

FAIR theory: Applying Open Science Principles to the Construction and Iterative
Improvement of Scientific Theories

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Abstract

22

23 Test test.

24 *Keywords:* meta theory, theory formation, cumulative science, formal models

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FAIR theory: Applying Open Science Principles to the Construction and Iterative Improvement of Scientific Theories

The FAIR Guiding Principles (hereafter: FAIR principles) were established to improve the reusability of research data by making them more findable, accessible, interoperable and reusable [REF] for both humans and computers. Since their inception in 2014, scholars have demonstrated their relevance for making other information artefacts more open, such as research software (Lamprecht et al., 2019) and computational workflows (S. R. Wilkinson et al., 2024). This paper argues that the FAIR principles can advance effective and transparent scholarly communication about theory. To this end, we introduce “FAIR theory”: a digital representation of a scientific theory, compliant with the FAIR principles. By improving the efficiency of scholarly communication, FAIR theory has the potential to foster and accelerate cumulative knowledge acquisition and ultimately advance social scientific research.

The Need for FAIR theory

The so-called “replication crisis” has prompted extensive reforms in social science (Lavelle, 2021; Scheel, 2022). Concern that undisclosed flexibility in analyses was a major factor for the abundance of non-replicable findings led to widespread adoption of open science practices like preregistration and replication (Nosek et al., 2015). These various practices ensure transparent and repeated testing of hypotheses. However, recent reviews show that most preregistered hypothesis tests are not supported by empirical evidence (Scheel, Schijen, & Lakens, 2021). Thus, increased rigor in testing has revealed that the root cause of the replication crisis is more fundamental: Psychological theories rarely produce hypotheses that are corroborated by evidence. Furthermore, theories are often so vague that they can accommodate findings that are mutually inconsistent, as the theory’s central claims evade falsification.

Scholars have been raising concerns about the state of theory in social science for

nearly 50 years (Paul E. Meehl, 1978; Robinaugh, Haslbeck, Ryan, Fried, & Waldorp, 2021). Two main concerns are that, first, social scientific theories lack precision compared to theories in the physical sciences (Szollosi & Donkin, 2021). and clarity In other words, social scientific theories lack *formalization*, which means that they do not make very accurate predictions, and are thus hard to falsify and difficult to understand on their own, without either substantial interpretation or additional background knowledge. A second concern is the lack of transparent and participative scholarly communication about psychological theory and their development over time.

Given these concerns, it is an imbalance that scientific reform initiated by the open science movement has focused primarily on improving deductive methods. The equally critical inductive processes of theory construction and theory improvement have been largely overlooked. The present paper restores balance by applying, for the first time, open science principles to psychological theory. We apply the FAIR principles to scientific theories, introducing the concept of *FAIR theory* to facilitate transparent scholarly communication and accelerate cumulative knowledge acquisition.

Theory and Scientific Progress

According to the *empirical cycle* (de Groot, 1961), a philosophical model of cumulative knowledge acquisition, research ideally follows a cyclical process with two phases (Figure 1). In the deductive phase, hypotheses derived from theory are tested on data. In the inductive phase, patterns observed in data are generalized to theoretical principles. In this model, theories are the vehicle of scientists' understanding of phenomena. Ideally, they are iteratively updated based on deductive testing and inductive theory construction.

In a progressive research program (Lakatos, 1971), this cycle is regularly completed to iteratively advance our understanding of the studied phenomena . There are, however, indications that contemporary psychology falls short of this ideal. Firstly, because

