

# **Title: A unified functional network target for deep brain stimulation in obsessive-compulsive disorder**

*Short title: Functional network target for DBS in OCD*

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## Abstract

**Background:** Multiple deep brain stimulation (DBS) targets have been proposed for treating intractable obsessive-compulsive disorder (OCD). Here, we investigated whether stimulation effects of different target sites would be mediated by one common or several segregated functional brain networks.

**Methods:** Seeding from active electrodes of four patient cohorts (N = 50) with DBS to anterior limb of the internal capsule or subthalamic nucleus zones, optimal functional connectivity profiles for maximal Yale-Brown Obsessive-Compulsive Scale improvements were calculated and cross-validated in leave-one-cohort-out and leave-one-patient-out designs. Second, we derived optimal target-specific connectivity patterns to determine brain regions mutually predictive of clinical outcome for both targets and others that were predictive for either target alone. Functional connectivity was defined using resting state functional MRI data acquired in 1,000 healthy participants.

**Results:** Optimal functional connectivity profiles showed commonalities and differences between target sites, while cross-predictions of improvements remained robust across cohorts and targets, suggesting a shared network. Connectivity to anterior cingulate cortex, insula, and precuneus, among other regions, was predictive regardless of stimulation target. Regions with maximal connectivity to these commonly predictive areas included insula, superior frontal gyrus, anterior cingulate cortex and anterior thalamus, as well as the original stereotactic targets.

**Conclusions:** Pinpointing the network modulated by DBS for OCD from different target sites identified a set of brain regions to which DBS electrodes associated with optimal outcomes were functionally connected – regardless of target choice. On these grounds, we elucidate potential brain areas that could prospectively inform additional or alternative neuromodulation targets for OCD.

## Introduction

For approximately 10% of obsessive-compulsive disorder (OCD) patients refractory to conventional first-line interventions (1), deep brain stimulation (DBS) and stereotactic ablation provide effective alternative treatment options (2,3).

Since the early times of stereotactic neurosurgery, disrupting dysfunctional networks has ranked prominently among treatment strategies for brain disorders (e.g., via *ansotomy* – or lesioning of the ansa lenticularis – as performed by Spiegel and Wycis or others in 1954 (5)). From as early as ~1888 on, an underlying rationale was to alleviate pathological communication between cerebral subsystems via their separation (4). Modern neuroimaging methods are now beginning to facilitate mapping such remote network-level effects in a more systematic and model-driven fashion (6–18). These more recent studies may confirm the impact of DBS on distributed pathological brain networks and allow anticipating essential network targets for maximized efficacy (19).

Severe forms of OCD have been termed *circuitopathies* (20,21), suggesting optimal treatment responses to result from rebalancing exactly the affected brain networks. Specifically, current views propose OCD symptoms to arise from i) excessive activity in direct over indirect cortico-striato-pallido-thalamo-cortical (CSTC) loops, and ii) abnormal frontostriatal connectivity (22–30). In this concept, the dACC, which mediates conflict processing between affective (orbitofrontal) and cognitive (dorsolateral prefrontal) inputs plays a central role (31). In case of OCD, an imbalance favoring limbic over associative loops leads to inappropriate dACC activity, inducing a pathological state of error/conflict signaling that may involve hyperdirect input to the associative/limbic STN (32).

While predominantly embedded within this circuitry, no definite agreement about one single optimal DBS target or network for OCD has yet been reached (33). Common targets include limbic/associative gray matter nuclei such as the anteromedial subthalamic nucleus (amSTN) (34–36), nucleus accumbens (NAcc) (37), or ventral striatum (VS)/ventral capsule (VC) (38). Other proposed targets include fronto-subcortical white-matter fiber bundles, such as the anterior limb of the internal capsule (ALIC) (39), bed nucleus of the stria terminalis (BNST) (40), inferior thalamic peduncle (ITP) (41,42), or ventral tegmental area projection pathway (VTApp; formerly termed superolateral medial forebrain bundle; sIMFB) (43).

Critically, similar response rates (in terms of global obsessive-compulsive symptomatology) (2), along with anatomic juxtaposition between some of these targets may imply a unifying pathophysiological mechanism (44). Moreover, given stimulation spread, anatomical diversity of local neuronal and fiber distributions, slight targeting imprecisions, and the wide span of electrode contacts (between 6 and 21 mm across ventral and dorsal contact centers), on average, differences in target definition are often merely semantic (12). Broadly, most targets can be divided into an ALIC/striatum zone (including ALIC, BNST, VS/VC, and NAcc), and one at the junction between brainstem and forebrain (including amSTN and VTApp). We will thus use the terms ALIC-DBS and STN-DBS to describe these zones, unless otherwise specified.

Recently, clinical outcome following stimulation to either region was shown to be associated with optimal treatment outcome when the same specific white matter tract (passing through both areas) was modulated (12). Based on this tract, improvements could be predicted across DBS targets and independent datasets from different DBS centers (12). The tract traversed the ALIC focally, would macroscopically center around the frontopontine projection pathway (also termed *Arnold's bundle*), and may functionally comprise hyperdirect prefrontal input to STN and midbrain structures.

However, structural network analyses (such as conducted with diffusion-MRI based tractography, dMRI) bear the short-coming of neglecting indirect connections across distributed functional (whole-brain) networks. Here, we hence build on a previous dMRI study (12) to investigate differences and commonalities of *functional* networks associated with either OCD target category. Based on resting-state functional MRI (rs-fMRI) data acquired across 1,000 healthy subjects, we first identify optimal networks in relation to clinical outcome across the whole patient group and cross-validate these via leave-one-cohort-out and leave-one-patient-out designs. Crucially, functional connectivity is modeled via rs-fMRI data from healthy subjects, while seed regions are obtained from patients. Second, we inspect specificity of and differences in optimal networks between STN vs. ALIC cohorts. We then compare a connectivity profile predictive for *both* targets (i.e., an *agreement map*) to a full-cohort map in terms of predictive utility. Considering the agreement map's superior ability in explaining outcome variance, it might represent the set of connections a DBS electrode should optimally display – regardless of target choice. Finally, we raise the question: Which voxels in the brain maximally display exactly this set of connections? We therefore *invert* the agreement map to identify brain regions with a connectivity profile maximally similar to this map and validate it against a meta-analytic connectivity pattern functionally linked to OCD. This step intends elucidating and confirming stimulation targets – or network *sweet-spots* – with connectivity profiles most strongly resembling the agreement map.

## Methods and Materials

### Patient cohort

Detailed methodological descriptions can be found in supplemental material. Analyses were based on retrospective data of the same 50 OCD patients from four centers as described in (12), comprising *Cologne* (N = 22 ALIC-DBS patients), *Grenoble* (N = 14 STN-DBS patients), *Madrid* (N = 8 NAcc-DBS patients), and *London* cohorts (N = 6 patients with bilateral electrodes targeting both STN and ALIC). Since, as shown in (12), placements of NAcc electrodes (Madrid sample) on average were similar to the ones of ALIC electrodes (Cologne and London samples), we subsumed NAcc electrodes conforming to an ALIC zone going forward. For more comprehensive clinical descriptions, see **Table 1** (largely adopted from (12)).

Pre- and postoperative OCD symptom severity was assessed on the Yale-Brown Obsessive-Compulsive Scale (Y-BOCS) (45). Information on imaging data, electrode localization and volumes of tissue activated (VTAs) estimation performed in Lead-DBS software (<http://www.lead-dbs.org>) (46,47) is detailed in supplements.

### DBS connectivity analysis

Functional connectivity analyses were based on a normative functional connectome, calculated from rs-fMRI scans of 1,000 healthy subjects (48) acquired within the Brain Genomics Superstruct Project (<https://dataverse.harvard.edu/dataverse/GSP>; (49)). Similarly to previous studies (6,9), voxel-wise correlations between time-series of voxels inside each bilateral VTA and of remaining whole-brain voxels were calculated per OCD patient, averaged across the normative sample, and Fisher-z-transformed (**Figure 1 A**). On a voxel-by-voxel basis, the resulting, patient-specific fingerprints were correlated with Y-BOCS improvements across patients to generate an *R-map model*

of “optimal” connectivity profiles for maximal Y-BOCS improvement (**Figure 1 B**). This R-map models regions to which DBS electrodes ideally should (red regions, **Figure 1 B**) or should not be connected (blue regions) to maximize Y-BOCS improvement. More details are displayed in the supplemental material.

