

1 **Title:** MTL functional connectivity predicts stimulation-induced theta power

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22
23 **Abstract/Summary**

24
25 Focal electrical stimulation of the brain incites a cascade of neural activity that
26 propagates from the stimulated region to both nearby and remote areas, offering the potential to
27 control the activity of brain networks. Understanding how exogenous electrical signals perturb
28 such networks in humans is key to its clinical translation. To investigate this, we applied
29 electrical stimulation to subregions of the medial temporal lobe in 26 neurosurgical patients
30 fitted with indwelling electrodes. Networks of low-frequency (5-13 Hz) spectral coherence
31 predicted stimulation-evoked increases in theta (5-8 Hz) power, particularly when stimulation
32 was applied in or adjacent to white matter. Stimulation tended to decrease power in the high-
33 frequency broadband (HFB; 50-200 Hz) range, and these modulations were correlated with
34 HFB-based networks in a subset of subjects. Our results demonstrate that functional connectivity

35 is predictive of causal changes in the brain, capturing evoked activity across brain regions and
36 frequency bands.

37

38 **Introduction**

39

40 Intracranial brain stimulation is increasingly used to study disorders of human behavior
41 and cognition, but very little is known about how these stimulation events affect neural activity.
42 Though several recent studies have demonstrated the ability to modulate human memory with
43 direct electrical stimulation (DES) of the cortex¹⁻⁷, none have described the mechanism by which
44 stimulation yields altered cognitive states. However, understanding how the brain responds to
45 these exogenous currents is necessary to ultimately develop therapeutic interventions that rely on
46 DES^{8,9}.

47 Specifically, investigators have long have asked whether the brain's intrinsic functional
48 or anatomical architecture can predict how mesoscale stimulation events propagate through the
49 brain. Early work focused on inferred connectivity through stimulation-evoked behavior in
50 rodents^{10,11}. More recently, Logothetis and colleagues demonstrated that the effects of electrical
51 stimulation propagated through known anatomical connections in the macaque visual system^{12,13}.
52 In humans, corticocortical evoked potentials (CCEPs), measured with intracranial EEG (iEEG),
53 have also been shown to propagate through anatomical and functional connections^{14,15}, as has the
54 fMRI BOLD response to stimulation¹⁶. These studies provide powerful evidence that the effects
55 of stimulation are determined by the connectivity profile of a targeted region. More broadly,
56 renewed interest in the idea of the brain as a controllable network¹⁷⁻¹⁹ raises a testable hypothesis
57 in need of empirical validation: to what extent does a brain's network architecture predict the
58 cascade of physiologic change that accompanies a stimulation event?

59 In this study, we asked whether the functional connectivity of a stimulated region predicts
60 where we observe changes in neural activity. To expand on prior work that has examined
61 network architecture and stimulation, we adopted a paradigm that (1) focuses on stimulation's
62 effect on low-frequency (theta) power, a cognitively-relevant electrophysiological biomarker,
63 and (2) simultaneously considers the structural and functional connectivity of a targeted region.
64 In 26 neurosurgical patients with indwelling electrodes, we stimulated different regions of the
65 medial temporal lobe (MTL) and asked whether low-frequency coherence predicted modulations
66 of theta power in distributed cortical regions. We showed that coherence was mostly predictive
67 of theta modulation when stimulation occurred in or near a white matter tract, but in those cases,
68 stimulation could evoke sustained increases in theta power even in distant regions. With this
69 initial finding in hand, we expanded our paradigm to consider additional measures of functional
70 connectivity and evoked power at higher frequencies. We principally considered the amplitude
71 envelope of the high-frequency broadband signal (HFB; 50-200 Hz)²⁰, shown to correlate with
72 the resting-state fMRI BOLD correlations that are widely used in network neuroscience²⁰⁻²³. We
73 demonstrated that while low-frequency coherence accurately predicts increases in low-frequency
74 power, HFB-based networks can explain decreases of HFB power. Taken together, functional
75 connectivity can predict the widespread changes in local spectral power induced by direct
76 electrical stimulation of the MTL.

77

78 **Results**

79 **Calculating a theta modulation index**

80 To determine how direct cortical stimulation propagates through brain networks, we
81 collected intracranial EEG (iEEG) data from 26 patients undergoing clinical monitoring for

82 seizures. Subjects rested passively in their hospital bed while we applied bipolar macroelectrode
83 stimulation at varying frequencies (10-200 Hz) and amplitudes (0.25 to 1.5 mA) to MTL depth
84 electrodes (see online Methods for details). Rectangular stimulation pulses were delivered for
85 500 ms, followed by a 3-second inter-stimulation interval (Figure 1A-C). Each subject received
86 at least 240 stimulation events (“trials”) at 1-8 distinct sites in MTL gray or white matter (mean
87 2.7 sites; see Supplementary Table 2 for stimulation locations). During a separate recording
88 session in which no stimulation occurred, for each subject we computed resting networks of low-
89 frequency (5-13 Hz) coherence, motivated by prior literature that shows robust iEEG functional
90 connectivity at low frequencies²⁴⁻²⁷. These networks reflect correlated low-frequency activity
91 between all possible pairs of electrodes in a subject, during a period when subjects are passively
92 waiting for a task to begin (Figure 2A).

93 For each stimulation trial, we computed theta power (5-8 Hz) in 900 ms windows before
94 and after each 500 ms stimulation event, and compared the pre- vs. post-stimulation power
95 across all trials with a paired *t*-test (Figure 1D). Next, we used linear regression to correlate the
96 strength of a stimulation site’s network connectivity to a recording electrode with the power *t*-
97 statistic at that electrode (Figure 2A-D). We included absolute distance as a factor in our
98 regression, to only consider how connectivity relates to stimulation beyond the brain’s tendency
99 to densely connect nearby regions²⁸. The result is a model coefficient that indicates, independent
100 of distance, the degree to which functional connectivity predicts stimulation-induced change in
101 theta power at a recording site. The regression was repeated using permuted connectivity/evoked
102 power relationships to generate a null distribution of model coefficients against which the true
103 coefficient is compared. We refer to the resulting z-score as the “theta modulation index,” or

104 TMI. High TMIs indicate functional network connectivity predicts observable stimulation-
105 related change in theta power at distant sites.

106 **TMI is correlated with proximity to white matter**

107 At a group level of stimulation sites, TMI was significantly greater than zero (1-sample t -
108 test, $t(71) = 4.0$, $P = 0.0002$; Figure 3A), indicating that stimulation in the MTL tends to evoke
109 network-driven change in theta power in distant regions. However, we noted substantial
110 heterogeneity between stimulation sites, with some showing little or no ability to modulate
111 network-wide theta activity, as reflected by TMIs near zero. To explain this heterogeneity, we
112 hypothesized that, as earlier work demonstrated^{13,15,29}, structural connections (i.e. white matter
113 tracts) may be key to the propagation of electrical stimulation throughout the brain.

114 To test whether structural connections play a role in stimulation propagation, we asked
115 whether TMI was correlated with the proximity of a stimulation site to white matter. If these
116 measures are correlated, it would indicate that functional connectivity is predictive of physiology
117 only insofar as white matter tracts are accessible. We binned stimulation sites according to
118 whether they were placed in gray matter ($n = 32$, lower 50th percentile of distances to white
119 matter), near white matter ($n = 33$, upper 50th percentile of distances to white matter), or within
120 white matter ($n = 7$, manually identified by a neuroradiologist; Figure 3A; see Supplementary
121 Figure 1 for anatomical placement of each white matter target). We found that TMI was
122 significantly increasing with white matter placement, relative to a permuted distribution
123 (permuted $P < 0.001$; Figure 3B). The TMI for gray matter sites was not significantly different
124 than zero (1-sample t -test, $t(31) = 1.4$, $P = 0.18$), while TMI for sites near or in white matter was
125 significant ($P < 0.05$). This relationship holds in a Pearson correlation agnostic to any electrode
126 categorization ($r = 0.33$, $P = 0.005$; Supplementary Figure 2). This finding does not mean gray

127 matter stimulation fails to induce theta activity, but it does suggest that stimulation far from
128 white matter tracts may result in theta activity that is uncorrelated with connectivity to remote
129 sites.

130 To account for the possibility that the theta response is sensitive to the pulse frequency or
131 amplitude of stimulation, we asked whether TMI differed in accordance with stimulation
132 parameters. Across all stimulation sites, TMI was marginally greater for trials delivered at a
133 subject's maximum versus minimum amplitude (paired T-test, $T(71) = 1.91$, $P = 0.061$;
134 Supplemental Figure 2A), but no difference was noted across pulse frequencies delivered at 10
135 Hz, 50 Hz, and 200 Hz (repeated measures ANOVA, $F = 0.16$, $P = 0.85$; Supplemental Figure
136 2B). Additionally, raw evoked power was significantly greater for maximum amplitude
137 stimulation ($T(71) = 3.52$, $P = 0.0008$), but did not reliably differ across pulse frequencies ($F =$
138 0.26 , $P = 0.77$; Supplemental Figure 2B). For the remainder of this study, all analyses consider
139 stimulation events aggregated across amplitudes and frequencies.