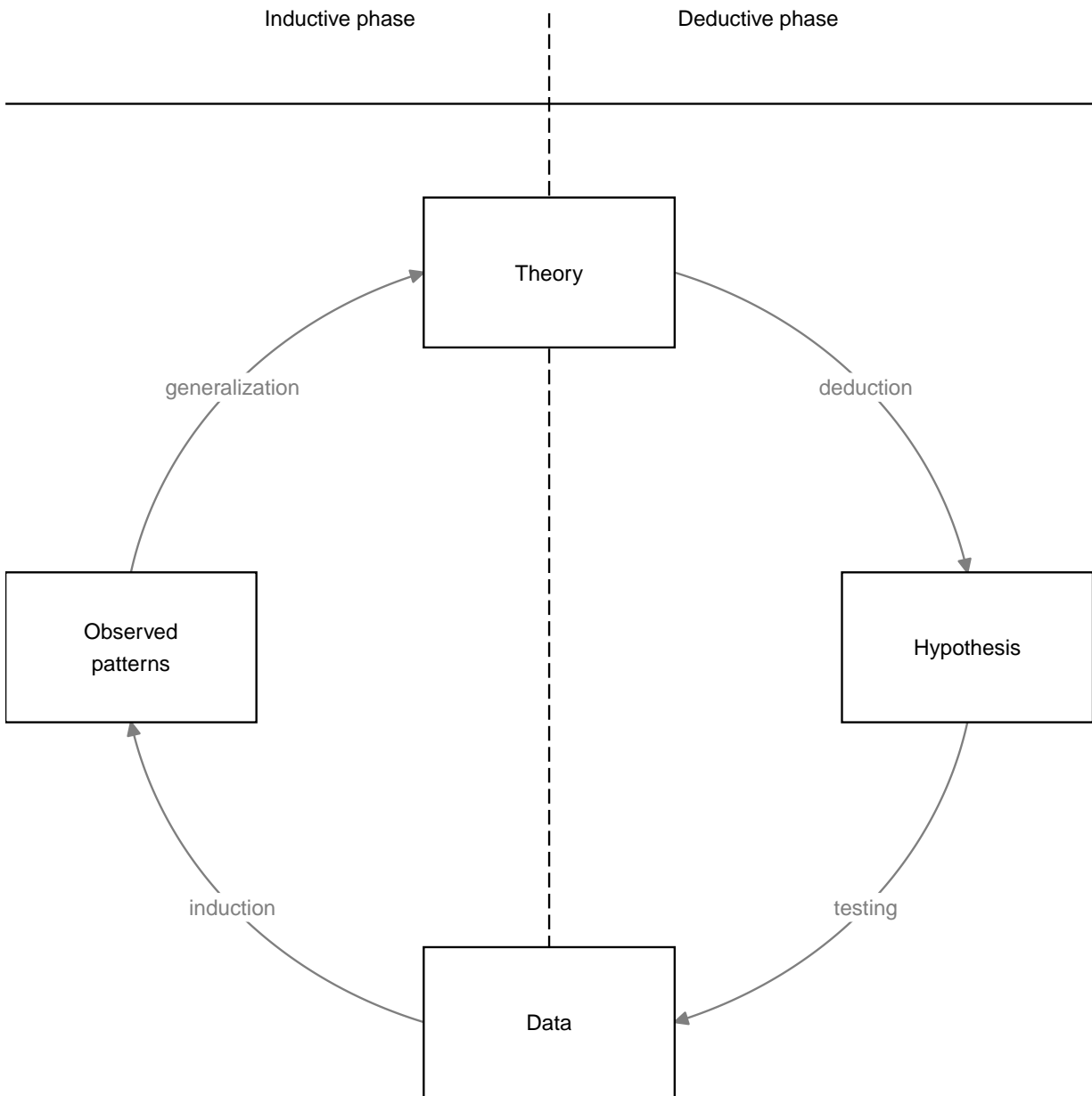


Figure 1. A take on the empirical cycle by De Groot

hypothesis-testing research is over-represented in the literature: According to Kühberger, Fritz, and Scherndl (2014), 89.6% of papers published in psychology report confirmatory hypothesis tests. Closer examination of reported deductive research reveals, however, that the link between theory and hypothesis is often tenuous (Oberauer & Lewandowsky, 2019; Scheel, Tiokhin, Isager, & Lakens, 2021). Only 15% of deductive studies referenced any theory, and theory was often not cited in relation to the hypothesis (McPhetres et al., 2021). The remaining 85% of deductive studies lacked an explicit derivation chain from theory to hypothesis. In the best case, such ungrounded hypotheses are rooted in researchers' implicit theories, in which case it is particularly important to make these explicit (Norouzi, Kleinberg, Vermunt, & Van Lissa (2024). Or, perhaps the hypotheses are not of substantive interest, such as null hypotheses that exist purely for the purpose of being rejected (Van Lissa et al., 2020), and researchers are simply testing them as part of a cultural ritual (Gigerenzer, Krauss, & Vitouch, 2004). Testing ad-hoc hypotheses not grounded in theory does not advance our principled understanding of psychological phenomena. Put differently: collecting significance statements about ad-hoc hypotheses is much like trying to write novels by collecting sentences from randomly generated letter strings (van Rooij & Baggio, 2021).

Theory thus has an uncomfortable and paradoxical role in contemporary psychology: The majority of papers ostensibly test hypotheses, but these are rarely derived from theory, and test results do not routinely contribute to the improvement of existing theories. The paradoxical role of theory in psychology is perhaps best described by Meehl's observation that theories in psychology "lack the cumulative character of scientific knowledge. They tend neither to be refuted nor corroborated, but instead merely fade away as people lose interest" (Paul E. Meehl, 1978).

Making Theory FAIR

The present paper addresses the lack of open science methods for theory development and suggests an improvement of the state of affairs by applying the FAIR principles to scientific theories. Merely publishing theory – as in a classic research article – does not make it open; to be open, theory should adhere to established open science standards. We apply the FAIR principles to digital representations of theory, introducing a FAIR metadata format to make theories *Findable* via a DOI, *Accessible* in a machine- and human-readable filetype, *Interoperable* within the data analysis environment, and *Reusable* in the practical and legal sense, so that they may be improved over time – at best, in a participative process. Digital representations of theory intentionally is a broad term, particularly including textual representations of a given theory, as well as formal representations, such as mathematical notation, algorithmic pseudo code, or a set of logical clauses. Following the original proposal of Lamprecht and colleagues, we adapt the FAIR principles for theory, see Table 1. We reflect on the necessary changes (which are minor), as well as on the current state and future of FAIR theory in the social sciences. The resulting principles provide guidance for instantiating theory as a FAIR information artifact, and we provide worked examples to encourage their adoption.

Table 1

Criterion	Original	Theory	Action
F1	(Meta)data are assigned a globally unique and persistent identifier	Theory (meta)data has a global, unique and persistent identifier	Rephrased
F2	Data are described with rich metadata	Theory is described with rich metadata	Rephrased
F3	Metadata clearly and explicitly include the identifier of the data it describes	Metadata clearly and explicitly include identifiers for all the versions of the theory it describes	Rephrased and extended
F4	(Meta)data are registered or indexed in a searchable resource	Theory and its associated metadata are included in a searchable repository	Rephrased, needs work
A1	(Meta)data are retrievable by their identifier using a standardized communications protocol	Theory and its associated metadata are accessible by their identifier using a standardized communications protocol	Rephrased
A1.1	The protocol is open, free, and universally implementable	The protocol is open, free, and universally implementable	Remain the same
A1.2	The protocol allows for an authentication and authorization procedure, where necessary	The protocol allows for an authentication and authorization procedure, where necessary	Remain the same, but less relevant

Table 1 continued

Criterion	Original	Theory	Action
A2	Metadata are accessible, even when the data are no longer available	Theory metadata are accessible, even when the theory is no longer available	Rephrased, but less relevant
I1	(Meta)data use a formal, accessible, shared, and broadly applicable language for knowledge representation	Theory and its associated metadata use a formal, accessible, shared and broadly applicable language to facilitate machine readability and reuse	Rephrased and extended
I2	(Meta)data use vocabularies that follow FAIR principles	(Meta)data use vocabularies that follow FAIR principles	Rephrased
I2S.1	-		
I2S.2	-		
I3	(Meta)data include qualified references to other (meta)data	(Meta)data includes qualified references to other (meta)data, including previous versions of the theory	Extended
I4S	-		

Table 1 continued

Criterion	Original	Theory	Action
R1	(Meta)data are richly described with a plurality of accurate and relevant attributes	Theory and its associated metadata are richly described with a plurality of accurate and relevant attributes	Rephrased
R1.1	(Meta)data are released with a clear and accessible data usage license	Theory (meta)data are released with a clear and accessible license	Rephrased
R1.2	(Meta)data are associated with detailed provenance	Theory (meta)data are associated with detailed provenance	Rephrased
R1.3	(Meta)data meet domain-relevant community standards	Theory (meta)data and documentation meet domain-relevant community standards	Rephrased

There are different definitions of theory, and many of those definitions are consistent with the FAIR theory principles. This paper defines theory as an integrated set of statements that explain phenomena consistently evidenced by patterns in data [bogen/woodward]. Meehl (1990) provides guidance as to what kinds of “statements” such a theory might contain: statements about the types of entities postulated (i.e., ontology), statements about causal connections between those entities, statements about the functional form of those connections, and statements about their specific numerical values Guest (2024).

