Sequential activation of premotor, primary somatosensory and primary motor areas in humans during cued finger movements

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HIGHLIGHTS

- Using onset of high gamma activities in human electrocorticography, the authors capture sequential activation of premotor, somatosensory and primary motor regions in preparation for voluntary movements.
- The fact that sensory neurons led motor neurons in activation in movement planning supports the notion of the efference copy.
- Premotor neurons, rather motor neurons, modulate the activities of sensory neurons in preparation for human voluntary movements.

ABSTRACT

Objective: Human voluntary movements are a final product of complex interactions between multiple sensory, cognitive and motor areas of central nervous system. The objective was to investigate temporal sequence of activation of premotor (PM), primary motor (M1) and somatosensory (S1) areas during cued finger movements.

Methods: Electrocorticography (ECoG) was used to measure activation timing in human PM, S1, and M1 neurons in preparation for finger movements in 5 subjects with subdural grids for seizure localization. Cortical activation was determined by the onset of high gamma (HG) oscillation (70–150 Hz). The three cortical regions were mapped anatomically using a common brain atlas and confirmed independently with direct electrical cortical stimulation, somatosensory evoked potentials and detection of HG response to tactile stimulation. Subjects were given visual cues to flex each finger or pinch the thumb and index finger. Movements were captured with a dataglove and time-locked with ECoG. A windowed covariance metric was used to identify the rising slope of HG power between two electrodes and compute time lag. Statistical constraints were applied to the time estimates to combat the noise. Rank sum testing was used to verify the sequential activation of cortical regions across 5 subjects.

Results: In all 5 subjects, HG activation in PM preceded S1 by an average of 53 ± 13 ms ($P = 0.03$), PM preceded M1 by 180 ± 40 ms ($P = 0.001$) and S1 activation preceded M1 by 136 ± 40 ms ($P = 0.04$).

Conclusions: Sequential HG activation of PM, S1 and M1 regions in preparation for movements is reported. Activity in S1 prior to any overt body movements supports the notion that these neurons may encode sensory information in anticipation of movements, i.e., an efference copy. Our analysis suggests that S1 modulation likely originates from PM.

Significance: First electrophysiological evidence of efference copy in humans.

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1. Introduction

Humans have remarkable ability to perform complex voluntary movements. These movements are results of complex interactions between different cortical areas including the primary motor cortex (M1), the supplementary motor area (SMA), the premotor cortex (PM), and the primary somatosensory cortex (S1) (Day et al., 1989; Ghez et al., 1997; Kutas and Donchin, 1974; Riehle and Requin, 1993; Rosenbaum, 1980; Wise, 1985; Rao et al., 1993; Christensen et al., 2007). Attempts have been made to capture the timing of activation of these areas in order to elucidate their roles in motor generation and control. For example, single neuron recordings in nonhuman primates showed the pathway by which visually guided reach movements are generated begins in the extrastriate visual cortex (Felleman and Van Essen, 1991) and passes through the parietal reach region (PRR) and area 5 to the dorsal premotor cortex (PMd) and then to the primary motor cortex (Johnson et al., 1991; Wise et al., 1997). In rhesus macaques, Soto and Fetz reported that cells in primary somatosensory cortex fire prior to active movements. This phenomenon demonstrated that S1 is an integral part of movement generation (Soso and Fetz, 1980). Also working with nonhuman primates, Nelson recorded changes in discharge rates in S1 that occurred early before movement onset; and hypothesized that S1 is modulated by other cortical region(s) before any movements take place (Nelson, 1987). In these animal studies it was demonstrated that S1 neurons were activated prior to the actual movements, which supported a concept important for somatosensory integration during voluntary movements known as the efference copy (EC).

Simply stated, an EC is an internal prediction of the sensory consequence of a volitional movement. The EC is thought to serve two important functions. Firstly, EC is used by the brain to distinguish sensory signals produced by one's own actions from those generated by external stimuli (Wolpert et al., 1995; Helmholtz, 2005). This influences sensations experienced during active movements (Chapman et al., 1987; Blakemore et al., 1998). Secondly, EC assists in control of movements when peripheral sensory feedback cannot reach the central nervous system fast enough (Miall and Wolpert, 1996). EC is an essential component of an internal forward model of motor planning and control, which has gained prominence due to emergent evidence from psychophysical and electrophysiological studies (von Holst, 1954; Jordan and Rumelhart, 1992; Miall and Wolpert, 1996). Electrophysiological evidence is however, confined to nonhuman primate models. Studies using single neuron recordings in these models have shown a change in firing rates of cells in S1 cortex prior to active movements (Soso and Fetz, 1980; Nelson, 1987), and the timing of this change is consistent with encoding of EC signals. A pathway from the superior colliculus to the frontal eye field via the thalamus has also been identified for conveying EC signals related to eye movements in monkeys (Sommer and Wurtz, 2002a). Evidence of EC has also been reported in a study of vestibular nuclei neurons encoding head angular velocity during passive and active head turns in rhesus monkeys (Roy and Cullen, 2001). Another phenomenon related to EC is the suppression of auditory response in self-generated sounds (Ford et al., 2001; Poulet and Hedwig, 2002).

Electrophysiological evidence of EC in humans is currently lacking. The timing of neural activation in different cortical regions relative to movement has been studied in humans using different imaging modalities. Using near-infrared spectroscopy (NIRS) signals and movement-related cortical potentials (MRCP), Sato et al., showed in humans that MRCP started approximately 1.5 s before electromyogram (EMG) change of finger movement. The NIRS data showed bilateral prefrontal cortex activations at approximately 0.5 s before EMG onset and bilateral dorsal premotor cortex activations at approximately 0.6 s before EMG onset. The hand area of the sensorimotor cortex was activated left-dominantly, peaking at approximately 3.7 s after EMG onset (Sato et al., 2012). Using magnetoencephalography (MEG) in humans, Nishitani and Hari reported, during imitation, left inferior frontal cortex led M1 in activation by 0.05–0.25 s (Nishitani and Hari, 2000, 2002). These modalities lacked the temporal resolution and spatial coverage to accurately study the interactions of different cortical regions.

Here, we used electrocorticography (ECoG) to record the timing of neuronal activation in premotor (PM), primary motor (M1) and S1 neural populations relative to cued finger movements in humans. An implanted subdural ECoG grid covers different cortical regions and is capable of detecting changes in activity produced by neuronal ensembles focal enough to distinguish somatotopic areas (Miller et al., 2007; Kubánek et al., 2009). Changes in high gamma (HG) power (>70 Hz) in ECoG have been shown to correlate with local cortical activity (Brindley and Craggs, 1972; Crone et al., 1998; Yoshor et al., 2007; Ray et al., 2008; Mesgarani and Chang, 2012). These broadband changes capture functionally specific cortical activity with a temporal resolution of milliseconds (Pei et al., 2011; Swann et al., 2013).

