Cortical activity during motor execution, motor imagery, and imagery-based online feedback

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Edited* by Riitta Hari, Helsinki University of Technology, Espoo, Finland, and approved January 14, 2010 (received for review November 26, 2009)

Imagery of motor movement plays an important role in learning of complex motor skills, from learning to serve in tennis to perfecting a pirouette in ballet. What and where are the neural substrates that underlie motor imagery-based learning? We measured electrocorticographic cortical surface potentials in eight human subjects during overt action and kinesthetic imagery of the same movement, focusing on power in “high frequency” (76–100 Hz) and “low frequency” (8–32 Hz) ranges. We quantitatively establish that the spatial distribution of local neuronal population activity during motor imagery mimics the spatial distribution of activity during actual motor movement. By comparing responses to electrocortical stimulation with imagery-induced cortical surface activity, we demonstrate the role of primary motor areas in movement imagery. The magnitude of imagery-induced cortical activity change was ~25% of that associated with actual movement. However, when subjects learned to use this imagery to control a computer cursor in a simple feedback task, the imagery-induced activity change was significantly augmented, even exceeding that of overt movement.

Results

Movement. In an initial set of experiments, the subjects (Table S1) performed an interval-based task in which they alternated between several seconds of either hand or tongue movement and several seconds of rest in response to visual cues. We examined changes in the power spectral density (PSD) of the ECoG potential from all electrodes during and after the cue period. As with EEG and MEG studies (17, 18), we found that, for a given movement type, the spatial distributions of motor-associated α and β rhythms (captured jointly here in an 8–32 Hz band) in lateral frontoparietal cortex overlap between overt movement and imagery. However, when comparing different movement types, we quantitatively show that these also overlap, demonstrating that the α/β-rhythm changes do not delineate local cortical function. In contrast, the spatial distribution of the high-frequency aspect of the ECoG signals does not overlap between movement types but does overlap between movement and imagery of the same type. This finding unambiguously establishes a shared representation for movement and imagery at the local population level. The role of primary motor areas in movement imagery was revealed by significant imagery-induced cortical surface activity at electrode sites where electrocortical stimulation produced movement. The magnitude of imagery-induced cortical activity was ~25% of that of actual movement. When this same imagery was used to control a cursor in a simple feedback task, we found an augmentation of spatially congruent cortical activity, even beyond that found during movement.

Brain–computer interface | electrocorticography | primary motor cortex | learning | plasticity

Imagined motor movement (“imagery”) has been shown to be crucial for motor skill learning in a variety of situations, ranging from learning new skills in sports (1) to overcoming the effects of neurological conditions (2, 3). Demonstration of residual cortical activity during imagery in motor areas in paraplegic individuals (4) and stroke victims (5) implies that motor imagery could play an important role in rehabilitation and prosthesis control (6). Because of this, many recent efforts to build “brain–computer interfaces” (BCIs) for paralyzed patients have relied on motor imagery to obtain volitional neural signals to control external devices (6–11).

Human brain imaging using hemodynamic markers (PET and fMRI) and extracerebral magnetic and electric field studies (MEG and EEG) have shown that motor imagery activates many of the same neocortical areas as those involved in planning and execution of motor movements (e.g., medial supplemenetual motor area, premotor cortex, dorsolateral prefrontal cortex, and posterior parietal cortex). The direct, population-scale, electrophysiologic correlate of these findings has not been quantified at the cortical surface. The relevance of primary motor cortex in motor imagery has remained an unresolved issue, there being evidence both supporting and against a role for primary motor areas during motor imagery (6, 12–19). Furthermore, the congruence of cortical electrophysiologic change associated with motor movement and motor imagery has not been determined.

Crone and colleagues, in 1998 (20), and others since then (21–24) have demonstrated that somatotopic motor function is revealed in the high frequencies of the cortical surface potential power spectral density (PSD). This cortical surface potential has been examined in humans with electrocorticography (ECoG), which employs the clinical placement of electrode arrays on the brain surface. Recent ECoG studies have demonstrated that these signals can characterize local cortical dynamics with very high spatiotemporal precision: for example, the independent dynamics of individual fingers can be resolved at the 20-ms time scale in single electrodes (25). This is done by capturing noise-like, 1/ν changes in the PSD of the cortical surface electrical potential (26), which directly correlate with population firing rate (27) and are plainly revealed at high frequencies (22, 25).