Some have defined a model as a “specific instantiation of theory narrower in scope and often more concrete, commonly applied to a particular aspect of a given theory” [REF Fried]. This invites the question: if a FAIR theory is a specific instantiation of theory, how does FAIR theory differ from a model? While there may be philosophical differences, these differences are largely irrelevant for the purpose of making theories more FAIR. When organizing knowledge, we think it is helpful to view theories and models derived from them as existing along a continuum of specificity, from broad to narrow, from general to specific, where a theory has a relatively broader scope and may contain one or more models as specific instances. For example, following Meehl, we could envision a theory that merely specifies how specific constructs are causally connected. From this theory, we could derive a more specific *statistical model* by assuming functional form (e.g., linear effects) and error families (e.g., normal distributions). This statistical model makes just enough assumptions to allow estimation of the remaining unknown parameters (e.g., regression slopes) from data. Or, we could derive an even more specific *generative/computational model*, which is completely parametrized (i.e., specific values of regression slopes are also assumed) such that an interpreter (e.g., the R programming language) can use the model to generate new data. Note that broadness and narrowness are relative terms, and one person’s theory may be another person’s model. This definition also implies that FAIR models are a special case of FAIR theory.

As an applied example, consider a comprehensive theory of disease spread and

144 pandemics which covers various psychological factors such as adherence to pandemic
 145 mitigation methods (e.g.,), pandemic-related social disruption (e.g., panic buying), or
 146 pandemic-related distress and related problems (e.g., anxiety) (Taylor, 2022). The theory
 147 may encompass a particular transmission *model* for disease spread including precise
 148 parameters for the process of infection (e.g., social distance, average duration of encounters,
 149 ventilation) and incubation times.

150 **The Role of Theory Formalization.** Concerns about the state of theory in the
 151 psychological literature revolve around two issues: theory formalization and theory (re-)use.
 152 Greater formalization increases theories' *empirical content* [REF] as it forces researchers to
 153 use precise statements, for example by specifying exact functional forms for relations or
 154 forces them to specify processes that would otherwise get away with their vague formulation
 155 in verbal theories. For example, the phonological loop in Baddeley's verbal description of his
 156 working memory model allows at least 144 different implementations, for example, depending
 157 on how decay rate, recall success, or rehearsal sequence are precisely implemented
 158 (Lewandowsky & Farrell, 2010). More precise theories makes them easier to falsify, which
 159 necessitates revising them, thus advancing our principled understanding of the phenomena
 160 they describe. FAIR theory does not require theories to be formal, and formal theory can be
 161 represented in a way that is not FAIR. It is therefore important for us to emphasize that
 162 FAIR theory imposes no restrictions on researchers regarding the manner in which theories
 163 are derived and written down. The guidelines introduced by FAIR theory primarily pertain
 164 to how theories are documented and shared in digital environments, with the aim of
 165 enhancing their reusability and extensibility. For example, it is possible to represent a
 166 collection of verbal propositions (perhaps derived through qualitative research) as a FAIR
 167 theory. Conversely, a directed acyclic graph (DAG) is a type of formal theory (for example,
 168 see the empirical cycle in Figure 1), but if it is embedded within a journal article as a
 169 bitmap image without any key words to help search engines index that article as a theory
 170 paper, it is not FAIR. FAIR theory is thus consistent with, but does not require, formal

theory (also see *Accessibility*).

Findability

Making theories Findable would allow researchers to easily identify relevant theories to inform their hypotheses, grounding their work in established theoretical foundations. Making theories Findable also increases the impact and reuse potential of theories across disciplines, either through direct application (where one discipline stumbles upon a problem that is already well-understood in another discipline), or through analogical modeling. In analog modeling, the structure of a theory from one discipline is applied to a phenomenon in another field. For example, predator-prey models have inspired theories of XXX, and the Eysenck model of atomic magnetism has inspired a network theory of depression. Findability also enables meta-research on theories, in the same way libraries and search engines have enabled scholars to study the literature via systematic reviews. In a similar way, it would become possible to compare all theories of a specific phenomenon, or to study structural properties of theories.

The four Findability criteria are applicable to theory with only minor adjustments, see Table 1. First, this requires assigning a globally unique and persistent identifier, or DOI, to each theory (F1). Of the many services that provide DOIs for scientific information artefacts, Zenodo and the Open Science Framework are commonly used in psychology. Second, Findable theory is described with rich metadata (F2). This includes citation metadata (e.g., referencing a scientific paper that documents the theory, or a psychometric paper that operationalizes specific constructs). It might further include domain-specific metadata, such as a reference to a taxonomy of psychological constructs (Bosco, Uggerslev, & Steel, 2017), ontology (Guyon, Kop, Juhel, & Falissard, 2018), or catalog of psychological phenomena [REF Noah Denny]. Metadata should also include identifiers for all the versions of the theory it describes (F3); Zenodo handles this by default by providing an overarching DOI for an information artifact which subsumes the DOIs of that artifact's versions. Finally, these

metadata should be registered or indexed in a searchable registry (F4). This final criterion is less straightforward. Ideally, FAIR theories should be indexed in search engines used by academics, like Google Scholar. At present, however, these search engines are designed to index traditional print publications. The *data paper* solves this problem for research data; the idea is that scholars publish a paper (or even preprint) as documentation for the data resource [REF McGillivray on data papers]. The data paper is indexed by search engines, and in turn points to the relevant information artifact. The same solution could be applied to theories - but it seems superfluous to generate papers whose only purpose is to redirect to a specific resource. Another solution is to manually index FAIR theories, for example by adding them to one's Google Scholar profile, or entering them in PURE.

At present, theories have poor findability, which impedes cumulative knowledge acquisition. One factor contributing to theories' lack of Findability is the lack of standardized metadata, or even a standardized keyword to signal the presence of theory within a paper - terms like "theory", "model", and "framework" are used interchangeably. To curb this trend, we suggest using the keyword "FAIRtheory" for all resources that constitute or reference a FAIR theory (separating the words FAIR and theory by a space or hyphen would lead them to be interpreted as separate tokens in many search engines. This would allow theoretical resources to be systematically indexed, tagged, and made searchable. Another factor contributing to the present lack of Findability is that the primary unit of dissemination and search in psychology is still the academic paper. A paper may contain multiple resources - such as materials, data, code, and theory - but if these are not merely described in text, and not instantiated as separate informational artefacts, their findability is limited. This would be achieved by modular publishing of theories as individually citable academic assets, with adequate metadata that is indexed in standardized repositories, similar to the current practice of publishing empirical data in standardized repositories (e.g., DataVerse). As with empirical data, these theories could still be connected to a specific paper which might serve as documentation and the canonical reference for the resource.

