1	Title: A human brain network linked to restoration of consciousness after deep brain			
2	stimulation			
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ABSTRACT

Disorders of consciousness (DoC) are states of impaired arousal or awareness. Deep brain stimulation (DBS) is a potential treatment, but outcomes vary, possibly due to differences in patient characteristics, electrode placement, or stimulation of specific brain networks. We studied 40 patients with DoC who underwent DBS targeting the thalamic centromedian-parafascicular complex. Better-preserved gray matter, especially in the striatum, correlated with consciousness improvement. Stimulation was most effective when electric fields extended into parafascicular and subparafascicular nuclei-ventral to the centromedian nucleus, near the midbrain-and when it engaged projection pathways of the ascending arousal network, including the hypothalamus, brainstem, and frontal lobe. Moreover, effective DBS sites were connected to networks similar to those underlying impaired consciousness due to generalized absence seizures and acquired lesions. These findings support the therapeutic potential of DBS for DoC, emphasizing the importance of precise targeting and revealing a broader link between effective DoC treatment and mechanisms underlying other conscciousness-impairing conditions.

78 INTRODUCTION

79

80 There are no proven treatments for patients with chronic disorders of consciousness (DoC).^{1, 2} DoC are caused by brain injuries including hypoxia, 81 82 ischemia, trauma, and intracerebral hemorrhage, resulting in impairments of arousal or awareness that vary widely in severity and prognosis.^{3, 4} These patients 83 84 commonly reside in long-term care facilities with no or severely limited ability to engage with their environments-for months, years, even decades. 85 86 87 Neuromodulation has been explored as a potential therapy to restore consciousness in patients with DoC for over 50 years. Pioneering work by 88 89 Hassler⁵ and McLardy⁶ in the 1960s was followed by larger case series of deep brain stimulation (DBS) in the 1990s^{7, 8} and beyond.⁹⁻¹⁹ Various stimulation targets 90 91 have shown some efficacy in uncontrolled trials, including the intralaminar thalamus,^{7-9, 12-14, 16, 17, 19} brainstem,^{8, 17} pallidum,^{5, 11} and nucleus accumbens.¹⁵ 92 93 However, evidence from randomized controlled trials is lacking. 94 Recent advances in understanding the brain networks underlying DoC²⁰ have 95 opened new avenues for diagnosis,^{1, 2} treatment,²¹ and outcome prediction.⁴ 96 97 Regardless of the cause, DoC involve a widespread suppression of excitatory neurotransmission.³ particularly in the "mesocircuit"²⁰ of the anterior forebrain, 98 99 which includes the frontal cortex, central thalamus, striatum, and brainstem. 100 101 Within the thalamus, the posterior intralaminar nuclei (centromedian [CM] and parafascicular [Pf] nuclei) are central components of the mesocircuit²⁰ and project 102 directly to the striatum.^{22, 23} Based on these connections, the CM-Pf complex was 103 104 targeted in a recent uncontrolled study of DBS for DoC,^{9, 12} the largest of its kind, motivated by findings in previous, smaller studies.^{7, 8, 17, 18} While no consistent 105 106 effects were seen at the group level, this study identified a subset of patients with 107 dramatic improvements in consciousness in the first year following implantationbeyond what has been seen in natural history studies of chronic DoC.^{24, 25} 108 109

- Here, we analyzed patient-level data to test whether these DBS "responders"
- 111 exhibited specific clinical characteristics, including MRI measures of brain tissue

- 112 integrity, that predicted treatment responsiveness. We also investigated whether
- 113 they were stimulated in a specific thalamic subregion, white matter tract, or
- 114 distributed functional brain network. Finally, we examined the external validity of
- 115 this network by testing its involvement in separate groups of patients with lesion-
- 116 or epileptic seizure-induced impairments of consciousness.
- 117

118 **RESULTS**

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120 Patient characteristics and clinical outcomes

121

122 The analysis cohort included 40 patients with DoC secondary to cardiac arrest or

- 123 traumatic brain injury who met criteria for DBS based on previously described
- 124 clinical, neurophysiologic, and neuroimaging evaluations.^{9, 12} Patients were
- assessed using the Coma Recovery Scale-Revised (CRS-R)²⁶ and classified as
- 126 having unresponsive wakefulness syndrome (UWS), a minimally conscious state
- 127 (MCS), or full consciousness.¹ Supplementary clinical measures included the
- 128 Disability Rating Scale (DRS) and Coma/Near-Coma (C/NC) scale.²⁷
- 129

130 After a median of 6 months post-injury (IQR=3.5-13, range=2-137), patients

- underwent unilateral DBS targeting the left (n=37) or right (n=3) CM-Pf (**Fig. 1**). At
- 132 12 months post-DBS, 11/40 patients were classified as improved and 29 as non-
- 133 improved. "Improved" patients were defined as those who transitioned from UWS
- 134 to MCS or fully conscious, or from MCS to fully conscious, as in previous work.^{9, 12}
- 135 Across all patients, the median increase in CRS-R scores was 2 (IQR=2-7,
- range=0-18; **Fig. 1**). Three patients died 3-6 months post-surgery. Secondary
- 137 analyses were performed to test consistency of results when using a different
- 138 outcome definition (each patient's change in CRS-R scores, as opposed to a
- 139 binary improved/non-improved classification), excluding patients with right-sided
- 140 DBS implants, and excluding patients who died (see **Supplementary Material**).
- 141
- 142 We first tested the hypothesis that clinical variables are associated with
- 143 improvement (**Table 1**). Compared to non-improved patients, those who improved
- had less severe baseline impairments on the DRS and C/NC scales (p<0.05, false
- 145 discovery rate [FDR]-corrected) and, at an uncorrected threshold, were a median

- of 20 years younger at the time of injury. In contrast, the groups did not differ by
- sex, side of DBS, etiology of DoC, duration between initial injury and DBS, DBS
- 148 lead model, or the programmed stimulation amplitude, frequency, or pulse width.
- 149



150

151 Figure 1: Clinical outcomes and DBS electrode localizations for the improved and

152 **non-improved patient groups. (A)** Raincloud plots²⁸ showing DoC severity measured

- 153 before and 12 months after DBS using the Coma Recovery Scale-Revised (CRS-R),
- 154 where higher scores correspond to better outcomes. Patients were categorized into
- improved (n=11) and non-improved (n=29) groups. **(B)** Three-dimensional DBS
- 156 electrode localizations. All patients underwent unilateral implantation. Leads are shown

with respect to the centromedian (CM) and parafascicular (Pf) nuclei, as defined by the 157 atlas of Krauth et al.²⁹ based on the histology work of Morel.³⁰ Leads are shown only for 158 patients in whom accurate image registrations were possible (n=10 improved and n=18159 160 non-improved). (C) Two-dimensional coronal views of the thalamus showing the center 161 of each patient's volume of tissue activated (i.e., the modelled electric field around DBS contact(s) chosen for stimulation). Pink and blue circles indicate locations for patients in 162 163 the improved and non-improved groups, respectively. Y coordinates indicate the coronal position (in mm) of each slice, in MNI 152 ICBM 2009b nonlinear asymmetric template 164 165 space. Stimulation coordinates for each patient are reported in **Supplementary Material**. Locations are displayed upon the BigBrain histological atlas³¹ registered to 166 MNI space.³². Abbreviations: A, Anterior, CL, Central lateral nucleus, CM, Centromedian 167 nucleus, CRS-R, Coma Recovery Scale-Revised, I, Inferior, L, Left, Lat., Lateral, LD, 168 Lateral dorsal nucleus, Md, Mediodorsal nucleus, Pf, Parafascicular nucleus, R, Right, 169 170 RN, Red nucleus, S, Superior, sPf, Subparafascicular nucleus, STh, Subthalamic 171 nucleus.

