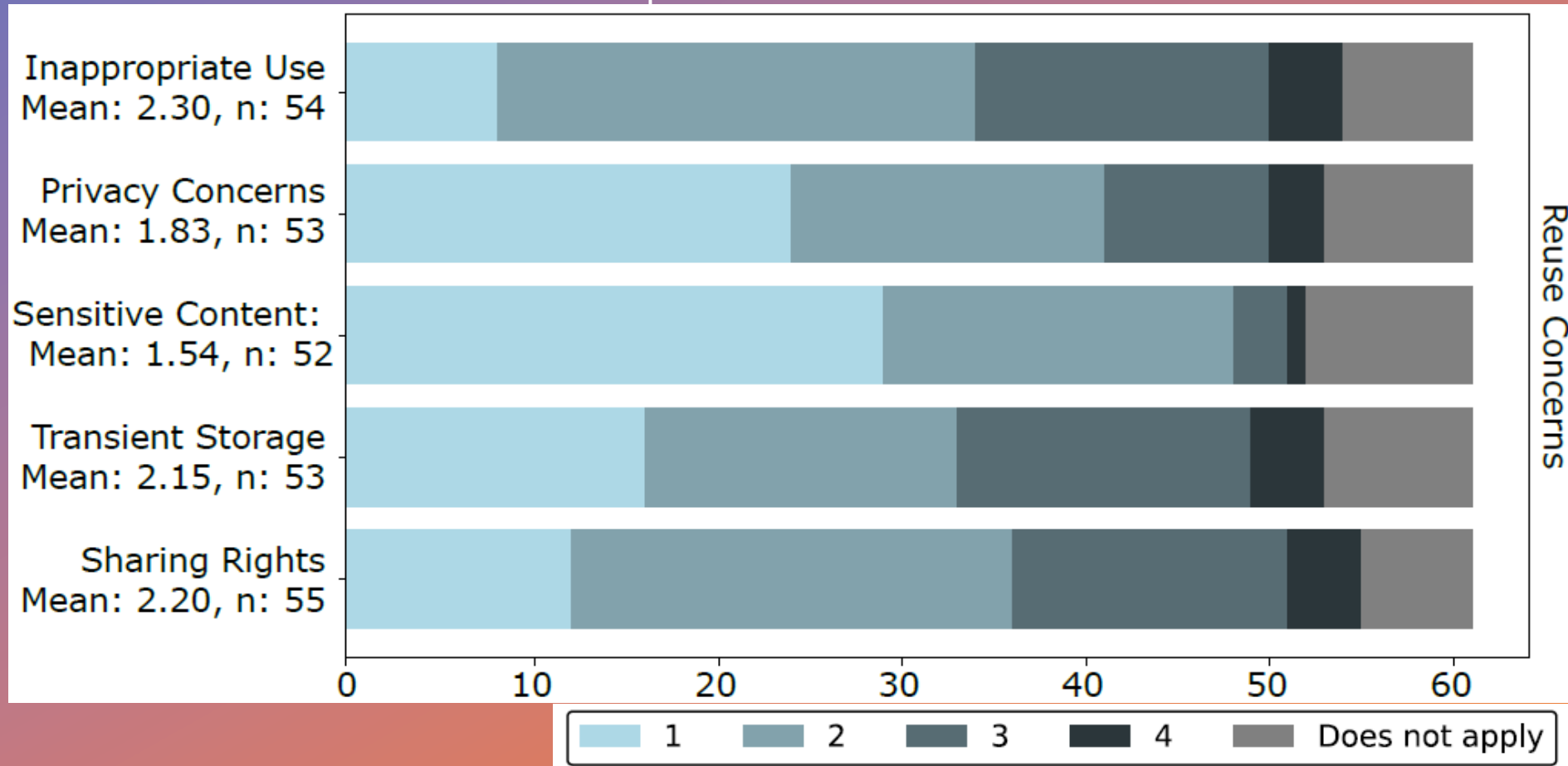
**Research Fields of Participants**



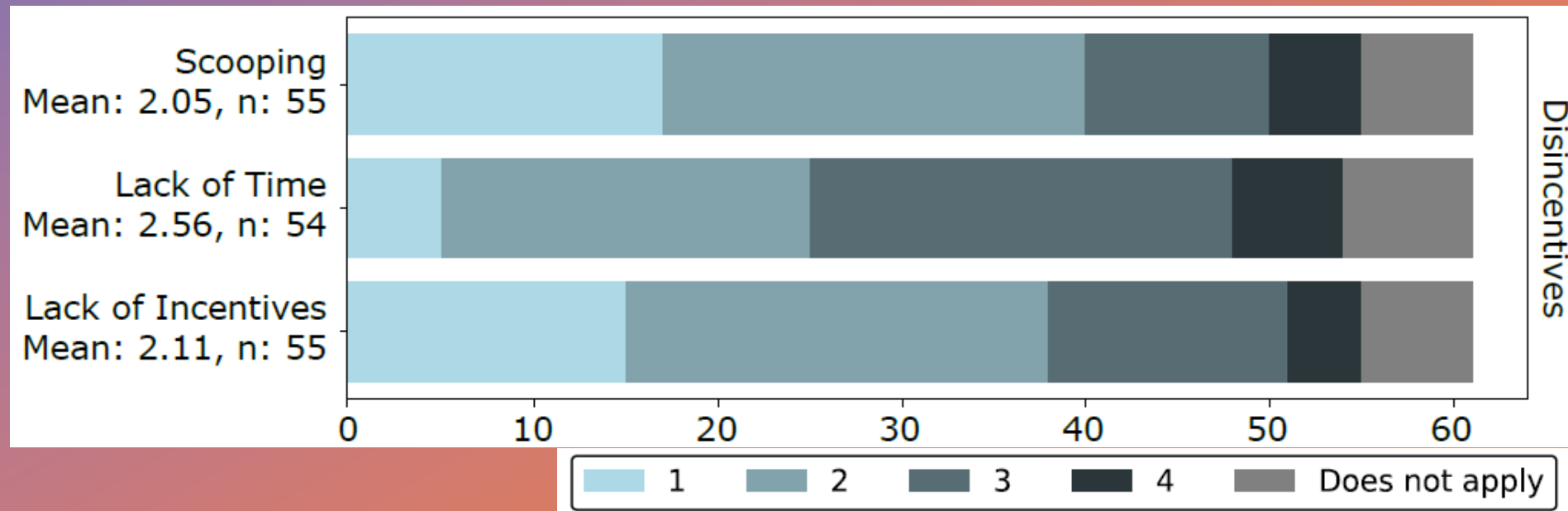
# Results – Sharing Data and Software



# Results – Reuse Concerns



# Results – Disincentives



# Training as a Solution for Perceived Barriers (Individual Barriers)

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## Insecurity, Complex Workflows, Lack of Time

- Training on research data and software management
- Software development with a focus on automation

## Scooping, Unclear Value

- Highlight increased citations and collective benefits

## Inappropriate Use, Sharing Rights, Privacy

- Training on software documentation and licensing
- Implementing restrictive mechanisms like synthetic datasets or access restrictions for sensitive data

# Field-Specific Needs



## Data-intensive fields (e.g., Geosciences, Biology, Linguistics, Physics)

- Focus on strategies for publishing and archiving large datasets

## Fields handling sensitive data (e.g., Cognitive Sciences, Psychology, Didactics)

- Focus on anonymization techniques

Research Field	Large Files	Transient Storage	Privacy
Biology	2	4	-
Chemistry	1	2	1
Cognitive Sciences	-	3	4
Computer Science	1	1	-
Didactics	-	-	3
Geosciences	3	5	1
Health Sciences	1	-	2
Linguistics	2	3	1
Mathematics	-	2	2
Nutritional Science	-	-	-
Physics	2	4	1
Psychology	1	2	3
Sports Sciences	1	3	2
Other	-	-	1

