Universal Transform or Multiple Functionality? Understanding the Contribution of the Human Cerebellum across Task Domains

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An impressive body of research over the past 30 years has implicated the human cerebellum in a broad range of functions, including motor control, perception, language, working memory, cognitive control, and social cognition. The relatively uniform anatomy and physiology of the cerebellar cortex has given rise to the idea that this structure performs the same computational function across diverse domains. Here we highlight evidence from the human neuroimaging literature that documents the striking functional heterogeneity of the cerebellum, both in terms of task-evoked activity patterns and, as measured under task-free conditions, functional connectivity with the neocortex. Building on these observations, we discuss the theoretical challenges these results present to the idea of a universal cerebellar computation and consider the alternative concept of multiple functionality, the idea that the same underlying circuit implements functionally distinct computations.

The cortico-cerebellar system is one of the most prominent networks in the human brain (Figure 1A). Of the 40 million axons that exit the neocortex and traverse the cerebral peduncles, the vast majority send collaterals to the pontine nuclei (Tomasch, 1969), and these, in turn, project to the granule cells of the cerebellar cortex. The 50 billion granule cells constitute more than half of the neurons in the human brain (Azevedo et al., 2009). Thus, this transmission point corresponds to an information expansion of at least 1:1,000, creating the system with the largest information bandwidth in the human brain. Each granule cell gives rise to a single axon, the parallel fibers that innervate Purkinje cells, the main computational elements of the cerebellar cortex. Purkinje cells deliver their output, via the deep cerebellar nuclei (DCNs) and thalamus, back to the neocortex. Unfolded, the cerebellar cortex has approximately the same surface area as one cerebral hemisphere (M.I. Sereno et al., 2014, Soc. Neurosci., abstract), and it accounts for approximately one-fifth of the entire energy budget of the human brain (Howarth et al., 2010). Based on these numbers alone, the "Kleinhirn" (i.e., little brain) is simply too large to ignore.

What is the function of this remarkable system? Based on the clinical symptoms observed in people with cerebellar damage, the historical focus has always been on the fine control of movement (Holmes, 1939). However, a paradigm shift can be associated with the publication of a paper by Leiner et al. (1986), who, based on evolutionary considerations, suggested that the cerebellum may have an important role in human cognition. Leiner et al. (1986) were impressed by the expansion of the cerebellar hemispheres in anthropoid apes and humans that paralleled the increase in size of the prefrontal cortex. Subsequent cross-species comparisons suggested that the expansion of the cerebellum disproportionally outstripped that of prefrontal areas (Barton and Venditti, 2013, 2014). More recently, analysis of intra-cranial volumes indicates that cerebellar size is the most prominent neuroanatomical difference between *Homo neanderthalensis* and early *Homo sapiens* (Kochiyama et al., 2018). Noteworthy is that these changes are specific to regions of the cerebellum that have been linked to non-motor functions (Balsters et al., 2010), supporting the argument that the cerebellum may have played a key role in the evolution of human cognition.

Anatomical tracing studies have revealed extensive communication between almost the entire neocortex and the cerebellum (Strick et al., 2009). Pontine projections arise not only from motor areas but also from prefrontal (Schmahmann and Pandya, 1997a), parietal, superior temporal (Schmahmann and Pandya, 1991), and parahippocampal areas (Schmahmann and Pandya, 1993). Moreover, projections from the dentate nucleus return to a similar set of cortical association areas (Dum and Strick, 2003; Kelly and Strick, 2003; Middleton and Strick, 1997). The presence of these loops strongly suggests a role of the cerebellum that extends well beyond motor control.

Looking beyond anatomy, two other lines of research have provided a compelling case for involvement of the cerebellum in cognition. First, the clinical picture of patients with cerebellar disorders as well as their performance on neuropsychological test have revealed a broad range of impairments, including deficits in executive function, language, and affect (Kansal et al., 2017; Schmahmann and Sherman, 1998; Tedesco et al., 2011; but see Alexander et al., 2012). Second, functional neuroimaging
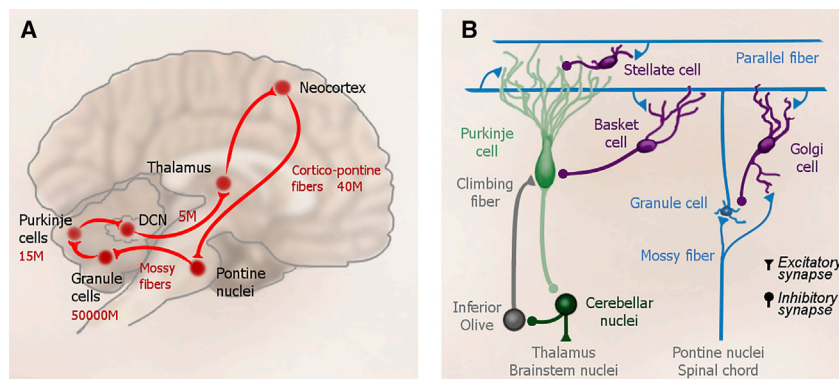
Figure 1. Cerebellar Circuitry
(A) The main connections of the cortico-cerebellar loop. Numbers indicate rough estimates of the number of projections or cells in the human brain in millions (M).
(B) Local circuit in the cerebellum. Every granule cell receives 4–5 mossy fibers and gives rise to a single parallel fiber. Each Purkinje cell receives input from ∼175,000 parallel fibers as well as from a single climbing fiber that originates in the inferior olive. Purkinje cells send inhibitory projections to cells in the deep cerebellar nuclei (DCNs). The inhibitory interneurons (mainly Golgi, stellate, and basket cells) complete the circuit.

studies involving healthy individuals have revealed engagement of the cerebellum in a surprisingly diverse set of tasks (Strick et al., 2009). What remains unclear, however, is how to best characterize the functional role of the cerebellum across all of these domains.