- 172 173

Optimal brain tissue integrity 174

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176 We next investigated whether DBS outcomes were associated with MRI measures

- of brain tissue integrity (Fig. 2) in patients with T1-weighted MRI available (n=8 177
- 178 improved patients versus n=18 non-improved). MRI scans were segmented into
- 179 whole-brain (gray matter, white matter, cerebrospinal fluid) and regional
- subcortical volumes,³³ then normalized by total intracranial volume and by age-180
- matched controls.³⁴ At the uncorrected significance threshold, patients who 181
- improved had greater preservation of whole-brain gray matter and larger volumes 182
- 183 of cerebellar gray matter and the dominant striatal projections of the CM-Pf.
- 184 including the putamen and caudate. These comparisons did not survive FDR
- 185 correction. The groups did not differ in thalamic, pallidal, or brainstem volumes.



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188 Figure 2: Comparison of MRI tissue volumes between improved and non-

189 **improved groups.** Violin plots showing whole-brain volumes of gray matter (GM), 190 white matter (WM), and cerebrospinal fluid (CSF) in the improved (n=8) and non-191 improved (n=18) groups. This analysis only included patients with T1-weighted MRI 192 available. The plot on the right shows comparisons of subcortical gray matter 193 volumes. The dashed horizonal line on each plot indicates the average value in age-194 matched controls from the Nathan Kline Institute-Rockland Sample (NKI-RS).³⁴ 195 Volumes from patients are expressed as z-scores measuring the distance, in units of 196 SD, away from the control mean. * = p < 0.05 (uncorrected).

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198 **Optimal stimulation site**

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Electrodes were localized using Lead-DBS software.³⁵ This revealed variability in 200 201 electrode placement across patients (Fig. 1), suggesting outcomes may be linked 202 to stimulation of specific thalamic subregions. To test this, we calculated electric 203 fields (E-fields) for each patient, which estimate the distribution and magnitude of stimulation based on the DBS settings.^{35, 36} Then, to identify sites linked to 204 205 therapeutic benefit (Fig. 3), we initially conducted a series of voxel-wise twosample *t*-tests to pinpoint potential sites of optimal stimulation ("sweet spots"). 206 207 This analysis compared E-field magnitudes between improved (n=10) and non-208 improved (n=18) groups. The *t*-tests identified a candidate sweet spot defined by 209 the largest surviving voxel cluster at a significance threshold of p < 0.05. This site 210 was in the inferior Pf, aligned with the axial anterior commissure-posterior 211 commissure (AC-PC) line, and extended into the subparafascicular nucleus 212 below, with MNI 152 ICBM 2009b coordinates (mm) of [X=-6.91, Y=-20.11, Z=-213 3.08]. Notably, there were no regions where the E-field magnitude was higher in 214 the non-improved group. To test the reliability of this candidate site, we subjected 215 the results to *k*-fold cross-validation. This confirmed the robustness of the findings 216 (t=2.05, p=0.05), validating the candidate sweet spot's association with benefit.

218 Figure 3: Anatomical localization and cross-validation of optimal stimulation

219 site. (A) K-fold (k=10) cross-validation showing that E-field peak locations are 220 associated with clinical outcomes in left-out patients (p=0.05). (B) Three-dimensional 221 views displaying the location of the stimulation "sweet spot", defined as the center-of-222 gravity of the largest cluster (p<0.05, uncorrected) following voxel-wise two-sample t-223 tests of E-field magnitudes between improved (n=10) and non-improved (n=18)224 groups. The sweet spot is shown in sagittal and coronal orientations with respect to the centromedian (CM), parafascicular (Pf), and subparafascicular (sPf) thalamic 225 226 nuclei, as defined by the atlas of Krauth et al.²⁹ based on the histology work of Morel.³⁰ (C) Two-dimensional views of the thalamus showing an unthresholded map 227 of *t*-scores where positive values indicate locations where E-field magnitudes were 228 229 higher in the improved relative to non-improved group, and negative values where 230 they were lower. The X, Y, and Z values indicate sagittal, coronal, and axial positions 231 (mm) in MNI 152 ICBM 2009b nonlinear asymmetric template space, respectively. (D)

Thresholded map (*p*<0.05, uncorrected) showing the peak site associated with
therapeutic benefit. Results are displayed upon the BigBrain histological atlas³¹
registered to MNI space.³² <u>Abbreviations</u>: A, Anterior, CL, Central lateral nucleus, I,
Inferior, L, Left, Lat., Lateral, Md, Mediodorsal nucleus, Med., Medial, P, Posterior,
PC, Posterior commissure, Pulv, Pulvinar nucleus, R, Right, RN, Red nucleus, S,
Superior.

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239 Optimal structural connectivity

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241 Having identified optimal DBS sites, we next hypothesized that DBS improvement 242 might be mediated by structural connections traversing areas of beneficial 243 stimulation.³⁵ To test this, we calculated white matter streamlines using a 244 normative diffusion MRI connectome acquired at an ultra-high resolution of 760 µm,³⁷ as in our recent work.^{38, 39} Like the "sweet spot" analysis, which analyzed 245 246 stimulations across voxels, here we compared stimulation magnitudes across 247 streamlines between improved (n=10) and non-improved (n=18) groups. This 248 involved conducting similar two-sample *t*-tests to pinpoint potential streamlines 249 associated with the rapeutic benefit, at an initial threshold of p<0.05. Improved 250 patients had stronger involvement of connections to distributed projection pathways of the ascending arousal network,⁴⁰ including the hypothalamus 251 (posterior, periventricular, and paraventricular hypothalamic nuclei), midbrain (red 252 253 nucleus, mesencephalic reticular formation, and ventral tegmental area), pons 254 (locus coeruleus, subcoeruleus, and medial parabrachial nucleus), medulla 255 (inferior and superior medullary reticular formation), cerebellar dentate nucleus, 256 and medial frontal cortex (Fig. 4). There were no connections with stronger 257 involvement in non-improved patients. The findings were robust to k-fold crossvalidation testing (t=2.29, p=0.03), confirming their association with benefit. 258 259 260 261 262

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Figure 4: Anatomical localization and cross-validation of optimal structural 267

268 **connectivity.** White matter fiber tracts more strongly involved in the improved (*n*=10)

269 relative to non-improved (n=18) group (p<0.05, uncorrected). Box plots in the upper left

corner show the results of K-fold (k=10) cross-validation, demonstrating that fiber tracts 270

are associated with clinical outcome in left-out patients (p=0.03). The most strongly 271

- connected subcortical nuclei are displayed using published atlases of the 272
- hypothalamus⁴¹ (https://zenodo.org/records/3942115), brainstem 273
- (https://www.nitrc.org/projects/brainstemnavig), ascending arousal network⁴⁰ 274
- (https://doi.org/10.5061/dryad.zw3r228d2), and cerebellum⁴² 275
- (https://www.diedrichsenlab.org/imaging/propatlas.htm). Results are displayed upon the 276
- BigBrain histological atlas³¹ registered to MNI space.³² Abbreviations: A, Anterior, AHA, 277

278 Anterior hypothalamic area, AN, Arcuate nucleus, DR, Dorsal raphe nucleus, iMRt,

279 Inferior medullary reticular formation, Lat., Lateral, LC, Locus coeruleus, MPB, Medial 280 parabrachial nucleus, mRt, Mesencephalic reticular formation, PA, Paraventricular

281 nucleus, PE, Periventricular nucleus, PH, Posterior hypothalamus, RN, Red nucleus, S,