# Barriers Beyond Training



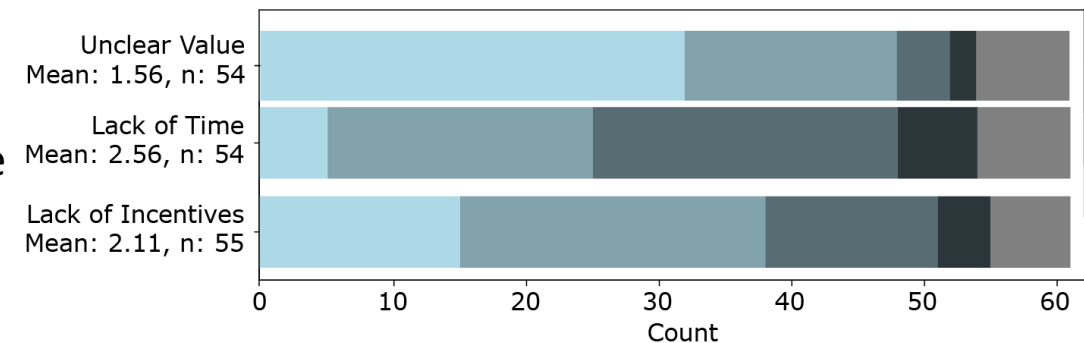
**Certain barriers to sharing research data and software are rooted in cultural and infrastructure frameworks**

## Infrastructure:

- Fields like Geosciences, Biology, and Physics face challenges with Large Files and Transient Storage
- Solution: Provide access to discipline-specific repositories for publishing and archiving large datasets

## Systemic Issues in Academia

- Barriers like 'Lack of Time' and 'Lack of Incentives' critical
- The perceived value, stated in the 'Unclear Value' barrier, is not seen as a major hindrance
- Discrepancy between the perceived value of data/software sharing and the lack of incentives in academia
- Academic reward structures do not align with the importance of open science practices





# Intermediate Conclusion



## Addressing Barriers

- Targeted training can help overcome individual perceived barriers
- Limitations are part of a larger structural context

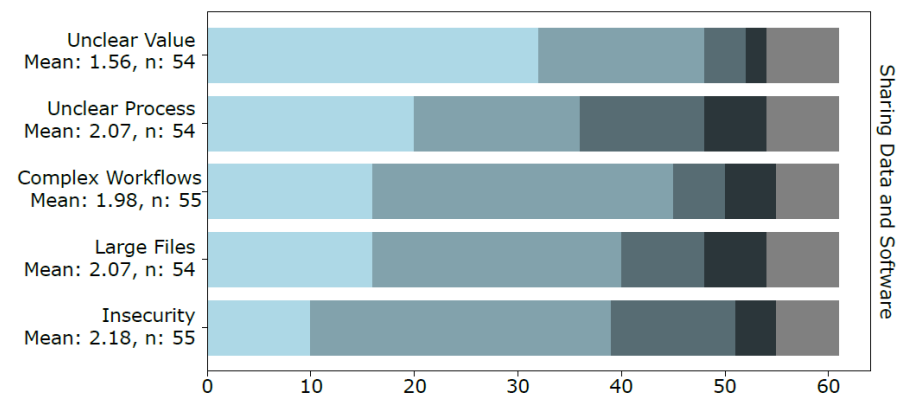
## Need for Support

- Establish appropriate training opportunities and IT infrastructure
- Enable policies from funding agencies, journals, and research institutions
- Promote cultural change towards open science

## Discipline-Specific Considerations

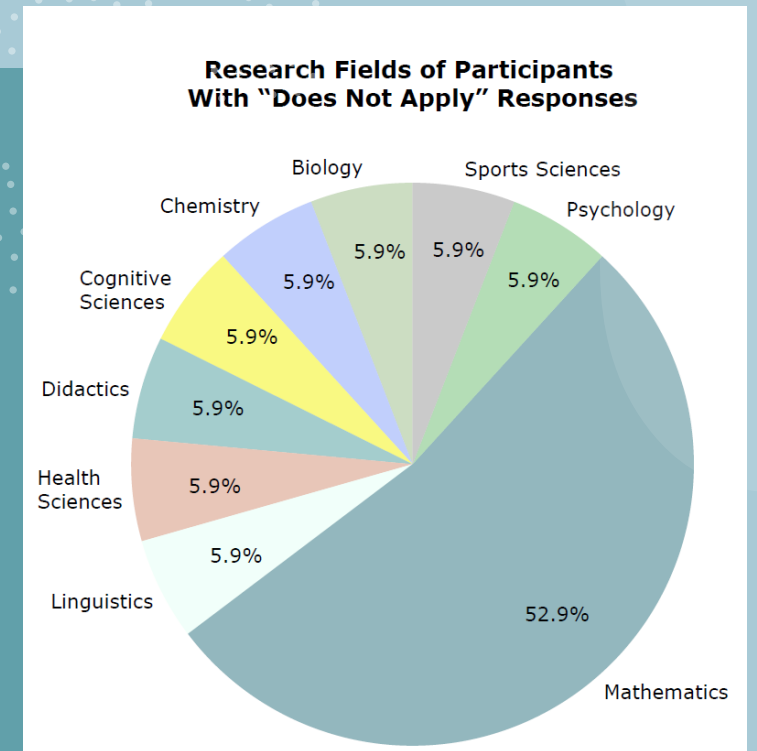
- Recommendations should be connected to discipline-specific open science demands
- Consider the role of data and software in the research process and associated research culture

