

Enhancing long-term memory with stimulation tunes visual attention in one trial

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Scientists have long proposed that memory representations control the mechanisms of attention that focus processing on the task-relevant objects in our visual field. Modern theories specifically propose that we rely on working memory to store the object representations that provide top-down control over attentional selection. Here, we show that the tuning of perceptual attention can be sharply accelerated after 20 min of noninvasive brain stimulation over medial-frontal cortex. Contrary to prevailing theories of attention, these improvements did not appear to be caused by changes in the nature of the working memory representations of the search targets. Instead, improvements in attentional tuning were accompanied by changes in an electrophysiological signal hypothesized to index long-term memory. We found that this pattern of effects was reliably observed when we stimulated medial-frontal cortex, but when we stimulated posterior parietal cortex, we found that stimulation directly affected the perceptual processing of the search array elements, not the memory representations providing top-down control. Our findings appear to challenge dominant theories of attention by demonstrating that changes in the storage of target representations in long-term memory may underlie rapid changes in the efficiency with which humans can find targets in arrays of objects.

medial-frontal cortex | visual attention | long-term memory | executive control | transcranial direct-current stimulation

The cognitive and neural mechanisms that tune visual attention to select certain targets are not completely understood despite decades of intensive study (1, 2). Attention can clearly be tuned to certain object features (similar to tuning a radio to a specific station, also known as an attentional set), but how this tuning occurs as we search for certain objects in our environment is still a matter of debate. The prevailing theoretical view is that working memory representations of target objects provide top-down control of attention as we perform visual search for these objects embedded in arrays of distractors (3–7). However, an alternative view is that long-term memory representations play a critical role in the top-down control of attention, enabling us to guide attention based on the more enduring representations of this memory store (8–16). To distinguish between these competing theoretical perspectives, we used transcranial direct-current stimulation (tDCS) to manipulate activity in the brain causally (17), and combined this causal manipulation of neural activity with electrophysiological measurements that are hypothesized to index the working memory and long-term memory representations that guide visual attention to task-relevant target objects.

To determine the nature of the working memory and long-term memory representations that control visual attention during search, we simultaneously measured two separate human event-related potentials (ERPs) (8, 18, 19). The contralateral delay activity (or CDA) of subjects' ERPs provides a measure of the maintenance of target object representations in visual working memory (20, 21). The CDA is a large negative waveform that is maximal over posterior cortex, contralateral to the position of a remembered item. This large-amplitude lateralized negativity is observed even when nonspatial features are being remembered,

and persists as information is held in working memory to perform a task. A separate component, termed the anterior P1, or P170, is hypothesized to measure the build-up of long-term memory representations. The anterior P1 is a positive waveform that is maximal over frontal cortex and becomes increasingly negative as exposures to a stimulus accumulate traces in long-term memory (8, 19, 22). This component is thought to reflect the accumulation of information that supports successful recognition of a stimulus on the basis of familiarity (23). For example, the anterior P1 amplitude can be used to predict subsequent recognition memory for a stimulus observed hundreds of stimuli in the past (i.e., across minutes to hours of time) (23) (additional information on the critical features of these ERP components is provided in *SI Materials and Methods*). We used simultaneous measurements of the CDA and anterior P1 to determine the role that working memory and long-term memory representations play in the tuning of attention following brain stimulation.

Our tDCS targeted the medial-frontal region in our first experiments (Fig. 1A) because anodal stimulation of this area results in rapid improvement of simple visual discriminations relative to baseline sham conditions (24). If it is possible to induce rapid improvements in the selection of targets among distractors as humans perform search, then the competing theories of visual attention would account for the accelerated tuning of attention in different ways. The theories that propose working memory representations provide top-down control of visual attention predict that the stimulation-induced improvement in visual search will be due to changes in the nature of the visual working memory representations indexed by the CDA component (Fig. 1B and C). Specifically, the CDA elicited by the target cue presented on each trial should increase in amplitude, relative to the sham condition, to explain the improvement of attentional selection during search. This type of modulation is expected if

Significance

Theories of attention propose that we rely on working memory to control attention by maintaining target presentations in this active store as our visual systems are used to search for certain objects. Here, we show that the tuning of perceptual attention can be sharply accelerated by noninvasive brain stimulation. Our electrophysiological measurements showed that these improvements in attentional tuning were preceded by changes in event-related potentials thought to index long-term memory, but not those potentials that index working memory. Our findings support the hypothesis that changes in the storage of target representations in long-term memory may underlie rapid changes in how target objects are selected by visual attention.

