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# A speech planning network for interactive language use

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## Abstract

During conversation, people take turns speaking by rapidly responding to their partners while simultaneously avoiding interruption<sup>1, 2</sup>. Such interactions display a remarkable degree of coordination, as typical gaps between turns are  $\sim 200 \text{ ms}^{3, 4}$  – approximately the duration of an eyeblink<sup>5</sup>. These response times are considerably shorter than those observed in simple word production tasks<sup>6</sup>, indicating that speakers often plan their responses while listening to their partners<sup>2</sup>. Although a distributed network of brain regions has been implicated in speech planning<sup>7–14</sup>, the neural dynamics underlying the specific preparatory processes enabling rapid turn-taking are poorly understood. Here we use intracranial electrocorticography to precisely measure neural activity as participants perform interactive tasks, and we observe a distinct class of cortical dynamics associated with speech planning. We localize these preparatory responses to a frontotemporal circuit that is anatomically segregated from regions involved in speech perception and production, and we find the core of this network to be comprised of language-critical caudal inferior frontal cortex<sup>15, 16</sup> (i.e., 'Broca's region') and caudal middle frontal gyrus -a region not normally implicated in speech planning 17-20. Using a series of motor tasks, we then show that this planning network is selectively active when preparing speech but significantly less recruited when planning nonlinguistic orofacial and manual actions. Finally, we uncover planning-related circuitry in natural conversation that is nearly identical to the network mapped with our interactive tasks, and we find this circuit to be most active prior to participant speech during unconstrained turn-taking. Therefore, we have identified a speech planning network central to natural language generation during social interaction.

Turn-taking during typical conversation can be divided into three major cognitive processes: perception of the opposing speaker's turn, planning of one's own turn, and production

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AUTHOR CONTRIBUTIONS

G.A.C. and M.A.L conceived the study and designed the experiments; G.A.C., C.K.K., J.D.W.G., and M.A.L conducted the research; G.A.C., C.K.K., and M.A.L performed data analyses; G.A.C., C.K.K., and M.A.L created the Figures; G.A.C. and M.A.L. wrote the initial draft of the manuscript; G.A.C., C.K.K., M.A.H., J.D.W.G., and M.A.L edited and reviewed the final manuscript. J.D.W.G. and M.A.L acquired funding; J.D.W.G., M.A.H., and M.A.L. supervised the project.

COMPETING INTERESTS

The authors declare no competing interests.

of the speech comprising that turn<sup>2, 21</sup> (Fig. 1a,b). While each of these steps are comprised of several subcomponents<sup>12</sup> (Extended Data Fig. 1a), speech planning is an especially multifaceted operation which encompasses various functions ranging from abstract conceptual and semantic processes to lower-level articulatory programming and motor initiation<sup>8, 12, 22–26</sup>. Accordingly, many cortical regions have been linked to aspects of planning, including inferior frontal gyrus<sup>7, 11, 13, 27</sup>, premotor cortex<sup>12</sup>, insula<sup>14</sup>, superior temporal gyrus<sup>9</sup>, supplemental motor area<sup>8, 28</sup>, and inferior parietal cortex<sup>10</sup>. However, because non-interactive language tasks – such as picture naming<sup>27, 29, 30</sup>, repetition<sup>7, 9, 11, 28</sup>, and reading<sup>7, 13</sup> – have typically been used to identify such candidate regions, their relevance to speech preparation during interaction is unknown. In this study, we delineate the neural substrates underlying the planning processes relevant for rapid turn-taking by measuring cortical activity while participants engage in structured interactive tasks as well as unconstrained conversation.

## Behavioral isolation of planning activity

A major barrier to the study of speech planning is that it is an internal process with little or no behavioral correlate<sup>31, 32</sup> which often overlaps with speech perception and production during natural conversation<sup>2, 33-35</sup>. Therefore, we first sought to experimentally isolate the neural activity related to planning during turn-taking. To do so, we employed an established question-answer paradigm in which a single word (i.e., the 'critical information' or CI) initializes speech planning by providing the information necessary for a correct response<sup>33, 36</sup> (Fig. 1c-f). An experimenter posed a battery of 39 to 94 CI questions (55.1  $\pm$  20.5, mean  $\pm$  SD; Extended Data Table 1) to 8 participants (i.e., neurosurgical patientvolunteers; Extended Data Table 2), in which the CI was presented either near the middle of the question ('early') or at its end ('late'). If speech planning is initiated by CI, then late trials should display longer response latencies than early trials as they provide relatively less planning time<sup>2, 33</sup>. We observed that median reaction times were  $64.6 \pm 31.5\%$  longer in late trials compared to early trials (median: 495 ms vs. 782 ms; p < 0.05 [n = 8 participants], signed-rank test; Extended Data Fig. 1b,c), indicating that the CI paradigm effectively isolates the core set of prearticulatory planning processes relevant to rapid turn-taking in our neurosurgical cohort.

We next examined cortical responses while participants answered CI questions. Previous work using this task reported widespread CI-related activity at the scalp using EEG<sup>33, 36</sup>, and we sought to leverage the temporal (> 10 ms) and spatial (> 5 mm) precision<sup>37, 38</sup> of intracranial electrocorticography (ECoG) to measure planning activity at distinct cortical sites. We implanted a total of 874 ECoG electrodes across the left, language-dominant hemispheres of 8 participants (64 to 224 electrodes in each participant; Extended Data Table 3) and found 790 electrodes (90.4%) to be suitable for further analysis (for details related to exclusion criteria, see Methods). We assayed neural activity by examining the high gamma frequency band (~ 70–150 Hz) of the local field potential. Using this method, we observed many individual electrodes whose activity profiles were temporally locked to specific phases of these interactions (Fig. 1c–f). For example, perception-related activity remained sustained throughout the duration of the experimenter's question (Fig. 1c,d), and production-related activity was largely restricted to the period comprising the participant's spoken answers

(Fig. 1e,f). In contrast, planning-related activity was observed immediately following CI presentation and generally returned to baseline prior to participant response (Fig. 1c–f). This profile was consistent across trials (Fig. 1g–j) regardless of CI position (i.e., early or late) or question type (Extended Data Fig. 1d,e).

Previous work has demonstrated that cortical language circuitry is highly multimodal<sup>39-41</sup> and that processing for interactive behaviors may be widely distributed rather than organized into discrete modules<sup>42</sup>. Therefore, it is possible that the dynamics exhibited by our electrodes may be heterogeneous and form a continuum at the population level. Instead, neural activity may cluster into distinct categories, indicating that individual sites are specialized for specific phases of spoken interactions. To differentiate these alternatives, we designed a general linear model (GLM) to quantify activity levels during the perception, planning, and production windows of the CI task (colored bars in Fig. 1c-f; Extended Data Figs. 1f and 2). We focused our analyses on significant increases in high gamma amplitude - a correlate of local neuronal activity 43-45 - and detected a total of 253 electrodes (32.0%) that were responsive during at least one of the defined task epochs (Fig. 1k-n, Extended Data Table 3). We then examined the organization of all task-responsive electrodes in threedimensional space according to their perception, planning, and production GLM weights (Fig. 10). If task-related activity represented a continuum, electrodes would be evenly distributed throughout this volume; however, we found that electrodes appeared to form three distinct clusters.

To test whether these putative clusters could have arisen from a continuum of neural activity as opposed to discrete response categories, we compared the recorded data to a 'continuum model' where activity related to perception, planning, and production (i.e., GLM weights) was randomly selected from independent unimodal distributions (see Methods) (Fig. 1p). In both observed and simulated data sets, we used an unsupervised approach (k-medoids) to define three activity clusters in GLM weight space (Fig. 10,p). Electrodes exhibiting significant positive responses within the perception, planning, or production windows were invariably confined to separate clusters (Fig. 1q), while the same analysis using our simulated 'continuum model' always resulted in clusters containing multiple response types (interquartile range [IQR] of misclustered sites: 20.1–40.9%) (Fig. 1r, Extended Data Fig. 3a). Additionally, we found that only 61 electrodes (24.1% of responsive sites) in our recorded data displayed 'mixed' selectivity (i.e., positive responses within more than one task window, Fig. 1s), significantly fewer than expected under the continuum model (IQR: 44.3–48.6%) (Fig. 1t, Extended Data Fig. 3b). Taken together, these results indicate that neural activity recorded with our ECoG electrodes is organized into discrete categories related to speech perception, planning, and production during turn-taking.

We next examined the relative frequency of neural responses (i.e., significant positive GLM weights) related to each window of the CI task. We found responses related to planning and production were most common, with 20.3% and 15.3% of electrodes displaying significant increases in activity during the planning and perception windows, respectively. These planning and production electrodes often exhibited significant negative perception-related GLM weights (Fig. 1k–n, Extended Data Fig. 4a, Extended Data Table 3), reflecting a decrease in high gamma activity during the perception window of the CI task (Extended

Data Fig. 4b). Meanwhile, only 4.6% of electrodes were responsive during the perception window, most likely resulting from sparse electrode coverage over auditory-related regions in temporal lobe<sup>46–49</sup> (Extended Data Fig. 5a–c). However, we performed additional analyses to ensure that the temporal overlap of the perception and planning windows in early CI trials (~50% of trials; Fig. 1c,e) did not bias our analysis against detecting perception responses. We reanalyzed our data set with a reduced GLM lacking a planning component and found no net increase of perception-related electrodes (Extended Data Fig. 3c–e). These results demonstrate that our statistical approach and behavioral paradigm effectively delineated the cortical dynamics related to speech perception, planning, and production.

## Spatial organization of planning responses

While ECoG has been previously used to map cortical regions related to speech perception<sup>46–49</sup> and production<sup>12, 50, 51</sup> with high spatial precision, the circuitry involved in speech planning remains poorly defined. For example, it is possible that planning-related sites are segregated to distinct cortical areas<sup>7–14, 27, 28</sup>; conversely, planning sites may instead be widely distributed across the cortical surface<sup>52</sup> and share extensive territory with perception and/or production related sites<sup>39–41</sup>. To resolve this issue, we first examined the activity recorded in arrays of ECoG electrodes within individual participants. We observed that electrodes which were responsive for the same windows of the CI task (i.e., same GLM class) appeared to be spatially clustered (Fig. 2a,b; Extended Data Fig. 5a).

We then determined the cortical location of all electrodes using precise anatomical reconstructions for each participant (Extended Data Table 3; see Methods) and observed spatially segregated cortical networks related to speech perception, planning, and production during the interactive CI task (Fig. 2c-e). Electrodes responsive during speech perception (Fig. 2c) and production (Fig. 2d) were largely restricted to well-characterized sensory and motor structures, respectively. Meanwhile, 95% of planning-related electrodes were grouped in a spatially segregated anterior frontotemporal region (Fig. 2e, Extended Data Table 3). Specifically, the highest densities of planning electrodes were clustered in caudal inferior frontal gyrus (cIFG) and caudal middle frontal gyrus (cMFG) with a small number of planning sites located in ventral speech motor and premotor cortices (vSMC) as well as anterior superior temporal gyrus (aSTG) (Fig. 2f). In contrast, the highest densities of production-related electrodes were tightly restricted to vSMC and dorsal SMC (Extended Data Fig. 5d). To ensure that these anatomical findings were not biased by the clinical condition of the participants, we generated separate coregistrations for tumor and epilepsy patients. This analysis revealed similar spatial patterns across groups (Extended Data Fig. 5e), indicating that the observed neuroanatomical organization is a robust feature of the human brain.

We found that many electrodes within vSMC, IFG, and cMFG exhibited significant negative perception-related GLM weights (Extended Data Fig. 4c–e), suggesting these planning and production-related structures are likely to be significantly less active during speech perception. However, some electrodes within IFG displayed perception-related responses (Fig. 2c), potentially reflecting a previously established role for IFG in language comprehension<sup>53</sup>.

#### Speech selectivity of cortical planning sites

To initiate speech planning in the CI task, one must first perceive and comprehend the spoken word comprising the CI. Therefore, a potential limitation of the task is that it does not completely disambiguate between the onset of speech planning and the cognitive processes involved in understanding the CI. We addressed this issue by designing a command-response (CR) task in which participants were instructed to perform a range of actions varying in their linguistic relevance (Extended Data Table 1), including: hand movements (i.e., button pressing) ('CR1', Fig. 3a), nonspeech orofacial behaviors ('CR2', Fig. 3b), repetition of words and pseudowords ('CR3', Fig. 3c), and performing linguistic operations (i.e., noun pluralization) ('CR4', Fig. 3d). Crucially, the structure of each CR task mirrors the CI task – a critical word or phrase is presented either early or late, and its comprehension is required to generate a motor response.

Using this behavioral approach, we further characterized electrodes that had displayed significant planning-related responses in the CI task (n = 124 electrodes from 6 participants; see Methods and Extended Data Table 2). Although we observed individual planning electrodes which responded maximally in each of the CR tasks (Fig. 3a-d), we found that only a subset of electrodes displayed significant increases in preparatory activity for hand (18.6%) and nonspeech orofacial movements (43.6%), while the majority were responsive for speech repetition (59.7%) and pluralization (80.7%) (Fig. 3e-h). At the population level, planning activity levels were lower for manual movements than for all other behaviors and higher while preparing plural nouns compared to all other actions (p < 0.005 [n = 124electrodes], Friedman Test with Dunn-Šidák correction); however, planning activity in the pluralization task did not differ in magnitude from the CI task (p > 0.99 [n = 124 electrodes], Friedman Test with Dunn-Šidák correction) (Fig. 3i). To test whether our results could have arisen due to task difficulty, we compared participant reaction times for late CI trials of the repetition task, pluralization task, and CI questions and found no significant difference (p = 0.3114 [n = 6 participants], Friedman Test; Fig. 3j). These results indicate that activity within the identified planning network is more likely to reflect speech preparation rather than other processes related to task engagement.

The CR tasks also allow us to assess the degree to which planning network dynamics are specific for preparing speech as opposed to representing more generic premotor processes supporting a wide range of behaviors<sup>9, 54–56</sup>. While electrode sites within network were generally more active while planning speech as opposed to nonspeech tasks (Fig. 3e–i), we hypothesized that subregions within the circuit may display variable degrees of speech selectivity. We explicitly tested this notion by first examining planning activity levels in relation to cortical location, and we observed higher levels of preparatory activity broadly within the planning network as linguistic relevance increased (Fig. 3k–n). However, upon further inspection, we found that planning electrodes within cIFG were more linguistically selective than electrodes located in the precentral gyrus (preCG) and cMFG (Fig. 3o–r). While most preCG and cMFG electrodes displayed significant planning responses in both speech tasks, electrodes in cIFG were generally responsive only for the pluralization task (Fig. 3q–s). This result demonstrates that our planning circuit may be further divided into subnetworks which are likely related to distinct speech planning processes.

#### Speech planning during natural conversation

We next aimed to compare the neural activity observed during task-based interaction to that generated during natural language use. To do so, we used the same electrode arrays from our structured tasks to record neural activity in participants engaging in unconstrained conversation with the experimenter for 2.8 to 16.3 minutes ( $7.5 \pm 4.0$  minutes; Extended Data Fig 6a) following task completion. In this unstructured context, we found that our neurosurgical cohort exhibited grossly normal turn-taking behavior; all participants were capable of rapid responses (i.e., gap durations < 250 ms) while 6 of 8 participants displayed modal gap values of 200 ms or less (Extended Data Fig. 6b), which is comparable to conversational response times observed in the general population<sup>2–4</sup>. Thus, these rapid interactions provide an important opportunity to investigate the neural mechanisms enabling naturalistic turn-taking.

During conversation, we observed that individual electrodes displaying significant planning responses in the CI task were often active prior to the onset of a participant's turn (Fig. 4a), suggesting these sites are engaged in speech preparation during both behaviors. However, unlike our interactive tasks – where perception of a single word (i.e., the CI) is necessary to initiate planning – a speaker is free to begin response preparation and to update their response plans at any time during conversation. Furthermore, behavior during conversation is also highly flexible, with factors such as turn duration and complexity (i.e., syntactic structure, semantic content, etc.), inter-turn gap duration, and planning strategy varying widely both within and across speakers<sup>3</sup>, <sup>35</sup>, <sup>36</sup>, <sup>57–59</sup>. Due to this behavioral variability, a supervised analysis of language-related neural activity during natural turn-taking is unfeasible as the content and timing of planning is uncontrolled.

