Reorganization of corticospinal output during motor learning

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Motor learning is accompanied by widespread changes within the motor cortex, but it is unknown whether these changes are ultimately funneled through a stable corticospinal output channel or whether the corticospinal output itself is plastic. We investigated the consistency of the relationship between corticospinal neuron activity and movement through in vivo two-photon calcium imaging in mice learning a lever-press task. Corticospinal neurons exhibited heterogeneous correlations with movement, with the majority of movement-modulated neurons decreasing activity during movement. Individual cells changed their activity across days, which led to changed associations between corticospinal activity and movement. Unlike previous observations in layer 2/3, activity accompanying learned movements did not become more consistent with learning; instead, the activity of dissimilar movements became more decorrelated. These results indicate that the relationship between corticospinal activity and movement is dynamic and that the types of activity and plasticity are different from and possibly complementary to those in layer 2/3.

The ability of the motor cortex to drive movement is presumed to be mediated by a direct projection from a subset of motor cortex neurons to motor circuits within the spinal cord. These corticospinal neurons are located within layer 5B of the motor cortex but are spatially intermingled with non-corticospinal neurons. Neuron activity within the motor cortex has been closely linked to movement, both specifically in corticospinal neurons and in the general motor cortex population, suggesting its role in guiding ongoing behavior. In particular, the motor cortex has been implicated in motor skill learning. Behaviorally, this function is evidenced by the requirement of an intact motor cortex to learn new movements and a deficit in dexterous and skilled movements following acute motor cortex inactivation, motor cortical lesions or corticospinal tract transaction. Moreover, motor skill learning induces plasticity of the motor cortex at multiple levels, including stimulation-evoked movement maps, activity of neurons during learned behavior and dendritic spine growth and turnover. The organization of the motor cortex according to complex movements further supports the notion that it develops circuits that facilitate learned movements.

Learning-related plasticity has been demonstrated within many components of the motor cortex. Connection strength in the motor cortex changes with motor learning, including inputs from the thalamus and intracortical connections. Learning-dependent dendritic spine growth has also been observed in both superficial and deep layer motor cortex neurons, including in corticospinal neurons. These forms of plasticity also depend on and interact with plasticity in local inhibitory interneurons and downstream structures like the striatum. Given this distributed reorganization within the motor cortex, a fundamental question arises as to whether circuits within the cortex operate through a functionally stable output to the spinal cord or whether the behavioral correlation of corticospinal activity itself changes with motor learning. These two possibilities represent separate schemas of motor cortex plasticity: intracortical circuits could assemble around a consistent output channel, or the output channel itself could be malleable.

Different lines of evidence lend credence to both possibilities. Individual layer 5 neurons within the motor cortex can be associated with specific aspects of movement, and changes in neuronal activity during learning can directly reflect changes in corresponding movements. These results indicate that the relationship between movement and the activity of layer 5 cells is dynamic during motor learning, and artificial feedback can alter muscle activity associated with corticospinal activity. Directly addressing this issue therefore requires specifically monitoring the activity of large ensembles of corticospinal populations and accompanying movements across learning. We approached this using targeted in vivo two-photon calcium imaging in a lever-press task previously used to examine plasticity within layer 2/3 of the motor cortex. By using Cre-dependent expression of calcium indicators and imaging the apical dendrites of layer 5B corticospinal neurons, we were able to track the activity of corticospinal neurons every day for 2 weeks while animals learned and performed the task. We found that a subset of neurons was selectively active during movement, but, unexpectedly, a larger number of neurons were selectively active during quiescence. The behavioral correlation of each neuron was plastic; cells could switch between silent, indiscriminately active, actively selective during movement (movement-active) and active selectively during quiescence (quiescence-active) across days. These changes resulted in a dynamic relationship between

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Received 7 November 2016; accepted 30 May 2017; published online 3 July 2017; doi:10.1038/nn.4596
corticospinal activity and movement kinematics, such that given movements early and late in learning were accompanied by different activity patterns. Moreover, the corticospinal activity patterns accompanying dissimilar movements diverged, but, unlike what was seen in layer 2/3, there was no stabilization in activity patterns exhibited during the learned movement. These results indicate that functional plasticity within the motor cortex during learning extends to the corticospinal output.

**RESULTS**

**Two-photon calcium imaging of corticospinal neurons during motor learning**

We used the Cre–FLEX system to selectively express the calcium indicator GCaMP6f in corticospinal cells in the motor cortex. This was achieved by dual injections of two adeno-associated viruses (AAV): an AAV encoding Cre recombinase (AAV2/9-CaMKII-Cre, which can be taken up by axonal terminals and infect neurons projecting to the injected area) into the C7 and C8 segments of the spinal cord and an AAV encoding Cre-dependent GCaMP6f (AAV2/1-Syn-FLEX-GCaMP6f) into the right caudal forelimb area of the motor cortex (Fig. 1a). The caudal cervical segments of the spinal cord were targeted because they contain motor neurons innervating muscles for forelimb control, and corticospinal cells projecting to these segments exhibit structural plasticity during the learning of a forelimb motor task. Fluorescent cells in layer 5B of the motor cortex were observed 2 weeks after the injections, and these cells projected via the pyramidal tract to the spinal cord (Fig. 1b). Fluorescently labeled axons were observed in the intermediate and ventral lamina of the cervical spinal cord, consistent with targeting motor circuitry within the spinal cord. Axons within the corticospinal tract typically did not extend beyond the thoracic spinal cord (in 3 of 4 mice), suggesting that labeled cells were specific to forelimb control (Fig. 1c). Many axon collaterals were observed in regions outside of the spinal cord, consistent with reports of these cells projecting to multiple areas (Fig. 1d).