140 Taken together, these results show that direct electrical stimulation of the MTL can
141 induce spectral power changes across a distributed network of regions, particularly if stimulation
142 occurs in or proximal to white matter. When this occurred, we discovered that functional low-
143 frequency coherence is predictive of where stimulation-related modulations in theta power are
144 observed.

145 **Network properties of MTL stimulation**

146 Having shown that stimulation in or near white matter sites induces distributed changes
147 in theta power, we next sought to characterize the directionality of change. Specifically, high
148 TMIs could be caused by increases in theta power at electrodes with strong functional
149 connectivity to the stimulation target, or decreases in theta power at electrodes with weak

150 connectivity to the stimulation target. To distinguish between these possibilities, we further
151 examined theta power changes in detail among the 16 stimulation sites that exhibited
152 individually significant ($P < 0.05$) TMI (see Supplementary Table 1 for statistics and anatomical
153 placement of each significant site). In this subset, we measured the average pre- vs. post-
154 stimulation theta power at the five electrodes with the strongest functional connectivity to the
155 stimulation site (controlled for distance), and the five electrodes with the weakest functional
156 connectivity. At strongly-connected sites, theta power change was significantly positive (1-
157 sample t -test, $t(15) = 5.6$, $P = 4.0 \times 10^{-5}$) and significantly greater than power change at weakly-
158 connected sites (paired t -test, $t(15) = 6.03$, $P = 1.7 \times 10^{-5}$; Figure 4B). No significant power
159 change was observed at sites with weak functional connectivity (1-sample t -test, $t(15) = 1.5$, $P =$
160 0.15). Notably, we observed that of the 16 significant sites analyzed here, 15 were placed in or
161 near white matter. We conclude that stimulation causes increased theta power at strongly-
162 connected sites and little to no change in power at weakly-connected sites.

163 Principles of network control theory suggest a relation between the connectivity profile –
164 or network topology – of a stimulation site and the ensuing change in brain activity. Network
165 “hubs,” or regions with strong connectivity to the rest of the brain, exert differential effects on
166 overall brain activity versus non-hubs, or regions with strong connections to only a few areas^{18,30}.
167 To directly test whether stimulation propagates differently from hub regions, we asked whether
168 stimulation-induced theta power correlated with the functional “hubness” of a stimulation site.
169 We again took our measure of stimulation-induced activity to be the theta power change at the 5
170 recording sites with the strongest functional connectivity to the stimulation site, and tested this
171 metric against the node strength of a stimulation site, an indicator of hubness. For this analysis,
172 we considered all stimulation sites in or near white matter, ($n = 40$) as these groups both

173 exhibited significant TMI (see Figure 3B). When weak hubs (lower tercile of hub scores; $n = 13$)
174 were stimulated, power change at connected recording sites was significantly greater than zero
175 (1-sample t-test, $t(12) = 3.6$, $P = 0.003$), but stimulation at strong hubs (upper tercile; $n = 14$)
176 evoked no significant power modulation ($t(13) = 0.15$, $P = 0.87$; Figure 4D). While
177 counterintuitive, this result could suggest that stimulation at a site with many connections may
178 disperse or blunt the effect of perturbation, yielding lesser activation in downstream regions.
179 Alternatively, hub stimulation does evoke widespread changes in brain activity, but these
180 changes tend to be outside the theta band assessed here.

181 Our choice of low-frequency (5-13 Hz) functional connectivity as the basis for predicting
182 distributed changes in theta power was motivated by prior studies that have shown strong,
183 cognitively-relevant connectivity at low frequencies particularly the theta and alpha bands ^{24,25,27}.
184 However, others have noted significant inter-regional connectivity in the beta and gamma bands
185 ³¹. As our study presented a unique opportunity to examine the causal nature of functional
186 connectivity, we asked whether functional connectivity in other frequency bands is also
187 predictive of downstream power modulations. Among all MTL electrodes placed in or near white
188 matter ($n = 40$), we asked whether TMI was significant for networks constructed from any
189 frequency to a maximum of 50 Hz. No frequencies outside the alpha/theta bands exhibited
190 significant group-level TMIs, after correction for multiple comparisons ($P < 0.05$, Benjamini-
191 Hochberg correction; Figure 4E). This demonstrates that functional networks constructed from
192 high frequencies (> 13 Hz) are not predictive of stimulation-induced theta activity.

193 **Alternative measures of connectivity**

194 Functional connectivity is a broad domain, generally referring to an array of measures
195 that fundamentally reflect timeseries correlations. In addition to the phase-based measures (i.e.

196 spectral coherence) used here, other correlations have also been shown to robustly capture inter-
197 regional functional dynamics in the human brain. Of particular utility in iEEG investigations is
198 the amplitude envelope of high-frequency broadband (HFB; 50-200 Hz), shown to reflect
199 neuronal population spiking activity³² and correlated with fMRI BOLD activation³³. The slow (<
200 1 Hz) fluctuations of this signal have also been shown to correlate with resting-state functional
201 connectivity (rsfMRI)^{23,34,35}. It has recently been demonstrated that stimulation perturbs brain
202 networks in accordance with measures of functional connectedness, including a modulation of
203 remote cortical excitability^{14,36}.

204 We therefore sought to determine whether these established measures of intrinsic
205 functional connectivity – HFB amplitude envelope correlation and rsfMRI connectivity – also
206 predicted the location of evoked theta power. To do this, we replicated our procedure for
207 computing TMI, but used HFB amplitude envelope correlation or atlas-based rsfMRI
208 connectivity as predictor variables (see Methods for details; see Figure 5A for example
209 adjacency matrices). As with the low-frequency coherence networks, the result is a z-scored
210 statistic (TMI) that reflects the degree to which a functional network predicts remote changes in
211 theta power. We emphasize that HFB envelope networks – though based on extracted high-
212 frequency power – are wholly distinct from 30+ Hz coherence networks assessed earlier; the
213 former is a correlation of the slow variation in a power timeseries, while the latter is a measure of
214 high-frequency phase consistency between two signals.

215 Though rsfMRI and HFB connectivity measures qualitatively recapitulated our earlier
216 finding – TMI increases with closeness to white matter – their ability to predict downstream
217 changes in evoked theta power was not significant across all stimulation sites (HFB-connectivity,
218 $T(71) = 1.07, P = 0.28$; atlas rsfMRI, $T(49) = 0.17, P = 0.87$; Figure 5B). We note that slightly

219 fewer stimulation sites were available for the rsfMRI analysis ($n = 50$), due to subjects where
220 atlas-based measures could not be estimated for a sufficient number of electrodes (see Methods
221 for details).

222 Though HFB envelope and rsfMRI connectivity did not strongly replicate our finding of
223 significant TMI using low-frequency coherence, several factors could account for this
224 discrepancy. First, stimulation within the unique architecture of the MTL may propagate
225 differently than the cortical surface stimulation used in many prior studies – it is possible that, at
226 the cortical surface, HFB/rsfMRI connectivity is better predictive of stimulation effects than low-
227 frequency coherence. Second, different measures of connectivity may differentially predict
228 different kinds of evoked responses. Low-frequency coherence successfully predicts low-
229 frequency power, but may fail to accurately predict modulations at higher frequencies.

230 **Evoked responses at higher frequencies**

231 While our choice to examine the effect of stimulation on theta frequencies was
232 theoretically motivated by a vast literature implicating theta oscillations and cognition³⁷, activity
233 in the HFB range is a useful marker of population neural activity³⁸, and cognitively-relevant
234 oscillatory dynamics are also observed in the alpha, beta, gamma bands (9-13 Hz, 15-25 Hz, 30-
235 60 Hz, respectively). To account for the possibility that stimulation evokes activity in these
236 higher frequency bands, we extended our analysis to consider the correlation between low-
237 frequency coherence and induced power in alpha/beta, gamma, and HFB ranges. Furthermore, to
238 address the possibility that HFB-based connectivity networks (Figure 5A) better predict induced
239 local HFB power, we asked about the correlation between HFB envelope connectivity and
240 induced power across frequency ranges (see Methods for details).

241 We first assessed whether stimulation evoked any detectable modulation of power in the
242 alpha/beta, gamma, and HFB bands, regardless of relationship to connectivity. To do this, we
243 averaged the pre-versus-post stimulation T-statistic across all electrodes in each subject's brain,
244 for each frequency band. The result is an average T-statistic reflecting the stimulation-evoked
245 whole-brain change in power at each frequency band. Across all stimulation sites, stimulation
246 significantly increases power in the theta, alpha/beta, and gamma bands, but significantly
247 decreases power in the HFB range (1-sample T-test, FDR-corrected $P < 0.05$; Figure 6A-C).
248 However, the power response to stimulation is not uniform across electrodes within a subject; for
249 electrodes that exhibited a strong theta response, evoked changes were weaker at higher
250 frequencies (Figure 6D), indicating a theta-specific effect.