To overcome this issue, we performed Principal Component Analysis (PCA) on continuous electrode signals (~ 10 minutes) recorded while participants engaged either in the interactive tasks or natural conversation. This approach allowed us to identify sets of electrodes displaying correlated signals during task performance and conversation independently and to assess whether network organization remains stable between these two contexts. When participants were engaged in the structured tasks, we found that electrode signals formed three main clusters in three-dimensional PC coefficient space (n = 6 participants; Fig. 4b, Extended Data Fig. 7a–f), indicating that neural activity during task-based interactions is organized into distinct classes. We then asked whether this organization is preserved in unstructured turn-taking (Fig. 4c) and found that electrode signals formed nearly identical clusters during conversation (Fig. 4d, Extended Data Fig. 7a–f). Only 5 of 200 electrodes were differently clustered between the two behaviors, significantly less than would be expected by chance (IQR: 120–128; p < 0.0001, permutation test; Extended Data Fig. 6d). Therefore, these results demonstrate that the cortical network active during language generation in task-based interactions is similarly organized during natural turn-taking.

Our next goal was to determine whether the PCA activity clusters observed during natural conversation corresponded to neural responses related to speech perception, planning, and production. These functional classifications were originally made using a GLM (e.g., Fig. 1), which detected elevated neural activity within predefined time windows of the CI task (Fig. 4e). Importantly, we found that electrode clusters in PC coefficient space

were overwhelmingly comprised of electrodes exhibiting a single class of response in the CI task (Fig. 4f, Extended Data Fig. 7a–f). We were consequently able to functionally categorize each PCA cluster as 'perception', 'planning', or 'production' in 17 out of 18 cases across participants (Fig. 4f, Extended Data Fig. 7a–f). We observed a high degree of correspondence between cluster categorization and GLM-derived activity classes, such that 93.8%, 94.7% and 96.1% of electrodes in perception, planning, and production clusters displayed significant perception, planning, and production responses, respectively (p < 0.0001, permutation test; Extended Data Fig. 6e). Therefore, using our unsupervised PCA analysis method, we uncovered categories of neural activity during natural conversation which correspond to the distinct response classes and circuits isolated with the CI task (Extended Data Fig. 6f).

We next examined the functional properties of electrodes within perception, planning, and production clusters in relation to task-based and natural turn-taking by assaying time-varying PC activity (i.e., PC scores). Specifically, we found that each category of cluster was highly weighted for an individual PC (Fig. 4g). We employed these maximally weighted PCs as an aggregate signal reflecting the overall activity of each cluster and further analyzed 2 perception-related PCs, 6 planning-related PCs, and 5 production-related PCs, excluding 4 clusters which primarily contained electrodes displaying mixed responses (Extended Data Fig. 7; see Methods). Across participants, we observed that planning PCs displayed a significant peak in average activity near the offset of experimenter speech which continued into the inter-turn gap during both the task and conversation; conversely, perception PCs were active during the experimenter turn but sharply decreased in activity at the offset of experimenter speech (p < 0.05, permutation test; Fig. 4h). Planning PCs likewise exhibited a peak in activity prior to participant turn onset for both the task and conversation, while production PC activity occurred later and typically during participant speech in both contexts (p < 0.05, permutation test; Fig. 4i). These results demonstrate that the functional properties of PCs related to speech perception, planning, and production during unconstrained turn-taking are congruent with the responses observed during the structured CI task.

Finally, we examined planning PC activity in individual turn-taking interactions to provide a first description of speech planning-related dynamics during natural conversation. Consistent with the average response profiles discussed above, we observed planning-related PCs to be most active during the experimenter turns and the silent gaps prior to participant turns; these responses differed dramatically from production-related PCs, whose activity was largely restricted to periods of articulation (Fig. 4j–l). Across participants, we likewise found that planning PCs displayed maximal activity levels during inter-turn gaps while the perception and production PCs were most active during the experimenter and participant turns, respectively (Fig. 4m–o; Extended Data Fig. 6g). Taken together, the functional properties displayed by the planning-related PCs are consistent with speech preparation during conversation, indicating that these signals represent the neural correlates of the speech planning processes enabling rapid turn-taking during natural language use.

### DISCUSSION

Conversational turn-taking requires a complex interplay of simultaneous speech perception and planning coupled with finely timed speech production. We used high-precision ECoG recordings to isolate planning-related dynamics and found planning responses to be functionally and anatomically distinct from dynamics underlying speech perception and production. We further observe that these distinct classes of neural activity exist during both structured interactions as well as unconstrained conversation, thus paving the way for future studies of naturalistic speech generation by providing the first explicit description of neural dynamics arising in natural language use. Although previous work has suggested that the neural circuitry related to human language is highly multimodal<sup>39–41</sup> and distributed<sup>12, 52, 60</sup>, our findings indicate that separate cortical modules perform distinct functions within the language network during ethologically-relevant interactive speech production<sup>12, 22</sup>. This conclusion is consistent with perturbation studies of human language circuitry<sup>61–64</sup> as well as investigations of cortically-dependent vocalizations in animals demonstrating the existence of anatomically-distinct subcircuits within larger vocal production systems<sup>65–67</sup>.

Our findings demonstrate that planning-related responses during interaction are largely restricted to a frontotemporal circuit centered on cIFG and cMFG which is located within the broader peri-Sylvian language network. The caudal portion of IFG (pars opercularis and pars triangularis) is thought to represent Broca's region<sup>68</sup>, which is classically considered to be crucial for speech production<sup>15, 16</sup>. However, more recent work has demonstrated that this region is active prior to – rather than during – articulation<sup>7, 27, 30</sup>. Using our interactive paradigm, we also observe that responses in cIFG occur primarily prior to speech initiation, and we show that this activity is specifically associated with planning speech. Unlike cIFG, cMFG had not previously been considered to be important for speech planning or production<sup>17–20, 69</sup>, although a case of pure apraxia of speech following its focal resection has been recently reported<sup>17</sup>. Here we show for the first time that cMFG is involved in speech-selective planning during interaction. This region displays robust anatomical connectivity with cIFG<sup>70</sup>, as well as other language-related regions<sup>17, 71</sup> and therefore represents an important cortical site for future study as its role in language generation has been largely overlooked.

It is likely that cIFG and cMFG perform distinct functions related to speech planning, as evidenced by differing levels of speech selectivity exhibited during the command-response tasks. However, future work is necessary to elucidate the specific planning subprocesses encoded by these and other nodes within the identified planning network and how these subprocesses unfold on a moment-to-moment level during turn-taking. Nevertheless, the isolation of this planning circuitry represents an important advance towards understanding how the human brain generates language in naturalistic contexts and complements parallel efforts in the wider field of neuroscience aimed at quantifying natural behavior broadly<sup>72</sup>. Furthermore, this work helps lay a foundation for future studies investigating the biological basis of communication disorders – such as stuttering and apraxia of speech – which disrupt social language use as a result of abnormalities in speech planning<sup>73–75</sup>.

# METHODS

#### **Participants**

Participants were patient-volunteers undergoing surgical treatment at the University of Iowa Hospitals and Clinics for medically intractable epilepsy or brain tumors who consented to participate in research. Patient participants were undergoing treatment with implanted ECoG electrodes for seizure focus determination or for required awake craniotomy. All procedures were approved by the University of Iowa Institutional Review Board. Patients were confirmed to be left-lateralized for language production with Wada testing and/or were confirmed to be right-handed or ambidextrous (Extended Data Table 2).

#### **Behavioral Tasks**

#### Three behavioral tasks were included in this experiment:

(1) Critical information (CI) task: CI questions were adapted from an established Dutch stimulus set<sup>33, 36</sup>. We developed 120 similarly structured English questions which were divided among three categories (animals, body parts, and antonyms). Each category features pairs of questions with the same CI in which the CI was presented either early or late (e.g., Fig. 1c–f). The full list of CI questions is provided in Extended Data Table 1.

CI questions were read to the participant by an experimenter, who instructed the participant to respond as quickly as possible. Questions were presented in random order and varied slightly in exact wording from experiment to experiment. All CI questions could be answered with a single word, although participants occasionally responded with a short phrase.

(2) Command-response (CR) tasks: The four CR tasks were performed in a randomized order interleaved with CI task trials in all cases except one, where it was completed as a separate block (Extended Data Table 2). A full list of all CR stimuli is presented below in Extended Data Table 1. In CR1 (hand movements), participants performed movements that engage muscles unrelated to speech production. Specifically, participants pressed a spring-loaded button using their right hand a variable number of times, ranging from 1 to 4. In CR2 (orofacial movements), participants were instructed to carry out nonspeech orofacial tasks, specifically smiling, mouth opening, tongue protrusion, and lip puckering. The movements were recorded with a video camera and time-stamped with a common time scale to the ECoG recordings. In CR3 (speech repetition), participants repeated monosyllabic real words and/or monosyllabic or disyllabic pseudowords. In CR4 (pluralization), participants were instructed to produce the irregular and/or regular plural form of common monosyllabic nouns (e.g., 'goose'  $\rightarrow$  'geese'). See Extended Data Table 2 for the specific tasks completed by each participant.

(3) Unconstrained conversation: Following completion of the CI and CR tasks, the experimenter engaged the participant in unconstrained conversation. Participants were given no specific instructions during this portion of the experiment other than to speak naturally.

Data Acquisition—For awake craniotomy patients, local field potentials were recorded with either a custom 64 channel grid engineered at the University of Iowa or with commercially available subdural grids manufactured by Ad-Tech Medical or PMT. Signals were amplified, bandpass filtered (0.7-800 Hz) and sampled at 2034.5 Hz using a multichannel amplifier and digital acquisition system (PZ2 preamplifier and RZ2 processer; Tucker-Davis Technologies). For chronically implanted epilepsy patients, electrophysiological signals from subdural electrode grids and strips (AdTech) were bandpass filtered (0.1-500 Hz) and recorded at 2000 Hz with a multichannel amplifier and digital acquisition system (Atlas system, Neuralynx). Analog input channels synchronized with the neural recordings additionally marked the timing of participant button presses and the output of a microphone which captured the speech acoustics of the experimenter and participant. Input channels were sampled at 48,828 Hz by the TDT system and 16,000 Hz by the Neuralynx system and downsampled offline to 12,000 Hz. In addition to the electrical signals, a video of the participant was also acquired at 24 fps during all experiments. The video was synced to the electrophysiological data after the experiment and provided a secondary high-quality audio recording channel of the experiment, which was sampled at 48 kHz.

#### **Behavioral Analysis**

The audio acquired with the electrophysiological acquisition system and/or camera was then annotated by a trained phonetician (G.A.C.) to determine the onsets and offsets of all experimenter questions/commands, CI, and participant responses for the CI and CR tasks. Trials where the participant failed to respond, requested additional information (e.g., asking for the question to be repeated), or produced a filled pause or disfluency (e.g., saying 'um' or stuttering) prior to responding were excluded from further analysis. Trials in the CR task where a participant failed to accurately complete the command (e.g., repeating a word when instructed to say the plural form of the word) were deemed incorrect and excluded from analysis. In the CR task, the onset and offset of hand movements were determined using the electrical signal from the button which was acquired simultaneously with the local field potentials and were defined as the onset of the first press and offset of the last press of the trial, respectively. For the CR trials requiring orofacial movement responses, the onset and offset of movements were estimated as the timing of the first and last frames of the video where movements were clearly observed.

The audio of the unconstrained conversation was also annotated to provide timestamps for the onsets and offsets of all experimenter and participant turns. Turns were considered continuous segments of speech that were uninterrupted by the other speaker. However, experimenter turns were occasionally truncated at silent pauses (i.e., unfilled pauses) if the experimenter spoke continuously for periods in excess of 10 seconds. Silent pauses produced by the participant were annotated within the participant's turns. Additionally, during the participant's speech, the experimenter sometimes produced backchannels (e.g., 'okay', 'yes', to display interest) that would not result in termination of the participant's turn. In such cases, we did not consider these events turn-taking on behalf of the participant as they did not yield the floor. Interactions where the participant responded with a backchannel were noted (e.g., Extended Data Fig. 6a) but were defined as turn-taking events on behalf

of the participant. Finally, interactions where speaker overlap rendered reliable annotation impossible were not considered for analysis.

#### Anatomical Reconstructions

**Electrode localization in intraoperative patients**—In intraoperative cases, electrode localization and coregistration was performed using intraoperative photographs and perioperative T1-weighted MRIs obtained over the course of the clinical workup. Magnetic resonance images were processed using the 'recon-all' pipeline in FreeSurfer to generate cortical surface meshes and to obtain surface-based coregistration to an anatomical template space<sup>76</sup>. Intraoperative photographs of the craniotomy and electrode arrangement were then aligned to renderings of the surface meshes according to a visual comparison of gyral anatomy. Image-based localization was carried out by two raters (G.A.C. and C.K.K.) and, when discrepant, reconciled by an additional rater (J.D.W.G.). Finally, a subset of electrode locations was identified in image RAS coordinate space by selecting the vertex within the surface mesh nearest the electrode location, as observed in the aligned intraoperative photograph. Remaining locations were determined through thin-plate spline (TPS) interpolation, according to the geometry of the electrode grid.

**Electrode localization in chronically implanted patients**—Patients with chronically implanted electrodes underwent CT and MR imaging before and immediately after implantation. Electrode locations were identified in post-implantation images based on characteristic metallic artifacts: localized magnetic susceptibility-related voids in MR images and punctate radiodensities in CT images. Electrode coordinates were transformed to the pre-implantation image through an initial linear image coregistration followed by a manually guided TPS warping. Coordinates aligned to corresponding anatomical landmarks in pre- and post-implantation imaging, identified through visual comparison of the linearly coregistered images, served as control points in the TPS warping

**Electrode coregistration**—For all patients, anatomical categorization of electrode sites was guided by surface-based coregistration and segmentation implemented in Freesurfer<sup>77</sup> using the DKT atlas<sup>78, 79</sup>. Following automatic parcellation, the location of each electrode according to the DKT atlas labels was confirmed by three raters (G.A.C., C.K.K., J.D.W.G.) and corrected if necessary. Electrode locations were then transformed into MNI space using symmetric diffeomorphic registration implemented in the ANTs toolbox<sup>80</sup> to obtain a nonlinear coregistration between the patient's pre-operative T1 image and an MNI-aligned template brain (CIT168 template)<sup>81</sup>.

For analyses where electrodes from individual participants were rendered a canonical cortical surface (e.g., Fig. 2c), electrodes were plotted on the gyral surface of the MNI152 brain. For analyses of electrode response density (e.g., Fig. 2f), the proportions of electrodes displaying a given response profiles in 1 cm diameter regions of the canonical cortical surface were calculated for areas with electrode coverage from at least 3 participants.

Prior to preprocessing each participant's electrophysiological data, nonresponsive channels were first identified and excluded (Extended Data Table 3). Such channels were clearly identifiable as they displayed signal variances markedly lower than active channels. Next, artifactual periods were identified in the active channels by iteratively thresholding at an absolute z-score threshold of 9 (calculated across the entire recording) and blanking all signals within 2 seconds of the suprathreshold period. Z-scores were then recomputed, and the procedure was repeated with undiscarded samples until all samples fell below the threshold. Any CI or CR trials occurring within one second of a blanked period were not considered for analysis.

Next, the electrophysiological data were preprocessed by filtering stationary and nonstationary line noise using adaptive thresholding applied to coefficients of the demodulated band transform<sup>82</sup> with a bandwidth parameter of 0.25 Hz. The data were then high-pass filtered at 5 Hz with a FIR filter and re-referenced to the common average signal on a per-grid basis<sup>83</sup>. Finally, high gamma band amplitude was calculated for each channel by averaging the analytical envelope across 8 frequency bands of logarithmically increasing center frequency (73–144 Hz) and standard deviations (4.68–6.62 Hz)<sup>50</sup> and resampled at 500 Hz. As artifacts were commonly observed at the beginning and end of recordings, these periods (the first and last ~5 seconds) were blanked in both the raw and high gamma signals. Finally, the high gamma signals for each channel were z-scored across the entire recording duration. Only electrodes located on the cortical surface were included in the analyses described in this paper. Any sites within white matter, not contacting the brain, or deep within cortical or subcortical structures were not considered, and any electrodes determined to have been located on seizure foci or tissue included in the subsequent resection were excluded from further analysis.