GCaMP6f-expressing dendrites were visible in vivo under a two-photon microscope, but somata were too deep to allow for consistent longitudinal imaging. Therefore, we imaged the apical trunks of dendrites passing through layer 2/3. The locations of these apical dendrites were stable across days and the same dendrites could be reliably identified each day (Fig. 2a). As dendrites of corticospinal neurons at various depths could be imaged in a single imaging plane, this approach had an added advantage of capturing larger ensembles of corticospinal neurons, compared to imaging at their somata. GCaMP6f fluorescence within these dendrites was observed as bright discrete points in a very low-noise background, allowing for automated region-of-interest creation (Fig. 2b).

In two mice with serendipitously bright and sparsely labeled corticospinal populations, we were able to track some dendrites to their respective somata. In these cases, we were able to image certain somata and their apical dendrites semi-simultaneously using a piezoelectric motor to rapidly move the objective lens vertically (~3.75 volumes per s, 8 planes per volume). With this approach, we found a high degree of overlap between calcium events in both somata and apical dendritic trunks (484 observed calcium events shared between dendrites and soma, 34 events unique to the soma and 14 events unique to the dendrites, across 36 neurons; Fig. 2c and Supplementary Fig. 1). We suggest from this that the vast majority of our observed calcium events in apical dendritic trunks were the result of back-propagating action potentials, which are known to induce calcium influx through voltage-gated calcium channels. Therefore, we posit that our apical dendrite signals can serve as a proxy for somatic spiking. This semi-simultaneous imaging of identified soma–dendrite pairs also confirmed that calcium events in sibling branches belonging to the same soma were highly correlated, in agreement with previous reports (Fig. 2c and Supplementary Fig. 1). Using data from verified sibling and nonsibling branches, we were able to set a cutoff value for similarity, which was then applied to mice with densely labeled corticospinal populations to categorize dendrites as likely originating from the same or different somata (Fig. 2d). We combined fluorescence traces from presumed sibling branches by weighted averaging (194 ± 68 ‘unique’ corticospinal neurons per mouse from 258 ± 87 imaged dendrites, mean ± s.d.). Calcium events were then detected within baseline-normalized traces through a thresholding process (Online Methods and Supplementary Fig. 2).

We performed apical dendrite imaging while mice were trained in a cued lever-press task previously used to examine functional and structural plasticity in layer 2/3 of the motor cortex (n = 8 mice). Mice were trained in the task in one approximately half-hour
Imaging apical dendrites of corticospinal neurons. (a) Left: coronal section of the motor cortex, illustrating deep corticospinal cells and prominent apical dendrites. Middle: schematic of imaging plane. Right: example in vivo two-photon images of corticospinal dendrites across days; blue-outlined images on top left are enlargements of the central regions outlined in blue. The same corticospinal dendrites could be readily identified each day. (b) Schematic of automated region-of-interest generation. Left: images aligned across days (green unfilled circles, imaged position; green filled circles, aligned position). Center left: active regions are detected by thresholding across all images from all days (white circles), and the centroids of those regions are stored (red dots). Center right: the shapes of active regions are defined as contiguous pixels above threshold on at least 50% of the frames in which the predetermined centroid is above threshold. Right: regions-of-interest are created as the borders of active shapes. (c) Left: example semi-simultaneous recordings from a corticospinal neuron soma and its four apical dendrite branches. Images are side-projection (left), dendrite plane (top right) and soma plane (bottom right); traces are min–max normalized fluorescence from dendrites (colors correspond to regions of interest) and soma (black). Right: histogram of L2 normalized fluorescence trace dot product among pairs of dendrites from different neurons (black, nonsibling branches) or the same neuron (blue, sibling branches) maximum normalized within each group. Red dashed line, cutoff for defining sibling branches in dense imaging. (d) Example fluorescence traces from dendrite imaging. Indicated blue traces are putative sibling dendrites above the similarity threshold.

Figure 3 Lever press task. (a) Schematic of task. Top: mouse holds a lever with its left forepaw. Bottom: lever trajectory (solid black horizontal lines, quiescence; red lines, movement) passing a threshold (horizontal dashed line) during an auditory cue (onset shown by black vertical line) results in a water reward (blue vertical line). (b) Rewarded movement stereotypy increases across days. Top: median correlations between rewarded movements of all pairs of days. Bottom left: rewarded movement correlation within days corresponding to the diagonal of the top plot (indicated by the black arrow); movements within days become increasingly stereotyped across time (Pearson’s correlation, $r = 0.40, P < 0.001$). Bottom right: rewarded movement correlation across adjacent days corresponding to the diagonal of the top plot (indicated by the gray arrow); movements across days become increasingly stereotyped across time (Pearson’s correlation, $r = 0.39, P < 0.001$). Error bars indicate s.e.m. across animals. (c) Mice perform one movement (‘learned movement’) more often after learning but retain variability. Top: histogram of the correlation between all movements and the learned movement (defined as the average movement across days 11–14) in the early and late stages of learning. Mice produce more movements that resemble the learned movement late in learning (two-sample Kolmogorov-Smirnov test, $P < 0.001$). Creating a template movement from days 1–4 did not result in a shifted distribution across learning (two-sample Kolmogorov-Smirnov test, $P = 0.06$), indicating that the shift in distribution is not an artifact of creating a template from the later days. Error shading indicates s.e.m. across animals. Middle: average lever trajectory from days 11–14 (learned movement) in an example animal. Bottom: example movements binned by correlation percentile to the learned movement. Gray, single movements; black, average of all movements within bin.