A Single Cerebellar Computation or Multiple Functionality?

The cerebellar circuitry is highly uniform across the entire cerebellar cortex, inspiring the belief that cerebellar function might be conserved when generalized to task domains beyond motor control. In the words of Leiner et al. (1986), "The hypothesis states that in the human brain the newest cerebrocerebellar loops could contribute to skilled mental performance in much the same way that the older loops contribute to skilled motor performance." Building on this notion, Schmahmann (1996) coined the term "universal cerebellar transform" to capture the idea of a single cerebellar computation.
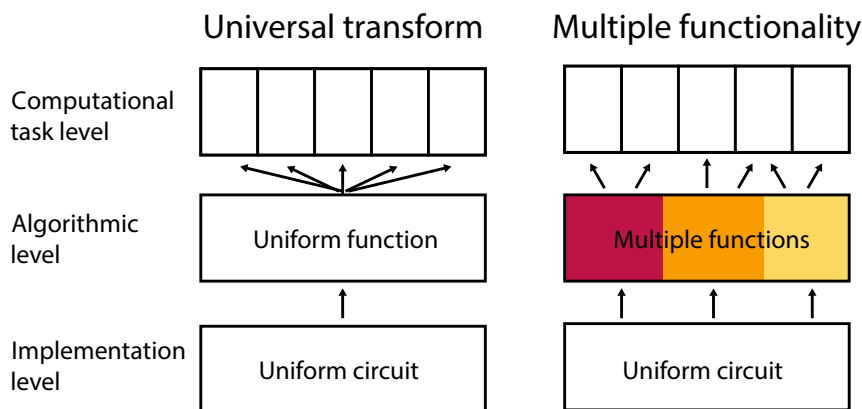
This idea contrasts dramatically with the common conception of the neocortex. Here, different regions can be distinguished based on their unique cytoarchitecture, myelination patterns, and gene expression (Toga et al., 2006). These features suggest a highly specialized organization with processing modules subserving specific functions that are supported by their local circuitry. Only in the most abstract sense would we ever consider the question "what is the function of the neocortex?". In contrast, given the uniform circuitry, the question "what is the function of the cerebellum?" seems much more sensible. But does the question have an answer?

As an approach, it is useful to consider the problem in terms of the three levels of analysis introduced by Marr (1982) (see also Dean and Porrill, 2016). We can distinguish between the computational level (which task the brain has to solve), the level of representations and algorithms (how the brain solves the task), and the level of implementation (how the requisite processes are realized in neuronal tissue). Given the heterogeneity of clinical deficits and functional activation patterns observed in the cerebellum, it is clear that commonalities cannot be found at the first level because each task will demand its own computational description. In contrast, the uniformity of the cerebellar circuitry suggests an invariance at the implementation level in terms of connectivity and plasticity. What is unclear, however, is whether we can identify a theory at the level of algorithms

and representation that explains how the uniform cerebellar circuitry supports the diverse set of computations required at the task level. Such a theory would conform to a universal cerebellar transform. Alternatively, it may be that, as the functional domain of the cerebellum diversified over the course of evolution, so too did the computations that are supported by its circuitry (Figure 2).

Candidate hypotheses for a universal transform at the algorithmic level have been motivated by an influential theory of how the cerebellum works at the implementation level. The distinctive anatomical and physiological features of the cerebellar circuitry (Figure 1B), well-conserved across species, were described in exquisite detail in the middle of the 20th century (Eccles et al., 1967). This work inspired the development of a circuit-level model of cerebellar function, commonly known as the Marr-Albus-Ito model (Marr, 1969; Albus, 1971; Ito and Kano, 1982). This model builds on two core ideas. First, information coming through the mossy fibers is massively expanded by projection onto the 50 billion granule cells, providing a detailed representation of the context via its near-infinite set of patterns. For example, the exact duration of a simple tone or light can be encoded in a repeatable, time-varying activity pattern that is transmitted to Purkinje cells via parallel fibers (Medina et al., 2000a, 2000b). Second, the unitary climbing fiber input from the inferior olive to the Purkinje cell serves as a teaching signal for supervised learning. The activation of a climbing fiber causes long-term depression of recently activated parallel fiber-to-Purkinje cell synapses (Ito and Kano, 1982). Thus, when the pattern that previously preceded a climbing fiber input is detected again, the Purkinje cell reduces its firing rate, disinhibiting cells in the DCN cell. The output of the DCN neurons can help to produce or shape the desired behavioral output as well as to inhibit the corresponding inferior olive neuron, reducing climbing fiber input to the cerebellum as learning unfolds. The Marr-Albus-Ito model has proven, with minor modifications, to provide a compelling account of the cerebellar role in simple sensory-motor learning tasks (e.g., Medina and Lisberger, 2008). At a more abstract level, the core tenet of the model is that each Purkinje cell learns to predict its climbing fiber input based on the context signaled by the complex pattern of parallel fiber activity.