#### Active Site Detection using Generalized Linear Modelling

To identify electrodes displaying signals significantly correlated to speech perception, planning, and production in the CI task, we used mass univariate generalized linear modelling (GLM) in an approach similar to the statistical parameter mapping procedure used in functional MRI<sup>84</sup> and ECoG<sup>85</sup> studies. For each participant, we first constructed four block regressors using the temporal structure of each CI trial (depicted in Extended Data Fig. 1f): 1) a regressor which was active from experimenter question onset to participant answer offset, which was included to regress out any activity that was globally correlated with trial engagement but not with any specific feature of the task (e.g., attention); 2) a regressor which was active from question onset to question offset to identify activity correlated with speech perception; 3) a regressor which was active from CI offset to 200 ms prior to answer onset (as articulatory movements not immediately resulting in an acoustic consequence can occur hundreds of milliseconds prior to the acoustic onset of speech<sup>31, 32</sup>) to identify activity correlated with planning; and 4) a regressor which was active from 200 ms prior to answer onset until answer offset to identify activity correlated with participant speech production. A 250 ms buffer where no regressors were active was included prior to trial onset and after trial offset. The regressors for each trial along with the associated high gamma signals (downsampled to 100 Hz for computational efficiency) were then

concatenated and GLM fitting was performed. This analysis was carried out using the 'glmfit' function in MATLAB (Natick, MA, USA) with a normal distribution and identity link function specified (therefore, general linear modelling was specifically performed).

To assess significance, we used a resampling method where the GLM was iteratively performed on a randomly selected subset of 80% of trials over 100 repetitions with replacement. This resampling technique was then repeated to apply the trial structure to random epochs of high gamma data whose duration matched the actual data. This allowed us to generate empirical distributions of actual regressor weights, weight p-values, and full-model correlation coefficients as well as null distributions for analogous values generated with shuffled signals. For an electrode's individual regressor weights to be considered significant, two criteria must have been met: 1) the two-tailed 95% confidence interval obtained from the empirical distribution of individual regressor weights could not include 0 or the mean shuffled weight for that regressor and 2) the logscale distribution of the weight p-values could not include the Bonferroni-corrected alpha value of 0.05 at the 95<sup>th</sup> percentile.

To confirm that the responses of individual electrodes were well-fit by the specific temporal structure of the CI questions, we used a jittering analysis. Specifically, we performed the resampling technique described above on high gamma signals that were incrementally shifted in uniform 500 ms steps from zero lag to an absolute maximum jitter of 10 seconds, and the mean full model R value over 100 repetitions was recorded at each step for all electrodes (Extended Data Fig. 2a). We then collected random subsets (with replacement) of R values from the jittered models such that all participants contributed data from an equal number of electrodes (n = 44, 75% of the count from the participant with the fewest electrodes). To quantify the temporal precision of each electrode's fit to the CI task data, we subtracted the median value of large jitters (-10 to -4 seconds) from the maximum R value of small jitters (-2 to 2 seconds), a value we refer to as delta (D). We included only large negative jitters in the calculation of D to avoid inadvertently fitting activity related to the subsequent trial at high positive jitters with the short duration regressor corresponding to the participant's answer (~0.5 s). We observed that the pooled distribution of D values across participants was bimodal, with electrodes whose activity was fit poorly to the temporal structure of the CI task forming a cluster that was distinct from electrodes whose activity fit the task structure well. To isolate these two distributions, we then fit the D distribution with a mixture of two Gaussians after excluding values above the 95<sup>th</sup> percentile to reject outliers (Extended Data Fig. 2b). We then used the Gaussian representing the distribution of poorly fit electrodes to empirically define a D threshold by calculating the value at which 95% would be excluded (1.96 standard deviations from the mean) (Extended Data Fig. 2b). This procedure was repeated 1000 times and the median D threshold was calculated across iterations. Any electrode whose D value was below this threshold value (0.126) was excluded from further analysis. After thresholding, we found that the number of rejected sites within DKT-defined regions was significantly anti-correlated with the number of responsive electrodes within each region (Spearman R = -0.8311, p < 0.0001; Extended Data Fig. 2c-e). Therefore, electrodes whose response profiles were not specifically matched to CI task structure were not evenly distributed across the brain, but

were more prevalent in regions outside of the language-related circuitry identified with this task.

Electrodes which met all significance testing criteria were then classified based on their mean regressor weights. Electrodes possessing a single positive significant perception, planning, and production weights were deemed unmixed perception, planning, and production sites, respectively. Electrodes possessing a combination of significant positive perception, planning, and production weights were categorized as 'mixed'. Significantly negative regressor weights did not affect electrode classification.

#### **Reduced GLM Analyses**

Because we observed far fewer perception electrodes than planning electrodes, we performed two additional GLM analyses to assess whether the overlapping structure of the perception and planning periods during early CI questions biased the GLM to detect planning-related activity at the expense of perception-related activity. To do so, we first performed the GLM analysis described above but omitted the planning regressor. For this reduced model, we held the D threshold and individual electrode D values constant in order to maintain consistent exclusion criteria with respect to the full model. We then compared the number of perception-related electrodes detected with the reduced model to that detected with the full model (Extended Data Fig. 3c). We then performed a parallel analysis with a reduced GLM which omitted the perception regressor. With these reduced models, we assessed 1) the number of planning electrodes (as detected with the full model) that gained a significant perception response when the planning regressor was omitted (Extended Data Fig. 3d), and 2) the number of perception electrodes (as detected with the full model) that gained a significant planning response when the perception regressor was omitted (Extended Data Fig. 3e). This analysis allowed us to assess whether the planning regressor also possessed explanatory value for perception-related activity and vice versa.

#### Analysis of CI Question Types

To assess whether planning-related electrodes displayed significantly elevated activity following CI in both 'early' and 'late' CI trials (Extended Data Fig. 1d) as well as in each of the three question categories (Extended Data Fig. 1e), we averaged high gamma signals aligned to CI for each of these categories. We then calculated peak-to-trough amplitude of each mean response by subtracting the 25<sup>th</sup> percentile value of the z-scored high gamma activity during the 2 seconds prior to CI onset from the 75<sup>th</sup> percentile value of the z-scored high gamma activity during the 2 seconds following CI onset. We then performed the same procedure using random timepoints rather than CI onset times to generate parallel shuffled measures of peak-to-trough response amplitudes for each condition. Data from all mixed and unmixed planning electrodes were pooled across participants to generate distributions of actual and shuffled response amplitudes for each condition, and differences between these distributions were assessed using a signed-rank test.

#### CI Clustering Analyses in GLM Weight Space

To assess whether functionally-relevant classes of electrodes (i.e., perception, planning, and production) formed distinct clusters in GLM weight space, k-medoids clustering was

performed on all responsive electrodes pooled across all participants with the 'kmedoids' function in MATLAB. Clustering was specifically performed in three dimensional GLM weight space (i.e., perception, planning, and production weights), and three clusters were assumed.

Using a simulation, we then assessed whether the observed clustering of functional electrode classes in GLM weight space could have resulted from subdividing continuously varying patterns of neural activity. To do this, we generated 253 simulated electrodes (i.e., the number of observed responsive electrodes in our data set) which were randomly assigned a perception weight, planning weight, and production weight. The simulated weights for each regressor were generated from independent unimodal normal distributions whose means and standard deviations matched the observed values across responsive electrodes (perception:  $-0.119 \pm 0.313z$ ; planning:  $0.245 \pm 0.238z$ ; production:  $0.220 \pm 0.377z$ ). Simulated weights that surpassed the 5<sup>th</sup> percentile of observed significant weights for each regressor (perception: 0.229z; planning: 0.164z; production: 0.205z) were considered "responsive" in the model and subjected to further analysis. This method allowed us to match the range of regressor weights of our simulated responsive electrodes to that observed in the actual data while assuming that regressor weights reflect underlying noncategorical, unimodal continua of neural activity as opposed to the discrete response profiles we hypothesized. Next, simulated responsive electrodes were classified as unmixed and mixed perception, planning, and production sites in the same manner as was performed for the actual data, and we quantified 1) the proportion of mixed electrodes, and 2) the proportion of "misclustered" unmixed electrodes. To identify misclustered electrodes, each simulated cluster first received an assignment as perception, planning, or production; this assignment was determined such that the maximal number of simulated unmixed perception, planning, and production electrodes fell the perception, planning, and production cluster, respectively, and each cluster contained at least 1 unmixed electrode whose type matched its assignment (i.e., the planning cluster must contain at least one unmixed planning electrode). The proportion of misclustered unmixed sites was then calculated by summing the number of simulated unmixed perception, planning, and production electrodes falling outside of the perception, planning, and production clusters, respectively, and dividing by the total number of simulated unmixed electrodes. This analysis was repeated over 100,000 iterations to generate empirical distributions of mixed and misclustered unmixed electrode proportions under the unimodal regime to compare against the observed results. Data from all iterations was considered unless the model failed to generate at least one unmixed electrode of each class.

#### Assessing the Speech-Selectivity of Planning Sites

To quantify the specificity of planning responses, another GLM analysis was performed on the CR task data. This analysis was identical to the GLM performed for the CI task, except four planning regressors (CR1-CR4) and three production regressors (CR1 [hand movements], CR2 [nonspeech orofacial movements], CR3&4 [speech]) were employed. As the temporal precision of each site was quantified with the CI task, no jittering analysis was performed. Also, as the CI in this task were longer in duration (e.g., "open your mouth" vs. a single word), CI onset was defined as the onset of the planning regressor. This GLM analysis

was performed on all unmixed planning and mixed planning-production electrodes (referred to as "planning electrodes" in the context of the CR task). Mixed planning sites possessing a significant positive perception GLM weight were not included in order to avoid biasing the analysis with unintended acoustic differences between the four CR trial types. The mean planning regressor weight was then used as a metric for selectivity for each of the four CR response types.

We assessed the speech-selectivity of planning electrodes within individual regions by calculating the proportion of planning electrodes in 1) caudal IFG (i.e., pars triangularis and pars opercularis), 2) caudal MFG, and 3) precentral gyrus which displayed a significant positive planning weight in each of the four CR tasks; the boundaries these regions were defined according to the DKT atlas<sup>78, 79</sup> (Extended Data Table 3). To assess the significance of these proportions, we performed a resampling analysis where the planning weights observed across all planning electrodes for each CR task were shuffled and randomly assigned to the electrodes in each of these three regions. The proportion of shuffled significant positive responses were then calculated within each region and the process was repeated 100,000 times in order to determine the distribution of proportions that would be expected by chance; due to differences in electrode coverage, this null distribution varied between regions. Any proportions greater than the 97.5<sup>th</sup> percentile and lower than the 2.5<sup>th</sup> percentile of each region's null distribution were therefore considered significantly different than chance.

#### Principal Component Analysis

#### Calculation of Task-Derived and Conversation-Derived Principal Components

-To reduce the dimensionality of neural activity, principal component analysis (PCA) using singular value decomposition was performed on the unaveraged high gamma signals from all CI-task responsive electrodes (i.e., each electrode is a variable with timepoints as observations). Specifically, we divided the task and conversation epochs of each participant's data set - which occurred sequentially during the experiments - and performed PCA (implemented with the 'pca' function in MATLAB) on the data from each period independently for all participants. The task period was defined as the epoch beginning 5 seconds prior to the onset of the first CI or CR trial and ending 5 seconds after the last CI or CR trial. In participants where the task began prior to the start of the electrophysiological recording or during a blanked artifactual period, the task period was defined to begin at the first unblanked sample. The conversation period was defined as the period beginning 5 seconds prior to the first experimenter turn onset and ending 5 seconds after the last participant turn offset. In cases where conversation was still ongoing when the recording was terminated or during a final artifactual period, the conversation period was defined to end at the last unblanked sample. Finally, prior to performing PCA, each electrode's high gamma signal was smoothed with a 250 ms mean boxcar filter and z-scored within the task and conversation epochs independently.

**Clustering of Task-Responsive Electrodes in PC Coefficient Space**—To assess network organization in an unsupervised manner, we mathematically identified clusters of task-responsive electrodes for each participant possessing at least 2 unmixed

electrodes of different classes (n = 6; see Extended Data Table 2) using k-medoids clustering (implemented with the 'kmedoids' function in MATLAB). To standardize this analysis across participants, clustering was specifically performed in three-dimensional PC coefficient space using the scalar coefficients (i.e., loadings) of the first three PCs while assuming three clusters, as visual inspection revealed at least three main electrode clusters in all cases (Extended Data Fig. 7a-f). The coefficients of the first three PCs were used because 1) these PCs individually explained at least 5% of the variance for the task data in all participants and for the conversation data in 7/8 participants while also cumulatively explaining at least 1/3 of the variance in all participants for both behaviors, 2) these PCs fell significantly above the linear decay phase in the scree plots for the task and conversation data sets in 7/8 and 6/8 participants, respectively, and 3) when all participant data were pooled, the first three PCs individually explained at least 5% of the variance and cumulatively explained at least 1/3 of the variance on average for both the task and conversation while also falling significantly above the linear decay phase (Extended Data Fig. 6c). Note that the linear decay phase was estimated for each participant using the 95% confidence interval of a line fit to the center of the scree plot (i.e., data from the middle 50% of PCs centered at the half maximum PC) via the 'fitlm' function in MATLAB thus allowing for the "elbow" of the scree plot to be empirically defined. This process was performed across participants by similarly fitting the pooled data from the middle 50% of the first 13 PCs (the number possessed by the participant with the fewest task-response electrodes, and consequently, PCs) from all participants (Extended Data Fig. 6c).

To assess whether the correlational structure among electrodes was stable between the task-based behavior and natural conversation, we then identified electrode clusters in the conversation time period which were analogous to those observed during the task period. Specifically, analogous clusters were defined as the task cluster and conversation cluster who possessed the highest number of common electrodes. For all six participants, each task-related electrode cluster was analogous to a single, unique cluster in the conversation data such that 18 pairs of analogous clusters were identified (Extended Data Fig. 7). We then calculated the number of electrodes which 'switched' clusters between the task and conversation data sets (i.e., electrodes not remaining within a pair of analogous clusters), and we determined whether observed number of electrode switches for each participant and the total across participants were significantly less than expected by chance using a permutation test where the conversation cluster membership was shuffled for each participant over 1,000 iterations (Extended Data Fig. 6d).

**Functional Categorization of Electrode Clusters**—To determine whether electrode clusters in PC coefficient space corresponded to the perception, planning, and production electrode classes defined using the CI task and GLM, we examined the GLM-defined classification of electrodes within each cluster. To do this, we tallied the number of unmixed perception, planning, and production electrodes (i.e., electrodes with a single significant positive GLM weight) in each cluster and functionally categorized clusters according to which electrode class was most numerous. Across the six participants, 17/18 analogous cluster pairs were thus defined as either a perception, planning, or production cluster using this method (analogous clusters always received the same designation); one cluster pair

received two categorizations (Extended Data Fig. 7d) and was not included in further analyses.

We next assessed whether the unsupervised clustering analysis successfully recovered the functional classes of neural activity identified with the GLM by assessing the degree to which electrodes within a cluster in PC coefficient space exhibited GLM-defined neural activity congruent with the functional categorization of that cluster. Specifically, we calculated the percentage of electrodes across participants in conversation-related perception, planning, and production clusters that displayed significant GLM-defined perception, planning, or production activity, respectively. To determine whether the observed values across participants were higher than expected by chance, we assessed significance using a permutation test where the GLM activity classes of each electrode were shuffled over 1,000 iterations (Extended Data Fig. 6e).

