Review

The self-regulating brain and neurofeedback: Experimental science and clinical promise

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ABSTRACT

Neurofeedback, one of the primary examples of self-regulation, designates a collection of techniques that train the brain and help to improve its function. Since coming on the scene in the 1960s, electroencephalography-neurofeedback has become a treatment vehicle for a host of mental disorders; however, its clinical effectiveness remains controversial. Modern imaging technologies of the living human brain (e.g., real-time functional magnetic resonance imaging) and increasingly rigorous research protocols that utilize such methodologies begin to shed light on the underlying mechanisms that may facilitate more effective clinical applications. In this paper we focus on recent technological advances in the field of human brain imaging and discuss how these modern methods may influence the field of neurofeedback. Toward this end, we outline the state of the evidence and sketch out future directions to further explore the potential merits of this contentious therapeutic prospect.

1. Introduction

Neurofeedback refers to a self-regulation technique that provides the individual with feedback about specific brain activity in connection with a related behavior. The underlying assumption at the core of this practice posits that through this type of feedback one can entrain, change, and regulate neural activity. This trend appeals to both researchers and practitioners, who wish to understand the neurobiological mechanisms as well as the therapeutic potential this approach may offer. Beyond electroencephalography (EEG), the advent of modern real-time brain imaging technology elucidates the time-course and location of brain activity and seems to open the road to new prospects, including the modulation of seemingly volitionless neural functions (Fig. 1). And yet, imaging-based neurofeedback has hardly transitioned from the cognitive neuroscience lab into the clinical trenches. In this paper we highlight the relative merits and current shortcomings of neurofeedback in the context of contemporary imaging technologies. We discuss how new modalities of brain imaging may provide a future trajectory to consider meaningful research resulting in potential inclusion in the clinical armamentarium.
An evolutionary derivative of biofeedback, in the 1960s neurofeedback emerged to employ neural feedback via EEG (Kamiya, 2011). To this day, specialty clinics and private institutions continue to offer variations on EEG neurofeedback (EEG-nf) for an array of disorders and impairments, although this intervention has been largely dismissed as placebo-driven (see next section). Beyond EEG-nf, the advent of new technologies for imaging the living human brain has vastly expanded the scope of neurofeedback, which today includes more novel methods such as functional magnetic resonance imaging (fMRI), functional near infrared spectroscopy (fNIRS), and magnetoencephalography (MEG) (TIMELINE and Table 1). Thus, current-day neurofeedback draws on diverse imaging methods to help drive volitional control over electromagnetic and hemodynamic alterations in brain activity (Cannon, 2015; Hammond, 2011; Thibault, Lifshitz, Birbaumer, & Raz, 2015). Within each imaging modality, moreover, researchers have developed distinct neurofeedback protocols that target different brain signals and their concomitant physiological processes (Hammond, 2011; Sulzer, Haller, et al., 2013). Whereas proponents of neurofeedback sometimes lump together these diverse protocols, research findings support some imaging techniques more than others. In this paper we explore the potential merits and shortcomings of modern neurofeedback techniques and contextualize their place in the current technological landscape.

2. EEG neurofeedback

More than half-a-century ago neurofeedback came on the scene promoting the main non-invasive technology of the day, EEG, to “image” the living human brain. Although EEG-nf may still hold some appeal as an alternative to conventional medical treatment, mounting evidence refutes the clinical superiority of feedback training over sham treatment (Arnold et al., 2013; Esmail & Linden, 2014; Lansbergen et al., 2011; Lofthouse, Arnold, Hersch, Hurt, & DeBeus, 2012; Logeman, Lansbergen, van Os, Bocker, & Kenemans, 2010; Perreau-
Linck, Lessard, Lévesque, & Beauregard, 2010; Sonuga-Barke et al., 2013; Thibault et al., 2015; Van Dongen-Boomsma, Vollebregt, Slats-Willemse, & Buitelaar, 2013; Vollebregt, van Dongen-Boomsma, Buitelaar, & Slaats-Willemse, 2014; Zuberer, Brandeis, & Drechsler, 2015). Nonetheless, EEG-nf remains a relatively active clinical field boasting international accreditation boards, specialized academic journals, and over a thousand practitioners (Thibault et al., 2015).