251 Given that stimulation does evoke changes in spectral power beyond the theta range, we
252 next asked whether functional connectivity networks predicted these changes. Corrected for
253 multiple comparisons, low-frequency (5-13 Hz) coherence networks only correlated with evoked
254 power in the theta range ($T(71) = 4.18$, FDR-corrected $P < 0.05$, Figure 7A). On average, HFB
255 envelope connectivity did not significantly predict power modulation at any frequency band.
256 However, given our earlier finding of decreases in power in the HFB band (Figure 6), we
257 hypothesized that a null average effect was obscuring heterogenous – but individually significant
258 – responses to stimulation. In other words, for specific subjects, HFB envelope networks could
259 predict increases or decreases in HFB power and yield significant correlations in positive or
260 negative directions. Indeed, HFB functional connectivity significantly predicted HFB power
261 decreases for 7 stimulation sites and power increases for 3 stimulation sites, a total count that
262 significantly exceeds the expected false positive rate (binomial test, FDR-corrected $P = 0.009$;
263 Figure 7B).

264 Taken together, functional connectivity measured by low-frequency coherence
265 significantly predicts stimulation-evoked power in the theta band, but not induced power at
266 higher frequencies. On average, HFB envelope networks do not significantly correlate with
267 evoked changes in any frequency band, even HFB power itself. However, the dynamics of
268 stimulation appear to be more complex in this high-frequency band; HFB power is often
269 decreased by stimulation – unlike the theta response – and for a significant number of
270 stimulation sites, both low-frequency coherence and HFB functional connectivity predict where
271 in the brain such decreases are observed.

272

273 **Discussion**

274 We set out to test a fundamentally simple hypothesis: Do functional connections in the
275 brain predict how focal electrical stimulation flows from one region to another? Though critical
276 to the future of brain stimulation and therapeutic development, this hypothesis has not seen
277 rigorous testing. Prior studies indicate that connectivity plays a role in how stimulation events
278 perturb distant brain regions^{13,14,16,29}, but fundamental assumptions of graph-theoretic models
279 remain untested¹⁷. More broadly, no prior studies have addressed whether iEEG-based functional
280 connectivity indicates anything about causal relationships in the brain, or whether is it merely a
281 correlative measure. In this manuscript, we specifically tested a hypothesis about the effects of
282 stimulation on theta power, given an especially rich literature showing the cognitive relevance of
283 theta oscillations^{39–42}. To account for possible dynamics outside this range, we extended several
284 key analyses to alpha/beta, gamma, and high-frequency broadband power, and further considered
285 whether functionally-derived measures of connectivity better capture the effects of stimulation-
286 induced power.

287 We discovered that (1) modulation of theta power is correlated with functional
288 connectivity, particularly if stimulation occurred in or near white matter, (2) stronger functional
289 connections yield greater theta power increases, (3) low-frequency coherence better predicts
290 downstream increases in theta power than HFB envelope or rsfMRI networks, and (4) in specific
291 cases HFB envelope networks do succeed in predicting modulations in HFB power. These results
292 suggest that stimulation evokes a heterogenous mixture of effects across frequencies, and that
293 functional networks may best predict the frequency on which they were based.

294 The meaning of functional connectivity is a subject of considerable debate. Correlated
295 activity between two parts of the brain may reflect direct connection between the two, an indirect
296 connection through a third region, or the activity of a third region independently driving activity
297 in each⁴³. Though most neuroscientists are aware of such limitations, functional connectivity is
298 often implicitly treated as a measure of causality nonetheless. Our use of targeted stimulation
299 allowed us to test whether this implicit assumption is true. Our results generally support the idea
300 that functional connectivity indicates causal relations in the brain; when stimulation occurs in or
301 near white matter, we could predict where power changes would occur based on distance-
302 independent measures of low-frequency functional connections. This finding aligns with
303 observations that intrinsic functional connectivity in MRI is constrained by white matter
304 anatomy⁴⁴. However, substantial variance in power modulation remained unexplained by
305 connectivity, and we also showed that propagation of gray matter stimulation – still rich with
306 functional connections – cannot be predicted in the same way.

307 HFB amplitude envelope networks and atlas-based rsfMRI networks failed to strongly
308 predict remote changes in theta power. However, earlier reports suggest that these functionally-
309 relevant measures do correlate with CCEPs and changes in cortical excitability^{14,36}. To explain

310 this discrepancy, we note that there are several key differences between those reports and the
311 present study. First, we solely examined the effect of stimulation within the medial temporal
312 lobe, which has a distinct architecture that may affect how stimulation propagates to other
313 regions – effects of stimulation at the cortical surface, as in prior studies, could differ markedly.
314 Relatedly, we used stimulation amplitudes that are lower than those typically used at the cortical
315 surface (< 2mA versus > 4mA). Finally, while HFB envelope networks did not successfully
316 predict remote changes in theta power, they more accurately correlated with remote decreases in
317 HFB power – it is possible that networks based on measures of cortical activation are better
318 predictors of how stimulation affects those same measures.

319 In this study, we also assessed the relationship between stimulation and the network
320 topology of a targeted region. Specifically, we asked whether the downstream effects of
321 stimulation differed between hubs and non-hubs, reflecting regions that are richly or sparsely
322 connected. Counterintuitively, we found that stimulation of non-hubs yielded greater increases in
323 theta power at downstream sites. It is possible that (1) hub stimulation does result in greater
324 distributed power changes, but outside the theta band, or (2) hub stimulation results in a dispersal
325 or blunting effect, causing widespread change but limiting the magnitude of the effect at any
326 single downstream site. Such a result is plausible if there is an interaction between the underlying
327 brain structure and the effect of stimulation – it has been demonstrated that stimulation less
328 effectively activates large-diameter axons, for example⁴⁵. Furthermore, principles of network
329 control theory postulate that stimulation of sparsely connected regions can be efficacious for
330 moving the brain to “difficult-to-reach” states, or states that require significant cognitive effort to
331 achieve^{17,18,30}. However, the mapping between spectral power and “brain states” in a cognitive

332 sense remains unclear; further empirical and theoretical work should aim to clarify how control
333 theoretic predictions can be tested with common intracranial techniques.

334 The findings from this study could be extended in several ways. A recent study by Keller,
335 et al. (2018) asked whether a multivariate model could predict how direct brain stimulation alters
336 remote cortical excitability³⁶. A similar approach could be adopted with these data, wherein
337 multimodal measures of connectivity – e.g. coherence, HFB-envelope, and white matter
338 proximity – could be used to predict the stimulation response across locations and frequency
339 ranges. Such an approach could reveal relationships that were obscured by the univariate
340 methods in this manuscript; gray matter targets, for instance, may induce widespread,
341 connectivity-related changes in specific frequencies that are predictable by a weighted
342 combination of functional networks. Additionally, our study as designed was agnostic to the
343 directionality of induced effects; especially in the setting of direct white matter stimulation, we
344 expect that our results reflect a combination of prodromic and antidromic propagation. In other
345 words, stimulation of MTL structures is potentially inducing activity in input and output regions,
346 though the undirected measures of functional connectivity used here are unable to tease those
347 effects apart.

348 We solely analyzed stimulation through the lens of changes in brain physiology.
349 However, with an eye towards the eventual therapeutic use of stimulation, the results here begin
350 to bridge prior studies of stimulation and behavior with underlying neural mechanisms. A recent
351 study reported decreases in episodic memory performance during stimulation at certain times,
352 associated with increases in cortical theta power³. Additionally, memory performance was noted
353 to increase with theta-burst stimulation of the perforant path, a major white matter tract of the
354 MTL⁴. Deep brain stimulation targeted to white matter tracts has also been shown to improve

355 outcomes in treatment-resistant depression⁹. Collectively, these findings are supported by the
356 results here – white matter stimulation appears to evoke remote increases in neural activity. Few
357 studies have deeply examined stimulation-induced changes in physiology with behavioral
358 enhancement, though our approach outlined here enables us to do exactly that in future work.

359 Here we demonstrated that functional connections in the human brain inform how
360 stimulation evokes remote changes in neural activity. This is powerful new evidence that, even in
361 the absence of knowledge about an individual’s structural connectome, functional connectivity
362 reflects causality in the brain – a finding with significant implications for how neuroscientists
363 interpret inter-regional correlations of neural activity. Furthermore, by showing that stimulation-
364 evoked changes interact with the functional hubness of a targeted site, we provided a critical data
365 point for the application of network control theory to real-world brain dynamics.

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372 **References**

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498 brain image registration. *Neuroimage* **54**, 2033–2044 (2011).
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502 **Supplementary Information** is linked to the online version of the paper at www.nature.com.