session each day for 2 weeks. During training, mice were head-fixed under a two-photon microscope and rested their right paw on a stationary block and their left paw on a lever attached to a force transducer (Fig. 3a). Imaging was conducted throughout each training session in the right motor cortex. The displacement of the lever was continuously recorded, allowing for a measurement of movement kinematics. The task structure consisted of a variable intertrial interval followed by an auditory cue, during which a press of the lever past the threshold produced a brief tone and a water reward. Mice learned this task over the course of 2 weeks and developed an increasingly stereotyped movement to achieve reward yet maintained some variability, which we took advantage of in later analyses (Fig. 3b,c and Supplementary Fig. 3). In two circumstances in which we were able to image dendrites and somata simultaneously across days, we confirmed that activity was reliably shared between compartments throughout learning (Supplementary Fig. 4). Below, we describe the activity of corticospinal neurons during the learning of this task, and in several cases we compared...
Corticospinal neuron activity is heterogeneously related to movement. (a) Example activity from a single mouse. Top: population average of all neurons (black), movement-active neurons (green) and quiescence-active neurons (red). Middle: single cells that are movement-active (green), quiescence-active (red) and indiscriminately active (yellow). Bottom: lever movements. Blue highlighted regions represent portions of the lever trace that were detected as movement. (b) Activity of all cells aligned to movement onset and offset (dashed lines). Top: activity of all recorded cells in all animals, min–max normalized for the average within each day and then averaged across days (1,553 cells), sorted by the coefficient of the first principal component of average activity across cells. Bottom: average activity across all cells, then averaged across animals. Error shading indicates s.e.m. across animals. (c) Average activity of active classes of cells aligned to movement onset and offset (dashed lines). Top: activity of all recorded cells that fell into each category on at least 1 d, min–max normalized within day and then averaged across days with that classification, sorted by the coefficient of the first principal component of average activity across all cells (413 movement-active cells, 760 quiescence-active cells, 1,026 indiscriminately active cells). Note that if a cell was classified differently across days, it appears under multiple classes and is averaged across the days with that classification. Bottom: average activity across all cells of a given classification averaged across days with that classification, then averaged across animals. Error shading indicates s.e.m. across animals.

Figure 4 Corticospinal neurons are heterogeneously related to movement. (a) Example activity from a single mouse. Top: population average of all neurons (black), movement-active neurons (green) and quiescence-active neurons (red). Middle: single cells that are movement-active (green), quiescence-active (red) and indiscriminately active (yellow). Bottom: lever movements. Blue highlighted regions represent portions of the lever trace that were detected as movement. (b) Activity of all cells aligned to movement onset and offset (dashed lines). Top: activity of all recorded cells in all animals, min–max normalized for the average within each day and then averaged across days (1,553 cells), sorted by the coefficient of the first principal component of average activity across cells. Bottom: average activity across all cells, then averaged across animals. Error shading indicates s.e.m. across animals. (c) Average activity of active classes of cells aligned to movement onset and offset (dashed lines). Top: activity of all recorded cells that fell into each category on at least 1 d, min–max normalized within day and then averaged across days with that classification, sorted by the coefficient of the first principal component of average activity across all cells (413 movement-active cells, 760 quiescence-active cells, 1,026 indiscriminately active cells). Note that if a cell was classified differently across days, it appears under multiple classes and is averaged across the days with that classification. Bottom: average activity across all cells of a given classification averaged across days with that classification, then averaged across animals. Error shading indicates s.e.m. across animals.

Corticospinal activity with layer 2/3 activity reanalyzed from our previous data (Online Methods).

Corticospinal neuron activity is heterogeneously correlated with movement

As a first step to examining the relationship between corticospinal neuron activity and movements, we characterized the activity patterns of individual neurons around movements (Online Methods). Cells could be movement-active, quiescence-active or active indiscriminately with regards to movement (Fig. 4a), although unexpectedly there were many more cells selectively active during quiescence than movement (Fig. 4b). This manifested as a decrease of global population-averaged activity during movement (Fig. 4b), in striking contrast to layer 2/3, which displayed a large increase in population activity during movement (Supplementary Fig. 5a).

We further investigated the heterogeneous response types by classifying cells as either movement-active, quiescence-active, indiscriminately active or silent (Online Methods). In accordance with the decrease in population activity around movement, there were roughly twice as many quiescence-active cells than movement-active cells (Fig. 4c). Averaging activity within classes established very different response profiles across movement- and quiescence-active cells (Fig. 4c). In particular, quiescence-active cells showed higher levels of activity during quiescence than movement-active cells. This excluded the possibility that quiescence-active cells and movement-active cells had the same level of spontaneous activity and were suppressed or activated by movement respectively. Furthermore, the quiescence-active population exhibited an increase in activity immediately after movement offset, suggesting a possible postinhibitory rebound or a function in stopping movement. There were many fewer quiescence-active cells in layer 2/3, although the average activity of each class was similar to that of corticospinal neurons (Supplementary Fig. 5b). Consequently, more corticospinal than layer 2/3 cells were active during quiescence, but the fraction of active cells during movement was comparable in both populations (Supplementary Fig. 5c).

Corticospinal activity is dynamic across learning

When examining corticospinal populations across time, we found that cells often switched movement-related classification. Individual neurons could move between being active and silent across days or even switch between being movement- and quiescence-active (Fig. 5a). On a daily basis, roughly 50% of cells were active, and there were twice as many quiescence-active and indiscriminately active cells as movement-active cells (Fig. 5b). Notably, the fraction of quiescence-active cells increased after the first 2 d, coinciding with a large increase in movement stereotypy (Fig. 3b). This was the converse of what we observed in layer 2/3, where movement-active cells increased early in learning without a significant change in the fraction of quiescence-active cells (Supplementary Fig. 5d). The classification of individual corticospinal cells was dynamic across days but became more stable for both the movement- and quiescence-active populations later in learning (Fig. 5c).
The dynamism of activity within single cells prompted us to investigate the activity of movement-modulated neurons across learning. Although neurons could alter their activity over days, more than half of the cells maintained their classification between first and second weeks (Fig. 6a). Of the cells that switched classifications across weeks, the transition to newly movement-active was less common than transitions away from movement-active, transitions to newly quiescence-active and transitions away from quiescence-active (Fig. 6a).

To determine overarching changes in activity, we quantified the average activity across all cells during all movement or quiescence epochs within each day. This showed a stable level of activity during both quiescence and movement in the first week and a slightly decreasing level of activity during both states in the second week of training (Fig. 6a). This decrease in activity in the second week appeared to be more exaggerated for periods around movement onset (Fig. 6a). The average activity during quiescence did not change in the first few days despite the increase in the fraction of quiescence-active cells; this is reminiscent of previous results in which more layer 2/3 neurons became movement-active early in learning, but the average activity during movement was stable, as it was balanced by each cell being active less often.

