



Structure-Behavior-Action Framework for Coherent Scientific Enterprise^{*}

Daniel Kristanto, Adam Craig, Carl Taswell[†]

Abstract

The scientific community aims to produce cumulative knowledge. However, systemic inefficiencies, such as the fragmentation of research findings and the replication crisis, often hinder this goal. Here, we examine the Structure-Behavior-Action (SBA) framework, a paradigm evolving from engineering systems design, to diagnose these challenges. While the engineering-based Structure-Behavior-Function (SBF) model describes deterministic machines, the SBA framework is uniquely suited to social systems because it accounts for cognitive agents. The framework posits that a system's structure constrains the behavior of its actors, which determines the system's outcomes. We argue that the current scientific enterprise suffers from flaws in two such structures: disjointed knowledge management systems and misaligned incentives that prioritize quantity over quality. For example, fragmented literature prevents comprehensive review, while pressure to publish encourages the production of substandard reports. To address these issues, we advocate for restructuring the scientific ecosystem. We discuss the use of decentralized information structures, such as the Nexus-PORTAL-DOORS-Scribe (NPDS) Cyberinfrastructure, to create accessible community knowledge spaces. Furthermore, we supplement traditional bibliometrics with Fair Acknowledgment of Information Record (FAIR) Metrics, which quantitatively measure the quality and reproducibility of individual reports. By redesigning these foundational structures, we can foster a more cooperative environment that ensures the cumulative advancement of science.

Keyphrases

Research ethics, scientific integrity, Structure-Behavior-Action Framework, NPDS Cyberinfrastructure, FAIR Metrics.

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The Challenge: Systemic Barriers to Scientific Enterprise

The scientific enterprise aims to produce cumulative knowledge that advances our understanding of the world and its complex phenomena. However, this goal is hindered by significant systemic barriers. These include the inherent complexity of the systems under study, particularly in the life and social sciences (Gernigon et al. 2024); practical and ethical barriers to experimentation (Rosenthal 1994); the ongoing crises of replication and reproducibility that have cast doubt on the reliability of findings across disciplines (Obels et al. 2020; Open Science Collaboration 2015; Baker 2016); and a systemic issue of research fragmentation (Muthukrishna and Henrich 2019; Gates et al. 2025; Ballezzi et al. 2015).

While all these factors are significant, we focus here on the twin challenges of research fragmentation and the reproducibility and replication crises. We argue that these are not independent failures, but coupled emergent behaviors of the scientific community. By research fragmentation, we mean the disjointed state of knowledge production where research communities become isolated silos. This phenomenon has been characterized structurally as the formation of citation clusters that limit the global diffusion of ideas (Gates et al. 2025), bibliometrically as a disconnection that hinders productivity (Ballezzi et al. 2015), and theoretically as a lack of integrative frameworks (Muthukrishna and Henrich 2019). Meanwhile, we distinguish the replication crisis from computational reproducibility (re-analyzing existing data to obtain the same results (Goodman et al. 2016; Taswell 1998)). Here, we refer to replicability as the widespread inability to obtain consistent results across independent studies using new data. Together, these behaviors lead to the undesirable outcome of a vast but disconnected literature where a proliferation of methods, inconsistent terminology, and a lack of theoretical integration make findings difficult to compare, synthesize, or cumulatively build upon.

In this perspective, we review these challenges of fragmentation and poor reproducibility and replicability using the Structure-Behavior-Action (SBA) framework to diagnose its structural origins and discuss appropriate interventions.

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The Framework: Structure-Behavior-Action

The paradigm of SBA evolves from the Structure-Behavior-Function (SBF) model, originally introduced as a formal model for engineering design processes (Gero 1990). In this context, designing a system involves understanding its components (*Structure*), how those components interact physically (*Behavior*), and their assigned purpose (*Function*). For simple artifacts lacking agency, this model is sufficient. For instance, in automotive engineering, specific structural variables (e.g., insulation thickness) dictate the thermal behavior of the system (e.g., heat flux), which serves the fixed *Function* of providing thermal comfort (Kristanto and Leephakpreeda 2018). Here, the system has no capacity for decision-making; it simply executes the function it was designed to perform.