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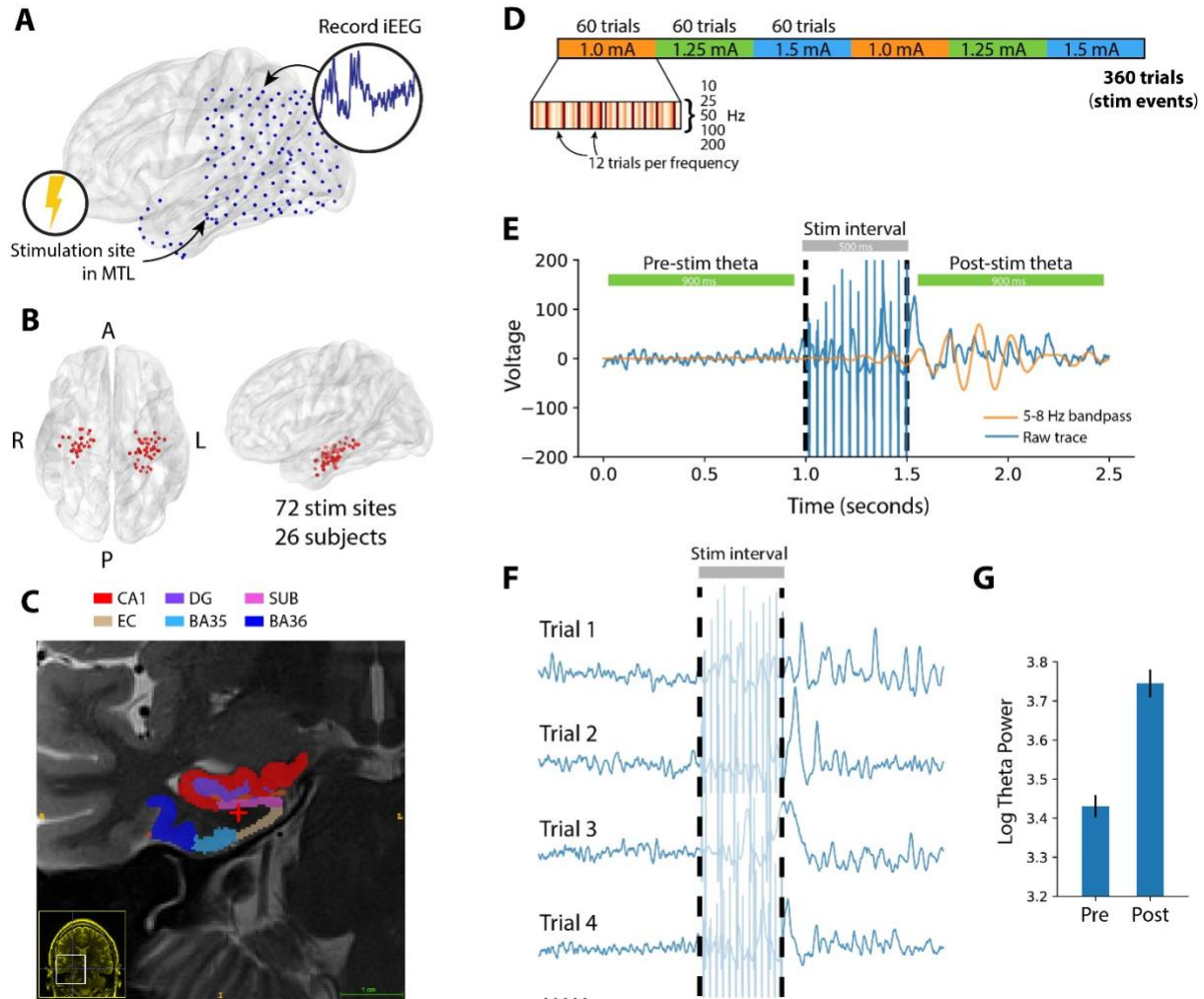
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517 executed all data analyses, J.K. analyzed data, and E.S. wrote the paper. J.S., R. Gorniak, S. Das.
518 performed anatomical localization of depth electrodes. M.S., G.W., B.L., R. Gross., B.J., C.I., K.Z., S.
519 Sheth, and S. Seger recruited subjects, collected data, and performed clinical duties associated with data
520 collection including neurosurgical procedures or patient monitoring.

521 **Data Availability:** Raw electrophysiological data used in this study is freely available at
522 http://memory.psych.upenn.edu/Electrophysiological_Data

523 **Competing Interests**

524 Michael J. Kahana and Daniel S. Rizzuto have started a company, Nia Therapeutics, LLC (“Nia”),
525 intended to develop and commercialize brain stimulation therapies for memory restoration. Each of them
526 holds more than 5% equity interest in Nia.



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Figure 1. Comparison of pre- vs. post-stimulation theta (5-8 Hz) power in an example subject. (A) Each of 26 subjects received a series of 500 ms bipolar stimulation events, at 1-7 sites within the MTL; an example subject schematic is shown here. (B) Anatomical distribution of all MTL stimulation sites in the 26-subject dataset. (C) T2 MRI and MTL subregion segmentation for an example subject. Stimulation location, in white matter, is indicated at the red cross. See Supplemental Figure 1 for subregion labels. (D) Schematic of a typical stimulation session. Each stimulation site receives stimulation at three amplitudes (within the 0.25-2.00 mA range) and five pulse frequencies (50-200 Hz; see Methods for details). During each session, amplitudes are delivered in 60-trial blocks, within which 12 stimulations are delivered at each

538 frequency. For the main results in this manuscript, effects are aggregated across all stimulation
539 parameters; see Supplemental Figure 2 for analysis of stimulation amplitude and frequency. **(E)**
540 Using the multitaper method, theta power (5-8 Hz) was measured in 900 ms windows preceding
541 and following each stimulation event, with 50 ms buffers before and after stimulation. In an
542 example stimulation event, the 5-8 Hz bandpass signal (orange) is overlaid on the raw bipolar
543 signal (blue), to emphasize a change in pre- vs. post-stimulation theta power. **(F)** Theta power is
544 extracted in the pre- and post-stimulation intervals for at least 240 events (“trials”) per
545 stimulation site. **(G)** The log-transformed theta power is aggregated for all pre- and post-
546 stimulation intervals separately, for later statistical comparison (Fig. 2).

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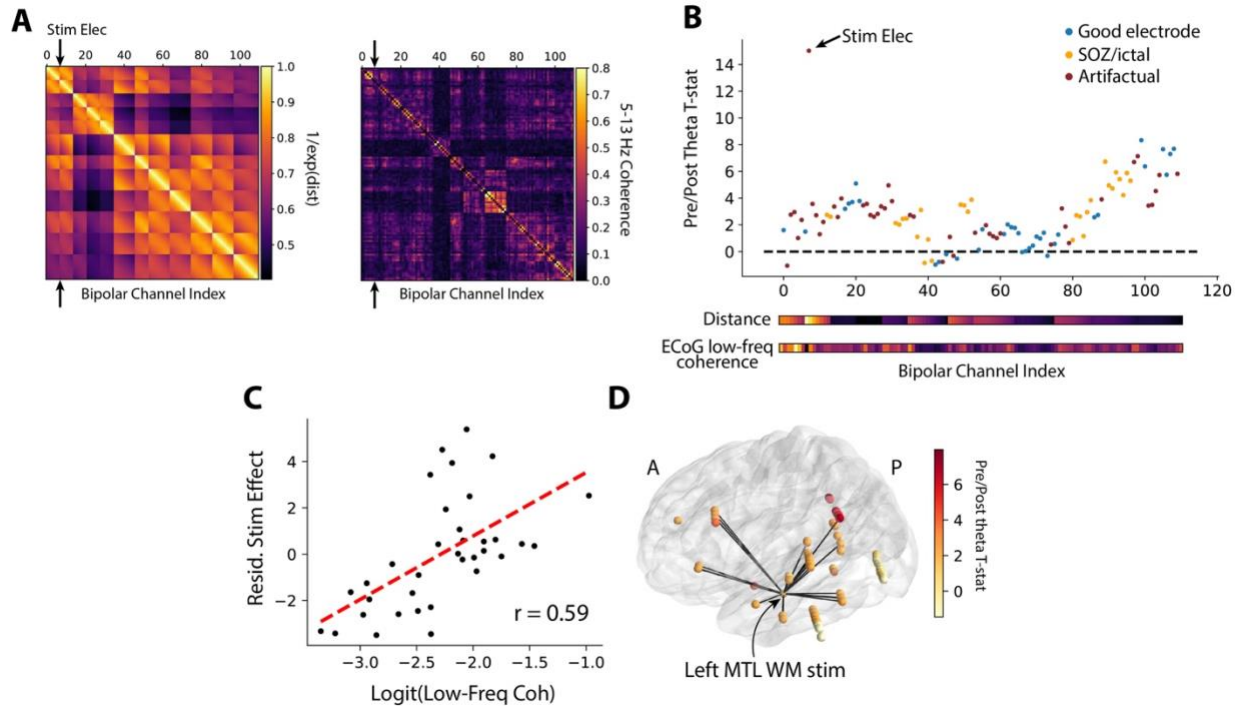
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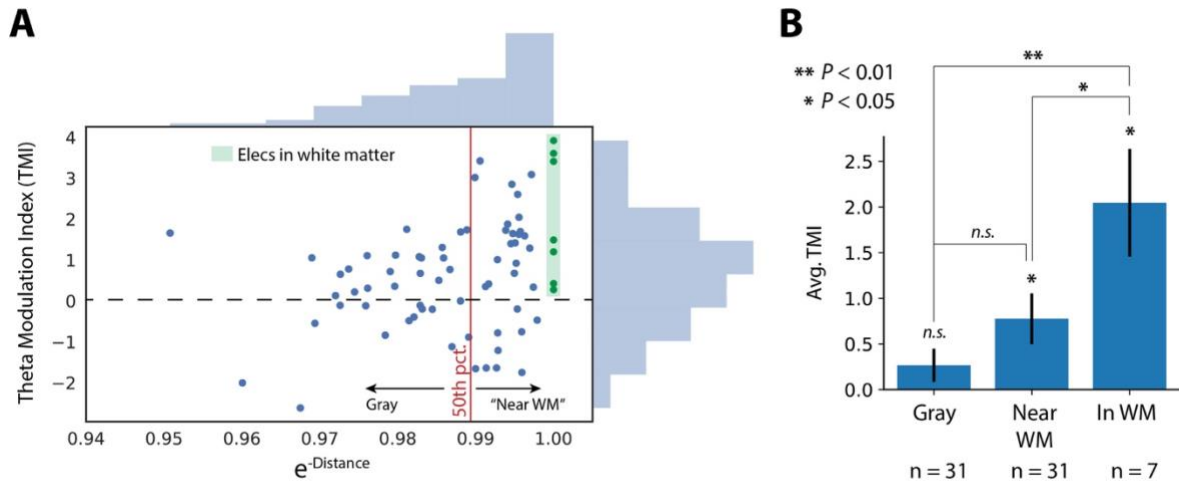


