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Context-sensitive computational mechanistic explanation in cognitive neuroscience

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Mainstream cognitive neuroscience aims to build mechanistic explanations of behavior by mapping abilities described at the organismal level *via* the subpersonal level of computation onto specific brain networks. We provide an integrative review of these commitments and their mismatch with empirical research findings. Context-dependent neural tuning, neural reuse, degeneracy, plasticity, functional recovery, and the neural correlates of enculturated skills each show that there is a lack of stable mappings between organismal, computational, and neural levels of analysis. We furthermore highlight recent research suggesting that task context at the organismal level determines the dynamic parcellation of functional components at the neural level. Such instability prevents the establishment of specific computational descriptions of neural function, which remains a central goal of many brain mappers – including those who are sympathetic to the notion of many-to-many mappings between organismal and neural levels. This between-level instability presents a deep epistemological challenge and requires a reorientation of methodological and theoretical commitments within cognitive neuroscience. We demonstrate the need for change to brain mapping efforts in the face of instability if cognitive neuroscience is to maintain its central goal of constructing computational mechanistic explanations of behavior; we show that such explanations must be contextual at all levels.

KEYWORDS

cognitive neuroscience, degeneracy, functional brain mapping, levels of analysis, many-to-many mappings, mechanistic explanation, neural reuse, structure-function relationships

Introduction

A brief history of the origins of modern functional brain mapping

The historical neuropsychological literature is full of case-studies of individuals with specific behavioral impairments following damage to local regions of the cerebral cortex. Perhaps most famously, [Broca \(1861\)](#) described a patient who, after damage to the left inferior frontal gyrus (IFG), permanently lost almost all of his expressive language

abilities while his receptive abilities remained largely intact. Other patients may exhibit fluent (but incoherent) expressive language paired with receptive impairments, typically following lesions in the left superior temporal gyrus (STG). From these observations, Wernicke (1874/1969) developed one of the earliest neurocognitive models of language, which centered on the specific functions of IFG and STG and their connection *via* association fibers. Wernicke argued that complex functions such as language are the result of the interaction of multiple, simpler, sensory, motor, and association processes, thereby effectively ending the search for localized higher-level “faculties” in the brain that had dominated the study of brain-behavior relationships up to that time (Fancher, 1996; Bergeron, 2007). Since then, researchers have continued to grapple with understanding the functions of brain parts and their relationship to behavior, and one of the main goals of cognitive neuroscience remains mapping functional descriptions onto neural activity. But what does successful mapping look like?

Functional brain mapping as the analysis, decomposition, and localization of function

The goal of mainstream cognitive neuroscience¹ is to “investigate brain-behavior interactions” and to “address both descriptions of function and underlying brain events” (Journal of Cognitive Neuroscience, 2021); it seeks “an understanding how the functions of the physical brain can yield the thoughts, ideas, and beliefs” of the mind (Gazzaniga et al., 2019, p. 4). In the mainstream cognitive neuroscience research literature, the term *function* is critical to the enterprise. Importantly, function is often implicitly and interchangeably used at multiple, typically three, distinct levels of analysis (Marr, 1982/2010; Craver, 2014; Krakauer et al., 2017; Zednik, 2018; see Garson, 2016, for discussion of conceptualizations of biological function). The first level concerns that of observable behavior or its disorders. Here, cognitive neuroscientists distinguish between various complex behavioral or cognitive categories described at the personal or organismal level; that is, at the level of

the behaving human being. For example, we talk about the abilities of (or impairments of) “paying attention,” “using a tool,” or “speaking.” These are the phenomena of which the discipline ultimately seeks a *mechanistic* explanation, in terms of a description of the parts and interactions of a system that gives rise to the phenomenon (Miłkowski, 2013; Craver and Tabery, 2015; Zednik, 2019).²

To this end, at the second, subpersonal, level the phenomenon described at the first level is decomposed into components that perform specific computational operations over representations and are organized together in a specific way (Bechtel and Abrahamsen, 2010). These components are therefore identified by their functions (Piccinini and Craver, 2011), the descriptions of which tend to be “human-interpretable” (Hasson et al., 2020). For instance, computational processes are hypothesized that implement, e.g., “attentional orienting,” “action selection,” or “linguistic retrieval,” which are *subcapacities* of the organismal-level capacity of paying attention, using a tool, and speaking, respectively. In other words, and following Wernicke’s lead, at the second level multiple latent, interacting, domain-specific or domain-general, functional components are postulated that explain the production of a particular complex behavior or cognitive ability observed at the first level. In general terms, it is widely accepted that computational operations are thought to involve inputs which feed into the manipulation of internal, typically – but not necessarily (Piccinini and Scarantino, 2011) – representational states to produce outputs that are then used downstream in further computational processes (Shea, 2018).