However, living organisms differ fundamentally from machines (see Table 1). They are not built for a single, static function but evolve to perform complex, adaptive operations necessary for life. Consequently, when applying this framework to biological contexts, we replace the passive concept of *Function* with the active concept of *Action*. This shift acknowledges that living systems possess varying degrees of agency. For example, in general biology, structural entities such as cells and organs support physiological behaviors that enable the *Action* of survival. Importantly, the precise evolutionary threshold where *function* (automatic process) transforms into *action* (agentic goal) remains an open question in the philosophy of biology.

As we move up the level of complexity to cognitive systems, this distinction becomes sharper. In human neuroimaging, for example, the *Structure* of the brain (e.g., physical white matter tracts) constrains the *Behavior* of the brain (e.g., functional connectivity between regions). Importantly, the coupling between this physical structure and dynamic behavior of the brain enables the *Action* of the individual, manifesting as measurable differences in behavioral performance on cognitive tasks (see Figure 1; adapted from Fotiadis et al. (2024)). Here, *Action* is no longer just biological survival; it is the execution of specific cognitive demands based on neural constraints.

Finally, we apply this framework to the highest level of complexity: human society. The multifaceted nature of society requires distinguishing between its material and cultural dimensions, as outlined in Table 1. In the material dimension, technological *Structures* (e.g., transportation infrastructure) constrain physical *Behaviors* (e.g., the movement of people and goods), which supports the *Action* of allowing communities to function logistically. In the cultural dimension, the *Structure* consists of incentive rules (e.g., rewards and punishments) that shape culturally transmitted *Behaviors* (e.g., traditions and fads), determining collective *Actions* (e.g., peaceful coexistence or conflict). This dynamic is further examined by Raza (2024), who investigates how social institutions, norms, and hierarchies (*Structure*) influence individual interactions (*Behavior*) and identities (*Action*).

Given the proven utility of the SBA framework across these diverse fields, we examine the scientific enterprise itself through this lens to address its current systemic crises of research fragmentation and poor reproducibility and replicability.

For instance, in automotive engineering, designing a *Structure* (e.g., the cabin body, insulation, heating ventilation air-conditioning system) that has the *Behavior* of moving and heating or cooling air serves the *Function* of providing thermal comfort inside a car cabin (Kristanto and Leephakpreeda 2018). Engineers manipulate structural variables (e.g., glazing thickness, insulator R-value indicating its ability to resist heat

flow) to influence the thermal *Behavior* of the system (e.g., heat flux, internal air temperature). Importantly, the environment (e.g., external solar load, ambient temperature) also influences the *Behavior* of the system, demonstrating that the same *Structure* can yield different *Behaviors* under different conditions.

However, a living organism does not consist of systems engineered to support a single function but rather of many levels of structure that co-evolved to perform all actions necessary for life. Therefore, in biological contexts, we replace *Function* with *Action*. For example, in a human neuroimaging study, *Structural* properties of the brain (such as thickness, myelination, and physical white matter tracts) supports the *Behavior* of the brain (such as functional activation of regions and functional connections between them), which enables the *Action* of the individual. See Fotiadis et al. (2024) for a review.

Moreover, the Structure-Behavior-Action (SBA) framework applies equally to society as a system, but the multifaceted nature of human interaction requires a further distinction among different dimensions. In the technological dimension, for example, the material *Structure* of society (e.g., transportation infrastructure) constrains physical *Behavior* (e.g., movement of people and goods), which supports the *Action* of allowing people to live, work, and shop within their communities. In the cultural dimension, the *Structure* consists of incentive rules (e.g., rewards and punishments) that shape culturally transmitted *Behaviors* (e.g., traditions and fads), determining collective *Actions* (e.g., peaceful coexistence). This dynamic is further examined by Raza (2024), who investigates how social institutions, norms, and hierarchies (*Structure*) influence individual interactions (*Behavior*) and identities (*Action*).

Given the proven utility of the SBA framework in different research fields, we propose applying it to study the scientific enterprise itself to address the systemic problems of research fragmentation and the replication crisis.

Diagnosing the Scientific Enterprise

Here, we use the SBA framework to trace the twin challenges of research fragmentation and the crises of reproducibility and replicability to their sources. First, we identify that the primary goal of the scientific enterprise is to produce cumulative knowledge that advances our understanding of the world and its complex phenomena. To achieve this goal, a specific *Action* is necessary: Research must function as a cumulative effort where actors, the researchers, work together to develop findings that build directly upon one another.