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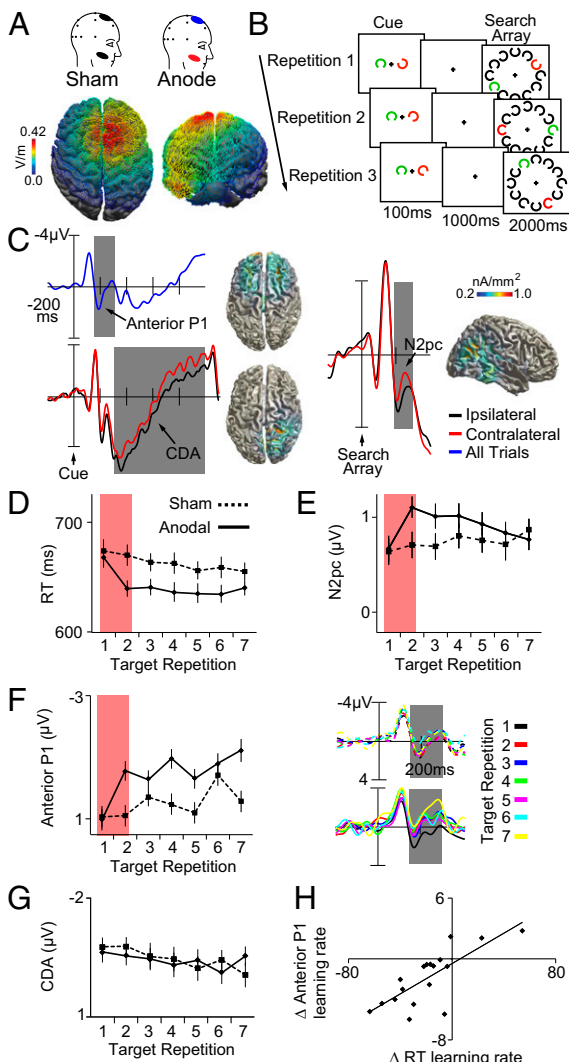


Fig. 1. tDCS model, task, and results of experiment 1. (A) Modeled distribution of current during frontocentral midline anodal tDCS on top and front views of a 3D reconstruction of the cortical surface. (B) Task-relevant cue (green Landolt C in this example) signaled the shape of the target in the upcoming search array. Subjects searched for the same target across a run of three to seven trials. Central fixation was maintained for the trial duration. (C) Representative anterior P1, CDA, and N2pc from repetition 1 in the sham condition show each component's distinctive temporal and spatial profile, with analysis windows shaded in gray. Mean RTs (D), N2pc amplitudes (E), anterior P1 amplitudes (F), and CDA amplitudes (G) are shown across target repetitions for sham (dashed line) and anodal (solid line) conditions. Error bars are ± 1 SEM. Red shading highlights dynamics across trials 1 and 2. Grand average ERP waveforms from the frontal midline electrode (Fz) synchronized to cue onset are shown across target repetitions for sham (dashed line) and anodal (solid line) conditions. The measurement window of the anterior P1 is shaded in gray. (H) Relationship between logarithmic rate parameter enhancements for mean anterior P1 amplitude and RT after anodal stimulation relative to sham.

working memory-driven theories of attention are correct based on previous evidence that the CDA is larger on trials of a short-term memory task when performed correctly compared with incorrect trials (20). In contrast, theories that propose long-term memory representations rapidly assume control of attention during visual search predict that the stimulation-induced improvement will be due to changes in the long-term memory representations indexed by the anterior P1 elicited by the target cue presented on each trial. Specifically, we should see the

anterior P1 exhibit a more negative potential as search improves following stimulation.