Participants selected „Does not Apply“ across all questions, which indicates that the defined limitations are irrelevant in some research contexts.



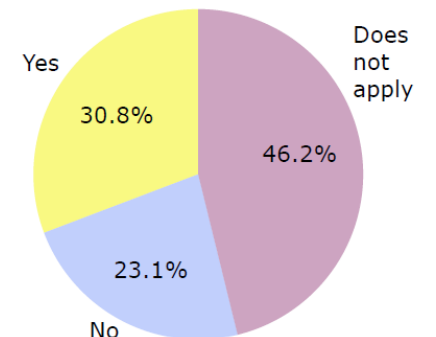
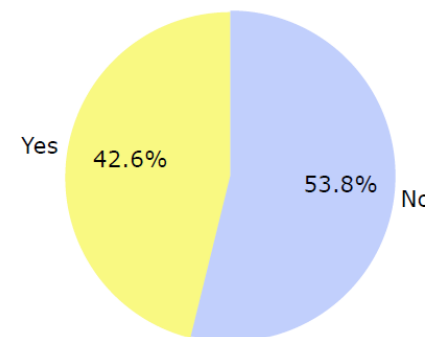
# Research Culture: When Barriers Do Not Apply

- A major proportion of participants who found at least one barrier inapplicable were conducting research in Mathematics



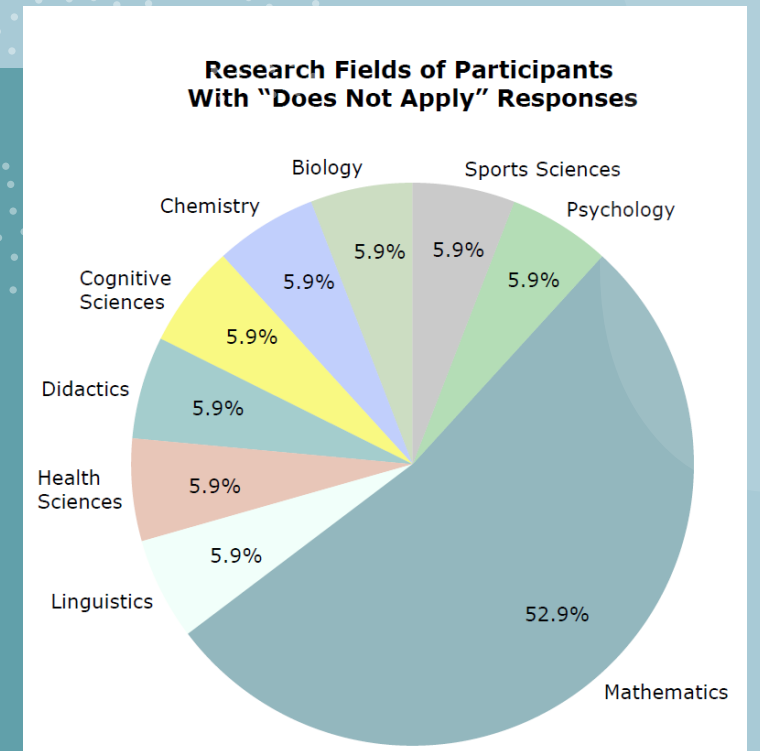
**Participants With "Does Not Apply" Responses**