We apply ECoG to address the problem of imagery-associated cortical activity. As with EEG and MEG studies (17, 18), we find that, for a given movement type, the spatial distributions of motor-associated α and β rhythms (captured jointly here in an 8–32 Hz band) in lateral frontoparietal cortex overlap between overt movement and imagery. However, when comparing different movement types, we quantitatively show that these also overlap, demonstrating that the α/β-rhythm changes do not delineate local cortical function. In contrast, the spatial distribution of the high-frequency aspect of the ECoG signals does not overlap between movement types but does overlap between movement and imagery of the same type. This finding unambiguously establishes a shared representation for movement and imagery at the local population level. The role of primary motor areas in movement imagery was revealed by significant imagery-induced cortical surface activity at electrode sites where electrocortical stimulation produced movement. The magnitude of imagery-induced cortical activity was ~25% of that of actual movement. When this same imagery was used to control a cursor in a simple feedback task, we found an augmentation of spatially congruent cortical activity, even beyond that found during movement.
Imagery. The subjects then repeated the movement task, except that instead of moving they were instructed to kinesthetically imagine making that movement (19, 29) during the cue period (nonmovement verified by EMG, dataglove measurement, and observation; Fig. S4). In 4 of the 5 subjects who received electrocortical stimulation (ECS) as part of their clinical care, imagery produced a spatially focal, statistically significant (P < 0.05) increase in HFB power in electrodes where ECS produced movement of the same type (Figs. 1 and 4 and Figs. S1, S2, and S5). This confirms that primary motor areas are activated during imagery.

The spatially broad decrease in power in the LFB and spatially focal increase in power in the HFB were observed during imagery and were significantly overlapping with their movement counterparts, but the magnitudes of the spectral changes were smaller (Figs. 1, 2, 4, and 5 and Figs. S1, S2, S6, and S7). Across the 8 subjects, 38 electrodes were selected to quantify the magnitude of spectral change: those in which there was a significant difference in the PSD for both the HFB and LFB associated with movement (P < 0.05 t test, Bonferroni-corrected for the number of electrodes in that subject; 21 electrodes for tongue, 17 for hand). In these electrodes, the magnitude of spectral change in the HFB for imagery was 26% of that during actual movement (Fig. 2). For the LFB, the relative magnitude was 49%. The LFB change was significantly larger than the HFB change (P = 0.005 by permutation resampling, 1010 iterations).

The spatial overlap between movement and imagery was strong and significant across subjects in both the HFB (P < 0.05 in 13 of 14 cases, and P < 0.01 in 9 of 14 cases, by reshuffling [Fig. S3]) and the LFB (P < 0.05 in 11 of 14 cases and P < 0.01 in 9 of 14 cases). The movement–imagery overlap was not significantly different for the HFB than the LFB on a case-by-case basis (P = 0.60, t test of difference in overlap metric).

Feedback. Four subjects (S1, S2, S6, and S8) participated in an imagery-based learning task, in which the magnitude of cortical activation at a particular electrode was used to control a cursor on a screen during motor imagery. Based on the cortical changes seen in the simple movement/imagery tasks, amplitudes at particular electrodes and frequencies were chosen as features that controlled the speed and direction of 1-dimensional movement of a cursor on a computer screen (Fig. 3 and Fig. S8). Subjects rapidly (5–7 min) learned to accurately control the cursor by using motor imagery. In this feedback task, subject 1 attained 94% target accuracy using imagined word repetition (imagining repeating the word “move”); subject 2, 74%/90% with hand/ tongue imagery; subject 6, 85% with shoulder imagery; and subject 8, 100% with tongue imagery. During this learning task, the spatial distribution of high-frequency ECoG activity was quantitatively conserved in each case, but the magnitude of the imagery-associated spectral change increased significantly (Figs. 3–5 and Figs. S6 and S7). In most cases, the spectral change associated with cursor control exceeded that observed during
after several (5–8) minutes of training, subjects 1 and 6 reported that motor imagery ceased and was replaced by thinking about moving the cursor up or down.