Each subject completed anodal and sham tDCS sessions on different days, with order counterbalanced across subjects ($n = 18$). Immediately after 20 min of tDCS over medial-frontal (experiments 1 and 2) or right parietal (experiment 3) regions of the head (see the current flow model for experiment 1 in Fig. 1A, and additional information about stimulation locations in *SI Materials and Methods*), we recorded subjects' ERPs while they completed a visual search task. In this search task, the target was cued at the beginning of each trial (Fig. 1B and C). The task-relevant cue signaled the identity of the target that could appear in the search array presented a second later. In experiments 1 and 3, the targets and distractors were Landolt-C stimuli, and in experiment 2, they were pictures of real-world objects. A task-irrelevant item was presented with each cue to balance the hemispheric visual input so that the lateralized ERPs that elicit the CDA could be unambiguously interpreted (25). The key manipulation in this task was that the target remained the same for three to seven consecutive trials (length of run randomized) before it was changed to a different object. These target repetitions allowed us to observe attentional tuning becoming more precise across trials.

We found that anodal medial-frontal tDCS in experiment 1 accelerated the rate of attentional tuning across trials, as evidenced by the speed of behavior and attention-indexing ERPs elicited by the search arrays (Fig. 1D and E). First, in the baseline sham condition, we observed that subjects became faster at searching for the target across the same-target runs of trials, as shown by reaction time (RT) speeding ($F_{2,34} = 6.031$, $P = 0.007$) (additional analyses of the sham condition and analyses to verify the absence of effects on accuracy are provided in Fig. S1A and *SI Materials and Methods*). However, following anodal stimulation, subjects' RTs dramatically increased in speed, such that search RTs reached floor levels within a single trial. This striking causal aftereffect of anodal tDCS was evidenced by a stimulation condition \times target repetition interaction on RTs ($F_{2,34} = 3.735$, $P = 0.042$), with this RT effect being significant between the first two trials of search for a particular Landolt C ($F_{1,17} = 6.204$, $P = 0.023$) but with no significant change thereafter ($P > 0.310$). Additionally, by fitting these behavioral RT data with a logarithmic function to model the rate of improvement (9), we found that anodal tDCS significantly increased the rate parameters of RT speeding ($F_{1,17} = 5.097$, $P = 0.037$).

Consistent with the interpretation that tDCS changed how attention selected the targets in the search arrays, we found that the N2-posterior-contralateral (N2pc) component, an index of the deployment of covert attention to the possible target in a search array (26), showed a pattern that mirrored the single-trial RT effects ($F_{1,17} = 4.792$, $P = 0.043$) (Fig. 1E; N2pc waveforms are provided in Fig. S1A). However, other ERP components indexing lower level perceptual processing or late-stage response selection during search were unchanged by the tDCS (Fig. S1C and D and Table S1). Our findings demonstrate that the brain stimulation only changed the deployment of visual attention to targets in the search arrays and did not change the operation of any other cognitive mechanism we could measure during the visual search task. Thus, by delivering electrical current over the medial-frontal area, we were able to accelerate the speed with which subjects tuned their attention to select the task-relevant objects causally.

To determine whether the tDCS-induced attentional improvements were caused by changes in working memory or long-term memory mechanisms of top-down control, we examined the putative neurophysiological signatures of visual working memory (i.e., the CDA) and long-term memory (i.e., the anterior P1) elicited by the target cues. Given the rapid tuning of attention following tDCS relative to sham, we might expect the flexible working memory system to underlie this effect. Contrary to this intuition, we found that the rapid, one-trial improvement in

attentional tuning following medial-frontal tDCS was mirrored by changes in the putative neural index of long-term memory but left the putative neural index of working memory unchanged (Fig. 1 *F* and *G*). Fig. 1*F* shows that the accelerated effects of attentional tuning caused by anodal stimulation were preceded by a rapid increase in negativity of the anterior P1 across same-target trials, mirroring the rapid, single-trial improvement in RT and the N2pc as the search array was analyzed. This effect was confirmed statistically by a significant stimulation condition \times target repetition interaction on the anterior P1 amplitude ($F_{2,34} = 3.797$, $P = 0.049$), and most dramatically between the first two trials of search ($F_{1,17} = 5.816$, $P = 0.027$), with no significant pairwise changes in anterior P1 amplitude thereafter ($P > 0.707$). Logarithmic model fits showed that the rate parameters of the anterior P1 significantly increased after anodal tDCS relative to the more gradual attentional tuning observed in the sham condition ($F_{1,17} = 5.502$, $P = 0.031$; anterior P1 analyses from the sham condition are described in *SI Materials and Methods*). Despite these causal changes in anterior P1 activity, neither the amplitude of the CDA ($F_{2,34} = 0.669$, $P = 0.437$) nor its rate parameters ($F_{1,17} = 1.183$, $P = 0.292$) significantly differed between stimulation conditions, showing the selectivity of medial-frontal tDCS on the putative neural metric of long-term memory (CDA waveforms are provided in Fig. S1*B*). We note that the absence of a stimulation-induced CDA increase is not due to ceiling effects. The single target cue gave us ample room to measure such a boost of the CDA, given that without brain stimulation, this memory load is far from eliciting ceiling amplitude levels for this component (20).