The third level at which function is invoked is that of the brain. Thus, following the characterization of a componential computational architecture at the second level (and sometimes *before* this characterization has taken place; see Agis and Hillis, 2017), an attempt is made to map the components onto the activity of the brain, effectively reifying that architecture; that is, the researcher tries to describe the coordinated computational operations in terms of physical neural processes (Piccinini and Shagrir, 2014; Burnston, 2021). There are at least two possible characterizations of this final step. Sometimes, the assumption is that a computational operation will directly map onto a specific brain “part,” where a part is broadly construed as a cell, localized assembly of cells or a distributed

1 While cognitive neuroscience is currently dominated by cognitivist assumptions of representation, computation and mechanism (which we here call the mainstream approach), 4E (embodied, embedded, enactive, extended) as well as ecological approaches to the study of the brain are becoming increasingly prevalent. Further, recent work aims to merge key insights from both traditions (e.g., Piccinini, 2022). In the present article, we have chosen to limit our analysis of research practices as they occur in the mainstream cognitive neuroscience literature. A similar analysis of neuroscientific research practices within 4E and ecological frameworks would be a rather different enterprise and is outside the scope of the current article (for examples of work in this tradition, see Gibson, 1966; Chiel and Beer, 1997; Barrett, 2011; van Orden et al., 2012; Dotov, 2014; Anderson, 2014; Kiverstein and Miller, 2015; de Wit et al., 2017; Hutto et al., 2017; Dewhurst, 2018; Bruineberg and Rietveld, 2019; de Wit and Withagen, 2019; van der Weel et al., 2019; Ryan and Gallagher, 2020; Raja, 2021; Raja and Anderson, 2021).

2 The topic of how cognitive neuroscience has come to decide on which explananda are worth pursuing is interesting on its own, and a point we return to later, in the section “Discussion: Implications for Cognitive Neuroscience” (see Fancher, 1996; Kästner, 2017 for discussion). Further, more detailed discussion of the role of mechanism in science is beyond the scope of this manuscript (see Cartwright et al., 2020). Our argument assumes a commitment to understanding cognition mechanistically, that is, being explicable in terms of the joint action of parts. Even if one denies a specific characterization of mechanism as discussed in the philosophy of science, we think it uncontroversial that there is at least an implicit and informal understanding that cognitive neuroscience, like the cognitive psychology it inherits, is often seeking mechanistic explanation.

network of cells (Glasser et al., 2016). For instance, using lesion-symptom mapping or functional magnetic resonance imaging (fMRI), cognitive neuroscientists may search for a neural region or network of regions whose function it is to implement the computations that instantiate attentional orienting (while sometimes considering the biological constraints of the part such as, e.g., neural response latencies or receptor types; e.g., Kravitz et al., 2013), and another region or network whose function it is to implement the computations for action selection. However, sometimes the assumption is that a computational operation requires additional decomposition before it can be mapped onto the brain. For instance, using computational modeling of different types of cells or cell networks, attentional orienting is broken down further into component operations (e.g., a number of different computations in a connectionist model). Either way, the goal is to mechanistically explain organismal-level phenomena, *via* organized, interacting computational components, in terms of functionally specialized brain cells, assemblies, or networks that combine into interacting brain parts.³ Though cognitive neuroscience has moved on from structure-function mapping understood in the modular sense described earlier (e.g., expressive language is due to Broca's area), many cognitive neuroscientists – implicitly or explicitly – still seek to discover *the* brain basis of our ability to pay attention, use a tool, or speak (Cabeza and Nyberg, 2000; see Zerilli, 2019 for a historical overview of this approach).

Given this, brain mapping continues to be a major effort, which can also be gleaned from its influence on the infrastructure of the field (e.g., the UCLA Brain Mapping Center; the journal *Human Brain Mapping*). Indeed, the idea of functional brain mapping is so deeply engrained in current thinking about brain-behavior relationships that it is implicit in our nomenclature and pervades textbooks and day-to-day public discourse about neuroscience. Thus, there is talk of the primary *visual* cortex or the dorsal *attention* network, which are believed to implement specialized computations underlying visual and attentional behavior, respectively. In a sense these efforts are clearly successful. For instance, it is possible to manipulate the activity of cortical regions using transcranial magnetic stimulation (TMS) and observe predictable effects on behavior, lesions to early visual regions result in predictable visual field deficits, and sophisticated computational models exist that impressively replicate various properties of real-world brain networks.

³ Note that different researchers may flesh out the second, computational level to a greater or lesser degree, and describe its connections to the third, neural level in more or less detail. Some researchers build full-fledged sophisticated computational models without paying much attention to the neural level, while other researchers attempt to characterize functions of brain regions using a combination of behavioral and neuroimaging methods, without fleshing out the computational operations of those regions in any detail.

However, at the same time, the practice of functional brain mapping as defined above has met serious empirical and theoretical challenges, particularly in recent years (Uttal, 2001; Pessoa, 2008; Anderson's, 2010, Anderson, 2014; Klein, 2012; Burnston, 2016b; Khalidi, 2017; Stanley et al., 2019; Viola, 2020; Zerilli, 2020). We are certainly not the first to note challenges in this domain: Previous neuroscientists critical of the functional brain mapping enterprise have called for a more effective integration of research findings at each level of analysis (Krakauer et al., 2017), have argued for a redescription of function at the level of local brain regions (Price and Friston, 2005; Poldrack, 2010), or have pointed out general limitations of current mapping efforts (Poldrack, 2006; Genon et al., 2018), among other criticisms. However, with some important exceptions to be discussed below, several of these researchers still aim to identify *the* componential computational operations of well-defined brain parts and in that sense maintain the central premises of functional brain mapping. For example, Shine et al. (2016, p. 26) state that “[i]n considering the computational capacities of independent brain regions, we will make the argument that computational specialization is not only abundant in the brain, but also that it would be difficult to imagine a working brain that did not contain such specialization.” However, the viability of this approach is in question.