**Discussion**

Similar to previous findings (20, 22, 23), we found a spatially broad decrease in power in the LFB (8–32 Hz) of the PSD during movement (Fig. 1), consistent with movement-induced desynchronization of the motor-associated frontoparietal α and β rhythms (30). With imagery, there is a similar, spatially broad, LFB decrease that significantly overlaps that associated with the overt movement. This might have been anticipated from MEG- and EEG-based imagery studies that found similar patterns of desynchronization between movement and imagery (17, 18).

The green line denotes $P_0$, the mean power across passive/active trials. The black line indicates a “discriminative index”; i.e., the smoothed difference between the mean power during the previous three active target trials and the previous three passive target trials. This index demonstrates that target accuracies (shown in C) were highest when the subject found a middle dynamic range. After the third run, the subject reported having ceased to perform imagery, and instead “thought about the cursor moving up or down to get it to move” at some point during the run. (C) Distribution of HFB (upper brain plots) and LFB activations, as well as target hit accuracies (% next to run number), during each of the four experimental runs. All activation maps are to the same scale (indicated by the color bar). The final activations are most prominent at the electrode that was used for cursor control. The number flanking each brain plot is the maximum (absolute value) activation.
desynchronization may reflect something entirely different, such as altered feedback between cortical and subcortical structures (31–33), with a time scale of interaction that corresponds to the peak frequency in the PSD. The 50% LFB decrease in power during imagery might then represent only a partial “release” of cortex by subcortical structures (partial decoherence of a synchronized corticothalamic circuit; see Fig. 1C), compared with a more complete release during actual movement or after feedback.

In contrast, HFB (76–100 Hz) change is specific for different movement types. With movement, we see characteristic spatially focal increases in activity that are somatotopically specific: the resulting hand and tongue activation maps do not overlap (Figs. 1 and 2 and Fig. S1). The electrodes with statistically significant HFB change include primary motor cortical sites, where electrocortical stimulation (ECS; passing current between paired electrodes) produced corresponding hand or mouth movement. This HFB change is reflective of a broadband PSD increase that is obscured at lower frequencies by the motor-associated α/β rhythms (22). Indeed, choosing a wider frequency band for the HFB to capture more of this broadband change (extending well above 100 Hz) has been shown to produce the same result in motor cortex during movement (25, 34, 35). Through both simulation and experiment, this broadband change has been specifically correlated with local population firing rate (25–27, 36).

During hand or tongue imagery, there is a similar increase in HFB power relative to rest in a spatial distribution of electrodes that significantly overlaps with the movement-related distribution. Coupled with recent findings that suggest that PSD amplitude changes linearly with neuronal firing rate (26, 27, 36), our finding of a 25% power increase with imagery, relative to movement, may imply that there is up to a 4-fold greater population firing rate during movement than imagery. Multiple mechanisms at the microcircuit level may be responsible for this. For example, a larger proportion of neurons in the population may be recruited during movement, of which only that subset that does not induce movement is active during imagery. This subset will not reflect the influence of recurrent somatic feedback, which is present during movement and absent during imagery. Given that motor imagery was recently shown to enhance corticospinal excitability (37), the 25% power increase may suggest that a subset of the neuronal population is activated during imagery, and that this subset “primes” the subset of neurons that send motor commands to the body. Alternately, the same population may be active throughout but with a higher firing rate in each neuron during movement or feedback. Recent work by Schieber and colleagues (38) showed that the relation between firing rate of single corticomotoneuronal cells and target muscle activity is not fixed but is context-dependent. This suggests that increases in the firing rate of large populations of motor cortical neurons may contribute to the HFB activity seen during imagery and feedback without inducing muscle contraction. Furthermore, although the precise relationship between fMRI-based BOLD activity and neuronal population activity is not straightforward (39), the ~25% HFB shift is similar to motor-imagery BOLD changes that were ~30% of those during movement (15).