2. Materials and methods

2.1. Subjects and recordings

This study was a retrospective analysis of previously recorded ECoG data. Consecutive subjects were recruited at Harborview Medical Center and Seattle Children's Hospital between December 2010 and December 2012. Subjects gave written informed consent according to a protocol approved by the institutional review boards at both institutions. Five (1 female and 4 male) subjects were included in this study. The inclusion criteria included: (1) subjects had electrode coverage of at least two of the three cortical areas of interest, i.e., PM, M1, and S1, based on cortical mapping, and (2) subjects responded to the visual cue in a timely fashion, i.e., performing finger movements during the action phase of the task and waiting for the next cue during the resting phase. One subject was excluded for inability to respond timely to the movement cue. All subjects had ECoG grids placed for extended clinical monitoring and localization of seizure foci, in the course of the treatment of medically-refractory epilepsy. All electrode grids were constructed of 2.4 mm diameter exposed platinum–iridium pads embedded in silastic (AdTech, Racine, WI). In 4 subjects (subjects 1–3 and 5), the electrode spacing was 10 mm while the electrode spacing the grid used in subject 4 was 7 mm. Electrodes were arranged as 8 × 8 or 4 × 8 array. ECoG signals were obtained by splicing the clinical system's recording sampled at either at 1.2 kHz or 2.4 kHz using either SynAmps2 (Neuroscan, El Paso, TX, USA) or g.USBamp (Guger Tec, Graz, Austria).

2.2. Motor tasks

Subjects perform a screening motor task, which involves gross motor movements of either their tongue or their hand contralateral to the side of the electrode placement for 3 s, followed by a similar period of rest. This process was repeated between 10 and 30 times for each motion. Based on the appearance of movement-induced HG activities among electrodes and their relationship with gyral anatomy, we determined the electrode coverage of cortical regions. Only subjects with electrode coverage of at least two of the three cortical regions were asked to perform the cued finger movement task.

In the cued finger task, subjects were shown a silhouette of a hand on a computer screen at the start of each trial (Fig. 1). The
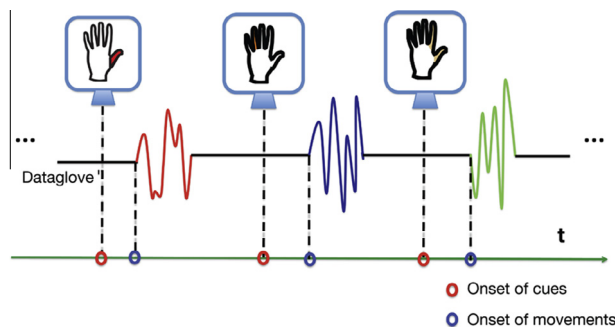
Fig. 1. Cued finger movement task. Subjects were shown a silhouette of a hand on a computer screen. At the start of each trial, subjects were instructed to flex the finger that was highlighted with a unique color on the hand silhouette or to make a pinching motion when both thumb and index fingers were highlighted. After 2 s of repetitive motion the screen became blank, signaling a rest period of 1 s. This was repeated between 10 to 30 times for each finger and same number of trials for the pinch. The trial types were alternated randomly so that each trial remained novel to the subject. Finger movements were recorded with a dataglove and time-locked with ECoG signals.

finger silhouette instructed the subject to flex the same finger that was highlighted with a unique color or make a pinching motion when both thumb and index fingers were highlighted. After 2 s of repetitive motion, the screen became blank signaling a rest period of 1 s. This was repeated between 10 and 30 times for each finger movement type. The trial types were alternated randomly so that each trial remained novel to the subject. A 22-degree-of-freedom dataglove (CyberGlove Systems, San Jose, CA) was used to capture finger movements sampled at 1.2 kHz. Neural signals and dataglove samples were collected and time-locked by the BCI2000 recording suite (Schalk et al., 2004).

2.3. Three independent methods of cortical mapping

Three independent methods for cortical mapping were utilized. In method 1, preoperative MR images were co-registered with postoperative CT scans using the statistical parametric Mapping software package (Penny et al., 2006). The cortical surface was first segmented from preoperative MR images using the FREESURFER environment (Dale et al., 1999). The electrode positions were calculated by segmenting them from a high-resolution postoperative CT scan registered with the preoperative MR imaging (Hermes et al., 2010). Each individual MR volume was registered to the Talairach brain atlas to allow an accurate localization of the electrode position (Talairach and Tournoux, 1988). The electrodes projected to the precentral gyrus of the common atlas are M1 electrodes (Brodmann area 4), whereas the electrodes projected to the postcentral gyrus are S1 electrodes (Brodmann areas 1–3). Electrodes located anteriorly to the M1 electrodes are PM electrodes.

In method 2, the accuracy of somatosensory mapping based on the common brain atlas, was verified by demonstrating that ECoG signals recorded by the electrodes in this region responded to sensory stimulation of the contralateral hand with increased HG power. In this experiment, the subject's hand contralateral to the grid implant site was repeatedly touched with an object. A signed squared cross-correlation value was calculated by comparing the distribution of HG power during hand stimulation with the distribution during rest (Miller et al., 2007); in subjects 1–4.

In method 3, attempts to record somatosensory evoked potentials (SSEPs) at the relevant electrodes and performed electrical cortical stimulation (ECS) at these electrodes was undertaken. SSEPs map the sensorimotor cortex by eliciting intrinsic cortical physiological responses to peripheral nerve stimulation at a rate of 5.1 Hz, with a pulse width of 200 ms and a current of 6 mA, most often of the median or posterior tibial nerves. The SSEP N20 waveform is recorded over the contralateral somatosensory cortex, whereas a P20 waveform with the opposite polarity is recorded over the primary motor cortex (Legatt and Kader, 2000; Ikeda et al., 2008). Referential recordings from adjacent electrodes over the pre- and postcentral gyri produce a "phase reversal" between the N20 and the P20, marking the site of the central sulcus, which is directly underneath or a few millimeters anterior to the area of highest amplitude. Electrode cortical stimulation was performed in awake subjects at the bedside by using an Ojemann stimulator (Integra, Plainsboro, NJ) to elicit movements by delivering up to 12 mA of current to pairs of implanted electrodes, and patients were monitored for movement or reproducible sensations of the hand, face, or leg.

2.4. Power spectral analysis

Offline data processing was performed with custom MATLAB software (Natick, MA). All ECoG data were re-referenced with respect to the common average. To minimize the trial-by-trial variability in task performance, the continuous recordings were segmented into epochs centered on movement onset, as measured by the dataglove. Time–frequency dynamic spectra were made using both a wavelet and the Hilbert transform approach. The wavelet approach used a Morlet wavelet (Goupillaud et al., 1984) to convolve with the voltage time-series to generate a time–frequency estimate for every frequency bin between 1 and 200 Hz. This approach allows for investigation of changes in the power spectrum density across all frequencies. For every electrode, we calculated statistically corrected activation weights (z scores) by comparing the power density during the activity phases with resting phases (Miller et al., 2007).