Outline of the review and further analysis

In contrast to previous critiques, we will suggest that the core problem cognitive neuroscience faces is an epistemological one: The present paper will integrate empirical and philosophical literature to show that the goal of giving *any* specific, computational description of *context-independently* defined brain parts is not possible, and therefore that the explanatory strategy of mainstream cognitive neuroscience is in need of revision. The section “Challenges to the Practice of Structure-Function Brain Mapping: An Integrative Review” provides an integrative review of the empirical literature on flexible neural tuning (section “Neural Tuning and Functional Brain Mapping”), plasticity and recovery of organismal-level function following brain lesions (section “Lesion Studies, Plasticity, and Functional Brain Mapping”), inter and intra-individual neural degeneracy (section “Degeneracy and Functional Brain Mapping”), neural reuse (section “Neural Reuse and Functional Brain Mapping”), and the neuroscience of enculturated skills such as reading and mathematics (section “Enculturated Skills and Functional Brain Mapping”). These areas are increasingly being investigated by researchers within cognitive neuroscience, and we summarize one consequence of these challenges that has been recognized in the field as “weak contextualism,” which denotes variable

mappings relative to the organismal level of description (section “Consequences of Weak Contextualism”). However, section “Strong Contextualism, Instability of Mapping, and Indeterminate Part Ontology for Cognitive Neuroscience” describes the fundamental incommensurability of the goals of mechanistic explanation and any context-independent – even weak contextual – descriptions of brain function, and highlights a consequence not yet widely recognized within the field, a type of instability in functional mapping that reflects a “strong contextualism.” Most significantly, we review recent research that forces a reconsideration of what constitutes a relevant neural “part” to begin with and show that the parcellation of functional components shifts depending on the task context we choose to study at the organismal level. Finally, the section “Discussion: Implications for Cognitive Neuroscience” provides a brief methodological and theoretical sketch of a cognitive neuroscience that can maintain its central goal of constructing robust computational mechanistic explanations of behavior by being sensitive to the fact that such explanations must be contextual at all levels.

Challenges to the practice of structure-function brain mapping: An integrative review

Below we provide an integrated review of research that present challenges to structure-function mapping (see [Ames and Fiske, 2010](#); [Anderson, 2014](#); [Seifert et al., 2016](#); [Hartwigsen, 2018](#); [Rule et al., 2019](#) for previous focused reviews of each of the topics discussed in this section). Afterward, we will highlight the consequences this literature has had on the current state of the art in cognitive neuroscience.

Neural tuning and functional brain mapping

Neuroscience has a long history of characterizing the neural tuning of brain regions, where manipulations of stimulus parameters of popular interest (e.g., line orientation, face-ness of an object, etc.) are correlated with changes in neural activity (firing rate, BOLD signal, etc.; see [Buzsáki, 2020](#), for a brief description of this history). However, neuroimaging results show that many neural regions are not statically tuned to particular types of stimuli in a stable manner ([Bair, 2005](#); [Clopath et al., 2017](#)). While flexibility in neural tuning at the single neuron level has been shown for some time (e.g., [Miller, 2000](#)), recent results show that the neural tuning of most brain regions appears capable of changing rapidly between different cognitive tasks. For instance, [Çukur et al. \(2013\)](#) had participants watch videos while lying in the scanner.

They were instructed to attend to different features of the videos, specifically vehicles or humans. As they did so, it was shown that the tuning characteristics of almost every region in cortex shifted depending on the goal of the observer (i.e., voxel activity was better explained by responding to humans when searching for humans and vice versa for vehicles). Other research has demonstrated more specific effects. For instance, ventral cortical regions, typically implicated in identifying objects, do not do so in an all-or-none fashion but shift their response tuning to object identity depending on the exact task participants are performing ([Harel et al., 2014](#)). In addition, recent advancements in multi-voxel pattern analysis (MVPA; [Kriegeskorte et al., 2008](#)) suggest that the “representational geometry” (i.e., the abstract, multidimensional space of neural activity patterns; see [Kriegeskorte and Diedrichsen, 2019](#)) of different network nodes is dynamically defined by context. For instance, geometry of areas in the dorsal stream is better described by action similarity in a task where participants make judgments about action, but category similarity when participants make judgments about category (see also [Anderson and Oates, 2010](#); [Gallivan and Culham, 2015](#); [Bracci et al., 2017](#)). Similarly, patterns of activity in some areas of the cortex are implicitly tuned to the category of objects depending on cues that are available in the environment ([Matheson et al., 2021](#)). The important thing to note here is that the tuning of single cells determines the representational geometry that is measured in these studies ([Kriegeskorte and Wei, 2021](#)). Thus, changes in neural representational geometry in different tasks reflect changes in neural tuning across most of the cortex depending on the context.

Overall, these neural tuning findings challenge functional mapping efforts because they suggest that the target – the response properties of a neural region to particular stimulus information – is a moving one that is shaped by task context.