There have been notable efforts to improve theories' findability through post-hoc curation. For example, Gray and colleagues introduced a format for representing theories, and post many examples on their website (Gray, 2017). Similarly, Borsboom and colleagues seek to establish a database of psychological theories [REF BORSBOOM]. Post-hoc curation is a notable effort but does not address the root cause of the lack of Findability, however. Ideally, Findability would be addressed ante-hoc, through documentation with rich metadata and modular publishing.

Accessibility

Transparent scholarly communication about theory requires that theories are accessible to all researchers and other stakeholders. If theories are accessible, researchers can reuse and refine them, thus accelerating cumulative knowledge acquisition. Making theories accessible also allows stakeholders (e.g., practitioners, policy makers, advocates) to inform themselves of the current scientific understanding of specific phenomena. While isolated empirical findings can appear fragmented and contradictory (Dumas-Mallet, Smith, Boraud, & Gonon, 2017), theories offer a top-down, big picture representation of the phenomena studied in a field. In other words, theories are an important instrument in science communication.

The Accessibility criteria serve to *regulate* access, not to maximize it. These principles apply to theory with minor changes, with the caveat that there might be less of a need to restrict access to theory than there is for (human subjects) data. Firstly, theory and its associated metadata should be accessible by their identifier using a standardized communications protocol (A1). This can be achieved, for example, by hosting theory in a version-controlled remote repository (such as git), and archiving that repository on Zenodo for long-term storage. The resulting resource will then have an identifier (DOI) which allows the theory to be accessed using a standardized communications protocol (download via `https` or `git`). Secondly (A2), theory metadata should be accessible, even when the theory is no longer available, which is also achieved via long-term storage (e.g., on Zenodo). Git

remote repositories allow for access control, and Zenodo allows for access control of individual files/resources. An unavailable theory typically refers to a theory that was abandoned in favor of a better or more general theory (such as the phlogiston theory, which was superseded by thermodynamics). In general, it makes sense to keep outdated theories, in order to be able to track the genesis of theories over time, yet, we require the availability of meta data as a minimum requirement.

At present, there are several impediments to theories' accessibility. To the extent that theories are still contained within papers, paywalls erected by commercial publishers constitute a barrier. Open Access publishing thus increases the accessibility of all academic output, including theory. A second impediment is more indirect: While open access publishing increases practical access to theories, accessibility also requires clear and explicit communication. This property of good theories has been dubbed "discursive survival [...], the ability to be understood" (Guest, 2024). The current prevalence of strategic ambiguity renders psychological theory difficult to understand (Frankenhuis et al., 2023). It is important to acknowledge the *indeterminacy of translation* (Quine, 1970): which holds that every communicative utterance has multiple alternative translations, with no *objective* means of choosing the correct one. It follows that an idea cannot be formalized to the point that it becomes unambiguously interpretable. This places a theoretical upper bound on theories' ability to be understood.

Successful communication requires shared background knowledge between sender and receiver (Vogt et al., 2024). The Kuhnian notion of "normal science", conducted within the context of a shared paradigm, provides shared background knowledge to facilitate mutual understanding (Kuhn, 2009). From a pragmatic perspective, these considerations indicate that, when striving to make theory accessible, it is important to be as explicit as possible (e.g., about assumptions and ontological definitions), while acknowledging that accessibility exists on a spectrum, and that it is impossible to eliminate all ambiguity. Rather, it may benefit scientific discourse to anticipate misunderstanding, and use it to drive further

explication of theory. In sum, efforts to communicate theory clearly, with as few dependencies on shared background knowledge as possible, including by formalization, embedding within shared contexts, and explication of assumptions, will advance its Accessibility.

A third impediment arises when theories have a “dependency on the author” (DOA). DOA occurs when a theory cannot be understood by independent scholars, thus requiring the original author for interpretation and clarification. We have heard DOA referred to apocryphally as the “ask Leon” phenomenon, as graduate students were supposedly told to ask Leon Festinger to explain to them how their misconstrual of cognitive dissonance theory had caused their experiments to yield null results. DOA relates to the discourse on “Great Man Theorizing” (Guest, 2024) because it enables gatekeeping: an author could insist that work requires their involvement or denounce work conducted outside their purview as illegitimate, which violates checks and balances of scientific research. DOA also renders theories immune to refutation, because the author can claim that the theory was misconstrued when confronted with falsifying evidence, thus making it a moving target (Szollosi & Donkin, 2021). The fact that DOA is inherently problematic is illustrated by cases where third parties identify logical inconsistencies within a theory (e.g., Kissner, 2008). This demonstrates that original authors are not the ultimate authority on their theories. DOA thus unduly impedes scientific progress, and authors should make good-faith efforts to make theories as accessible as possible; both in terms of availability and in terms of interpretability.

Interoperability

Interoperability pertains to the property of information artefacts to “integrate or work together [...] with minimal effort” (M. D. Wilkinson et al., 2016). The original interoperability principles can be rephrased somewhat to apply to theory. Firstly, theory and its associated metadata should use a formal, accessible, shared and broadly applicable

language to facilitate (human- and) machine readability and reuse (I1). The common practice of instantiating theory as lengthy prose or multi-interpretable bitmap image falls short of this ideal. Instead, FAIR theory should, ad minimum, be instantiated as as a type of data that is human- and machine-readable with as few interpretative steps as possible, as previously recommended (Van Lissa et al., 2021). Depending on the level of formalization of the theory, different formats may be appropriate, such as verbal statements in plain text, mathematical formulae, and statements expressed in some axiomatic system. Examples of the latter include pseudo-code, interpretable computer code, and Gray’s theory maps (Gray, 2017). While a theory represented as a bitmap image is not very interoperable, the same image represented in the DOT language for representing graphs does meet this ideal.