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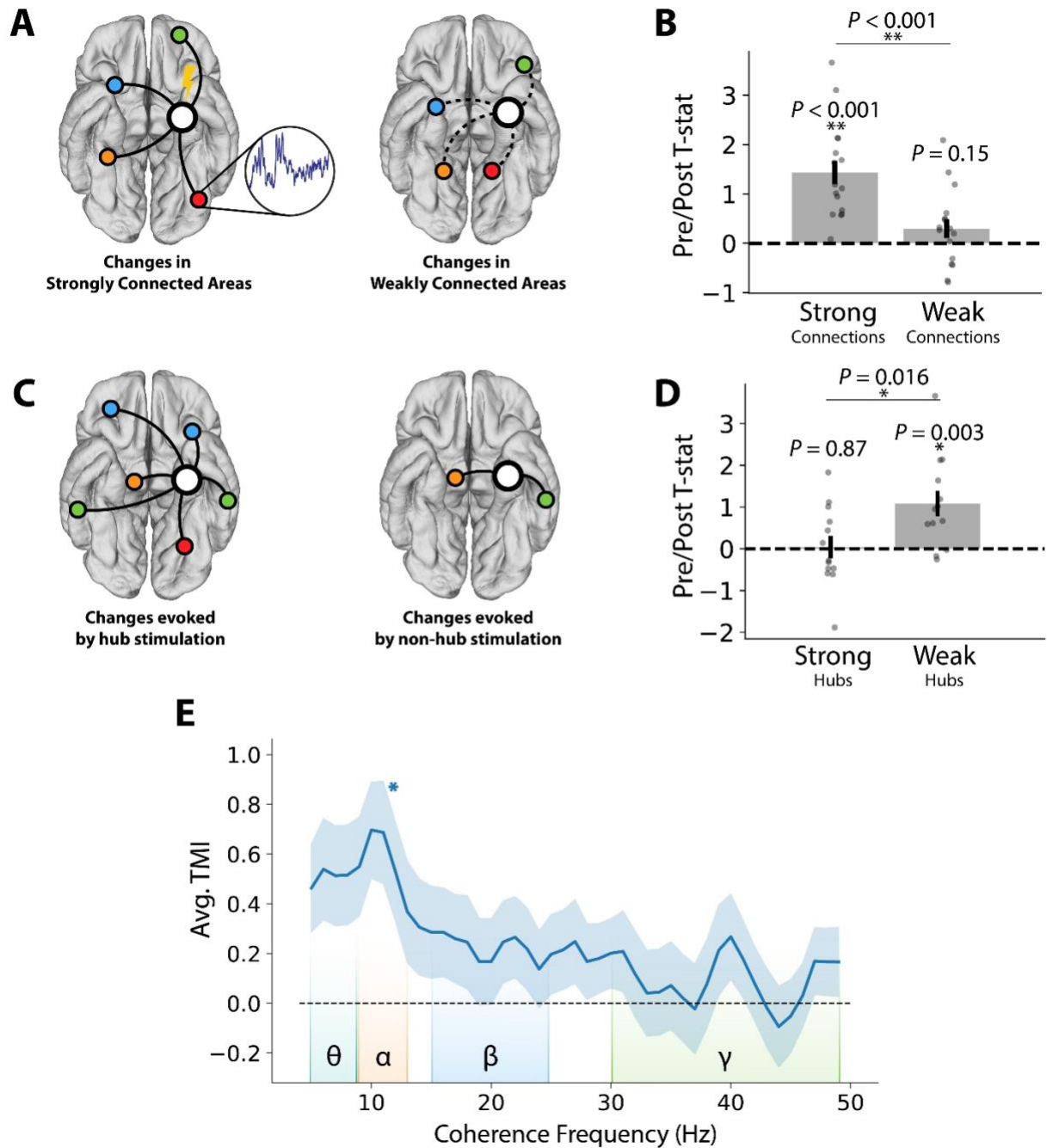
561 **Figure 2. Method for determining theta modulation index (TMI).** (A) For each subject,
562 Euclidean distances (left matrix) and functional connectivity (right matrix) are measured for all
563 possible electrode pairs. Distances are linearized as $e^{-(\text{distance})}$, with 1.0 representing no separation
564 between two electrodes. Functional connectivity is the averaged 5-13 Hz multitaper coherence in
565 1-second windows extracted from a baseline period. (B) Pre- and post-stimulation theta power
566 (Fig. 1C) is compared with a paired t-test to generate a t -statistic for each electrode. Electrodes
567 are excluded from analysis if they exhibited significant post-stimulation artifact (red, see
568 Methods for details) or were placed in the seizure onset zone or exhibit high inter-ictal spiking
569 (orange). (C) Multiple linear regression is used to correlate the logit-transformed functional
570 connectivity (between a recording electrode and the stimulation electrode) with the power t -
571 statistic, independent of distance. To demonstrate this, the distance-residualized t -statistic (“Stim
572 Effect”) is plotted against functional connectivity in the example subject. The z-scored version of
573 this correlation is referred to as the “Theta Modulation Index,” or TMI. (D) Rendering of the

574 power t -statistic as color on each electrode in the example subject, plotted with the top 10% of
575 functional connections to the stim electrode (red lines).

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 607 **Figure 3. Proximity to white matter predicts TMI.** (A) Correlation between a stimulation
 608 site's distance from nearest white matter with the site's TMI. The 50th percentile of white matter
 609 distances divides sites classified as "gray matter" versus "near white matter." Stimulated contacts
 610 in white matter are highlighted in green. See Supplemental Figure 2 for the Pearson correlation
 611 of these data ($r = 0.33$, $P = 0.005$). (B) TMI increases with closeness to white matter, as
 612 determined by a permutation test ($P < 0.001$, see Methods) and by noting that TMIs for sites in
 613 or near white matter are significantly greater than zero (1-sample t -test, $P < 0.05$) while gray
 614 matter sites are not ($P = 0.15$). Electrodes placed in white matter have greater TMIs than
 615 electrodes near white matter (2-sample t -test, $P < 0.05$) or gray matter ($P < 0.01$). Error bars
 616 show ± 1 SEM; * $P < 0.05$; ** $P < 0.01$.