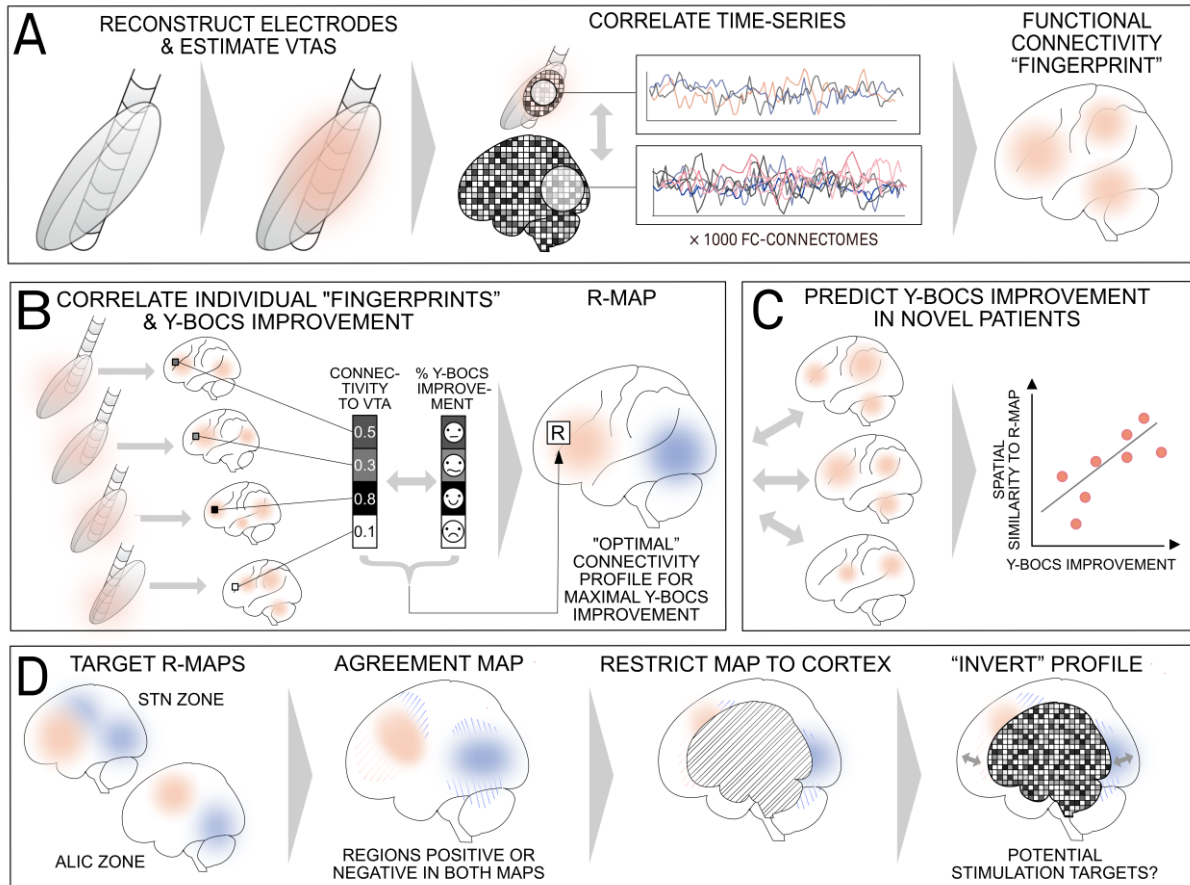
If the R-map itself represents a model of optimal connectivity, higher similarity of individual fingerprints to the model would link to better clinical outcome. To test this assumption, spatial similarity metrics between individual *fingerprints* and the *R-map* were used to predict clinical improvements of *novel* patients (**Figure 1 C**). If not otherwise specified, circularity of analyses was avoided by calculating R-maps on the basis of *one* part of the sample to predict outcomes in a *different* part, specifically using i) leave-one-cohort-out, and ii) leave-one-patient out designs.

Based on *target-specific* R-maps (i.e., one distinct map for ALIC and STN zones, respectively), an *agreement map* was derived with the aim of quantifying overlaps of optimal connectivity across DBS targets. R-maps from these two cohorts were multiplied with each other while retaining only voxels that were either positive or negative simultaneously in *both* maps (**Figure 1 D**). Voxels negative on both maps were kept negative despite their (absolute) values being multiplied.

To identify subcortical target *sweet spots* that could define optimal neuromodulation sites, we calculated which voxels throughout the whole brain had connectivity profiles that were maximally similar to the resulting agreement map. Since the agreement map itself included subcortical targets, we first restricted its profile to cortical (and cerebellar) regions (this step did not alter results; see supplements). Using a voxel-wise precomputed connectome of the same rs-fMRI dataset (49), we then identified brain-wide voxels with a cortical connectivity profile most similar to the (cortical) agreement map, termed *inverted map* (**Figure 1 D; Figure 6**). These regions



exhibit a connectivity profile that could be considered “optimal” based on data from cohorts operated with two different DBS targets.



**Figure 1. Schematic of methodological pipeline.** (A) First, deep brain stimulation (DBS) electrode localizations were reconstructed in stereotactic standard space, and volumes of tissue activated (VTAs) (red) were estimated based on individual stimulation parameters. Using a normative high-resolution resting-state functional MRI (rs-fMRI) connectome derived from data of the Brain Genomics Superstruct Project (Holmes et al., 2015), average time-series of voxels within bilateral VTAs and of every other whole-brain voxel were correlated. Per patient, this analysis was performed 1,000 times via data of each subject in the normative sample (N = 1,000 healthy subjects) and subsequently averaged, resulting in one functional connectivity fingerprint for each patient. (B) Depending on respective DBS electrode placement and individual stimulation parameter settings, these fingerprints varied between patients. Voxel-wise correlation between patients' fingerprints and individual Yale-Brown Obsessive-Compulsive Scale (Y-BOCS) change scores resulted in an R-map, whose spatial distribution by itself estimated an “optimal” connectivity profile of DBS electrodes. Hence, voxels with positive correlations (red) depict areas to which connectivity seeding from the VTA were associated with beneficial outcome, and vice versa for anticorrelated voxels (blue). (C) Probing this R-map as a predictive model, clinical outcomes of

novel patients (not used to define the model) were estimated based on spatial correlation (i.e., similarity) between the R-map and each novel patient's functional connectivity fingerprint. **(D)** For each of the two target zones (anterior limb of the internal capsule, ALIC, and subthalamic nucleus, STN), one separate R-map model was estimated. Based on both, an agreement map was calculated by discarding voxels in which the signs on both R-maps were conflicting and multiplying their absolute values when signs were in agreement. The resulting map may constitute an optimal connectivity map that applies for both target sites. Finally, we aimed to determine which regions in the brain were maximally connected to this map. To do so, connectivity profiles of every single voxel were compared (by spatial correlation) to the cortical surface of the agreement map. The resulting "inverted" map peaked at regions maximally connected to the agreement map. FC, functional connectivity.

For correlation analyses comparing predicted and empirical improvements, p-values were obtained via Monte-Carlo random permutations ( $\times 1,000$ ). This procedure is free from assumptions about distributions (i.e., Student t for R-values), which are typically violated in small sample sizes. Scatterplots were visualized with 95% confidence bounds (gray or light-red areas).

Finally, we validated our inverted connectivity profile against a meta-analytic brain network functionally implicated in OCD. Specifically, by computing a voxel-wise spatial correlation coefficient across the brain, we tested how this inverted map compared to a pattern of activations linked to the search term "obsessive compulsive" entered into NeuroSynth ([www.neurosynth.org](http://www.neurosynth.org); term-based meta-analyses) (50). This search engine automatically derives meta-analytic whole-brain activation maps based on text mining across neuroscientific publications to avoid biased study selection.

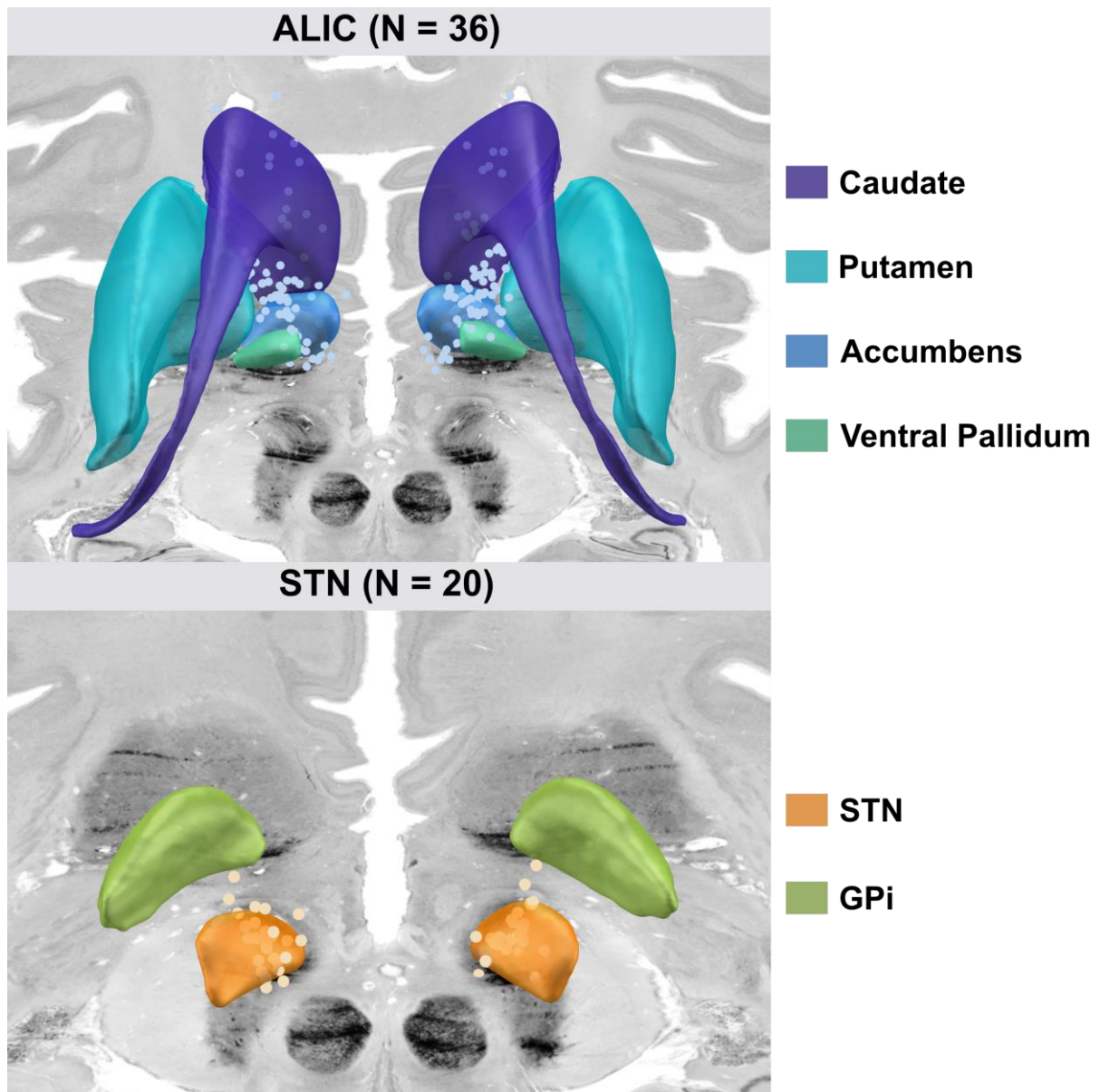
## Results

Age, as well as relative pre- to post-surgical Y-BOCS improvements were comparable across centers (see **Table 1** for details). Electrode localization confirmed accurate placement in both ALIC and STN target zone cohorts (**Figure 2**).