Because the 1-cm spacing of the clinical ECoG array samples the cortical surface at roughly the scale of the somatotopic representation of different limbs in motor areas (40), this spacing is appropriate to reveal congruence in large-scale activation between motor imagery and overt movement, and imagery-based feedback. Our results show that these representations are quantitatively overlapping, suggesting that motor imagery involves the same distributed circuits as movement. The presence of imagery-based HFB spectral change in electrodes where electrocortical stimulation produces movement identifies a concrete role for primary motor cortex in imagery. The exposed cortical surface is sometimes dismissed as not belonging to primary motor cortex (i.e., that Brodmann area 4 is only within the bank of the central sulcus, and that only Brodmann area 6 is exposed and is a purely premotor region). This is a view that has been refuted: first, the Betz cells that define primary motor cortex can be found on the convexity of the precentral gyrus (41); second, the border of Brodmann area 4 extends rostrally to the vertex of the precentral gyrus (particularly dorsomedially) (42); and third, primary motor cortex can be robustly identified with ECS on the exposed precentral gyrus (43). In this context, our result cements the notion that motor imagery involves subliminal activation of the motor system (13) and also
confirms that primary motor cortex is activated during motor imagery, validating claims by earlier noninvasive studies (16–19).

It has been known that the firing rate of single motor cortical neurons could be operantly conditioned with the help of explicit feedback (6, 44), and recent ECoG studies have demonstrated that imagery-associated feedback signals could be conditioned during a simple target task (8–11). Our study directly establishes that activity of widespread neuronal populations in motor cortex are augmented with simple imagery-based feedback (Figs. 3–5 and Fig. S6). The simplicity of this feedback task allows direct quantitative comparison with motor-associated representations. The dramatic augmentation, particularly in primary motor cortex, is significant because it demonstrates a dynamic restructur-

ing of neuronal dynamics across whole populations in motor cortex, on very short time scales (<10 min).

Methods

Subjects. Simple motor and imagery tasks were studied in eight patients (two females, ages 12–48, Table S1) who underwent craniotomy and placement of peri-rolandic intracranial electrode arrays for 5–7 days to localize seizure foci before surgery for medically refractory epilepsy. Array locations and duration of implantation were determined solely by clinical criteria. Experiments were performed for 3–5 days at Children's Hospital (subject 6) and Harbormont Hospital (all others) at the University of Washington. Subjects gave informed consent for participation, in a manner approved by the University of Washington Institutional Review Board.

Stimulation. In five of the eight patients (patients 1–4 and 7), electrocortical stimulation mapping (45, 46) of motor cortex was performed for clinical purposes. Each of these patients underwent stimulation mapping to identify motor and speech cortices as part of his/her clinical care. In this mapping, 5- to 10-mA square-wave current pulses (1 ms in length) were passed through paired electrodes for up to 3 s (or less if a response was evoked) to interrupt function, induce sensation, and/or evoke motor responses (Fig. S5). We denoted an electrode as a primary motor site if ECS produced simple movement.

Recordings. The platinum electrode arrays (Ad-Tech) were configured as 4 × 8 or 8 × 8-electrode grids. The electrode pads had 4-mm diameter (2.3 mm exposed) and 1-cm interelectrode distance, and were embedded in Silastic (Fig. S9). ECoG signals were split into two identical sets. One set was fed into the clinical EEG system (XLTEK), and the other set was recorded with Syn-

amp32 (Neuroscan) biocompatible amplifiers (Harbormont Hospital; at 1,000 Hz) or g.USBamp (g.tec) amplifiers (Children's Hospital; at 1,200 Hz). ECoG signals were acquired from the experimental amplifiers using the general-purpose BCI2000 software (47), which was also used for online signal processing and stimulus presentation. The signals at Harbormont Hospital were bandpass-

filtered from 0.3 to 200 Hz. Surface EMG electrodes were placed for clinical purposes in a subset of patients (see Fig. S4 for illustration). Finger position was controlled online and movement-related sensory signals were acquired using a dataglove (5DT) in a subset of patients (Fig. S5).

Tasks. A series of three experiments were performed: interval-timed active motor movement, interval-timed motor imagery, and a cursor-to-target movement task to provide feedback on motor imagery. First, subjects performed simple, repetitive movements (22) of the hand (synchronous flexion and extension of all fingers; i.e., clenching and releasing a fist at a self-paced rate of ≈1 Hz), the tongue (opening of mouth with protrusion and retraction of the tongue; i.e., sticking the tongue in and out, also at 1 Hz), shoulder (shrug at 1 Hz), or simple vocalization (saying the word “move”). These movements were performed in an interval-based fashion, alternating between 3-s movement blocks and rest. The side of actual or imagined somatic movement was always contralateral to the side of cortical grid placement. There were 30 text cues (“Hand,” “Tongue,” “Shoulder,” “Move”) for each movement modality, delivered visually as text (2.5 cm high) at a distance of 75–100 cm from the subject. Cues of different type were interleaved randomly so no particular movement type could be anticipated.