Lesion studies, plasticity, and functional brain mapping

Second, findings of neural plasticity and recovery have long complicated the functional brain mapping literature. After (sometimes extensive, e.g., [García et al., 2017](#); [Bowren et al., 2021](#)) brain damage, patients may be able to partially or fully recover the behavioral or cognitive ability that was lost ([Kolb and Gibb, 2013](#)). This trajectory of recovery can continue for years ([Hartwigsen and Saur, 2019](#)) and has even been observed following lesions in early sensory areas (in adult cats; see [Jiang et al., 2021](#)). One account of this phenomenon, consistent with the assumption that local neural regions perform specialized computational operations, is that the regained organismal-level function is supported by different computational processes. Perhaps a compensatory mechanism, involving different cognitive strategies, restores function at the

organismal level (Dixon et al., 2008; see De Brigard, 2017 for a related discussion focusing on brain and cognitive strategy changes associated with healthy aging).

Another more commonly offered account of recovery that is consistent with functional brain mapping is that, as a result of plasticity, the remaining undamaged brain regions are able to reorganize themselves to accommodate new functions; that is, neural parts are repurposed to perform new computational operations. However, if recovery is indeed a matter of repurposing (i.e., the specific computational operation formerly realized by the damaged area is now performed by another area), one would expect to lose the function that the newly colonized area was responsible for. Alternatively, it is commonly postulated that the remaining intact tissue is utilized more efficiently after recovery, allowing for the former and the new computational operation to be implemented alongside each other, but this notion of redundancy (Friston and Price, 2003) begs the question why the colonized tissue was utilized less efficiently before, given that neural tissue is notoriously expensive to maintain. To our knowledge, observations of loss of one function accompanying recovery of another function appear to be largely absent from the patient literature.

On the contrary, several recent treatment studies have reported gains in the *language* domain as a result of *upper extremity* movement therapy in stroke patients (Harnish et al., 2014; Primařin et al., 2015; see Anderlini et al., 2019 for review; and Stoll et al., 2021 for related work in limb apraxia). Furthermore, brain damage is often associated with multiple co-occurring deficits (e.g., patients presenting with both limb apraxia and aphasia following left hemisphere lesions), similarly suggesting that the impacted area supports diverse behavioral domains (Behrmann and Plaut, 2014; Goldenberg and Randerath, 2015). So called “crossed” cases of classical clusters of deficits have moreover been reported in patients with atypical lateralization of function (e.g., limb apraxia presenting together with aphasia following a *right* hemisphere lesion; Raymer, 1999; see Vingerhoets et al., 2013 for findings of co-lateralization in neurologically intact individuals). Each of these findings is hard to explain under the assumption that impacted functions are implemented by specialized neural parts, and may be interpreted as further evidence that the functional significance of any given region is sort of a chameleon which does not permit a context-free definition (Price et al., 2016; Price, 2018).⁴

4 A common alternative interpretation of the co-occurrence of symptoms is that lesions “do not color within the lines.” That is, that they may impact multiple – smaller – spatially co-located functionally specialized regions. However, this line of reasoning does not seem readily capable of accommodating the above-reported “crossed” cases and the impact of treatment in one behavioral domain on performance in another behavioral domain.

Degeneracy and functional brain mapping

It is increasingly recognized that the phenomenon of neural degeneracy – the notion that structurally different neural processes can produce equivalent behaviors at the organismal level – plays an important role in the brain (Edelman and Gally, 2001; Price and Friston, 2002; Noppeney et al., 2004, 2006; Figdor, 2010; Bateson and Gluckman, 2011; Sporns, 2011; Marder et al., 2015; Anderson, 2016; Seifert et al., 2016; De Brigard, 2017; Viola, 2020). Both inter- and intra-individual cases of degeneracy have been observed (Anderson, 2016). In the domain of numerical cognition, Tang et al. (2006) reported markedly different patterns of activation during simple arithmetic between native Chinese and native English speaking individuals, with left perisylvian activation in the former cultural group, and a network of “visual” and “premotor” regions in the latter, despite equivalent stimuli (Hindu-Arabic numerals) and equivalent performance at the behavioral level. In the language domain, Biduła et al. (2017) reported many different variants and degrees of language lateralization in neurologically intact individuals with normal language abilities, involving, in addition to more or less typical and atypical lateralization of classic language areas, right hemisphere components of the “default mode network” as well as an atypical role for the cerebellum. Finally, Merabet et al. (2008) provide evidence for intra-individual degeneracy in neurologically intact individuals by showing the existence of multiple, different neural substrates for braille reading (see De Brigard, 2017 for additional examples related to healthy aging). Degeneracy is clearly widespread, both in the intact and the lesioned brain (Fotopoulou, 2014; Price, 2018).⁵ Indeed, we would argue that functional recovery following brain damage can be considered a prime example of both inter and intra-individual degeneracy (see also Mogensen, 2011; Abrevaya et al., 2017; Hartwigsen and Saur, 2019).