**Table 1: Patient demographic details and clinical results of the four cohorts.**

	ALIC-DBS Cohort (Mean ± SD)	STN-DBS Cohort (Mean ± SD)	NAcc-DBS Cohort (Mean ± SD)	Combined DBS Cohort (Mean ± SD)
Center	University Hospital Cologne	University Hospital Grenoble	Hospital Clínico San Carlos Madrid	University Hospital London
Reference(s)	(7,84)	(116)	(85)	(70)
N patients (females)	22 (12)	14 (9)	8 (4)	6 (1)
N electrodes	44	28	16	24
Age	41.7 ± 20.5	41 ± 9	35.3 ± 10.4	45.5 ± 10.5
Y-BOCS baseline	31.3 ± 4.4	33.4 ± 3.7	30 ± 7.75	36.2 ± 1.8
Y-BOCS after DBS <i>(12 months postop)</i>	20.7 ± 7.7	19.6 ± 10.6 <i>(12 months postop)</i>	14.75 ± 7.2 <i>(3 months postop of best (optimized phase in (70)) contact)</i>	14.3 ± 4.1
Absolute Y-BOCS improvement	9.6 ± 6.5	13.8 ± 10.8	15.1 ± 9.6	21.83 ± 5.7
%-Y-BOCS improvement	31.0 ± 20.5 %	41.2 ± 31.7 %	47.8 ± 23	60.2 ± 12.7 %

Abbreviations: **ALIC**, anterior limb of the internal capsule; **DBS**, deep brain stimulation; **NAcc**, nucleus accumbens; **SD**, standard deviation; **STN**, subthalamic nucleus; **Y-BOCS**, Yale-Brown Obsessive-Compulsive Scale.



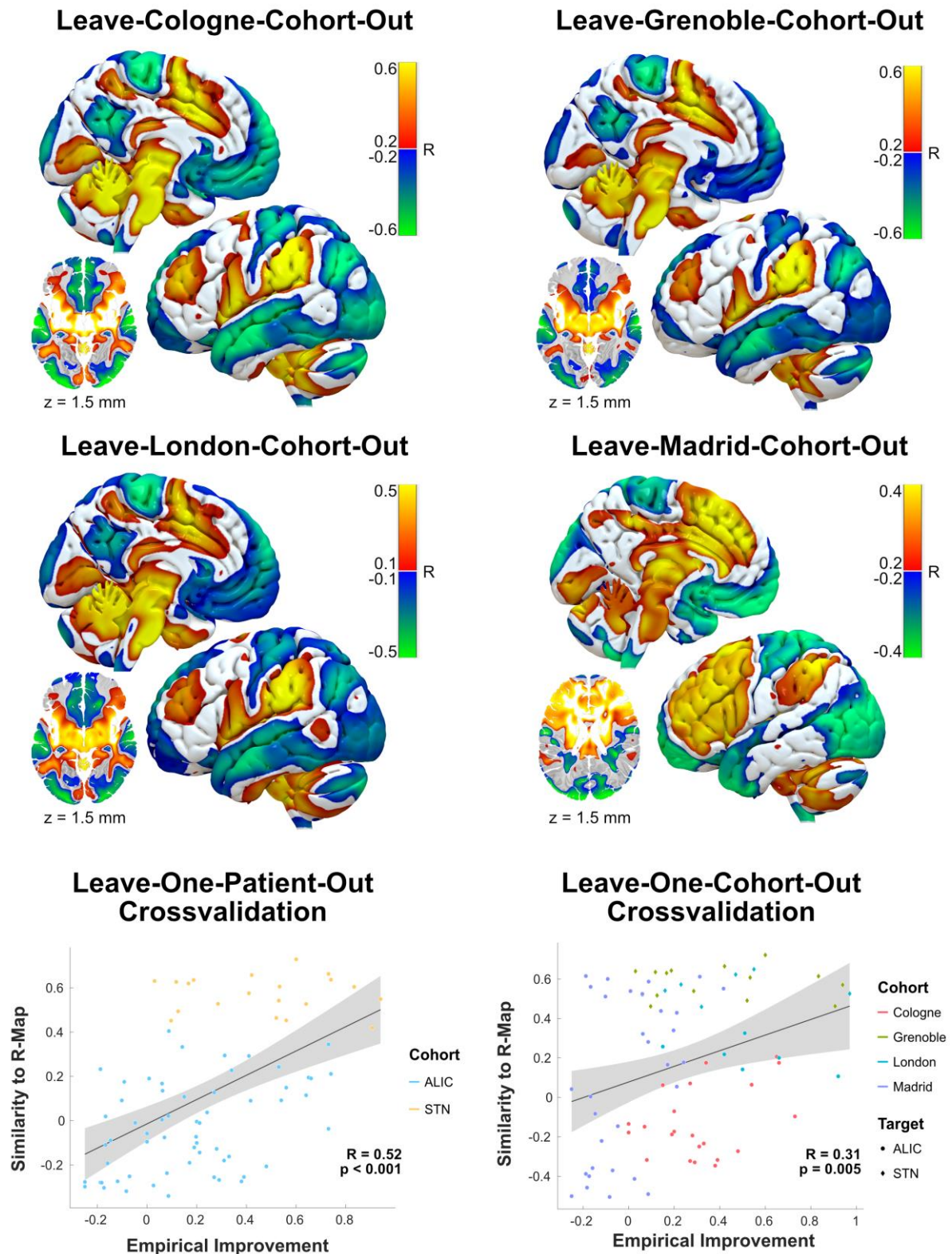
**Figure 2. Overview of lead electrode placement.** Active electrode contacts from the ALIC and STN cohorts are shown. Note that patients from the Madrid cohort were implanted with a Medtronic 3391 type lead which spans across 21 mm from ventral to dorsal contact centers, hence the larger spread of contacts extending to dorsal in this cohort (top panel). Subcortical structures defined by CIT-168 Reinforcement Learning atlas (114) (anterior limb of the internal capsule zone, ALIC) and DISTAL atlas (115) (subthalamic nucleus zone, STN), with coronal and axial planes of the BigBrain template (116) as background. GPi, globus pallidus internus.

First, an R-map model accounting for the overall effect *regardless* of DBS target was calculated based on the full patient cohort. Then, we cross-validated this model using leave-one-patient-out (**Figure 3** bottom left), as well as leave-one-cohort-out designs (**Figure 3** bottom right). Leaving one cohort out did not largely affect the spatial R-map

configuration. The four maps were used to cross-predict variance in clinical improvements of the left-out cohort via spatial correlation between left-out patients' connectivity fingerprints and the respective R-map model (**Figure 3** bottom right side;  $R = 0.31$ ,  $p = 0.005$ ). The R-map calculated on all patients (see **Figure 4** bottom left) associated positive connectivity between electrodes and insula as well as cingulate cortex most prominently with clinical response. Notably, instead of showing how the two targets involved the *exact* same network, our aim here was to investigate both across-target commonalities *and* differences. Notwithstanding robust findings across cohorts, we must emphasize that higher numbers of ALIC-DBS as STN-DBS patients may have driven effects. The following analyses investigated this further.