Since an increase in HG power in ECoG has been shown to correlate with local cortical activity (Brindley and Craggs, 1972; Crone et al., 1998), we employed the Hilbert transform approach to construct a time course of the HG power for all electrodes. The Hilbert transform of a voltage time-series produces a complex signal. The HG power time series can be computed by taking the norm squared of the Hilbert transform. Similar to wavelet approach, a signed squared cross-correlation value was calculated by comparing the distribution of HG power during movement with the distribution during rest (Miller et al., 2007). The statistical significance associated with the HG power distribution was estimated using a balanced, unpaired t-test. This allows selection of electrodes by applying a threshold. This step helped focus attention on the areas with HG activations.

2.5. A short-time windowed covariance (STWC) metric

With the HG power time-series, the time differences of neuronal activation between two electrodes could be estimated. This was achieved by capturing the rising phases of HG power of both electrodes and computing the time lag between these two events. A STWC metric was employed for this task (Blakely et al., 2014). To reduce computational cost, the HG power time series was down sampled by a factor of 10. To avoid aliasing, the time series were filtered with a low pass filter with a cutoff frequency of 5 Hz before down sampling. Each down-sampled signal was then normalized before the computation of the STWC metric (Van Albada and Robinson, 2007). A covariance calculation is performed with a small sample window of the source electrode x centered on sample t with multiple windows of a destination electrode y:

$$Q(x,y,t,\tau,\delta) = \sum_{i=t-\frac{\tau}{2}}^{t+\frac{\tau}{2}} \frac{(x_i - \overline{x})(y_{\delta+i} - \overline{y_\delta})}{\tau + 1}, \quad t \in [1, n] \text{ and } \delta \in [-\Delta, \Delta],$$

where, τ is the window size, n is the number of ECoG samples in a trial and δ is the number of samples by which the window on y is shifted (a negative value means the window on y is shifted ahead of window on x in time, whereas a positive value indicates a delay and Δ is the maximum of δ). For a pair of ECoG electrodes and a fixed window size τ, Q, the covariance, becomes a function of t and δ. For all subjects, we selected the number of samples in a half second to be the fixed window size τ and Δ. The selected windows in channel y are the same length of the channel x's window but shifted from $-\Delta$ to Δ, giving $(2 \times \Delta + 1)$ covariance measures for each sample point t. The covariance values were then averaged over trials. For each paired of electrodes, the STWC metric generates a two dimensional map, $Q_0(t, \delta)$, where $t \in [1, n]$ and $\delta \in [-\Delta, \Delta]$.

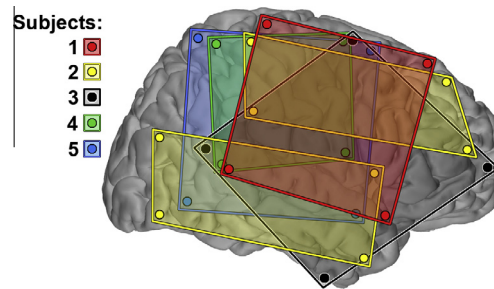### 2.6. Statistical significance of each value in Q_0

Statistical significance of each value of Q_0 was tested by applying a bootstrap method using surrogate signals generated with phase randomization (Nur and Wol, 2001). For two ECOG signals from a pair of electrodes, phase randomization was performed 500 times for each signal generating 500 pairs of surrogate signals. Applying the STWC metric to each pair of surrogate signals led to a new two dimensional map Q_i, where, $i \in [1, 500]$ By performing the phase randomization, we destroyed any temporal relationships existing between two ECoG frequency distributions and hence established a NULL hypothesis that there is no temporal relation between the two HG power time series. The statistical significance of each covariance value in Q_0 can be estimated by comparing it with the 500 trials in Q_i. Multiple testing when computing P value for each covariance value was also adjusted (Nichols, 2012). A covariance value with an associated P value <0.05 was deemed significant.

2.7. The time lag of HG activation between two electrodes

The task of estimating the time lag between the HG onsets in two electrodes involves finding statistically significant Q values and their corresponding t's and δ's. Having identified all statistically significant Q values ($P < 0.05$) and their corresponding t's and δ's, Q-value weighted means of both t and δ, $\bar{t}$ and $\bar{\delta}$, respectively were computed. $\bar{t}$ is now the time point where electrode x is activated in response to motor command and is the lag by which electrode y becomes activated relative to electrode x (a negative $\bar{\delta}$ indicates electrode y activates before electrode x whereas a positive $\bar{\delta}$ indicates the opposite). Both values can then be converted to time measurements by multiplying the sampling frequency and with the expressed relative to the onset of finger movements.

2.8. Statistical significance of the STWC-derived time lag of HG activations

To further limit the influence of noise, the statistical significance of a time lag of HG activation of two electrodes was estimated. The sequence of HG activation for any electrode pair

Table 1
Study subject demographics.

Subject	Subject demographics		
	Gender	Age (years)	Seizure etiology
1	Male	20	Post-trauma
2	Male	54	Post-trauma
3	Female	45	Focal cortical dysplasia
4	Male	18	Vascular malformation
5	Male	20	Focal cortical dysplasia

Fig. 2. The electrocorticography coverage of five subjects. All subjects had the 64-channel subdural electrode coverage of the sensorimotor area. This was achieved with two 4×8 electrode grids (yellow) in subject 2 and with one 8×8 electrode grid in the others. All electrode locations were registered, projected and plotted on the surface of right hemisphere of a common brain atlas. Only electrodes located at the corner of a grid are displayed here. Subjects 2 and 3 had left-sided coverage. All electrode grids were constructed of 3 mm diameter platinum pads embedded in silastic. In 4 subjects (subjects 1–3 and 5), the electrode spacing was 10 mm while the electrode spacing the grid used in subject 4 (green) was 7 mm. sub, subject.

can be estimated from the STWC metric or from the timing of activation of each electrode in the pair relative to a third electrode. These multiple estimates of the activation sequence between two electrodes were then subjected to binomial testing. The final activation time of an electrode is the mean of all statistically significant estimates ($P < 0.05$) for this electrode.