If the better long-term memory representations indexed by the anterior P1 were the source of the improved search performance, then the size of the stimulation-induced boost of the anterior P1 elicited by the cue should be predictive of the search performance that followed a second later. Consistent with the prediction, we found that an individual subject's anterior P1 amplitude change across the same-target runs following medial-frontal stimulation was highly predictive of the accelerated rates at which the subjects searched through the visual search array that followed the target cues ruled out the working memory explanation of the rapid changes in attentional tuning we observed, and were consistent with the hypothesis that changes in the nature of the long-term memory representations that control attention were the source of this dramatic improvement.

In experiment 2, we replicated the pattern of findings from experiment 1 using a search task in which the targets and distractors were pictures of real-world objects (Fig. 2 and Fig. S2). These results demonstrate the robustness and reliability of the pattern of effects shown in experiment 1. Specifically, brain stimulation resulted in attention being rapidly retuned to the new targets after one trial, as evidenced by RTs hitting the floor by the second trial in a run. Again, this change in RT was mirrored by stimulation changing the anterior P1, and not the CDA, consistent with accounts that posit an important role for long-term memory in the guidance of attention.

Next, we sought to provide converging evidence for our conclusion that the stimulation was changing subjects' behavior by changing the nature of subjects' long-term memory, consistent with previous functional interpretations of the anterior P1. So far, we have drawn conclusions using our analyses across the fairly short runs of same-target trials. However, we next looked at the learning that took place across the entire experimental session, lasting almost 3 h. If our interpretation of the anterior P1 underlying accelerated attentional tuning is correct, then we should see that the anterior P1 is sensitive to the accumulative effects of learning across the entire experimental session and that these long-term effects change following stimulation. To assess the cumulative effects of learning across these long experimental sessions, we examined how behavior, the anterior P1,