Secondly, theory (meta)data should use vocabularies that follow FAIR principles (I2). This is essentially a call to establish standardized ontologies, which are themselves a type of theory (Paul E. Meehl, 1990). Thirdly, theory (meta)data should include qualified references to other (meta)data, including previous versions of the theory (I3). The first part of this principle allows for nested theories; for example, a theory that specifies causal relationships between constructs could refer back to an ontological theory from which those constructs are derived. This can be achieved by linking the DOI of those nested theories (“Contributing Citations and References,” n.d.). The second part of this principle allows for tracing the provenance of a theory; keeping track of its prior versions and other theories that inspired it. This can be achieved by using Git for version control and Zenodo for archiving. Git tracks the internal provenance of a theory repository; Zenodo is used to archive relations to external references (e.g., papers that influenced the theory, previous theories that inspired it, models based upon the theory).

As the original definition of interoperability was somewhat narrow (M. D. Wilkinson et al., 2016), the concept has recently been further refined in terms of facilitating “successful communication between machines and between humans and machines”, where “A and B are considered X-interoperable if a common operation X exists that can be applied to both”

Table 2

Property	X-interoperability
1) Ontology	Variable selection
2) Causal connections	Model specification, covariate selection, causal inference
3-8) Functional Form	Deriving specific hypotheses
9) Numerical Value	Simulating data

(Vogt et al., 2024). This definition invites the question: *interoperable for what?* Suitable answers for FAIR theory may be: this theory is X-interoperable for deriving testable hypotheses, or for the purpose of selecting relevant control variables, or for the purpose of indicating the conditions necessary for observing a particular phenomenon. This revised definition implies that theories have specific properties that incur affordances in terms of X-interoperability; for example, Table 2 illustrates the affordances of Meehl’s nine properties of strong theories (properties 3-8 are grouped because they all refer to functional form).

With regard to the state of interoperability in contemporary psychology, Kurt Lewin’s adage “there’s nothing as practical as a good theory” (Lewin, 1943) implies that ought to be highly X-interoperable in psychological researchers’ day-to-day work. But, as we argued, this is not the case. The examples of X-interoperability offered in Table 2 illustrate that much can be gained by integrating theory directly into analysis workflows, and by making theory X-interoperable within software used for analysis. For example, interoperable theory could be used to select control variables for causal inference (Cinelli, Forney, & Pearl, 2022), or to preregister the inferential procedure that would lead to specific modifications of a theory after analyzing empirical data (Peikert, Van Lissa, & Brandmaier, 2021), or to derive machine-readable hypotheses (Lakens & DeBruine, 2021) which could be automatically evaluated through integration testing (Van Lissa, 2023). Furthermore, theories can be X-interoperable with each other to enable nesting, or using one theory to clarify elements of

another theory. For example, it should be possible to embed a theory about emotion regulation (e.g., Gross, 2015) within a theory of emotion regulation development (Morris, Silk, Steinberg, Myers, & Robinson, 2007).

Reusability

If take cumulative knowledge acquisition to be a goal of scientific research, then Reusability is the ultimate purpose of making theory FAIR. Applied to FAIR theory, reusability requires that theory and its associated metadata are richly described with a plurality of accurate and relevant attributes (R1) with a clear and accessible license for reuse (R1.1), detailed provenance (R1.2), and (meta)data which meets domain-relevant community standards (R1.3). As we will argue below, the most appropriate license for theory reuse is likely to be CC0 (no rights reserved), although this should be combined with a culture of comprehensive (theory) citation to meet other open science requirements [REF TOP guidelines]. Criterion R1.2 is met by version control with Git and archival on Zenodo. Domain-relevant community standards, to a large extent, remain to be established - and this paper is the first step towards further work in that area.

If we consider the current state of Reusability in psychological theory, there appears to be a norm *against* theory reuse: “[Theories are] like toothbrushes — no self-respecting person wants to use anyone else’s” (Mischel, 2008). This norm impedes scientific progress. Cumulative knowledge acquisition requires reusable theories that are continuously updated based on insights from new data (de Groot, 1961). In our workshops on FAIR theory, we similarly notice reluctance to the notion of reusing and adapting theories, reflected in questions such as “who owns theory”, and “who determines how a theory may be reused or changed”? These questions imply a norm against modifying theory without consent from the author reminiscent of the aforementioned problem of dependency on the author.

Licensing theories for reuse provides an unambiguous answer to such questions. In

determining what license is appropriate for theory, it is important to consider that copyright law limits authors' rights based on the *idea-expression dichotomy* (Bently, Davis, & Ginsburg, 2010), which holds that copyright explicitly does not “*extend to any idea, procedure, process, system, method of operation, concept, principle, or discovery*”. Copyright may, however, extend to creative works expressing that idea (e.g., writing, visual illustrations). It thus seems that vague, ambiguous verbal explanations of theories - in other words, those that fall short of the Accessibility criterion - are more likely to qualify for copyright protection than formal theories. If copyright limits Reusability and does not cover ideas in their purest form (like formal theories), then it might be counterproductive and possibly misleading to adopt a license that assumes copyright protection. Furthermore, even if copyright would apply, academic research is covered under “fair use” exemptions, so copyright would pose few restrictions to Reusability in scholarly communication. Given these considerations, the CC0 (no rights reserved) license seems most appropriate for FAIR theory; it explicitly waives all rights and encourages reuse. In principle, CC0 does not require attribution. Nevertheless, is essential that scholars do comprehensively cite theory, including prior work that new theories are based on, even in absence of legal obligations to do so, to meet the definition of Reusability (R1.2, Table 1 and to comply with other definitions of open scholarship (Aalbersberg et al., 2018).

Additional considerations

We can take inspiration from the field of computer science for well-established processes for iteratively improving information artefacts, like computer code (which we also have successfully applied in the domain of reproducible research findings, see XXX). Using version control systems, like git, would enhance the reusability of FAIR theory by thoroughly documenting every modification in a traceable and reversible manner. Git also facilitates diffuse and adversarial collaboration, as independent researchers can create independent versions of existing theories through “forking”, or suggest modifications to existing theories

via “pull requests”. In sum, version control using Git enables systematic, collaborative, and transparent theory development, enables studying the provenance of a theory and investigating how well different iterations of the theory explain empirical evidence (Van Lissa, 2023).

Even if scholars wish to diverge substantially from prior theory, explicitly referring back to it enables clear comparison of the differences (Ram, 2013).

From a meta-science perspective, FAIR theory facilitates studying the state of theory in a particular subfield, and comparing theories’ substantive and structural properties.