The key elements of the model, pattern expansion, supervised learning, and timed prediction have provided a springboard for theorists considering a general characterization of cerebellar

Universal transform

Multiple functionality

Computational
task level

Algorithmic
level
Uniform function

Multiple functions

Implementation
level
Uniform circuit

Uniform circuit

Figure 2. Schematic of the Universal Transform and Multiple Functionality Hypotheses, Considered across Marr's Three Levels of Analysis
At the computational level, each task demands a different computational description. At the implementation level, the cerebellar circuitry is remarkably uniform. The idea of a universal transform holds that, at the algorithmic level, we can formulate a general idea of how cerebellar circuits contribute to diverse functions. In contrast, the multiple functionality models posits that different tasks rely on variable contributions from a number of cerebellar functional modules, each of which requires a distinct algorithmic description.

function at the algorithmic level. These ideas include prediction (Miall et al., 1993), internal models (Ito, 2008; Wolpert et al., 1998), timing (Ivry, 1997; Ivry and Keele, 1989), and automatization (Balsters and Ramnani, 2011; Ramnani, 2014). These hypotheses, by their nature, are very general to encompass how a singular computation might apply across motor control, attention, working memory, language, and social cognition. The major challenge in evaluating such domain-general hypotheses of cerebellar function has been to translate them into testable experimental predictions.

For example, the timing hypothesis proposes that cerebellar circuits are involved whenever the computations require a representation of the precise temporal relationship between stimuli, events, or motor commands (Ivry, 1997). In some task domains, testing this hypothesis has been relatively straightforward; for example, in motor control, individuals with cerebellar dysfunction would be expected to be impaired when explicit timing of the movements is required (Spencer et al., 2003). In perception, deficits would be predicted for judgments that are based on temporal properties of the stimuli (Ackermann et al., 1997; Ivry and Keele, 1989). In other domains, especially in those where the cerebellum acts on internal, unobservable representations, it is much more difficult to derive critical tests. For example, although cerebellar activity is consistently observed during semantic retrieval (Petersen et al., 1989), it is not clear why generating a matching verb for the noun "apple" would require manipulation of information in a precisely timed manner.

A related conceptualization of cerebellar function is that of a predictive forward model for state estimation (Ito, 2008). This hypothesis has provided an appealing account of a number of phenomena in motor control and learning; deficits of patients with cerebellar damage in accounting for intersegmental dynamics or in learning to move in a novel environment can be understood as a failure of an internal forward model (Diedrichsen and Bastian, 2014; Tseng et al., 2007). Again, there have been some successes in extending this idea to cognitive domains. For example, there is evidence that one way in which the cerebellum supports language is by using forward models to generate semantic expectancies based on the linguistic context (Lesage et al., 2012; Moberget et al., 2014). However, many unresolved questions remain. What is the error signal used to shape a forward model of semantics? How are the predictions of a cerebellar forward

model distinct from predictions that can be generated by neocortical circuits alone? Clearly, prediction is not unique to the cerebellum; most neural activity can be understood as some form of prediction (Friston, 2009).

In summary, the uniformity of the cerebellar circuitry has been a powerful argument that, at some level, there exists a common computational principle that applies across various task domains. However, the remarkable heterogeneity of the human cerebellum at the task level poses an important challenge to the universal transform hypothesis because it requires identifying a common principle that holds across a large number of disparate task domains. Notably, in the 20 years since formulation of the universal cerebellar transform idea, very little progress has been made toward systematic evaluation and comparison of domain-general hypotheses of cerebellar function.

This may be partly due to the fact that we currently lack an understanding of the computations underlying different mental activities that is detailed enough to generate strong and falsifiable predictions concerning cerebellar function. However, it is also important to consider the more fundamental question of whether we should expect to observe a common theory of cerebellar function at the algorithmic or representational level. The idea of a universal transform is predicated on the assumption that there is a one-to-one relationship between the implementation and algorithmic levels; i.e., that a uniform circuit implies a uniform function. However, this assumption may be incorrect. It is widely recognized that the same algorithmic process can be implemented in many different but functionally equivalent ways, a concept referred to as "multiple realizability" (Fodor, 1975; Putnam, 1988). Similarly, we may need to consider the possibility of "multiple functionality," the idea that the same circuit, at the implementation level, can be used to realize quite different computations. For example, the predictive forward mode may have great explanatory power to capture the function of the cerebellum in motor control, but an entirely different concept may be needed to describe the role of the cerebellum in language comprehension. Thus, to describe cerebellar involvement in a broad range of tasks, it may require a (finite) set of modules, each of which requires its own functional description (Figure 2, right).

At this point, it is unclear whether a universal transform or a set of multiple functions will provide the more useful description of cerebellar processing. We do believe that two prerequisites are
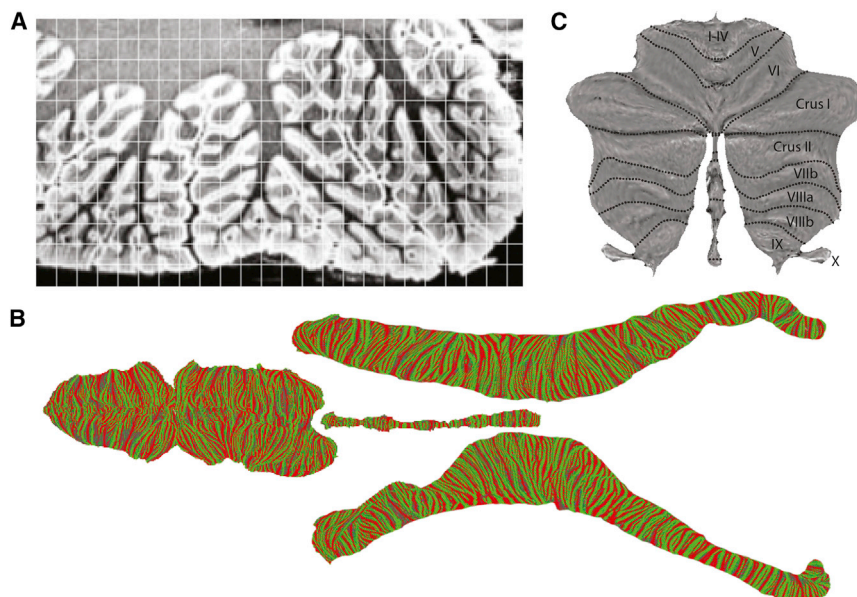
Figure 3. The Surface of the Cerebellar Cortex

(A) A section showing inferior lobules VIIIa, VIIIb, and IX of the left hemisphere of a human cerebellum scanned at 150-μm resolution. The scan clearly visualizes the granular cell layer in medium gray and the molecular layer in white. The superimposed grid indicates a typical sampling of a functional scan at 1.5-mm resolution.