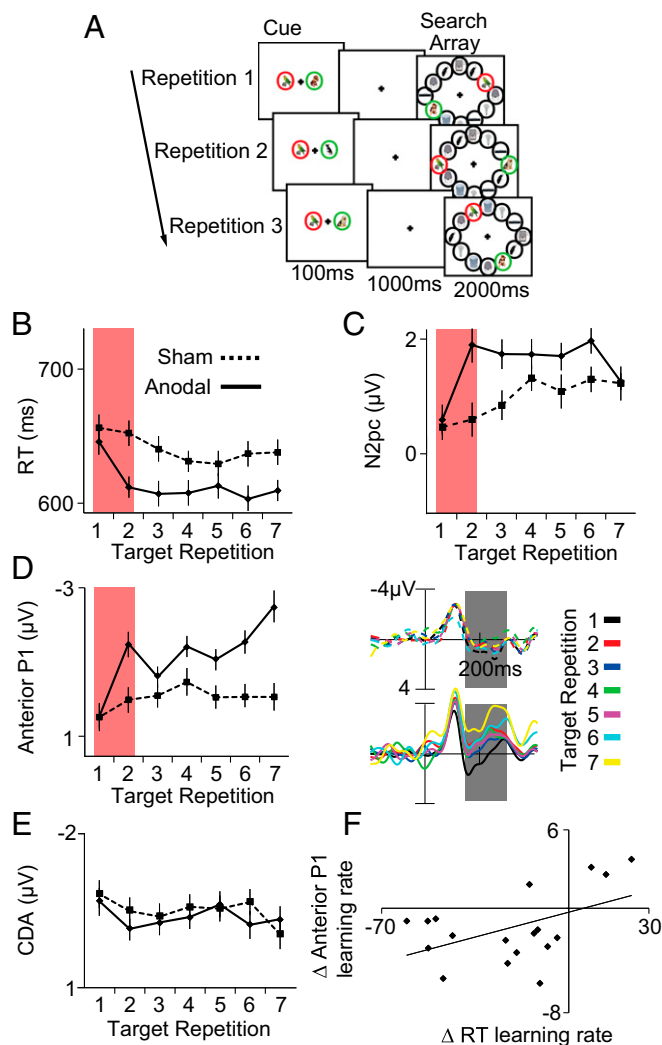


Fig. 2. Task and results of experiment 2. (A) Task in experiment 2 was identical to that of experiment 1 with the exception that Landolt-C stimuli were replaced with real-world objects. Mean RTs (B), N2pc amplitudes (C), anterior P1 amplitudes (D), and CDA amplitudes (E) are shown across target repetitions for sham (dashed line) and anodal (solid line) conditions. Error bars are ± 1 SEM. Red shading highlights dynamics across trials 1 and 2. Grand average ERP waveforms from the frontal midline electrode (Fz) synchronized to cue onset are shown across target repetitions for sham (dashed line) and anodal (solid line) conditions. The measurement window of the anterior P1 is shaded in gray. (F) Relationship between logarithmic rate parameter enhancements for mean anterior P1 amplitude and RT after anodal stimulation relative to sham.

and the CDA changed across the beginning, middle, and end of experiments 1 and 2 (Fig. 3 and *SI Materials and Methods*); that is, we averaged the same-target runs together in the first third, second third, and final third of sessions across all of our subjects. Fig. 3 shows the learning we observed across these long sessions. The RTs were slowest at the beginning of the experiment, when faced with a new target, but as subjects accumulated experience with the set of eight possible targets, we saw the RTs at the beginning of the same-target runs become progressively faster. This accumulation of experience across the entire session that sped RT was mirrored by systematic changes in the amplitude of the anterior P1. The anterior P1 became progressively more negative across the experiment, as we would expect if the magnitude of the negativity were indexing the quality (i.e., strength or number) of the long-term memories for these targets that

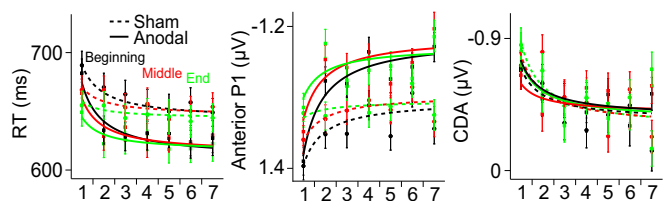


Fig. 3. Within-session dynamics of experiments 1 and 2. Mean RT, anterior P1 amplitude, and CDA amplitude as a function of target repetitions binned according to the first third (black), middle third (red), and last third (green) of runs, collapsed across experiments 1 and 2. Logarithmic model fits are shown for sham (dashed line) and anodal (solid line) tDCS conditions. Error bars are ± 1 SEM.

accumulated across the entire experiment. In contrast, the CDA showed no change across the entire experiment, indicating that the role of working memory in updating the target at the beginning of the same-target runs does not change with protracted learning. For example, it is likely that working memory representations were reactivated to help reduce proactive interference from the target representations built up during the previous run of trials, consistent with influential theoretical proposals (27). Our medial-frontal tDCS boosted these learning effects measured with the anterior P1 and search RTs while leaving the CDA unchanged, consistent with our interpretation of the findings across the shorter same-target runs. Thus, this cumulative learning across the entire experimental session allowed us to observe how the dynamics of the memory representations underlying the focusing of attention evolved over the long term. These results lend further support to the hypothesis that contributions from

long-term memory are driving the causal boost of attentional tuning we observed following brain stimulation.

To determine whether the effects of experiments 1 and 2 were specific to medial-frontal stimulation, in experiment 3, we stimulated the posterior parietal region in a new group of subjects (order of anodal and sham conditions was counterbalanced, $n = 18$) (Fig. 4A). This region of the dorsal visual stream plays a role in memory (28) and generating top-down attentional control signals (29), so that it provides a useful contrast with our medial-frontal stimulation, which appeared to influence attentional selection by changing the long-term memory representations. We specifically targeted the right parietal region because previous studies show that disrupting activity in right parietal cortex can influence attention (30, 31).