Making Theories FAIR Accelerates Scientific Progress. Adopting the FAIR principles for theories can address key challenges in the current research landscape, where theories often remain isolated and underutilized. By making theories findable, accessible, interoperable, and reusable, researchers can ensure that their work is grounded in a shared, transparent, and cumulative body of knowledge. This approach enhances scholarly communication, allowing for greater scrutiny, replication, and collaboration across disciplines, ultimately leading to faster, more reliable, and more impactful scientific progress.

Making a Theory FAIR

Open science infrastructure is an area of active development, and as such, the approach proposed here should not be considered definitive, but rather, as one proposal for a FAIR-compliant implementation of theory. At the time of writing (2024), the integration of GitHub and Zenodo makes for a particularly user-friendly approach. Nevertheless, it is important to stress that alternatives to GitHub (Gitlab, Bitbucket, etc.) and Zenodo (e.g., institutional repositories) exist. The principles described here (using version control and archiving major versions) could be implemented in a different workflow.

The process described here can be largely automated in R using the `theorytools` package; see the package vignette on FAIR theory, `vignette("fairtheory", package =`

"theorytools").

1. Implementing the Theory

Given that we structured our argument around the importance of FAIR theory for cumulative knowledge production through scientific research around the *empirical cycle*, we decided to use it as an example for this tutorial. Note that, while the empirical cycle is not explicitly referred to as a “theory” by De Groot, he derives it from “a theory of thinking” by Selz (REF). We can thus consider it a meta-theory of the process of theory construction. The empirical cycle is described on page 28 of De Groot and Spiekerman (1969):

Phase 1: ‘Observation’: collection and grouping of empirical materials;
(tentative) formation of hypotheses.

Phase 2: ‘Induction’: formulation of hypotheses.

Phase 3: ‘Deduction’: derivation of specific consequences from the hypotheses, in the form of testable predictions.

Phase 4: ‘Testing’: of the hypotheses against new empirical materials, by way of checking whether or not the predictions are fulfilled.

Phase 5: ‘Evaluation’: of the outcome of the testing procedure with respect to the hypotheses or theories stated, as well as with a view to subsequent, continued or related, investigations.

If we compare it to the levels of theory formalization (Guest & Martin, 2021), De Groot’s theory is either at the “theory” or “specification” level. It consists of a series of natural language statements. We can increase the level of formalization, and present an “implementation” in the human- and machine-readable DOT language:

digraph {


```
449  observation;
450  induction;
451  deduction;
452  test;
453  evaluation;
454
455  observation -> induction;
456  induction -> deduction;
457  deduction -> test;
458  test -> evaluation;
459  evaluation -> observation;
460
461 }
```

462 This language describes the model as a directed graph. Note that the code has been
463 organized so that the first half describes an ontology of the entities the theory postulates,
464 and the second half describes their proposed interrelations. This follows the first two
465 properties of good theory according to Meehl (Paul E. Meehl, 1990).

466 We can now write this implementation of the empirical cycle to a text file, say
467 `empirical_cycle.dot`.

468 2. Creating a Project Folder

469 Create a new folder and copy the theory file from the previous step into it. To help
470 meet the Interoperability and Reusability criteria, add two more files: A README.md file
471 with instructions for future users of your theory, and a LICENSE file with the legal
472 conditions for reuse. We recommend the CC0 license, but other options are available, see
473 <https://choosealicense.com>.

What’s in a README?. The readme should contain information to help people get started with using your FAIR theory. We suggest the following elements:

- Title, prefaced with **# FAIR theory: The Theory's Name**
- Description: A plain-text description of the theory and its scope
- Interoperability: Most README files contain a section labeled “Getting Started”, “Instructions”, or “How to Use”. From a FAIR perspective, such a section might be better labeled “Interoperability”, or “How to Use (Interoperability)”. We propose explicitly addressing the theory’s X-interoperability, telling users exactly what they can use the theory for, and how. For example, our example is implemented in the DOT language for describing graphs, so we would could provide instructions here on how to plot a DOT graph.
- Contributing: Pertaining to the Reusability criterion, this section should tell users the *social expectations regarding reuse and contributions*.
- License: The legal complement to the preceding section, this section should refer readers to the LICENSE file to learn about the *legal conditions of reuse*.
- Citing this work: Tell users how to cite the theory. Note that this section is redundant with the Zenodo archive, which has a preferred citation field. The disadvantage of redundant information is that you may have to maintain this section of the README going forward. The advantage is that documenting related works in the README makes it more readily accessible to users. We suggest a compromise: to retain this section, but refer the reader to the Zenodo page.
- Related works: This section should refer to the work that the FAIR theory is derived from, or documented in. Again, this is redundant with metadata entered in Zenodo (step 5). We nevertheless recommend using this section to refer to Zenodo, and/or to document one canonical reference for the theory that is unlikely to change going forward. For example, we referenced the original empirical cycle paper here:

This repository contains an implementation of the "empirical cycle",
a model proposed by De Groot and Spiekerman (1969, p. 28). See Zenodo for other related
> De Groot, A. D., & Spiekerman, J. A. A. (1969). Methodology:
Foundations of inference and research in the behavioral sciences.
De Gruyter Mouton. <https://doi.org/10.1515/9783112313121>

3. Version Control the Repository

The field of computer science provides well-established processes for creating
information artefacts that can be iteratively improved. In particular, the practice of version
control offers extensive benefits for scientific work Van Lissa et al. (2021). To version control
our project, we initiate a Git repository in the project folder. We subsequently create a
remote repository to host a copy of this local Git repository on GitHub, which will in turn
be archived. Note that the repository must be set to “Public” to take advantage of GitHub’s
Zenodo integration.

Push the local files to the Git remote repository, and keep them synchronized going
forward.

4. Archive the Theory on Zenodo

The process of archiving a GitHub repository on Zenodo is documented in a vignette in
the `theorytools` R-package, so that it can be kept up-to-date. We present a brief summary
of the instructions at the time of writing here. First, create a Zenodo account with your
existing GitHub account. Then in Zenodo, go to the GitHub section under your account.
Following the instructions on the page, activate Zenodo for your theory repository. Then,
create a new release of the GitHub repository. Choose a tag and release title using our
adapted semantic versioning, starting with version 1.0.0, if you intend to share your theory

with the broader scientific community. After publishing the release, you should be able to see the archived version in your Zenodo account, along with a DOI.