282 Superior, SCh, Suprachiasmatic nucleus, sMRt, Superior medullary reticular formation,

- 283 SubC, Subcoeruleus, Ve, Vestibular nuclei complex, VSM, Viscero-sensory-motor nuclei
- 284 complex, VTA PBP, Ventral tegmental area (parabrachial pigmented nucleus complex).
- 285

Optimal functional connectivity 286

- 287
- 288 In the next analysis, we investigated blood-oxygen-level-dependent (BOLD)
- connectivity of DBS sites using a normative resting-state fMRI dataset acquired in 289
- 1,000 healthy adults.^{43, 44} This involved overlaying each patient's E-fields on the 290

291 normative fMRI data and calculating connectivity with all brain voxels. Connectivity 292 strengths were then compared between improved (n=10) and non-improved 293 (n=18) groups using voxel-wise two-sample *t*-tests, resulting in a map where 294 positive *t*-scores indicate higher connectivity in the improved group (**Fig. 5A**). 295 296 The primary purpose of this analysis was to calculate a whole-brain, spatially 297 continuous (i.e., unthresholded) map for subsequent comparison with external 298 datasets of patients with consciousness-impairing brain lesions and seizures (see 299 below). To identify the cortical networks most implicated in improvement, we 300 calculated the mean *t*-score within each of the seven canonical "Yeo atlas" 301 networks.⁴³ Networks showing a positive mean *t*-score (i.e., more associated with 302 improvement) were, in descending order, the visual, dorsal attention, 303 frontoparietal, and default-mode networks. In contrast, the somatomotor and 304 ventral attention networks showed a negative mean *t*-score (non-improvement). 305 Applying a voxel-wise threshold of p < 0.05 (uncorrected), the improved group

306 showed stronger connectivity with the hypothalamus, midbrain, pons, dorsal

307 cerebellum, posterior hippocampus, and parieto-occipital fissure (Fig. 5A).

308

309 Alignment with consciousness-impairing lesion network

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Our previous analyses aimed to identify the brain network underlying favorable 311 312 DBS outcomes. To assess the external validity of these findings, we tested 313 whether effective DBS for DoC modulates the same networks disrupted by brain 314 lesions that cause impaired arousal in a separate cohort. Specifically, we tested 315 whether the DBS improvement network (Fig. 5A) overlapped the pathological 316 brain networks in 45 patients who had acute thalamic or brainstem lesions from 317 stroke or trauma. As previously described,^{45, 46} these patients were ordinally ranked based on the severity of their arousal impairment post-injury using scores 318 319 from 1-6, with lower scores indicating more severe impairment. For each patient, 320 we examined the brain network connected to their lesion using the same normative fMRI dataset described earlier,^{43, 44} producing 45 lesion connectivity 321 322 maps, each linked to a patient's arousal score. We then calculated the spatial 323 similarity (Pearson correlation) between these lesion connectivity maps and our 324 DBS improvement network, predicting that higher similarity would correlate with

325 worse outcomes (lower arousal scores). Consistent with this, we found that 326 lesions associated with worse arousal impairments were connected to the positive 327 regions in our DBS improvement network (Spearman rho=-0.5, p=0.0007; Fig. 328 **5B**). Similarly, when restricting the analysis to patients with coma (n=14) and those who were awake (n=15),^{45, 46} we found that coma patients' lesion 329 connectivity maps had higher similarity to our DBS improvement network than 330 331 awake patients (*t*=3.8, *p*=0.0006, permutation-based two-sample *t*-test). 332 333 Alignment with consciousness-impairing seizure network 334 335 In a final analysis, we tested whether effective DBS for DoC modulates the same network disrupted by absence seizures, which are brief lapses of awareness 336 marked by generalized spike-wave discharges on scalp EEG.⁴⁷ We used findings 337 338 from a previous study of 15 patients with absence epilepsy who underwent simultaneous EEG-fMRI,⁴⁸⁻⁵⁰ a technique that can measure whole-brain BOLD 339 340 signal changes time-locked to epileptiform EEG events. This analysis produced a 341 group-level brain map showing areas linked to absence seizure-related disruption 342 of awareness (Fig. 5C). Like in our earlier analysis of arousal-impairing brain lesions, we aimed to compare this EEG-fMRI map with our DBS improvement 343 344 network, hypothesizing that the brain network where DBS improves 345 consciousnesses overlaps with the network where seizures disrupt it. Using spinpermutation testing,⁵¹ we found that areas of BOLD signal suppression during 346 generalized spike-wave discharges overlapped the positive regions in our DBS 347 348 improvement network (Spearman rho=-0.43, p=0.026); this BOLD suppression is 349 thought to contribute to the transient lapses of awareness during absence seizures^{47, 52} and is most prominent in areas of the default-mode network.⁴⁸⁻⁵⁰ 350 351 352 353 354 355 356 357 358

359

360 Figure 5: Optimal functional connectivity and alignment with brain networks

- 361 **disrupted in other consciousness-impairing conditions. (A)** A DBS improvement
- 362 network was calculated by comparing functional connectivity of stimulation sites
- between improved (n=10) and non-improved (n=18) groups, using normative fMRI
- data.^{43, 44} Results are *t*-scores (unthresholded), where positive values indicate regions of
- higher connectivity in the improved group and negative values indicate lower
- 366 connectivity. Cortical views are shown using the $f_s LR_32k$ template
- 367 (<u>https://balsa.wustl.edu/QXj2</u>) and subcortical views using the BigBrain histological
 368 atlas³¹ registered to MNI space.³². (B) We spatially compared the DBS improvement
- atlas³¹ registered to MNI space.³². (B) We spatially compared the DBS improvement
 network (from A) to patterns of functional connectivity seen in a separate group of 45
- 370 patients with acute-onset, arousal-impairing lesions.^{45, 46} Axial views are frequency maps
- 371 showing all lesion locations. For each lesion, we calculated a whole-brain functional
- 372 connectivity map. We then assessed spatial similarity between each map and our DBS
- improvement network. The scatter plot shows a significantly negative association with
- arousal outcomes in the lesional group (measured as a 6-point rating, where lower
- 375 values indicate more severe impairment; for details, see^{45,46}). (C) We also assessed
- 376 spatial similarity between the DBS improvement network and areas of blood-oxygen-

377 level-dependent (BOLD) signal change during generalized spike-wave discharges in 15 patients with absence epilepsy scanned using simultaneous EEG with functional MRI 378 379 (EEG-fMRI).⁴⁸⁻⁵⁰ The EEG trace shows an example of these discharges recorded inside the MRI scanner (for methodological details, refer to⁴⁸⁻⁵⁰). EEG-fMRI results are 380 381 displayed as group-level z-scores where positive values indicate increased BOLD signal 382 during discharges and negative values indicate decreased BOLD signal. The 383 histogram/density plot shows the results of spin permutation testing (10,000 spins).⁵¹ 384 demonstrating a significantly negative correlation between the EEG-fMRI map and the 385 DBS improvement network. Abbreviations: Conn., Connectivity, EEG, 386 Electroencephalogram, ECG, Electrocardiogram, Func., Functional, L, Left, R, Right. 387