2.9. The timing of HG activation of each electrode

The estimation of activation time for an electrode site was therefore based on the *statistically* selected STWC results. Each STWC metric yielded one estimate of activation time for two electrode sites, x and y, involved in computation. The HG activation time of site x is $\bar{t}$ whereas the activation time of site y is $\bar{t} + \bar{\delta}$. The final activation time of an electrode site was then calculated by finding the mean of all statistically significant estimates for this site. Of the four subjects with reliable dataglove recordings, this time of activation was calculated relative to the onset of movements. In one subject for whom the dataglove recording was suboptimal, the time of activation was calculated relative to the onset of cues.
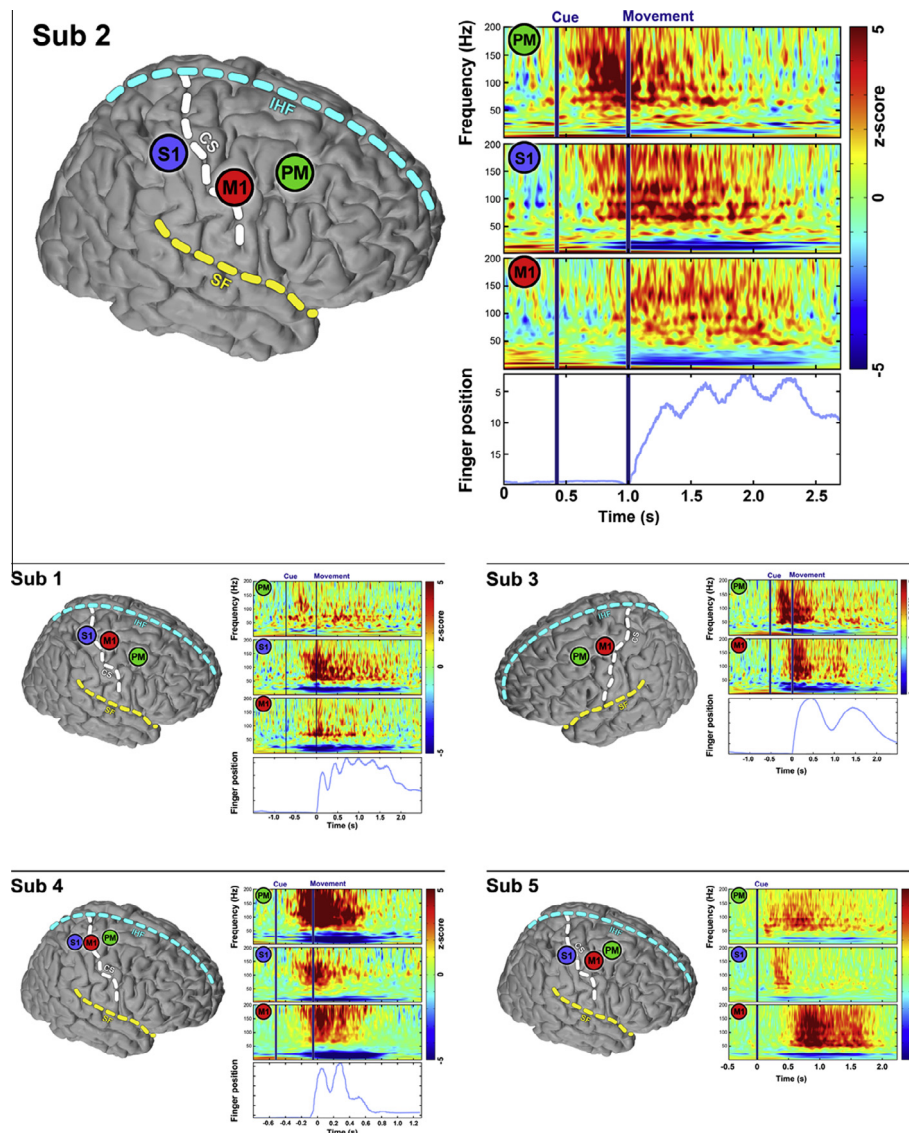
3. Results

Five subjects (aged 18–54 years; 1 female) with sensorimotor ECoG grid coverage met the study inclusion criteria. Demographics of these subjects are listed in Table 1. The composite grid coverage for all 5 subjects is shown in Fig. 2.

3.1. Changes in power spectral density

The primary study goal was to quantitatively estimate the relative timing of cortical activations in PM, S1, and M1 neurons, as measured by HG activity. Subjects were visually cued to flex particular digits or perform a pinch between thumb and forefinger. Signal dynamics were documented by the temporal and spatial characteristics of power spectral density (PSD) changes of ECoG signals (Fig. 3). In cued finger movements, these averaged PSDs showed an increase in HG power associated with movement onset and a decrease in β power (13–30 Hz) (Crone et al., 1998; Miller et al., 2007). Notably, spectral changes began before finger movements in all three regions. Moreover, these changes occurred sequentially in different cortical areas, beginning with PM, followed by S1, and ending with M1 sites. Finger movements were recorded with a dataglove for 4 of the 5 subjects. Sequential

Fig. 3. Sequential changes in power spectral density (PSD) among electrodes from three regions of the brain. The results from all 5 subjects are included. The subplot from subject 2 is shown in larger plot on the first row for better illustration. In each subplot, on the left; three electrodes from three different cortical regions, PM, S1, and M1 (except for subject 3, who has no electrode coverage of S1). On the right, the averaged power spectral density map shows that all three electrode sites exhibit decreased beta rhythms and increased broadband high gamma associated with movement. These changes all began before the movement (shown by averaged dataglove tracing over trials). Furthermore the broadband changes in PM electrodes occurred first and were followed by similar changes in S1 then M1. CS, central sulcus, SF, sylvian fissure, IHF, interhemispheric fissure.

changes in PSD among electrodes from three brain regions were observed in other somatotopically related electrodes across other individuals.

3.2. High gamma activation times estimated with a covariance metric

First the electrodes that had HG broadband changes associated with the finger movements were identified. For each electrode, the signed squared cross-correlation value of HG change associated with movements by comparing the distribution of HG power during movement with the same distribution during rest (Miller et al., 2007) was estimated. Only electrodes with statistically significant increase in HG power were included in the analysis. A short time windowed covariance (STWC) metric was used to identify the statistically significant STWC values ($P < 0.05$) yield the time lag of HG activation between a pair of sites and a specific timeframe when the correlation occurs. Combining these two values leads

to two HG activation times, one for each electrode of the pair. An example of estimating HG activation times of two electrode sites using STWC is depicted in Fig. 4. Note the PSD changes of these two electrodes related to finger movements shown in Fig. 3.