When we analyzed groups of cells separately depending on how they transitioned between classes, we found a number of noteworthy dynamic features. First, the activity of stably movement-active cells during movement increased in the first week and decreased in the second week, and the activity increase was roughly uniformly distributed while activity decreases were biased toward movement onset (Fig. 6b). Stably quiescence-active cells, on the other hand, maintained consistent levels of activity during the first week and declined in activity during both quiescence and movement in the second week (Fig. 6b). These changes in stably classified neurons indicate that even consistently modulated cells shaped their activity throughout learning.

When we considered cells that switched classification, a noteworthy asymmetry emerged: cells that transitioned away from being movement-active became quiescence-active or indiscriminately active, while cells that transitioned away from being quiescence-active largely became silent (Fig. 6b). Likewise, cells that became newly movement-active were previously silent, while cells that became newly quiescence-active were previously movement-active or indiscriminately active (Fig. 6b). This presents the possibility that active cells can be repurposed by transitioning away from movement-active or toward quiescence-active but that the transition toward movement-active or away from quiescence-active involves turning activity on and off entirely.

Learning induces decorrelation in activity accompanying dissimilar movements

A fundamental question is whether these activity changes are due to changes in movements or whether the relationship between neuronal activity and movement is itself altered. We investigated this by comparing activity patterns that accompanied individual movements across learning. Because mice maintained variability of movements throughout training even with an overall increase in stereotypy (Fig. 3c), pairs of movements could be identified across all days that were similar or dissimilar to each other. For example, a movement on the first day could be similar to some and dissimilar to other movements on the last day. This allowed us to determine the association between activity and movement within and across days.

We found that, both within the early and within the late stages of training, the similarity of activity patterns was related to the similarity of the movements that they accompanied. This was the case in both corticospinal and layer 2/3 neurons (Fig. 7a). Notably, we previously found that in layer 2/3, this activity–movement relationship was absent when comparing movements across stages of training, indicating that novel associations between activity and movement developed with learning (Fig. 7a). Unexpectedly, this same shift in the activity–movement relationship was also observed in corticospinal cells despite their more direct connectivity to movement-generating circuitry (Fig. 7a). Notably, this effect was also observed in the
Figure 6 Changes in activity across time. (a) Left (majority classification): fraction of all recorded cells divided by their majority classification within weeks (that is, the largest number of days with a given classification. For example, if a cell was classified as movement-active on 3 d of a week, then the cell’s majority classification for that week is movement-active). Center (average activity): average \( \Delta F/F \) values across all cells during all movement and quiescence epochs. Activity during both movement and quiescence was stable in the first week while activity in both states decreased in the second week (Pearson’s correlation coefficient of values \( z \)-scored within animal, movement week 1: \( r = 0.23, P = 0.1 \); movement week 2: \( r = -0.51, P < 0.001 \); quiescence week 1: \( r = -0.23, P = 0.1 \); quiescence week 2: \( r = -0.26, P = 0.0497 \)). Right (movement-aligned activity): average movement-aligned activity across all cells and across groups of days denoted by colored lines. Error bars indicate s.e.m. across animals. Vertical dashed lines indicate onset and offset of movement. (b) Plots as in a, for different groups of cells according to their classification by week. Cell populations are indicated by pie charts and correspond to cells stably movement-active (top left), stably quiescence-active (top right), switching out of movement-active (center left), switching to movement-active (center right), switching out of quiescence-active (bottom left) and switching to quiescence-active (bottom right). Activity during movement for stably movement-active cells increased in the first week and decreased in the second week, while activity during quiescence for stably quiescence-active cells did not change in the first week and decreased in the second week (Pearson’s correlation coefficient of values \( z \)-scored within animal; stably movement-active cells during movement week 1: \( r = 0.53, P < 0.001 \); stably movement-active cells during movement week 2: \( r = -0.48, P < 0.001 \); stably quiescence-active cells during quiescence week 1: \( r = 0.14, P = 0.3 \); stably quiescence-active cells during quiescence week 2: \( r = -0.53, P < 0.001 \)). Error bars indicate s.e.m. across animals.

no-task animals (Supplementary Fig. 6c), indicating that this process may be a constant and general feature of the motor cortex. The activity–movement relationship appeared to drift evenly over time in both layer 2/3 and corticospinal cells (Supplementary Fig. 7).

Even though the activity–movement relationship changed in both layer 2/3 and corticospinal neurons, the nature of changes was distinct. Specifically, in layer 2/3, similar movements became associated with increasingly similar activity with training, resulting in a more consistent activity–movement relationship after learning (Fig. 7a). In contrast, corticospinal activity did not become more consistent for similar movements; instead, dissimilar movements became associated with more distinct activity patterns (Fig. 7a). This change was not
The motor cortex is thought to play a fundamental role in motor learning and is capable of extensive plasticity. It has been unclear, however, whether activity within the motor cortex operates through a stable output to the spinal cord or whether the corticospinal output of the motor cortex is itself plastic. We addressed this issue by developing a method for longitudinally imaging the activity of corticospinal neuron populations across learning. We found that certain corticospinal neurons were active selectively during movement, while a larger fraction of corticospinal neurons were selectively active during quiescence. Moreover, the activity of corticospinal neurons was dynamic across days, such that different cells were active during movement or quiescence. These changes ultimately lead to novel associations between corticospinal activity and movement. Notably, the changing relationship between activity and movement in corticospinal neurons was seen in both task and no-task animals, while task-learning specifically induced a decorrelation of activity across dissimilar movements. We note, however, that our no-task mice were not completely free of learning, as they were put in a novel environment under head fixation with a lever. Therefore, it is possible that these animals still exhibited certain learning-related changes.