388 Secondary analyses

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390 Results were similar across three secondary analysis designs: (i) using a finer 391 outcome measure based on the change in total CRS-R scores (instead of a binary 392 improved/non-improved classification), (ii) excluding three patients who died within 393 12 months of DBS, and (iii) excluding three patients with right- (instead of left-) 394 sided DBS implants (Supplementary Material). Specifically, all secondary 395 analyses showed that improvement was linked to less severe baseline impairment 396 (DRS and C/NC scores), younger age at injury, and larger volumes of the 397 putamen and cerebellum (p<0.05, uncorrected). Patients who improved also 398 continued to show larger volumes of whole-brain gray matter and the caudate 399 when using CRS-R scores and when excluding patients who died (p < 0.05, 400 uncorrected), but not when excluding patients with right-sided implants (p>0.05). 401 Regarding optimal structural connectivity, the fiber tract distributions linked to DBS 402 improvement were like those in the primary analysis; the findings remained robust 403 to k-fold cross-validation when using CRS-R scores (Pearson r=0.42, p=0.028), 404 and when excluding patients with right-sided implants (t=2.29, p=0.03), but did not 405 reach significance when excluding patients who died (t=1.18, p=0.25). Finally, for 406 optimal stimulation sites, the peak location associated with improvement remained in the ventral parafascicular nucleus, with a Euclidean distance of <0.7 mm from 407 408 the primary analysis peak in all secondary designs. However, unlike the primary 409 analysis, these secondary analyses did not survive cross-validation (p>0.05). 410 411 412 413

414 **DISCUSSION**

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416 We investigated the optimal preoperative clinical features, brain tissue integrity, 417 stimulation target, and connectivity for CM-Pf DBS in DoC. Our findings suggest 418 the importance of selecting patients based on their latent capacity for recovery and neuroplasticity^{13, 53} and stimulation targets based on their position and 419 connectivity within the ascending arousal network.^{40, 54-56} The results shed light on 420 potential mechanisms through which disparate stimulation targets trialed over the 421 past 50 years may exert therapeutic effects for DoC,⁵⁻¹⁹ and provide foundational 422 423 data to design future prospective and controlled trials.

424

425 Patients who improved with DBS were younger and had less severe baseline 426 impairment. These findings align with prior studies on the influence of age and cognitive reserve on recovery after brain injuries.^{13, 19, 57} In stroke⁵⁸ and TBI,⁵⁹ 427 428 older patients show slower response to rehabilitation than younger patients with comparable injuries. Similar trends occur in DBS for Parkinson's disease¹⁹ and 429 transcranial magnetic stimulation for depression,⁶⁰ where younger age predicts 430 better outcomes (though conflicting findings exist).⁶¹ This may reflect reduced 431 capacity for brain plasticity and repair⁶² in older age, limiting stimulation-induced 432 changes in neurogenesis,⁶³ myelination,⁶⁴ gene expression,^{65, 66} and 433 neurotransmitter release,⁶⁷ all of which may contribute to DoC recovery.^{3, 21} 434 435

436 Improvement was seen in patients with more intact gray matter volumes on 437 structural MRI, at the whole-brain level and more specifically in the striatum and 438 cerebellum. Like younger age, gray matter preservation may be essential for 439 treatment effects of DBS to take hold, whether in the short-term (i.e., acute 440 functional changes requiring intact neuronal assemblies, like the synaptic transmission of stimulation) or longer term (i.e., chronic functional and/or structural 441 neuroplastic changes).^{68, 69} In keeping with this hypothesis, we previously found 442 443 that effective CM-Pf DBS for DoC is associated with longitudinal gray matter volume increases up to seven years post-DBS, including in the striatum.⁷⁰ 444 445

446 Preservation of the striatum and improved capacity for recovery is consistent with
447 the mesocircuit model of DoC proposed by Schiff.²⁰ In this model, DoC emerge

from a diffuse suppression of synaptic input from the cortex to striatal medium spiny neurons, leading to a loss of inhibitory projections to the global pallidus and in turn a tonic inhibition of the thalamus, culminating in a breakdown of anterior forebrain arousal.²⁰ Increasing activity within this circuit, for example by DBS, is thus thought to restore consciousness. Preserved volume in improved patients suggests the striatum may play a gating role⁷¹ in restoring mesocircuit activity and thus could serve as a preoperative marker to identify optimal DBS candidates.

456 Despite targeting the CM-Pf region in all patients, there was natural variability in 457 electrode positions and stimulation. Correlating this variability with outcomes, we 458 found that improved patients had greater stimulation of the inferior Pf and subparafascicular nucleus. Although the CM and Pf are often described as a 459 unitary complex, the two nuclei can be distinguished anatomically,⁷² functionally,^{73,} 460 ⁷⁴ and connectomically,^{22, 23, 75} which may confer differential stimulation effects. In 461 462 primates, the CM and Pf are dominant sources of glutamatergic input to the 463 striatum, but their projection profiles differ: the CM projects to striatal territories 464 receiving input from sensorimotor cortex, particularly the caudal putamen, while 465 the Pf projects to association and limbic territories, including the anterior putamen, caudate, and nucleus accumbens.^{22, 23, 75} Their extra-striatal projections also 466 467 differ, with the Pf having more input to the hypothalamus, amygdala, and ventral tegmental area, and to prefrontal, anterior cingulate, and frontal eye field 468 469 regions.^{22, 23} Among the posterior intralaminar nuclei, the subparafascicular nucleus has the densest descending projections to the brainstem, including the 470 471 inferior olivary nucleus, peripeduncular area, reticular core, and raphe nuclei.²³ 472

The CM is a more recent evolutionary development, with maximal expansion in primates. In smaller-brained species (e.g., rodents), the CM is not clearly distinct from the Pf.⁷² This suggests that the Pf may have a more conserved role in arousal, a function of all vertebrate brains, even those lacking a distinct CM.

Hence, one explanation for the sweet spot being in the Pf and subparafascicular
nucleus stems from the preferential effects that DBS of this region may have upon
associative and limbic areas of the striatum and cortex, and the resulting influence
upon several components of the mesocircuit hypothesized to underlie DoC,

482 including striatal inputs to the global pallidus (potentially counteracting abnormal inhibition of the thalamus)²⁰ and frontal/prefrontal systems involved in polysensory 483 484 integration (potentially supporting improved awareness and higher-order cognition).³ The subparafascicular extension of the sweet spot may reflect an 485 486 added benefit of modulating the dense projections of this nucleus to the brainstem, beyond the Pf's striatal and cortical connections.²³ Alternatively, this 487 488 ventral emphasis may point to the importance of modulating structures below the 489 thalamus, like the midbrain, a potentially effective target in its own right.^{8, 17} 490

491 Our results differ from recent studies of thalamic DBS for patients with TBI, which

492 targeted the 'wing' of the central lateral (CL) nucleus,^{13, 76} specifically the medial

dorsal tegmental tract (DTTm). Like the CM-Pf, the CL is similarly thought to play

494 a key role in arousal regulation via its striatal and frontal connections.⁷⁷

495 Experimental studies in non-human primates show that CL stimulation can

496 facilitate task performance⁷⁸ and awaken animals from anesthesia.⁷⁹ Following an

497 earlier case study,¹³ a recent randomized trial in 6 patients with TBI—none with

498 DoC—found improvement in executive function after CL/DTTm DBS.⁷⁶

499

500 Direct comparisons of CL and CM-Pf stimulation are limited. One study found that 501 stimulation of the CL/DTTm, but not the CM-Pf, improved behavioral performance 502 in macagues, with the authors suggesting that CL-specific improvements may 503 stem from its selective projections to striatal medium spiny neurons and broad effects on the frontal lobe,⁷⁸ unlike the CM-Pf's more variable striatal/cortical 504 505 connections (as discussed above). However, another possibility is that therapeutic 506 effects of these two targets converge not at the level of the CL and CM-Pf 507 complex per se, but at the level of the CL and Pf specifically-perhaps, via a 508 shared influence upon the dorsal striatum and associative and limbic regions of 509 frontal and anterior cingulate cortex,^{80, 81} unlike the CM's pattern of sensorimotor connections.22,23 510