We found that unlike medial-frontal stimulation, right parietal tDCS had no effect on the overall tuning of attention or the memory representations controlling search performance. Fig. 4B–E shows the overlap between stimulation conditions for the RTs (no stimulation condition \times target repetition interaction: $F_{2,34} = 0.029$, $P = 0.955$) and the amplitudes of the N2pc ($F_{2,34} = 0.139$, $P = 0.807$), CDA ($F_{2,34} = 0.814$, $P = 0.439$), and anterior P1 ($F_{2,34} = 0.393$, $P = 0.663$) across target repetitions. Because subjects again searched for the same target across the runs of trials in experiment 3, we did observe main effects of target repetition on RTs ($F_{2,34} = 6.190$, $P = 0.015$) and the amplitudes of the N2pc ($F_{2,34} = 4.053$, $P = 0.045$), CDA ($F_{2,34} = 5.292$, $P = 0.024$) and anterior P1 ($F_{2,34} = 6.320$, $P = 0.006$). These effects were due to the steady speeding of RTs, declining CDA amplitude, and increasing amplitudes of the anterior P1 and N2pc across same-target trials. The effects of target repetition indicate that the roles played by working memory and long-term memory in tuning attention across

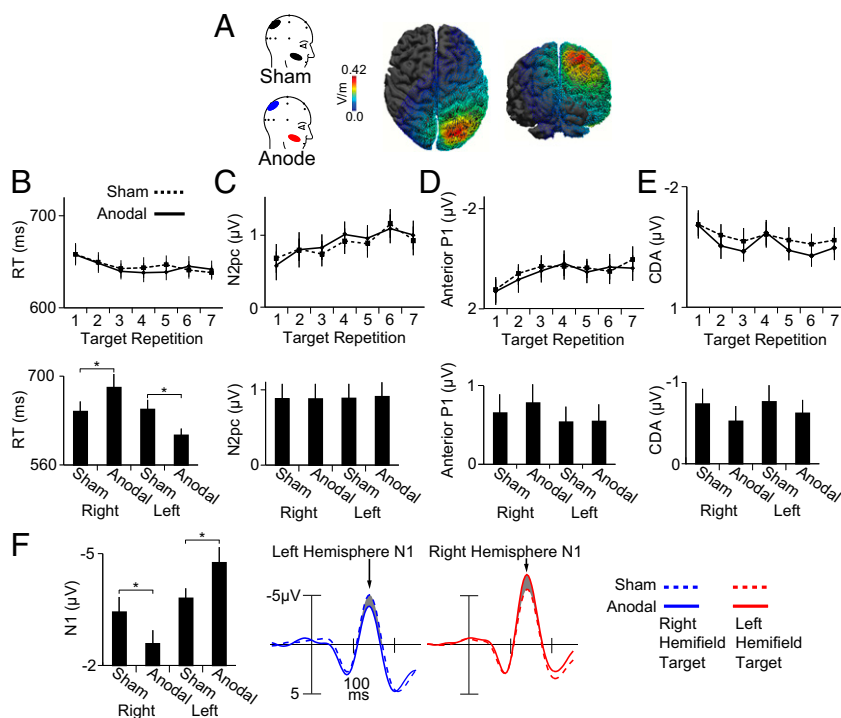


Fig. 4. tDCS model and results of experiment 3. (A) Modeled distribution of current during right parietal anodal tDCS on top and rear views of a 3D reconstruction of the cortical surface. Mean RTs (B), N2pc amplitudes (C), anterior P1 amplitudes (D), and CDA amplitudes (E) are shown across target repetitions for sham (dashed line) and anodal (solid line) conditions. Bar graphs show data collapsed across target repetitions for each stimulation condition based on whether the target color appeared in the left or right visual hemifield. Error bars are ± 1 SEM. (F) Mean N1 amplitudes are illustrated as in B–E. The waveforms are search array-locked grand average potentials at lateral occipital sites (OL/OR) contralateral to right (blue) and left (red) hemifield target colors shown across sham (dashed line) and anodal (solid line) conditions. OL, occipital left; OR, occipital right. $*P < 0.05$.

trials in the baseline sham condition were unchanged following right parietal stimulation (Fig. 4 *B–E* and Figs. *S3D* and *S4*).