**Research teams often use data or software to obtain results**      **Responsibility for data or software generation**



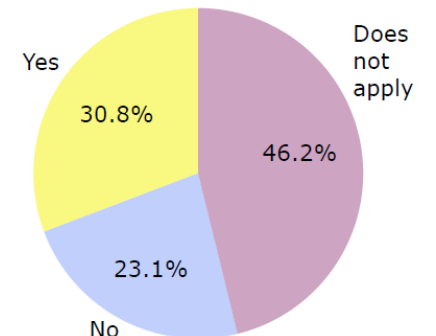
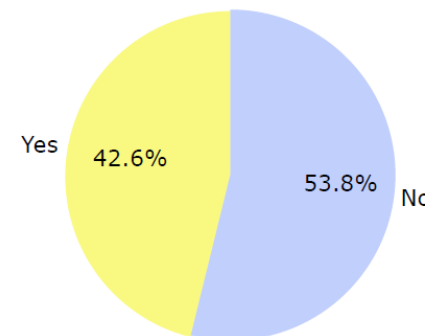
# Research Culture: When Barriers Do Not Apply

- Research culture characterized by methodologies where open science practices are less dependent on data / software
- Research data and software management is not integral to research activities



**Participants With "Does Not Apply" Responses**

**Research teams often use data or software to obtain results**      **Responsibility for data or software generation**



# Conclusions

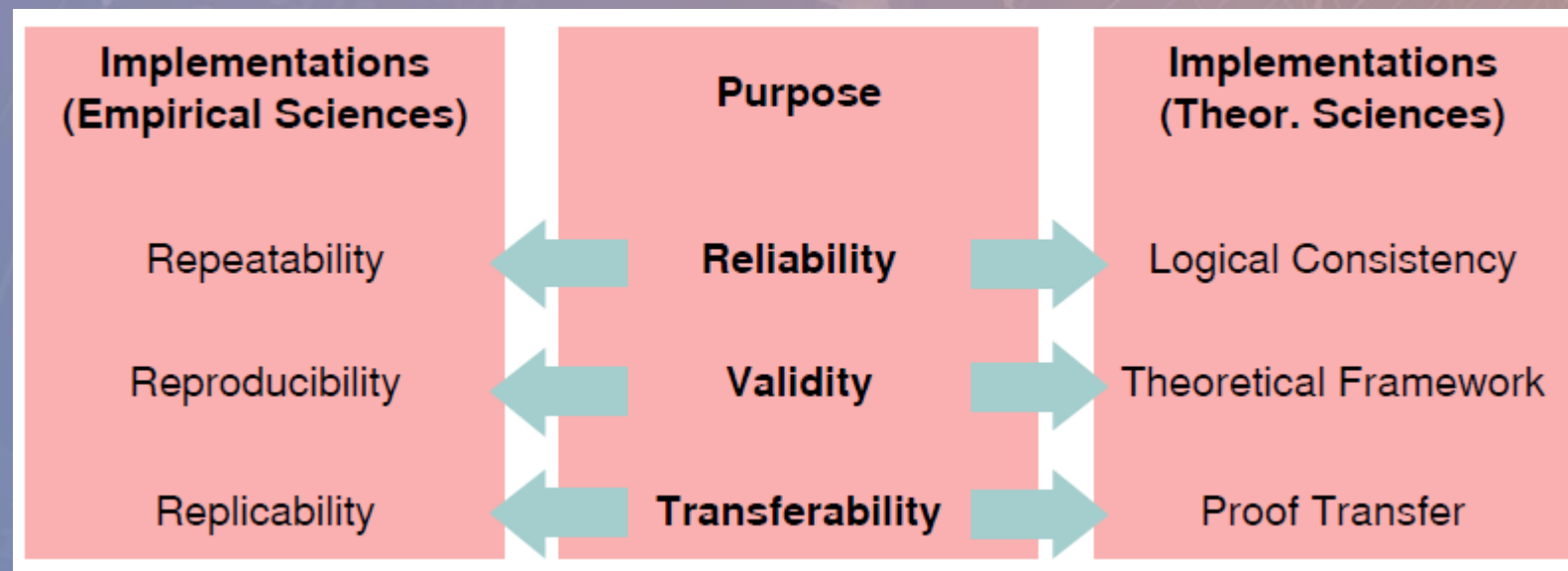


- Researchers recognize the value of Open Science but face significant barriers
- **Key Barriers :**
  - **Knowledge Barriers:** Lack of education in data and software management
  - **Infrastructure Barriers:** Insufficient digital infrastructure for storing and sharing data
  - **Cultural Barriers:** Lack of incentives and academic rewards for open science practices
- **Recommendations:**
  - **Education:** Enhance training in research data and software management
  - **Infrastructure Investments:** Research institutions must invest in digital infrastructure
  - **Recognition Systems:** Revise academic hiring and rewards to prioritize open science
  - **Research culture:** Investigate discipline-specific needs, especially in fields with a strong focus on theoretical research, where open science practices differ from empirical sciences.