Functional Categorization of Principal Components—To examine neural activity related to speech perception, planning, and production during conversation, we assessed time-varying PC scores (i.e., the linear combination of electrodes summed in time according to their PC coefficients) as a proxy for the aggregate activity of electrodes within perception, planning, and production-related clusters. To first determine which PCs corresponded to each electrodes class, we examined the distribution of PC coefficients displayed by electrodes within each cluster category for each participant. Specifically, we determined which of the first three PCs displayed the highest average coefficient value for electrodes in perception clusters, planning clusters, and production clusters, and defined these PCs as perception, planning, and production PCs, respectively. In two participants where the vast majority of task-related responses were defined as planning (472, 100%; 510, 97.5%), we considered the PC with the highest average coefficient across all planning-related electrodes as the planning PC (Extended Data Fig. 7g,h). Finally, to avoid assessing PCs representing mixed response profiles, we did not include any clusters containing 50% mixed response electrodes (e.g., a planning cluster containing 5 unmixed planning and 5 mixed planningproduction electrodes) in this analysis (n = 3/17 clusters; see Extended Data Table 2 and Extended Data Fig. 7a-f). Using this approach, we isolated 2 perception PCs, 6 planning PCs, and 5 production PCs across 8 participants in both the task and conversation data sets (Extended Data Fig. 7).

We then assessed the average response profiles of the perception, planning, and production PCs by calculating 1) the mean planning and perception PC scores aligned to experimenter speech offset in the CI task and conversation, and 2) the average planning and production PC scores aligned to participant speech onset in the CI task and conversation. Periods of significantly elevated average PC activity (i.e., PC scores) were identified using a permutation test where average responses for each PC were calculated when aligned to trial number-matched random timepoints over 1,000 iterations; any time bins of the average response which fell above the 95<sup>th</sup> percentile value of the shuffled average responses were considered significant.

Lastly, we quantified PC activity with respect to unconstrained turn-taking behavior in natural conversation by calculating the median z-scored PC score amplitude during all

1) experimenter turns, 2) intra-turn gaps, and 3) participant turns. Consistent with the previous analyses, participant turn onset was defined as 200 ms prior to the acoustic onset of participant speech to account for silent articulatory movements. Due to this correction, any gaps less than 200 ms in duration were not considered for analysis. Finally, for all participants possessing multiple functionally categorized PCs (n = 4), the difference in median amplitude between PCs was assessed for significance using a signed-rank test (Extended Data Fig. 6g).

# **Extended Data**



#### Extended Data Fig. 1 |. Behavior during the CI task.

a, Description of subprocesses assumed to occur during the perception, planning, and production windows of the CI task.
b, Histograms of reaction times (RT) in early and late CI trials for all participants.
c, Median RT values for early and late CI trials for all participants.
d,e, Histograms depicting the distribution of average peak-to-trough response amplitudes for all electrodes displaying planning-related responses when aligned to CI onset in early and

late trials (**d**) and different CI question types (**e**); median values for each distribution are indicated. Observed data (in black) are compared with a null distribution (in grey) consisting of randomly chosen timepoints (see Methods). **f**, Schematics displaying GLM regressor structure for early (top) and late (bottom) trials of the CI task.



# e Most Rejected Sites are Outside of Regions Displaying Task-Related Activity



#### Extended Data Fig. 2 |. GLM temporal jittering analysis.

**a**, Full model R values for GLM fits of jittered high gamma activity from participant 436; each line represents data from an individual electrode. **b**, Example distribution of

pooled D values with the fit of two Gaussians overlaid (black). The Gaussian distributions corresponding to well-fit electrodes (blue) and poorly fit electrodes (red) as well as the 95th percentile of the D distribution for poorly fit electrodes (dashed line) are indicated. D values above the 95th percentile of the pooled distribution were deemed outliers (white bars) and not fitted. **c**, Table summarizing the number of electrodes rejected by the jittering analysis in each participant. **d**, Table reporting the anatomical locations of electrodes rejected by the jittering analysis and electrodes displaying significant activity in the CI task. **e**, Scatterplot depicting the proportion of rejected electrodes within a region as a function of the proportion of responsive electrodes in a region.

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a,b, Histograms depicting the distribution of the proportion of misclustered electrodes responsive during a single task window (i.e., unmixed electrodes) (a), and the proportion of electrodes displaying more than one significant positive weight (i.e., mixed electrodes) (b) across 100,000 iterations of the continuum model simulation. The median of each distribution as well as the values observed in the actual data (dashed line) are indicated. Gold arrows indicate the bin of each distribution containing the measurements corresponding to the example iteration depicted in panels p, r, and t of Figure 1. c, Table

reporting the number of electrodes displaying perception-related responses using either the full model or the reduced GLM lacking a planning regressor. **d,e**, Scatterplots depicting perception (**d**) and planning (**e**) GLM weights in the full model and reduced models lacking a planning regressor or perception regressor, respectively. Significant positive weights are denoted with filled points and nonsignificant or significant negative weights are denoted with unfilled points; the x-coordinates of each point are randomly jittered by 25% to better visualize filled versus unfilled status. No planning electrodes displayed significant perception responses in the reduced GLM lacking a planning regressor, and no perception electrodes displayed significant planning responses in the reduced GLM lacking a planning regressor.

а	Summary of significant GLM weights in task-responsive electrodes											
		# Electrode	es with Sig.	# Electrode	es with Sig.	# Electrodes with Sig.						
	GLM Class	Perceptic	on Weight	Planning Weight		Production Weight						
		Positive	Negative	Positive	Negative	Positive	Negative					
	Perception	36	0	11	10	11	4					
	Planning	11	75	160	0	45	17					
	Production	11	41	45	8	121	0					

b

Analyze first 500 ms of question



Trial



#### Extended Data Fig. 4 |. Additional analyses of task-related activity changes.

**a**, Table reporting the number of perception, planning, and production-related electrodes displaying significant positive and negative weights for each GLM regressor. **b**, Histogram depicting mean high gamma amplitude in the first 500 ms of CI questions for all unmixed perception, planning, and production electrodes. **c-d**, Canonical cortical surfaces displaying electrodes with significant positive (colored) or negative (black) GLM weights in the perception (**c**), production (**d**), and planning (**e**) windows of the CI task across all participants. Electrode diameter is scaled to the absolute magnitude of the GLM weight, and electrodes not displaying a significant weight for a given regressor are indicated with small white circles.



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#### Extended Data Fig. 5 |. Anatomical analysis of responses.

a, Cortical reconstructions for all participants displaying the location of all electrodes;
the size of each electrode depicts the actual size of its recording area on the cortical surface. GLM classification is indicated by electrode color. b, Canonical cortical surfaces showing electrode locations from all participants as standard-sized white circles. c, Number of electrodes sampling each area of the canonical cortical surface (1 cm diameter spatial smoothing) after pooling electrodes from all participants. d, Proportion of electrodes displaying significant production-related responses in the CI task (1 cm diameter spatial

smoothing). **e**, Canonical cortical surfaces displaying electrodes with significant responses related to speech perception, production, and planning in tumor (top) and epilepsy (bottom) patients separately. Electrodes not displaying a significant response for a process are depicted as small white circles. Electrode diameter is scaled to GLM regressor weight.



Extended Data Fig. 6 |. Additional conversation-related analyses.

**a**, Table reporting additional turn-taking behavioral measures for each participant. **b**, Histograms of gap durations (time between experimenter turn offset and participant turn

onset) during unconstrained conversation for each participant; bins are centered on 100 ms increments with a width of 100 ms. c, Scree plots for the PCA analysis of high gamma signals in the task (left) and conversation (right) periods of the recordings; data from each participant are represented by thin lines and the average across participants is denoted with a thick black line. The 95% confidence interval of the linear decay phase across participants (see Methods) is also indicated. d, The observed number of electrodes whose cluster membership was not stable (i.e., switched clusters) between the task and conversation with a histogram depicting the distribution of electrode cluster switches expected by chance. e, The observed percentage of electrodes in perception, planning, and production clusters (in conversation-derived PC coefficient space) displaying significant perception, planning, and production responses (per the GLM), respectively, with histograms depicting the percentages expected by chance for each cluster type. f, Canonical cortical surfaces displaying the locations of all electrodes in perception, planning, and production clusters across participants (n = 6) in the task (left) and conversation (right). g, Table reporting summary statistics for PC activity (i.e., time-varying PC score) during unconstrained conversation for each participant.

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#### Extended Data Fig. 7 |. PCA results for individual participants.

**a-f**, For 6 participants possessing sufficient numbers of electrodes belonging to multiple GLM classes (see Methods): scatterplots depicting electrode distributions in PC coefficient space in the task and conversation periods (top row). Bar graphs depicting the PC coefficients for all electrodes in perception clusters, planning clusters, or production clusters from the PCA performed on task data and conversation data (bottom rows). **g,h**, For 2 participants possessing mainly planning electrodes (see Methods and Extended Data Table 2): bar graphs depicting the PC coefficients for all planning-related electrodes from the

PCA performed on task data and conversation data. In the bar graphs, the functional categorization of PCs is indicated by filled bars with either green (perception), blue (planning), or red (production). Any clusters rejected due to a high proportion (50%) of mixed electrodes are indicated with gray filled bars.

#### **Extended Data Table 1**

List of all task stimuli

Critical Information Task								
Question Type	Early critical information	Late critical information						
	What animals, who <b>cluck</b> , are often seen on farms or in zoos?	Which animals, who are often seen on farms or in zoos, <b>cluck</b> ?						
	What animals, who <b>moo</b> , are often found either on farms or in zoos?	Which animals, who are often seen either on farms or in zoos, <b>moo</b> ?						
	Which animals, who <b>oink</b> , are often found either on farms or in zoos?	Which animals, who are often found either on farms or in zoos, <b>oink</b> ?						
	Which animals, who <b>neigh</b> , are often found either on farms or in zoos?	What animals, who are often seen either on farms or in zoos, <b>neigh</b> ?						
	What animals, who <b>quack</b> , are often found either on farms or in zoos?	Which animals, who are often seen on farms or in zoos, <b>quack</b> ?						
	What animals, who <b>meow</b> , are often seen either on farms or in zoos?	What animals, who are often found on farms or in zoos, <b>meow</b> ?						
	What animals, who <b>bark</b> , are often seen either on farms or in zoos?	Which animals, who are often found either on farms or in zoos, <b>bark</b> ?						
	Which animals, who <b>ribbit</b> , are often found on farms or in zoos?	Which animals, who are often found on farms or in zoos, <b>ribbit</b> ?						
	What animals, who <b>tweet</b> , are often seen on farms or in zoos?	Which animals, who are often seen on farms or in zoos, <b>tweet</b> ?						
ú	What animals, who <b>roar</b> , are often seen on farms or in zoos?	What animals, who are often seen on farms or in zoos, <b>roar</b> ?						
<u>Animals</u> *	Which animals, who <b>howl at the moon</b> , are often seen either on farms or in zoos?	What animals, who are often found either on farms or in zoos, <b>howl at the moon</b> ?						
	Which animals, who <b>laugh like people</b> , are often found on farms or in zoos?	Which animals, who are often found either on farms or in zoos, <b>laugh like</b> <b>people</b> ?						
	What animals, who can <b>imitate speech</b> , are often found either on farms or in zoos?	What animals, who are often found on farms or in zoos, <b>can imitate speech</b> ?						
	What animals, who <b>hiss</b> , are often found on farms or in zoos?	What animals, who are often found either on farms or in zoos, <b>hiss</b> ?						
	Which animals, who <b>squeak</b> , are often seen either on farms or in zoos?	Which animals, who are often seen on farms or in zoos, <b>squeak</b> ?						
	Which animals, who <b>baa</b> , are often seen either on farms or in zoos?	What animals, who are often found on farms or in zoos, <b>baa</b> ?						
	Which animals, who <b>gobble</b> , are often seen on farms or in zoos?	What animals, who are often seen on farms or in zoos, <b>gobble</b> ?						
	Which animals, who <b>crow at sunrise</b> , are often seen either on farms or in zoos?	What animals, who are often seen either on farms or in zoos, <b>crow at sunrise</b> ?						
	Which animals, who <b>hee-haw</b> , are often found on farms or in zoos?	What animals, who are often seen either on farms or in zoos, <b>hee-haw</b> ?						
	What animals, who <b>honk</b> , are often found on farms or in zoos?	Which animals, who are often found on farms or in zoos, <b>honk</b> ?						

	The opposite of <b>soft</b> is what frequent word?	Which common word is the opposite of <b>soft</b> ?		
	The opposite of <b>nice</b> is which common word?	What word is the opposite of <b>nice</b> ?		
	The opposite of <b>sad</b> is what familiar word?	Which familiar word is the opposite of <b>sad</b> ?		
	The opposite of <b>hot</b> is what common word?	What word is the opposite of hot?		
	The opposite of <b>hungry</b> is which familiar word?	Which familiar word is the opposite of <b>hungry</b> ?		
	The opposite of <b>fast</b> is what common word?	Which word is the opposite of <b>fast</b> ?		
	The opposite of <b>small</b> is which word?	What word is the opposite of small?		
	The opposite of <b>first</b> is what word?	Which word is the opposite of <b>first</b> ?		
	The opposite of <b>rich</b> is what frequent word?	Which frequent word is the opposite of <b>rich</b> ?		
	The opposite of <b>easy</b> is which frequent word?	What common word is the opposite of <b>easy</b> ?		
	The opposite of <b>pretty</b> is what common word?	Which frequent word is the opposite of <b>pretty</b> ?		
	The opposite of <b>wet</b> is which frequent word?	What frequent word is the opposite of <b>wet</b> ?		
	The opposite of <b>strong</b> is which common word?	What common word is the opposite of <b>strong</b> ?		
Antonyms **	The opposite of <b>bad</b> is which common word?	What frequent word is the opposite of <b>bad</b> ?		
	The opposite of <b>tall</b> is which familiar word?	What familiar word is the opposite of <b>tall</b> ?		
	The opposite of <b>late</b> is which common word?	Which frequent word is the opposite of <b>late</b> ?		
	The opposite of <b>heavy</b> is which word?	What frequent word is the opposite of <b>heavy</b> ?		
	The opposite of <b>old</b> is what word?	What familiar word is the opposite of <b>old</b> ?		
	The opposite of <b>false</b> is what familiar word?	Which familiar word is the opposite of <b>false</b> ?		
	The opposite of <b>happy</b> is what word?	What word is the opposite of happy?		
	The opposite of <b>clean</b> is which word?	What familiar word is the opposite of <b>clean</b> ?		
	The opposite of <b>war</b> is what word?	What word is the opposite of war?		
	The opposite of <b>big</b> is what familiar word?	What familiar word is the opposite of <b>big</b> ?		
	The opposite of <b>stop</b> is what familiar word?	What familiar word is the opposite of <b>stop</b> ?		
	The opposite of <b>rare</b> is what familiar word?	What familiar word is the opposite of <b>rare</b> ?		
	The opposite of <b>float</b> is what familiar word?	What familiar word is the opposite of <b>float</b> ?		
	The opposite of <b>hate</b> is what word?	What word is the opposite of hate?		
	The opposite of <b>life</b> is what word?	What word is the opposite of life?		
Body Parts **	How many <b>fingers</b> does a healthy person have?	A healthy human has how many <b>fingers</b> ?		
	How many <b>toes</b> does a man have?	A human has how many toes?		