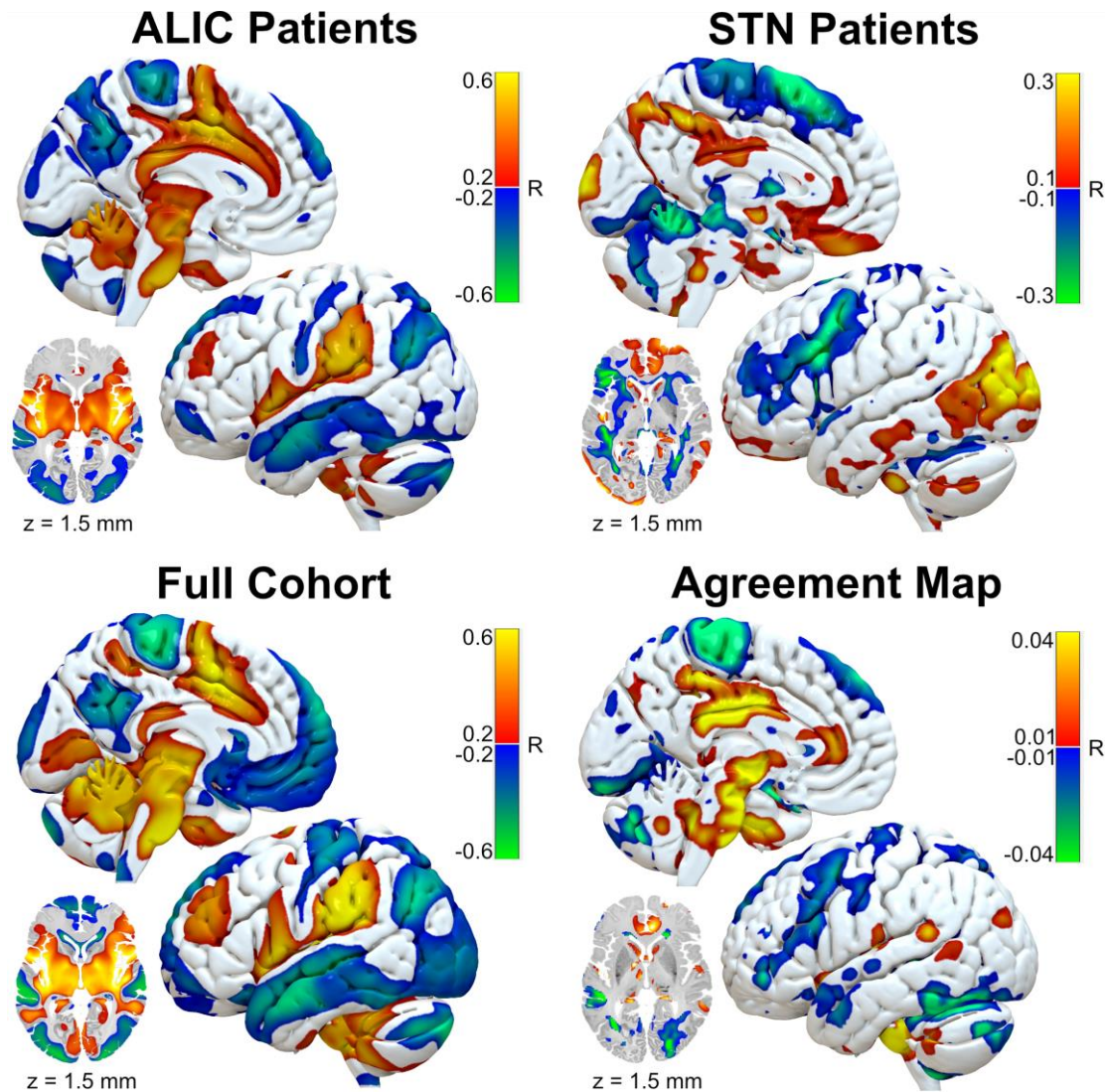
Apart from robustness of an *overall* network predictive of clinical improvement across cohorts (**Figure 3**), we were also interested in *differences* between the two target zones when analyzed separately. Thus, R-map models were derived from two sub-sets of the full sample split based on target region (ALIC vs. STN zone; **Figure 4**). While connectivity to some brain regions was associated with optimal clinical outcomes regardless of DBS target (such as anticorrelations to sensorimotor cortex or positive connections to striatum, insula and specific cingulate regions), others did not overlap. For instance, regions such as the orbitofrontal cortex (OFC) were highlighted by the STN but not the ALIC R-map, and regions such as the supramarginal gyrus showed the opposite pattern.





**Figure 3. Leave-One-Patient-Out and Leave-One-Cohort-Out cross-validations of the R-map model.** Patients were divided by DBS centers into four cohorts. The first four panels show correlation maps between functional connectivity fingerprints (seeding from bilateral VTAs in each patient to the rest of the brain) and clinical improvements. Maps are displayed in form of brain surface overlays together with an additional axial slice at  $z = 1.5$  mm, with yellow to red colormap indicating positive R-values and blue to green colormap indicating negative R-values. Results of the leave-one-patient-out cross-validation are shown on the bottom left panel, while results of the

leave-one-cohort-out cross-validation are presented on the bottom right panel. Here, each map was calculated on data of three out of the four cohorts and used to cross-predict outcomes in patients from the respectively left-out cohort. Axial planes show the BigBrain template (116) as background. ALIC, anterior limb of the internal capsule zone; STN, subthalamic nucleus zone.

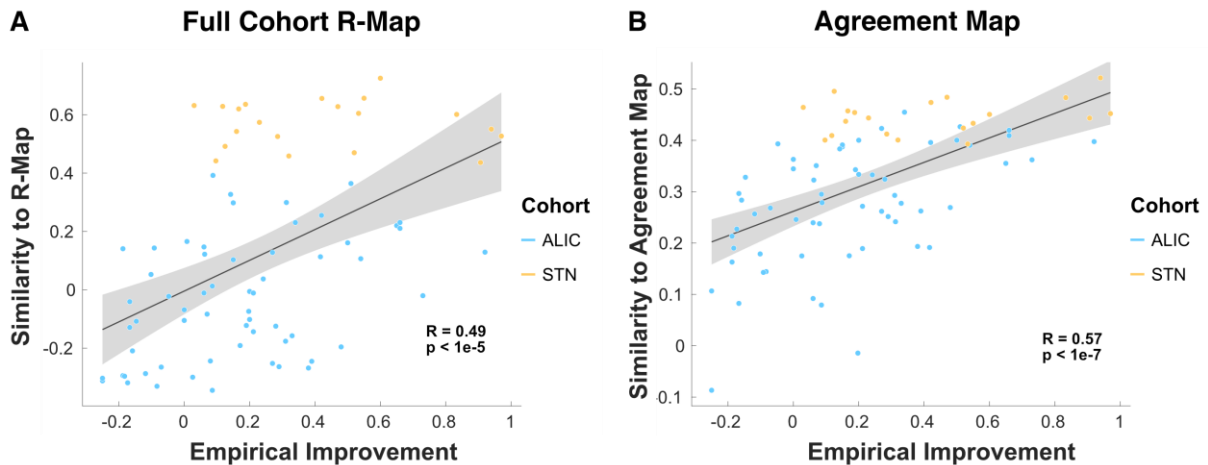


**Figure 4. Functional connectivity profile of cohorts divided by target zone.** R-maps were calculated by correlating functional connectivity fingerprints (seeding from bilateral VTAs in each patient to the rest of the brain) and clinical improvements. R-maps from anterior limb of the internal capsule (ALIC; top left) and subthalamic nucleus (STN; top right) zone cohorts separately, from full cohort combined (bottom left), and agreement map based on the ALIC and STN maps (bottom right) are displayed. Maps are shown as brain surface overlays along with an additional axial slice at  $z = 1.5$  mm, with yellow to red colormap indicating positive R-values and blue to green colormap indicating negative R-values. Anatomical regions of these results are reported in supplementary Table S1. Axial planes show the BigBrain template (116) as background.

We hypothesized that areas convergently identified by both maps could be more specific to the true DBS effect (while others could have resulted from indirect connections revealed in rs-fMRI connectivity analyses or spurious correlations). Hence, in addition, an agreement connectivity map between ALIC-DBS and STN-DBS cohorts was calculated by rejecting areas without agreement across the two maps (**Figure 4** lower right). Of note, while the full cohort map highlighted the dACC, the STN map was only mildly positively correlated here (but correlated to other ACC regions). As a consequence of coefficient multiplications, the agreement map almost showed a gap in dACC but highlighted rostral and posterior ACC.

To test whether this approach could improve utility of models, we computed the amount of variance explained using either the full-cohort R-map or the aforementioned *agreement map* (**Figure 5**). While the full-cohort R-map accounted for ~24% of the variance in clinical improvements ( $R = 0.49$ ,  $p < 1e-5$ ), the agreement map did so for ~32% ( $R = 0.57$ ,  $p < 1e-7$ ). In the ALIC cohort, the full map explained ~18% ( $R = 0.43$ ,  $p < 1e-3$ ), and the agreement map ~27% ( $R = 0.52$ ,  $p < 1e-4$ ) of variance. In the STN cohort, neither map accounted for a significant amount of variance in pre- to postoperative Y-BOCS change, but the association was stronger in the agreement map ( $R = -0.07$  vs.  $R = 0.31$ ). We must emphasize that differences in predictive utility were not significant ( $p = 0.127$ ,  $0.211$ ,  $0.095$  for combined, ALIC and STN targets, respectively) and are meant as direct head-to-head comparisons of each map's utility that at best may lead to anecdotal evidence.