However, the current *Structure* of the scientific enterprise fails to elicit the *Behavior* necessary to support this desired *Action*. Instead, existing structural constraints drive researchers toward *Behaviors* that inhibit cumulative progress, thereby preventing the system from achieving its collective goals. This systemic dysfunction arises from specific flaws in two foundational structures: the technological organization of knowledge and the cultural incentives for researchers.

The first structural flaw lies in how information technology supports the organization of scientific knowledge. Currently, the scholarly publishing system scatters records of past research across for-profit journals, divides them by discipline, and often hides them behind paywalls (Day et al. 2020). Moreover, this system also lacks the structural mechanisms to systematically associate or compare findings across different studies. This technological fragmentation constrains researcher behavior, causing individuals to pick out and skim only a few articles before designing new projects. This inadequate engagement with the literature frequently leads to the unintentional duplication of old studies and

150 a failure to adopt the best available methods, reinforcing the silos of
151 research fragmentation.

152 The impact of this technological structure on research conduct is
153 profound. Consider the specific mechanism of literature retrieval: the
154 current infrastructure generally presents knowledge as a list of isolated
155 documents returned by keyword searches. When a search for a specific
156 methodology returns thousands of results, the structure imposes an
157 overwhelming information load. Faced with this challenge, researchers
158 are forced to adapt their behavior by using heuristics to filter the studies.
159 The most common cognitive adaptation is the habit of sorting studies
160 by citation count, journal prestige, or familiar author names rather than
161 methodological relevance. This structural constraint directly creates a
162 "rich-get-richer" dynamic, where established but potentially less relevant
163 methods are repeatedly selected and reinforced simply because
164 they are visible, while innovative but less indexed solutions remain
165 ignored.

166 The second structural flaw is the system of incentives, which rewards
167 researchers primarily for the quantity of publications and citations
168 rather than the quality of their work (Edwards and Roy 2017). These
169 misaligned incentives drive undesirable behaviors, where researchers
170 prioritize self-promotion and "salami slicing" results over rigorous inquiry.
171 This behavioral adaptation is a rational strategic response to the
172 structural requirement for novelty. In the current economy of prestige,
173 the structure is defined by publication venues that prioritize novel and
174 positive results over robust verifications. Consequently, a researcher
175 who dedicates resources to replicating a foundational study faces a
176 losing proposition: if the replication is successful, it may be deemed
177 not novel and rejected; if it fails, it may invite conflict. The rational
178 behavioral response to this structure is the "file-drawer effect," where
179 researchers systematically suppress negative or ambiguous results in
180 favor of p-hacking or parameter tweaking to achieve the statistically
181 significant and novel findings required for career advancement. The
182 aggregate action of these individual behaviors is a scientific record
183 populated by unverified, fragile effects that cannot support cumulative
184 discovery.

185 Moreover, this results in an avalanche of substandard papers and
186 a "natural selection of bad science," characterized by increased false
187 discovery rates (Smaldino and McElreath 2016). This systemic generation
188 of unreliable findings acts as a primary driver of the crises of
189 reproducibility and replicability. Consequently, funding allocation often
190 favors high-visibility metrics over rigorous methodology, creating an
191 environment that can inadvertently pressure researchers to compromise
192 their integrity.

193 The Solution: Restructuring the Scientific Enterprise

194 To address the systemic dysfunction diagnosed through the SBA
195 framework, we advocate for a fundamental restructuring of the scientific
196 enterprise, beginning with the information systems used to organize
197 knowledge.

198 We suggest the shift from static, isolated records to decentralized
199 infrastructures, for example the Nexus-PORTAL-DOORS-Scribe (NPDS)
200 Cyberinfrastructure (Taswell 2007; Taswell 2010). This technology
201 creates accessible community knowledge spaces that separate resource
202 registration from publishing, ensuring data remains robust and discoverable.
203

204 Second, we suggest supplementing the incentives used to evaluate

205 and reward scientific contributions. We advocate moving beyond simple
206 citation counts to measures that prioritize quality, such as the Fair
207 Attribution of Indexed Reports, or Fair Acknowledgment of Information
208 Records, (FAIR) Metrics (Craig, Athreya, et al. 2023). Unlike traditional
209 metrics, this system goes deeper by quantitatively analyzing the reproducibility
210 of a report, rigorously distinguishing between correctly attributed factual
211 claims and those that are misattributed.