(B) Unfolded representation of an entire human cerebellum (M.I. Sereno et al., 2014, Soc. Neurosci., abstract). For flattening, the surface is cut into 4 pieces. Color indicates local curvature, with green indicating the crest of a folium.

(C) Simplified surface at the level of cerebellar lobules for the display of volume-averaged function data (Diedrichsen and Zotow, 2015).

essential for significant advance to be made on this question. First, we need to obtain a much more detailed picture of the functional heterogeneity within the cerebellum, understanding the importance of different areas for different tasks. Second, we need to understand how activity within each cerebellar sub-region is coordinated with corresponding neocortical regions. Functional imaging research of the human cerebellum is starting to provide considerable insight into these two important questions. For the remainder of this paper, we review these findings in detail, returning, in the final section, to the question of how this information can help with understanding cerebellar function at a more general level.

Functional Imaging of the Human Cerebellum: Methodological Considerations

Extensive neuroimaging literature has revealed prominent activation of the human cerebellar cortex across diverse task domains. Before reviewing these data, we need to consider two methodological issues. First, it is important to understand which neural processes in the cerebellar cortex lead to changes in the fMRI signal. fMRI measures the blood oxygenation level-dependent (BOLD) signal. How this signal relates to neural processing depends, to a large extent, on the mechanisms that govern metabolism and blood flow, and these vary considerably between the neocortex and cerebellum (Vaishnavi et al., 2010). The careful work of Caesar et al. (2003) and Mathiesen et al. (2000) has revealed important insights concerning the regulation of blood flow in the cerebellum. Mossy fiber input and the resultant granular cell activity cause substantial increases in cerebellar blood flow (Caesar et al., 2003; Mathiesen et al., 2000). In contrast, even large increases in the activity of Purkinje cells, either through changes in simple or complex spike firing rates, produce no measurable change in blood flow (Thomsen et al., 2004, 2009), suggesting that Purkinje cells may not be able to trigger a vasodilatory response (Attwell and Iadecola, 2002).

These empirical observations seem reasonable when considering the energy expenditure in the cerebellum. At rest, approximately 80% of energy use in the cerebellar cortex is related to signal transmission in the granular cell layer (Howarth et al., 2010). Although Purkinje cells contribute another 15% to composite energy use, the high baseline firing rate of simple spikes (50–80 Hz) and low frequency of complex spikes should result in relatively stable energy demands. In contrast, granule cells have a large dynamic range (1–600 Hz). Taken together, this body of work suggests that blood flow changes in the cerebellum most directly track activity modulation in the mossy fiber-granule cell system and likely tells us very little about the activity and computation of Purkinje cells.

The second important issue is that of spatial resolution. With the advent of high-field imaging and improvements in gradient design, fMRI studies with less than 2-mm isotropic resolution are now commonplace. In the neocortex, this level of resolution allows for excellent localization of the source of hemodynamic changes. Based on a standard anatomical image, the surface of the neocortex can be reconstructed, and the activity patterns can be projected onto the surface of the individual brain. Indeed, this approach has become a standard in many laboratories, enabling precise analyses of the functional organization of the cerebral cortex (Dale et al., 1999).

In contrast, a surface-based analysis approach remains elusive for cerebellar fMRI data. Given the intricate folding of the cerebellar folia, complete unfolding of the human cerebellum is extremely challenging (Figure 3A), with complete unfolding of the surface only possible when the scanning resolution is better than 200 μm (M.I. Sereno et al., 2014, Soc. Neurosci., abstract; Figure 3B). Even when fMRI data are acquired at 1-mm resolution, the BOLD signal will mix activation signals across neighboring folia. Thus, a surface-based display of the cerebellar cortex would cause activation of a single folium to appear distributed over various locations on the flattened surface.

Given this problem, we developed a hybrid solution to create a surface-based visualization of volume-averaged cerebellar activity data. This flat map averages across neighboring folia that cannot be cleanly resolved (Diedrichsen and Zotow, 2015; Figure 3C). The display is designed to be proportional, with the
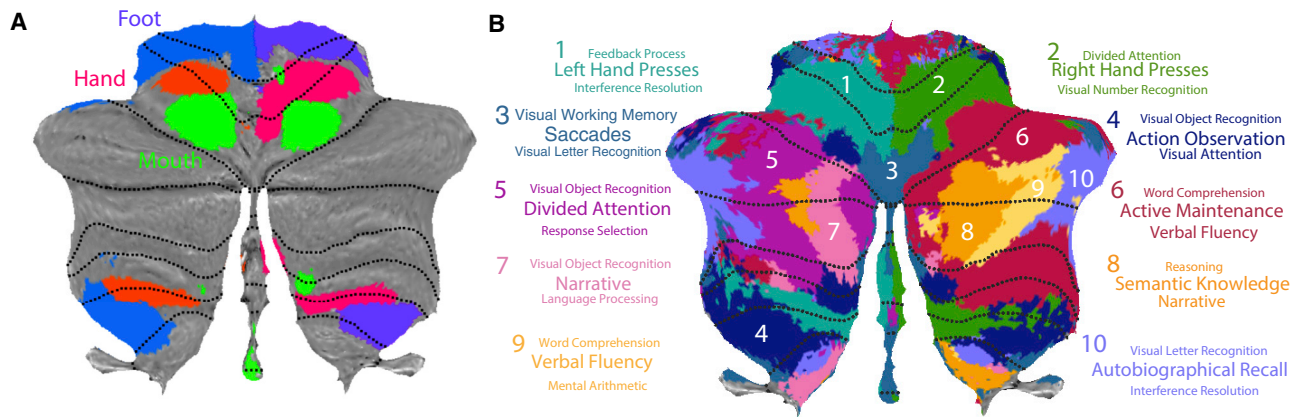
Figure 4. Functional Organization of the Human Cerebellum
(A) Somatotopic representation of foot (blue), hand (red), and tongue (green) movement in the cerebellum (Diedrichsen and Zotow, 2015).
(B) Functional parcellation of the cerebellum based on a multi-domain task battery (King et al., 2019). Parcellation and underlying task contrasts are available at http://www.diedrichsenlab.org/imaging/mdtb.htm.