Neural reuse and functional brain mapping

Fourth, there is extensive evidence from neurologically intact individuals showing that some, if not most, brain regions are implicated in many different behaviors, suggesting that they can be reused in different contexts. That is, regions are typically capable of participating in a diverse array of functions. The above-mentioned study by Merabet et al. (2008)

5 It is important to exclude the possibility that what seems like degeneracy at the neural level is in fact driven by organismal-level variability between or even within participants in cognitive or behavioral strategies when performing a task (Gardner et al., 2013; Berneiser et al., 2018).

showed, using TMS, that the occipital (i.e., “visual”) cortex of blindfolded sighted participants became causally involved in tactile perception following an intense 5-day braille reading training program, providing evidence for its functional perceptual capacity beyond the visual modality (see also Murray et al., 2015). Similarly, in congenitally blind individuals who have never possessed sight, the occipital cortex has been shown to be sensitive to non-perceptual stimulus attributes such as the grammatical structure of spoken sentences and the difficulty of math equations (Bedny, 2017). Much evidence has been marshaled in the last decade or so showing that even gross functional distinctions at the organismal level, for instance the difference between emotional processes and cognitive ones (Pessoa, 2008) or between perception and action (Cisek, 2007), do not hold at the neural level, due (in part) to reuse.

The consistency of reports of functional heterogeneity suggests that reuse is not a curiosity but a general feature of the nervous system (Anderson, 2010, 2014). Importantly, despite empirical advancements showing that some regions can be functionally further subdivided (e.g., Broca’s area; Fedorenko and Blank, 2020) reuse is observed regardless of the level of granularity at which these analyses are performed (i.e., whether the brain was parsed into, say, 10 or a 1,000 regions); thus, reuse may be reduced (Poldrack, 2006) but does not go away at an increased spatial resolution (Anderson et al., 2013; Uddin et al., 2014). For instance, it is clear that, at the resolution of the entire brain, the entire brain is reused to support different behaviors, but neural reuse can be observed even at the resolution of single neurons, with neurons involved in either sensory or motoric functions depending on the behavioral and concomitant neural context in the roundworm *C. elegans* (Bargmann, 2012). The fact that reuse phenomena do not disappear at smaller resolutions presents a major challenge to determining structure-function mappings.

Enculturated skills and functional brain mapping

Fifth and finally, Dehaene and Cohen (2007); (see Menary, 2015; Jones, 2018 for discussion) suggest that cognitive tasks that require enculturation and formal schooling in order to be displayed (like reading, writing, and mathematics) are supported by the neuronal “recycling” of neural regions that originally evolved for other purposes, yet have the right structure to implement those tasks – again a type of reuse, though reuse in this case is defined at longer timescales (see Borra and Luppino, 2018 for additional examples). The visual word form area in occipitotemporal cortex is a good example of such a region. The recycling account is convincing because these abilities have arguably emerged too recently (i.e., within the last

few thousand years) for evolution to have generated specialized cortical regions to support them.

Consequences of weak contextualism

This brief review integrates some of the most significant challenges facing cognitive neuroscience. Within the field, it is increasingly recognized that all of these phenomena suggest that the functional role of a brain region is context-dependent. However, while context-dependence is recognized by the field, we argue that this is recognition of a type of “weak” contextualism. By weak contextualism we mean that most researchers accept that behavioral context shapes a region/network’s functional role in organizing the organismal-level behavioral phenomenon, but still maintain that the functioning of the part itself is context-independent. That is, it is thought that *functions of brain parts are not stable when defined relative to the organismal level, but functions defined at the computational and neural level are* (compare to Burnston’s, 2016a, 2019 “absolutism”). Researchers sympathize with weak contextualism when they make continued calls for context-independent computational descriptions of brain regions while recognizing that this computation implements a cognitive subcapacity that contributes to many different organismal-level behaviors. For example, Vingerhoets et al. (2013) report evidence in neurologically intact individuals that skilled action and expressive language involve strongly overlapping neural components. To account for this overlap, they postulate that these components implement the production of complex (i.e., precise, articulated, coordinated, nested) learned movement, an operation that is common to both tool use and speech, explaining its involvement in each of those contexts (see Knops et al., 2009; de Wit et al., 2012; Parkinson et al., 2014 for similar conceptualizations of shared informational or computational demands across different behavioral domains). Thus, with this approach, the functional description of neural parts is context-independent and is abstract enough to account for their contextual involvement in a wide range of tasks (Price and Friston, 2005; Shine et al., 2016; Humphreys et al., 2021; see also Anderson’s, 2010 early “working” vs. “use” conceptualization of neural reuse). Note that this type of contextualism, though now often accepted in the field, is already quite far removed from the traditional structure-function accounts described in the section “Introduction,” which have typically characterized the function of brain regions relative to the organismal level (e.g., the fusiform face area is important for face identification) or at a minimum in terms of cognitive subcapacities directly related to specific organismal-level abilities (e.g., a region involved in “attentional selection”). Regardless, an acceptance of weak contextualism would still allow us to find *the* computational function of a brain part (as was suggested in the case of Vingerhoets et al., 2013).

Strong contextualism, instability of mapping, and indeterminate part ontology for cognitive neuroscience

While there may be sympathies toward weak contextualism within cognitive neuroscience, we argue that there is, inescapably, a form of “strong” contextualism that is not widely acknowledged (and therefore one that is far from accepted). Our argument builds on arguments within the philosophy of science regarding the consequences of seeking mechanistic explanation. Again, by mechanistic explanation, we mean the goal of providing a description of the parts and interactions of a system that give rise to a phenomenon (e.g., Craver, 2014), and in cognitive neuroscience the typical approach is to seek a mapping of behavior to computation to brain. Indeed, some philosophers have argued that a type of strong contextualism is an unavoidable consequence of seeking mechanistic explanations in general (i.e., not just a problem for cognitive neuroscience; see Lee and Dewhurst, 2021 for discussion) and therefore it is an unavoidable consequence of the explanatory goals of cognitive neuroscience. Here, we highlight two critical arguments to demonstrate strong contextualism; one regarding the functional mapping of the computational level to the brain level, and one regarding the parcellation of brain parts. We hope to show that these two epistemological issues demand methodological and theoretical re-orientation within the field that is much more significant than the demands of weak contextualism.