While EEG-nf proponents purport to effectively treat a range of psychological and neurological disorders (Arns, Heinrich, & Strehl, 2014; Cannon, 2015; Hammond, 2011; Tan et al., 2009), it appears that influences other than the feedback itself bring about improvements in clinical endpoints across a range of disorders (Thibault et al., 2015). For example, in the most researched application of EEG-nf—treatment of attention deficit hyperactivity disorder (ADHD)—all double-blind (Arnold et al., 2013; Lansbergen et al., 2011; Vollebregt et al., 2014), and multiple single-blind (Perreau-Linck et al., 2010; Van Dongen-Boomsma et al., 2013), sham-controlled studies demonstrate comparable clinical outcomes in patients receiving veritable-feedback compared with sham-feedback. These studies often compare “relevant” information—i.e., genuine EEG-nf—with feedback from unrelated brain signals—e.g., neural activity from the brain of another participant (Box 1 depicts some of the pertinent ideas and terminology). Notably, double-blind sham-controlled studies, the gold-standard across most clinical research domains, are all but absent from clinical EEG-nf experiments beyond ADHD.

A number of reviews and meta-analyses have addressed the literature regarding EEG-nf treatment for ADHD (Arns et al., 2014; Arns, Ridder, & Strehl, 2009; Lofthouse et al., 2012; Sonuga-Barke et al., 2013), epilepsy (Tan et al., 2009), depression and anxiety (Esmail & Linden, 2014), and a slew of other disorders (Hammond, 2011). Some of these accounts highlight the dearth of convincing evidence (Esmail & Linden, 2014), and a slew of other disorders (Hammond, 2011). This state of affairs suggests that the benefits of EEG-nf for ADHD rely heavily on factors associated with demand characteristics and other contextual parameters rather than contingent feedback per se (Beyerstein, 1990).

<table>
<thead>
<tr>
<th>Underlying Signal</th>
<th>Electrical activity from pyramidal cells perpendicular to the scalp (mainly gyri)</th>
<th>Magnetic fields produced by pyramidal cells perpendicular and tangential to the cortical surface</th>
<th>Blood oxygenation level dependent contrast (which indirectly relates with neuronal activity)</th>
<th>Volume of oxygenated and/or deoxygenated blood (which indirectly relates with neuronal activity)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Typical Feedback Signal Source</td>
<td>One central electrode or a multi-electrode cap</td>
<td>Sensors over sensorimotor cortex</td>
<td>Single brain regions, 3mm x 3mm voxels</td>
<td>Several sensors over sensorimotor cortex</td>
</tr>
<tr>
<td>Feedback delay</td>
<td>&lt; 50 ms</td>
<td>&lt; 50 ms</td>
<td>~1.5 s (plus 4-6 s hemodynamic delay)</td>
<td>~0.5 s (plus 4-6 s hemodynamic delay)</td>
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<td>Resolution</td>
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<tr>
<td>Relevant Literature</td>
<td></td>
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<td>Main Applications</td>
<td>Pediatric ADHD, epilepsy, various psychological disorders</td>
<td>Brain computer interfaces (Experimental)</td>
<td>Psychological conditions (chronic pain, depression, schizophrenia, etc.) (Experimental)</td>
<td>Brain computer interfaces, stroke rehabilitation (Experimental)</td>
</tr>
</tbody>
</table>

Table 1 – Advantages, shortcomings, and applications of neurofeedback imaging modalities.
neural activity (Raz, 2014). Add to that the allure of brain science (Ali, Lifshitz, Cucherat, Arzimanoglou, & Ryvlin, 2008) and the multi-faceted nature of neurofeedback becomes evident: whereas EEG-nf alters both brain patterns and behavioral measures, current findings hardly support a direct link between the specific feedback and these observed alterations.

3. fMRI

Based on fMRI, real-time functional MRI neurofeedback (rtfMRI-nf) offers a non-invasive method to modulate regional

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neural activity. rtfMRI provides a way to collect, process, and display blood oxygen level-dependent (BOLD) fMRI data within a short delay (e.g., 1–2 sec; Weiskopf, 2012). Because of this nearly-concurrent acquisition-processing-display cycle, rtfMRI-nf training putatively permits individuals to see and thereby self-regulate fMRI signals from their own brains. Unlike other EEG-based neurofeedback, rtfMRI-nf training seems to yield precise localization and modulation of relevant brain structures (deCharms, 2008).