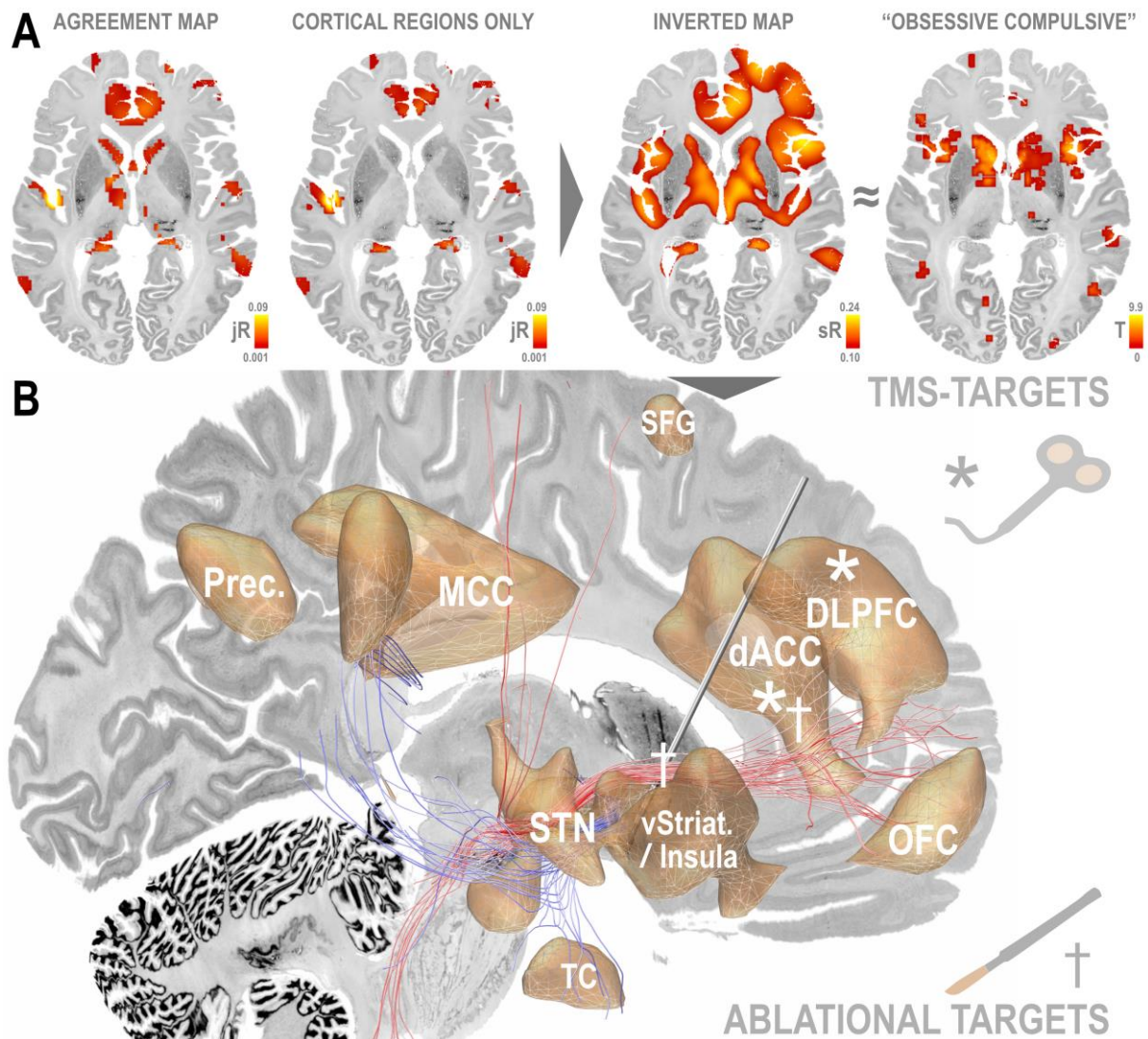




**Figure 5. Amount of variance in clinical improvement that could be explained using the full cohort R-map (left) vs. agreement map (right) models. (A)** Correlation between clinical improvement and spatial similarities between network fingerprints and the R-map model calculated across all patients of the full cohort. **(B)** Correlation between clinical improvement and spatial similarities between network fingerprints and the agreement map model. Please note that the aim of the figure is to directly compare the full cohort R-map model with the agreement R-map model. Correlations shown in this figure do not represent actual (out-of-sample) predictions since all data points were used to calculate both models. ALIC, anterior limb of the internal capsule; STN, subthalamic nucleus.

By integrating the agreement R-map with a preprocessed version of the functional connectome (48), we “*inverted*” its connectivity profile to find regions with connectivity profiles of strongest similarity to this map (**Figure 6**). For each brain voxel, this map estimated via spatial correlation *how similar* its connectivity profile was to the agreement map. If the agreement pattern indeed consisted of an optimal across-target connectivity profile of effective DBS, the inverted map would highlight regions *exhibiting* exactly this optimal connectivity profile. The resulting *inverted* map comprised the two DBS targets of study (ALIC and STN). Peak voxels resided at similar locations (~2 mm distance) to average active stimulation contacts in both regions (see **Table 2**; ALIC: DBS stimulation site:  $\pm 10.73, 9.20, -2.48$  mm; peak voxel on inverted map:  $\pm 11.46, 7.36, -2.37$  mm; STN: DBS stimulation site:  $\pm 10.04, -11.52, -7.84$  mm; peak voxel on inverted map:  $\pm 9.40, -10.01, -6.91$  mm). However, the inverted map also

highlighted common cortical targets of transcranial magnetic stimulation (TMS; e.g., dACC, OFC, and dlPFC) or stereotactic ablation (e.g., the ACC) for treating OCD. The most prominent anatomical regions in R-maps for ALIC and STN cohorts, as well as agreement and inverted maps are summarized in supplementary **Table S1**.



**Figure 6. “Inversion” of the agreement map to estimate potentially optimal stimulation target regions. (A)** The agreement map (regions a connection to which was associated with clinical improvements in both anterior limb of the internal capsule, ALIC, and subthalamic nucleus, STN, zone DBS cohorts, expressed as multiplied R-values across the two maps, jR) was restricted to cortical regions to avoid that subcortical connectivity regions would bias subsequent analyses. Based on a preprocessed version of the N = 1,000 functional connectome, each brain voxel’s connectivity profile was spatially compared to the cortical agreement map (by means of spatial correlation, sR). This inverted map identifies which brain regions are maximally connected to the

agreement map and showed strong similarity with automated meta-analyses for the term “obsessive compulsive” (red-yellow colormap;  $R = 0.30$ ,  $p < 1e-17$ ; obtained from neurosynth.org). **(B)** The inverted map – as a potential set of neuromodulation targets – is shown in 3D in synopsis with an ALIC-DBS electrode, the optimal structural connectivity target published on the same cohort of patients (12), as well as with remarks about transcranial magnetic stimulation (TMS) and lesioning targets that showed efficacy for treatment of obsessive-compulsive disorder, in the past. The parasagittal plane represents a slice from the Bigbrain dataset (116). dACC, dorsal anterior cingulate cortex; dlPFC, dorsolateral prefrontal cortex; MCC, medial cingulate cortex; OFC, orbitofrontal cortex; Prec., precuneus; SFG, superior frontal gyrus; TC, temporal cortex; vStriat., ventral striatum.

**Table 2: Peaks of inverted map vs. average DBS stimulation sites in ALIC vs. STN DBS cohorts.**

	Average Coordinate Active DBS contacts [mm]	Peak within structure on Inverted Map [mm]	Euclidean Distance [mm]
<b>ALIC</b>			
Left hemisphere	-11.02/8.91/-2.41	-10.81/5.78/-2.94	3.2
Right hemisphere	10.45/9.28/-2.54	12.03/8.93/-1.84	1.8
<b>Average coordinate</b>	<b>±10.73/9.20/-2.48</b>	<b>±11.46/7.36/-2.37</b>	<b>1.9</b>
<b>STN</b>			
Left hemisphere	-9.97/-11.84/-7.99	-9.41/-10.23/-7.37	1.8
Right hemisphere	10.10/-11.20/-7.64	9.40/-9.78/-6.41	2.0
<b>Average coordinate</b>	<b>±10.04/-11.52/-7.84</b>	<b>±9.40/-10.01/-6.91</b>	<b>1.9</b>

Abbreviations: **ALIC**, anterior limb of the internal capsule; **DBS**, deep brain stimulation; **STN**, subthalamic nucleus. Note that average coordinates were calculated using nonlinear flips established within Lead-DBS (due to use of an asymmetric, more realistic MNI template).

An additional – crucially circular – analysis was amended to test a suggested technique for potential future independent comparisons on out-of-sample data. Namely, we calculated weighted overlaps between each patient’s VTA and the inverted map. Results of this analysis significantly correlated with clinical improvements ( $R = 0.54$ ,  $p < 1e-6$ ).

Last, an automated meta-analytic activation pattern derived from 98 studies, detected via NeuroSynth ((50); [www.neurosynth.org](http://www.neurosynth.org); uniformity tests) for the term “*obsessive compulsive*”, showed striking similarity to the inverted map in terms of spatial correlation strength (**Figure 6**;  $R = 0.30$ ,  $p < 1e-17$ ).

## Discussion

Four major points can be drawn from this study: First, an overall optimal functional connectivity model yielded robust predictions of Y-BOCS outcomes across OCD cohorts both in a leave-one-cohort-out and a leave-one-patient-out fashion, on group-level. In other words, variance in clinical outcomes could be cross-predicted across the two DBS targets (ALIC and STN zones) and four DBS centers based on normative rs-fMRI profiles seeding from patient-specific, active DBS electrode contacts. Second, while target-specific optimal functional connectivity profiles (ALIC vs. STN) differed in several brain regions, a considerable set of areas was shared between targets. Third, when restricting our model to these regions (in form of an agreement map), the predictive utility for both cohorts increased. Hence, by pinpointing the optimal network model from two separate nodes (i.e., via different DBS targets), the resulting model might have become more robust and specific. Last, inverting the agreement map revealed potential targets for neuromodulation in OCD. Naturally, this included the two DBS targets of study (ALIC and STN), with ALIC also a known target for stereotactic ablation, but also common cortical targets for TMS (such as the dACC, OFC, and dlPFC) or stereotactic ablation (such as the dACC).

### Common circuitry of effective DBS in OCD

First and foremost, the potential across-target neuromodulation *sweet spots* identified here via agreement map inversion integrate with important aspects of the hypothesized etiology of obsessive-compulsive symptoms, involving aberrant frontosubcortical limbic projections (23,24,27,29). The STN receives direct cortical input from most if not all regions of the PFC: After originating from specific prefrontal structures, its associative-limbic domains traverse parts of the ALIC and send axon collaterals to the

amSTN and other midbrain structures (51,52). Among these hyperdirect input areas, the dACC plays a prominent role for OCD pathophysiology (31,51,53).