511

512 Stimulation was most beneficial when delivered to white matter pathways

513 connecting the thalamus to the brainstem, hypothalamus, cerebellar dentate

514 nucleus, and, to a lesser degree, the medial frontal cortex. This distribution

515 overlaps the recently proposed default ascending arousal network (dAAN),⁴⁰

516 which is thought to sustain resting wakeful states and connects to several cortical 517 networks, particularly ones anchored in frontal and parietal cortex involved in 518 awareness and cognition, including the default-mode, frontoparietal control, and 519 dorsal attention networks.⁸² The dAAN is thought to dynamically interact with 520 these cortical networks to provide a neuroanatomic basis linking the two 521 foundational components of human consciousness: arousal and awareness.⁴⁰ 522 The pattern of optimal connectivity shows intriguing similarities to the neural 523 circuitry underlying circadian regulation of arousal,⁸³ which has been implicated in 524 the pathophysiology and recovery of DoC.^{84, 85} Orexin-expressing neurons in the 525 526 dorsomedial hypothalamus send dense projections to the locus coeruleus, the 527 main site for norepinephrine synthesis, which has widespread excitatory effects.⁸³ 528 Activity in the locus coeruleus shows circadian variation, promoting arousal during wakefulness and being inhibited during sleep, and is under direct control by 529 projections from the hypothalamus.⁸³ Hypothalamic lesions disrupt this rhythm, 530 causing somnolence, altered body temperature, and coma-like states.^{46, 86} 531 532 Patients with DoC similarly show abnormal daily rhythms in EEG, temperature, and hormones.^{84, 85} The strength of circadian variation correlates with DoC 533 534 severity, predicts recovery, and may even have therapeutic effects when 535 exogenously entrained via, for instance, bright light stimulation.^{84, 85} These 536 findings may relate to the integrity of neural circuits driving these circadian 537 rhythms, consistent with our observation of enhanced connectivity from DBS sites 538 to the hypothalamus and locus coeruleus in patients who improved. 539

540 Our findings may aid with understanding pathophysiology, predicting outcomes, 541 and developing new treatments, both for DoC and other conditions. Brain areas 542 showing stronger connectivity with effective DBS sites overlapped with the networks underlying acute-onset lesions causing arousal impairments^{45, 46} and 543 epileptiform events associated with transient lapses of awareness.⁴⁸⁻⁵⁰ This 544 545 suggests that effective stimulation targets for DoC might have similar benefits for 546 a broader landscape of consciousness-impairing conditions. The findings may also aid with selecting cortical targets for non-invasive therapies like transcranial 547 magnetic stimulation or transcranial direct current stimulation.²¹ 548

549

550 The choice of stimulation paradigm is an important consideration when comparing our findings to prior work. We used a "medium" stimulation frequency of 20-40 Hz, 551 motivated by positive results in earlier studies,^{8, 18} while others used higher 552 frequencies of 100-185 Hz,^{10, 13, 76} likely having divergent effects.⁸⁷ The optimal 553 554 paradigm for DoC, and whether it differs by target, brain state, or species, is uncertain. Several authors^{78, 79, 88, 89} have hypothesized that stimulation efficacy 555 556 may partially depend on the intent of DBS-e.g., awakening from anesthesia or sleep versus enhancing arousal in an already conscious subject-as well as 557 558 correspondence between the stimulation frequency and resonant (intrinsic) 559 frequencies of the intended brain state. For example, 150-225 Hz stimulation of the CL facilitated task performance in awake macagues,⁸⁹ echoing the high 560 frequency of excitatory input required to trigger dendritic electrogenesis in 561 neocortical neurons.⁸⁹ In contrast, another study found that CL stimulation at a 562 lower (50 Hz), but not higher (200 Hz), frequency was effective at rousing 563 macagues from propofol/isoflurane-induced anesthesia.⁷⁹ perhaps reflecting 564 565 entrainment of central thalamic neurons that fire at similar frequencies (20-40 Hz) during wakeful states.⁹⁰ Conversely, low frequency (10 Hz) optogenetic 566 567 stimulation of the central thalamus elicited spindle-like oscillations and behavioral arrest in rats,⁸⁸ resembling patterns seen at the onset of sleep.⁹¹ 568

569

570 Hence, a possible reason for the efficacy of 20-40 Hz DBS seen here could be the 571 alignment with thalamic oscillations during wakefulness-akin to its rousing effects in anesthetized subjects.⁷⁹ Consistent with this, Arnts et al. recently reported that 572 573 30-50 Hz DBS of the CM-Pf produced stronger treatment effects than 130 Hz stimulation in one patient with DoC¹⁹ and another with akinetic mutism.⁹² 574 575 However, reports of contrary findings (e.g., restoration of consciousness during 130-180 Hz DBS in anesthetized macaques)^{93, 94} and potential interactions with 576 other stimulation parameters (e.g., amplitude)^{78, 89, 94} highlight the need for further 577 578 work to elucidate optimal paradigms and treatment mechanisms.

579

580 This study has limitations due to its clinical and retrospective nature. Patients had

various DBS device models, with subtle differences in stimulation parameters,

though not significantly associated with outcome. As in previous studies,^{10, 11, 15, 18,}

⁵⁸³⁷⁶ the time between injury and DBS varied, and it is possible that natural recovery

584 may have contributed to improvement in some patients; however, the lag time to DBS did not significantly differ between improved and non-improved groups, 585 586 arguing against this as a major confound. Patients showed structural brain 587 abnormalities, including cortical injuries and diffuse atrophy, presenting challenges 588 for accurate image registration and raising guestions about using normative data^{37, 95} to assess connectivity in patient brains.^{35, 96} We mitigated this by 589 590 excluding patients with severe abnormalities preventing accurate template 591 alignment and assessed the predictive utility of our findings using cross-validation 592 techniques and examining relevance to patients with consciousness-impairing lesions^{45, 46} and seizures.^{48, 49} However, future replications will be important. 593

594

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018	COMPETING INTERESTS
619	
620	AELW, MR, HF, FLWVJS, JT, SBS, JL, MMJC, KB, MUF, RJ, JEI, PWC, DF,
621	ADB, BLE, and DC have no competing interests to report. MDF has intellectual
622	property on the use of brain connectivity imaging to analyze lesions and guide

- 623 brain stimulation, has consulted for Magnus Medical, Soterix, Abbott, Boston
- 624 Scientific, and Tal Medical, and has received research funding from Neuronetics.
- 625 AH reports lecture fees for Boston Scientific and is a consultant for
- 626 Neuromodulation and Abbott. JDR has received past consulting payments from
- Medtronic, Corlieve, ClearPoint, Medtronic, and NeuroPace, and currently 627
- 628 consults for Turing Medical.
- 629

(10

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631

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634

635 DATA AVAILABILITY

636

637 Anonymized patient-level clinical information is available in Supplementary

- Material. The optimal stimulation site ("sweet spot") and structural connectivity 638
- 639 results will be available within Lead-DBS software upon publication (www.lead-
- 640 dbs.org). Normative functional MRI and diffusion MRI data are publicly available:
- 641
- 642 https://datadryad.org/stash/dataset/doi:10.5061/dryad.nzs7h44g2
- 643 https://dataverse.harvard.edu/dataset.xhtml?persistentId=doi:10.7910/DVN/25833 644

CODE AVAILABILITY 645

- 646
- Code used to analyze the dataset is openly available within Lead-DBS software 647
- (https://github.com/leaddbs/leaddbs). 648
- 649
- 650 **METHODS**
- 651