Dynamic functional mappings

First, our integrative review leads to the conclusion that behavioral context not only determines the contribution of a brain part to the organization of behavior (i.e., weak contextualism), but that *context determines which computational operation that brain part implements* in support of the behavior. That is, the computation a region performs is not a specialization of that region, but rather is determined by the behavioral and neural context in which the region finds itself, and can shift when the context changes (Sporns, 2011; Klein, 2012; Anderson, 2014, 2015a; Burnston, 2016b, 2019, 2021; Khalidi, 2017; and see Mesulam, 1990; McIntosh, 1999, 2000, 2004 for early arguments in this direction). Klein (2012) illustrates the idea clearly with a discussion of the function of pistons in trucks with engine brakes: “Most of the time, [pistons] compress a fuel-air mixture to the point of detonation and transmit the generated power to the crankshaft. On trucks equipped with engine brakes, the pistons also have a second function: when the engine brake is engaged, the pistons use power from the wheels to compress air in the cylinder, slowing

the truck. *Which function the piston performs depends on things external to it: whether it is powering or slowing the truck depends on the ignition system and the valve timing*” (p. 955, italics added; in this metaphor, this is the neural context). Notice that the function of the piston depends on whether we are interested in explaining the “going” or “stopping” action of the truck (the behavioral context); specifically, it is causally transmitting explosive force to the crankshaft in one instance and is reacting to a vacuum in the cylinder in the other. Here, we have one part (the piston) that is not simply performing an abstract function useful to both stopping and going (cf. Vingerhoets et al., 2013), but that has a functional description that is dependent on our explanatory goals. Thus, under the strong contextualism view, there is nothing specialized about the functional role of a piston – there is *no* single computational operation performed by the part that plays a role across phenomena. Thus, the mapping of the part to the computation is unstable. We argue that this conclusion holds for brain parts. The empirical evidence collected in the last section is consistent with the idea that, when seeking mechanistic explanations, brain regions are best modeled with different computations in different tasks (regardless of whether the computations are described mathematically or verbally). For example, in one context a brain region’s activity might be best modeled as multiplying an input signal whereas in another it is best modeled with addition; in one context a network might be best described as an “integrator,” while in another it is a “filter,” etc. Thus, *functions of brain parts are not stable regardless of whether they are defined relative to the organismal, or computational, or neural level.*

Note that we are not denying that neurons (and neural networks) have physiological, morphological and other neuroanatomical (e.g., topological) characteristics that ensure they can do some things and not others, in the same way we wouldn’t deny a piston’s physical characteristics that allow it to do some things and not others. Our point is that a structure’s participation in any given behavior, while obviously – and importantly – constrained by its properties, is not determined by those properties, and we cannot describe a part’s functional contribution in a context-independent way *at any level* within our mechanistic explanation.

Importantly, not even extreme abstraction of the putative computational function will allow us to recover a stable structure-function mapping for a part. This is the case given that we can choose to study an infinite number of phenomena at the behavioral level and that we will evolve skills in the future that we have not characterized yet. For instance, while the “production of complex learned movement” attributed to a neural network may apply to explanations of skilled action and expressive language (Vingerhoets et al., 2013), we are unable to rule out that the same neural network supports some behavior that does not require “complex learned movement,” simply because we haven’t studied all existing behavioral phenomena,

nor can we know what phenomena will arise in the future. Thus, unlike weak contextualism, strong contextualism reveals an instability in structure-function mapping that prevents us from ever finding *the* computational function of brain parts.

Dynamic part ontologies

The second component of our argument relates to brain parts themselves. Because the identification of parts plays a central role in mechanistic explanation (Kaiser, 2018), identifying brain parts (defined as networks, anatomical regions, subregions, circuits, or single cells) is a cornerstone of cognitive neuroscience. Thus, a *central* issue in a cognitive neuroscience that takes mechanistic explanation seriously is determining the right “part ontology” (cf. Stanley et al., 2013; Viola and Zanin, 2017). However, strong contextualism shows that context determines not only a part’s computational role in any given behavior, but also determines what we should even consider to be a relevant part in the first place – that is, *context determines the appropriate ontology of parts* (cf. Poldrack and Yarkoni, 2016; Genon et al., 2018; Uddin et al., 2019). Strong contextualism challenges the idea that there is any context-independent parcellation of the brain that cognitive neuroscience can use in its mapping efforts. That is, changing our explanatory goals (i.e., which behavioral phenomenon we seek to explain) results in a redefining of the causally relevant parts that give rise to the phenomenon (see Craver and Kaplan, 2020). Indeed, it follows from the strong contextualism described here that (1) the boundaries of functionally relevant brain parts can shift every time we identify and want to explain a new behavioral phenomenon of interest and map the requisite computations onto brain parts, (2) these boundaries need not follow any obvious structural boundaries, and (3) they might shift even within participants, depending on context. These are not mere speculations. Consistent with point (2), functional boundaries within the brain need not follow any obvious neuroanatomical boundaries. For example, King et al. (2019) reported that functional subdivisions of the cerebellum did not coincide with lobular boundaries. Further, parcellation requires thresholding and clustering approaches, and it is well-known that different approaches will lead to different neuroanatomical maps (Sporns, 2012). Additionally, if, as we have argued above, the computational operation of a part is not context-independently predetermined by its material properties alone, then there is no reason to assume that its size, shape, or position *would* be context-independently definable for neural explanations of behavior. To see this, consider the following instructive metaphor in which a fictional researcher is interested in mapping function to structure in the extraneural human body. In describing high-fiving, the researcher identifies the hand as a whole as a functional part, while in describing the feeling of soft materials for pleasure, (parts of) individual finger segments constitute the relevant parts. In line with this analysis, and consistent with (1) and (3), Salehi et al. (2020)