# Transdisciplinary Discussions



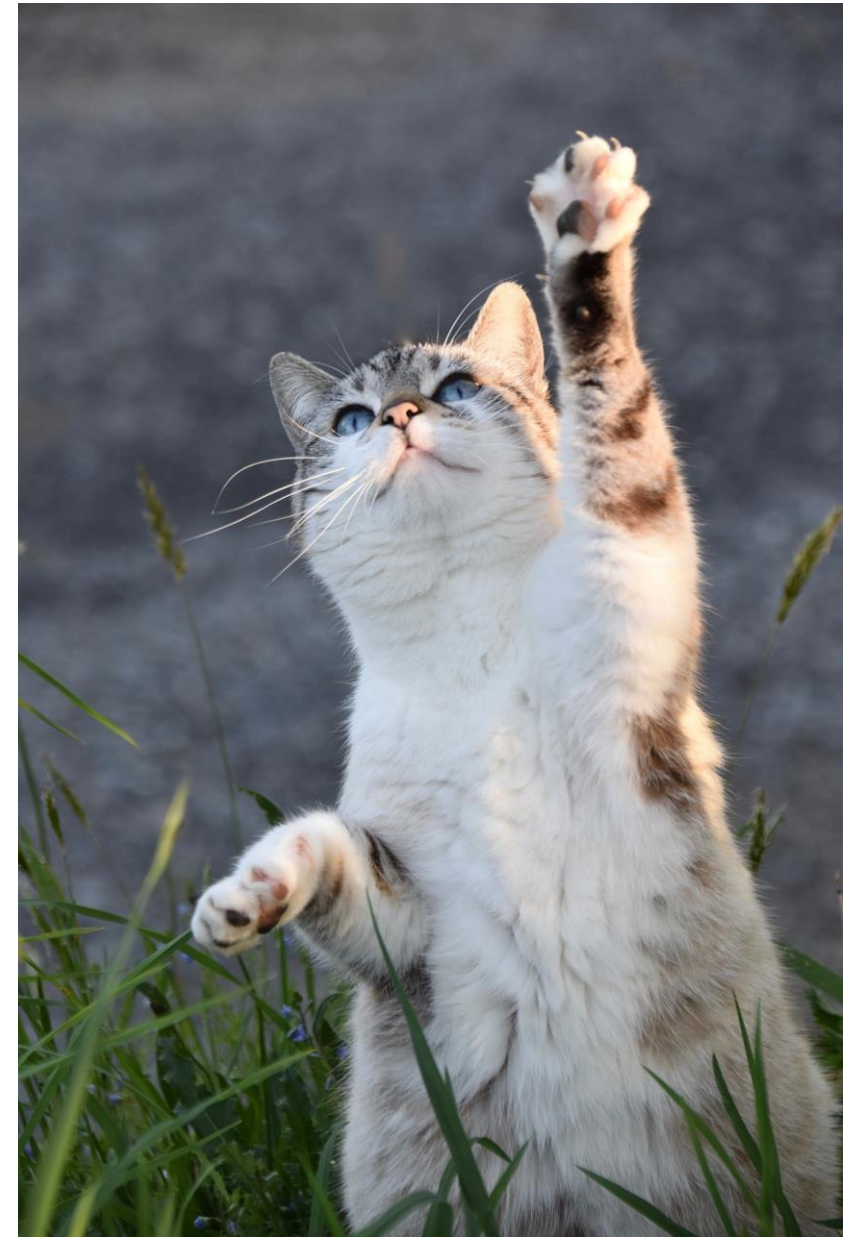
Would prioritizing the purpose over the implementation enhance transdisciplinary discussions about Open Science?



# Thankyou!

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