Following the overt movement experiment, each subject performed an interval-timed active movement task involving identical movement rather than executing the movement. The imagery was kinesthetic rather than visual (19, 29) (“imagine yourself performing the actions like you just did”; i.e., “don’t imagine what it looked like, but imagine making the motions”).

One half of the subjects (1, 2, 6, 8) also performed an imagery-based brain–computer interface task (the others did not because clinical conditions limited the time available for participation in our experiments). The activation (Aactiv) was calculated in 2-Hz bins at each electrode, from the basic motor or imagery task. Experimenters then visually selected a feedback feature, a particular electrode and frequency combination, to be used for online cursor control. That frequency-range power from a partic-

ular electrode was coupled to the speed of a computer cursor using a simple linear relation (Fig. 5). Targets were presented in random order, in one of two locations on the perimeter of the screen (e.g., up/down or left/right). The subject was instructed to imagine (again, via kinesthetic imagery) a particular movement to move a cursor toward one target (the “active” direction), and to rest (or ‘idle’) to move the cursor to the other target (the “passive” direction). The active and passive targets were presented in a block-randomized fashion, and in approximately equal numbers. A trial was terminated when the cursor hit any target or when movement time exceeded a fixed duration (7.2 s). A trial was followed by a 1-s “reward” period during which the target turned yellow if the correct target was hit. The reward period was followed by a 1-s rest period during which the screen was blank before the next trial began. A trial was considered a “miss” if the wrong target was hit or the timeout length was reached. Targets were presented continuously for ~2 min (i.e., a run), and the experiment typically consisted of three to five such runs.

Signal Analysis. A high-frequency band (HFB) (76–100 Hz) and a low-

frequency band (LFB) (8–32 Hz) were chosen for analysis, as described in previous studies (22). This LFB range was chosen to capture the full width of motor-associated alpha rhythm spectral change, as well as variability in the peak of the motor rhythm, across brain areas and subjects (23). Samples of the filtered signal were processed between movement and rest in the movement task, imagery and rest in the imagi-

ty task, and upper and lower targets in the feedback task (illustrated in Fig. S10). During mixed hand–tongue tasks, hand (tongue) movement cues were compared only to rest cues following hand (tongue) movement because the beta rebound following movement is task-specific (30). The cursor velocity was determined online each 40 ms based on the power in a particular frequency range electrode combination calculated over the prior 280 ms.

ECoG “activation,” Aactiv, was quantified as

\[ A_{activ} = \frac{\left( \bar{m} - r \right)^3 \cdot N_{m} - N_{r}}{|m - r|^{2} \cdot \sigma_{m}^{2}} \]

where \( m \) denotes distribution of samples of power (with mean \( m \)) for movement/imagery/active-feedback cues and \( r \) denotes distribution of power for resting/idle/passive-feedback target cues. The combined (joint) distribution is denoted \( m_{r} \cdot N_{m} \) and \( N_{r} \) are the number of samples of type \( m \), \( r \), respectively, and \( N_{m} \cdot N_{r} = N_{m} \cdot N_{r} \). The signed squared cross-correlation coefficient \( r^{2} \) comparing two distributions of the normalized power in the ECoG HFB or LFB. In other words, \( A_{activ} \) is the per-

centage of the variance in the data that can be explained by the fact that two subdistributions \( m \) and \( r \) have different means, \( m \) and \( r \), with an attached sign to indicate whether there is an increase or decrease associated with the action (vs. rest). The imagery/movement ratio is \( m - r \) for imagery, divided by \( m - r \) for movement in the same electrode.

To quantify power shift during feedback, the ratios of the activation metrics, \( A_{activ} \), were used instead, because the trial numbers and durations were not balanced, and a statistical measure was considered to be more appropriate. However, the values of the resulting ratios involving feedback are considered meaningful only in the sense that one case is larger than another.