recently found that the boundaries of functional brain parts shift depending on the behavioral state of the participant. More specifically, they “demonstrate that the parcels are indeed consistent for a given condition, but reproducibly reconfigure across conditions, even when starting with the same initial atlas each time” (p. 2). This evidence shows that the foundational assumptions of cognitive neuroscience that it can map *the* brain parts to the computations that support organismal level behavior are not tenable.

Overall, strong contextualism presents an underappreciated characterization of brain-behavior interactions, revealing that context shifts are associated with instability of mapping between organismal, computational, and neural levels. It also leads to the counterintuitive conclusion that context determines part ontology (and not the reverse; see section “Discussion: Implications for Cognitive Neuroscience” below for more details). This epistemological consideration lays bare a unique challenge to cognitive neuroscience’s goal of “an understanding how the functions of the physical brain can yield the thoughts, ideas, and beliefs” of the mind (Gazzaniga et al., 2019, p. 4).

Discussion: Implications for cognitive neuroscience

As discussed above, much of the current infrastructure of cognitive neuroscience has been and continues to be shaped by structure-functional brain mapping, and this has real consequences for how money, time, and physical resources get distributed, which includes how undergraduate and graduate students are introduced to the field and how knowledge is classified and disseminated. In our opinion, strong contextualism should force the field to seriously reconsider its approach to the study of how organismal-level functioning maps onto the computational level and how this maps onto the neural level. Below, we gesture toward a number of implications that researchers should consider when pursuing functional brain mapping.

Strong contextualism forces a shift in the goal of mainstream cognitive neuroscience toward developing relevant part ontologies and seeking computational mechanistic explanation *in context*. First, we must recognize that by deciding that we are interested in a particular organismal-level phenomenon in a particular situation (say, visual decision making in a dual rather than a single task setting), we “fix” the part ontology which best helps explain the behavior computationally – though of course we don’t know in advance what that part ontology is (in the same way that deciding whether we are interested in going or stopping helps fix the part ontology for the best explanation of truck behavior). Thus, cognitive neuroscience should abandon the aim of describing the context-independent specialized functions of well-defined brain parts. Rather, in specific contexts it should seek to answer the empirical question of what are *relatively stable parcellations*,

likely at the level of the individual or at a minimum of a small subgroup of individuals, that together with context-dependent computational descriptions lead to robust explanations of the behavioral phenomena of interest (Salehi et al., 2020; Viola, 2020). That is, we switch from asking “What does neural part X do, and is it important for behavior Y?” to “Given behavior Y, what neural parts – maybe X, maybe Z – and associated putative computational functions are needed to explain it?”

Second, contrary to previous proposals that attempt to deal with weak contextualism (Price and Friston, 2005; Poldrack, 2010), our analysis suggests that the determination of a part ontology should actually take place at a relatively *low* (rather than high) level of abstraction, rendering it more useful for explanations in specific contexts (Klein, 2012; Anderson, 2014; Burnston, 2016b, 2021). For instance, the hippocampus is a behaviorally promiscuous region that appears in explanations within many areas of cognitive neuroscience, most commonly in accounts of spatial or mnemonic behaviors (Jeffery, 2018, for a brief history) but other types of tasks as well, leading some to suggest there is an impasse in theorizing about hippocampus function (Ekstrom and Ranganath, 2018; see Humphreys et al., 2021 for a similar discussion centering on the angular gyrus). Consistent with weak contextualism, one approach to this staggering complexity has been to offer relatively abstract functional descriptions that could account for the hippocampus’s role in all of these phenomena, for instance pattern separation/completion (Yassa and Stark, 2011), scene construction (see Maguire et al., 2016 for debate), or context equivalency (Maurer and Nadel, 2021). Notice that these are all descriptions of functions that are abstract enough to be equally applicable across many parts of the brain and clearly show conceptual overlap (in at least the verbal accounts of their computational roles); further, it is hard to imagine identifying a behavioral phenomenon that would not implicate these types of computational functions. In contrast, strong contextualism points to why such functional promiscuity exists, and suggests that the hippocampus may be best understood, mechanistically, as a part that implements one type of computational operation that supports some behaviors in some tasks (e.g., memory in memory tasks), but other operations in other tasks (e.g., in navigation tasks). Importantly, the hippocampus, as a brain part defined in a context-independent way, may not be the right “part” for explanations of some other tasks which may require further decomposition into additional parts of the hippocampus itself (e.g., in imagination tasks). Another natural consequence from this analysis is that a functional part may crosscut gross neuroanatomical boundaries. As a hypothetical example, in some contexts but not others subregions of both the angular gyrus and the supramarginal gyrus may together constitute a functional part. Ultimately, the part ontology will depend on whether we are interested in explaining memory, navigation, or imagination.