Whereas EEG-nf requires many sessions to alter electrical activity, with rtfMRI-nf individuals can selectively modify fMRI BOLD signal within 30 min of training and sometimes even sooner (Brühl et al., 2014; Canterberry et al., 2013; Greer, Trujillo, Glover, & Knutson, 2014; Gröne et al., 2015; Hui, Zhang, Ge, Yao, & Long, 2014; Lawrence et al., 2014; Paret et al., 2014; Sulzer, Sitaram, et al., 2013; Zhang, Yao, Zhang, Long, & Zhao, 2013; Zhang, Zhang, et al., 2013). This striking difference between EEG-nf and rtfMRI-nf deserves further discussion. Learning based on neuro-hemodynamics may occur more quickly because unlike the absence of brain receptors to detect electrical changes, baroreceptors constantly inform the central nervous system about ongoing blood volume (Birbaumer, 2011). Some animal data (e.g., in rats) concerning baroreceptor activation support this speculative account (Dworkin, 1988). However, regardless of hypothetical insights (or lack thereof) regarding possible underlying mechanisms, humans seem to modulate neural hemodynamics faster than they do brain electrical fluctuations. This fact alone makes rtfMRI-nf an attractive vehicle to explore.

The operationalization of rtfMRI-nf affords more focal investigations of higher (e.g., cognitive-behavioral) brain functions and neuroplasticity (deCharms et al., 2004; Weiskopf et al., 2007). For example, healthy individuals showed the ability to self-regulate brain activity in neuroanatomical structures often associated with affect [e.g., insula, amygdala, prefrontal cortex (PFC), and anterior cingulated cortex (ACC)] (Hamilton, Glover, Hsu, Johnson, & Gotlib, 2011; Johnston et al., 2011; Posse et al., 2003; Sitaram, Caria, et al., 2007; Zotev et al., 2011). These findings poised rtfMRI-nf as a potentially effective intervention for several clinical domains, including pain regulation (e.g., deCharms et al., 2005), tinnitus (e.g., Haller, Birbaumer, & Veit, 2010), Parkinson’s disease (e.g., Subramanian et al., 2011) and depression (e.g., Linden et al., 2012; Young et al., 2014). Although these clinical studies represent mostly nascent efforts in line with pilot data and usually draw on small samples and largely unreplicated assays, the emerging tenor from these preliminary findings seems to speak favorably to the clinical potential of rtfMRI-nf. However, a careful evaluation of the literature reveals several caveats to quell this burgeoning assurance.
### Table 2 – Behavioral and clinical rtfMRI-nf experiments.

<table>
<thead>
<tr>
<th>Brain area</th>
<th>Behavioral or clinical target</th>
<th>Relevant rtfMRI-nf experiments</th>
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<tbody>
<tr>
<td>ACC</td>
<td>Pain modulation</td>
<td>deCharms et al. (2005), Emmert et al. (2014), Rance et al. (2014), Guan et al. (2015)</td>
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<td></td>
<td>Smoking cessation</td>
<td>Li et al. (2012), Canterberry et al. (2013)</td>
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<td></td>
<td>Emotional prosody</td>
<td>Gröne et al. (2015)</td>
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<td></td>
<td>Cognitive interference</td>
<td>Mathiak et al. (2015)</td>
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<td>Insula</td>
<td>Emotional valence</td>
<td>Caria et al. (2010), Lawrence et al. (2014)</td>
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<td></td>
<td>Schizophrenia</td>
<td>Ruiz et al. (2013), Cordes et al. (2015)</td>
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<tr>
<td></td>
<td>Psychopathy</td>
<td>Sitaram et al. (2014)</td>
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<tr>
<td></td>
<td>Pain modulation</td>
<td>Emmert et al. (2014), Rance et al. (2014)</td>
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<td></td>
<td>Phobia</td>
<td>Zilverstand et al. (2015)</td>
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<td>Motor cortex</td>
<td>Finger tapping speed (stroke rehabilitation)</td>
<td>Berman et al. (2011), Hui et al. (2014)</td>
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<td></td>
<td>Parkinson’s disease</td>
<td>Subramanian et al. (2011)</td>
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<td></td>
<td>Reaction time</td>
<td>Scharnowski et al. (2015)</td>
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<td></td>
<td>Precision grip</td>
<td>Blefari, Sulzer, Hepp-Reymond, Kollias, and Gassert (2015)</td>
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<tr>
<td>Amygdala</td>
<td>Emotion identification</td>
<td>Zotev et al. (2011)</td>
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<td></td>
<td>Depression</td>
<td>Young et al. (2014)</td>
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<tr>
<td>Prefrontal cortex</td>
<td>Emotional prosody</td>
<td>Rota et al. (2009)</td>
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<tr>
<td></td>
<td>Depression</td>
<td>Linden et al. (2012)</td>
</tr>
<tr>
<td></td>
<td>Working memory</td>
<td>Zhang, Yao, et al. (2013)</td>
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<tr>
<td></td>
<td>Affect</td>
<td>Sarkheil et al. (2015)</td>
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<tr>
<td>Auditory cortex</td>
<td>Chronic tinnitus</td>
<td>Haller et al. (2010)</td>
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<tr>
<td>Functional Connectivity</td>
<td>Smoking cessation</td>
<td>Kim et al., (2015)</td>
</tr>
</tbody>
</table>