surface area of each cerebellar region on the flat map corresponding, approximately, to its gray-matter volume. Note that the actual cerebellar surface is ~15 cm wide and ~1.2 m long (M.I. Sereno et al., 2014, Soc. Neurosci., abstract). This means that, although 1 cm in the horizontal direction on the flat map corresponds to the comparable distance on the surface of the cerebellar cortex, 1 cm distance in the vertical direction on the flat map corresponds to ~10 cm on the cerebellar cortex.

Imaging Studies of Functional Heterogeneity

One of the earliest reports of functional activation in the cerebellum was in the seminal positron emission tomography (PET) study by Petersen et al. (1989), who set out to describe the brain's language network through a series of nested contrasts. One cerebellar region, the superior motor representation, was activated during the overt production of words, relative to passive viewing or listening, consistent with a role in speech production. Surprisingly, a second cerebellar region, localized to *crus* I and II in the right hemisphere, was more activated when the participants were required to generate a semantic associate to the stimulus word compared with when they simply repeated the stimulus word. Because this contrast controlled for motor demands, this result suggested a role of the human cerebellum in language. Interestingly, this finding was so unexpected, violating preconceptions of cerebellar function, that the authors opted to ignore the cerebellum in their initial report (Petersen et al., 1988). However, subsequent work not only replicated this result but showed that cerebellar involvement is a much broader set of domains, including attentional control (Allen et al., 1997), working memory (Desmond et al., 1997), emotion processing (Baumann and Mattingley, 2012), and social cognition (Svoboda et al., 2006). Indeed, it soon became recognized that it was rare to find task contrasts that did not engage the cerebellum.

In an attempt to make sense of this plethora of results, a number of groups have conducted meta-analyses of the cerebellar activation patterns (E et al., 2014; Van Overwalle et al., 2014; Stoodley and Schmahmann, 2009). One important insight

from this work is that there are clearly differentiable loci for motor and non-motor tasks. Motor activity tends to be restricted to two topographic areas, a superior region spanning lobules IV–VI and an inferior region in lobule VIII. In contrast, lobule VII is activated by more cognitive tasks. Although they provide a reasonable first-pass overview of the functional organization of the human cerebellum, these meta-analyses have important limitations. First, they are based on the reported foci of highest activity, ignoring information about the extent and shape of the activation. Second, many of the studies included in these analyses provide relatively imprecise information about the coverage of the cerebellum (given that the focus is often on the cerebral cortex), making it hard to interpret the absence of activation. Finally, the fact that each task was studied in different sets of participants with different normalization methods makes direct comparisons across tasks difficult.

An alternative approach that mitigates these concerns is to study a broad task battery in the same set of participants. Stoodley et al. (2012) conducted the first study of this kind with respect to the cerebellum, testing seven participants on a set of five different tasks. A much larger-scale effort is contained in the Human Connectome Project (HCP), which includes data from 1,000 participants performing seven tasks (Barch et al., 2013). The projection of the activity patterns from these tasks onto the flat map representation of the cerebellar cortex reveals a complex but coherent functional organization (Diedrichsen and Zotow, 2015; Guell et al., 2018a). For example, the two motor representations can be shown to have an ordered arrangement for hand, foot, and tongue movement (Figure 4A). Foot movements occupy lobules I–IV and the lateral aspects of lobules VIIIb. Hand movements elicit activity on the boundary of lobules V and VI and in the medial aspects of lobule VIII, areas that also contain a representation of individual finger movements (Wiestler et al., 2011). Tongue movements produce activity in lobule VI and medial aspects of lobule VIIIa. However, even this extensive dataset has important limitations in that the number of tasks remains relatively small. Furthermore, different tasks are typically performed in different runs, again making direct comparisons

across tasks problematic, given that a common baseline measurement is lacking.

To make a more direct assault on the question of functional heterogeneity in the human cerebellum, we developed a rich, multi-domain battery involving 26 tasks entailing 47 unique task conditions (King et al., 2019). Participants were scanned while performing one of two subsets of 17 tasks during each of the four 80-min sessions, preceded by training to ensure that the participants were proficient in performing each task and in flexibly shifting between tasks every 35 s.

The diverse task battery resulted in activation patterns that encompassed almost the entire surface of the cerebellar cortex. To provide a concise summary of these data, we subdivided the cerebellum into 10 regions, each with a specific activation profile across all task conditions. This map reveals a picture of both the functional organization of the human cerebellum as well as its functional heterogeneity (Figure 4B). Activity in regions 1 and 2 was associated with left and right finger movements, corresponding to the hand areas in lobules IV–VI and VIII. We did not include foot movements in our battery, which likely accounts for the relative absence of activity in lobules I–IV. Region 3 encompasses the human equivalent of the oculomotor *vermis* observed in the macaque (Nitschke et al., 2005; Ohtsuka and Noda, 1995). Activity in this region strongly correlated with the number of eye movements made in each task (King et al., 2019) and is, as shown recently, retinotopically organized (van Es et al., 2019). Interestingly, the activity level in this region also appeared to be strongly modulated by the demands on visuo-spatial attention, suggesting that the recruitment of this area goes beyond that required by the actual eye movements.