Note that we are not denying that the hippocampus (and its neural contexts) have anatomical structure and physiological

properties that we can characterize – and in fact, in moving away from focusing on context-independent functional descriptions, the description of such constraints takes on an increased importance (see also Anderson, 2015a; Bolt et al., 2017). For instance, there is evidence that, given its physiology, the hippocampus implements sequence generation (Buzsáki and Tingley, 2018). However, the question is about the context-independent relevance of any given anatomical or physiological feature for explanations within cognitive neuroscience. We are suggesting that the degree to which a particular part (e.g., hippocampus) with particular properties will make a stable appearance in computational *mechanistic explanations* of particular behaviors across contexts, and whether the mechanistic explanation (with a particular part ontology) results in strong predictions for behaviors of interest, is an empirical question to be addressed in cognitive neuroscience under the specter of strong contextualism.

Finally, the implications of strong contextualism extend beyond strategies for empirical work; it affects how we theorize and interpret results. Indeed, it is recognized by many that the behavioral and neural sciences are in a “theory crisis” (Eronen and Bringmann, 2021). Our review suggests that one source of the problem is failing to recognize the instability of mapping “paying attention,” “using a tool,” and “speaking” to computational operations and brain parts. Strong contextualism, then, suggests an additional way to address the theory crisis: By developing formal computational mechanistic explanations in context, cognitive neuroscience can explicitly test effective part ontologies with associated computational operations and functional roles based on the phenomena it seeks to understand, *where the phenomena it seeks to understand are open to revision and new classification*. Doing so opens up cognitive neuroscience not only to better “part ontologies” but also to more useful cognitive ontologies (Poldrack, 2010). Such an approach may lead to further advancements in which historic ontologies may be dismantled in favor of ones with potentially greater accuracy and hence hopefully wider reaching clinical and scientific impact (e.g., see Renoult et al., 2019, for an example of the dismantling of the episodic vs. semantic distinction and challenges this poses to conceptions of “memory”; see Anderson, 2011 for challenges to the reification of “attention”; see Buzsáki, 2020, for challenges to the distinction between perception and action, etc.). Thus, we foresee an iterative process in which respecting the notion that context fixes the part ontology that best explains behavior allows us to remain open to other functional mappings in other contexts, and searching for empirically useful part ontologies will also shape how we identify and characterize the behavioral phenomena we wish to explain (see also Anderson, 2015b). This is a suggestion that goes beyond stating we need to understand behavior better before we should seek mechanistic explanations (Krakauer et al., 2017); rather, it opens up the possibility of characterizing both part and behavioral ontology differently than we do now, making way for new theoretical approaches and insights into brain-behavior

interactions in a way that allows cognitive neuroscience to feed back into the behavioral sciences.

We have encountered resistance to the arguments presented here. Strong contextualism is counterintuitive and there are two common reactions to these arguments that we want to briefly address before we conclude. First, one might accept weak contextualism but conclude the strong version is a step too far. Again, as we have shown, many researchers hold that there is a single abstract computational process that each region performs, and we just have to wait for technology and/or methodology to catch up before we can successfully map that region to a computation that maps to behavior. We have already addressed the limitations of this conclusion but want to highlight here that our arguments regarding strong contextualism are not technological or methodological, but primarily epistemological (and our integrative review provides empirical support for the epistemological issues we have highlighted). Advances in technology or methodology without theoretical advances will not address the problems identified in the sections “Challenges to the Practice of Structure-Function Brain Mapping: An Integrative Review” and “Strong Contextualism, Instability of Mapping, and Indeterminate Part Ontology for Cognitive Neuroscience.”

Second, one might argue that contextualized explanation is already the current state of the art, in which researchers in disconnected (siloes) subfields of cognitive neuroscience identify context-sensitive functions for brain parts, no one is *really* looking for context-independent explanations in cognitive neuroscience, and everyone already acknowledges that there is no principled way of discovering *the* mapping between a structure and its computations. If this was indeed the state of the art, there would be no reason for the anatomical parcellations to be similar across silos, as accepting strong contextualism means accepting empirically defined parcellations that will be contextually driven. However, parcellations *are* similar across silos (e.g., hippocampus is mainly structurally defined in both memory and navigation research), suggesting that researchers do think there is a meaningful context-independent parcellation of the brain that can be mapped to computations and behaviors of interest.

Ultimately, the efforts of cognitive neuroscience will be judged by their utility in fulfilling its goals as stated above (Gazzaniga et al., 2019; Journal of Cognitive Neuroscience, 2021). It is our contention that taking strong contextualism seriously, both when it comes to determining computational

operations, and when it comes to determining the size, shape and location of the neural parts that instantiate these operations, will position us better to mechanistically explain the clinical behavioral observations that this article started with, and even to more fully understand what it means for us to pay attention, use a tool, or speak to one another.

Author contributions

All authors listed have made a substantial, direct, and intellectual contribution to the work, and approved it for publication. Authors have contributed equally to this manuscript.

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Conflict of interest

The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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