One of the early and most notable rtfMRI-nf studies implemented a careful design and reported robust findings that subsequently sparked considerable enthusiasm (deCharms et al., 2005). This study demonstrated that individuals who suffered from chronic pain and received veritable rtfMRI-nf learned to modulate neural activity in the rostral ACC. Moreover, these individuals were subsequently able to govern ACC activity even without feedback, resulting in the desired decreases in subjective pain. Because these effects were absent in control participants, this pivotal report presented important evidence in support of rtfMRI-nf. However, enthusiasm turned into skepticism after several independent replication efforts, including by the original authors, were unable to corroborate the reported findings (Birbaumer, Ruiz, & Sitaram, 2013; Sulzer, Haller, et al., 2013).

One follow-up study included 59 chronic pain patients, rather than the 12 patients in the original study, and found that, while only those receiving veritable feedback gained control over neural activity, pain ratings decreased equally between experimental and sham participants (Sulzer, Haller, et al., 2013). The failure to replicate the original findings may relate to the inflated likelihood of false positives when employing small sample sizes—a general concern across the neurosciences (Button et al., 2013). Alternatively, some neurofeedback experts point out that in the replication effort, researchers provided participants in both the veritable feedback and control conditions with the same mental strategies—derived and optimized during previous neurofeedback experiments (Sulzer, Haller, et al., 2013). Thus, this attempt to replicate intimates that an effective mental strategy may show comparable effects regardless of whether one learns it through neurofeedback or through verbal instruction (Sulzer, Haller, et al., 2013). The promise of this key experiment, therefore, has been largely reneged since advocates had originally heralded its findings.

Most rtfMRI-nf accounts seldom probe behavioral change; instead, they report proof-of-concept experiments demonstrating that participants can alter blood flow to select cortical regions. Out of over 70 published rtfMRI-nf studies, very few reported that participants were unable to modulate brain hemodynamics (Berman, Horovitz, Venkataraman, & Hallett, 2011; Hampson et al., 2011). Using sham-feedback as a control, other rtfMRI-nf experiments demonstrated that participants increased their ability to modulate a particular brain region throughout training (Caria et al., 2007; Caria, Sitaram, Veit, Beglioni, & Birbaumer, 2010; Chiew, Laconte, & Graham, 2012; Hui et al., 2014; Lawrence et al., 2014; McCaig, Dixon, Keramatian, Liu, & Christoff, 2011; Rota et al., 2009; Rota, Handjarias, Sitaram, Birbaumer, & Dogil, 2011; Yoo, Lee, O’Leary, Panych, & Jolesz, 2008; Young et al., 2014; Zotev, Phillips, Young, Drevets, & Bodurka, 2013; Zotev et al., 2011). Many other studies often lacked necessary controls or appropriate analyses to determine that veritable neurofeedback was the primary factor accounting for the observed brain alterations (Thibault et al., 2015); Fig. 3 depicts some common control conditions. Thus, these collective findings raise some concerns about the specificity of neurofeedback, either sham or veritable. As a case in point, a recent study reported that participants receiving sham-neurofeedback expressed increased activation of bilateral insula, ACC, motor areas, and prefrontal areas (Ninaus et al., 2013) — arguably the four most common regions trained in rtfMRI-nf (Table 2). Experimental designs that lack a no-feedback or sham-feedback control cannot dissociate whether the change from baseline relies on feedback or on factors such as mental strategy and the attention and motivation that comes through participation in.
neurofeedback experiments. Thus, well-controlled designs are sorely missing, albeit desperately necessary, to tease apart the specific benefits of rtfMRI-nf.