How many legs does a healthy person have?         A typical man has how many legs?           How many heads does a typical person have?         The average woman has how many heads?           How many mouths does a human have?         A woman has how many mouths?           How many eers does a human have?         A person has how many eers?           How many eers does a human have?         A person has how many eers?           How many eers does a healthy woman         A healthy mana how many shoulders?           How many eers does a man have?         A healthy mana how many shoulders?           How many eers does a typical man have?         A typical person has how many shoulders?           How many feed does a typical man have?         A typical person has how many elbows?           How many eers does a typical man have?         The average human has how many elbows?           How many feed does a typical man have?         The average human has how many elbows?           How many feed does a typical human         The average human has how many elbows?           How many feed does a typical human         The average man has how many feed feed feed feed feed feed feed fee		How many <b>arms</b> does a human being have?	The average human being has how many <b>arms</b> ?
Image: stand s		How many legs does a healthy person have?	A typical man has how many legs?
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How many mouths does a human have?         A woman has how many mouths?           How many exes does a human being have?         A person has how many exes?           How many exes does a woman have?         A healthy human being has how many shoulders?           How many shoulders does a woman have?         A healthy man has how many shoulders?           How many shoulders does a woman have?         A healthy man has how many shoulders?           How many telbows does the average person         A typical human has how many ebbows?           How many ebbows does a typical man have?         The average human being has how many           How many epebrows does a typical human         The average human being has how many           How many hearts does a typical human         A healthy human being has how many           How many notes does a typical woman         A healthy human has how many noses?           How many noses does a typical woman         A healthy human has how many noses?           How many noses does a typical woman         The average man has how many noses?           How many tops nails does the average         A healthy human has how many tops mains?           Morean finger nails does a healthy man         The average man has how many tops many have?           Mow many tops nails does the average?         A healthy human has how many tops may?           Mow many tops nails does the average?         A healthy human being has how many finger nails tore		How many <b>heads</b> does a typical person have?	The average woman has how many <b>heads</b> ?
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CR1 - Hand Movements       Click the button two times with your right       With your right hand, click the button two times.         CR1 - Hand Movements       Click the button three times with your right       With your right hand, click the button three times.         Click the button four times with your right       With your right hand, click the button three times.         Click the button four times with your right       With your right hand, click the button four times.         CR2 - Hand Movements       Stick out your tongue so it can be seen on the camera, stick out your tongue.         CR2 - Hand Movements       Open your mouth so it can be seen on the camera, open your mouth.		Click the button <b>one time</b> with your right hand.	With your right hand, click the button <b>one time</b> .
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Click the button four times with your right hand, click the button four times.       CR2 - Hand Movements     Stick out your tongue so it can be seen on the camera, stick out your tongue.       Open your mouth so it can be seen on the camera, open gour mouth.     So it can be seen on the camera, open your mouth.	CKI - Hand Wovements	Click the button <b>three times</b> with your right hand.	With your right hand, click the button <b>three times</b> .
Stick out your tongue so it can be seen on the camera, stick out your tongue.       So it can be seen on the camera, stick out your tongue.         Open your mouth so it can be seen on the camera, open camera.       So it can be seen on the camera, open your mouth.		Click the button <b>four times</b> with your right hand.	With your right hand, click the button <b>four times</b> .
Open your mouth so it can be seen on the camera.       So it can be seen on the camera, open your mouth.		Stick out your tongue so it can be seen on the camera.	So it can be seen on the camera, <b>stick out your tongue</b> .
	CR2 - Hand Movements	<b>Open your mouth</b> so it can be seen on the camera.	So it can be seen on the camera, <b>open</b> your mouth.

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Make a smile so it can be seen on the camera.         So it can be seen on the camera, make a smile.           CR3a ****. Monosyllabic Pseudoword Repetition         Repeat the sound ta so it can be heard clearly.         So it can be heard clearly, repeat the sound ta.           Repeat the sound ta so it can be heard clearly.         Repeat the sound ta so it can be heard clearly.         So it can be heard clearly, repeat the sound ta.           Repeat the sound da so it can be heard clearly.         Repeat the sound da so it can be heard clearly.         So it can be heard clearly, repeat the sound da.           CR3b **** - Disyllabic Pseudoword Repetition         Repeat the sound kaka so it can be heard clearly.         So it can be heard clearly, repeat the sound data.           Repeat the sound data so it can be heard clearly.         So it can be heard clearly, repeat the sound data.         So it can be heard clearly, repeat the sound data.           Repeat the sound data so it can be heard clearly.         So it can be heard clearly, repeat the sound data.         So it can be heard clearly, repeat the sound data.           CR3c - Monosyllabic Word Repetition         Say the word cat so it can be heard clearly.         So it can be heard clearly, say the word clearly.           CR3c - Monosyllabic Word Repetition         Say the word cat so it can be heard clearly.         So it can be heard clearly, say the word clearly.           CR3c - Monosyllabic Word Repetition         Say the word cat so it can be heard clearly.         So it can be heard clearly, say the word clearly.		<b>Pucker your lips</b> so it can be seen on the camera.	So it can be seen on the camera, <b>pucker your lips</b> .
CR3a****. Monosyllabic Pseudoword Repetition         Repeat the sound ka so it can be heard clearly.         So it can be heard clearly, repeat the sound ka.           Repeat the sound ga so it can be heard clearly.         Repeat the sound ga so it can be heard clearly.         So it can be heard clearly, repeat the sound ga.           Repeat the sound da so it can be heard clearly.         Repeat the sound da so it can be heard clearly.         So it can be heard clearly, repeat the sound ga.           Repeat the sound da so it can be heard clearly.         Repeat the sound kaka so it can be heard clearly.         So it can be heard clearly, repeat the sound kaka.           Repeat the sound kaka so it can be heard clearly.         So it can be heard clearly, repeat the sound kaka.         So it can be heard clearly, repeat the sound gaga.           Repeat the sound dada so it can be heard clearly.         So it can be heard clearly, repeat the sound gaga.         So it can be heard clearly, repeat the sound gaga.           CR3b "#**" - Disyllabic Pseudoword Repetition         Say the word cat so it can be heard clearly.         So it can be heard clearly, repeat the sound gaga.           CR3c - Monosyllabic Word Repetition         Say the word cat so it can be heard clearly.         So it can be heard clearly, say the word clearly.           Say the word dog so it can be heard clearly.         So it can be heard clearly, say the word clearly.         So it can be heard clearly, say the word clearly.           CR4a - Regular Pluralization         Say the plural of cat so it can be heard cle		Make a smile so it can be seen on the camera.	So it can be seen on the camera, <b>make a smile</b> .
CR3a         Repeat the sound ta so it can be heard         So it can be heard clearly, repeat the sound ta.           Pseudoword Repetition         Repeat the sound ga so it can be heard         So it can be heard clearly, repeat the sound ga.           Repeat the sound da so it can be heard         So it can be heard clearly, repeat the sound da.         So it can be heard clearly, repeat the sound da.           Repeat the sound kaka so it can be heard         So it can be heard clearly, repeat the sound da.         So it can be heard clearly, repeat the sound da.           Repeat the sound kaka so it can be heard         So it can be heard clearly, repeat the sound data.         So it can be heard clearly, repeat the sound tata.           Repeat the sound gaga so it can be heard         So it can be heard clearly, repeat the sound dada.         So it can be heard clearly, repeat the sound dada.           Repeat the sound dada so it can be heard         So it can be heard clearly, repeat the sound dada.         So it can be heard clearly, repeat the sound dada.           Repeat the sound dada so it can be heard clearly.         So it can be heard clearly, repeat the sound dada.         So it can be heard clearly, repeat the sound dada.           CR3c - Monosyllabic         Say the word dog so it can be heard clearly.         So it can be heard clearly, say the word cat.           Say the word dow so it can be heard clearly.         So it can be heard clearly, say the word clearly.         So it can be heard clearly, say the pural of cat.           CR4a -		Repeat the sound <b>ka</b> so it can be heard clearly.	So it can be heard clearly, repeat the sound <b>ka</b> .
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Repeat the sound da so it can be heard clearly.         So it can be heard clearly, repeat the sound da.           CR3b ****         Repeat the sound kaka so it can be heard clearly.         So it can be heard clearly, repeat the sound kaka.           Pseudoword Repetition         Repeat the sound gaga so it can be heard clearly.         So it can be heard clearly, repeat the sound dada.           Repeat the sound gaga so it can be heard clearly.         Repeat the sound dada so it can be heard clearly.         So it can be heard clearly, repeat the sound gaga.           Repeat the sound dada so it can be heard clearly.         So it can be heard clearly, repeat the sound dada.         So it can be heard clearly, repeat the sound dada.           CR3c - Monosyllabic Word Repetition         Say the word cat so it can be heard clearly.         So it can be heard clearly, say the word cat.           Say the word dogs oi t can be heard clearly.         So it can be heard clearly, say the word cearly.         So it can be heard clearly, say the word cat.           Say the word docs so it can be heard clearly.         So it can be heard clearly, say the word cearly.         So it can be heard clearly, say the word cearly.           Say the word docs so it can be heard clearly.         So it can be heard clearly, say the plural clearly.         So it can be heard clearly, say the plural of cat.           Say the plural of cat so it can be heard clearly.         So it can be heard clearly, say the plural of cat.         So it can be heard clearly, say the plural of cat.	Pseudoword Repetition	Repeat the sound <b>ga</b> so it can be heard clearly.	So it can be heard clearly, repeat the sound <b>ga</b> .
CR3b ****         Repeat the sound kaka so it can be heard clearly.         So it can be heard clearly, repeat the sound kaka.           Pseudoword Repetition         Repeat the sound gaga so it can be heard clearly.         So it can be heard clearly, repeat the sound gaga.           Repeat the sound gaga so it can be heard clearly.         So it can be heard clearly, repeat the sound gaga.           Repeat the sound dada so it can be heard clearly.         So it can be heard clearly, repeat the sound dada.           So it can be heard clearly.         So it can be heard clearly, repeat the sound dada.           So yt he word cat so it can be heard clearly.         So it can be heard clearly, say the word cat.           Say the word dog so it can be heard clearly.         So it can be heard clearly, say the word cog.           Say the word dog so it can be heard clearly.         So it can be heard clearly, say the word cog.           Say the word dog so it can be heard clearly.         So it can be heard clearly, say the word cog.           Say the word duck so it can be heard clearly.         So it can be heard clearly, say the word cog.           Say the plural of cat so it can be heard clearly.         So it can be heard clearly, say the plural of cat.           Say the plural of dog so it can be heard clearly.         So it can be heard clearly, say the plural of dog.           Say the plural of dow so it can be heard clearly.         So it can be heard clearly, say the plural of dog.           Say the plural of foot so		Repeat the sound <b>da</b> so it can be heard clearly.	So it can be heard clearly, repeat the sound <b>da</b> .
CR3b **** - Disyllabic Pseudoword Repetition         Repeat the sound tata so it can be heard clearly.         So it can be heard clearly, repeat the sound tata.           Repeat the sound dada so it can be heard clearly.         Repeat the sound dada so it can be heard clearly.         So it can be heard clearly, repeat the sound dada.           Repeat the sound dada so it can be heard clearly.         So it can be heard clearly, repeat the sound dada.         So it can be heard clearly, repeat the sound dada.           CR3c - Monosyllabic Word Repetition         Say the word dog so it can be heard clearly.         So it can be heard clearly, say the word dog.           Say the word dog so it can be heard clearly.         So it can be heard clearly, say the word dog.         So it can be heard clearly, say the word dog.           Say the word dow so it can be heard clearly.         So it can be heard clearly, say the word dclearly.         So it can be heard clearly, say the word dog.           Say the word dow so it can be heard clearly.         So it can be heard clearly, say the plural of cat.         So it can be heard clearly, say the plural of cat.           Say the plural of dog so it can be heard clearly.         So it can be heard clearly, say the plural of dog.         So it can be heard clearly, say the plural of dog.           Say the plural of duck so it can be heard clearly.         So it can be heard clearly, say the plural of duck.         So it can be heard clearly, say the plural of duck.           CR4a - Regular Pluralization         Say the plural of foot so it can be he		Repeat the sound <b>kaka</b> so it can be heard clearly.	So it can be heard clearly, repeat the sound <b>kaka</b> .
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CR4a - Regular PluralizationSay the plural of cat so it can be heard clearly.So it can be heard clearly, say the plural of dog.Say the plural of dog so it can be heard clearly.So it can be heard clearly, say the plural of dog.So it can be heard clearly, say the plural of dog.Say the plural of cow so it can be heard clearly.So it can be heard clearly, say the plural of cow.So it can be heard clearly, say the plural of dog.Say the plural of duck so it can be heard clearly.So it can be heard clearly, say the plural of duck.So it can be heard clearly, say the plural of duck.CR4b - Irregular PluralizationSay the plural of foot so it can be heard clearly.So it can be heard clearly, say the plural of foot.Say the plural of foot so it can be heard clearly.So it can be heard clearly, say the plural of foot.Say the plural of foot so it can be heard clearly.So it can be heard clearly, say the plural of foot.Say the plural of man so it can be heard clearly.So it can be heard clearly, say the plural of foot.Say the plural of man so it can be heard clearly.So it can be heard clearly, say the plural of man.Say the plural of mouse so it can be heard clearly.So it can be heard clearly, say the plural of man.		Say the word <b>duck</b> so it can be heard clearly.	So it can be heard clearly, say the word <b>duck</b> .
CR4a - Regular Pluralization       Say the plural of dog so it can be heard clearly.       So it can be heard clearly, say the plural of dog.         Say the plural of cow so it can be heard clearly.       So it can be heard clearly, say the plural of cow.         Say the plural of duck so it can be heard clearly.       So it can be heard clearly, say the plural of duck.         So it can be heard clearly, say the plural clearly.       So it can be heard clearly, say the plural of duck.         CR4b - Irregular Pluralization       Say the plural of foot so it can be heard clearly.       So it can be heard clearly, say the plural of foot.         Say the plural of foot so it can be heard clearly.       So it can be heard clearly, say the plural of foot.         Say the plural of foot so it can be heard clearly.       So it can be heard clearly, say the plural of foot.         Say the plural of foot so it can be heard clearly.       So it can be heard clearly, say the plural of foot.         Say the plural of man so it can be heard clearly.       So it can be heard clearly, say the plural of man.         Say the plural of mouse so it can be heard clearly.       So it can be heard clearly, say the plural of man.		Say the plural of <b>cat</b> so it can be heard clearly.	So it can be heard clearly, say the plural of <b>cat</b> .
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CR4b - Irregular       Say the plural of <b>goose</b> so it can be heard       So it can be heard clearly, say the plural of <b>goose</b> .         Say the plural of <b>foot</b> so it can be heard       So it can be heard clearly, say the plural of <b>foot</b> .         Say the plural of <b>man</b> so it can be heard       So it can be heard clearly, say the plural of <b>foot</b> .         Say the plural of <b>man</b> so it can be heard       So it can be heard clearly, say the plural of <b>foot</b> .         Say the plural of <b>man</b> so it can be heard       So it can be heard clearly, say the plural of <b>man</b> .         Say the plural of <b>mouse</b> so it can be heard       So it can be heard clearly, say the plural of <b>mouse</b> .		Say the plural of <b>duck</b> so it can be heard clearly.	So it can be heard clearly, say the plural of <b>duck</b> .
CR4b - Irregular Pluralization       Say the plural of foot so it can be heard clearly.       So it can be heard clearly, say the plural of foot.         Say the plural of man so it can be heard clearly.       So it can be heard clearly, say the plural of man.       So it can be heard clearly, say the plural of man.         Say the plural of mouse so it can be heard clearly.       So it can be heard clearly, say the plural of mouse.       So it can be heard clearly, say the plural of mouse.		Say the plural of <b>goose</b> so it can be heard clearly.	So it can be heard clearly, say the plural of <b>goose</b> .
Pluralization       Say the plural of man so it can be heard clearly.       So it can be heard clearly, say the plural of man.         Say the plural of mouse so it can be heard clearly.       So it can be heard clearly, say the plural of mouse.	CR4b - Irregular	Say the plural of <b>foot</b> so it can be heard clearly.	So it can be heard clearly, say the plural of <b>foot</b> .
Say the plural of <b>mouse</b> so it can be heard clearly, say the plural of <b>mouse</b> .	Pluralization	Say the plural of <b>man</b> so it can be heard clearly.	So it can be heard clearly, say the plural of <b>man</b> .
		Say the plural of <b>mouse</b> so it can be heard clearly.	So it can be heard clearly, say the plural of <b>mouse</b> .

\* For intraoperative patients, usually only the first 7 stimuli were used and the words "or in zoos" were omitted.

\*\* For intraoperative patients, usually only the first 7 stimuli were used.

\*\*\* Stimuli only used for 436, who was the first participant of the study. Data from these questions were not included in the response amplitude analysis reported in Extended Data Figure 1.

\*\*\*\* For some participants, the word "say" was used instead of "repeat".