Crucially, this hyperdirect trajectory largely matched the structural connectivity profile associated with optimal ALIC-DBS (7,12,54,55) and STN-DBS (12) outcomes for OCD. Extending these findings, the inverted pattern in the present study revealed optimal connectivity of agreement profiles between DBS electrodes – irrespective of target choice – with the ACC. These convergent results may hint at a fundamental, overarching network substrate of successful DBS in OCD. However, direct input from dACC to amSTN may be one link (or *edge*) in a network (or *circuitopathy*) consisting of additional brain regions (or *nodes*).

Functionally, the dACC consistently links to OCD symptom severity across resting-state and symptom-provocation studies, with reductions in excessive dACC activation following successful interventions (56–64). According to the *control signal theory of dACC dysfunction* (31), the dACC integrates input from other prefrontal structures such as the OFC, ventromedial PFC (vmPFC), and dlPFC involved in executive planning in response to unpleasant stimuli and/or *conflict* (compare with regions identified here; **Figure 6**). The dACC then emits a control signal to effectors such as the pre-supplementary motor area (pre-SMA) and SMA. Unlike in healthy individuals, in OCD, conflict resolution does not lead to normalized dACC activity and concurrent control signal extinction, potentially due to ongoing OFC or striatal activation. Instead, a vicious circle of repetitive behaviors (compulsions), accompanied by emotional distress (obsessions) follows. In line with this, recalibrating OCD-related dACC hyperactivity may thus provide a functional avenue for interrupting the maladaptive control signals chain reaction, restoring cognitive control and goal-directed behavior.

Indeed, OCD-DBS likely mirrors effects comparable to a long-standing tradition of neurosurgical approaches, which disrupt frontostriatal and/or frontosubthalamic connectivity to rebalance OCD-related dACC hyperactivity. While disconnecting white-matter bundles via bilateral *capsulotomy* led to successful OCD symptom alleviation as early as in the 1950s (65) – with new traction gained via gamma-knife surgeries or MR-guided focused ultrasound (66) – the long-term therapeutic efficacy of surgical dACC lesioning (*cingulotomy*) (67,68) has since been extensively confirmed. Similarly, efficacy of DBS may rely on normalizing frontostriatal connectivity in treatment-refractory OCD patients (37). Moreover, efficacious DBS sites targeted to the amSTN connected mainly to the dACC – alongside the lateral OFC and dlPFC (69). Li et al. expanded on this finding in elucidating the association of these connections with clinical improvements (12). Beyond that, positron emission tomography studies similarly indicated decreased prefrontal metabolism in OCD following both STN-DBS (70) and ALIC-DBS (71). Finally, the dACC (and dlPFC) count among the few noninvasive neuromodulation targets for OCD approved by the US Food and Drug Administration (72,73).

Beyond the classical CSTC pathway, the inverted across-target circuitry of effective DBS complements emerging evidence on more widely distributed dysfunctional brain systems (e.g., limbic, salience, or default mode networks) contributing to OCD pathogenesis (21,26,74). For instance, not only does the insula functionally connect to the CSTC circuitry (75,76), but compared to healthy controls, OCD patients also consistently display altered, task-related insular activation (77–79). Accordingly, a recent study showed that DBS to the ALIC increased the impact of vmPFC onto the amygdala, decreasing amygdalar impact onto the insula (80). Switching DBS off reversed this pattern drastically, along with sudden depressive or anxious exacerbations. Still, it remains to be established whether such nodes could

become involved in effective therapy of obsessions or compulsions per se rather than in that of OCD comorbidities (81).

Apart from optimal *post*-surgical DBS electrode connectivity, other amounts of clinical outcome variance may be explained by patient selection – since not all patients show uniform symptom profiles and may thus differentially respond to treatment. This raises a need for preoperative baseline predictors of response to DBS in OCD (2,82,83). Here, the most impaired patients (London cohort), on group level, appeared to benefit most from the intervention. However, statistical tests of a potential interaction may be biased in this particular cohort due to the retrospective study character, small sample size (N = 6) and patient-wise implantation of four electrodes, increasing stochastic likelihood of hitting a specific beneficial network. So far, no study has identified OCD symptom severity as a reliable candidate marker (2,83).

### **Target-specific circuitry**

Beyond an *agreement* set of optimal connections across the two DBS targets, our *target-specific* models pointed out that, in addition, each may modulate separate networks. Such target-specific therapeutic networks may potentially be harnessed for tailoring treatment strategies to different disease subtypes or symptom domains (44,84,85). In other words, while both ALIC- and STN-targets may effectively reduce obsessive-compulsive symptoms registered by the Y-BOCS, each may complementarily lead to symptom changes across other domains – conceivably as a result of co-modulating target-specific networks.

Initial evidence indeed points towards behaviorally selective, target-specific DBS effectiveness sustained by differential OCD networks. In a prospective clinical study, ALIC-DBS preferentially improved depressive symptoms (potentially mediated via connectivity to the medial OFC), whereas amSTN-DBS selectively ameliorated



cognitive flexibility (potentially mediated via connectivity to the lateral OFC, dACC, and dlPFC) (69). Similarly, immediate improvements in mood and anxiety followed NAcc-DBS even in non-responders (86), while specific effects of STN-DBS on Y-BOCS outcome (but not on depression, anxiety, or other neuropsychological symptoms) were observed (36). Pertaining to specific OCD subtypes, refractoriness to NAcc-DBS was shown for perfectionism, hoarding, or symmetry (86). Still, since psychiatric DBS studies often rely on small samples, follow-up investigations should confirm the concept of symptom-specific substrates in OCD – and the one of shared network topologies.

The proposed unifying theory could henceforth – after sufficient prospective validation and replication via patient-specific connectomes – contribute to advancements in DBS programming, inform noninvasive stimulation targets (87), or add to a harmonized understanding of effective neuromodulation for OCD in a broader sense (88,89).

## **Limitations**

There are several limitations to this study. First, the retrospective design warrants careful interpretation and prospective validation of interventional outcomes. Nevertheless, pooling a large sample allowed testing for robustness of effects across DBS centers and surgeons.

Second, non-linearly warping electrode sites into template space inevitably introduces inaccuracies. We sought to maximally counteract this bias by use of a state-of-the-art neuroimaging pipeline that includes brain-shift correction and subcortical refinement (47). Furthermore, we applied multispectral normalization, with an accuracy close to manual expert delineation as demonstrated by two independent groups (90,91), as well as electrophysiologically confirmed (92–98), phantom-validated

electrode localization methods (99). Crucially, results of each processing step underwent meticulous inspection and manual refinement by an expert user.

Third, the finite element VTA model applied in our analyses as well as the very *concept* of binarized E-field thresholding in general, are based on approximations that cannot truthfully incorporate every aspect of interactions between electrical current and neural tissue (100,101). While the approach applied here may represent a valid – albeit simplified – approximation (102), future studies may benefit from exploring more advanced concepts (101,103,104).

Fourth, estimates of therapeutic connectivity seeding from DBS electrodes in individual patients were based on normative connectomes. This concept has been widely applied in DBS, compared with individualized connectivity in Parkinson's Disease (105) or OCD (7), and may increase signal-to-noise ratios (85,106,107). However, its limited ability of accounting for patient-specific differences renders our study into a “broad lens view” on the brain, with a need for replication via individualized rs-fMRI data (108). Nevertheless, normative connectome-based models may represent first-order approximations of DBS network effects (6,9,11,12,109–111). Moreover, first proposed models have been prospectively validated by groups world-wide, underscoring reproducibility and clinical utility (54,105) (55,112). Importantly, despite group-level significance, results should not guide surgical targeting or DBS programming without thorough replication and refinement.

Last, although widely accepted as an interventional assessment in OCD and thus facilitating across-study comparisons (113), the explanatory value accessible via the Y-BOCS summary score is naturally limited to the effect of DBS onto global obsessive-compulsive symptomatology. While unfortunately, more comprehensive

neuropsychological assessments were unavailable, investigating postoperative network changes along different symptom axes represents a valuable future avenue.

## **Conclusions**

In summary, we identify functional connectivity correlates of effective OCD-DBS in form of a distributed brain network, highlighting *differences* but also *commonalities* between ALIC and STN target zones. The distinctive *target-specific* connectivity features may point toward behaviorally selective therapeutic circuitries. Yet, fundamental across-target overlap allowed for significance in leave-one-cohort-out and leave-one-patient-out designs. Furthermore, a set of *agreement regions* outperformed a full-cohort model in terms of predictive utility for clinical improvements in both targets. Finally, we elucidate brain-wide regions exhibiting the most optimal connectivity profile as potential candidate targets for successful neuromodulation in OCD.

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## **Disclosures**

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## ***Supplemental Information***

### **Patient cohort**

Data of 50 obsessive-compulsive disorder (OCD) patients from four centers were included into our analyses, comprising 22 anterior limb of the internal capsule (ALIC) deep brain stimulation (DBS) patients from University Hospital of Cologne, 14 subthalamic nucleus (STN) DBS patients from Grenoble University Hospital, eight patients from Hospital Clínico San Carlos in Madrid with bilateral electrodes targeting the nucleus accumbens (NAcc), and six patients from the National Hospital for Neurology and Neurosurgery in London with electrodes implanted to both STN and ALIC. Patients operated in Cologne, Grenoble and Madrid centers had received two DBS leads each (N = 44 patients with N = 88 leads), while the six patients from the London cohort were implanted with four DBS leads each (N = 6 patients with N = 24 leads). All patients gave written informed consent, and study protocols received ethical clearance from each local Ethics Committee.