The roles of “mental strategy” and “sustainability of effect” remain critical lacunae in the understanding of rtfMRI-nf. Mental strategy — the cognitive scheme one follows to achieve the desired goal — seems at least partly responsible for the rapid changes in fMRI BOLD signal. For example, a yet-to-be-published rtfMRI-nf experiment, discussed in an authoritative review (Sulzer, Haller, et al., 2013), showed that participants who received explicit instructions for mental strategies, compared to individuals left to explore and develop them implicitly on their own, were more successful at neural self-regulation. Moreover, comparing verifiable feedback with no feedback (Zhang, Yao, et al., 2013; Zhang, Zhang, et al., 2013), inversely proportional feedback (Sulzer, Sitaram, et al., 2013), or feedback from a distinct brain region (Paret et al., 2014), revealed similar neural changes from baseline to the first trial in both experimental and control groups. On the other hand, findings from both human and animal studies suggest that explicit techniques are unnecessary and that contingent feedback alone is responsible for neural regulation (Caria et al., 2010; Koralek, Jin, Long, Costa, & Carmena, 2012). Thus, the degree to which explicit mental strategies relate to learned brain modulation remains unclear.

In certain experiments, a minimum level of instruction may help participants avoid injurious strategies. For example, because both negative and positive emotions stimulate amygdala activation (e.g., Garavan, Pendergrass, Ross, Stein, & Risinger, 2001; Hamann, Ely, Hoffman, & Kilts, 2002; Yang et al., 2002), some experimenters take care to instruct participants to adopt positive mental strategies (Young et al., 2014; Zotev, Phillips, Young, et al., 2013). Similar considerations surface concerning up-regulation of the anterior insula, which responds strongly to disgust (e.g., Phillips et al., 1997; Wicker et al., 2003), and to modifying ACC activity, which relates to a wide spectrum of cognitive processes (e.g., Bush, Luu, & Posner, 2000; Etkin, Egner, & Kalisch, 2011). The potential for maladaptive mental strategies may present less of a concern for protocols aiming to modify sensorimotor areas through imagined movements.

Sustainability refers to the period of time, after feedback is over, during which participants can demonstrate the learned benefits of rtfMRI-nf. As it turns out, few studies examine sustainability in neurofeedback (Thibault et al., 2015). Whereas some experiments demonstrate that participants retain control over target brain regions (Caria et al., 2007; Mathiak et al., 2015; Robineau et al., 2014; Scharnowski et al., 2015; Zotev, Phillips, Young, et al., 2013; Zotev, Phillips, Yuan, Misaki, & Bodurka, 2013; Zotev et al., 2011), other studies report no such retention (Berman, Horovitz, & Hallett, 2013; Greer et al., 2014; Hamilton et al., 2011; Ruiz et al., 2013; Sulzer, Sitaram, et al., 2013). Thus, the conundrums of mental strategy and sustainability still beg resolution.

Relatively few accounts correlate rtfMRI-nf training with changes in either perception or behavior (Table 2). For example, two reports suggested that volitional dampening of ACC activity would decrease cigarette cravings (Canterberry et al., 2013; Li et al., 2012); however, both experiments feature deficient designs that make it difficult to determine what — e.g., feedback or mental strategy — actually accounted for the decreased desire to smoke (Thibault et al., 2015). Another example draws on analgesia: whereas one study demonstrated that modulation of ACC activity was independent of behavioral parameters such as perceived pain intensity (Rance, Ruttorf, Nee, Rudi Schad, & Flor, 2014), another study found that control over activity in both the ACC and the anterior insula decreased pain perception (Emmert et al., 2014). According to some studies, modulating activity in the insula appears to influence the perceived valence of emotional stimuli (Caria et al., 2010; Ruiz et al., 2013); yet one of these studies had no control group (Ruiz et al., 2013) and another experiment suggested that control over activity in the insula area hardly related to valence ratings (Lawrence et al., 2014).

Beyond the ACC and insula, training motor areas can improve motor control in both healthy participants (Hui et al., 2014) and individuals diagnosed with Parkinson’s disease (Subramanian et al., 2011); however, in the former study the control group also demonstrated improved finger-tapping speed while the latter experiment used no sham-feedback control condition. Altering prefrontal blood flow can improve detection of emotional prosodic intonations (Rota et al., 2009) and verbal working memory (Zhang, Yao, et al., 2013); however, again, the latter experiment demonstrated that sham-neurofeedback enhanced performance on four of the five working memory tasks, even while impairing the ability of participants to modulate target brain regions. On the other hand, control over brain activity in the amygdala appears to bear on emotional measures. Sham-controlled experiments demonstrated that successful amygdala modulation reduced ratings of anxiety and increased ratings of happiness in depressed patients (Young et al., 2014) and inversely correlated with difficulty identifying feelings and susceptibility to others’ anger in healthy participants (Zotev et al., 2011). While neurofeedback may account for some of these outcomes, only experimental designs featuring more robust controls would permit ascertaining such specificity (Fig. 2). Thus, at the end of the day, the literature provides a mixed bag with little by way of conclusive evidence.

