1 **Coordinated cerebellar climbing fiber activity is gated by behavioral context in a**
2 **voluntary motor regime**
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Abstract

The prevailing model of cerebellar learning states that climbing fibers (CFs) are both driven by, and serve to correct erroneous motor output. However, this model is grounded largely in studies of reflexive behaviors where CFs respond with high fidelity to the same unconditioned stimuli that evoke erroneous movement. To test whether this model applies to more flexible learning regimes, we have developed a cerebellar-dependent voluntary motor learning paradigm compatible with both mesoscale and single dendrite resolution calcium imaging in mice. Here, we find that CFs are preferentially driven by, and more time-locked to correctly executed movements, exhibiting widespread correlated activity within parasagittal processing zones that depends on behavioral context. Together, these complimentary CF activity patterns are well-suited to drive learning based on successful movement, and suggest that classical supervised learning models must be extended to account for the context-dependent gating of olivary signaling independent from motor errors.

Introduction

A key role of the cerebellum is to form predictive associations between sensory inputs and motor outputs. These sensorimotor predictions are critical for generating well-timed and accurate movements, and in the absence of cerebellar function, the lack of such predictive motor output severely impairs our ability to generate coordinated responses to stimuli in the external world.

Classic models posit that the cerebellum generates sensorimotor predictions according to a supervised learning rule[1-3]. According to such models, projections from the inferior olive called climbing fibers are thought to signal motor errors, thus providing information to Purkinje cells about discrepancies between the expected consequences of a motor command and subsequent sensory feedback. To correct erroneous motor output, the climbing fibers instruct heterosynaptic long-term depression[4-6] by producing powerful regenerative calcium transients[7-9] in Purkinje cell dendrites called complex spikes[10, 11]. In so doing, climbing fibers are thought to appropriately update the cerebellar forward internal model with revised sensorimotor predictions.

This error-signaling framework provides a compelling explanation for climbing fiber activity in a variety of simple behaviors, such as classical conditioning (e.g. eyeblink conditioning) or adaptation (e.g. vestibulo-ocular reflex gain changes)[12-14] paradigms. These behaviors typically make use of an unconditioned stimulus that drives both climbing fibers and the motor response that requires modification. Hence, there is a yoked relationship between sensory input and motor output that allows the climbing fibers to respond with high fidelity to movement error according to the causal association between stimulus and movement.

However, many forms of motor learning do not involve modifications to motor programs
driven by an unconditioned stimulus. Instead, the correct association between sensory input and
motor output must be learned through experience. Behavioral regimes with such requirements
are typically classified under the domain of voluntary motor learning, and cerebellar dysfunction
is also known to impair this type of learning[15]. However, it is unclear whether the current model of
supervised error signaling can account for either climbing fiber activity or the cerebellar
contribution to learning in these more flexible regimes.

To test how the climbing fiber system is engaged under conditions where the sensory and
motor signals necessary to drive learning are not innate, we have established a cerebellar-
dependent behavioral paradigm compatible with population level *in vivo* calcium imaging,
optogenetic and electrophysiological approaches. Using this paradigm, we reveal two key
features of cerebellar climbing fiber (CF) driven complex spiking. First, we find that complex
spiking cannot be accounted for by a simple error-based supervised learning model. Instead,
complex spiking can signal learned, task specific information necessary for flexible control of
complex motor behaviors. Second, mesoscale population level recordings reveal that while
complex spiking is correlated within and across parasagittal zones, these correlations also
depend on behavioral context. While previous measurements have shown increased correlations
in complex spiking in response to sensory input or motor output[16-19], and have suggested an
important role for synchrony in downstream processing and motor learning[20], our results reveal
that such modulation can be contextual, and vary for identical movements depending on
behavioral relevance. Hence, these data reveal key features of cerebellar CF activity that differ
significantly from classically studied reflexive behaviors, and suggest an extension to current
models of complex spiking in order to account for the cerebellar contribution to voluntary motor
learning regimes.

Results:

To determine the conditions under which climbing fiber driven complex spiking occurs
during a voluntary motor learning regime, we first designed an appropriate sensorimotor learning
task. An important feature of the task design is to specifically engage neurons near the dorsal
surface of the cerebellum, thereby allowing for visualization of complex spikes at the population
level via calcium imaging. Hence, we used optogenetics to functionally map the dorsomedial
surface of the mouse cerebellum according to the motor output produced by spatially defined
populations of Purkinje cells (PCs). To do so, we expressed Archaerhodopsin (Arch) using a
transgenic approach (methods) in cerebellar PCs (Fig. 1A), and used an external fiber coupled
laser to transiently suppress PCs near the dorsal surface of the cerebellum (Fig. 1B). Consistent
with previous work[21], we identified a population of superficial PCs in lobule simplex capable of

driving ipsilateral arm movements either during optogenetic silencing with Arch or following a brief
train of electrical simulation (Fig. 1, Supp. Fig. 1, Supp. movie). *In vivo* single unit recording
confirmed that our optogenetic stimulation was effective in silencing superficial PCs at the
threshold for driving arm movement (~20 mW) up to a depth of 500 μm (Fig. 1B).
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A cerebellar dependent voluntary motor learning task
Based on these results, we designed a sensorimotor association task involving forelimb
movement, with the rationale that the cerebellar contribution to learning in such a task should
necessarily involve the PCs capable of driving such movements. This task contains two discrete
phases; one where motor output is reactive ('cue reaction'), and another where motor output
becomes predictive with learning ('cue prediction'). In both task variations, head restrained, water
deprived mice are trained to self-initiate trials by depressing a lever, and to release the lever in
response to a visual cue (Fig. 1C,D).
In the cue reaction task, trial-to-trial variability in the timing of the visual cue (Fig. 1E,G-
"random cue delay") is imposed to necessitate that mice employ reactive forelimb responses. In
this regime, lever release reaction times are reflective of the latency for sensory integration[22] and
remain constant throughout each session (ΔRT = 1.1 $\pm$ 9.3 ms, n=26 sessions, 7 mice).
In the cue prediction task, the visual cue is presented with the same delay on every trial.
In this regime, mice learn to predict the timing of the cue and adjust their motor responses to more
closely approximate the time of cue (Fig. 1F,G, ΔRT 500ms cue delay = 154.5 $\pm$ 13.6 ms, n=44
sessions, 10 mice). Moreover, learning occurs within a single training session (90 min., 300-600
trials), is retained across days (Supp. Fig 2A), and can be extinguished by returning mice to the
cue reaction paradigm (Supp. Fig 2C,D). In addition, we find that learning depends on the
duration of the cue delay: maximal learning occurs at short cue delays (e.g. 500 ms) and little or
no learning occurs for cue delays greater than 1.5 seconds (Fig. 1H; one-way ANOVA, main effect
of cue delay, p = 4.8×10^{-6}. Post hoc comparison to cue reaction: 500 ms, p=1.9×10^{-11}, 1 s,
p=8.4×10^{-11}, 1.5 s, p=0.359, 2 s, p=0.005). This dependence on a short interval delay is consistent
with the temporal characteristics of canonical cerebellar-dependent sensorimotor associative
learning[23].
To test whether animals adjust the timing or kinematics of their forelimb movement to
predict the timing of the visual cue, we measured lever kinematics during a subset of learning
sessions. Alignment of the lever trajectory to the onset of the cue revealed that animals altered
the time of lever release, releasing the lever sooner in anticipation of the visual cue (Fig. 1I; 20%
rise time occurs 116 $\pm$ 18 ms earlier, p = 1.2×10^{-5}, paired t-test). However, despite substantial
variability in the kinematics of release (Supp. Fig. 3), independent alignment of both press and
release demonstrated that there was no change in movement kinematics across learning

sessions (Fig 1I-J; First 1/3 of session 20-80% release rise time= 18.9 ± 0.6 ms, last 1/3 of session