#### **Extended Data Table 2**

Participant information

<u>Code</u>	Age	<u>Sex</u>	<u>Type</u>	Handedness	<u>Wada</u> <u>Testing</u>	<u>Pathology/</u> Diagnosis	Tumor location/ seizure loci	Tasks Completed	Data Rejected	Notes
436	58	F	Acute	RH	N/A	Grade II oligodendroglioma (tumor)	Left middle frontal gyrus	CI questions; conversation	N/A	N/A
442	35	F	Chronic	70+	LH language dominant	epilepsy	Multifocal onset	CI questions; CR1, CR2, CR3a, CR3c, CR4a, CR4b; conversation	N/A	CR and trials no interleav
463	65	М	Acute	ambidextrous	N/A	Glioblastoma (tumor)	Left parahippocampal gyrus	CI questions; CR1, CR2, CR3a, CR4b; conversation	Excluded from CR analysis as only 2 pluralization trials (CR4) were completed correctly; planning and perception electrode clusters rejected from PC functional analysis ( 50% mixed electrodes)	N/A
472	32	М	Acute	RH	N/A	Vascular malformation, focal cortical dysplasia TIIIc (tumor)	Left temporal pole	CI questions; CR1, CR2, CR3a, CR4b; conversation	Excluded from PCA clustering analysis (n = 0 unmixed perception and n = 0 unmixed production electrodes)	N/A
477	24	F	Chronic	80+	LH language dominant	epilepsy/mild gliosis	Left parietal seizure focus	CI questions; CR1, CR2, CR3a, CR3b, CR4a, CR4b; conversation	Production electrode cluster rejected from PC functional analysis ( 50% mixed electrodes); mixed electrodes); mixed planning- production electrode cluster rejected from PC functional analysis	N/A

<u>Code</u>	Age	<u>Sex</u>	<u>Туре</u>	<u>Handedness</u>	<u>Wada</u> <u>Testing</u>	<u>Pathology/</u> <u>Diagnosis</u>	<u>Tumor location/</u> <u>seizure loci</u>	Tasks Completed	Data Rejected	Notes
486	53	М	Acute	RH	LH language dominant	Glioblastoma (tumor)	Left inferior frontal gyrus	CI questions; CR1, CR2, CR3a, CR4b; conversation	N/A	N/A
494	30	М	Acute	70+	LH language dominant	Hippocampal sclerosis (epilepsy)	Left hippocampus	CI questions; CR1, CR2, CR3a, CR4b; conversation	N/A	N/A
510	54	М	Acute	RH	N/A	Anaplastic Oligodendroglioma (tumor)	Left inferior frontal gyrus	CI questions; CR1, CR2, CR3a, CR4b; conversation	Excluded from PCA clustering analysis (n = 0 unmixed perception and n = 1 unmixed production electrodes).	N/A

### Extended Data Table 3a

All electrode locations and GLM classifications

	Electrode		Rejected		MNI Coordinates					
Patient	<u>Number</u>	<u>GLM</u> Designation	Dy Jittering Analysis?	<u>Trial</u>	Perception	<u>Planning</u>	Production	X	¥	Z
	1	Excluded - Artifactual	N/A					-57.53	-37.93	35.:
	2		YES	N.S.	N.S.	0.25757	N.S.	-58.11	-33.17	34.′
	3	Excluded - Artifactual	N/A					-58.71	-28.40	33.9
4361	4	Excluded - Artifactual	N/A					-59.30	-23.64	33.0
Intraoperative	5	Production	NO	N.S.	N.S.	N.S.	0.5188129	-59.83	-18.89	32.2
	6		YES	N.S.	N.S.	N.S.	N.S.	-60.23	-14.14	31.
	7		YES	N.S.	N.S.	N.S.	0.3881277	-60.43	-9.41	30.
	8		YES	N.S.	N.S.	N.S.	N.S.	-60.36	-4.70	28.

	Electrode		Rejected	GLM Weights (z)					MNI Coordinates		
Patient	<u>Number</u>	<u>GLM</u> Designation	by Jittering Analysis?	Trial	Perception	<u>Planning</u>	Production	X	Y	Z	
	9		YES	N.S.	N.S.	N.S.	N.S.	-60.07	-0.02	27.2	
	10		YES	-0.246621886	N.S.	0.20485	0.3595874	-59.66	4.66	25.6	
	11	Planning	NO	0.299508389	N.S.	0.44039	N.S.	-59.18	9.33	23.9	
	12	Excluded - Artifactual	N/A					-58.69	14.00	22.3	
	13	Planning	NO	N.S.	N.S.	0.7416	N.S.	-58.25	18.68	20.7	
	14		YES	N.S.	N.S.	N.S.	0.2736352	-53.93	-37.04	40.5	
	15		YES	N.S.	N.S.	N.S.	0.2957145	-54.53	-32.28	39.6	
	16	Production	NO	N.S.	-0.202617537	N.S.	0.5448633	-55.16	-27.51	38.8	
	17	Production	NO	N.S.	N.S.	N.S.	0.4011025	-55.79	-22.74	38.0	
	18	Production	NO	N.S.	N.S.	N.S.	0.543027	-56.36	-17.98	37.1	
	19		NO	N.S.	N.S.	N.S.	N.S.	-56.83	-13.22	36.2	
	20		YES	N.S.	N.S.	N.S.	N.S.	-57.09	-8.48	35.1	
	21		YES	N.S.	N.S.	N.S.	N.S.	-57.01	-3.77	33.7	
	22		YES	N.S.	N.S.	N.S.	N.S.	-56.73	0.92	32.1	
	23		YES	N.S.	N.S.	0.22356	N.S.	-56.33	5.60	30.5	
	24	Planning	NO	N.S.	-0.204891664	0.35086	N.S.	-55.86	10.27	28.8	
	25	Planning	NO	N.S.	N.S.	0.48775	N.S.	-55.39	14.95	27.1	
	26	Planning	NO	N.S.	N.S.	0.83801	N.S.	-54.97	19.62	25.5	
	27		YES	N.S.	N.S.	N.S.	0.2793434	-50.29	-36.16	45.4	
PatientNumberCLAM DesignationNumbersityTrialPerceptionPlanningProductionXXZ28287YESN.S.N.S.N.S.N.S.N.S.S.O50.90-31.3944.029ProductionNON.S0.206478079N.S.0.529417-51.54-26.6243.330ProductionNON.S0.204478079N.S.0.417418-52.18-21.8442.931ProductionNON.S0.24181095N.S.0.881716-52.77-17.0642.932ProductionNON.S0.24181095N.S.0.962780-53.23-12.3241.033ProductionNON.S.N.S.N.S.0.962780-53.23-12.3241.033ProductionNON.S.N.S.N.S.0.962780-53.23-12.3241.034ProductionNON.S.N.S.N.S.0.962780-53.23-12.3241.035ProductionNON.S.N.S.N.S.0.5938367-53.43-2.8635.236141YESN.S.N.S.N.S.0.260964-52.9316.335.237142YESN.S.N.S.N.S.0.260964-52.9316.442.939PlanningNON.S.N.S.N.S.0.260964-52.9316.242.939Planning<		Ele	ectrode	Rejected		GLM Weigh	nts (z)		MN	I Coordin	ates
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28          YES         N.S.         N.S.         N.S.         N.S.         -50.00         -31.30         44.0           29         Production         NO         N.S.         -0.20478079         N.S.         0.529417         -51.54         -26.62         43.3           30         Production         NO         N.S.         -0.20478079         N.S.         0.4174418         -52.18         -21.84         42.4           31         Production         NO         N.S.         -0.241816995         N.S.         0.8817166         -52.77         -17.08         42.4           32         Production         NO         N.S.         O.241816995         N.S.         0.8817166         -52.77         -17.08         42.4           33         Production         NO         N.S.         N.S.         N.S.         0.9027801         -53.30         -12.32         41.4           34         Production         NO         N.S.         N.S.         N.S.         0.499066         -53.50         -75.8         39.3           36         I.         YES         N.S.         N.S.         N.S.         0.5938367         -53.49         -2.86         36.2           37	Patient	<u>Number</u>	<u>GLM</u> Designation	by Jittering Analysis?	<u>Trial</u>	<b>Perception</b>	<u>Planning</u>	Production	X	Y	Z
29         Production         NO         N.S.         -0.206478079         N.S.         0.5294417         -51.54         -26.62         43.3           30         Production         NO         N.S.         N.S.         N.S.         0.4174418         -52.18         -21.84         42.5           31         Production         NO         N.S.         -0.241816995         N.S.         0.8817166         -52.77         -17.08         42.6           32         Production         NO         N.S.         N.S.         N.S.         0.9627801         -53.23         -12.32         41.6           33         Production         NO         N.S.         N.S.         N.S.         0.4099064         -53.50         -7.58         39.9           34         Production         NO         N.S.         N.S.         N.S.         0.4099064         -53.50         -7.58         39.9           35         YES         N.S.         N.S.         N.S.         N.S.         0.5393837         -53.49         -28.69         35.3           36         YES         N.S.         N.S.         N.S.         N.S.         -52.21         11.22         33.0           37         YES         N.S. </td <th></th> <td>28</td> <td></td> <td>YES</td> <td>N.S.</td> <td>N.S.</td> <td>N.S.</td> <td>N.S.</td> <td>-50.90</td> <td>-31.39</td> <td>44.0</td>		28		YES	N.S.	N.S.	N.S.	N.S.	-50.90	-31.39	44.0
30         Production         NO         N.S.         N.S.         N.S.         0.4174418         -52.18         -21.84         42.5           31         Production         NO         N.S.         -0.241816995         N.S.         0.8817166         -52.77         -17.08         42.0           32         Production         NO         N.S.         N.S.         N.S.         0.9627801         -53.23         -12.32         41.0           33         Production         NO         N.S.         N.S.         N.S.         0.4099064         -53.50         -7.58         39.8           34         Production         NO         N.S.         N.S.         N.S.         0.5938367         -53.49         -2.86         38.3           35         YES         N.S.         N.S.         N.S.         N.S.         -53.27         1.84         36.5           36         YES         N.S.         N.S.         N.S.         0.2699964         -52.93         6.53         35.3           37         YES         N.S.         N.S.         0.2887         N.S.         -51.71         15.90         32.4           38         Planning         NO         N.S.         N.S.         0		29	Production	NO	N.S.	-0.206478079	N.S.	0.5294417	-51.54	-26.62	43.7
31         Production         NO         N.S.         -0.241816995         N.S.         0.8817166         -52.77         -17.08         42.0           32         Production         NO         N.S.         N.S.         N.S.         0.9627801         -53.23         -12.32         41.0           33         Production         NO         N.S.         N.S.         N.S.         0.4099064         -53.50         -7.58         39.4           34         Production         NO         N.S.         N.S.         N.S.         0.5938367         -53.49         -2.86         38.3           35         YES         N.S.         N.S.         N.S.         N.S.         -53.27         1.84         36.5           36         YES         N.S.         N.S.         N.S.         0.2609964         -52.93         6.53         35.3           37         YES         N.S.         N.S.         0.22927         N.S.         -52.71         11.22         3.6           38         Planning         NO         N.S.         N.S.         0.38464         N.S.         -51.72         20.58         30.4           40         Excluded- Artificenual         N/A         Inconoon         N.S.		30	Production	NO	N.S.	N.S.	N.S.	0.4174418	-52.18	-21.84	42.9
32         Production         NO         N.S.         N.S.         N.S.         0.9627801         -53.23         -12.32         41.4           33         Production         NO         N.S.         N.S.         N.S.         0.4099064         -53.50         -7.58         39.1           34         Production         NO         N.S.         N.S.         N.S.         0.4099064         -53.50         -7.58         39.1           34         Production         NO         N.S.         N.S.         N.S.         0.5938367         -53.49         -2.86         38.2           35         YES         N.S.         N.S.         N.S.         N.S.         -53.27         1.84         36.5           36         YES         N.S.         N.S.         N.S.         0.2609964         -52.93         6.53         35.3           37         YES         N.S.         N.S.         0.28887         N.S.         -52.71         11.22         33.6           38         Planning         NO         N.S.         N.S.         0.38464         N.S.         -51.72         20.58         30.7           40         Excluded- Artifactual         N/A		31	Production	NO	N.S.	-0.241816995	N.S.	0.8817166	-52.77	-17.08	42.0
33         Production         NO         N.S.         N.S.         N.S.         0.4099064         -53.50         -7.58         39.4           34         Production         NO         N.S.         N.S.         N.S.         0.5938367         -53.49         -2.86         38.4           35         YES         N.S.         N.S.         N.S.         N.S.         N.S.         -53.27         1.84         36.5           36         YES         N.S.         N.S.         N.S.         N.S.         0.2609964         -52.93         6.53         35.3           37         YES         N.S.         N.S.         0.22927         N.S.         -52.52         11.22         33.6           38         Planning         NO         N.S.         -0.277229163         0.28887         N.S.         -51.72         20.58         30.7           40         Excluded - Artificeual         N/A         -         -46.63         -35.28         50.4           41         NO         N.S.         N.S.         N.S.         N.S.         -47.23         -30.51         49.3           42         Excluded - Artificeual         N/A         -         -47.87         -25.73         48.6 <th></th> <td>32</td> <td>Production</td> <td>NO</td> <td>N.S.</td> <td>N.S.</td> <td>N.S.</td> <td>0.9627801</td> <td>-53.23</td> <td>-12.32</td> <td>41.0</td>		32	Production	NO	N.S.	N.S.	N.S.	0.9627801	-53.23	-12.32	41.0
34         Production         NO         N.S.         N.S.         N.S.         N.S.         0.5938367         -53.49         -2.86         38.3           35         YES         N.S.         N.S.         N.S.         N.S.         N.S.         -53.27         1.84         36.3           36         YES         N.S.         N.S.         N.S.         N.S.         0.2609964         -52.93         6.53         35.3           37         YES         N.S.         N.S.         N.S.         0.22927         N.S.         -52.52         11.22         33.0           38         Planning         NO         N.S.         -0.277229163         0.28887         N.S.         -52.11         15.90         32.0           39         Planning         NO         N.S.         N.S.         0.38464         N.S.         -51.72         20.58         30.0           40         Excluded- Artifactual         N/A          -46.63         -35.28         50.4           41         NO         N.S.         N.S.         N.S.         N.S.         -47.23         -30.51         49.5           42         Excluded- Artifactual         N/A          -47.87         -25.7		33	Production	NO	N.S.	N.S.	N.S.	0.4099064	-53.50	-7.58	39.8
35       YES       N.S.       N.S.       N.S.       N.S.       -53.27       1.84       36.5         36       YES       N.S.       N.S.       N.S.       N.S.       0.2609964       -52.93       6.53       35.3         37       YES       N.S.       N.S.       N.S.       0.22927       N.S.       -52.52       11.22       33.4         38       Planning       NO       N.S.       -0.277229163       0.28887       N.S.       -51.72       20.58       30.4         39       Planning       NO       N.S.       N.S.       0.38464       N.S.       -51.72       20.58       30.4         40       Excluded- Artifactual       N/A       Image: No       N.S.       N.S.       N.S.       -46.63       -35.28       50.4         41       NO       N.S.       N.S.       N.S.       N.S.       N.S.       -47.83       -30.51       49.5         42       Excluded - Artifactual       N/A       Image: No       N.S.       N.S.       N.S.       0.6803557       -48.51       -20.96       47.7         43       Production       NO       N.S.       N.S.       N.S.       0.8914032       -49.08       -16.19		34	Production	NO	N.S.	N.S.	N.S.	0.5938367	-53.49	-2.86	38.5
36         YES         N.S.         N.S.         N.S.         O.2609964         -52.93         6.53         35.3           37         YES         N.S.         N.S.         0.22927         N.S.         -52.52         11.22         33.0           38         Planning         NO         N.S.         -0.277229163         0.28887         N.S.         -52.51         15.90         32.0           39         Planning         NO         N.S.         -0.277229163         0.28887         N.S.         -51.72         20.58         30.4           40         Excluded- Artifactual         N/A         Image: Constraint of the second secon		35		YES	N.S.	N.S.	N.S.	N.S.	-53.27	1.84	36.9
37       YES       N.S.       N.S.       0.22927       N.S.       -52.52       11.22       33.4         38       Planning       NO       N.S.       -0.277229163       0.28887       N.S.       -52.11       15.90       32.4         39       Planning       NO       N.S.       N.S.       0.38464       N.S.       -51.72       20.58       30.4         40       Excluded- Artifactual       N/A       Image: Constraint of the second sec		36		YES	N.S.	N.S.	N.S.	0.2609964	-52.93	6.53	35.3
38       Planning       NO       N.S.       -0.277229163       0.28887       N.S.       -52.11       15.90       32.0         39       Planning       NO       N.S.       N.S.       0.38464       N.S.       -51.72       20.58       30.4         40       Excluded- Artifactual       N/A       Image: Constraint of the second se		37		YES	N.S.	N.S.	0.22927	N.S.	-52.52	11.22	33.0
39       Planning       NO       N.S.       N.S.       0.38464       N.S.       -51.72       20.58       30.4         40       Excluded - Artifactual       N/A       Image: Constraint of the second sec		38	Planning	NO	N.S.	-0.277229163	0.28887	N.S.	-52.11	15.90	32.0
40       Excluded - Artifactual       N/A       Image: Solution of the second		39	Planning	NO	N.S.	N.S.	0.38464	N.S.	-51.72	20.58	30.4
41       NO       N.S.       N.S.       N.S.       N.S.       -47.23       -30.51       49.5         42       Excluded - Artifactual       N/A       Image: Constraint of the second s		40	Excluded - Artifactual	N/A					-46.63	-35.28	50.4
42       Excluded - Artifactual       N/A       Image: Second S		41		NO	N.S.	N.S.	N.S.	N.S.	-47.23	-30.51	49.5
43         Production         NO         N.S.         N.S.         N.S.         0.6803557         -48.51         -20.96         47.7           44         Planning- Production         NO         N.S.         -0.37831231         0.27894         0.8914032         -49.08         -16.19         46.7           45         Production         NO         N.S.         N.S.         N.S.         0.9633869         -49.52         -11.43         45.7		42	Excluded - Artifactual	N/A					-47.87	-25.73	48.0
44         Planning- Production         NO         N.S.         -0.37831231         0.27894         0.8914032         -49.08         -16.19         46.7           45         Production         NO         N.S.         N.S.         N.S.         0.9633869         -49.52         -11.43         45.7		43	Production	NO	N.S.	N.S.	N.S.	0.6803557	-48.51	-20.96	47.3
45 Production NO N.S. N.S. N.S. 0.9633869 -49.52 -11.43 45.7		44	Planning- Production	NO	N.S.	-0.37831231	0.27894	0.8914032	-49.08	-16.19	46.7
		45	Production	NO	N.S.	N.S.	N.S.	0.9633869	-49.52	-11.43	45.3