Region 4 was most strongly engaged in tasks involving action observation. Activity here was quite pronounced when participants passively watched videos of knot-tying (Cross et al., 2012; King et al., 2019). Interestingly, this finding seems to contradict a recent meta-analysis (Van Overwalle et al., 2014) that found no evidence of cerebellar activation during "mirroring" tasks. The discrepancy may relate, in part, to incomplete coverage of the cerebellum in some of the studies included in the meta-analysis. Even when studies report "whole-brain" imaging, the inferior aspects of lobules VIII and IX are often cut off and subsequently filled in with smoothing procedures. This can lead to a substantial loss of statistical power in these regions.

The remaining regions were associated with higher-level cognitive processes. Regions 5–9 cover the medial and midlateral aspects of *crus* I, *crus* II, and lobule VIIb. Tasks with high demands on attentional processes (e.g., divided attention and active maintenance) contributed substantially to the activation patterns in regions 5 and 6. In contrast, tasks involving language-related processing (e.g., narrative, language processing, semantic knowledge, word comprehension, and verbal fluency) figured prominently in the activity profile for regions 7, 8, and 9. Consistent with prior studies, language-related functions were more prominent in *crus* I and II in the right cerebellar hemisphere. Regions 7–9 also included the cerebellar component of the default-mode network (Raichle et al., 2001), brain regions that tend to be more active during rest than during task performance. Finally, region 10, located in the most lateral aspects of lobule VII,

loaded heavily on tasks that involved autobiographic memory and recall.

We do not intend to suggest that this parcellation provides the "final" functional map of the human cerebellum, nor do we argue that a subdivision into 10 regions has a special status. Rather, the map offers one visualization that captures the functional heterogeneity of the cerebellum. Combined with other meta-analyses and re-analyses of multi-task datasets (Guell et al., 2018b), this functional map raises a number of interesting questions. How do we characterize the apparent asymmetries between the cerebellar hemispheres? Are the lateral aspects of lobule VIII better characterized as a foot or as an action observation region? Do emotional pictures elicit reliable vermal activity (e.g., region 3), as has been claimed elsewhere (Guell et al., 2018a), or is this activity better explained by the demands on visual attention or eye movements? A combination of condition-rich experimentation and data integration across studies will ultimately provide us with a much more detailed picture of the functional heterogeneity of the cerebellar cortex.

Does the Cerebellum Have Distinct Functional Regions?

Current functional maps of the cerebellum have already revealed a number of important organizational principles. For example, these maps suggest that functional regions within the cerebellum do not respect lobular boundaries. This observation is especially salient in the parcellations derived from our multi-domain task battery (Figure 4B), where many of the boundaries appear to be unrelated to the lobular divisions. To quantify this observation, we developed a simple method to evaluate the degree to which a boundary separates functionally distinct regions (King et al., 2019). This method involves comparing the similarity of the functional profiles of two voxels within a region relative to two voxels across a boundary. Because the functional similarity of two voxels is highly dependent on their spatial distance, with closer voxels being more similar, we matched the distance for within- and between-region voxel pairs. This novel criterion clearly showed that lobular boundaries only weakly mark functional segregation. Indeed, many lobular boundaries did not demark a functional change that was larger than the corresponding within-lobule change.

Given this finding, it is possible that the cerebellum does not have any distinct functional regions but, rather, is better characterized as continuous gradients of functional differentiation based on slow variation in the inputs (Guell et al., 2018b). To test this idea, we asked whether the functional parcellation derived from our multi-domain task battery (Figure 4B) defined real functional boundaries or whether it simply cut up an underlying continuous map in an arbitrary way. Of course, it is a trivial exercise to show that, given a limited task set, one can always find boundaries across which the activation profiles change more than within a boundary. To be a real or meaningful functional boundary, the parcellation needs to be predictive of functional differences when tested with a completely new set of tasks. Using data from independent tasks within the multi-domain task battery, we were able to show this to be the case. This analysis allowed us to conclude that the nature of the input to the cerebellum changes rather abruptly across the newly identified functional boundaries. Interestingly, these boundaries

neither coincide with lobular boundaries nor do they resemble Zebrin zones that have been identified through immune-histological staining in rodents and non-human primates (Sugihara and Shinoda, 2004).

The identification of functional boundaries and the fact that these do not align with the macro-anatomical landmarks have important practical implications for studies of the human cerebellum. References to anatomical localization in the cerebellum are, because of a lack of other defining criteria, often made by reference to lobules (Schmahmann et al., 2000; Diedrichsen et al., 2009), with within-lobule divisions limited to distinguishing between *vermis* and hemisphere. Similarly, regions of interest for functional and anatomical analyses also tend to be based on the lobules of the cerebellum (e.g., Kansal et al., 2017). Our quantitative analysis of functional variation shows that lobular divisions have minimal predictive utility. Indeed, the functional specialization within the hemispheric aspect of lobule VII changes multiple times between paravermal and the most lateral regions (Figure 4B). Thus, analyzing data using lobules as the basic modular units mixes signals from separate functional regions. The use of functionally defined and carefully evaluated regions of interest (ROIs) should therefore lead to clearer insight into functional differences between different cerebellar regions.