= 19.3 ± 0.8 ms; p=0.513, paired t-test), indicating that mice release the lever sooner but with the

same kinematics after learning.

To test the necessity of the lobule simplex for this learning, we first locally applied lidocaine

in order to block spiking in cerebellar cortical neurons, including PCs. However, consistent with

the critical tonic inhibition PCs provide to the deep cerebellar nuclei, we observed significant motor

deficits including slowed movement, dramatically fewer initiated trials, and dystonic limb

contractions (not shown). Thus, this manipulation was not appropriate to define the necessity of

lobule simplex in our task.

To avoid significant impairment of motor output, we next used a pharmacological

manipulation to selectively disrupt excitatory synaptic transmission in lobule simplex without

abolishing the spontaneous pacemaking of PCs.   For these experiments, we optimized

concentrations of the excitatory synaptic transmission blockers NBQX (300 μM), CPP (30 μM)

and MCPG (30 μM) to avoid visible impairment of motor output, and then tested this cocktail

during behavior. Following local drug application to lobule simplex, mice retained the ability to

release the lever in response to the visual cue, maintaining stable kinematics across sessions

(Fig. 2A,B) that were not different from control sessions (Fig. 1I,J, and Supp. Fig. 3; Press: control

range = 28.97 ± 2.5 ms, NBQX range = 25.33 ± 1.9 ms, p=0.2612; Release: control range = 6.16

± 1.2 ms, NBQX range = 4.13 ± 1.06 ms, p=0.2215, unpaired t-test). These data suggest that

motor output was not significantly impaired as a result of this pharmacological manipulation.

However, despite stable movement kinematics, learning was severely impaired when the cue was

presented with a constant 500 ms delay (Fig 2A,C,D) (NBQX ΔRT with 500 ms cue delay = 48.8

± 26.7, control ΔRT with 500 ms cue delay = 154.5 ± 13.6 ms, ΔRT with cue reaction = 1.1 ± 9.3,

unpaired t-test NBQX vs control 500 ms p=4.6×10^{-4}, unpaired t-test NBQX vs control cue reaction

p=0.05). Single unit PC recordings during drug application revealed that this manipulation

significantly reduced complex spike rates, consistent with a reduction in the amplitude of

postsynaptic glutamatergic currents produced by climbing fiber input (control 0.60 ± 0.07 Hz,

NBQX/CPP/MCPG 0.27 ± 0.04 Hz, n = 9, p=8.5×10^{-4}, paired t-test, Supp. Fig 4). We also found

a modest reduction in simple spiking, consistent with reducing excitatory synaptic input from the

mossy fibers and parallel fibers (control 66.8 ± 17.8 Hz, NBQX/CPP/MCPG 45.9 ± 13.0 Hz, n =

8, n = 9, p=0.36, paired t-test). Hence, these data support that hypothesis that synaptic

transmission in lobule simplex is necessary for predictive sensorimotor learning to occur in the

cue reaction phase of our task.

Complex spiking is not driven by motor errors in a voluntary motor learning task

To test how the olivary system is engaged in this behavior, we imaged climbing fiber
evoked calcium signals in mice performing the cue reaction phase of our task. In this regime, all
sensorimotor signals are in place to drive learned, predictively timed forelimb movements, but no
learning can occur due to the variable cue timing (Fig. 1G,H). Hence, we can test what task
features drive complex spiking in a stable regime. We began by using a mesoscale, single photon
microscope to image populations of PCs virally expressing GCaMP6f (methods). Image
segmentation in a subset of experiments with the most superficial GCaMP expression revealed
structures consistent in size and shape with PC dendrites in cross section (Supp. Fig. 5). Calcium
transients in these structures occurred at 0.5 ± 0.01 Hz, consistent with the baseline rate of
complex spiking we measured with single unit electrophysiology (Supp. Fig. 5). These results are
also consistent with several *in vivo* calcium imaging studies using a variety of GCaMP variants
showing that PC responses are dominated by the large dendritic calcium transients produced by
climbing fiber driven complex spikes[16-19, 24-29].

We next identified regions of interest, defined by the expression pattern of GCaMP in each
animal, that were restricted to the region within lobule simplex capable of driving ipsilateral arm
movements (methods). Consistent with the role of this region in the control of arm movements,
we observed task modulated PC calcium transients associated with the timing of lever release
(Fig. 3A-B). Surprisingly, when we segregated trials according to outcome, we observed
significantly larger calcium transients associated with lever releases when animals correctly
released the lever in response to the visual cue as compared to trials where lever release occurred
prior to the cue (Correct = 0.156 ± 0.023 ΔF/F; Early = 0.092 ± 0.018 ΔF/F, n=17 sessions, 10
animals, p=4.60×10^{-5}, paired t-test). These enhanced responses were distributed widely across
dorsal, superficial lobule simplex, and were consistent across sessions and animals (Fig. 3C).
Thus, in this voluntary motor task, our data suggest that complex spikes preferentially encode
correctly timed movements, and not motor errors.

There are many differences between the correct and early release trials that could
contribute to the observed outcome-dependent difference in complex spiking. One possible
difference between trial types could be the kinematics of the forelimb movement. Hence, in a
subset of our experiments, we measured movement kinematics by comparing lever trajectories
on correct and early release trials (Fig. 4A-B). These data, however, revealed closely matched
movement kinematics between correct and early lever release trials (Press: correct = 26.8 ± 1.6
ms, early = 25.1 ± 1.3 ms; Release: correct = 18.9 ± 0.8 ms, early: 22.1 ± 1.5 ms, n=14 sessions).

Another difference between correctly timed and early lever releases is that reward is only
delivered on correct trials, and this results in extended licking. Indeed, we find that lick rates were
significantly higher and prolonged on correct trials (Fig. 4C; correct = 10.5 ± 0.7 Hz, early = $1.5 \pm$
0.4 Hz, p=3.74×10^{-9}, paired t-test, n=9 animals, 15 sessions). Moreover, in some recording

sessions we observed a late, sustained calcium transient that occurred during licking (Fig. 4D).