652 Methods overview: a pragmatic analysis approach

653

This cohort is the largest sample of patients with DoC undergoing DBS reported to
date, making it a valuable dataset to address key scientific and clinical questions.
However, this patient population presents unique challenges, including abnormal
neuroanatomy¹⁻³ and low DBS response rates (~30% in previous studies^{9, 12, 18, 97}),
making some analysis conventions unsuited to such a rare population. Given
these challenges, we made pragmatic decisions in our analysis and reporting.
Accurate brain alignment to a common template is required for group-level

- analysis of optimal DBS locations and connectivity.³⁵ We excluded patients with
- 663 severe brain abnormalities causing poor template alignment, as determined by
- two neuroimaging experts blinded to clinical outcomes. However, these patients
- 665 were retained for other analyses not requiring template alignment, including
- 666 clinical variables associated with improvement. For clarity, **Supplementary**
- 667 **Material** details the specific analyses each patient's data contributed to, and the
- sample size for each analysis is noted throughout the results section.
- 669

Patients did not undergo advanced MRI connectivity sequences like functional or
diffusion MRI. For analysis of optimal connectivity profiles, we used normative
connectivity data acquired in healthy participants, an approach we have previously
used to generate robust predictive models of DBS outcome in Alzheimer's
disease,³⁸ epilepsy,⁹⁸ Parkinson's disease,⁹⁹ and more.³⁵

675

Finally, for hypothesis tests involving multiple comparisons, we report both
uncorrected and false discovery-rate (FDR)-corrected *p*-values. This approach
ensures transparency while balancing statistical rigor with the importance of
preserving the exploratory insights afforded by this unique dataset.

680

681 Study design and ethics

682

This was a retrospective analysis of a previous clinical study^{9, 12} of patients with
 DoC who underwent DBS at Dubrava University Hospital in Zagreb, Croatia. The
 clinical study received approval from the institutional review board of Dubrava

686 University Hospital and the School of Medicine at the University of Zagreb, and

687 informed consent was obtained from patients' families or caregivers. Approval for

retrospective data analysis performed in the current study was received from the

689 Mass General Brigham institutional review board.

690

691 Patients and DBS surgery

692

The analysis cohort included 40 patients. A description of study procedures and 693 outcomes from 32 of these patients has been previously published.^{9, 12} Patients 694 695 were selected based on neurophysiologic, clinical, and neuroimaging evaluations.¹² Briefly, inclusion criteria included: (i) meeting clinical diagnostic 696 criteria for UWS or MCS;^{100, 101} (ii) a minimum DoC diagnosis duration of 6 weeks; 697 698 (iii) obtainable somatosensory evoked potentials (SSEP) via median nerve 699 stimulation, with or without SSEPs from tibial nerve stimulation; (iv) periods of 700 desynchronized scalp EEG activity observed during 12-24 hours of monitoring; (v) 701 sufficient hemodynamic and respiratory stability to undergo study procedures; and (vi) absence of significant lesions (e.g., hemorrhages or infarctions) in the 702 brainstem, diencephalon, or basal ganglia.¹² The last criterion was based on the 703 704 hypothesis that recovery potential depends on the integrity of subcortical nuclei 705 and their dynamic interactions with cortical networks.³ For MRI examples of 706 patients who met and did not meet this criterion, see figure 1 in Chudy et al.¹². 707

The DBS procedure involved unilateral implantation of the CM-Pf. The rationale 708 709 for unilateral, as opposed to bilateral, implantations was based on experience in prior studies, and to reduce surgical risk.^{8, 18} Most cases (37/40 patients) were 710 711 implanted on the left (typically dominant) hemisphere, also motivated by 712 experience in earlier studies^{8, 18} The remaining 3/40 were implanted on the right 713 due to left-sided injuries or anatomical variations that made the right thalamus a more surgically feasible target. For group-level analyses, we flipped these right 714 715 DBS leads to the left hemisphere by mirroring patients' MRI/CT scans about the x axis as a first step, prior to further image processing steps.³⁵ 716 717

Device models varied and included Medtronic lead models 3387 or 3389, as well
as Boston Scientific Vercise leads; model type was not significantly associated

720	with outcomes (Table 1). Surgical target coordinates were defined on
721	preoperative CT using the Schaltenbrand-Bailey ¹⁰² atlas: 4.5 mm anterior to the
722	posterior commissure, 1 mm inferior to the inter-commissural line, and 4 mm
723	lateral to the third ventricular wall. The lateral coordinate was defined with respect
724	to the third ventricular wall (rather than the inter-commissural line) to account for
725	widened ventricles due to brain atrophy typically seen in this patient group. ^{9, 103}
726	
727	Three days post-surgery, DBS devices were programmed to deliver a stimulation
728	paradigm optimized per patient to elicit the strongest arousal reaction, as
729	previously described. ^{9, 12} Briefly, this involved testing each electrode contact using
730	a stimulation frequency of 20-40 Hz, pulse width of 120-330 $\mu s,$ and amplitude of
731	2-4.5 V or 2.5-5.5 mA. An arousal reaction was defined by eye opening (if the
732	patient's eyes were closed) with mydriasis and change in facial expression, with or
733	without head turning and elevation of blood pressure and heart rate. ^{9, 12}
734	Stimulation parameters for each patient are reported in Supplementary Material .
735	Stimulation was administered for 30 minutes every 2 hours during the day and
736	ceased at night with the aim of promoting circadian (sleep-wake) cycles.8

737

738 Clinical outcomes

739

Outcomes were tracked using the Coma Recovery Scale-Revised (CRS-R).²⁶ 740 741 Total scores range from 0-23, with higher scores indicating a higher level of consciousness across auditory, visual, motor, oromotor, communication, and 742 743 arousal subscales.²⁶ Supplementary clinical measures included the Disability Rating Scale (DRS) and Coma/Near-Coma (C/NC) scale.²⁷ Patients were 744 745 classified as having unresponsive wakefulness syndrome (UWS), a minimally conscious state (MCS), or full consciousness.¹ We defined "improved" patients as 746 747 those who transitioned from UWS to MCS or conscious, or from MCS to conscious, within 12 months post-DBS, while "non-improved" patients were those 748 749 who did not change states, as in our previous work.^{9, 12} Dichotomization of the cohort into improved and non-improved groups was intended to enhance our 750 751 sensitivity to factors driving clinically significant improvements in consciousness.¹⁰⁴ However, we performed secondary analyses using each 752

24

753 patient's change in CRS-R scores to test consistency across a different definition 754 of improvement.

- 755
- 756

DBS electrode localization and stimulation-induced electric fields

757

DBS electrodes were reconstructed using Lead-DBS software (Fig. 1).³⁵ Given the 758 759 study's retrospective nature, the types of pre- and post-operative imaging data varied (MRI, CT, or both). Electrode localizations were therefore optimized per patient, 760 761 based on available data. When a T1 MRI scan was available (28/40 patients), it was 762 used as the reference image for non-linear spatial warping to template space; in 763 other cases, we used CT. MRI scans were acquired on a 1.5 T Siemens Avanto or 764 Aera scanner using a volumetric T1-weighted magnetization-prepared rapid acquisition gradient echo (MPRAGE) sequence with a voxel resolution ≤1mm³. CT 765 766 scans were acquired using a Siemens scanner with slice thickness ≤0.5mm. 767

768 We first linearly co-registered the post- to the pre-operative image using Advanced Normalization Tools software,¹⁰⁵ then calculated nonlinear spatial warps to the 769 770 Montreal Neurological Institute (MNI) 152 ICBM 2009b nonlinear asymmetric brain 771 template. To accommodate the heterogenous imaging modalities (MRI or CT) 772 available for these nonlinear warps, we used a recently developed deep learningbased tool, EasyReg,¹⁰⁶ which can perform robust, modality-agnostic registrations,¹⁰⁷ 773 774 unlike classical techniques that rely upon optimization of similarity metrics between images.¹⁰⁶As shown in Billot et al.,³³ this strategy can cope with CT scans, despite 775 776 their low soft-tissue contrast. This is because synthetic images with low contrast-tonoise ratio are regularly seen during model training.¹⁰⁶ To further optimize the 777 778 performance of EasyReg for CT, we followed the approach of Billot et al.,³³ and 779 stretched the histogram of CT values in the soft-tissue interval (0<HU<80) using the 780 piece-wise linear "tone-mapping" function implemented in Lead-DBS software.³⁵ 781

782 Accuracy of image registrations was reviewed by two authors (AELW and AH).

blinded to clinical outcomes. Twelve patients were excluded from further analysis 783

involving DBS localizations, primarily due to severe brain atrophy and/or grossly 784

785 enlarged ventricles resulting in poor template alignment. However, these patients

were retained for other analyses not requiring alignment to MNI space. Details of the
 specific patients included in each analysis are provided in Supplementary Material.