212 This incentive structure extends to the evaluation process itself
213 through reproducible peer review (Craig and Taswell 2024). We support
214 publishing reviews as citable, independent references cross-linked to
215 the original report. This approach elevates peer review from an invisible
216 administrative task to a recognized scientific contribution, thereby
217 incentivizing high-quality critique.

218 Finally, achieving this structural transformation requires coordinated
219 support from all agents within the system. It demands active participation
220 not only from researchers but also from universities, funding
221 bodies, and journals, along with external engagement from industry
222 and society. By aligning these stakeholders to redesign the scientific
223 endeavor, we can foster the collective action of producing cumulative
224 knowledge. For a detailed summary of how the SBA framework diagnoses
225 these systemic issues and maps out the necessary structural
226 interventions, please refer to Table 2.

227 Conclusion

228 In this perspective, we utilized the SBA framework to diagnose the
229 current state of the scientific enterprise. We argue that the persistent
230 challenges of research fragmentation and the crises of reproducibility
231 and replicability are not isolated problems but systemic issues stemming
232 from deep-seated flaws in the underlying structures of academia. To
233 remedy this, we examine specific examples of restructuring, ranging
234 from the NPDS Cyberinfrastructure for knowledge organization to FAIR
235 Metrics for incentive reform, designed to encourage strategic behaviors
236 of collaborative synthesis and rigorous methodology. Importantly,
237 achieving the desired collective action of a cumulative science requires
238 more than just technical solutions; it demands the synchronized
239 commitment of all elements within the system, including researchers,
240 universities, funders, and journals, alongside external support from society
241 and industry.

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Table 1: The evolution of the framework from Structure-Behavior-Function (SBF) in deterministic systems to Structure-Behavior-Action (SBA) in complex adaptive and cognitive systems.

System Type	Domain	Structure (Components)	Behavior (Processes)	Function → Action (Goal)
<i>Deterministic Systems (No agency)</i>				
Artifact	Engineering (SBF)	Physical components (e.g., hinges, engine parts)	Physical dynamics (e.g., rotation, heat flow)	Function: Assigned utility (e.g., thermal comfort, transportation).
<i>Adaptive Living Systems (Emergent agency)</i>				
Organism	General Biology (SBA)	Anatomical entities (e.g., cells, tissues, organs)	Physiological processes (e.g., metabolism, circulation)	Action: Life-sustaining activities (e.g., survival, reproduction).
Cognitive	Neuroscience (SBA)	Brain anatomy (e.g., white matter tracts, cortical thickness)	Functional activation (e.g., regional coactivation, connectivity)	Action: Cognitive execution (e.g., perceptual decision-making, motor control).
<i>Complex Social Systems (Collective decision-making)</i>				
Society (Material)	Sociology / Economics	Technological infrastructure (e.g., transport networks, buildings)	Physical behaviors (e.g., movement of people, supply chains)	Action: Logistic outcomes (e.g., distribution of goods, economic growth).
Society (Cultural)	Sociology / Anthropology	Incentive structures (e.g., rewards, laws, social norms)	Cultural practices (e.g., traditions, fads, compliance)	Action: Social outcomes (e.g., protecting vulnerable groups, peaceful coexistence).

Table 2: Applying the Structure-Behavior-Action (SBA) framework to the scientific enterprise: Diagnosing systemic barriers and proposing structural solutions.