However, we found no correlation between the early, peak calcium transients and lick rates across

trial types (Fig. 4E,F; Correct R^2=0.182, p=0.11; Early R^2=5.97x10^{-5}, p=0.98, ANOVA). Thus, we

conclude that the enhanced responses as measured by the early peak calcium transient on

correctly timed lever releases are not due to differences in arm movement kinematics or licking,

but are instead linked to the context of the movement.

Behavioral context gates the probability and synchrony of complex spiking

To investigate the dynamics of complex spiking at the single cell level, we used video-rate

two-photon microscopy to image the dendrites of PCs during behavior. Using this approach, we

isolated the dendrites of individual PCs [25] (Fig.5, Supp. Fig. 5), and measured both spontaneous

and task-evoked calcium transients. As with the single photon imaging experiments, spontaneous

calcium transients measured during inter-trial intervals occurred at a rate of 0.6 $\pm$ 0.01 Hz,

consistent with our electrophysiological measurements of spontaneous complex spiking (Supp.

Fig. 5E,F). The two-photon data also replicated the main finding that the mean calcium transients

were significantly larger on correct release trials within individual PC dendrites (Correct = 0.110 $\pm$

0.004 ΔF/F; Early = 0.032 $\pm$ 0.002 ΔF/F, n=962, p=1.65^{-75}, paired t-test) (Fig. 5A,B). Moreover,

lever releases that were rewarded but occurred too fast to be responses to the visual cue (reaction

times less than 200 ms) produced calcium transients that were significantly smaller than those

produced by correct lever release trials ('Too fast' 0.028 $\pm$ 0.002 ΔF/F, p=2.78^{-100}, paired t-test)

and not significantly different from responses on early release trials (p=0.134). Hence, these data

support the analysis in Figure 4 that the enhanced complex spiking on correct trials is not due to

licking or reward per se.

In addition to larger mean responses across trials, population level analysis revealed that

more PCs were driven on correct release trials (Fig 5C). Of the total population of PCs that

responded to any lever press or release (962 of 1051), most (87%) PCs exhibited complex spikes

on correct release trials, whereas less than half (39%) responded during early lever releases (Fig

5D). Approximately half (49%) of PCs exhibited complex spikes in response to lever press. The

increased number of active dendrites on correct trials is unlikely to be a thresholding artifact, since

there was no significant difference in the average response amplitude of correct-only and

correct+early responsive dendrites to correct trials (Correct only 0.126 $\pm$ 0.006 ΔF/F, n=489;

Correct + Early 0.123 $\pm$ 0.007 ΔF/F, n=324, p=0.8 Student's t-test). These data also indicate that

the measured increase in number of active dendrites is not due to contamination of fluorescent

signals from neighboring dendrites. Hence, these single dendrite measurements explain our

single photon results by demonstrating that both more PCs exhibit complex spikes on correct

trials, and those dendrites that exhibit complex spikes in response to both trial types have larger mean calcium transients on correct trials.

The difference in the average release-evoked response on correct versus early trials may be due to a difference in the amplitude of the calcium transients on each trial. For instance, complex spikes are not "all-or-none" events, and the number of spikelets can vary from event to event[30-32], altering the size of the evoked calcium transient[33]. Thus, we extracted single trial events associated with lever releases across trial types, and compared these with each other and to the spontaneous calcium transients. However, this analysis revealed no significant difference between the amplitude of calcium transients evoked on single trials in response to either correct or incorrect lever releases (Correct 0.279 ± 0.005 ΔF/F n=15,586, Early 0.262 ± 0.005 ΔF/F n=7,705), and between movement-evoked and spontaneous activity (Spontaneous 0.296 ± 0.004 ΔF/F n=239,904) (p=0.3, one-way ANOVA) (Fig. 5E). Thus, if there are variable numbers of evoked spikelets across trials, such differences are not systematically correlated with either trial outcome or task engagement. Moreover, the equivalent amplitude of lever-evoked and spontaneous events suggests that the majority of the measured signal is from calcium influx due to complex spiking, and is not substantially influenced by additional input from the parallel fibers. Indeed, the size and shape of our segmented ROIs is consistent with several studies showing that complex spikes produce a calcium signal throughout the entire PC dendrite[7, 9], whereas parallel fiber calcium signals are much smaller and confined to isolated regions of spiny dendritic branchlets[34-37].

To assess how equivalent single event responses can produce larger mean responses on correct trials, we generated a peristimulus time histogram of complex spike events for each trial type aligned to the time of release (Fig. 6A). This analysis revealed a significantly enhanced complex spike rate at the time of lever release for trials when movement was correctly timed (Fig 6A,B; correct spike rate=2.31 ± 0.01 Hz, early spike rate= 1.39 ± 0.01 Hz, p=4.54x10^{-38}, n=962, paired t-test). In addition, complex spiking on correct trials in single PC dendrites occurred with shorter latency (Fig 6C; correct latency=30.44 ± 0.98 ms, early latency=45.72 ± 1.02 ms, p=6.37x10^{-27}, n=962, paired t-test) and less temporal jitter across trials (Fig 6D; correct jitter (S.D of latency) =65.91 ± 1.75 S.D., n=2140, early jitter 93.29 ± 0.61 S.D., n=1195, p=1.07x10^{-31}, Student's t-test). These results also suggest the possibility of enhanced synchrony at the population level on correct release trials. Indeed, in some sessions, we measured enhanced population synchrony across dendrites (Fig 6E; 5 out of 20 sessions have significantly elevated synchrony on correct trials, open circles, p=5.11x10^{-3}, Mean across the population is not significant, p=0.1305, paired t-test.). However, in many experiments, we observed no such increase in population synchrony on correct trials. We thus considered the possibility that

enhanced synchrony was location specific, and might obey spatial structure that was not well

demarcated at the scale of the field of view in our two-photon experiments.

Context determines correlated population activity at the mesoscale level

To investigate whether correlated complex spiking was manifest at the mesoscale, we

returned to single photon imaging and analyzed experiments that exhibited widespread GCaMP

expression across superficial lobule simplex (n=5 animals, 7 sessions). Using an unsupervised,

iterative pixel-clustering approach[27] (methods), we first assessed correlations between pixels

across all trials (Fig. 7A). Despite non-uniform, unpatterned GCaMP expression, this analysis

revealed spatial patterns of correlated activity that were organized across parasagittal bands

oriented in the rostro-caudal axis. These bands were 221 ± 15 μm wide on average across 39

measured zones, consistent with previous anatomical and physiological measurements of

cerebellar microzones[38, 39]. Hence, these data support the longstanding hypothesis that the

anatomical pattern of climbing fiber projections into the cerebellar cortex establishes functionally

distinct parasagittal processing modules.