DBS electrodes were localized using the PaCER¹⁰⁸ or TRAC/CORE³⁵ algorithm in
Lead-DBS.³⁵ We then calculated electric fields (E-fields) for each patient's
stimulation settings using the finite element method (FEM) implemented in
FieldTrip/SimBio.³⁶ E-fields represent the first derivative of the estimated voltage
distribution applied to voxels in space; the field's magnitude is strongest near
active electrode contacts and diminishes rapidly with distance.³⁸

795

796 Analysis of optimal brain tissue integrity

797

798 We assessed whether whole-brain and subcortical tissue volumes differed 799 between improved (n=8) and non-improved (n=18) groups. Each patient's T1-800 weighted MRI scan was segmented into whole-brain volumes of gray matter. white matter, and cerebrospinal fluid using SynthSeg software.³³ Additionally, we 801 802 segmented regional subcortical volumes and carefully inspected the results for 803 accuracy. To adjust for inter-patient variability in brain size, each segmented 804 volume was normalized by the total intracranial volume. We additionally 805 normalized by tissue volumes from age-matched samples of T1-weighted MRI scans from the Nathan Kline Institute-Rockland Sample (NKI-RS).³⁴ This 806 807 longitudinal, community-ascertained neuroimaging study includes >1,500 808 individuals aged 6-85 years. For each DoC patient, we selected a subset of T1-809 weighted MRI scans from NKI-RS control participants whose ages matched the 810 patient's age within a ±2-year range. On average, we found 92 matching control 811 participants per patient (range=46-126). The NKI-RS scans were processed using SynthSeg,³³ normalized by total intracranial volume and used to convert each 812 patient's whole-brain and subcortical volumes into z-scores. 813 814 815 Analysis of optimal stimulation sites

- 816
- To identify optimal stimulation sites, we compared E-fields between improved
- (n=10) and non-improved (n=18) groups.³⁵ As in prior studies,^{38, 39} we focused on
- voxels covered by >25% of E-fields with a magnitude >200 V/m, a commonly

820 assumed estimate of the voltage required to activate axons.^{38, 109} We initially 821 conducted a series of voxel-wise *t*-tests, resulting in a map of *t*-scores (*t*-map) 822 where positive values indicate higher E-field magnitudes in the improved relative 823 to non-improved patients. To identify a candidate site of optimal stimulation, the t-824 map was thresholded at p < 0.05 (uncorrected). We then subjected this candidate site to k-fold cross-validation with k=10, where k was the number of groups into 825 826 which the dataset was randomly split.^{35, 38} A k of 10 was used to align with previous similar DBS studies.^{38, 99} We iteratively used k-1 folds for training and the 827 828 remaining fold for testing. In each iteration, the *t*-map was recalculated, leaving 829 out the E-fields of patients in the test fold. The clinical outcomes for the left-out 830 patients were then estimated by calculating the peak value of a voxel-wise multiplication of their E-field distributions with the derived *t*-map. These estimates 831 832 were then compared between the improved and non-improved patients using a 833 two-sample *t*-test. The intuitive interpretation of this analysis is that positive values 834 in the *t*-map represent better stimulation locations. By testing whether E-fields 835 from left-out patients more strongly overlapped with the positive sites in the *t*-map. 836 we evaluated the robustness and potential predictive utility of our findings.^{35, 38} 837

838 Analysis of optimal structural connectivity

839

To identify white matter tracts associated with improvement, we utilized the fiber 840 841 filtering approach in Lead-DBS software with normative structural connectome 842 (**Fig. 4**)^{35, 38}. Given the potential importance of small and intricate connections 843 within and around the thalamus, and to the brainstem, we used a state-of-the-art, 844 ultra-high resolution (760 µm) diffusion-weighted MRI dataset acquired across 18 hours,³⁷ as detailed in our recent work.^{38, 39} Like the analysis of optimal stimulation 845 sites, which analyzed stimulations across voxels, here we examined stimulations 846 across streamlines of the normative connectome in the same, mass-univariate 847 848 fashion. For each streamline and E-field pair, we recorded the peak magnitude 849 that the streamline traversed. Then, we performed the same *t*-tests on the E-field 850 magnitudes between improved and non-improved groups, yielding a *t*-value for 851 each streamline, with positive t-scores indicating exposure to E-fields that were 852 higher in the improved group. To identify a candidate network of optimal structural connections, we again applied a threshold of p < 0.05 (uncorrected), then tested 853

854	robustness of this network using the same k -fold cross-validation (k =10)
855	procedure. Specifically, we iteratively assigned <i>t</i> -scores to streamlines, each time
856	leaving out E-fields of patients in the test fold. We then computed the peak
857	overlap between the left-out E-fields and the t-weighted streamlines for each
858	patient, comparing the results between groups using a two-sample t-test. ^{35, 38} To
859	define subcortical nuclei traversed by the observed fiber tracts, we compared
860	results to atlases of the hypothalamus ⁴¹ (<u>https://zenodo.org/records/3942115</u>),
861	brainstem (https://www.nitrc.org/projects/brainstemnavig), ascending arousal
862	network ⁴⁰ (https://doi.org/10.5061/dryad.zw3r228d2), and cerebellum ⁴²
863	(https://www.diedrichsenlab.org/imaging/propatlas.htm).
864	
865	Analysis of optimal functional connectivity
866	
867	We investigated blood-oxygen-level-dependent (BOLD) connectivity of DBS sites
868	using a normative, sex-balanced sample of resting-state fMRI scans acquired in
869	1,000 healthy adults (500 males, 500 females) from the Brain Genomics
870	Superstruct Project.43,44 The fMRI data and pre-processing pipeline are publicly
871	available.95 For each patient's E-field location, we calculated the mean BOLD
872	time-course (in the normative scans) using a weighted average across all voxels
873	with E-field magnitudes >200 V/m, then measured connectivity with every brain
874	voxel using Fisher's r-to-z transformed Pearson correlations. Connectivity
875	strengths were then compared between improved and non-improved groups using
876	voxel-wise <i>t</i> -tests in permutation analysis of linear models (PALM) software. ¹¹⁰
877	The results of this analysis was a spatially continuous (i.e., unthresholded) map of
878	brain areas showing greater functional coupling with DBS sites linked to
879	improvement (positive <i>t</i> -scores) or non-improvement (negative <i>t</i> -scores).
880	
881	Comparison with consciousness-impairing lesion network
882	
883	We hypothesized that the brain network underlying DBS improvement in DoC may
884	overlap with the pathological circuits underlying consciousness-impairing brain
885	lesions. In other words, we reasoned that the network where stimulation improves
886	consciousness may reflect the network where lesions disrupt it. To test this, we
887	studied a group of 45 patients—unrelated to those who underwent DBS in the