Recent studies have increasingly leveraged functional connectivity assays to index the effects of rtfMRI-nf (Emmert et al., 2014; Fukuda, Mitsu, & Hiroshi, 2015; Greer et al., 2014; Kim, Yoo, Tegethoff, Meinlschmidt, & Lee, 2015; Sarkheil et al., 2015; Shen, Zhang, Yao, & Zhao, 2015; Sitaram et al., 2014; Zhang, Yao, Shen, Yang, & Zhao, 2014). One study reported a link connecting functional connectivity, rtfMRI-nf, and behavior (Hui et al., 2014). Compared to sham-feedback controls, verifiable neurofeedback altered brain connectivity in experimental participants: they improved their ability to modulate the prefrontal cortex and increased finger tapping speed (Hui et al., 2014). However, findings from other functional connectivity neurofeedback studies were less conclusive. Two analyses demonstrated post-training functional connectivity changes in participants receiving veritable feedback only, but experimental and control groups improved similarly on behavioral measures (Sarkheil et al., 2015; Shen et al., 2015). Thus, the overarching tenor emerging from these collective studies suggests that functional connectivity analyses scarcely bind neurofeedback with a change in behavior. If and when future experiments establish and cement this relationship, resorting
to functional connectivity analysis may help unravel the neural mechanisms suberving self-regulation.

4. fNIRS

An emerging functional neuroimaging technology, fNIRS offers a relatively non-invasive, safe, and affordable way to monitor brain activity (Ferrari & Quaresima, 2012). fNIRS technology typically uses a portable cap with approximately 50 channels to measure the concentration of deoxygenated blood in superficial brain areas. While fNIRS probes the same underlying hemodynamics as fMRI (Cui, Bray, Bryant, Glover, & Reiss, 2011), each method possesses distinct strengths and weaknesses when applied to neurofeedback. On the one hand, fNIRS is relatively inexpensive, resistant to movement artifacts, and highly portable. On the other hand, whereas fNIRS offers spatial resolution on the order of centimeters at surface regions, fMRI provides millimetric precision throughout the brain (Table 1 provides a more detailed comparison).

Experiments with fNIRS-nf typically aim to show increased control over motor regions by having participants imagine tapping their fingers or clenching their hands. Four studies reported that fNIRS-nf increased activity in a variety of motor areas (Kanoh, Susila, & Miyamoto, 2011; Kober et al., 2014; Mihara et al., 2012, 2013). However, simply moving a limb or covertly tensing a muscle can also modulate motor cortex activity and researchers rarely measure electromyography (EMG) signals to detect such covert muscle tensing. Thus, subtle, possibly subconscious, muscular activity may account for the observed increases in neural function. Proponents of fNIRS claim that this technique highlights how contingent feedback seems key: brain modules change activity as a function of genuine, but not sham, feedback (Kober et al., 2014; Mihara et al., 2013). However, veritable feedback may teach participants to, perhaps involuntarily and without awareness, increase muscle tension rather than develop new mental techniques. Thus, fNIRS-nf of the motor cortex may, in some ulterior fashion, operate as a form of EMG-biofeedback.

Beyond improvement over motor control, one fNIRS-nf experiment reported the successful regulation of prefrontal neural activity to treat ADHD (Marx et al., 2015). However, whereas participants who received fNIRS-nf improved on various measures of ADHD, comparable measures were apparent in controls participants, who received EEG-nf and EMG-biofeedback. Moreover, Marx et al. (2015) shy away from reporting whether participants learned to regulate the fNIRS signal. Thus, the link between improved symptomatology and fNIRS-nf remains tenuous.

5. MEG

MEG is an imaging modality that measures the amplitude of magnetic signals emitted from both superficial and deeper cortical areas. To record such activity, participants typically sit in a magnetically shielded room while placing their head inside a helmet-shaped concave dewar, which features hundreds of magnetometers — sensors capable of detecting small changes in the magnetic field. Because such specialty apparatuses require an expensive set-up and substantial support facilities, MEG-nf largely remains the purview of brain researchers with access to advanced imaging centers. In line with EEG, MEG too provides a direct — rather than indirect (e.g., in the case of fMRI and fNIRS) — measure of neural activity. However, unlike with electrical signals, the skull and cephalic tissues hardly distort the magnetic field, thereby allowing magnetometers to accurately measure minute variations in neural activity. In addition, compared to other imaging modalities, MEG affords an impressive spatial resolution and signal-to-noise ratio (Table 1 provides a detailed comparison).