To test how complex spiking is modulated during behavior within and across parasagittal

zones, we divided trials according to outcome and analyzed brief epochs surrounding the time of

lever release (Fig. 7B). This analysis revealed that activity amongst pixels within a zone exhibited

higher correlations for most zones on correct lever release trials as compared to early release

trials ($r_{correct}$ = 0.77 $\pm$ 0.01, r_{early} = 0.69 $\pm$ 0.017, p = 1.60x10^{-5}, n=39 clusters, paired t-test) (Fig.

7C). Because higher spike rates can cause larger correlations, we tested whether the higher

complex spike rates on correct trials could account for the higher correlations. While increased

complex spiking did produce larger correlations, early release trials still had weaker correlations

than correct release trials across equivalent activity levels as assessed by linear fits to

bootstrapped distributions paired from each trial type (methods) (Fig. 7D). Hence, the enhanced

correlations on correct trials are not an artifact of enhanced complex spike rates on correct trials.

We also found enhanced correlations between neighboring and non-neighboring zones on correct

lever release trials (Fig. 7E,F). However, these cross-zone correlations were significantly lower

on average than those within zones ($r_{neighboring}$ = 0.69 $\pm$ 0.02, n=31 cluster pairs, p=0.004, r_{non-}

$_{neighboring}$ = 0.64 $\pm$ 0.06, n=50 cluster pairs, p=3.88x10^{-4}, Student's t-test). Hence, these data reveal

a precise spatial and temporal organization of climbing fiber activity, with highly correlated

complex spiking within parasagittal zones that is task specific, differing for movements with the

same kinematics depending on behavioral context.

Discussion

We have established a cerebellar-dependent sensorimotor association task that involves
PCs near the dorsal surface of lobule simplex. This behavior has several hallmarks of cerebellar
learning, including dependence on short delay intervals for establishing the learned sensorimotor
association and the requirement for excitatory synaptic transmission in the cerebellar cortex.
However, this behavior also differs from most cerebellar dependent learning regimes in that the
movement requiring modification is not driven by an unconditioned stimulus. In our regime, the
proper association between a neutral visual stimulus and movement must be learned before a
cerebellar-dependent sensorimotor timing prediction can be established.   By breaking the
relationship between the specific motor output involved in learning and the unconditioned
stimulus, we have revealed patterns of climbing fiber driven complex spiking that are not
consistent with the traditional model of error-based supervised learning.
In reflexive associative learning paradigms such as conditioned eye-blink learning, the
inferior olive is activated by unconditioned sensory input, even in naïve animals. However, in our
regime, the cerebellar contribution to motor learning has different requirements, first necessitating
that the olivary system establish responses to instructive sensorimotor input that previously had
no behavioral relevance, and then utilizing these responses to guide changes in motor output. An
important question arising from this finding is whether the initial learning that produces enhanced
complex spiking for correctly timed movements occurs at the level of the olive, or if it is inherited
from upstream brain regions such as the neocortex, colliculus, or elsewhere. Hence, future
experiments designed to test how olivary neurons can flexibly represent learned, task-specific
sensorimotor information will be necessary to reveal the origin of such initial learning.
Revealing how climbing fiber activity is learned may also reveal how it is gated. Previous
work has demonstrated that climbing fiber activity can be gated during different phases of
movement[40]. In particular, olivary neurons show reduced excitability during discrete stages of
reaching[41, 42] and locomotion[43-45], and this phenomena has been proposed to involve either
nucleo-olivary inhibition[46], or an inhibitory pathway from the external cuneate nucleus that can be
activated by the red nucleus[47, 48]. Our results demonstrate a new form of gating, where the same
movement produces differential climbing fiber activity depending on behavioral context. While
our data do not distinguish whether complex spiking is actively enhanced during correct release
trials or suppressed during incorrect lever releases, our expectation is that learning the initial task
contingencies establishes an association whereby the cued movement (either through sensory or
reward-related circuits) can drive enhanced climbing fiber activity. In turn, both the higher
probability of complex spiking and enhanced correlation at the population level are ideally suited
to facilitate context-dependent motor learning.
At the population level, our mesoscale imaging of cerebellar complex spiking has revealed
correlated activity within parasagittal bands of approximately 200 μm. This result provides an

expanded view of climbing fiber activity, and is consistent with several studies indicating
enhanced synchrony in complex spiking across much smaller areas for PCs oriented in the
parasagittal axis[27, 49-51]. As previously mapped by the anatomical projection pattern of climbing
fibers[52-54], electrophysiology[39, 55] and two photon imaging[17, 27, 56], these parasagittal zones are
typically 100-200 μm wide, and are commonly referred to as 'microzones.' Microzones are also
related to overlapping, sometimes larger parasagittal compartments defined by the expression of
molecular markers such as Aldolase C, the glutamate transporter EAAT4, and others[56-60].
Because PCs across microzones can respond to different sensorimotor input, and have different
spiking[61] and synaptic[62] properties, these zones are thought to constitute discrete computational
units in the cerebellum.

To date, it has remained challenging to test how cerebellar processing is organized within
and across microzones *in vivo* during behavior, in large part due to the spatial restrictions imposed
electrophysiological and multiphoton imaging approaches. Recent measurements have revealed
that within parasagittal bands consistent in size with microzones, complex spiking can become
more synchronous during both motor output[19] and sensory input[16, 17, 28]. These observations have
lent support to the idea that cerebellar processing is functionally segregated to enable
sensorimotor learning by clustering responses of like kind onto nearby PCs that in turn converge
onto the same deep cerebellar neurons. Our mesoscale imaging results provide a fundamental
extension to this idea by demonstrating that correlated complex spiking within parasagittal zones
is not simply enhanced by sensorimotor input, but rather can be enhanced for the same action
according to behavioral context. This feature is likely to be essential for the cerebellar climbing
fiber system to instruct learning in flexible, voluntary regimes where behavioral context is required
for accurate motor performance[63]. Precisely timed spiking across a large population of nearby
PCs that receive common parallel fiber input and converge to the same DCN neurons would
provide an ideal substrate to maximize the impact of plasticity. However, other notable models
could also play a role in learning. In particular, complex spiking can robustly inhibit nuclear
neurons[64, 65], and enhanced climbing fiber synchrony would magnify this effect via convergence
of simultaneously active PC axons in the DCN[66]. Because synchronous inhibition has been
shown to play a key role in synaptic plasticity at nuclear neuron synapses[67, 68], correlated complex
spiking across populations of nearby PCs could place the DCN as the central site of learning
under such conditions[69].

Finally, our data suggest that in certain regimes climbing fibers do not signal erroneous
motor output as described by classical models of supervised learning. An alternative to these
models has recently emerged, suggesting the possibility that climbing fibers may instead operate
according to different learning rule[70]. Specifically, conditioned eyeblink learning data has pointed
toward a reinforcement learning rule, where climbing fibers instead signal sensory prediction error

in order to instruct motor learning. Interpreted within this context, our results are consistent with

the possibility that correctly timed lever releases represent positive prediction errors for a

movement that accurately predicts future reward delivery. Hence, future experiments specifically

designed to test what learning rules are common across different cerebellar behavioral paradigms

will be necessary. For the present, the results described here provide an important new

demonstration that complex spiking can signal learned, task specific information necessary for

flexible control of complex motor behaviors in a manner that does not depend on motor errors.