### **Imaging**

For all patients, high-resolution structural T1-weighted images were acquired on a 3.0-Tesla MRI scanner before surgery. Postoperatively, computed tomography (CT) was obtained in 33 patients to evaluate lead placement, while eleven patients from the Grenoble cohort and the six London patients received postoperative MRI. Across the four centers, pre- and postoperative OCD symptom severity was assessed on the Yale-Brown Obsessive-Compulsive Scale (Y-BOCS) (1). In the London cohort, multiple stimulation settings with combined “optimized” settings of both targets (and respective Y-BOCS improvement scores) vs. the ones where each target was probed alone (STN

vs. ALIC zones) were available (2). When calculating maps for patients stimulated in each target alone, the latter scores and stimulation settings were used. Instead, when calculating overall models across all patients, the “optimized” settings were used.

## **Electrode localizations and estimation of stimulation effects**

DBS electrodes were localized using default parameters in Lead-DBS software (<http://www.lead-dbs.org>) and details are reported elsewhere (3,4). Briefly, postoperative CT and MRI scans were linearly coregistered to preoperative T1 images using Advanced Normalization Tools (ANTs, <http://stnava.github.io/ANTs/>; (5)). This registration was further corrected for brain shift as implemented in Lead-DBS.

Electrodes were then pre-localized using either the PaCER algorithm (6) for postoperative CT, or the TRAC/CORE algorithm (7) for postoperative MRI, and automatic pre-localizations were manually refined by an expert user (NL). Volumes of tissue activated (VTAs) were estimated based on patients’ individual stimulation parameters using a finite element method (FEM) (4). A volume conductor model was constructed via a four-compartment mesh that includes grey and white matter, lead contacts, and insulating parts. The electric field (E-field) distribution was simulated using an adaptation of the FieldTrip-SimBio pipeline integrated into Lead-DBS (<https://www.mrt.uni-jena.de/simbio/>; <http://fieldtriptoolbox.org/>), and thresholded at 0.2 V/mm to obtain binarized VTAs (4).

Based on preoperative acquisitions, volumes were then normalized into ICBM 2009b Nonlinear Asymmetric (“MNI”) template space using the SyN approach implemented in ANTs, with an additional subcortical refinement step (“Effective: Low Variance + Subcortical Refinement” preset in Lead-DBS). This specific method was top performer for subcortical image registrations in a recent comparative study that involved >10,000 nonlinear warps and a variety of normalization techniques (8), as

confirmed by a second comparative study from a different group (9). Coregistration and normalization results were reviewed and refined if needed. Subsequently, DBS electrodes were reconstructed via Lead-DBS software and warped into MNI space.

## **DBS connectivity modeling**

This section is dedicated to a more detailed description of our approach to modeling functional connectivity R-maps. Specifically, functional connectivity analyses were based on a normative functional connectome, calculated from resting-state functional MRI (rs-fMRI) scans of 1,000 healthy subjects (10) that had been acquired within the Brain Genomics Superstruct Project (<https://dataverse.harvard.edu/dataverse/GSP>; (11)).

For each of the 50 OCD patients, normative whole-brain functional connectivity seeding from bilateral VTAs was estimated similarly as done in previous work (12,13): First, voxel-wise correlations between mean time-series of voxels inside each bilateral VTA and time-series of all remaining whole-brain voxels were calculated. This analysis was repeated 1,000 times per patient, using data of each of the 1,000 healthy subjects in the normative connectome. Second, the resulting 1,000 connectivity *fingerprints* for each patient were averaged and Fisher-z-transformed. Given individual differences in electrode position, stimulation parameter settings, and consequently, divergent VTA seeds, this approach resulted in 50 average connectivity fingerprints (i.e., one for each OCD patient).

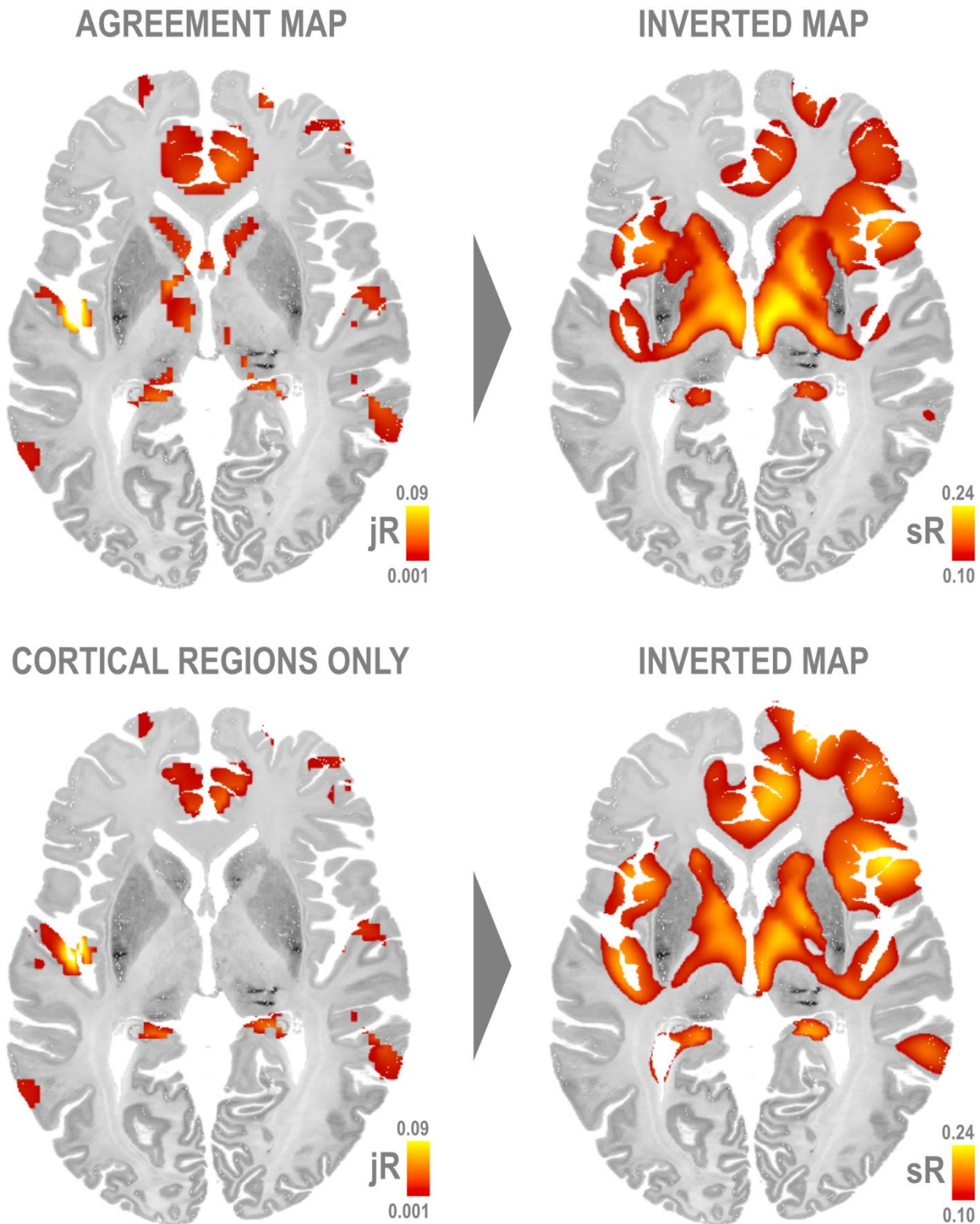
On a voxel-by-voxel basis, individual fingerprints were then correlated with Y-BOCS improvements across patients to generate an *R-map model* of “optimal” connectivity profiles for maximal Y-BOCS improvement. This R-map models regions to which DBS electrodes optimally should or should not be connected to maximize Y-BOCS improvement. Rs-fMRI derived functional connectivity being bipolar (i.e.,

anticorrelations across time series existing (14)), blue regions can also mean that the BOLD signal obtained from the electrode seeds should be negatively correlated with the one recorded in these regions for optimal Y-BOCS improvement.

### **Inverted map calculation**

As described in the methods section, the inverted map highlights brain regions which have cortical connectivity profiles most similar to the optimal ones defined by the agreement map. Specifically, similarity between each brain voxel's connectivity profile (based on a precomputed resting-state functional MRI connectome) and the (cortical) agreement map was calculated. Since the goal here was to define the optimal (subcortical) neuromodulation sites, we chose to restrict the agreement map to cortical regions when calculating the inverted map based on it in order to avoid circularity. Had subcortical regions been included in the agreement map, the generated inverted map could have been biased, since the subcortical voxels will have maximum connectivity ( $R=1$ ) to themselves. However, again, this step did not alter results. To demonstrate this, in **Figure S1**, we show that the result stays largely unchanged when we do not exclude the subcortical regions before calculating the inverted map. These two versions of the inverted maps (based on the full agreement map and cortical only agreement maps) were highly similar (spatial correlation  $R=0.96$ ,  $p<1e-17$ ).





**Figure S1.** Inverted maps based on the full agreement map and its cortical regions only. Top left panel shows the full agreement map calculated based on the R-maps from the ALIC and STN cohorts. Bottom left shows the same map restricted to cortical (and cerebellar) regions. Inverted maps shown in the right panels were calculated by testing each brain voxel's connectivity profile (based on a precomputed resting state functional MRI connectome) with its similarity to either version of the agreement map. The analysis shows that restricting the agreement map to cortical regions did not alter results. ALIC, anterior limb of the internal capsule; STN, subthalamic nucleus.