Heterogeneity of corticospinal activity
The observation that corticospinal neurons can be either active selectively during movement or quiescence corroborates previous findings dating back to the earliest recordings of motor cortex activity. Moreover, it has been suggested that quiescence-active neurons are found exclusively in intermediate and deep layers of the motor cortex, reinforcing our observed differences between previously recorded layer 2/3 activity and layer 5B corticospinal cells. We report here a larger fraction of quiescence-active neurons than typically reported in the

Figure 7 Cell-type-specific differences in the relationship between movement and activity. (a) Pairwise correlation in population activity as a function of correlation of accompanying movements. Left, corticospinal; right, layer 2/3. The interaction between movement correlation and activity correlation becomes stronger over time for both corticospinal and layer 2/3 cells (paired Wilcoxon signed-rank test of the fitted slope for black vs. gray lines; corticospinal, \( P = 0.008 \); layer 2/3, \( P = 0.02 \)). In corticospinal cells, this derived from less-correlated activity for negatively correlated movements (paired Wilcoxon signed-rank test for negatively correlated movement bins for black vs. gray lines, \( P = 0.009 \)). In layer 2/3 cells, activity became more correlated for similar movements (paired Wilcoxon signed-rank test for positively correlated movement bins for black vs. gray lines, \( P < 0.001 \)). The activity patterns after learning were novel compared to those before learning (paired Wilcoxon sign-rank test of the fitted slope for gray vs. blue lines; corticospinal, \( P = 0.008 \); layer 2/3, \( P = 0.02 \)). Error bars indicate s.e.m. across animals. (b) Pairwise correlation in population activity for movements, separated by type of movements. Left: pairwise correlation of corticospinal population activity on pairs of trials with dissimilar movements (from data in the purple box in a, left). Correlation in activity does not depend on the type of movement made (paired Wilcoxon signed-rank test; black line, \( P = 0.9 \); gray line, \( P = 0.4 \); blue line, \( P = 0.5 \)). Right: pairwise correlation in layer 2/3 population activity on pairs of trials with similar movements (from data within the orange box in a, right). Correlation in activity is higher specifically for learned movements late in learning (paired Wilcoxon signed-rank test; black line, \( P = 0.5 \); gray line, \( P = 0.02 \); blue line, \( P = 1 \)). Error bars indicate s.e.m. across animals. (c) Schematic of population-specific changes in relationship between activity and movement. Boxes, spaces of potential activity patterns; circles, activity patterns associated with given movements within each day; days progress from black to gray. Used activity drifts across time in both populations. In layer 2/3 this is accompanied by a more consistent activity pattern specifically for the learned movement (smaller circle). Conversely, in corticospinal neurons, different movements associate with more separable activity patterns (separation of gray circles).
motor cortex, such that the population-average activity of corticospinal neurons decreased during movement. While the reason for this apparent discrepancy is unclear, two possibilities are that this balance is specific to corticospinal neurons and not the general deep layer population or that the movement in our task was particularly effective in eliciting activity during quiescence. As we demonstrated in the no-task mice, not all movements elicited the same balance of activity.

The diversity of corticospinal activity may not be unexpected given the heterogeneity of the cellular properties within the corticospinal neuron population. The functions of different response types and their relationship to heterogeneity in cellular properties, however, are unknown. It is possible that movement- and quiescence-active corticospinal cells have unique descending connections or other intrinsic differences and are effectively segregated into unique subtypes. Toward this end, it has been observed that axonal conduction velocity and response to neuromodulators can differ between corticospinal cells with different response types. On the other hand, movement- and quiescence-active cells might not be independent cellular subtypes but may instead have flexible roles in circuit dynamics. This notion is supported by observations that corticospinal cells can switch between movement- and quiescence-active responses for different types of movements within a day and across days. The function of activity during quiescence has yet to be deduced, but it may be involved in specifically halting movement, as suggested by work on the vibrissa motor cortex, or it may be an inherent aspect of generating activity with particular dynamics.

## Motor cortex output is flexibly associated with movement

A main finding from the current work is that corticospinal activity changes with time to create a novel relationship between activity and movement. We previously found that same phenomenon in layer 2/3, and extending this result to corticospinal neurons indicates that the motor cortex does not utilize a consistent functional output. We note that this flexibility does not necessitate a corresponding change in how downstream motor circuitry is influenced by motor cortical input. It is possible, for example, that the relationship between corticospinal and spinal cord activity is stable but degenerate, so that one subset of possible activity patterns for a given movement is observed before learning and another subset after learning. Indeed, it has been documented that motor cortex can reversibly switch between multiple activity states, and activity of a given muscle can be accompanied by different patterns of motor cortex activity based on the context of movement both generally in the motor cortex and specifically in corticospinal neurons. It should also be noted that not all motor cortex activity generates movement, and indeed population activity can evolve within ‘movement-null’ space without overt effects on movement.

A drift across a space of functionally degenerate activity is supported by recent work in zebra finches, in which a stereotyped song is accompanied by a changing pattern of premotor activity across days, while inhibitory interneuron activity and the local field potential retained consistent patterns. The authors suggest that drifts in population activity may actively develop degeneracy, contributing to a more robust circuit that can tolerate input noise and output variability, which nevertheless relates to stable motor output. Our results are consistent with this finding, suggesting that activity drifts may be a common principle across species. Such degeneracy could be beneficial because it requires less reliance on any given collection of neurons, making the system more robust to noise and insult. Alternatively, it might be important to allow for a movement to be associated with multiple inputs. For example, a given forelimb movement may be triggered by many different sensory inputs, in many different contexts and toward many different aims. It may be maladaptive to have only one required ‘target’ pattern of output activity that must be generated in each of these cases; instead, degeneracy may allow for each of these contexts to utilize one of many possible activity patterns to produce the same movement. It will therefore be an important issue in the future to differentiate degeneracy that is stable over time from remapping between motor cortex activity and movement.

Another possible functional benefit of activity drifts relates to the fundamentally dynamic nature of motor systems. The demands of motor systems constantly change, based on many factors including muscle fatigue, muscle strengthening, injury and external forces as subtle as a long sleeve shirt or heavy shoes. Accordingly, the motor control system may always maintain variability of representations so that it can adapt to unpredictable changes.