Author Contributions

WH, ES, AM, MJ, and CH designed experiments, WH, ES, AM, BT, MAH, and MJ conducted

experiments, WH, ES, ZX, BT, AM, MJ, and CH analyzed data, and CH wrote the manuscript.

Figure Legends

Figure 1. Cerebellar sensorimotor task for head-fixed mice. Ai) Confocal image of a sagittal

section from PCP2-Cre x Arch mouse showing expression of the inhibitory opsin Arch in PCs at

low (left) and high (right) resolution. ML; molecular layer, PCL; Purkinje cell layer, GCL; granule

cell layer. Aii) Schematics of a cranial window implanted over lobule simplex (left) and

configuration for optogenetic stimulation (right). Bi) *Left*, representative single unit recording of

PC simple spike raster (top) and spike rate histogram (bottom) aligned to onset of Arch activation.

Right, Location of the stimulation sites that evoked ipsilateral arm movements (green). Bii)

Average normalized firing rates from superficial (top) or deep (bottom) PCs at low (left) or high

(right) stimulation powers (n=5 mice; 10-20 mW – superficial: n=6 cells; deep: n=5 cells; 30-50

mW – superficial: n=10 cells; deep: n=4 cells). Note that the lower powers used to map motor

responses were only sufficient to strongly suppress neurons above 500 μm. C) Schematic of

configuration of behavioral task. Head-fixed mice were trained to release a lever in response to a

visual cue. D) Schematic of trial structure. The delay between lever press and release cue was

either randomized (cue reaction) or constant (cue prediction) from trial to trial. E) Average

reaction time (from cue onset) as a function of trial number for an example cue reaction session.

Each point is the average of 5 trials. F) Same as E, for a cue prediction session. G) Summary of

reaction times for all cue prediction sessions with a 0.5 s cue delay (gray; n=44 sessions, 10 mice)

and all cue reaction sessions (black; n=26 sessions, 7 mice). Error bars are $\pm$SEM across

sessions. H) Average change in reaction time from the beginning to the end of cue prediction and

reaction sessions (methods). I) *Top*, average lever kinematic traces from an example 0.5 s cue

prediction session aligned to the cue show that lever releases occur sooner in the last 1/3 of trials

(black) as compared to the first 1/3 of trials (gray). Shaded area is ±SEM across trials. *Bottom,*

same example session as above aligned separately to press and release showing no difference

in kinematics from beginning (gray) to end (black) of the session. J) Summary of average lever

release times (20-80% rise time) across sessions (n=15 sessions, 4 mice)

Figure 2. Learning requires synaptic transmission in lobule simplex. A) Average lever kinematic

traces from an example 0.5 s cue prediction session after application of NBQX (300 μM), CPP

(30 μM) and MCPG (30 μM). *Top*, traces aligned separately to press and release showing no

difference in kinematics from beginning (blue) to end (black) of the session. *Bottom*, traces

aligned to the cue show that lever releases do not occur sooner in the last 1/3 of trials (black) as

compared to the first 1/3 of trials (blue). Shaded area is ±SEM across trials. B) Summary of

average lever release times (20-80% rise time) across NBQX 0.5s cue prediction sessions (n=10

sessions, 6 mice). C) Summary of all 0.5 s cue prediction sessions where NBQX (n=14) CPP

(n=14) and MCPG (n=8) were applied locally to LS (blue; n=6 mice) compared to summary of cue

reaction sessions (black; replotted from Fig 1G). D) Average change in reaction time from the

beginning to the end of NBQX 0.5s cue prediction sessions (blue; n=10 sessions, 6 mice)

compared to cue prediction and reaction sessions (replotted from Fig. 1H)

Figure 3. Mesoscale single photon imaging during cue reaction sessions. A) Field of view

(images separated by 100 ms) showing the average fractional change in fluorescence aligned to

lever release for correct (top) and early (bottom) trials from an example session. B) Average

timecourse of fluorescent changes from the experiment in A for correct (left, black) and early trials

(left, red), and the subtraction of early from correct trials (right). Error bars are ±SEM across trials.

C) Summary of the average peak fluorescent transients for all correct and early trials across

sessions (n=10 animals, 17 sessions). Colors represent different animals, and each point

represents an imaging session. Note that all points lie below the diagonal. *Inset*, schematic of

ROIs for all sessions. Outline color corresponds to mouse from scatter plot, and fill color indicates

imaging depth identified by post-hoc histology.

Figure 4. Lever dynamics and licking do not explain differences in complex spiking across trial

types. A) Average lever kinematics from a representative cue reaction session for correct (black)

and early (red) trials aligned to threshold crossing for both press and release. B) Summary of rise

times for press (left) and release (right) on correct and early trials across sessions (n=14 sessions,

3 mice). C) Average timecourse of licking for correct (top) and early (bottom) release trials across

sessions (n=15 sessions, 9 mice). Error bars are ±SEM across sessions. D) Two example

sessions illustrating the relationship between release-evoked calcium transients (black and red

lines) and licking (blue bars) for correct (left) and early (right) release trials. E) Summary of the relationship between lick rates and the magnitude of fluorescent transients for correct (black) and early (red) release trials (n= 15 sessions, 9 mice). Error bars are ±SEM across trials. F) Summary of the mean ratio of early and correct trial-evoked peak calcium transient and lick rates for the same sessions in E.

Figure 5. Complex spiking produces larger mean response in individual dendrites and enhanced population responses when movement is correctly timed. A) *Left*, average GCaMP fluorescence from an example 2-photon imaging session (30 Hz acquisition rate). Scale bar = 100 μm. *Middle*, pixel mask of individual PC dendrites extracted from the same session. Right, average calcium responses measured from the highlighted dendrite segregated by trial event. Dotted line represents time of lever press (blue) or lever release (correct: black; early: red). Shaded area is ±SEM across trials. B) Left, average timecourse of calcium response for all significantly responsive dendrites aligned to lever press (blue), or release (correct releases (black), early releases (red), and too-fast releases (green) that occurred < 200 ms following the visual cue; n= 11 animals, 20 sessions, 962 dendrites). Shaded area is ±SEM across dendrites. *Middle*, summary scatter plot comparing the average peak amplitude of the calcium transient for each significantly responsive PC dendrite (gray) and the average of all dendrites (red) for correct and early release trials. *Right*, same as middle for correct and too fast release trials. C) Pixel masks of dendrites significantly responsive to lever release on correct trials, lever release on early trials, and lever press for the example session in A. Color map represents average ΔF/F. D) Summary across experiments of the fraction of dendrites responsive to correct lever releases vs early releases (top, left), to only correct or only early releases (top, right), to lever press vs release (bottom, left), and only to press or release (bottom, right). E) *Left,* average amplitude of correct and early release-evoked calcium events across all dendrites normalized to the amplitude of spontaneous calcium events. *Right,* mean spontaneous (gray) and correct (black) and early (red) release lever-evoked calcium events across all responsive dendrites Error bars are ±SEM across cells.