Healthy participants using MEG-nf learned bidirectional control of neural activity, in a subset of the 9—15 Hz range over the motor cortex (often called the sensorimotor- or μ-rhythm), within 64 min (Mellinger et al., 2007), and for neural sources involved in motor imagery in eight sessions of 12—18 min (Florin, Bock, & Baillet, 2014). However, as in the case for fNIRS-nf, here too participants may have unknowingly relied on limb muscles to increase cortical activation. This option receives further credence in findings coming from stroke patients, who manifest impaired muscular control and take much longer—13 to 22 sessions of one-to-two hours—to govern the MEG signal (Buch et al., 2008). Moreover, with one exception (Gallegos-Ayala et al., 2014), patients lacking control of all muscles (e.g., completely locked-in patients) have been entirely unsuccessful in maintaining control over neuroimaging signals (De Massari et al., 2013). Rather than neuro-feedback, therefore, muscle tension — perhaps outside of common awareness — may drive such improvements.

Two recent MEG-nf studies included behavioral measures and employed no-feedback (Boe et al., 2014) and sham-feedback (Okazaki et al., 2015) controls. One study demonstrated that only participants provided with feedback learned to increase source level activity at the primary sensorimotor cortex in three 40-min sessions (Boe et al., 2014). This experiment included EMG sensors placed on the flexor and extensor muscles of the fingers — reducing, but not fully eliminating the potential of muscle contamination from bodily sources. Interestingly, both control and experimental groups improved finger dexterity equally on the only behavioral task included. The other experiment separated participants into two groups and provided half of each group with sham-feedback (Okazaki et al., 2015). Over a single 10-min feedback session, one group attempted to increase posterior alpha lateralization to the left, while the other group aimed to lateralize the same activity to the right. One experimental group succeeded at increasing alpha lateralization and expressed an associated impairment in visual detection threshold. This account reported an absence of additional predicted changes in brain and behavior and scantily established a benefit to MEG-nf. Thus, whereas early MEG studies may have sparked excitement concerning new clinical prospects, more experiments would need to carefully control for the aforementioned caveats in order to verify the effectiveness of MEG-nf as a viable technique to modify brain and behavior.
6. Future steps

The breadth of neurofeedback techniques has greatly expanded with the advent of novel real-time imaging modalities. As we have discussed, however, the influence of neurofeedback on behavior remains largely indeterminate. Functional connectivity now pervades the neurofeedback literature as multi-modal imaging approaches the fore (Zotov, Phillips, Young, et al., 2013; Zotov, Phillips, Yuan, et al., 2013) and higher magnetic fields (e.g., MRI scanners operating at 7 T) rekindle enthusiasm amongst researchers (Gröne et al., 2015). Scientists are beginning to unravel how feedback training rearranges neural circuits; they may even begin to elucidate the feasibility of concurrent regulation over multiple distinct brain signals. And yet, for neurofeedback to earn the application of a solid intervention, we must establish the clinical significance of its purported therapeutic effects.

Whether feedback, rather than other ulterior factors, propels the neural changes requires further elucidation. Proof-of-concept neurofeedback experiments abound, yet few studies extend these efforts with more robust experimental paradigms. To establish whether a specific neurofeedback signal accounts for learned brain modulation, control conditions are imperative (Fig. 2 depicts some key ideas). In addition to presenting control participants with sham-feedback, researchers should also aim to match less obvious factors between control and experimental groups. For example, some neurofeedback practitioners argue that the amount of positive feedback participants receive influences training outcome (Arns et al., 2014).

Providing sham-feedback based on the brain activity from previously tested experimental participants would control for the level of positive feedback. Such control participants, however, may lose motivation if they experience an absence of control over the feedback signal and begin to believe they are receiving non-contingent feedback. Many clinical studies highlight that the majority of research participants can correctly guess their group assignments (i.e., experimental vs placebo; Shapiro & Shapiro, 1997). In some contexts, moreover, positive outcome relies more heavily on the belief of veritable treatment than on the actual treatment administered (Dar, Stronguin, & Etter, 2005). Providing a second control group with feedback from a different area of their brain could help equalize the belief that they are receiving veritable feedback from the region of interest. To confirm the eftiveness of the control groups, researchers could collect phenomenological assays, including sense of control over the signal, motivation, and believed group assignments. Moreover, double-blind designs wherein both researchers and participants are unaware of group assignment would obviate any potential influence experimenters could unknowingly exert on participants (and vice versa). Such experimental designs hold the potential to thrust neurofeedback beyond the proof-of-principle stage while unraveling how ulterior factors, such as mental strategy and exposure to neurofeedback may contribute to neural self-regulation.