The overlap metric and overlap significance between patterns of brain activation were determined using a reshuffling technique (illustrated in Fig. S3). The dot product of the activations between two cases was determined (e.g., \( \sum A_{activ}(f_{j} \cdot \text{Imagery}) \cdot A_{activ}(f_{j} \cdot \text{Movement}) \), where \( f_{j} \) denotes the \( j \)th electrode).

Then, a distribution of surrogate overlap dot-products was generated by randomly reshuffling the channel labels for the imagery task, and then recomputing. The associated overlap metric was the value of the appropriate dot product in z-score units from the surrogate distribution. The associated \( P \) value was the probability that a surrogate dot product value was larger than the actual dot product.

Electrode Localization and Template Brain Mapping. Electrode locations were estimated based upon their relation to skull table and landmarks from the sagittal (lateral) and coronal anterior–posterior skull x-rays using the LOC package (48),
We thank the patients and staff at Harborview Hospital, Seattle, for their time and dedication, and Dora Hermes for helpful discussions. This work was supported by National Science Foundation Grants 0622522, 0642848, and 0930908; the Packard Foundation; National Institutes of Health Grant R01-61-3925; the National Aeronautics and Space Administration Graduate Student Researchers Program; and the National Institute of General Medical Sciences Medical Scientist Training Program.


ACKNOWLEDGMENTS. We thank the patients and staff at Harborview Hospital, Seattle, for their time and dedication, and Dora Hermes for helpful discussions. This work was supported by National Science Foundation Grants 0622522, 0642848, and 0930908; the Packard Foundation; National Institutes of Health Grant R01-61-3925; the National Aeronautics and Space Administration Graduate Student Researchers Program; and the National Institute of General Medical Sciences Medical Scientist Training Program.
Supporting Information
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Fig. S1. Spectral changes in cortical surface potentials during paired hand and tongue movement and imagery for subjects 2–5 (denoted S# in middle of brain plots). This is the same as Fig. 1 C and D, for additional subjects. In each map, the activation (HFB, top two rows in each quadrant; LFB, bottom two rows, activation calculation described in Fig. S10) is scaled to the maximum absolute activation (value indicated in the upper left corner of each map) and plotted to a template cortex. The overlap between maps is quantified according to the bootstrapped overlapping method described in Fig. S3. This overlap is shown on the adjacent bars, where yellow denotes hand–tongue overlap for movement, light blue denotes hand movement–imagery overlap, and pink denotes tongue movement–imagery overlap (corresponding to arrows). The circle-slash indicates a $P > 0.01$ significance associated with the overlap (i.e., a >1% chance that the overlap seen is due to chance).

Fig. S2. Actual and imagined tongue movement activations for subjects 6–8 (HFB, top two rows; LFB, bottom two rows), plotted (as in Fig. 1 and Fig. S1, but no hand movement/imagery task was performed). The HFB overlap metric (significance) for subjects 6/7/8 was 2.20 ($P = 0.03$)/1.96 ($P = 0.04$)/3.73 ($P = 2 \times 10^{-4}$). The LFB overlap metric (significance) for subjects 6/7/8 was 3.59 ($P = 8 \times 10^{-5}$)/1.16 ($P = 0.10$)/2.17 ($P = 0.02$).
Fig. S3. Method for quantifying overlap in spatial distribution. (A) We would like to quantify the overlap between two patterns of activation, one for actual movement, and one for imagery. Subject 2 is shown for illustration, during hand movement and imagery. (B) To do this, we assess the activity at each electrode (numbering on brain plot indicates electrode number). (C) The activity at each electrode is quantified independently using the approach illustrated in Fig. S10. (D) The overlap can be quantified by taking the dot product across electrodes (in this way, the activities are paired together at each electrode). (E) Given a similar distribution of activation values, we can probe all of the possible conformations that they might have taken, by resuffling the electrode positions for one of the cases (movement), and recalculating the dot product. This is done $10^6$ times, and a distribution of surrogate dot products is generated. (F) We can use this surrogate distribution to estimate a $P$ value for the significance of the overlap (in this example, $P = 2 \times 10^{-6}$). The distribution of surrogate values tends to be approximately Gaussian (in this case, kurtosis = 3.15). Because of this, the value of our actual overlap, in $z$-score units (of the surrogate distribution; here, 4.44), can be used as a quantitative measure of overlap.