3.3. Cortical mapping

This step identified electrodes in three cortical regions, PM, S1, and M1. Three independent cortical mapping methods yielded congruent results. Results of brain mapping from all 5 subjects are shown in Fig. 5. HG activation times for all electrodes with sufficient HG response to the finger movements using STWC were computed. The estimated HG activation times of each selected electrodes from all subjects are shown in Fig. 6. Note in all subjects, the HG activation began with PM, followed by S1, and lastly by M1 sites after the electrodes have been assigned to one of the sites based on cortical mapping.
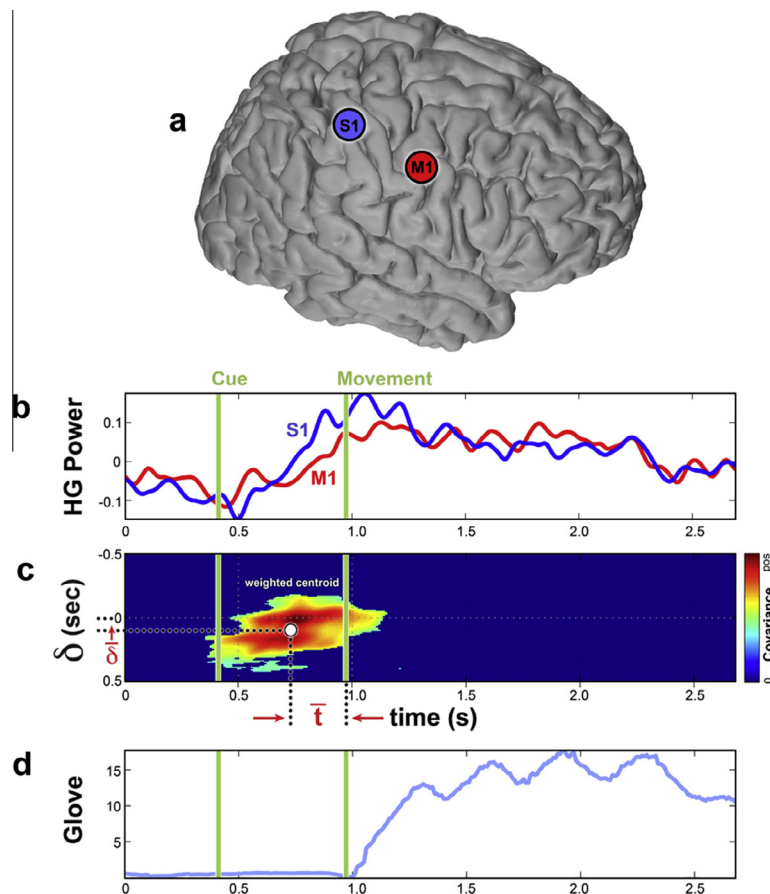
Fig. 4. An example of estimated HG activation times for a pair of electrodes using the short-time windowed covariance (STWC) metric. (a) The location of the two electrodes on the brain. (b) The two normalized HG power time series averaged over trials. (c) The 2-D map $Q(t, \delta)$ after statistically insignificant covariance values has been set to zero. (d) Averaged dataglove tracing over trials. Computing a weighted centroid of all statistically significant covariance values leads to two time estimates $\bar{\delta}$ and $\bar{t}$. $\bar{\delta}$ is the estimated time lag of the HG activation between these two electrodes. Here the $\bar{\delta}$ has a positive number, which indicates the S1 site activates earlier than the M1 site. $\bar{t}$ is the HG activation time of the S1 site relative to the onset of movements. The HG activation time of the M1 electrode relative to the onset of movements is then $\bar{t} + \bar{\delta}$. (d) The averaged finger position recorded with dataglove.

3.4. The activation time of each cortical region

The activation time of each region was computed for each subject by determining the mean HG activation times among the sites localized to each region, Fig. 7 and Table 2. Among the 4 subjects (subjects 1, 2, 4, and 5) who had electrode coverage of all three cortical regions, the HG activation began in PM, was followed by S1 and lastly by M1 sites. In subject 3, the HG activation in PM led that in M1 neurons.

To demonstrate that finger movement type has little influence on the results, subject 5 was asked to perform the same motor task with each finger movement repeated 30 times in random order. The activation time of each of the three cortical regions was then computed using the same method outline above but with only signals from the same finger movement included in the analysis. As shown in Fig. 8, the activation of S1 consistently preceded that of M1 in all six types of finger movements.

3.5. Time lags of the cortical activation among three regions

The time lag of HG activation between two cortical regions was calculated by finding the difference of the mean of HG activation time of electrodes in each region. The time lag of cortical HG activation was calculated for each subject and then averaged across subjects. The HG activation in the PM region leads that in the S1 region by an average of 53 ± 13 ms ($P = 0.03$, paired-sample t-test),

PM leads M1 by 180 ± 40 ms ($P = 0.001$) and activation in S1 leads M1 by 136 ± 40 ms ($P = 0.04$; Fig. 9).

4. Discussion

The purpose of this study was to characterize the relative activation times of three cortical regions: PM, S1, and M1, in preparation of cued finger movements in human subjects. Increase in HG activities was used as a measure of the cortical activation and estimated the activation time of these three regions in relation to each other and relative to the movements. Using a data-driven approach with a novel covariance metric and a series of statistical constraints, a significant temporal relationships of changes in HG power demonstrated the sequential HG activation of PM, S1 and M1 regions in preparation for movements.

4.1. Electrophysiological evidence of efference copy in humans

A significant study result was the identification of early involvement of sensory cortex in motor planning. S1 cortex was activated not only before peripheral muscles and movements, but also before the activation of M1 cortex. This activation of S1 prior to motor efference, and well in advance of any sensory input, is consistent with models of movement that require prediction of sensory changes prior to a planned (finger) movement, i.e., EC. In humans, psychophysical studies have demonstrated that sensory predictions
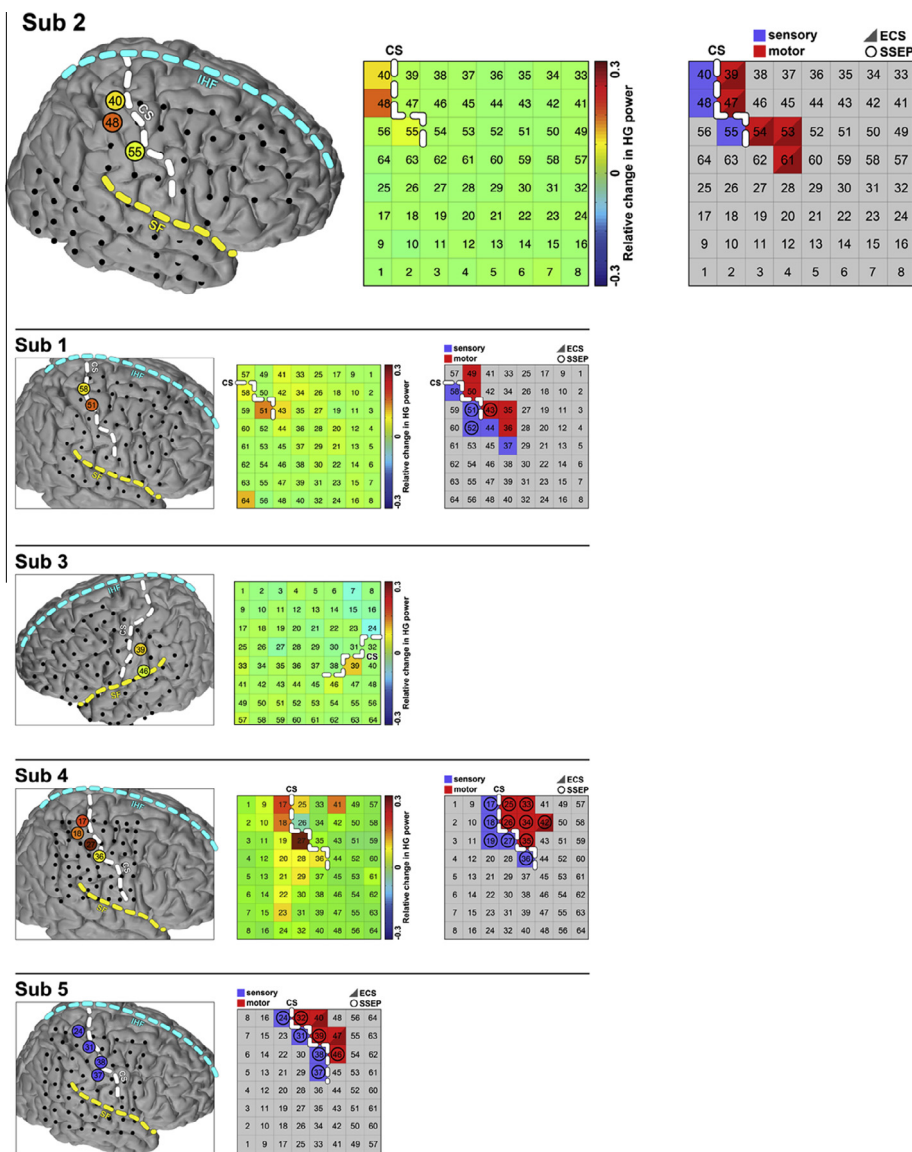
Fig. 5. Confirmation of the cortical mapping. The results from all five subjects are shown here. The results from subject 2 are shown in the first row in bigger plot for better illustration. In this subplot, on the left, all 64 electrodes were registered, projected and plotted with a common brain atlas. Electrodes 40, 48 and 55 were localized to S1. In the middle, changes in HG power among all electrodes when subject's contralateral hand was touched with an object relative to resting baseline. Electrodes 40, 48 and 55 had highest increase of HG power associated with sensory stimulation of the contralateral hand. On the right, the location of central sulcus (CS) relative to the grid is again confirmed based on the electrical cortical stimulation (ECS, triangle), where cortical stimulation of electrodes 39, 47, 54, 53 and 61 produced motor response. In this subject, somatosensory evoked potentials (SSEPs, circle) yielded reliable mapping results. The results from the other subjects are shown in the panels below. In subject 3, neither ECS or SSEPs identified any electrode covering hand motor or sensory area. The grey grid is therefore omitted for this subject. In subject 5, the confirmation was only conducted with ECS and SSEPS, the green grid is therefore omitted for this subject. Note that, these three different mapping techniques yielded congruent identification of central sulcus (CS) in all 5 subjects. Sub, subject.