Given the lateralized application of tDCS in experiment 3, we examined the data based on whether the target appeared in the left or right visual field. We found that parietal stimulation caused lateralized, bidirectional effects on search performance. Relative to sham, subjects were faster at searching for targets after anodal stimulation, but only on trials in which the target color appeared contralateral (i.e., in the left visual field) to the location of the stimulating electrode on the head (i.e., over the right hemisphere) (Fig. 4*B*). This effect was evidenced by a stimulation condition \times target color laterality interaction on search RTs ($F_{1,17} = 12.098$ $P = 0.003$) and a main effect of stimulation condition on contralateral search RTs ($F_{1,17} = 6.014$ $P = 0.025$). In contrast, RTs were slower when target colors appeared ipsilateral (i.e., in the right visual hemifield) with respect to the location of tDCS ($F_{1,17} = 4.276$ $P = 0.054$) (Fig. 4*B*). These results suggest that parietal stimulation facilitated and impeded overall search behavior depending on the location of the target in the visual field.

We found that the lateralized, bidirectional effects of parietal tDCS on search performance were caused by directly influencing perceptual processing, not changing the memory representations controlling attention. The amplitude of the posterior N1 component, a neural index of perceptual processing (32), was significantly modulated by stimulation condition and in a pattern mirroring that of the behavior (stimulation condition \times target color laterality interaction: $F_{1,17} = 10.494$ $P = 0.005$; stimulation condition main effects: contralateral, $F_{1,17} = 4.755$ $P = 0.044$; stimulation condition main effects: ipsilateral, $F_{1,17} = 4.573$ $P = 0.047$) (Fig. 4*F* and Fig. *S3A*). In contrast, our indices of the memory representations of the targets and of the deployment of attention were not significantly changed by tDCS [i.e., no stimulation condition \times target color laterality interaction: N2pc ($F_{1,17} = 0.041$ $P = 0.843$), CDA ($F_{1,17} = 0.107$ $P = 0.748$), anterior P1 ($F_{1,17} = 0.169$ $P = 0.686$)] (Fig. 4 *C–E* and Fig. *S3 B–D*).

In sum, our parietal stimulation protocol did not change the nature of the memory representations controlling attention but directly influenced the perceptual processing of the objects in the search array. These observations were evidenced by lateralized changes in the early visual ERPs and the behavioral responses to the task-relevant items contralateral vs. ipsilateral to the stimulation. Thus, the effects observed in experiments 1 and 2 are not a ubiquitous pattern observed following stimulation of any cognitive control structure. Instead, when we stimulated the posterior parietal region of the visual stream, we observed changes in early visual responses of the brain and similarly spatially mapped patterns of performance.

Our findings from experiments 1 and 2, that stimulation over medial-frontal areas can rapidly improve attentional selection of targets, may seem surprising because the medial-frontal cortex is not commonly thought to be a crucial node in the network of regions that guide attention (29, 33). This region is most frequently discussed as critical for the higher level monitoring of task performance, response conflict, and prediction error (34, 35). However, a variety of studies across species and methods have found connections between regions of medial-frontal cortex and both attention and memory processes. First, human neuroimaging research shows that the cingulate opercular network, including anterior cingulate and presupplementary cortex, is engaged during the implementation of a task set, visuospatial attention, and episodic memory (36–38). Second, studies using animal models show that attentional selectivity in the visual domain appears to reside in dorsomedial areas of prefrontal cortex (39), such as the anterior cingulate gyrus. Third, both the dorsomedial and right dorsolateral prefrontal cortices respond strongly in memory recognition tasks with specific activity bordering the anterior cingulate at or near Brodmann's areas 6, 8, and 32 (40), including supplementary and presupplementary

motor areas. The right dorsolateral prefrontal cortex, which also appeared to be in the path of our current-flow modeling, has been causally linked to human long-term memory processes (41). Given the set of regions in this path, the specificity of our empirical observations is striking. However, future work is clearly needed to dissect the contributions of the group of medial-frontal and medial-prefrontal regions within the path of the current used here.

Our results present evidence from causal manipulations of the healthy human brain that suggest the rapid reconfiguration of the top-down control of visual attention can be carried out by long-term memory. This conclusion seems counterintuitive, given that the active storage of objects in working memory can strongly control attention (7, 18, 42) and that the dominant theories of attention focus exclusively on the role of working memory in guiding attention (3–6). The present findings do not suggest that working memory representations do not control attention across the short term; indeed, we observed the neural index of storage of the target in working memory that was concurrent with the large changes in the putative index of long-term memory. The critical implication of the present findings is that the rapid improvements in attentional control following brain stimulation were most closely related to our ERP measure of long-term memory and not working memory. These results are surprising to us, given that effects of long-term memory on attentional control are typically observed in tasks in which improvements evolve slowly across protracted training (10, 12–14, 16, 43), or even a lifetime of semantic associations (11). Here, we show that the time course of improvement need not be diagnostic of the type of memory representation involved.