	Electrode			GLM Weights (z)					MNI Coordinates		
Patient	Number	<u>GLM</u> <u>Designation</u>	by Jittering Analysis?	Trial	Perception	<u>Planning</u>	Production	X	Y	Z	
	46	Production	NO	0.233344565	N.S.	N.S.	0.7261057	-49.79	-6.68	44.:	
	47	Production	NO	N.S.	-0.232255414	N.S.	0.4176748	-49.84	-1.95	43.:	
	48		YES	N.S.	N.S.	N.S.	N.S.	-49.71	2.76	41.0	
	49		YES	N.S.	N.S.	N.S.	0.2463998	-49.46	7.46	40.	
	50		YES	N.S.	N.S.	0.20577	0.2490346	-49.14	12.15	38.4	
	51	Planning	NO	N.S.	N.S.	0.37714	N.S.	-48.78	16.84	36.5	
	52		YES	N.S.	N.S.	0.30163	0.2393926	-48.44	21.53	35.2	
	53		YES	N.S.	N.S.	N.S.	N.S.	-42.97	-34.39	55.2	
	54		YES	N.S.	N.S.	N.S.	N.S.	-43.56	-29.62	54.:	
	55		YES	-0.272407936	N.S.	N.S.	N.S.	-44.19	-24.84	53.4	
	56		NO	N.S.	N.S.	-0.16439	N.S.	-44.80	-20.07	52.:	
	57	Production	NO	-0.265519531	-0.207795467	N.S.	1.0455253	-45.34	-15.30	51.:	
	58	Production	NO	N.S.	N.S.	N.S.	0.8295242	-45.76	-10.54	50.4	
	59	Production	NO	N.S.	N.S.	N.S.	0.7187525	-46.03	-5.79	49.1	
	60	Production	NO	N.S.	N.S.	N.S.	0.4652766	-46.14	-1.05	47.5	
	61		YES	N.S.	-0.187775583	N.S.	0.2325704	-46.09	3.67	46.	

Potient	Ele	ectrode	Rejected		GLM Weigh	nts (z)		MN	I Coordin	ates
Patient	Number	<u>GLM</u> Designation	by Jittering Analysis?	Trial	Perception	<u>Planning</u>	Production	X	Y	Z
	62	Planning	NO	N.S.	N.S.	0.28976	N.S.	-45.92	8.38	44.
	63		YES	-0.210930735	N.S.	N.S.	N.S.	-45.68	13.08	43.
	64		YES	N.S.	N.S.	N.S.	N.S.	-45.40	17.78	41.
	1	Excluded - Artifactual	N/A					-59.36	-7.75	-21.
	2	Planning- Production	NO	N.S.	N.S.	0.30705	0.2866195	-58.31	-4.77	-17.
	3	Production	NO	N.S.	-0.16580427	N.S.	0.2391814	-57.37	-1.83	-12.
	4	Excluded - Artifactual	N/A					-56.45	1.12	-8.
	5		YES	N.S.	N.S.	N.S.	0.2794977	-55.52	4.06	-4.
	6		YES	N.S.	N.S.	N.S.	0.2729939	-54.60	7.00	-0.
	7	Production	NO	N.S.	N.S.	N.S.	0.4822546	-53.69	9.93	3.
463L Intraoperative	8	Production	NO	-0.338820362	N.S.	N.S.	0.7582374	-52.83	12.86	8.
	9	Production	NO	-0.483763704	N.S.	N.S.	0.6479538	-52.02	15.76	12
	10	Perception- Production	NO	N.S.	0.548440014	N.S.	0.2279239	-51.28	18.64	16.
	11		YES	N.S.	N.S.	N.S.	N.S.	-50.62	21.50	20.
	12	Excluded - Artifactual	N/A					-50.01	24.34	25.
	13		YES	N.S.	N.S.	N.S.	N.S.	-49.32	27.21	29.
	14		YES	N.S.	N.S.	N.S.	0.2744417	-61.17	-12.28	-18.
	15		YES	N.S.	N.S.	N.S.	0.2382969	-59.98	-9.26	-14.

	ectrode	Rejected	GLM Weights (z)					MNI Coordinates		
Patient	Number	<u>GLM</u> Designation	by Jittering Analysis?	Trial	Perception	<u>Planning</u>	Production	X	Y	Z
	16	Production	NO	N.S.	N.S.	N.S.	0.3501959	-58.85	-6.25	-9.
	17		YES	N.S.	N.S.	0.16509	N.S.	-57.73	-3.25	-5.
	18		YES	N.S.	N.S.	N.S.	N.S.	-56.61	-0.24	-1.
	19		NO	N.S.	-0.180036929	N.S.	N.S.	-55.49	2.76	2.
	20	Planning	NO	0.404202996	-0.3524339	0.15759	N.S.	-54.39	5.76	7.
	21	Planning- Production	NO	N.S.	-0.18931978	0.20421	0.5966911	-53.33	8.74	11.
	22	Planning- Production	NO	N.S.	-0.12673648	0.15018	0.3963801	-52.33	11.71	15.
	23	Perception- Planning	NO	N.S.	0.605285025	0.28684	N.S.	-51.40	14.65	19.
	24	Perception- Planning- Production	NO	N.S.	0.53447814	0.62091	0.440486	-50.57	17.56	23.
	25	Planning	NO	N.S.	N.S.	0.47296	N.S.	-49.82	20.45	28.
	26	Planning	NO	N.S.	-0.365201192	0.22575	N.S.	-49.05	23.34	32
	27	Excluded - Artifactual	N/A					-63.18	-16.88	-15.
	28		YES	N.S.	N.S.	0.14738	N.S.	-61.84	-13.80	-10.
	29	Perception- Planning- Production	NO	N.S.	0.484041428	0.47432	0.345482	-60.48	-10.72	-6.
	30	Perception- Planning- Production	NO	-0.316957988	0.35337835	0.5149	0.5180013	-59.12	-7.64	-2.
	31	Planning- Production	NO	N.S.	N.S.	0.18953	0.3234403	-57.76	-4.57	1.
	32	Production	NO	N.S.	N.S.	N.S.	0.6807492	-56.42	-1.49	5.

	Ele	ectrode	Rejected	GLM Weights (z)					MNI Coordinates		
Patient	<u>Number</u>	<u>GLM</u> Designation	Jittering Analysis?	Trial	Perception	<u>Planning</u>	Production	X	¥	Z	
	33	Production	NO	N.S.	-0.239291929	N.S.	0.897202	-55.10	1.58	10.1	
	34	Production	NO	-0.381041128	N.S.	N.S.	1.1042034	-53.82	4.63	14.3	
	35	Production	NO	-0.33679518	N.S.	N.S.	0.7685204	-52.59	7.67	18.5	
	36	Planning	NO	0.324823236	N.S.	0.25807	N.S.	-51.44	10.68	22.7	
	37	Planning- Production	NO	N.S.	N.S.	0.48176	0.3520618	-50.38	13.67	26.9	
	38	Planning	NO	N.S.	N.S.	0.45348	N.S.	-49.45	16.61	31.2	
	39	Planning- Production	NO	N.S.	-0.247049633	0.22404	0.3369419	-48.60	19.53	35.4	
	40	Production	NO	0.200086812	-0.221967324	N.S.	0.2624638	-65.19	-21.47	-12.0	
	41		YES	N.S.	N.S.	N.S.	N.S.	-63.70	-18.35	-7.8	
	42	Excluded - Artifactual	N/A					-62.12	-15.20	-3.0	
	43		YES	N.S.	N.S.	N.S.	N.S.	-60.52	-12.04	0.5	
	44		YES	N.S.	0.493532474	0.18991	0.4389757	-58.93	-8.89	4.7	
	45	Production	NO	-0.518154137	N.S.	N.S.	0.4919819	-57.37	-5.75	8.9	
	46	Production	NO	N.S.	N.S.	N.S.	0.4533271	-55.84	-2.61	13.1	
	47	Production	NO	N.S.	N.S.	N.S.	0.6668852	-54.34	0.51	17.3	
	48		YES	N.S.	N.S.	N.S.	0.2230687	-52.88	3.62	21.5	
	49		YES	N.S.	-0.223508435	N.S.	N.S.	-51.49	6.71	25.7	
	50	Planning- Production	NO	0.261051214	N.S.	0.19476	0.2545687	-50.19	9.77	30.0	

	Electrode		Rejected	GLM Weights (z)					MNI Coordinates		
Patient	Number	<u>GLM</u> Designation	by Jittering Analysis?	Trial	Perception	<u>Planning</u>	<b>Production</b>	X	Y	Z	
	51	Production	NO	N.S.	-0.3082805	N.S.	0.2524143	-49.06	12.78	34.2	
	52	Planning- Production	NO	0.195069386	-0.310855838	0.16372	0.2745816	-48.18	15.71	38.4	
	53		YES	N.S.	N.S.	N.S.	0.2768078	-67.00	-26.00	-8.8	
	54		YES	N.S.	0.150897445	N.S.	N.S.	-65.37	-22.84	-4.7	
	55		YES	-0.22297392	N.S.	N.S.	N.S.	-63.61	-19.63	-0.:	
	56	Perception	NO	N.S.	0.589007833	N.S.	N.S.	-61.83	-16.41	3.(	
	57		YES	N.S.	N.S.	N.S.	N.S.	-60.06	-13.20	7.5	
	58	Production	NO	-0.365247602	N.S.	N.S.	0.6033718	-58.31	-10.00	12.0	
	59	Production	NO	-0.410804439	N.S.	N.S.	1.0408236	-56.60	-6.81	16.	
	60	Production	NO	N.S.	N.S.	N.S.	0.5821413	-54.92	-3.63	20.4	
	61	Production	NO	-0.304387994	N.S.	N.S.	0.7779056	-53.28	-0.46	24.0	
	62	Planning- Production	NO	0.239540178	-0.327519099	0.2512	0.5230921	-51.69	2.69	28.	
	63	Planning- Production	NO	N.S.	N.S.	0.29172	0.9096786	-50.20	5.81	33.	
	64		YES	N.S.	-0.114595064	N.S.	0.2344248	-48.89	8.88	37.2	
	1	Excluded - Artifactual	N/A					-66.30	-13.74	11.3	
472L Intraoperative	2		NO	0.440167675	-0.252981596	N.S.	N.S.	-65.08	-8.80	9.	
	3	Planning	NO	0.632257972	-0.338431832	0.2988	N.S.	-63.86	-3.83	7.	
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Electrode			Rejected	GLM Weights (z)					MNI Coordinates		
Patient	Number	<u>GLM</u> Designation	by Jittering Analysis?	Trial	Perception	<u>Planning</u>	Production	X	Y	Z	
	4	Excluded - Artifactual	NO					-62.65	1.14	6.0	
	5	Planning	NO	0.478953792	-0.351789372	0.30301	N.S.	-61.44	6.12	4.2	
	6		YES	0.253921029	-0.252644121	N.S.	N.S.	-60.22	11.09	2.4	
	7		YES	N.S.	N.S.	0.15893	N.S.	-59.01	16.08	0.3	
	8		YES	N.S.	N.S.	N.S.	N.S.	-57.80	21.07	-1.1	
	9		YES	N.S.	N.S.	N.S.	N.S.	-56.60	26.09	-2.9	
	10		YES	N.S.	N.S.	N.S.	N.S.	-55.40	31.12	-4.7	
	11		YES	N.S.	N.S.	N.S.	N.S.	-54.21	36.18	-6.0	
	12	Excluded - Artifactual	N/A					-53.03	41.26	-8.5	
	13		YES	N.S.	N.S.	N.S.	N.S.	-51.83	46.32	-10.3	
	14		NO	0.637884357	-0.676395326	N.S.	N.S.	-65.99	-11.81	16.3	
	15		YES	0.224192086	-0.218652105	N.S.	N.S.	-64.76	-6.92	14.0	
	16		NO	0.378802324	-0.202053885	N.S.	N.S.	-63.52	-2.01	12.9	
	17	Planning	NO	1.011357061	-0.537163882	0.73877	-0.792162	-62.29	2.90	11.2	
	18	Planning	NO	0.435273735	-0.259550212	0.43448	N.S.	-61.06	7.81	9.4	
	19		YES	N.S.	N.S.	N.S.	N.S.	-59.83	12.73	7.3	
	20		YES	N.S.	N.S.	N.S.	N.S.	-58.60	17.64	6.0	
	21		YES	N.S.	N.S.	0.47115	N.S.	-57.38	22.58	4.2	
	22		YES	N.S.	N.S.	0.27843	N.S.	-56.16	27.53	2.5	
	23	Planning	NO	N.S.	N.S.	0.83896	N.S.	-54.94	32.50	0.7	
	24		YES	0.38447788	-0.224609399	N.S.	N.S.	-53.74	37.51	-1.0	

	Ele	ectrode	Rejected	GLM Weights (z)					MNI Coordinates		
Patient	<u>Number</u>	<u>GLM</u> <u>Designation</u>	by Jittering Analysis?	<u>Trial</u>	Perception	<u>Planning</u>	Production	X	Y	Z	
	25		YES	N.S.	N.S.	N.S.	N.S.	-52.54	42.54	-2.9	
	26		YES	0.473273848	-0.538356945	0.16755	N.S.	-51.34	47.57	-4.7	
	27	Excluded - Artifactual	N/A					-65.70	-9.81	21.2	
	28		NO	0.382639476	-0.277674307	N.S.	N.S.	-64.45	-4.97	19.0	
	29		YES	0.286842858	-0.213347758	0.13488	N.S.	-63.20	-0.13	17.9	
	30	Planning	NO	1.068651264	-0.618537545	0.62218	-0.747139	-61.95	4.70	16.3	
	31		YES	N.S.	N.S.	N.S.	N.S.	-60.69	9.53	14.0	
	32	Planning	NO	N.S.	N.S.	0.46617	N.S.	-59.44	14.37	13.0	
	33		YES	N.S.	N.S.	N.S.	N.S.	-58.20	19.22	11.3	
	34		YES	N.S.	N.S.	0.23722	N.S.	-56.95	24.08	9.0	
	35		YES	0.311999454	-0.246335168	0.26925	-0.272267	-55.71	28.95	8.0	
	36		YES	0.470987114	-0.301629547	N.S.	-0.313468	-54.48	33.85	6.2	
	37		YES	N.S.	-0.344413712	0.3831	N.S.	-53.25	38.78	4.5	
	38		YES	0.449066103	-0.459650249	0.19744	N.S.	-52.04	43.76	2.7	
	39		YES	N.S.	-0.272628412	N.S.	N.S.	-50.83	48.76	0.9	
	40	Planning	NO	0.539036948	-0.50031757	0.17977	N.S.	-65.41	-7.81	26.1	
	41		YES	N.S.	-0.262695065	0.27019	N.S.	-64.15	-3.02	24.5	
	42	Excluded - Artifactual	N/A					-62.87	1.74	22.9	