While more rigorous controls will likely help elucidate the putative relation between neurofeedback and brain modulation, additional conditions would further link feedback training to behavioral changes. For example, it would behoove neurofeedback proponents to study clinical populations and measure their behavior. In such experiments, researchers could provide some participants with inversely proportional feedback [i.e., present positive feedback to control participants for modulating the same signal of interest as the experimental group, yet in the opposing direction: thus, reversing the reward contingency (Box 1)]. If inverse-controls and experimental participants learn to modulate brain signals in opposing directions while manifesting opposite behavioral changes, these findings would help cement the link between neurofeedback, brain self-regulation, and behavior.

Experimental designs leveraging sham control groups pose researchers with some ethical conundrums; from withholding effective treatment to engaging in potentially harmful procedures. Neurofeedback, however, has yet to amass the credence commensurate with the appellation of “evidence-based medicine” (Sulzer, Haller, et al., 2013; Thibault et al., 2015). As such, employing sham controls falls short of qualifying as withholding effective treatment. Moreover, researchers employing sham controls could administer proper neurofeedback to all participants, after completion of the experiment, to ensure that all eventually receive veritable treatment – as is common in many clinical trials.

Implementing inverse feedback may appear unethical because it could theoretically worsen symptoms or impair behavior. And yet, the neurofeedback literature hardly supports these concerns. A few early EEG-nf experiments leveraged inverse controls for intractable epileptics (Lubar et al., 1981; Sterman & MacDonald, 1978) and found comparable decreases in seizure frequency after veritable and inverse feedback sessions (Thibault et al., 2015). As for rtfMRI-nf, only one study employs an inverse feedback control group and the authors do not report any adverse effects (Sulzer, Sitaram, et al., 2013). Interestingly, looking across experiments we find many cases of distinct protocols encouraging regulation in opposing directions. For example, many experiments train participants to up-regulate anterior insular activity to increase valence ratings (Caria et al., 2010; Lawrence et al., 2014; Ruiz et al., 2013; Sitaram et al., 2014), while others encourage down-regulation to decrease pain perception (Emmert et al., 2014; Rance et al., 2014) or subdue a phobia (Zilverstand, Sorger, Sarkheil, & Goebel, 2015). Likewise, multiple experiments train down-regulation of the ACC to mitigate pain (deCharms et al., 2005; Guan et al., 2015) or inhibit cigarette cravings (Canterberry et al., 2013; Li et al., 2012), while others encourage ACC activity to heighten valence ratings (Gröne et al., 2015) or limit the effects of cognitive interference (Mathiak et al., 2010). Thus, researchers are already training regulation in opposite directions, but shy away from directly comparing the behavioral effects of up- versus down-regulation using a specific pattern of brain activity. To date, the reported negative effects of rtfMRI-nf plateau at mental fatigue and the kind of physical discomfort typical to brain imaging environments (Hawkinson et al., 2012; Sulzer, Haller, et al., 2013). In the absence of adverse reactions, therefore, ethical arguments against the use of inverse control conditions remain tenuous.

If well-controlled studies reinforce the connection between feedback, brain activity, and behavior, follow-up studies could then explore the feasibility of implementing such methods. The current literature concerning neurofeedback, however,
largely evades the obligation of following circumspect exper-
imental designs.

7. Conclusion

Neurofeedback is in vogue. Despite a vigorous revival over the past decade, recent accounts highlight a host of methodological and interpretational caveats that pervade the relevant scientific literature. Although neurofeedback draws on a diverse array of imaging methods, rudimentary questions regarding key concepts, such as sustainability and specificity, continue to linger largely unaddressed. While neurofeedback appears to help some participants gain the capacity for brain modulation, the relative contribution of specific feedback compared to ulterior factors remains unclear. At the moment, sparse behavioral measures, little follow-up sessions, and many methodological caveats preclude formal endorsement of neurofeedback as a clinical treatment vehicle. Although the jury is still out, additional judicious experiments and more compelling findings will have to further demonstrate the seductive, albeit yet unconfirmed, clinical promise of neurofeedback.

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