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Fig. S4. Verifying imagery. Where possible, subjects wore a dataglove (5DT, Irvine, CA) or were monitored using clinical EMG sensors to verify that movement was absent during imagery. Subject 1 wore the dataglove, which recorded index finger position during the interval-based finger movement task. During the motor movement task, the squared correlation of index finger position with the cue (movement vs. rest) was significant ($r^2 = 0.84$, $P < 0.001$). There was no significant correlation of finger position with cue during the hand motor imagery task (second panel). Subject 8 had a sublingual differential EMG pair in place during a tongue motor task, a tongue motor imagery task, and an imagery-based feedback task. The EMG signal was band-stop-filtered from 57 to 63 Hz, 117 to 123 Hz, and 177 to 183 Hz to reject line noise (third-order Butterworth digital filter, forward and backward to eliminate phase distortion). It was then high-pass-filtered at 20 Hz noise (third-order Butterworth digital filter, forward and backward) and squared to obtain the “filtered EMG power.” The power in the EMG was significantly correlated with the tongue motor movement task (cross-correlation $r^2 = 0.93$, $P < 0.001$) but was not correlated with either the tongue imagery or imagery-based feedback task.
Fig. S5. Functional localization of primary motor areas by electrocortical stimulation (ECS), in five subjects (denoted S#). For clinical reasons, it was performed for these 5 subjects, and not the other 3. Language and nonrelevant functionally identified sites are not shown. For subjects 1, 2, 4, and 7, imagery of a type of movement produced significant HFB change (P > 0.05) at a site where ECS induced movement of that type, as shown with circled black electrodes (see Fig. 1 and Fig. S1). In the case of subject 3, imagery produced significant HFB change only at a frontal site, and not at the site that induced motor movement (motor movement also produced HFB change at that frontal site, but not at the ECS site). Electrocortical stimulation (ECS) to create transient lesions or induce overt movements, extra- or intraoperatively, is the established method to clinically localize function in the brain (1–3). This process of stimulation is critical to minimizing risk in neurological surgeries that involve resection of seizure focus, tumor, or vascular malformation (4–6), and predicts functional outcome (6, 7).

Effective mapping requires that all electrodes be stimulated with varying amounts of amperage and requires ongoing and consistent patient participation. Stimulation often leads after discharges that can induce seizures or provide misleading functional responses from distally stimulated cortex. The stimulation mapping process in the context of subdural electrode arrays is often incomplete either because it is prematurely aborted by an induced seizure, or stopped once an acceptable surgical margin is obtained. The entire array is rarely surveyed. For these reasons, many of the activation maps in this study are devoid of stimulation locations or contain incomplete stimulation information. In this sense, there are only positive stimulation results, without negative control. The meaningful question in the context of this study is whether motor imagery produces neuronal population activity, revealed by changes in the HFB, in primary motor areas (i.e., Brodmann areas 4 and 6) that are identified by ECS. As shown in this study by subjects 1, 2, 4, and 7, imagery does result in cortical activity in primary motor areas.

Fig. S6. Augmentation of cortical activity during learning. (A) Site of feedback, shown with a black dot, and electrode positions for subjects 6, 2, 8, and 1. The box above each template brain indicates the subject, the paired motor/imagery/feedback modality, and the frequency range used for feedback. The feedback accuracy is the percentage of time the subject was able to hit the correct target using this feedback feature. “Speech” for subject 1 was repetition of the word “move.” (B) HFB (76–100 Hz) ECoG-based brain activation maps for actual movement, imagined movement, and feedback-based BCI cursor control. Activation in each map is scaled to the maximum activation (absolute-value, noted by the number above each template brain). Subject 1 did not perform a speech (word repetition) imagery task. (C) As in B, for the LFB (8–32 Hz). (D) The relative activation (also shown in Fig. 4) during each of the three conditions is shown for the HFB (Upper) and LFB (Lower) on the left side of each set of axes, for each subject. Relative activation was computed by dividing the imagery- and feedback-related activation by the activation for actual movement, and normalizing the magnitude of activation for actual movement to 1 for all plots. The bar plots show that in most cases, the magnitude of feedback-related activation after learning exceeds the magnitude of activation for actual movement. The overlap between each of the conditions is shown with the purple/pink/magenta bars, for the HFB (Upper) and LFB (Lower) on the right side of each set of axes, for each subject.