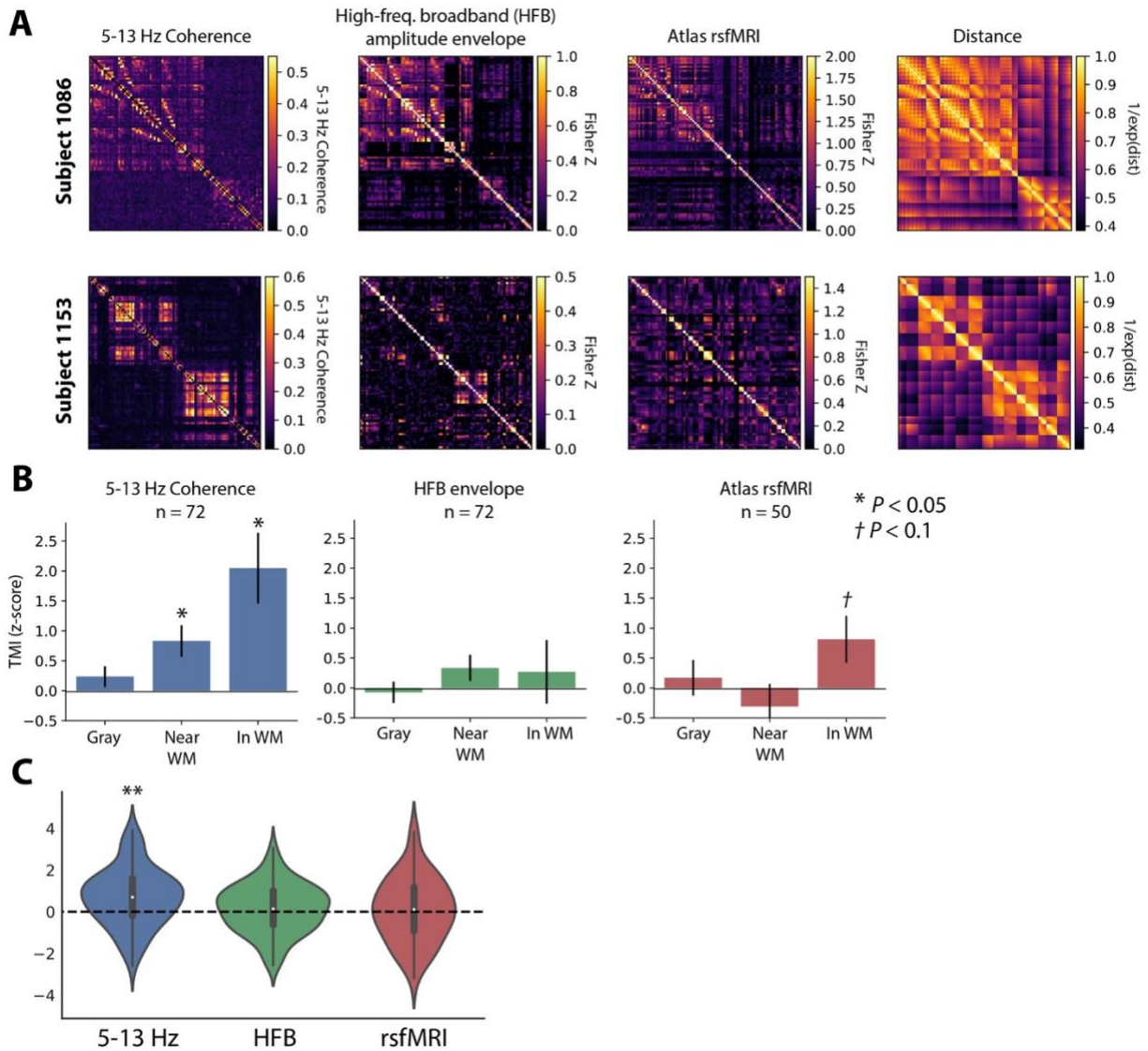


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619 **Figure 4. Network properties of stimulation-induced theta.** (A) Schematic of a stimulation
620 site and its most strongly-connected areas (left) or weakly-connected areas (right). (B) For each
621 of 16 stimulation sites with significant TMI ($P < 0.05$), the average post- vs. pre-stimulation
622 theta T-statistic is computed for the five strongest-connected electrodes and the five weakest-
623 connected electrodes (controlled for distance). Strongly-connected regions are typically areas of

624 the lateral temporal, prefrontal, or inferior parietal cortices. Changes at strongly-connected
625 recording sites are significantly greater than changes at weakly-connected sites (paired *t*-test,
626 $t(15) = 6.03, P = 1.7 \times 10^{-5}$). **(C)** Schematic of a hub-like stimulation site (left) and a non-hub
627 stimulation site (right). Hub scores are calculated as the node strength, or average of all
628 connection weights to a given electrode. **(D)** For each of 40 stimulation sites in or near white
629 matter, the average post- vs. pre-stimulation theta T-statistic is computed for the five strongest-
630 connected recording electrodes. Stimulation of a weak hub (lower tercile of hub scores, $n = 13$)
631 yields significantly greater change in connected regions than stimulation of a strong hub (upper
632 tercile of hub scores, $n = 14$) (2-sample *t*-test, $P = 0.016$). **(E)** Average TMI across all in or near-
633 white matter stimulation sites, as a function of functional connectivity frequency. TMI is greatest
634 for networks constructed from theta or alpha coherence (5-13 Hz). Corrected for multiple
635 comparisons across all frequencies, TMI is significantly greater than zero at 11 Hz. Error bars
636 show +/- 1 SEM; * $P < 0.05$; ** $P < 0.001$.

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659 **Figure 5. Alternative measures of connectivity.** (A) Example adjacency matrices for two
660 subjects, reflecting functional connectivity from low-frequency (5-13 Hz) coherence, correlated
661 high-frequency broadband envelope (HFB; 50-200 Hz), and atlas-based resting state fMRI
662 (rsfMRI). Matrices are organized as in Figure 2A. For reference, adjacency matrices of linearized
663 Euclidean distance are shown at the far right. Colormap ranges are selected to visually
664 emphasize network structure. Values in the HFB envelope and rsfMRI networks are Fisher z-
665 transformed correlation coefficients. (B) TMI is computed as in Figures 2-3, using adjacency

666 matrices for each of the three measures. TMI was binned by distance from white matter,
667 organized as in Figure 3B. In addition to significant TMI in and near white matter using 5-13 Hz
668 coherence networks (* $P < 0.05$), we noted marginally significant TMI in white matter using
669 atlas-based rsfMRI networks ($P < 0.1$). (C) Distribution of TMI for all stimulation electrodes
670 regardless of distance from white matter. TMI is significantly greater at the group level for 5-13
671 Hz coherence ($P = 0.0002$). Note that the total count of stimulation electrodes is lower for
672 rsfMRI connectivity ($n = 50$) analyses, due to subjects where atlas-based rsfMRI could not be
673 extracted for a stimulation electrode; see Methods for details.

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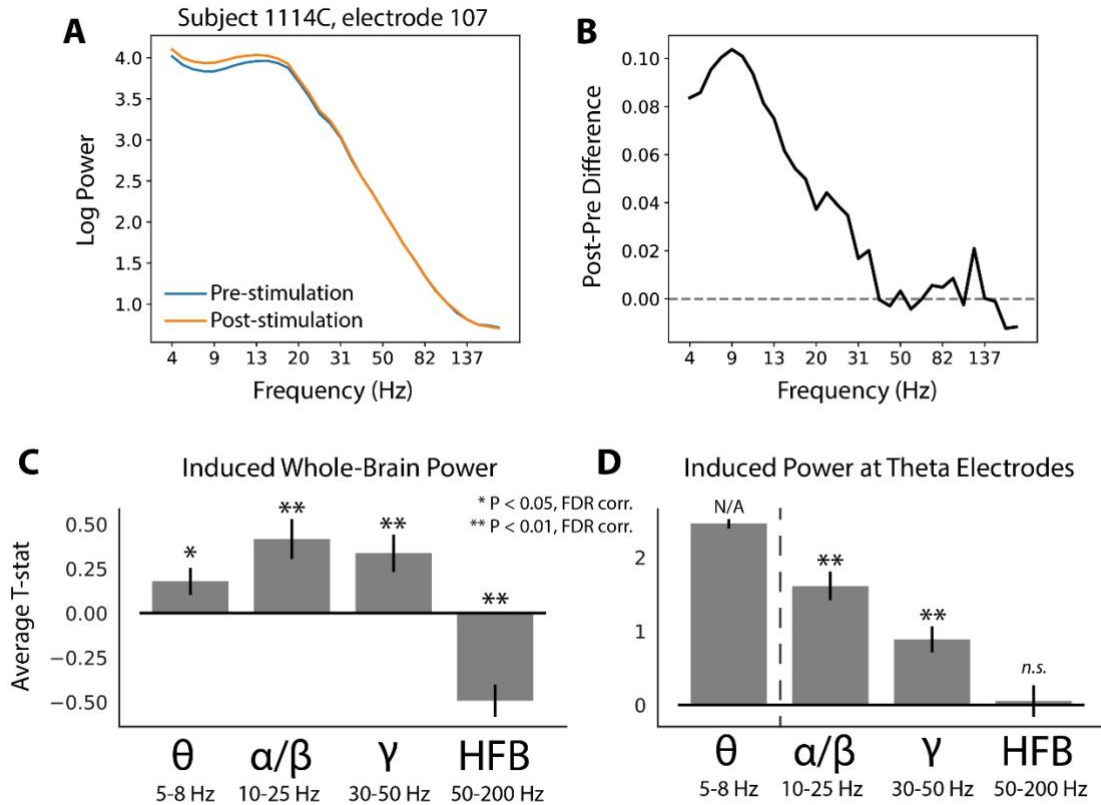
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 690 **Figure 6. Stimulation-induced power across frequency bands.** (A) Stimulation-induced
 691 power spectrogram for an example electrode from a single subject (stimulation in left MTL white
 692 matter, recording electrode in left inferior parietal cortex). (B) Post-minus-pre stimulation
 693 difference in power from the electrode in (A). (C) Whole-brain stimulation-induced power was
 694 measured by computing a T-statistic on the pre- vs. post-stimulation spectral power at each
 695 electrode in a subject, and then averaging across electrodes to get an estimate of whole-brain
 696 change in power. On average, stimulation elevated whole-brain power in the theta (5-8 Hz),
 697 alpha/beta (10-25 Hz), and gamma (30-50 Hz) bands (1-sample T-test, FDR-corrected $P < 0.05$).
 698 Stimulation decreased power in the HFB range (50-200 Hz). (B) For each subject/stimulation
 699 site, electrodes were classified by whether they exhibited a significant ($T > 2$) change in theta
 700 power induced by stimulation (at least 1 theta-responsive electrode was found for 47 of the 72
 701 stimulation sites). The stimulation-induced change at higher frequencies was computed for this

702 subset of electrodes, to determine whether the power response was specific to theta. Across all
703 subjects/stimulation sites, significant responses were also observed in the alpha/beta and gamma
704 range at theta-responsive electrodes, but no effect was observed in the HFB range. The bar for
705 induced theta power is delineated by a dashed line and shown as a reference only, since theta
706 power was the basis for selecting these electrodes for further analysis.