5. Entering Meta-Data

By default, Zenodo assumes that GitHub repositories contain software and documents them as such. To document our archive as a FAIR theory requires adding some extra information on Zenodo. Supplying the following information helps improve the Findability of a theory:

- Set the *resource type* to **Model**
- Verify that the *title* is prefaced with **FAIR theory**:
- Add the *keyword* **fairtheory**
- Optionally, submit the theory to the “FAIR Theory Community” to increase its findability
- List the DOIs/identifiers of *related works*. Use the **Relation** field as appropriate. For example:
 - **Is documented by** can be used to reference a theory paper you wrote, in which you introduce this FAIR theory
 - **Is derived from** could be used to reference a paper or book chapter that introduced an existing theory that was not previously made FAIR. We used **Is derived from** to reference De Groot and Spiekerman’s empirical cycle.

6. Making Changes

Aside from technical solutions to However, “[i]n systems with many dependencies, releasing new [...] versions can quickly become a nightmare.” The tension lies in balancing regular versioned updates with maintaining interconnected theories: changes risk breaking links between theories, but assuming every change does so leads to an overly sparse network.

In software development this problem is colloquially called “dependency hell” and usually addressed via semantic versioning. Traditional version control only tracks changes but semantic versioning tracks what those changes mean to other users of the theory by using a specific format of version numbers. We propose the following adaptation of semantic versioning for theories:

Given a version number MAJOR.MINOR.PATCH, increment the:

MAJOR version when you make incompatible changes, i.e., change the meaning or derived empirical statements
MINOR version when you expand the set of empirical statements in a backward compatible way
PATCH version when you make backward compatible bug fixes, or clarifications

Semantic versioning and clearly documented interdependencies allow for tracing the provenance of a theory; keeping track of its prior versions and other theories that inspired it.

Automating these Steps

R-users can use the `theorytools` package to partly automate the preceding steps, for example, using following code (see the package documentation for more information):

```
install.packages("theorytools")
library(theorytools)
# Use worcs to check if GitHub permissions are set:
library(worcs)
check_git()
check_github()
# Create the theory repository:
fair_theory(path = "c:/theoryfolder/empirical_cycle",
            title = "The Empirical Cycle",
```

```

theory_file = "empirical_cycle.dot",
remote_repo = "empirical_cycle",
add_license = "cc0")

```

Note that this function also automatically provides basic FAIR theory metadata to Zenodo.

Forking Different Implementations of a Theory

De Groot’s empirical cycle has inspired several authors, but not all of them have interpreted his work the same. For example, Wagenmakers and colleagues ([wagenmakersCreativityVerificationCyclePsychological2018?](#)) write “*De Groot’s “empirical cycle,” shown here in Figure 6*” - but Figure 6 diverges substantially from De Groot’s description, and from our implementation of it. An important advantage of FAIR theory is that we can implement different versions of a theory, compare them, and document their cross-relationships. We can take work that has been done before - in this case, the repository created above, and create an independent copy that we can modify as we wish, while retaining cross-references to the original. This is achieved by “forking the repository”, “cloning” the forked repository to our local computer, making any changes we want, and then completing steps 4-5 of “Making a Theory FAIR”.

We have implemented Wagenmakers and colleagues’ version as a DOT graph to illustrate some clear deviations from the original. First, the phases of the cycle have been renamed. While this change was not described in the paper, we assumed that the labels are meant to illustrate the phases, not substantially change the ontology. We represent this change by adding labels to the original DOT graph. Note, however, that the labels suggest a focus on empirical psychology that was absent in the original formulation, which was more general. Furthermore, the label “knowledge accumulation” invites the question of exactly *how* knowledge accumulates upon evaluation of a prior experiment. As this lack of

cumulative knowledge acquisition appears to be precisely where contemporary research practice falls short, this ambiguity invites further improvement of the theory. Second, the authors mention an explicit change: *“We added the Whewell-Peirce-Reichenbach distinction between the context of discovery and the context of justification”*. The DOT graph below shows our implementation of this version of the empirical cycle, by adding subgraphs.

```

digraph {
    subgraph cluster_discovery {
        label="Discovery";
        induction [label="New hypothesis"];
        deduction [label="New prediction"];
    }
    observation [label="Old knowledge and old data"];
    subgraph cluster_justification {
        label="Justification";
        test [label="Test on new data"];
        evaluation;
    }
    observation -> induction [label="Speculate & explore"];
    induction -> deduction [label="Deduce"];
    deduction -> test [label="Design new experiment"];
    test -> evaluation [label="Statistical analysis"];
    evaluation -> observation [label="Knowledge accumulation"];
}

```

The first author was inspired by De Groot too, but they conceive of the empirical cycle in yet another way. First, notice that the nodes in De Groot's formulation mostly refer to processes. This invites the question of what the deliverables are in each phase, or in other words: what actually changes when going through the cycle, except the scholar's mind. In our implementation below, we account for this difference by having the nodes refer to specific deliverables; the edges now refer to processes. Second, De Groot's strict distinction between processes of observation, induction, and deduction is not widely supported by philosophy of science. For example, many have argued that observation is value-laden, and as such, involves induction. The derivation of hypotheses from theory is also not purely deductive, as auxiliary assumptions are often made (which are, again, an inductive process). Furthermore, if the testing procedure is not explicitly defined before seeing the data, it incurs some inductive bias as well [REF Peikert]. With these alterations, we implement the empirical cycle as follows:

```
digraph {
    theory;
    prediction;
    data;
    test;
    results;

    theory -> prediction [label="deduction"];
    prediction -> test [label = "implement inferential procedure"];
    data -> results;
    test -> results [label = "apply to data"];
    results -> theory [label="interpretation and generalization"];
```


650 }

651 Using FAIR theory to Perform Causal Inference

652 Some have argued that *causal explanations* are a property of good theory [REF Meehl,
653 etc?]. According to Pearl and colleagues, explicit assumptions about the direction of
654 causality allow one to perform causal inference even on cross-sectional data. Any formal
655 theory that is explicit about direction of causality could thus be used to guide causal
656 inference, and could even be integrated into the analysis environment.