Cerebro-cerebellar Connectivity

Given the cytoarchitectonic homogeneity, the functional diversity of the cerebellum must arise from variation in the inputs to the cerebellar cortex, most prominently, the massive projections from the cerebral cortex via the pontine nuclei (Schmahmann and Pandya, 1997b). For example, Kelly and Strick (2003), using a combination of retrograde and anterograde transsynaptic tracers, showed that M1 is connected reciprocally to the two motor areas in lobules IV–VI and VIII, whereas prefrontal area 46 has extensive connections with the lateral aspect of *crus* II. Although studies such as these provide important insights into cerebrocerebellar networks, current anatomical methods can only reveal a small piece of the overall connectivity puzzle at a time. Additionally, this work can only be performed in non-human primates; we need to keep in mind that there may be substantial inter-species differences, especially for the cerebellar hemispheres.

An important breakthrough in the study of human cerebrocerebellar connectivity was provided by "functional connectivity" studies of the human brain. Biswal et al. (1995) made the important observation that fMRI time series of specific pairs of regions correlated with each other, even when participants simply rested in the scanner without performing a task. Common activity fluctuations are interpreted as evidence that these two areas are functionally connected even though they may not share direct anatomical connections. Although this approach was originally developed to explore functional connectivity in the cerebral cortex, a number of studies have employed this method to study the otherwise inaccessible long-range connections between the human neocortex and cerebellum (Habas et al., 2009; Krienen and Buckner, 2009; O'Reilly et al., 2010).

In a now seminal paper, Buckner et al. (2011) used resting-state fMRI data from 1,000 participants to produce a detailed map of cerebellar organization based on cortically defined networks (Figure 4). The map clearly showed the two motor regions

of the cerebellum with body part-dependent connectivity and confirmed the connectivity of lobule VII to prefrontal and parietal association areas. Subsequent studies using slightly different methodologies have shown similar pictures of the organization of the cerebellum (Marek et al., 2018; Ji et al., 2019).

From these resting-state studies, a number of novel insights can be deduced that could not be easily anticipated from animal studies. Analyzing the proportion of the cerebellum dedicated to each network, it is remarkable that only 20%–30% of the cerebellar territory falls into what might be considered core motor networks. In contrast, fronto-parietal cortical networks (Figure 5, networks 7, 8, 12, and 13) account for a disproportionally (compared with the neocortex) large part of the cerebellum (Marek et al., 2018), suggesting a prominent role of the cerebellum in higher-level aspects of action planning and cognitive control. Furthermore, the hemispheric aspect of lobule IX is consistently correlated with the precuneus and anterior cingulate (network 16, Figure 4), which form part of the default-mode network.

Based on the pattern of resting-state connectivity, Buckner et al. (2011) suggested a basic organizational principle by which the neocortex is "represented" three times on the cerebellar cortical sheet. Going from lobule IV to *crus* I, there appears to be a progression from motor (networks 3 and 4) to premotor (networks 6, 7, and 12) to prefrontal regions (networks 8, 13, 16, and 17). This sequence reverses across *crus* I to lobule VIII, with a second reversal evident in the representation of more cognitive networks in lobule IX. This triplicate organization has also been observed in task-related data (Guell et al., 2018a).

The idea of a three-fold cortical-to-cerebellar mapping, however, is likely overly simplistic. Although the repeated representation of cortical networks looks compelling on a parasagittal slice (Figure 6 in Buckner, 2013), the projection onto a flattened map of the cerebellum reveals a more complex organization (Figure 5). For example, two of the areas assigned to "network 12" are actually continuous along the paravermal cerebellum. The tertiary representation in lobule IX is much more limited and dominated by the default-mode network. Furthermore, the neocortex also shows an equally complex repetition of networks, with most networks consisting of a frontal, parietal, and sometimes medial component.

Instead of a three-fold mapping of the cortex onto the cerebellum, it is possible that each cerebellar area receives input from one and only one of the cortical components of each network. To take network 8 as an example, the lateral aspect of *crus* I might be dominated by input from the dorsolateral prefrontal cortex, whereas VIIb might be dominated by input from the medial or parietal component. Alternatively, each cerebellar region may combine inputs from these anatomically separate regions. Because of the strong correlations of the fMRI data among the cortical components of the same network, this question will not be easily answered using task-free fMRI.

Determining the exact pattern of connectivity has important consequences for understanding cerebellar function. If each cerebellar region is connected, in a reciprocal fashion, with only one cortical region (Kelly and Strick, 2003), then the focus of a computational hypothesis would have to be on how the cerebellar circuit modulates the local computations in a single
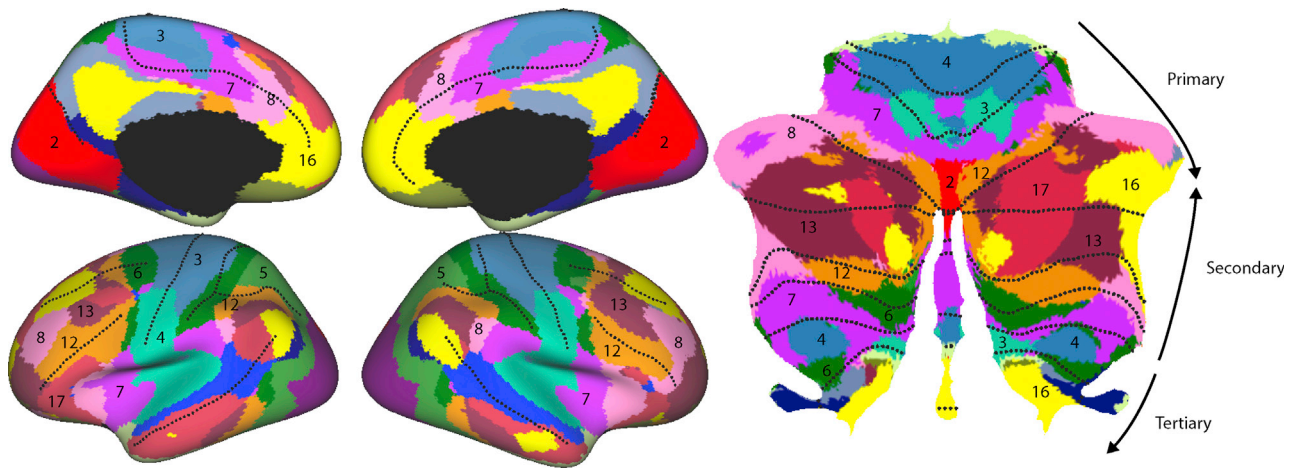
Figure 5. Functional Connectivity between the Neocortex and Cerebellar Cortex, Using Cortical Parcellation into 17 Regions
It has been proposed that the neocortex is "represented" 3 times in the cerebellum, with each representation running from motor to more cognitive regions (Buckner et al., 2011).