Fig. S7. Spectral changes in ECoG following shoulder imagery feedback for subject S6 (feedback from 13 to 21 Hz) on a medial electrode (red circle; same as in Fig. S6 A–C).
Fig. S8. Changes in cortical activity during brain–computer interface control for subject 8 during tongue motor imagery (details as in Fig. 5). Three consecutive runs with feedback are shown. The subject reported being disinterested in the last run.

Fig. S9. Coregistration of electrodes, comparing anatomical photographs, subject 8. (A) Electrode positions interpolated from x-ray (on the AFNI template) estimated with the LOC package. (B) Electrode positions shown from an intraoperative picture. (C) The aperture in B projected onto C, with the feedback electrode shown with a black circle. (D) The location of the electrode on the exposed brain (intraoperative picture). SF, Sylvian fissure; CS, central sulcus.
Fig. S10. Method for calculating and displaying activation. (A) An electrode from primary motor cortex from subject 1 shown for illustration. (B) A subject performed an interval-based motor movement or kinesthetic imagery [Neuper C, et al. (2005) Brain Res Cogn Brain Res 25:668–677] task. The raw signal (re-referenced with respect to the common average) from each electrode is divided into 3-s epochs based on the stimulus to move (or to imagine moving) the hand or tongue, and resting. A subset of each epoch, the time corresponding to 0.5–2 s into that epoch (shown in the horizontal lines below the raw data), is selected for analysis, to account for behavioral lag. (C) The power spectral density of each epoch was calculated every 1 Hz using a fast Fourier transform with 0.25-s Hann-windowed segments, stepping with an overlap of 0.1 s (Welch’s averaged periodogram method). Each power spectrum is then normalized by dividing the power at each frequency bin by the average power at that frequency bin, taken across all epochs. (D) Total power in a low-frequency band (LFB) (8–32 Hz) and a high-frequency band (HFB) (76–100 Hz) is calculated for each epoch of each type. The plot shows samples of HFB power during hand movement and rest. (E) The activation, $A_{mr}$, is calculated for each movement or imagery modality. In the case of a mixed modality experimental run, the distribution of samples of power for each modality are compared only with the samples of power for rest epochs that followed epochs of that kind (i.e., hand movement epochs are compared with rest epochs that followed hand movement, and not those that followed tongue movement). Activation was quantified as the signed square of the cross-correlation to reflect an increase or a decrease in the band-limited PSD for movement (m) with respect to rest (r) (or, in the case of a target task, whether one target represents an increase or decrease in power with respect to the other target). This metric indicates how much of the variation in the joint data set $\sigma^2_{mr}$ can be accounted for by the fact that subdistributions $m$ and $r$ might have different means, $\mu_m$ and $\mu_r$, and that $N_m$ and $N_r$ are the number of samples of type $m$ and $r$, respectively, and $N_{mr} = N_m + N_r$. (F) The activation obtained for each electrode is used to scale a spherical Gaussian kernel ($\sigma = 5$ mm) centered at the Talairach location obtained for that electrode using the LOC package [Miller KJ, et al. (2007) J Neurosci Methods 162:303–308], and all kernels are linearly superimposed to obtain a cortical map. The activation is scaled to the maximum absolute activation (indicated to the upper left of the map), with the color scale shown (dark red, maximum increase in activation; dark blue, maximum decrease in activation; gray, no activation). These maps are generated for each subject, each band (HFB or LFB), and each movement or imagery modality.
Table S1. Patient descriptions and characteristics

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Correction

NEUROSCIENCE

The authors note that, in the Acknowledgments, “National Institutes of Health Grant R01-61-3925” should read “National Institutes of Health Grants R01-61-3925, EB006356, and EB000856.” Also, on page 4434, right column the equation appeared incorrectly.

\[ A_{mr} = \frac{(m-r)^3 N_m - N_r}{|m-r| \sigma_{mr}^2 N_m^2 N_r} \]

should instead appear as

\[ A_{mr} = \frac{(m-r)^3 N_m N_r}{|m-r| \sigma_{mr}^2 N_m^2 N_r} \]

This error does not affect the conclusions of the article.

www.pnas.org/cgi/doi/10.1073/pnas.1002462107