888 current study—who had acute-onset lesions in the thalamus or brainstem due to 889 stroke or head trauma. The patients were obtained from two sources: one study of patients with lesion-induced coma,⁴⁵ and another of patients with variable 890 outcomes ranging from coma to no impairment (i.e., awake).⁴⁶ In the latter study, 891 patients were ordinally ranked using scores from 1-6.46 based on clinical 892 definitions of Plum and Posner,¹¹¹ with lower scores indicating more severe 893 impairment (coma=1; stupor=2; obtunded=3; somnolent=4; lethargic=5; 894 895 awake=6). We combined the two datasets by assigning all coma patients a score 896 of 1 (in both studies) while retaining the original rankings from the second study 897 for patients with outcomes less severe than coma (i.e., scores from 2-6). Using the same normative resting-state fMRI data^{43, 44, 95} and processes described 898 earlier, we used binary lesion masks as seeds and calculated functional 899 900 connectivity with all brain voxels to create a lesion connectivity map for each 901 patient. We then calculated a similarity score between each patient's lesion 902 connectivity map and our DBS improvement network (Fig. 5A) using spatial 903 (Pearson) correlations. Finally, we tested whether higher similarity to our DBS 904 improvement network was associated with worse outcomes (i.e., lower arousal 905 scores) using a rank-based, non-parametric Spearman correlation (Fig. 5B). 906

Since the latter study⁴⁶ included both patients with coma and others with more variable levels of impairment, we conducted an additional analysis focusing solely on patients with lesion-induced coma (n=14) and those who were awake (n=15), comparing the groups using a two-sample *t*-test.

911

912 **Comparison with consciousness-impairing seizure network**

913

914 In a final analysis, we explored whether effective DBS sites for DoC modulate the 915 same network that is disrupted by absence seizures, which are brief lapses of 916 awareness marked by generalized spike-wave discharges (GSW) on scalp EEG. 917 We used findings from a previous study of 15 patients with absence epilepsy who underwent up to 60 mins of EEG-fMRI.⁴⁸⁻⁵⁰ GSW timings were manually marked 918 on the EEG and used as regressors in a whole-brain fMRI analysis to identify 919 920 discharge-related BOLD signal changes. Event-related independent component 921 analysis (eICA)^{48, 112} was employed to detect BOLD patterns deviating from the

922 canonical hemodynamic response function (HRF), which is often seen with epileptiform events.^{113, 114} The eICA was performed on temporally concatenated 923 924 fMRI data from all patients, covering a 32-second window before and after GSW onset.^{48, 112} Thirteen brain components significantly associated with GSW were 925 926 identified (*F*-test; p<0.05, Bonferroni-corrected), each represented by a spatial 927 map (z-scores) and a BOLD time-course. Positive z-scores indicated regions with 928 increased BOLD signal (activation) and negative z-scores indicated decreased 929 signal (deactivation). We averaged all z-score maps together to create one map 930 representing overall patterns of activation/deactivation (Fig. 5C). Finally, both this 931 map and our DBS improvement network were warped to FreeSurfer's fsaverage5 932 template.¹¹⁵ Spatial similarity was then measured using a Spearman correlation,

- 933 with significance assessed via spin-permutation testing (10,000 spins).⁵¹
- 934

935 Statistical analyses

936

937 Analyses were performed using MATLAB version R2023b and RStudio version

938 2022.07.01. For comparison of clinical variables and MRI tissue volumes between

939 improved and non-improved groups, we used non-parametric, permutation-based

940 two-sample *t*-tests (10,000 permutations) for continuous variables and Fisher's

941 exact tests for categorical variables. Statistical procedures for the remaining

942 analyses are described in the methods. Significance was defined using an alpha

943 of 0.05 (two-tailed). For hypothesis tests involving multiple comparisons, we report

both uncorrected and Benjamini-Hochberg¹¹⁶ FDR-corrected p-values.

945

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	DoC improved (<i>n</i> =11), median (IQR) or proportions	DoC non-improved (<i>n</i> =29), median (IQR) or proportions	<i>p</i> -value uncorrected (FDR-corrected), effect size (95% CI)
Age at injury (years)	19 (16-33)	39 (24.5-54)	p=0.03 (0.14) Hedge's g=-0.78 (-1.63, -0.09)
Sex (Male:Female)	8:3	21:8	<i>p</i> =1.0 (1.0), OR=1.0 (0.18, 7.43)
Time from injury to DBS (months)	6 (3-12)	6 (3.5-14)	p=0.67 (0.94), Hedge's g=-0.19 (-0.53, 0.56)
C/NC score before DBS [#]	2 (1-2)	3 (2-3.5)	<i>p</i> =0.0009 (0.01), Hedge's <i>g</i> =-1.21 (-2.11, -0.58)
DRS score before DBS [#]	2.2 (1.8-2.6)	3.2 (2.6-3.5)	<i>p</i> =0.003 (0.02), Hedge's <i>g</i> =-1.09 (-2.28, -0.47)
CRS-R score before DBS [§]	7 (6-9)	4 (4-6)	p=0.06 (0.21), Hedge's g=0.67 (0.06, 2.33)
DoC state prior to DBS (UWS:MCS)	8:3	26:3	<i>p</i> =0.32 (0.75), OR=0.32 (0.04, 2.85)
Implant side (Left:Right)	10:1	27:2	<i>p</i> =1.0 (1.0), OR=0.75 (0.04, 48.02)
Cause of injury (CA:TBI)	8:3	20:9	<i>p</i> =1.0 (1.0), OR=1.19 (0.21, 8.63)
DBS lead model (M3387:M3389:BSCI)	4:2:5	10:10:9	p=0.6 (0.94)
Stimulation amplitude $(V)^{\dagger}$	3.25 (3-3.5)	3.25 (2.3-3.7)	p=0.67 (0.94), Hedge's g=0.2 (-0.33, 0.79
Stimulation amplitude $(mA)^{\dagger}$	4.5 (4-5)	4.5 (4-5)	<i>p</i> =0.75 (0.95), Hedge's <i>g</i> =0.21 (-1.24, 1.05)
Stimulation frequency (Hz)	40 (25-40)	30 (25-30)	<i>p</i> =0.13 (0.36), Hedge's <i>g</i> =0.57 (-0.22, 1.6)
Stimulation pulse width (µs)	210 (210-210)	210 (180-210)	p=0.38 (0.76), Hedge's g=0.34 (-0.07, 0.8)

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1291Table 1: Comparison of demographic and clinical variables between patients who1292improved (n=11) and those who did not improve (n=29) with DBS. Results are1293reported using uncorrected and false discovery rate (FDR)-corrected p-values, the latter1294corrected for 14 clinical and demographic variables tested.

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1296 Abbreviations: BSCI, Boston Scientific Vercise lead model; CA, Cardiac arrest; CI,

1297 Confidence interval; CNC, Coma/near-coma; M3387, Medtronic lead model 3387;

1298 M3389, Medtronic lead model 3389; MCS, Minimally conscious state; OR, Odds Ratio;

- 1299 DRS, Disability Rating Scale; TBI, Traumatic brain injury; UWS, Unresponsive
- 1300 wakefulness syndrome.
- 1301
- [#]Note: a higher score on the C/NC scale and DRS indicates more severe impairment. 1302
- [§]Note: a higher score on the CRS-R scale indicates less severe impairment. 1303
- 1304 [†]Note: comparisons of stimulation amplitude were performed separately in patient sub-
- 1305 groups for whom amplitude was recorded as voltage (V; n=6 improved versus n=20 non-
- 1306 improved) or milliamps (mA; n=5 improved versus n=9 non-improved).

-3 Non-improved

Improved 3