The relationship between movement similarity and activity similarity became stronger for both layer 2/3 and corticospinal neurons but in opposite ways. In layer 2/3, this occurred through increasingly consistent activity patterns, especially for the learned movement, while corticospinal cells acquired more distinct activity patterns for dissimilar movements in general. This suggests possible complementary roles for layer 2/3 and corticospinal cells, with layer 2/3 establishing learned patterns of activity that feed into a corticospinal system that retains degeneracy while separating spaces of activity for different movements. These components together could make up a circuit that establishes consistent interpretations of important inputs, uses that to operate a flexible output command, and ensures that the range of output commands are sufficiently differentiable by downstream targets. It will be of interest to determine how these changes are then carried downstream, especially given that descending corticospinal connectivity is known to be malleable.

## METHODS

Methods, including statements of data availability and any associated accession codes and references, are available in the online version of the paper.

Note: Any Supplementary Information and Source Data files are available in the online version of the paper.

## ACKNOWLEDGMENTS

We thank A. Kim, T. Loveland and L. Hall for technical assistance and thank current and former members of the Komiyama lab, especially S. Chen, J. Dahlén, B. Danskin and H. Liu, for comments and discussions. This research was supported by grants from NIH (R01 DC014690-01, R21 DC012641, R01 NS091010A, U01 NS094342 and R01 EY025349), Human Frontier Science Program, Japan Science and Technology Agency (PRESTO), New York Stem Cell Foundation, David & Lucile Packard Foundation, Pew Charitable Trusts and McKnight Foundation to T.K. A.J.P. was supported by the Neuroplasticity of Aging Training Grant (AG000216), a Newton International Fellowship, a Human Frontier Science Program fellowship and an EMBO fellowship. J.L. was supported by the Swiss National Science Foundation.

## AUTHOR CONTRIBUTIONS

Conceptualization, A.J.P. and T.K.; methodology for spinal cord injections and histology investigation, J.L.; longitudinal simultaneous dendrite and soma imaging, N.G.H. and K.O.N.; other methodology and investigation, A.J.P.; software and writing for the original draft, A.J.P.; analysis, A.J.P. and T.K.; writing review and editing, A.J.P. and T.K.; supervision and funding acquisition, T.K.

## COMPETING FINANCIAL INTERESTS

The authors declare no competing financial interests.

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ONLINE METHODS

Animals. All procedures were in accordance with protocols approved by the UCSD Institutional Animal Care and Use Committee and the guidelines of the National Institutes of Health. All mice were male and acquired from Charles River Laboratory (C57Bl/6 wild-type mice). All surgeries and experiments were carried out in adult mice (6 weeks or older). All animals were group-housed before surgery and singly housed afterwards in disposable plastic cages with standard bedding, nestlets and a running wheel; they were kept in a room on a reversed light cycle (12 h). All experiments were performed at approximately the same time each day during the dark period.

Surgery. Surgeries consisted of two consecutive parts, the first being a spinal cord injection and the second being cortical injection and cranial window preparation. Mice were anesthetized with isoflurane and fixed on a bite bar with a nose clamp over a heating pad. The back was shaved from below the shoulder blades to the top of the neck and cleaned with iodine and alcohol. A midline incision was made in the skin from below the shoulder blades to the middle of the neck. Fatty tissue was removed as necessary to expose the trapezius muscles. The trapezius muscles were then cut along the midline at the shoulder blades to expose the spine. The spinothalamic process on the T2 vertebra was identified and separated from attached musculature. The spine was then fixed using custom metal wedges held by a stereotaxic frame. This was accomplished by lifting the spine from the T2 spinous process while placing the wedges under the trapezius muscles to support the spine from underneath. Fatty tissue over the spine was then removed and muscles directly overlying the spine were cut. A laminectomy was performed in the range of the C7 to C5 vertebrae, exposing the C6 to C8 segments of the spinal cord. A viral solution of AAV2/9-CaMKII-Cre (University of Pennsylvania Vector Core Facility) was injected into two sites on the left side of the spinal cord, each injection being 200 nL and placed 400 µm from the midline, 700 µm from the surface and separated by 600 µm rostrocaudally. After injecting, the wedges fixing the spine were removed, the trapezius was sutured with 5-0 Vicryl sutures and the skin was sutured with 5-0 silk sutures. Immediately after the spinal cord injections, cortical injections and cranial windows were then prepared as previously described. Skin overlying the skull was removed, the skull was scraped clean and a custom headplate was glued to the skull and fixed with dental cement. A craniotomy was then performed over the right caudal forelimb area of the motor cortex as stereotaxically defined. A viral solution of 1:4 diluted AAV2/1-Syn-FLEX-GCaMP6f (University of Pennsylvania Vector Core Facility) was injected into the cortex in five sites in a plus (+) shape, each injection being 40 nL and placed 700 µm from the surface, separated by 500 µm, centered at 1.500 µm lateral and 300 µm anterior from bregma. A glass window consisting of a base and concentrically attached smaller plug was held against the skull and brain respectively; the gap between plug and skull was filled with 1.5% agarose, and the base was fixed in place with dental cement. Batyril (10 mg/kg) and buprenorphine (0.1 mg/kg) was injected subcutaneously at the end of surgery. Animals did not display motor deterrents following surgery and were often observed running on their wheels within 1 d of surgery.