precede motor acts (Johansson and Cole, 1992; Flanagan et al., 2003; Gritsenko et al., 2007) and lesion studies have demonstrated the importance of EC in movement control (Duhamel et al., 1992; Gaymard et al., 1994; Leiguarda and Marsden, 2000). A study conducted with functional magnetic resonance imaging (fMRI) in humans has shown that S1 neurons are active in voluntary movements in the absence of peripheral sensory inputs (Christensen et al., 2007). In the same study, Christensen et al., hypothesized that S1 is likely modulated by the premotor area based on the correlation of blood oxygen level dependent (BOLD) signals in these two regions (Christensen et al., 2007). Here, we use human electrophysiological evidence of the temporal characteristic of S1 activation to support the notion of both EC and the internal forward model.

Electrophysiological evidence of EC is not only important for understanding human motor physiology, EC is also implicated in human cognitive and behavioral functions. The existence of EC implies that the sensory system is not only influenced by environment but is also modulated by the brain. It is through EC one can distinguish the owner of a motor action, i.e., the sense of agency (Farrer et al., 2003; Tsakiris et al., 2006). EC explains why we cannot tickle ourselves (Blakemore et al., 2000).

While we believe that we have captured neuronal activities related to EC, it is possible that in the experiments described only one particular type of EC at a specific time point, i.e., the moment immediately before the movement onset is captured. As shown in previous studies, neuronal firing can either increase or decrease prior to the movement depending on the regions in S1 (Nelson,
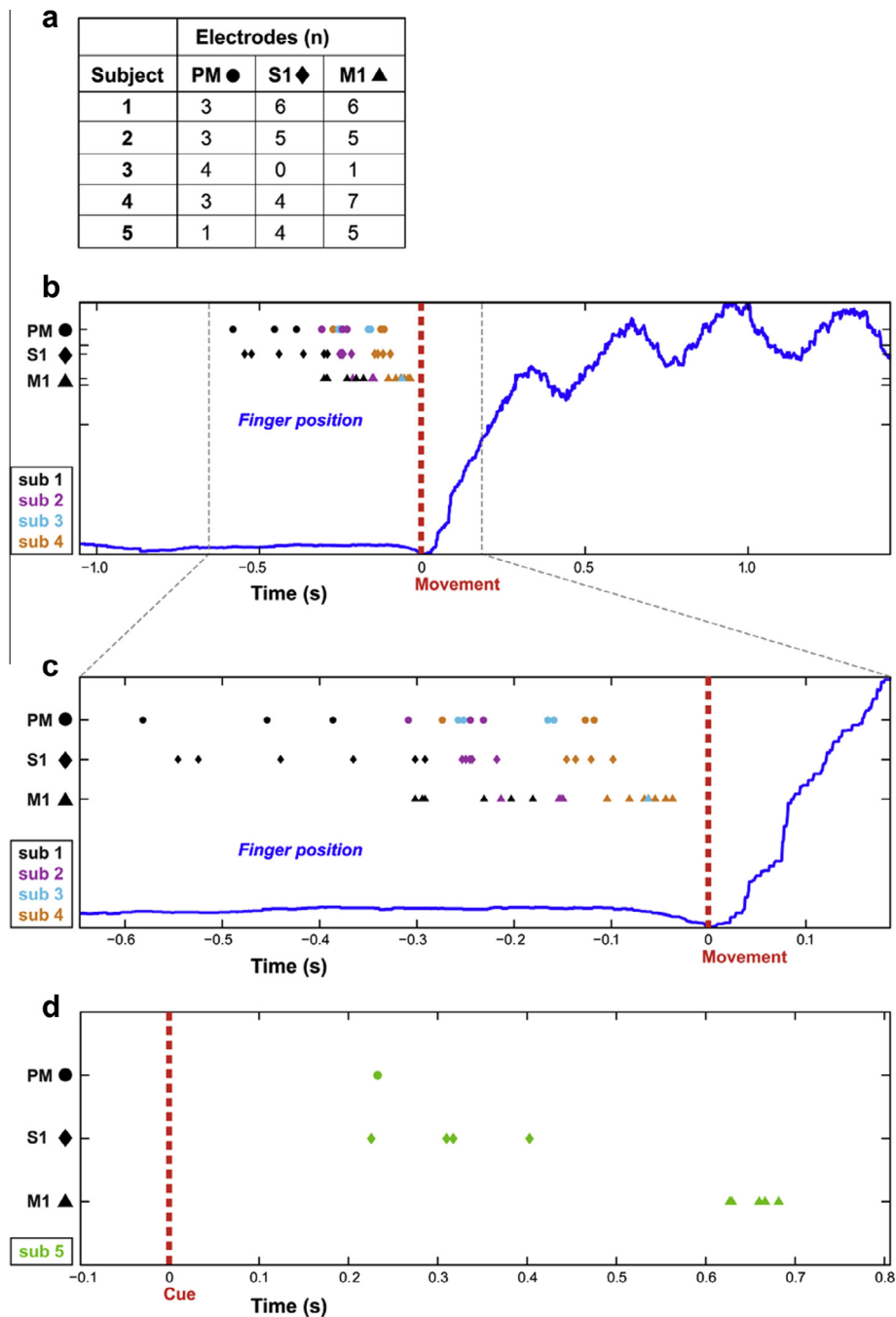
Fig. 6. The activation times for all electrodes with HG response related to movement from all 5 subjects. The shape of each electrode symbol corresponds to its location relative to three cortical regions: The circular electrodes are from PM, the diamond electrodes are from S1 and the triangular electrodes are from M1. The color of electrodes indicates which subject they belong to. (a) Are numbers of electrodes with sufficient change in HG power in each cortical regions among all 5 subjects. (b and c) Are results from subjects 1 to 4 and the activation times were computed relative to the onset of finger movement (at time zero). C is a magnified portion of B to better illustrate salient features. (d) The results from subject 5 and activation times computed relative to onset of visual cue (at time zero).