Structure (Constraints)	Behavior (Researcher Strategy)	Realized Action (Systemic Outcome)
(A) Current System: Diagnosis of Systemic Barriers		
Knowledge Organization: Records of research scattered across for-profit journals, divided by discipline, and locked behind paywalls.	Heuristic Filtering: Researchers pick out and skim only a few articles before designing new projects (inadequate literature review).	Research Fragmentation: New projects unknowingly duplicate old ones and fail to adopt the best available methods.
Incentives: Productivity measured by publication-focused quantitative metrics (e.g., publication counts, impact factors).	Strategic Compliance: Researchers prioritize self-promotion and "salami slicing" (file-drawer effect); avoidance of citing competitors.	Replication & Reproducibility Crisis: Funding flows to the media-savvy; "natural selection of bad science" leads to unreliable findings.
(B) Suggested System: Structural Interventions		
NPDS Cyberinfrastructure: Independent, decentralized community knowledge spaces separate registration from publishing.	Strategic Exploration: Researchers quickly explore the global state of knowledge, identifying gaps and standardizing methodologies.	Cumulative Advancement: Science builds effectively on prior work instead of retreading the same ground.
FAIR Metrics: Evaluation based on accuracy, reproducibility, and correct attribution of claims rather than citation counts.	Rigorous Verification: Researchers rigorously verify prior work, citing accurately and focusing on answering open questions.	Valid Science: Funding organizations can distinguish rigorous contributions and allocate resources to verifiable science.

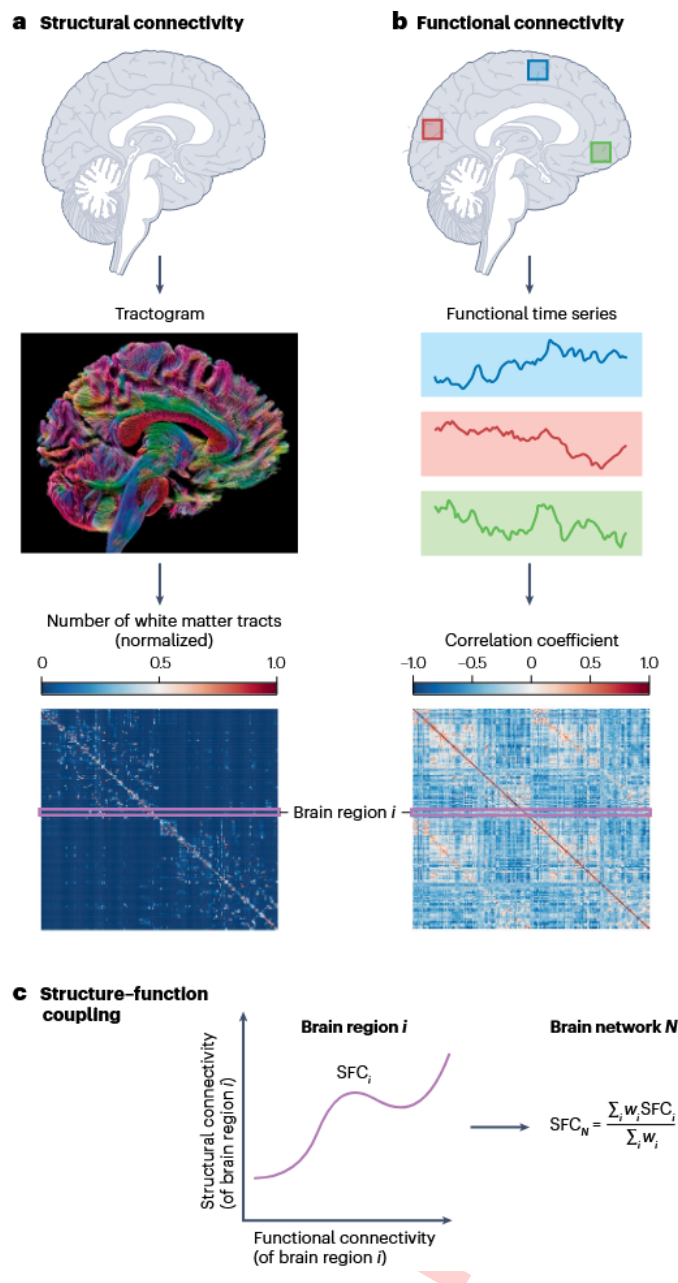


Figure 1: The Structure-Behavior-Action framework applied to neuroimaging. (a) Structural connectivity (white matter tracts) forms the physical *Structure*. (b) Functional connectivity (synchronized activity) represents the dynamic *Behavior*. (c) The coupling between structure and function predicts inter-individual differences in behavioral performance, representing the *Action*. Figure adapted from [Fotiadis et al. \(2024\)](#).