Our results can also be interpreted within theoretical models that take a broader view of top-down control and do not rely on a conceptual dichotomy between working memory and long-term memory processes that guide attention (44). Neuroimaging research has identified multiple control mechanisms that configure downstream processing consistent with behavioral goals. Most relevant here is the network consisting of the anterior insula (also referred to as the frontal operculum) and dorsal anterior cingulate cortex (also referred to as the medial superior frontal cortex). This network is thought to integrate information over protracted time scales, in an iterative manner, similar to the dynamics and functional properties of the anterior P1. Further, the cingulate opercular network carries various critical control signals, including the selection and maintenance of task goals and the making and monitoring of choices (38, 45, 46). It is possible that our medial-frontal stimulation changed the functioning of this control network, causing the improvements we observed in attentional control.

Finally, our findings provide evidence from causal manipulations of the human brain to support the slowly growing view that the nature of top-down attentional control involves the interplay of different types of memory representations (8, 15, 47–49). Moving forward, we believe that such a view moves theories of attention nearly into register with models of learning, automaticity, and skill acquisition (9, 50–52). Ideally, this perspective will serve to unify, rather than further hyperspecialize, theories of information processing in the brain.

Materials and Methods

Subjects. Eighteen subjects participated in each experiment (additional subject information is provided in *SI Materials and Methods*). All had normal color vision and normal or corrected-to-normal visual acuity, and gave informed consent to participate in the study approved by the Vanderbilt University Institutional Review Board.

tDCS. The tDCS was administered using a battery-driven, constant-current stimulator (Mind Alive, Inc.) through a pair of conductive rubber electrodes (active, 19.25 cm²; reference, 52 cm²). The electrodes were placed in

saline-soaked synthetic sponges and held in place by a headband. The reference (or cathodal) electrode was placed on the center of the right cheek (Fig. 1A).

Current was applied at 2.0 mA for 20 min over the medial-frontal region (site FCz, from the International 10–20 System) for experiments 1 and 2, and over the right parietal region (site P2) for experiment 3. A sham tDCS condition was administered using an identical procedure, but stimulation only lasted 30 s, ramping up and down at the beginning, middle, and end of the 20-min period to simulate the periodic tingling sensation often endorsed by subjects on active stimulation days. Debriefing questions confirmed that subjects were blinded, and could not distinguish between sham and anodal stimulation.

Stimuli and Task. Following the active or sham stimulation, subjects performed a cued visual search task while their EEG was recorded so that we could extract their ERPs using our standard methods (18, 19) (additional details are provided in *SI Materials and Methods*). Each trial of the task began with fixation (1,200–1,600 ms). Next, two cue stimuli were presented for 100 ms, followed by a 1,000-ms interval during which we measured the CDA and anterior P1. Then, the search array was presented for 2,000 ms (additional metrics describing these stimuli are provided in *SI Materials and Methods*). The intertrial interval was 1,200–1,600 ms, randomly jittered with a rectangular distribution.

In all experiments, a target matched the shape of the Landolt C (experiments 1 and 3) or picture (experiment 2) of the task-relevant cue. Every search array contained an item that matched the color of the cue object (i.e., the possible target), but on target absent trials, this object had a different shape (half of trials). Subjects responded as quickly and accurately as possible to the search array by pressing one button on a handheld gamepad with their right hand for target present and a different button with their right hand for target absent. The task-relevant color in experiments 1 and 3 (i.e., red or green) and the task-relevant real-world object category in experiment 2 (i.e., dog or bird) were counterbalanced across subjects to rule out physical stimulus confounds.

Target presence and possible target location were randomly selected on each trial. The same target was cued across a run of three, five, or seven trials, randomly varying in length, with the identity of the target randomly selected for each run without repetition in adjacent runs. Each subject completed 720 trials in each condition (sham and anodal).

Additional details about the methods used in this study are provided in *SI Materials and Methods*.

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