Figure 6. Complex spiking occurs with higher peak rates and greater synchrony when movements are correctly timed. A) Peri-release time histogram of calcium events on correct (black) and early (red) release trials (n= 20 sessions, 11 mice, 962 dendrites). Shaded area is ±SEM across dendrites. B) Summary of average event rate for correct and early releases across sessions. Error bars are ±SEM across dendrites. C) Same as B for average event latency relative to release. D) Summary of average standard deviation (S.D.) across events within single dendrites across trials. Error bars are ±SEM across dendrites. E) Same as D for measurement of S.D. across

dendrites within each experiment. Open circles are statistically significant for reduced correct trial

jitter across dendrites. Error bars are ±SEM across trials.

Figure 7. Complex spiking is correlated across parasagittal zones, with higher correlations on

correct lever releases. A) *Left*, Average GCaMP fluorescence for three example experiments.

ROIs from lobule simplex (white boxes) were selected for unsupervised clustering analysis (meta

K-means, methods). *Right*, voxels are colored according to cluster identity for each example

experiment. M=medial, R=rostral, C=caudal, L=lateral. Note the alignment of the clustered zones

along the rostro-caudal axis. B) Within cluster voxel-by-voxel correlations between correct and

early lever release trials from an example session in A (bottom) for each of the six identified

clusters (yellow). C) Summary of average correlation coefficient for all voxel pairs within an area

across experiments for correct and early release trials. (n=5 animals, 7 sessions, 39 clusters).

Colors denote clusters from the same session. Error bars are ±SEM across voxel pairs. D)

Summary of the relationship between peak $\Delta F/F$ and correlation coefficients for correct (black)

and early (red) trials. Lines are linear fits to each cluster. E) Same as C for pairs of voxels across

neighboring clusters. F) Same as C for pairs of voxels across non-neighboring clusters.

Supplemental Figures

Figure 1. Mapping of motor output driven by Arch and electrical stimulation of superficial

cerebellar cortex A) Schematics of the results of Arch stimulation over a grid of sites in lobule

simplex (Sim) and Crus I for all five mice. Colors denote outcome of stimulation. B) Same as A

for three mice tested with electrical microstimulation.

Figure 2. Retention and extinguishment of learning for cue prediction regime. A) Average

reaction times relative to cue for the first (left) and second (right) of sessions where animals

performed 0.5 s cue prediction on consecutive training days (n=17 sessions, 5 mice). Error bars

are ±SEM across sessions. B) Same as A for 2 s cue prediction sessions (n=10 sessions, 6

mice). C) *Left,* example cumulative distributions of reaction times for cue reaction sessions

immediately surrounding a cue prediction session (session 0). Trials from the cue prediction

session include only the last 1/3 of the session when the animal was anticipating the cue timing.

Right, summary of distribution skewness across cue reaction sessions immediately surrounding

cue prediction sessions. D) Same as A for pairs of 0.5 s prediction sessions separated by 3 or

more cue reaction sessions (n=20 sessions, 10 mice).

Figure 3. Performance on cue prediction task in control and NBQX sessions. A-E) Summary across experiments of average percent correct lever releases, baseline reaction time from the visual cue, mean reaction time variance, mean duration to press, number of trials performed, and the duration of behavioral sessions. n=44 0.5 s, n=9 1.0 s, n=6 1.5 s, n=30 2.0 s, n=26 Δt, n=14 0.5 s NBQX F_i) Individual (top) and average (bottom) lever trajectories aligned to lever press (left) and release(right) segregated by duration into quartiles for a representative 0.5 s cue prediction session. F_{ii}) Same as F_i for a representative 0.5 s cue prediction session in NBQX. G) Summary of press (left) and release (right) quartiles across experiments for control (top; n=15 sessions) and NBQX (bottom; n=14 sessions) experiments. H) Summary of the ratio between press and release ranges (difference between 1st and 4th quartile) for control (black) and NBQX (blue) experiments.

Figure 4. Single unit recordings during NBQX application to LS. Top, summary of complex spike rates in control and after application of NBQX for individual cells (black; n=9 cells, 5 mice) and the average of all cells (red). Error bars are $\pm$SEM across cells. Bottom, same as top for simple spike rates. Note that NBQX strongly reduces complex but not simple spike rates.

Figure 5. Segmentation of PC dendrites from imaging experiments. A) Example mesoscale imaging field of view (left) is segmented to mask individual dendrites (right) used for subsequent single cell analysis. Scale bar = 100 μm. B) Example timecourse of raw fluorescence (top) from the circled dendrite in A showing individual calcium transients identified (gray circles) according to peaks in the first derivative of the fluorescence trace (bottom). C) Same as A for an example two photon field of view. Note that while PC somata are visible (round, top right), image segmentation does not extract activity from these structures. Scale bar = 100 μm. D) Same as B for the experiment in C. E) Top, example time course of a single unit PC recording from an awake mouse illustrating detection of complex spikes (gray circles). Bottom, overlay of individual simple (left) and complex (right) spike waveforms (gray) and the average waveform (black). F) Summary of spike rates across all two photon dendrite imaging sessions (blue; n=1051 dendrites) and single unit recordings (n=11 units).

Methods:

Mice

All experimental procedures using animals were carried out with the approval of the Duke University Animal Care and Use Committee. All experiments were performed during light cycle using adult mice (>p60) of both sexes, randomly selected from breeding litters. All mice were housed in a vivarium with normal light/dark cycles in cages with 1-5 mice. Imaging experiments

were performed using Tg(PCP2-Cre)3555Jdhu mice (Jackson Labs, 010536; n=12). Optogenetic mapping experiments were conducted in PCP2-Cre animals crossed with Ai35(RCL-Arch/GFP) (Jax 012735; n=5 mice). Behavioral experiments where imaging was not performed were conducted in wild type C57BL/6J mice (n=18). We used two exclusion criteria for animals in this study: (1) poor recovery or other health concerns following surgical intervention or (2) missed virus injection, as determined by *in vivo* imaging and post-hoc histological analysis.

Surgical Procedures

3-10 hours prior to surgery, animals received dexamethasone (3mg/kg) and ketoprofen (5mg/kg). Surgical procedures were performed under anesthesia, using an initial dose of ketamine/xylazine (50mg/kg and 5mg/kg) 5 minutes prior to surgery and sustained during surgery with 1.0-2.0% isoflurane. Toe pinches and breathing were used to monitor anesthesia levels throughout surgeries. Body temperature was maintained using a heating pad (TC-1000 CWE Inc.). Custom-made titanium headplates (HE Parmer) were secured to the skull using Metabond (Parkell). For imaging experiments, a 3mm diameter craniotomy was made over the lobule simplex approximately 1.4mm lateral and 2.8mm posterior to lambda, and glass cover slips consisting of two 3mm bonded to a 5mm coverslip (Warner Instruments No. 1) with index matched adhesive (Norland No. 1) were secured in the craniotomy using Metabond. Buprenex (0.05mg/kg) and cefazolin (50mg/kg) were administered following surgery twice a day for two days. Following a minimum of 4 recovery days, animals were water deprived for 3 days, or until body weight stabilized at 85% of initial weight, and were habituated to head restraint (3-5 days) prior to behavioral training.