Behavior. Animals were trained in a lever-press task as previously described. Mice were water restricted to a maximum of 1–2 mL per d beginning 3 d after surgery for 2 weeks before training. Mice were then trained in the lever-press task during two-photon imaging for 1 session per d, lasting approximately 0.5 h. Mice rested their body and hindlimbs in a tube, and placed their right forelimb on a movable lever. The lever consisted of a handle glued to a piezoelectric flexible force transducer (LCL-113G, Omega Engineering). Voltage from the force transducer, which was linearly proportional to the lever displacement, was continuously monitored using a data acquisition device (LabJack) and software (LabVIEW, National Instruments). Presses of the lever were defined as displacement through two thresholds within a short time (~1.5 mm to ~3 mm below resting position within 200 ms). The task structure consisted of a variable intertrial interval, followed by a cue period during which lever presses triggered water reward. Cue periods and rewards were paired with separate tones, and a failure to press the lever within the cue period resulted in a short burst of white noise. The cue period was reduced during the first two sessions from 30 s to 10 s, and the intertrial interval was increased during the first three sessions from 2–4 s to 5–7 s and then to 8–12 s to encourage discrete movements.

Mice in the no-task condition underwent the same preparations, training and task structure as defined above, except that water rewards were not contingent on lever press and instead were delivered on every trial after 0.5–2 s of the cue tone.

Immunofluorescence. Mice were anesthetized and transcardially perfused with ice-cold 0.1 M PBS (pH 7.4), followed by perfusion with ice-cold 4% paraformaldehyde (PEA) solution. Isolated brains and spinal cord were postfixed overnight at 4 °C in 4% PEA and cryoprotected in 30% sucrose solution for at least 24 h at 4 °C.

Microtome-cut (Thermo Scientific Microm HM 430) 60-µm free-floating brain (coronal) and brainstem (sagittal) sections were collected in PBS and stored at 4 °C. Cryostat-cut (Leica CM 1900) 20-µm spinal cord sections were collected on microscopy slides (Fisherbrand Superfrost Plus) and stored at ~80 °C.

Antibodies were diluted in staining buffer consisting of 0.1% (wt/vol) bovine serum albumin (BSA, OmniPur) and 0.3% (vol/vol) Triton X-100 (Alfa Aesar) in PBS. Primary antibodies were incubated overnight at 4 °C, then washed three times in PBS. Secondary antibodies were incubated for 1 h at RT (20–22 °C) followed by washing three times in PBS. Tissue was mounted using CC/Mount (Sigma). Primary antibodies used were guinea pig anti-NeuN (1:1,000, Synaptic Systems, No. 266004) and chicken anti-GFP (1:400, Aves, No. 1020). Secondary antibodies used were goat anti-guinea pig Alexa Fluor 594 (1:1,000, Invitrogen, A11076) and goat anti-chicken Alexa Fluor 488 (1:1,000, Invitrogen, A11093).

Images of brain and spinal cord sections were taken with a Zeiss Imager M2 with the Apotome.2 attachment, controlled with AxioVision 4.8 software. Adjacent images were stitched with Microsoft Image Composite Editor Version 1.4.4.0 and color levels were postprocessed using Adobe Photoshop CS6.

Two-photon imaging. Two-photon imaging was conducted through a 16× 0.8 NA objective (Nikon) mounted on a commercial two-photon microscope (B-scope, Thorlabs) and using a 925-nm laser (Ti:sapphire laser, Newport). Images were acquired with Scanimage 4.1 (Vidrio Technologies) at a rate of ~28 Hz, covering ~340 µm × 340 µm with 512 × 512 square pixels. Frame triggers, lever voltage and start times of trials were recorded with Ephus (Vidrio Technologies), allowing for alignment between behavior and imaging. Drifts in the imaging field during imaging were manually monitored and corrected. Images were motion corrected offline by maximizing 2D cross-correlations between raw images and an average reference image. For semi-simultaneous dendrite and soma imaging, a z-stack was first collected consisting of 51 slices spanning 400 µm to identify dendrite branches and corresponding somata, with each slice being an average of 100 motion corrected frames. Multiple z-planes were then imaged semi-simultaneously using a piezo stage mounted to the objective (Physik Instruments). Eight z-planes separated by 50 µm were used, to allow sufficient time for the piezo stage to travel the full range of 400 µm.

Automated region of interest generation. Hundreds of dendrites were imaged within each field, each of which could be sparsely active, motivating us to use automated region of interest generation (Fig. 2b). This was done by first registering maximum projections from all sessions together through affine alignment to account for slight differences in fields across days. All subsequent steps were performed on movies smoothed by 50-frame moving averages. Each frame was registered according to the affine transformation matrices calculated during session alignment, ensuring that each pixel across the experiment represented the same location in the brain. Any pixels along the edges that were not imaged within and across all sessions were not used any further. Active portions of the field were then defined on a frame-by-frame basis across the entire experiment by subtracting the average image within each session from all frames of that session and thresholding the average-subtracted frames by one s.d. of all average-subtracted pixel values. Dendrites were often closely neighboring and the point-spread resulted in overlapping thresholded regions across dendrites, so time-invariant thresholding was not informative. Instead, we took advantage of the temporal diversity of activity across dendrites to define co-varying thresholded pixels. Our approach was similar in intent to methods using independent components analysis (ICA), but we found ICA to be insensitive and not easily and accurately segmentable, while our method (detailed below) was highly sensitive and able to segment the images cleanly as verified by visual inspection.

We began by finding the centroids of all discrete active spots within each frame and summing all active centroids across the entire experiment. Active spots
Baseline estimation was performed as previously described. Briefly, a recursive other being 1× the noise to define baseline. Noise was estimated as the s.d. of thresholds were defined, one being 3× the noise to find active portions and the event detection in order to remove signals within the trace caused by calcium sibling branch ROIs. further analysis can therefore correspond either to a single ROI or combined L2 norm to take advantage of the highest signal to noise ratios. Each ‘cell’ in weighted averaging of all sibling branch traces according to their across-session window within each session and L2-normalized across all sessions. Based on originiated from the same cell. Activity similarity between ROIs was calculated by the baseline trace.

Baseline-normalized fluorescence traces within ROIs were subject to calcium events above the fifth percentile for those of classified cells were considered indiscriminately active. A minimum of 5 fluorescence events was imposed to define any cell as active, and cells not fitting these criteria were classified as silent.