1987; Chapman, 1994). The interplay between the EC and peripheral sensory afferent is also complex, which can lead to different sensory perceptions by the subject. Comparing with externally generated stimuli, some studies reported reduction (i.e., gating) (Blakemore et al., 1998; Baess et al., 2009) of neuronal activities in S1 following self-initiated motor acts, while others suggest enhancement of these activities (Hughes and Waszak, 2011).

4.2. PM, rather than M1, modulates S1 before movements

This study demonstrates the existence of EC-related S1 activation and also shows that this activation precedes the activation

of M1 and is preceded by the activation of PM. The timing of these events suggest that, immediately prior to movements, S1 neurons are likely modulated by PM rather than M1 neurons. Christensen et al., have proposed this relationship based on the correlation of BOLD signals in fMRIs between these two regions during human voluntary movements (Christensen et al., 2007). The results obtain in nonhuman primates are different. In two separate studies conducted in awake macaques, the timing of neuronal activation in precentral (Fetz et al., 1980) and postcentral (Soso and Fetz, 1980) gyri was documented during active and passive elbow movements. Similar to our observations, they showed that the neurons were activated in both precentral and postcentral gyri
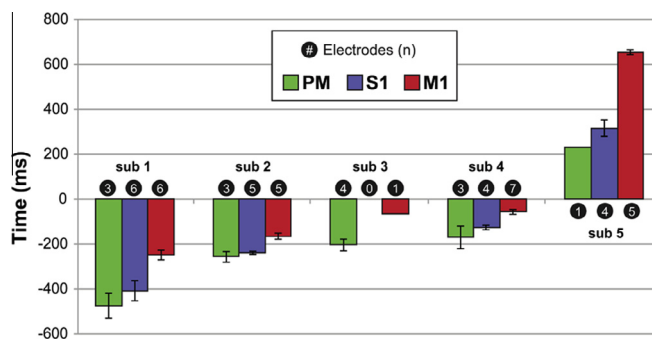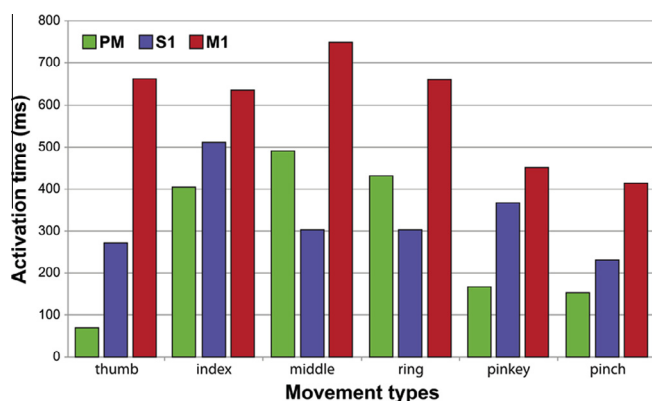
Fig. 7. Sequential activation of PM, S1 and then M1. The activation time for each cortical region is the mean activation time of all electrodes (*n*) in the region. Subject 3 had no S1 coverage. Error bars are the standard error (SE) of the mean of the sample. No error bar is calculated for a sample of one. The activation times for subjects 1–4 were calculated relative to the onset of movement (negative values) whereas the activation times for subject 5 are relative to the onset of cue (positive values). sub = subject.

Table 2
Means and standard errors of the means (SE) of the HG activation time (ms ± SE) of the electrodes localized to each cortical region.

Subject	Region HG activation time (ms ± SE)		
	PM	S1	M1
1	−474 ± 57	−411 ± 45	−250 ± 21
2	−261 ± 24	−241 ± 6	−163 ± 13
3	−209 ± 26		−62
4	−172 ± 50	−126 ± 11	−59 ± 12
5	233	314 ± 36	653 ± 11

Subject 3 had no S1 electrodes. SE was not computed for the results obtained with only one electrode.

Fig. 8. The activation time of three cortical regions computed for each finger movement type. The sequential action of PM, S1, then M1 were maintained in flexion of thumb, index, and little fingers and thumb-index pinch. With middle and ring finger flexion, the order of activation among the three regions was S1, then PM, and lastly M1.

before EMG activities. In these experiments, the average onset of neurons in precentral gyri preceded average activation of postcentral neurons. Future experiments that utilize the same recording modality and similar motor tasks need to be conducted both in humans and primate models to elucidate the cause for this discrepancy. A study of the modulation of cortical potentials evoked by electrical stimulation of a cutaneous nerve in an instructed delay task showed early modulation in PM and M1 prior to the active movement, consistent with the predictions of an EC signal (Seki and Fetz, 2012).

It is important to emphasize that the phenomenon reported here is limited to the transient period immediately prior to finger movements in these particular motor tasks. S1 neurons can be modulated by M1 neurons in other settings. For example, a study has shown that electrical stimulation of M1 neurons changes the firing patterns of S1 neurons and hence the sensory perception (Tsubokawa et al., 1991). In addition, the motor task we used is based on visual cues. For future studies it will be important to investigate timing of cortical activations during spontaneous and reactive (cued) movements. Type of cues may also influence pattern of cortical activation. Similar studies can be conducted with auditory cues.

The sequential activation of PM, S1 and M1 neurons documented here suggests a possible pathway by which the EC is transmitted from PM to S1 rather than from M1 to S1. The EC-related activation is not limited to these two regions. A fMRI study showed that the forward model of motor control involved the cerebellum (Kawato et al., 2003). One well-studied pathway for visually guided reach movements begins in the extrastriate visual cortex (Felleman and Van Essen, 1991) and passes through the parietal reach region (PRR) and area 5 to the dorsal premotor cortex and then to the primary motor cortex (Johnson et al., 1991; Wise et al., 1997). Other structures possibly involved include cortical regions such as supplementary motor area (SMA) (Fried et al., 2011) and subcortical structures such as basal ganglia and thalamus. Our methodology, when combined with depth electrodes, can be applied to documenting interactions of these structures to further elucidate the pathway for EC.