657 In this example, we illustrate how to use DAGs for causal inference, including the
658 detection of a violation of the initial model and subsequent adaptation of the DAG. We
659 could use that to illustrate updating FAIR theory:

660 <https://currentprotocols.onlinelibrary.wiley.com/doi/full/10.1002/cpz1.45>

661 We can find more examples of causal inference with DAGs in these tutorials:

662 <https://www.r-bloggers.com/2019/08/causal-inference-with-dags-in-r/>

663 <https://www.r-bloggers.com/2018/08/applications-of-dags-in-causal-inference/>

664 Discussion

665 The replication crisis has brought the inadequacies of contemporary theoretical
666 practices in the social sciences into focus. Psychological theories often fall short of all FAIR
667 principles: they are hard to find and access, have limited interoperability, and are rarely
668 reused. These limitations impede cumulative knowledge production in our field, leading to
669 an accumulation of “one-shot” empirical findings, without commensurate advancement in our
670 principled understanding of psychological phenomena. We argued that applying the FAIR
671 principles to theory offers a structured solution to these shortcomings. We demonstrated how
672 to create, version-control, and archive theories as digital information artifacts. We

introduced the `theorytools` R-package to partly automate these processes, reducing barrier of entry for researchers, and creating a FAIR resource for theory construction tools and documentation that can be continuously updated as best practices develop further.

Making theory FAIR allows researchers to more easily find a relevant framework; access and understand it; interact with it in a very practical manner, for example, by deriving predictions from it, or using it to select control variables; and reuse it, contributing changes to existing theories or splitting of in a new direction. Whereas the idea of theory can be quite nebulous to empirical social scientists, FAIR theory makes theoretical work practical and tangible, incorporating theory into scholars' workflows. Having a concrete object to iterate upon facilitates the systematic improvement and iterative refinement of psychological theories, thus substantially increasing the efficiency of research. While FAIR theory does not directly resolve the problem of strategic ambiguity, it does provide a framework within which scholars can increase the precision and formalization of theories. FAIR principles also facilitates new ways of collaboration, leveraging tools like Git for version control and Zenodo for archiving to document provenance and facilitate contributions from diverse researchers.

Strengths

One important strength of FAIR theory is that it provides much-needed open science methods for the underemphasized inductive phase of the empirical cycle. Most extant open science methods focus on increased rigor in testing, but provide little guidance as to what to do with the newly collected empirical evidence. By providing much-needed open science methods for theory construction, FAIR theory helps restore the balance between inductive and deductive research and contributes to closing the “open empirical cycle” [REF Hoijtink].

Our approach aligns closely with contemporary developments in open science, such as modular publishing, interdisciplinarity, meta-research, and team science. The advantage of modular publishing is that authors can be credited for theory development. Given the

current emphasis on empirical papers [REF], theoretical papers can be hard to publish. FAIR theories, by contrast, can be readily disseminated as citable information artifacts, thus changing the incentive structure to favor theory development.

Interdisciplinarity benefits from FAIR theory’s accessibility across different fields; thus, theoretical frameworks can be reused, adapted, or used for analogical modeling [REF Oisin paper]. Meta-research benefit from the fact that FAIR theory enables studying the structure, content, and development of theories over time. In terms of team science, FAIR theory facilitates collaboration by ensuring that all contributors have access to the same information and clarifying any remaining areas of contention or misunderstanding. Version control provides a framework to resolve parallel developments from multiple collaborators in a non-destructive manner. This facilitates collaboration across geographical boundaries, and adversarial collaboration, where others strive to falsify a theory or identify its inconsistencies, and democratizes collaboration with as-of-yet unknown collaborators via platforms like GitHub, where researchers outside one’s network can identify issues or suggest improvements to theories.

Finally, FAIR theory plays an important role in science communication, because theory synthesizes contemporary scientific understanding about a phenomenon. Theory bridges the gap between academic research and practitioners by summarizing actionable insights, relieving practitioners from the need to sift through extensive empirical literature. By providing a mechanism for iterative improvement based on emerging evidence, FAIR theory also supports effective evidence-based decision making.

Limitations

One important limitation of the present work is that, while we build on well-established information architecture like Zenodo, it is unlikely that the proposed metadata standard is definitive. Community adoption can reveal areas of further

improvement. Furthermore, at the time of writing, dedicated indexing systems for FAIR theory are non-existent. Using the Zenodo search function and submitting theories to the “FAIR Theory Community” on Zenodo can help overcome this limitation in the short term.

Another limitation is the learning curve associated with tools like Git and Zenodo. The `theorytools` R-package mitigates this limitation by automating key steps in the process. Moreover, the initial investment in time can be offset by long-term productivity gains and increased impact of FAIR theory. One barrier to adoption of FAIR theory is cultural resistance to sharing and modifying theories, also known as the “toothbrush problem”. Education might help address this limitation; with this in mind, we are developing open educational materials on theory development.

One limitation of scope is that FAIR theory does not directly resolve problems related to strategic ambiguity [REF] and lack of theory formalization [REF]. However, our work does establish a framework within which theories can be further formalized. The example of the empirical cycle demonstrates how FAIR principles can guide theory formalization and foster cumulative progress. Another limitation of scope is that FAIR theory does not resolve other related issues in social sciences, such as the measurement crisis [REF] and lack of standardized ontologies for psychological constructs [REF]. However, our work here provides a template for addressing such challenges, and any advancements in the areas of measurement and ontology will serve to amplify the value of FAIR theories, particularly when such resources are cross-referenced in the metadata (e.g., on Zenodo).

Future Directions

One remaining issue that intersects with FAIR theory is the measurement and operationalization of psychological constructs. Aside from the aforementioned “theory crisis”, there has been talk of a “measurement crisis”: it is not always clear how theoretical constructs are operationalized, and many existing instruments have poor psychometric

properties [REF]. Additionally, the “jingle-jangle” fallacy is prevalent in the social sciences: the same term is often used for distinct constructs, and conversely, different terms are used to refer to the same construct. FAIR theory can help address the measurement crisis: since theories can reference other theories and resources, it is possible to extend a structural theory with a theory of

FAIR theory incorporates theory into open science workflows, facilitates scholarly communication about theories, making it easier to share theories with less opportunity for ambiguity and misunderstanding. FAIR Theories are easier to find, and facilitate sharing, reusing, and updating open theories. More efficient and transparent communication about theory democratizes and accelerates cumulative knowledge acquisition, removes barriers for knowledge exchange with the global scholarly community, opens theory development to diverse perspectives, and enables (distributed and adversarial) collaboration.

Conclusion

FAIR theory is a major step forwards towards more transparent, collaborative, and efficient theory construction. It provides much-needed open science methods for the inductive phase of the empirical cycle, closing a critical gap in the scientific process. FAIR theory makes theory more tangible, enabling scholars to incorporate it in their day-to-day workflows in order to derive hypotheses, select control variables, and contribute new data-driven insights. This paves the way for more theory-driven scholarship, and accelerates cumulative knowledge acquisition in the social sciences and beyond.

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