## **Prominent anatomical regions of maps**

In order to extract spatial information about the anatomical structures highlighted by the different statistical maps, we parcellated each target-specific R-map, agreement map and inverted map according to a Talairach atlas version transformed into MNI space (15). We selected this atlas because of its detailed description and the availability of Brodmann areas labeling. The atlas is freely available to download from the FMRIB Software Library (FSL). The peak value in addition to voxel coordinates (x/y/z in MNI space) were next extracted for each parcel using an in-house MATLAB code implemented in Lead-DBS. We reported highest 30-40% of parcels in each of the resulting statistical maps. For ALIC and STN targets, results were manually analyzed using 3Dslicer software ([www.slicer.org](http://www.slicer.org)).

**Table S1: Anatomical regions associated with optimal clinical improvement across the four different models. Regions identified with optimal connectivity profiles across the brain are listed, as well (Inverted Map column)**

Reg.	Hem.	ALIC R-map	STN R-map	Full Cohort R-map	Agreement Map	Inverted Map
<i>Positive Peak Coordinates X/Y/Z (R-Value)</i>						
ACC (BA 24, 32)	RH	5/4/39.5 (0.54)	18/12/48 (0.26)	6/6/39.5 (0.60)	9/41/0 (0.05)	8/32/22 (0.27)
	LH	-8/5/37 (0.53)	-16/40/18 (0.08)	-6.5/10/31 (0.53)	-3/37/10 (0.03)	-1/36/16 (0.24)
ALIC	RH		13/9.5/-2.5 (0.48)	16/4/5.5 (0.31)		12/9/-2 (0.20)
	LH		-13.5/7.5/-2.5 (0.49)	-16.5/7.5/2 (0.26)		-11/6/-3 (0.24)
CBM	RH	5/-56/-17 (0.45)	18/-50/-28 (0.16)	4/-71.5/-14.5 (0.58)	14/-44/-34 (0.09)	18/-48/-36 (0.10)
	LH	-2.5/-56/-16 (0.46)	-14/-58/-32 (0.17)	-4.5/-60/-16 (0.51)	-44/-46/-46 (0.06)	-16/-64/-20 (0.06)
Cun. (BA 17)	RH	18/-82/10 (0.18)	8.5/-102/9 (0.30)	14/-83/10 (0.49)		
	LH	-14/78/14 (0.21)	-8/-104/3 (0.48)	-11/-77.5/11 (0.42)		
dlPFC (BA 46, 9, 9/46)	RH	32.5/44.5/29 (0.35)		40/54.5/27 (0.42)		26/44/34 (0.26)
	LH	-34/42/34 (0.34)		-40/50/27 (0.45)		-26/38/30 (0.18)
IFG (BA 10, 46)	RH		24/24.5/-17 (0.31)			42/12/-18 (0.19)
	LH		-26/31/-14.5 (0.21)			-38/12/-16 (0.30)
Ins. (BA 13)	RH	44/4/-2 (0.6)	48/-2/-8 (0.32)	39.5/-25/6.5 (0.59)	47/-3.5/-3.5 (0.07)	36/14/-10 (0.34)
	LH	-40.5/-18.5/2 (0.53)	-44/4/-10 (0.35)	-46.5/-4/6 (0.61)	-47/-9.5/-1 (0.10)	-38/12/-10 (0.33)
IPL (BA 7, 39)	RH	65.5/-27/30.5 (0.5)		61/-35/47 (0.50)		60/-42/30 (0.26)
	LH	-67/-29/26 (0.5)		-52.2/-41/29 (0.58)		-64/-40/32 (0.15)
MCC (BA 23, 31)	RH	8.5/-15/34 (0.53)	8/-41/39.5 (0.29)	6/-22/41.5 (0.48)	12/-44/44 (0.13)	18/-56/32 (0.26)
	LH	-8/-29/39.5 (0.53)	-1.5/-33.5/40 (0.20)	-11/-35.5/40 (0.61)	-12/-42/42 (0.17)	-10/-72/34 (0.23)
MD	RH	8/-20/2 (0.41)	6/-10/10 (0.08)	6/-20/2 (0.55)		4/-18/4 (0.2)
	LH	-6/-20/2 (0.38)	-6/-16/2 (0.09)	-6/-20/2 (0.54)		-4/-10/0 (0.2)
Prec. (BA 7, 31)	RH	14/-66/39.5 (0.46)	8/-50/44 (0.34)	14/-64/42 (0.58)	14/-66/32 (0.13)	18/-56/32 (0.26)
	LH	-14/-63/38.5 (0.50)	-1/-57/47.5 (0.35)	-12/-44/44 (0.60)	-16/-68/30 (0.09)	-10/-72/34 (0.23)
SFG (BA 9, 11)	RH		22/50/-26 (0.35)			22/42/32 (0.26)
	LH		-8/56/-34 (0.31)			-26/38/30 (0.18)
SMA (BA 6)	RH	5/-3/66.5 (0.47)		3.5/-1.5/66 (0.56)		
	LH	-9.5/-1.5/70 (0.43)		-6.5/-4/70 (0.54)		
STN	RH	9/-14/-7 (0.46)	10/-12/-8 (0.31)	9.5/-14.5/-7 (0.51)	10/-10/-8 (0.16)	9/-10/-6 (0.24)

	LH	-9.5/-17.5/-8 (0.43)	-6/-12/-6 (0.34)	-7/-16/-7 (0.50)	-6/-12/-8 (0.15)	-9/-10/-7 (0.24)
Uncus (BA	RH	22/-10/-36 (0.52)	16/0/-28 (0.32)	20/-2/-40 (0.51)	22/-05/-36 (0.08)	22/0/-36 (0.32)
28)	LH	-17/-4/-38.5 (0.43)	-14/2/-26 (0.25)	-14/0/-40 (0.55)	-21/-6/-35 (0.07)	-20/-4/-36 (0.25)
<b>Negative Peak Coordinates X/Y/Z (R-Value)</b>						
	RH	21.5/-87/-39 (- 0.53)	13/-49/-17 (-0.34)	32/-90/-40 (-0.62)	32/-83/-43 (-0.07)	16/-78/-44 (- 0.24)
CBM	LH	-30/-92/-22 (-0.45)	-1/-49/-17 (-0.35)	-22/-90/-38 (-0.62)	-34/87/-43 (-0.02)	-20/-90/-38 (- 0.08)
IFG (BA 10,	RH	49/25/4.5 (-0.25)	48/26/4 (-0.21)	47/27/-19 (-0.37)	62/24/20 (-0.01)	64/12/32 (-0.10)
46)	LH	-48/18.5/10 (-0.45)	-54/48/-6 (-0.28)	-39.5/23/-18 (- 0.44)	-58/20/20 (-0.04)	-60/18/30 (-0.16)
IPL (BA 7,	RH	35/-33.5/56 (-0.36)		53/-63/41.5 (-0.38)		
39)	LH	-43/-35/56 (-0.37)		-44/-67/45 (-0.42)		
MTG (BA	RH	46/-7/30 (-0.59)	38/-64/22 (-0.16)	66/0/-20.5 (-0.58)	56/-44/-16 (-0.01)	56/-44/-16 (- 0.10)
19, 39)	LH	-64/-14.5/-17 (- 0.50)	-36/-58/18 (-0.40)	-65.5/-9/-14 (-0.57)	-56/-43/-14.5 (- 0.04)	-52/-42/-16 (- 0.20)
OL (BA 18,	RH			26/-95/-4.5 (-0.52)	35.5/-85/-5 (-0.04)	
19)	LH			-29/-96/-1 (-0.52)	-39/-80/-3.5 (-0.04)	
PCL (BA 3,	RH	5/26.5/63 (-0.49)	2/-25.5/71 (-0.15)	3.5/-36/66.5 (- 0.60)	6.5/-24/67.5 (- 0.14)	12/-22/72 (-0.18)
4)	LH	-4.5/-35.5/64 (- 0.46)	-1/-31/73.5 (-0.22)	-3/-30/63 (-0.59)	-3.5/-30/67.5 (- 0.11)	-10/-26/72 (- 0.19)
Prec. (BA	RH	4/-61.5/41 (-0.50)		1/-56/34 (-0.48)		
7, 31)	LH	-5/-57.5/39.5 (- 0.45)		-1/-62.5/34 (-0.44)		
SFG (BA 9,	RH	7.5/65/22 (-0.41)	2/15.5/64 (-0.40)	11/65.5/13.5 (- 0.52)	14/34/42 (-0.01)	-10/60/34 (-0.11)
11)	LH	-12/66/22 (-0.53)	-4/7/64.5 (-0.44)	-14.5/66/20 (-0.57)	-8/52/44 (-0.05)	-10/52/44 (-0.12)

Abbreviations: **ACC**, anterior cingulate cortex; **ALIC**, anterior limb of the internal capsule; **BA**, Brodmann area; **CBM**, cerebellum; **Cun.**, cuneus; **dIPFC**, dorsolateral prefrontal cortex; **Hem.**, hemisphere; **IFG**, inferior frontal gyrus; **Ins.**, insula; **IPL**, inferior parietal lobule; **LH**, left hemisphere; **MCC**, midcingulate cortex; **MD**, medial dorsal nucleus of thalamus; **MTG**, middle temporal gyrus; **OL**, occipital lobule; **PCL**, paracentral lobule; **Prec.**, precuneus; **Reg.**, region; **RH**, right hemisphere; **SFG**, superior frontal gyrus; **SMA**, supplementary motor area; **STN**, subthalamic nucleus.

## Supplemental References

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