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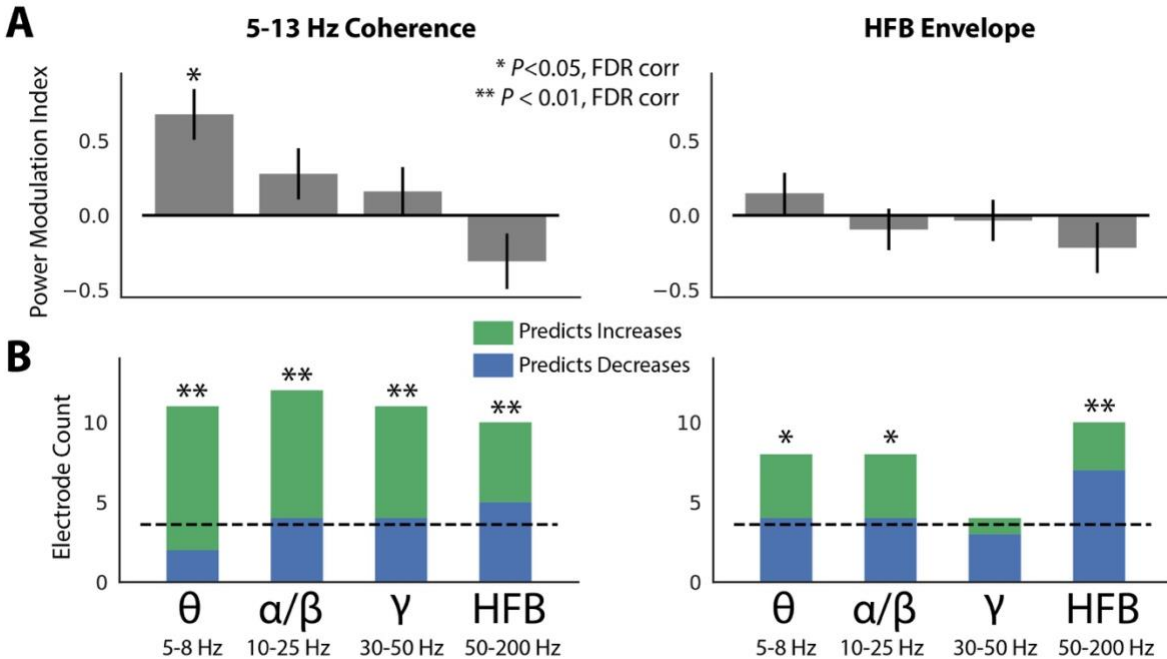
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718 **Figure 7. Power response at higher frequencies.** (A) The average power modulation index –
 719 reflecting the degree to which functional connectivity predicts changes in spectral power – was
 720 computed for theta (5-8 Hz), alpha/beta (10-25 Hz), gamma (30-50 Hz), and HFB (50-200 Hz)
 721 bands for all 72 MTL stimulation sites. Functional connectivity was measured as 5-13 Hz
 722 coherence (left) and HFB amplitude correlation (right; see Methods for details). Across all
 723 stimulation sites, 5-13 Hz coherence significantly predicted changes in theta power, as
 724 demonstrated in Figure 2 (corrected $P < 0.05$). HFB amplitude correlations did not significantly
 725 predict power changes in any band. (B) To account for the fact that connectivity could predict
 726 decreases or increases in power, each stimulation site was classified by whether its functional
 727 connectivity profile significantly predicted decreases (blue) or increases (green) power (two-
 728 tailed $P < 0.05$). The count of significant stimulation sites are depicted as stacked bars; the
 729 expected false positive rate ($P = 0.05$) is indicated as a dashed line. Note that though HFB
 730 amplitude correlations do not significantly predict changes in HFB power on average (panel A),

731 the count of stimulation sites where connectivity significantly predicts changes in HFB power is
732 significantly greater than chance (7 decreases, 3 increases; binomial test, corrected $P < 0.01$).

733
734 **Methods**

735
736 Participants

737 Twenty-six patients with medication-resistant epilepsy underwent a surgical procedure to
738 implant subdural platinum recording contacts on the cortical surface and within brain
739 parenchyma. Contacts were placed so as to best localize epileptic regions. Data reported were
740 collected at 8 hospitals over 4 years (2015-2018): Thomas Jefferson University Hospital
741 (Philadelphia, PA), University of Texas Southwestern Medical Center (Dallas, TX), Emory
742 University Hospital (Atlanta, GA), Dartmouth-Hitchcock Medical Center (Lebanon, NH),
743 Hospital of the University of Pennsylvania (Philadelphia, PA), Mayo Clinic (Rochester, MN),
744 National Institutes of Health (Bethesda, MD), and Columbia University Hospital (New York,
745 NY). Prior to data collection, our research protocol was approved by the Institutional Review
746 Board at participating hospitals, and informed consent was obtained from each participant.

747
748 Electrocorticographic recordings

749 iEEG signal was recorded using depth electrodes (contacts spaced 3.5-10 mm apart) using
750 recording systems at each clinical site. iEEG systems included DeltaMed XITek (Natus), Grass
751 Telefactor, and Nihon-Kohden EEG systems. Signals were sampled at 500, 1000, or 1600 Hz,
752 depending on hardware restrictions and considerations of clinical application. Signals recorded at
753 individual electrodes were first referenced to a common contact placed intracranially, on the
754 scalp, or mastoid process. To eliminate potentially confounding large-scale artifacts and noise on
755 the reference channel, we next re-referenced the data using a bipolar montage. Channels
756 exhibiting highly non-physiologic signal due to damage or misplacement were excluded prior to
757 re-referencing. The resulting bipolar timeseries was treated as a virtual electrode and used in all
758 subsequent analysis.

759
760 Anatomical localization

761 To precisely localize MTL depth electrodes, hippocampal subfields and MTL cortices were
762 automatically labeled in a pre-implant, T2-weighted MRI using the automatic segmentation of
763 hippocampal subfields (ASHS) multi-atlas segmentation method⁴⁶. Post-implant CT images
764 were coregistered with presurgical T1 and T2 weighted structural scans with Advanced
765 Normalization Tools⁴⁷. MTL depth electrodes that were visible on CT scans were then localized
766 within MTL subregions (including white matter) by neuroradiologists with expertise in MTL
767 anatomy. All localizations in this manuscript refer to the bipolar midpoint of two recording
768 contacts or the anode/cathode stimulation contacts.

769
770 Functional connectivity estimation

771 To obtain coherence values between electrode pairs, we used the MNE Python software package
772⁴⁸, a collection of tools and processing pipelines for analyzing EEG data. The coherence (C_{xy})
773 between two signals is the normalized cross-spectral density (Equation 1); this can be thought of

774 as the consistency of phase differences between signals at two electrodes, weighted by the
775 correlated change in spectral power at both sites.

776

$$777 \quad C_{xy} = \left| \frac{S_{xy}}{S_{xx}S_{yy}} \right| \quad (1)$$

778 Where S_{xy} is the cross-spectral density between signals at electrodes x and y ; S_{xx} and S_{yy} are the
779 auto-spectral densities at each electrode. Consistent with other studies of EEG coherence^{49,50}, we
780 used the multitaper method to estimate spectral density. We used a time-bandwidth product of 4
781 and a maximum of 8 tapers (tapers with spectral energy less than 0.9 were removed), computing
782 coherence for frequencies between 4-50 Hz, avoiding the 60 Hz frequency range that may be
783 contaminated by line noise. Inter-electrode coherences were computed for a series of 1-second
784 windows extracted sequentially from 10-second baseline periods of a non-stimulation task, in
785 which subjects wait passively before beginning a verbal free-recall task. Each subject typically
786 had 24-72 such baseline periods, but all had a minimum of 10 (i.e. the minimum total number of
787 windows used for network estimation was 100). To construct the low-frequency networks used
788 in the majority of this paper, cross-spectra were first averaged across all baseline period
789 windows, normalized by the average power spectra, and then averaged between 5-13 Hz. For the
790 analysis in Figure 4E, networks are constructed for each frequency between 4-50 Hz with no
791 averaging over bands.

792

793 Stimulation paradigm

794 At the start of each session, we determined the safe amplitude for stimulation using a mapping
795 procedure in which stimulation was applied at 0.5 mA, while a neurologist monitored for
796 afterdischarges. This procedure was repeated, incrementing the amplitude in steps of 0.5 mA, up
797 to a maximum of 1.5 mA (chosen to be below the afterdischarge threshold and below accepted
798 safety limits for charge density⁵¹). For each stimulation session, we passed electrical current
799 through a single pair of adjacent electrode contacts in the MTL. Stimulation was delivered using
800 charge-balanced biphasic rectangular pulses (pulse width = 300 μ s) at (10, 25, 50, 100, or 200)
801 Hz frequency and (0.25 to 2.00) mA amplitude (0.25 mA steps) for 500 ms, with a minimum of 3
802 seconds between stimulation events. During a session, subjects were instructed to sit quietly and
803 did not perform any task. An average of 2.7 stimulation sites were selected for each subject, with
804 a minimum of 240 trials delivered for each. In a typical stimulation session, a given target would
805 receive 360 total stimulation events, in blocks of 120 trials at each amplitude, with 24 randomly-
806 spaced trials at each frequency within the block (Figure 1D). For all analyses in the main text,
807 effects were aggregated across stimulation parameters; see Supplemental Figure 2 for
808 consideration of stimulation frequency and amplitude.