For imaging experiments, the glass cover slip was removed following behavioral training, and mice were injected (WPI UMP3) with AAV1.CAG.Flex.GCaMP6f.WPRE.SV40 (UPenn vector core, titer = 9.40×10^{12} or 7.60×10^{12}). 150 nL virus diluted 1:1-1:5 in ACSF was injected at a rate of 30nl/min and a depth of 150 μm at 1-3 sites in dorsal lobule simplex. Imaging was performed beginning 14 days following injection.

For in vivo pharmacology experiments, a 3mm craniotomy was performed over lobule simplex of headposted mice, and the dura mater was peeled back at the center of the craniotomy. 10uL of combined NBQX (300 μM) and CPP (30 μM) was applied 20 minutes prior to behavioral training. In 8 of 14 experiments, MCPG (30 μM) was also included to block metabotropic glutamate receptors. Because there no significant differences in performance across these two groups, all data were pooled for analysis. During behavior, a second drug application of 10uL was administered 20 and 40 minutes into the task, and depending on the duration of training, a third dose was applied after 1 hour. Craniotomies were subsequently covered with silicone elastomer (WPI, Inc.) prior to returning animals to their home cage.

Behavior

During behavioral training, animals were head-fixed and placed in front of a computer monitor, lever and reward delivery tube. Animals were trained to self-initiate trials by depressing the lever using their right forepaw, and required hold the lever in the down position for up to 5 s on the cue reaction paradigm before performing the cue prediction paradigm. Lever releases within one second of the release cue were rewarded (0.01 M saccharine). Animals used for imaging experiments performed the cue reaction task with a mean peak percent correct of 80.1 $\pm$ 2.6%. Behavioral parameters including lever press, lever release and licking were monitored using Mworks (http://mworks-project.org) and custom software written in MATLAB (Mathworks). Licking was measured with electrical contact circuit, and the visual cue was presented as a 90° orientation change from a high contrast vertical grating. Inter-trial intervals ranged from 3 to 5 seconds.

Calcium Imaging

Wide-field imaging: Single photon, mesoscale imaging was performed using a customized microscope (Sutter SOM) affixed with a 5x objective (Mitutoyo, 0.14NA) and CMOS camera (Qimaging, Rolera em-c^2). Data were collected at a frame rate of 10Hz, with a field of view of 3.5x3.5mm at 1002x1004 pixels.

Two-Photon Imaging: Two-photon imaging was performed with a resonant scanning microscope (Neurolabware) using a 16x water immersion objective (Nikon CFI75 LWD 16xW 0.80NA). Imaging was performed using a polymer to stabilize the immersion solution (MakingCosmetics, 0.4% Carbomer 940). A Ti:Sapphire laser tuned to 920nm (Spectra Physics, Mai Tai eHP DeepSee) was raster scanned via a resonant galvanometer (8 kHz, Cambridge Technology) onto the brain at a frame rate of either 30Hz with a field of view of 250 by 500um and 264 by 796 pixels, or 15.5Hz with a field of view of 500 by 500um and 529 by 796 pixels. Data were collected through a green filter (510 $\pm$ 42 nm band filter (Semrock)) onto GaAsP photomultipliers (H10770B-40, Hamamatsu) using Scanbox software (Neurolabware). A total of 12 mice were used for imaging experiments (10 mice for wide-field, 11 mice for two-photon, 12 mice total).

Single Unit Recordings and Optogenetics

Acute single unit recordings were performed in awake animals by performing a craniotomy over lobule simplex and inserting a multi-electrode silicone probe (Neuronexus, A4x8-5mm-100-400-177-A32, 4 shanks, 8 site/shank at 100 μm spacing) using a Cerebus multichannel acquisition system (Blackrock Microsystems, Salt Lake City). Continuous recording data was bandpass filtered with a 2-pole Butterworth between 250 Hz and 5 kHz and referenced against an electrode

with no spikes using Spike2. Single units were isolated by amplitude thresholding. Template

waveforms were defined and characterized by their width and peak, and PCA of waveforms was

done in off-line in Spike2.

Complex spikes and simple spikes were discriminated as in de Solages et al.[71], first on

the basis of their stereotypical waveform. Complex spikes typically had a multi-wavelet form,

including a large positive peak within 6ms following spike initiation. Manual identification of 10-

20 complex spikes was used to generate a mean template waveform, which was then compared

to all other spikes for a given unit using the Spearman rank order correlation coefficient. The

combination of Spearman coefficients and the magnitude of the positive waveform deflection was

used to segregate complex and simple spikes. The presence of a post-complex spike pause (20

ms or more) in simple spike firing was verified by cross-correlogram for all isolated single units.

<ins>Histology</ins>

Mice were deeply anesthetized with ketamine/xylazine (200mg/kg & 30mg/kg respectively, IP)

then perfused with PBS then 4% paraformaldehyde. 100um sagittal sections were cut using a

vibratome (Pelco 102). Cerebellar slices were then mounted using a mounting medium (Southern

Biotech Fluoromount-G or DAPI Fluoromount-G) and imaged with a fluorescence microscope

(Nikon Eclipse 80i).

<ins>Data Analysis</ins>

Behavior: Behavior sessions were only analyzed within the time range of active task

performance. Accordingly, the last trial of each session was determined by the occurrence of

either two consecutive failed trials (in which the mouse did not release the lever in response to

the cue) or two consecutive trials with post-ITI duration to press longer than 1.5x the session

average.

Reaction times were measured using lever releases within the reward window as well as

those occurring 200ms prior to the reward cue to account for predictive responses. Reaction times

are plotted as a binned average of five trials. Initial reaction times were calculated according to

the y-intercept of the linear regression of the first fifty trials. Final reaction time at the end of each

session was calculated by averaging the reaction times of trials 200-250.

The symmetry of reaction time data about the sample mean was measured to test the

extinguishment of learning after following cue prediction sessions according to skewness.

Sample skewness was defined by

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$$s = \frac{E(\mu - X)^3}{\sigma^3}$$

679   where μ is the mean of X, $E(*)$ is the expected value of the quantity $*$, and σ is the standard

680   deviation of X. Thus for a given sample x, this calculates:

681
$$\frac{\frac{1}{n}\sum_{i=1}^{n}(x_i - \bar{x})^3}{\left(\sqrt{\frac{1}{n}\sum_{i=1}^{n}(x_i - \bar{x})^2}\right)^3}$$

for $x_i \in x$, where $\bar{x}$ is the sample mean.