4.3. Implications for neuroprothesis design

Better understanding of how sensory information is integrated in initiation and control of movements is important for developing robust neuroprostheses. Historically, brain computer interface (BCI) device development has focused on using brain activities in the conventional motor cortices, although volitional control of postcentral cells has also been demonstrated (Moritz and Fetz, 2011). Evidence has emerged that both pre and post-central gyri continue to be activated in individuals with chronic spinal cord injury during attempted movements (Shoham et al., 2001; Cramer et al., 2005; Hotz-Boendermaker et al., 2008). The study confirms that sensory cortex activation is an integral part of motor planning that is independent of feedback from peripheral afferents. These signals may still be present in individuals with disease processes that selectively affect M1 such as focal ischemic stroke or motor neuron diseases. A natural extension of this study is to investigate interactions of these cortical regions during performance of imagined movements.

The presence of EC-related S1 activation provides a potential means for the implementation of somatosensory feedback in BCI applications. Currently application of BCI mostly relies on visual feedback information to users. There is, however, an increasing need to include other forms of sensory feedback to BCI devices (Schwartz et al., 2006; Nicolelis and Lebedev, 2009; Johnson et al., 2013). Some groups have proposed adding sensory modalities such as vibrotactile (Chatterjee et al., 2007), proprioceptive (Ramos-Murguialday et al., 2012), and auditory (Nijboer et al., 2008) feedbacks to BCI devices to increase their efficacy and reduce user training time. By identifying the electrophysiological correlate of the EC using ECoG, we have separated anticipatory sensory signals produced by BCI users from those generated by external stimuli as potential inputs for BCI devices. Comparison of these two signals in real time may lead a sensory feedback paradigm.

4.4. Study limitations

It is important to consider the limitations associated with human ECoG studies. All subjects included here were undergoing surgical
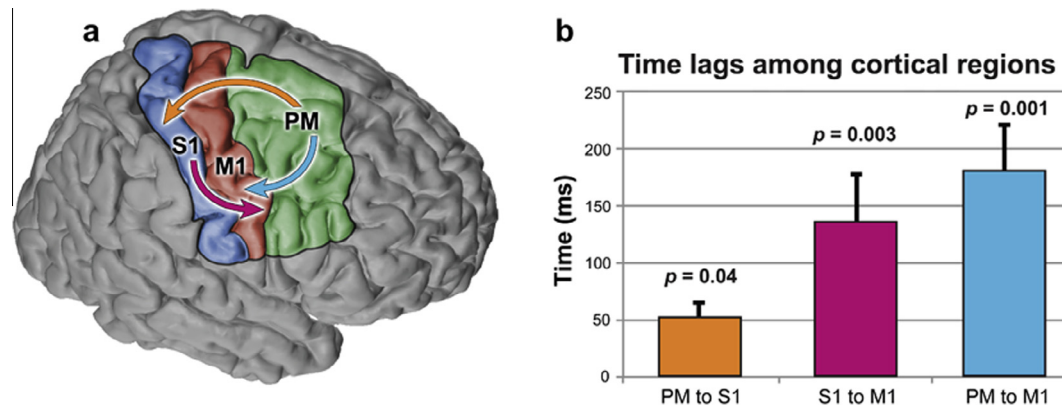
Fig. 9. Time lags of cortical activation among the three cortical regions, PM, S1, and M1. (a) The sequential activation of the three cortical regions; arrows originate in region that leads the region where the arrow points. (b) Bar graph shows that, across all 5 subjects, the HG onset in PM leads that in S1 by an average of 53 ± 13 ms ($P = 0.03$, paired-sample t-test), PM leads M1 by 180 ± 40 ms ($P = 0.001$) and S1 lead M1 by 136 ± 40 ms ($P = 0.04$). Error bars are standard errors of the means.

treatment of medically intractable epilepsy. The intracerebral pathology involved with this subject population should be kept in mind when making extensive generalizations of the study results.

Study analysis revealed that the sequence of activation from PM, then S1 and finally M1 is consistently preserved across subjects. Consistency in results was achieved when human physiological noise data including ECoG and dataglove recordings were averaged over a large number of repeated finger movements (>90 trials). For example, despite a high sampling rate (1.2 kHz), it is not possible to accurately define finger position from dataglove recording during one movement trial. When these recordings were averaged over a large number trials, a definite pattern of finger movement emerges, Fig. 3.

In subject 5, our results showed that PM lagged behind S1 in activation when only signals from the middle finger movements (30 trials) were included in the analysis, Fig. 8. We believe that this aberrant result likely reflects the impact of signal noise in analysis rather than true timing of activation among these cortical regions.

We defined PM electrodes as those anterior to M1. PM subdivisions have been well studied in nonhuman primates and include areas such as PM dorsal rostral; PMDr (F7), PM dorsal caudal; PMDc (F2), PM ventral, dorsal; PMVr (F5), PM ventral, caudal; PMVc (F4) and frontal eye field (FEF) (Weinrich et al., 1984). The function of each region is complex and has undergone intense investigation. Sommer and Wurtz demonstrated that FEF is involved in EC during eye movements (Sommer and Wurtz, 2002b). Depending on the grid coverage in each patient, the number and the actual location of a PM electrode vary among these PM subdivisions. Additionally, the intracerebral pathology involved with this subject population makes it unreliable to infer the function of a small cortical region based on anatomy alone.

Our analysis revealed that the actual timing of cortical activations varied considerably among the 5 subjects, Table 2. The most essential study finding is the relative temporal relationship of HG onsets among the three cortical regions rather than the absolute timing of each event. A factor that may contribute to the variability in estimates of actual onset of HG power may be the subject's ability to react to the visual cue. This ability is affected by but not limited to factors such as patient age, medical history, underlying brain pathology, medications, time of the day, and hospital environments. Together these variables likely account for some of the variability among subjects. A second factor is our method of using the STWC metric for the estimation of timing of HG onset contributes to the variability of actual activation times. The STWC metric was designed and implemented and associated with statistical constraint to minimize the impact of noise on electrophysiological signal results. By design, this metric is more robust in estimating the time lag between changes in HG power in two cortical regions (Blakely et al., 2014) but is less accurate in estimating the time lag between the change in HG power and finger movements, which leads to more variability in the results. The inclusion of additional study subjects could certainly address these study limitations, although our conclusions would likely remain the same.

5. Conclusion

We have used ECoG to measure the time of activation in human PM, M1 and S1 neurons in preparation for finger movements. These measurements have revealed that the onset of finger movements is preceded by the increase in HG power in all three cortical regions. Furthermore, the increase of HG power begins in PM and is followed by S1 and then at last by M1. The timing of these events suggests that the anticipatory sensory activation of S1 is likely related to the EC and this activation of S1 is a result of modulation by PM rather than by M1. These findings provide the first electrophysiological evidence of EC in humans and strengthen the argument for an internal forward model for the control of human voluntary movements. Collectively, the results further better understanding of human motor and cognitive function and can help unlock the answers to disorders such as motor apraxia and mental illness such as schizophrenia. Additionally, a better understanding of how sensory information is integrated in initiation and control of movements is a key component in the development of robust neuroprosthesis for stroke or spinal cord injury patients.

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