809

810 In most subjects, a post-stimulation voltage deflection artifact briefly contaminates a subset of
811 recording contacts. To identify and remove channels exhibiting this artifact, the average voltage
812 in the 350 ms prior to stimulation is compared with a paired t-test to the average voltage in the
813 350 ms after stimulation, across all trials, for each channel. The same procedure is done with a
814 levene test for different variances. Any electrode with a significantly different pre-vs.-post mean
815 voltage or voltage variance ($P < 0.01$) is excluded from further analysis (see “Estimating theta
816 modulation index”). On average, this procedure excludes 28% of channels. Regardless of

817 stimulation artifact, any bipolar pair is excluded from analysis if it shares a common contact with
818 the stimulated pair. See Supplemental Figure 5 for a representative example of this artifact.

819

820 Spectral power analysis

821 We used the multitaper method to assess spectral power in the pre- and post-stimulation intervals
822 (-950 to -50 ms relative to stimulation onset, and +50 to +950 ms after stimulation offset; Figure
823 1B). We avoided the Morlet wavelet method to obviate the need for buffer periods that extend
824 into the stimulation window. As in “Functional connectivity estimation,” we used the MNE
825 Python software package. For each trial, theta power was taken as the average PSD from 5-8 Hz,
826 using a time-bandwidth product of 4 and excluding tapers with < 90% spectral concentration. To
827 compute a T-statistic at each electrode, the pre- vs. post log-transformed power values were
828 compared with a paired t-test (Fig. 1D, Fig. 2B). We avoid calculating significances for
829 individual electrodes because sequential trials are non-independent events; T-statistics are only
830 used for later correlation analysis (see “Theta modulation index”).

831

832 For analyses that considered spectral power at higher frequencies (Figure 6 and Supplemental
833 Figure 4), we used the following bands: alpha/beta (10-25 Hz), gamma (30-50 Hz), and high-
834 frequency broadband (50-200 Hz). Power was otherwise computed exactly as described for theta.
835 To measure whole-brain evoked power (Supplemental Figure 4A), we took the average T-
836 statistic across all electrodes in each subject’s brain, subject to the same exclusion criteria
837 described in “Estimating theta modulation index.” Additionally, we excluded electrodes with T-
838 statistics greater than 10 from the whole-brain average, to account for raw power values that are
839 potentially corrupted by post-stimulation artifact which survives our exclusion procedure (their
840 inclusion does not significantly change the main results).

841

842 Estimating theta modulation index

843 To examine the relationship between stimulation and functional connectivity, we developed an
844 index that reflects the correlation between theta power modulation and connectivity, independent
845 of distance. To do this, we first construct low-frequency (5-13 Hz) networks as described in
846 “Functional connectivity estimation,” and take the logit transform to linearize coherence values
847 that fall between 0 and 1. We also construct adjacency matrices that reflect the normalized
848 Euclidean distance between all possible pairs of electrodes (Fig. 2A), and linearize the distances
849 by taking the reciprocal of their exponential (i.e. a Euclidean distance of zero would correspond
850 to 1.0). For each stimulated electrode, we take that electrode’s distance and connectivity to all
851 other electrodes as predictors of the theta power t-statistic (see “Spectral power analysis) in a
852 multiple linear regression. This controls for the effect of distance from a stimulation target,
853 which is correlated with power and functional connectivity. Next, we permute the order of the
854 predictors 1000 times and re-run the regression for each. The true coefficient for functional
855 connectivity is compared to the distribution of null coefficients to obtain a z-score and p-value
856 for each stimulation site. The z-score is referred to as the theta modulation index, or TMI.

857

858 Prior to computing TMI, we excluded electrodes placed in the seizure onset zone or exhibiting
859 significant inter-ictal spiking, as determined by a clinician. Electrodes with high post-stimulation
860 artifact (see “Stimulation paradigm”), and stimulated electrodes themselves, were also excluded.
861 Subjects were discarded if less than 10 electrodes remained after all exclusions.

862

863 To analyze the relationship between TMI and white matter category (Fig. 3), we first binned
864 electrodes according to their distance from nearest white matter. Distance were measured as the
865 linearized Euclidean distance from a stimulation electrode (i.e. bipolar midpoint of the
866 anode/cathode) to the nearest vertex of that subject's Freesurfer white matter segmentation⁵²
867 based on T1 MRI. The 50th percentile of white matter distances marked the division between
868 stimulation electrodes categorized as "near" white matter versus in gray matter. Seven
869 stimulation electrodes were identified by expert neuroradiologists as being placed within white
870 matter (see Supplementary Figure 2 for exact placements). To ask whether TMI increases with
871 white matter category, permuted the white matter labels for each electrode 1000 times and took
872 the minimum T-statistic between gray vs. near and near vs. in categories at each permutation. We
873 then compared the minimum T-statistic in the true data to the distribution of null statistics to
874 generate a p-value.

875

876 Network properties of stimulation

877 To determine how the network structure of a stimulation site affected downstream alterations in
878 theta power (Fig. 4), we first analyzed the relationship between pre- vs. post-stimulation theta
879 power and the strength of functional connectivity to a stimulation site (Fig. 4A-B). For each
880 stimulation site with a significant TMI ($P < 0.05$), we ranked all other electrodes by the strength
881 of their functional connectivity to that site, residualized on Euclidean distance ($e^{-\text{dist}}$). We then
882 took the average power T-statistic (see "Spectral power analysis") across the 5 strongest-
883 connected sites and the 5 weakest-connected sites, to assess whether theta power changes
884 correlated with the strength of a functional connection.

885

886 To assess whether the effects of stimulation differ between hubs and non-hubs (Fig. 4C-D), we
887 measured the node strength⁵³ for each stimulation site in or near white matter ($n=38$), using our
888 low-frequency coherence networks (see "Functional connectivity estimation"). The node strength
889 reflects the sum of all connection strengths to a given node (for this paper, we normalized node
890 strength by the total number of possible connections for a given site, yielding strengths in the
891 range from 0 to 1). For all stimulation sites, we binned hub scores by tercile, and took the highest
892 tercile as "strong hubs," the weakest tercile as "weak hubs" ($n=13$ for each). For stimulation at
893 all strong and weak hubs, we took the average power T-statistic for the 5 strongest-connected
894 electrodes. These values were used to assess whether hub stimulation tends to cause greater
895 power changes in connected regions. The relationship between coherence frequency and theta
896 modulation index (Figure 4E) was assessed by re-estimating the TMI (see "Estimating theta
897 modulation index") using spectral coherence networks observed for each frequency between 4-
898 50 Hz, spaced by 1Hz, for all stimulation electrodes placed within or near white matter. The
899 average TMI across sites/subjects was 1-sample t -tested against zero and p-values were FDR
900 corrected for multiple comparisons (corrected $P < 0.05$). For visualization purposes only, the
901 displayed TMI/frequency curve was smoothed with a 3-point moving average window.

902

903 Alternative connectivity metrics

904 *HFB amplitude envelope correlation*: Networks of correlated high-frequency broadband (HFB;
905 50-200 Hz in this manuscript) amplitude envelopes were computed in a manner similar to Foster,
906 et al. (2015)²³. The general approach is to low-pass filter HFB spectral power during a resting
907 period, and the resulting timeseries are correlated between recording electrodes to construct an
908 adjacency matrix. Specifically, in a non-stimulation memory task, we extracted 240-second (4

909 minute) resting periods between any task events. Resting periods were identified by searching for
910 the maximum amount of time between task events; in some subjects, 240-second intervals were
911 not available but timeseries were still extracted for that length in the best-possible period. Signals
912 were bipolar re-referenced and notch filtered, sequentially band-passed in 10 Hz windows from
913 50-200 Hz, Hilbert transformed and normalized to the mean amplitude, and then averaged across
914 bands. Finally, to estimate the slow variation in this signal as a basis for inter-regional
915 correlation, we low-pass filtered the HFB amplitude (< 1 Hz), and computed the Pearson
916 correlation coefficient between the resulting signals between all possible pairs of electrodes
917 within a subject, yielding an adjacency matrix of correlations. The resulting correlations were
918 Fisher z-transformed and then used as predictors of modulations in power (see “Estimating a
919 theta modulation index”).

920

921 *Atlas-based rsfMRI:* We used an independent dataset of resting state functional MRI from the
922 Human Connectome Project (HCP)⁵⁴ to estimate functional connectivity between recording sites
923 in each patient. For each patient, we mapped the location of subdural and depth electrodes to the
924 HCP grayordinate space⁵⁵. For subdural electrodes, we assigned vertices on the cortical surface
925 mesh within 3mm (geodesic distance) of each recording site to a region of interest (ROI). The
926 coordinates in the native space of each subject were then mapped to the standard fs_LR mesh
927 (i.e., HCP surface space). The location of subcortical contacts in native space were transformed
928 to MNI space using Advanced Normalizations Tools⁵⁶, with spherical ROIs centered at each
929 bipolar midpoint. Adjacency matrices for each subject were constructed by computing the
930 average connectivity (fisher-transformed timeseries correlations) between all grayordinates from
931 each pairwise combination of ROIs, provided by the group-averaged (n=897 subjects) dense
932 connectome. These adjacency matrices were subsequently used to determine whether fMRI
933 defined networks provide a scaffold for the propagation of brain-wide theta power following
934 direct electrical stimulation.

935

936