cortical region. In contrast, if it turns out that each cerebellar region integrates information from a combination of cortical regions, then theoretical accounts would need to consider how the cerebellum modulates or gates communication between these regions. It is also possible that the basic rules of connectivity vary across cerebellar regions. For example, cerebellar regions that project back to the neocortex through the thalamic nuclei ventral lateral nucleus (VL) and ventral posterolateral nucleus (VPL) may communicate with a focal cortical area, whereas cerebellar regions that project through the thalamic laminar nuclei may modulate the interactions of many regions (Gornati et al., 2018).

Summary and Outlook
We have reviewed convergent evidence that highlights the functional diversity of the human cerebellum. This diversity makes the formulation of a domain-general theory of cerebellar function, at best, very challenging. It is also important to keep in mind that an algorithmic account of cerebellar function may entail multiple computational concepts and that these may differ across domains, an idea we termed multiple functionality.

The relative merits of the universal transform and multiple functionality hypotheses will, in the end, be an empirical question. For now, we think there is considerable value in carefully developing hypotheses of cerebellar function for specific cognitive domains without being limited, *a priori*, by the assumption that the function is somehow analogous to those established for motor control. For example, most of our hypotheses and experiments in the sensorimotor domain focus on the role of the cerebellar circuit in the adult organism. However, in the cognitive domain, the cerebellum may play a more important role in development than in mature function (Badura et al., 2018). Furthermore, although damage to the cerebellum in adulthood frequently results in rather subtle symptoms on cognitive and affective measures (Alexander et al., 2012), the same damage in the developing brain may have much more profound conse-

quences. Thus, in cognitive and social domains, the cerebellum may help set up cortical circuitry during certain sensitive phases of development. When established, the cortical circuits may no longer require substantial cerebellum-based modulation. This hypothesis may be important for understanding why cerebellar dysfunction has been attributed to neuropsychiatric developmental disorders such as autism (Wang et al., 2014) and schizophrenia (Moberget et al., 2018), even though damage to the cerebellum in adulthood will not result in the symptoms associated with these disorders.

When exploring cerebellar function in each task domain, there are two critical issues that must be addressed. First, cerebellar activity should be studied in the context of the activity patterns in the cerebral cortex. In isolation, the study of cerebellar activity may lead to interesting punctuated insights, such as "the cerebellum represents reward" (Wagner et al., 2017). However, to gain a deeper understanding of cerebellar function, we need to compare cerebellar and cortical representations (Wagner et al., 2019). Do the pontine nuclei simply transmit information from the neocortex to the cerebellum in a non-selective manner, or are specific aspects of cortical representations emphasized and other aspects omitted? The circuitry in the pontine nuclei suggests that these subcortical nuclei can perform non-linear integration and gating of cortical input (Schwarz and Thier, 1999). Thus, the information reaching the cerebellum may differ in informative ways from the way it is represented in the neocortex.

Identifying these differences is likely to yield important insight into the role of the cerebellum. For example, if a cerebellar area is especially important in a specific phase of skill activation, then we would expect different activity time courses for the relevant cerebellar and cortical regions: disproportionately higher activity in early phases of learning when the cerebellum is involved in initial acquisition and disproportionately higher activity in later phases when it is important for the performance of automatized behaviors. To perform such experiments and analyses, a full

model of cortical-cerebellar connectivity is required, allowing the researcher to identify the relevant pairs of cortical and cerebellar regions.

Second, it will be important to understand what information is carried by the climbing fiber system. According to the Marr-Albus-Ito model, the climbing fiber input specifies the "learning goal" for the cerebellar circuit and, therefore, plays a pivotal role in shaping the output of the cerebellum. Although the climbing fiber input has traditionally been assumed to represent an error signal, new evidence suggests that it may be better conceptualized as a general teaching signal that may sometimes also relate to reward rather than error (Heffley et al., 2018). At present, we have virtually no insight concerning the information content of the climbing fiber system in the "cognitive" regions of the human cerebellum. Thus, we do not know what these cerebellar circuits are being instructed to learn. Understanding the learning goal (or cost function) will likely provide an important key to understanding cerebellar function in the domain of cognition.

In summary, careful investigation of cerebellar function within well-specified task domains will provide a clearer picture of the functional diversity of this major subcortical structure. Looking across domains, we may ultimately discover a universal cerebellar transform. It is likely, however, that this computation will not be easily captured in the functional terms we can intuitively describe: ideas such as timing, automatization, prediction, error correction, or internal models. Rather, a common principle may only emerge in terms of a more abstract language describing the population dynamics of neuronal networks.

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