Movement trajectories were measured using a lever affixed to a rotary encoder (US

Digital) with a 60mm radius and 1250 pulses per revolution. Encoder values were collected at

1000 Hz, and trajectories were calculated by up-sampling at 200 μs (5000 Hz) using nearest-

neighbor interpolation. Pulse number was converted to degrees, and the vertical displacement in

millimeters was calculated using the chord length of the leaver displacement angle. 20-80% rise

times were calculated using normalized average trajectories. Presses and releases were sorted

according to duration in a window from 200ms before threshold crossing to the time of a threshold

crossing half way between the top and bottom of the total lever displacement.

Wide-field imaging: Imaging analysis were performed using custom MATLAB code. Regions of

interest (ROIs) were selected within empirically defined arm movement region of lobule simplex

according to the spread of GCaMP expression. Window location and the lobule identity were

identified according to folia patterns visible through the cranial window, landmarks recorded

during surgery, and post-hoc histology. Baseline fluorescence (F) was measured on a trial-by-trial

basis during the inter-trial interval (ITI) as the mean fluorescence 900 to 200ms before trial

initiation. Normalized fluorescence (ΔF/F) was calculated according to the cumulative activity

within an ROI, or on a pixel by pixel basis. Lever hold times <200ms in total or > 1 s after the

visual cue were rare, and thus excluded from analysis. Lever releases between 175 and 1000ms

following the visual cue were classified as correct, and releases prior to visual cue were classified

as early. Lever releases $\leq$200 ms following the visual cue were considered too fast to be reactions

to the cue based on average reaction time distributions. Peak ΔF/F was measured in the time

window from 100ms before to 400 ms after lever release.

Meta K-means analysis: Images were first registered to reduce motion artifacts, and then

thresholded at 70% of maximum intensity for each frame to remove background noise. Images

were downsampled 5-fold in both X and Y. Baseline F was defined as the averaged fluorescence

across the entire movie for each pixel, and used to normalize change in fluorescence (ΔF/F). ΔF/F

was then re-normalized to the maximum ΔF/F during the entire movie for each pixel. The repeated

k-means clustering algorithm (meta-k-means) separated pixels to cluster centroids based on a

pairwise correlation distance function using Pearson's linear correlation coefficient (r). The final

clusters were determined by first thresholding all the k-means results at 800 out of 1000 runs, and

then by merging highly correlated clusters based on Dunn's index. Clusters occupying less than

3% of the total imaging field were excluded from further analysis. Intra-cluster and inter-cluster

correlation coefficients were calculated between 100 ms before and 300 ms after lever release

for both correct and early trials. To test for differences in the relationship between spike rates and

correlation coefficients between trial types, paired clusters from each distribution were resampled

with replacement and fit with a line to measure the y-intercept 1000 times. Statistical significance

was computed according to the 95% confidence interval of the distribution of y-intercept difference

values (difference between y-intercepts 0.0375 ± 0.0013, 95% confidence interval [0.0349,

0.0402]).

Two-photon imaging: Motion in the XY plane was corrected by sub-pixel image registration. To

isolate signals from individual PC dendrites, we utilized principal component analysis (PCA)

followed by independent component analysis (ICA). Final dendrite segmentation was achieved

by thresholding the smoothed spatial filters from ICA. A binary mask was created by combining

highly correlated pixels (correlation coefficient > 0.8) and removing any overlapped regions

between segmented dendrites. Fluorescence changes (ΔF) were normalized to a window of

baseline fluorescence (F) between 500ms and 100ms preceding trial initiation. Responses were

categorized as significant (p<0.05) according to a two-tailed t-test if the ΔF/F in a 200 ms window

surrounding lever press or release was statistically different from a 500 ms window immediately

preceding lever movement. To extract events on single trials used for estimation of complex spike

rates and event amplitudes, the first derivative of the raw fluorescence trace was thresholded at

2.5 standard deviations from baseline. Notably, image segmentation using these criteria did not

extract PC soma, which were clearly visible in many single- and two-photon imaging experiments.

For single trial amplitude measurements, only events well separated in time (greater than 650ms

from the next event) were considered; however, for measurements of rates, all events separated

by at least one frame were included.

Additional statistics and data sharing: Data are presented as mean ± s.e.m., unless stated

otherwise. Statistical tests were two-sided, and analyses of variance (ANOVA) were performed

when more than two groups were tested. Differences were considered statistically significant

when *P* < 0.05. No correction for multiple comparisons was applied. No statistical methods were

used to predetermine sample sizes. Data collection and analysis were not performed blind to the

conditions of the experiments, but data collection relied on automatized measurements and

subsequent analysis was based on code uniformly applied across experimental conditions. Data
analysis code are available on GitHub (access available upon request). The data that support the
findings of this study are available from the corresponding author upon reasonable request.
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Figure 1

Figure 2

Figure 3

Figure 4

A

Field of view   All dendrites   Correct / Early / Press

0.2 ΔF/F
200 ms

B

Responsive dendrites
(n = 962 of 1051)

0.025 ΔF/F
200 ms

p = 1.65e−75

Early ΔF/F

Correct ΔF/F

p = 2.78e−100

Too Fast ΔF/F

Correct ΔF/F

C

Correct   Early   Press

ΔF/F

D

Early responsive (%)   Correct responsive (%)

Early only (%)   Correct only (%)

Press responsive (%)   Release responsive (%)

Press only (%)   Release only (%)

E

Early (norm)   Correct (norm)

spontaneous
correct
early

0.1 ΔF/F
200 ms

Figure 5

Figure 6

Figure 7

A

Arch: 5 mice
10-50mW, one Step pulse for 1sec

Right arm •
Right & Left arms ●
Very weak arm movements •
Whiskers or mouth or nose •
nothing •

B

Electrical Stim. 3 mice
0.1mA-1mA
50Hz for 500ms or 1sec

Right arm •
Right & Left arms ●
Very weak arm movements •
Whiskers or mouth or nose •
nothing •

supplemental figure 1

A

500 ms cue prediction
Day1

500 ms cue prediction
Day2

n = 17

B

2 s cue prediction
Day1

2 s cue prediction
Day2

n = 10

C

cue

session
-1
0
1
2
3

D

500 ms cue prediction
Day1

next 500 ms cue prediction session after
> 3 cue reaction extinction sessions

n = 20

supplemental figure 2

supplemental figure 3

supplemental figure 4

A mesoscale imaging

Field of view   1P dendrites

B raw F trace   2 F

1st derivative   500 ms

C 2P imaging

Field of view

2P dendrites

D raw F trace

1st derivative   1 s

E 50 μV
2 s

400 μV
40 ms

F 0.6 ± 0.01 Hz   0.6 ± 0.06 Hz
n = 1051   n = 11

2P dendrites   single units

Event rate (Hz)

supplemental figure 5