Fluorescence analysis. Traces for each ROI were created by averaging enclosed pixels and subtracting background fluorescence. Background subtraction was critical for extracting the activity of single dendrites, as the signal from neighboring dendrites could invade an ROI. Within each raw imaging frame, background signal was estimated by interpolating fluorescence values across each ROI from the surrounding fluorescence (inpaint_nans Matlab function, J. D’Errico, MATLab File Exchange). Two traces were then produced for each ROI, one averaging the raw ROI pixels and one averaging the background-estimated ROI pixels. The changes in fluorescence (ΔF as defined below) of the background trace were then subtracted from the raw trace, producing a final background-subtracted trace. This process errs on the side of over-estimating background signal, because the point spread function decreases superlinearly from the source but interpolation was linear and because some signal originating from within the ROI was included. The amplitudes of calcium events are therefore somewhat reduced, but contaminating signals are effectively eliminated.

Apical dendrites often have multiple branches within the superficial layers, which led us to combine the traces of ROIs with very similar activity that likely originated from the same cell. Activity similarity between ROIs was calculated by the dot product of baseline-normalized traces, LOESS-smoothed with a 3.4 s window within each session and L2-normalized across all sessions. Based on the semi-simultaneous imaging of somata and their dendrites (Supplementary Figure 1), any ROI groups with a normalized dot product over 0.8 were considered putative sibling branches. Final traces for each neuron were derived by weighted averaging of all sibling branch traces according to their across-session L2 norm to take advantage of the highest signal to noise ratios. Each ‘cell’ in further analysis can therefore correspond either to a single ROI or combined sibling branch ROIs.

Baseline-normalized fluorescence traces within ROIs were subject to calcium event detection in order to remove signals within the trace caused by calcium indicator dynamics instead of neuronal activity (Supplementary Fig. 2). Two thresholds were defined, one being 3× the noise to find active portions and the other being 1× the noise to define baseline. Noise was estimated as the s.d. of negative fluorescence values mirrored about zero to simulate the noise distribution. Active portions of the trace were identified by a 1 s LOESS-smoothed trace crossing the active threshold and extended backwards to begin when the baseline threshold was last crossed by the unsmoothed trace. Periods of the smoothed trace with negative slopes during active portions were set to inactive to eliminate fluorescence changes not associated with action potentials. All remaining active portions were considered calcium events and set to the difference between the maximum and minimum values within each event, with all other points set to zero.

Layer 2/3 data. All analyses involving layer 2/3 were performed on previously published data, with fluorescence thresholding updated to utilize the method described above and classification updated to utilize the method described below.

Movement analysis. Voltage from the piezoelectric lever was continuously recorded at 10 kHz during each session and parsed into movement and quiescence epochs as previously described. Briefly, movement was first identified by velocity threshold. Movement epochs were then refined by combining nearby epochs, eliminating small epochs and refining the start and end times of movement epochs according to when the lever position respectively left or entered a baseline defined by adjacent quiescent epochs. Visual inspection confirmed accurate demarcation of behavior.

To utilize the fullest extent of data, movements used for analyses were not restricted to only those movements that led to a reward. In this case, movements made during the intertrial interval or unsuccessful movements during the response period were also analyzed. For analyses involving extraction of individual movements and accompanying activity, only movements longer than 2 s and with at least 1 s of preceding quiescence were used, and only the first 2 s of those movements and accompanying activity were analyzed.

Movement-related classification. Cells were classified as movement-active or quiescence-active on each day. The fraction of movement frames that contained activity in each ROI was first calculated. This value was compared to a shuffled distribution, in which movement and quiescence epochs were kept intact but shuffled relative to each other 10,000 times. Activity during shuffled movement epochs was compared to activity during actual movement epochs. Actual values that were above the 97.5 percentile of the shuffled distribution were classified as movement-active while actual values that were below the 2.5 percentile of the shuffled distribution were classified as quiescence-active.

Among the cells that were not classified as movement-active or quiescence-active, those which had both an average ΔF/F and number of fluorescence events above the fifth percentile for those of classified cells were considered indiscriminately active. A minimum of 5 fluorescence events was imposed to define any cell as active, and cells not fitting these criteria were classified as silent.

Stability of classification (Fig. 5c) was defined by z-score in order to control for differing total numbers of classified cells on each day. This was done by creating a chance distribution of overlap by shuffling classification across all cells 1,000 times and calculating the z-scored overlap as (real overlap – shuffled overlap mean)/(shuffled overlap s.d.).

Pairwise activity correlation analysis. For analyses comparing population activity accompanying movement (Fig. 7a,b and Supplementary Figs. 6c, 7 and 8), the first 2 s of movement and activity were extracted for all movements that lasted longer than 2 s and had at least 1 s of preceding quiescence, regardless of whether they were rewarded movements or not. Pairwise activity correlation was calculated by concatenating the temporal activity of all cells during the corresponding movement and finding the Pearson’s correlation coefficient between pairs of these population activity vectors. The correlation between corresponding pairs of movements was calculated as the Pearson’s correlation coefficient of lever trajectories.

For analyses relating to the type of movement performed (Fig. 7b), the ‘learned’ movement was defined as the average lever trajectory across all movements from days 11–14. Each movement was then correlated to the learned movement, and pairs of movements were segregated by having a positive or negative maximum correlation with the learned movements.
Statistics. Statistical tests were chosen to avoid assumptions about data distributions, and therefore data was not tested for normality. All corticospinal data uses \( n = 8 \) animals for all days, while layer 2/3 data uses \( n = 7 \) mice for days 1–11, 6 mice for days 12–13 and 5 mice for day 14. No statistical methods were used to predetermine sample sizes, but our sample sizes are similar to those reported in previous publications.\(^{19}\) No randomization was used; mice used for layer 2/3 cells were from a previous experiment, mice used for corticospinal cells learning the task were prepared first in this experiment and mice used for corticospinal cells not learning the task were prepared second in this experiment. No blinding was used because no blinding was possible with our experimental structure. A Supplementary Methods Checklist is available.

Data and code availability. The data collected for this study and code used for analyses are available upon reasonable request from the corresponding author.

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