Rejected GLAI Weights (2) MINI	MNI Coordinates		
Patient     Number     GLM Designation     by Jittering Analysis?     Trial     Perception     Planning     Production     X	¥	Z	
43 YES 0.494673404 -0.418444201 N.S0.367371 -61.60	6.50	21.4	
44 YES N.S. N.S. N.S. N.S60.33	11.26	19.8	
45 YES N.S. N.S. N.S. N.S59.06	16.02	18.2	
46 YES N.S. N.S. N.S. N.S57.79	20.80	16.6	
47 YES N.S. N.S. N.S. N.S56.53	25.59	15.0	
48 YES 0.343138386 N.S. 0.35942 N.S55.27	30.39	13.4	
49 YES N.S. N.S. 0.28328 N.S54.01	35.21	11.8	
50 Planning NO N.S0.395152957 0.39462 N.S52.76	40.06	10.1	
51 Planning NO N.S0.327776527 0.26532 N.S51.53	44.97	8.4	
52 YES N.S. N.S. 0.29785 N.S50.32	49.96	6.6	
53 YES N.S. N.S. N.S. N.S65.10	-5.88	31.1	
54 YES N.S. N.S. N.S. N.S63.82	-1.13	29.5	
55 YES 0.339569431 N.S. N.S. N.S62.54	3.57	28.0	
56         YES         N.S.         N.S.         N.S.         -61.25	8.27	26.5	
57 YES N.S. N.S. N.S. N.S59.96	12.96	25.0	
58 YES 0.428636207 -0.320898333 0.17322 -0.314109 -58.67	17.67	23.5	
59         Planning         NO         0.360666621         -0.268568031         0.35077         N.S.         -57.39	22.39	21.9	
60 Planning NO 0.408381151 -0.344766467 0.56884 N.S56.11	27.12	20.4	

Patient	Ele	Electrode		GLM Weights (z)					MNI Coordinates		
Patient	<u>Number</u>	<u>GLM</u> Designation	Jittering Analysis?	<u>Trial</u>	Perception	<u>Planning</u>	Production	X	¥	Z	
	61	Planning	NO	0.595820839	-0.569480192	0.35142	-0.421541	-54.83	31.86	18.	
	62		YES	0.524204866	-0.442951814	N.S.	-0.552984	-53.56	36.62	17.	
	63		YES	N.S.	N.S.	0.21636	N.S.	-52.30	41.41	15.	
	64		YES	N.S.	N.S.	0.16891	N.S.	-51.05	46.25	14.	
	1		YES	0.199330532	N.S.	N.S.	N.S.	-51.32	21.99	-11.	
	2		YES	N.S.	N.S.	N.S.	N.S.	-51.60	25.01	-6.	
	3		YES	N.S.	N.S.	N.S.	N.S.	-51.96	28.03	-1.	
	4		YES	N.S.	N.S.	N.S.	N.S.	-52.02	31.10	3.	
	5	Excluded - Artifactual	N/A					-51.45	34.30	7.	
	6	Excluded - Artifactual	N/A					-50.11	37.65	12.	
	7		YES	N.S.	N.S.	0.21509	N.S.	-48.15	41.12	17.	
	8	Planning	NO	0.433222481	-0.29695292	0.27275	N.S.	-45.88	44.66	22.	
486L Intraoperative	9	Excluded - On resected tissue	N/A					-54.18	17.45	-7.	
	10	Excluded - On resected tissue	N/A					-54.45	20.48	-2.	
	11	Excluded - On resected tissue	N/A					-54.86	23.48	2.	
	12		YES	0.351479452	-0.235094562	N.S.	N.S.	-55.01	26.54	6.	
	13		YES	N.S.	N.S.	0.25713	N.S.	-54.46	29.73	11.	
	14	Planning	NO	0.423703077	-0.293359211	0.3931	N.S.	-52.87	33.13	16.	
	15		YES	0.283747429	N.S.	N.S.	N.S.	-50.60	36.66	21.	
	16		YES	N.S.	-0.275819064	N.S.	N.S.	-48.07	40.25	25.	

Electrode			Rejected	GLM Weights (z)					MNI Coordinates		
Patient	Number	<u>GLM</u> Designation	Jittering Analysis?	Trial	<b>Perception</b>	<u>Planning</u>	<b>Production</b>	X	Y	Z	
	17	Excluded - On resected tissue	N/A					-56.99	12.93	-4.0	
	18	Excluded - On resected tissue	N/A					-57.18	15.98	0.8	
	19	Excluded - On resected tissue	N/A					-57.46	19.00	5.0	
	20		YES	N.S.	N.S.	0.22818	N.S.	-57.54	22.06	10.4	
	21	Planning- Production	NO	N.S.	-0.171387499	0.21219	0.2306402	-57.00	25.25	15.2	
	22		NO	0.250686174	-0.259133504	N.S.	N.S.	-55.09	28.71	20.0	
	23		YES	N.S.	N.S.	N.S.	N.S.	-52.57	32.30	24.8	
	24		NO	0.343076244	-0.366714191	N.S.	-0.230496	-49.87	35.91	29.0	
	25	Excluded - On resected tissue	N/A					-59.51	8.47	-0.4	
	26	Excluded - On resected tissue	N/A					-59.51	11.55	4.4	
	27	Excluded - On resected tissue	N/A					-59.50	14.63	9.2	
	28	Production	NO	N.S.	-0.187201696	N.S.	0.2421053	-59.23	17.76	14.1	
	29		NO	N.S.	-0.215874396	N.S.	N.S.	-58.29	21.03	18.9	
	30		NO	0.377271915	-0.332261691	N.S.	N.S.	-56.39	24.48	23.7	
	31		YES	N.S.	N.S.	N.S.	N.S.	-53.87	28.06	28.5	
	32		YES	N.S.	-0.169079508	N.S.	N.S.	-51.14	31.69	33.4	
	33	Production	NO	N.S.	N.S.	N.S.	0.683511	-61.68	4.07	3.1	
	34	Production	NO	N.S.	N.S.	N.S.	0.3177786	-61.40	7.21	8.0	

	Electrode		Rejected	GLM Weights (z)					MNI Coordinates		
Patient	Number	<u>GLM</u> Designation	Jittering Analysis?	Trial	<b>Perception</b>	<u>Planning</u>	<b>Production</b>	X	Y	Z	
	35		YES	N.S.	N.S.	N.S.	N.S.	-61.01	10.36	12.8	
	36		YES	N.S.	N.S.	N.S.	N.S.	-60.29	13.58	17.3	
	37		YES	N.S.	N.S.	0.18086	0.2351348	-59.00	16.92	22.0	
	38		YES	N.S.	N.S.	0.23901	N.S.	-57.03	20.38	27.4	
	39		YES	N.S.	N.S.	N.S.	N.S.	-54.57	23.95	32.3	
	40		NO	0.386937203	-0.213720789	N.S.	-0.26783	-51.92	27.56	37.1	
	41	Production	NO	N.S.	-0.16174871	N.S.	0.3531824	-63.53	-0.26	6.7	
	42		YES	N.S.	N.S.	N.S.	N.S.	-62.95	2.94	11.0	
	43		YES	N.S.	N.S.	N.S.	N.S.	-62.17	6.17	16.4	
	44		YES	N.S.	N.S.	N.S.	N.S.	-61.05	9.46	21.3	
	45		YES	N.S.	N.S.	N.S.	N.S.	-59.46	12.86	26.2	
	46		YES	N.S.	N.S.	N.S.	N.S.	-57.37	16.35	31.1	
	47		YES	N.S.	N.S.	0.21809	0.288971	-54.92	19.91	36.0	
	48	Planning	NO	0.484438908	N.S.	0.37849	N.S.	-52.35	23.50	40.9	
	49	Production	NO	N.S.	N.S.	N.S.	0.542823	-65.12	-4.54	10.2	
	50	Production	NO	N.S.	N.S.	N.S.	0.405237	-64.25	-1.29	15.1	
	51	Production	NO	N.S.	N.S.	N.S.	0.4717783	-63.13	2.01	20.1	
	52	Planning- Production	NO	N.S.	N.S.	0.30772	0.2769672	-61.69	5.37	25.0	

	Ele	ectrode	Rejected		GLM Weigh	nts (z)		MN	I Coordin	ates
Patient	Number	<u>GLM</u> Designation	by Jittering Analysis?	Trial	<b>Perception</b>	<u>Planning</u>	Production	X	Y	Z
	53		YES	N.S.	N.S.	N.S.	N.S.	-59.86	8.81	29.
	54		YES	N.S.	N.S.	N.S.	N.S.	-57.64	12.32	34.
	55		YES	N.S.	0.386769066	N.S.	N.S.	-55.16	15.89	39.'
	56		YES	N.S.	N.S.	N.S.	N.S.	-52.63	19.47	44.
	57	Planning- Production	NO	N.S.	N.S.	0.16054	0.8579636	-66.50	-8.77	13.
	58	Production	NO	N.S.	N.S.	N.S.	0.4956224	-65.37	-5.48	18.3
	59		YES	N.S.	N.S.	0.18186	0.3830288	-63.99	-2.13	23.
	60		YES	N.S.	N.S.	N.S.	N.S.	-62.31	1.28	28.
	61		YES	N.S.	N.S.	0.2198	N.S.	-60.32	4.75	33.
	62		YES	N.S.	0.267075774	0.16413	N.S.	-58.02	8.27	38.
	63		YES	N.S.	N.S.	0.21974	N.S.	-55.52	11.85	43.4
	64	Planning	NO	0.862298192	-0.523659597	0.60115	N.S.	-53.01	15.42	48.
	1	Planning	NO	N.S.	N.S.	0.32852	N.S.	-41.71	31.14	32.
510L	2		YES	0.208456962	-0.129416211	0.11041	-0.202355	-44.88	26.49	33.4
Intraoperative	3		YES	N.S.	N.S.	N.S.	N.S.	-48.11	21.86	34.
	4	Planning	NO	N.S.	N.S.	0.45096	N.S.	-51.10	17.19	34.

	Ele	ectrode	Rejected by Jittering Analysis?     GLM Weights (z)					MN	I Coordin	ates
Patient	<u>Number</u>	<u>GLM</u> <u>Designation</u>	by Jittering Analysis?	<u>Trial</u>	Perception	<u>Planning</u>	Production	X	¥	Z
	5	Planning	NO	N.S.	N.S.	0.47576	N.S.	-53.65	12.42	35.2
	6	Planning	NO	0.353484667	-0.23533349	0.50965	N.S.	-55.53	7.42	35.2
	7		YES	N.S.	N.S.	0.22106	0.2977129	-56.99	2.33	35.1
	8		YES	N.S.	N.S.	0.13175	N.S.	-58.20	-2.76	34.9
	9		YES	N.S.	N.S.	0.27999	N.S.	-45.32	31.51	27.8
	10	Planning	NO	0.48716049	N.S.	0.43181	N.S.	-48.11	26.73	28.4
	11	Planning	NO	0.42298043	N.S.	0.25925	-0.32905	-50.91	21.95	29.0
	12	Planning	NO	N.S.	N.S.	0.92591	N.S.	-53.51	17.16	29.:
	13	Planning	NO	0.426277094	N.S.	0.48511	N.S.	-55.70	12.28	29.8
	14	Planning	NO	0.504105748	N.S.	0.56732	N.S.	-57.35	7.28	29.9
	15		YES	N.S.	N.S.	0.21407	N.S.	-58.59	2.21	29.8
	16		YES	N.S.	N.S.	N.S.	N.S.	-59.60	-2.86	29.1
	17		YES	N.S.	N.S.	N.S.	N.S.	-48.80	31.85	22.8

	Ele	ectrode	Rejected		GLM Weigh	nts (z)		MN	I Coordin	ates
Patient	<u>Number</u>	<u>GLM</u> Designation	by Jittering Analysis?	Trial	Perception	<u>Planning</u>	Production	X	Y	Z
	18	Planning	NO	N.S.	N.S.	0.49852	N.S.	-51.18	26.92	23.3
	19	Planning	NO	0.266358871	N.S.	0.60408	N.S.	-53.47	21.97	23.7
	20	Planning	NO	0.392209656	N.S.	0.7459	N.S.	-55.64	17.02	24.1
	21	Planning	NO	0.642420831	-0.333914246	0.56251	N.S.	-57.46	12.05	24.3
	22	Planning	NO	0.473355443	N.S.	0.26827	N.S.	-58.84	7.05	24.4
	23		YES	N.S.	N.S.	0.22275	N.S.	-59.85	2.02	24.5
	24		YES	N.S.	N.S.	N.S.	N.S.	-60.68	-3.00	24.5
	25		YES	0.359298982	N.S.	N.S.	N.S.	-51.02	31.90	17.5
	26	Planning	NO	N.S.	N.S.	0.41302	N.S.	-53.35	26.93	17.9
	27	Planning	NO	N.S.	N.S.	0.43256	N.S.	-55.43	21.89	18.3
	28	Planning	NO	0.50620564	N.S.	0.49832	N.S.	-57.32	16.83	18.5
	29	Planning	NO	0.693185708	-0.480408783	0.66969	-0.363006	-58.87	11.79	18.8
	30	Planning	NO	0.311544952	-0.236325395	0.27764	N.S.	-59.96	6.79	19.(
	31		YES	N.S.	-0.217762971	0.18866	N.S.	-60.73	1.81	19.1
	32	Planning- Production	NO	N.S.	-0.566568899	0.24758	0.4254089	-61.40	-3.17	19.3
	33	Planning	NO	0.484665653	N.S.	0.19299	-0.390847	-52.45	31.82	11.9
	34	Perception- Planning	NO	N.S.	0.481418562	0.28298	N.S.	-54.70	26.82	12.4
	35	Planning	NO	N.S.	N.S.	0.51738	N.S.	-56.68	21.74	12.7
	36	Planning	NO	N.S.	N.S.	0.69998	N.S.	-58.40	16.64	13.0
	37	Planning	NO	0.474669863	-0.326314496	0.58215	N.S.	-59.79	11.55	13.2

	Ele	ectrode	Rejected		GLM Weigh	nts (z)		MN	I Coordin	ates
Patient	Number	<u>GLM</u> Designation	by Jittering Analysis?	Trial	<b>Perception</b>	<u>Planning</u>	Production	X	Y	Z
	38	Planning	NO	N.S.	-0.283509145	0.21357	N.S.	-60.62	6.57	13.5
	39	Planning	NO	N.S.	-0.401969205	0.30129	N.S.	-61.20	1.63	13.8
	40	Production	NO	0.489013025	-0.521472245	N.S.	0.868548	-61.76	-3.31	14.0
	41	Planning	NO	0.632453312	-0.33482728	0.5754	-0.36381	-53.38	31.71	6.3
	42	Planning	NO	N.S.	N.S.	0.44637	N.S.	-55.49	26.69	6.1
	43	Planning	NO	N.S.	N.S.	0.37837	N.S.	-57.33	21.61	7.1
	44	Planning	NO	0.305345451	N.S.	0.4285	N.S.	-58.88	16.51	7.4
	45	Planning	NO	0.824084787	-0.716412383	0.82188	N.S.	-60.05	11.44	7.3
	46	Planning	NO	N.S.	-0.291786065	0.16488	N.S.	-60.78	6.45	8.1
	47		YES	N.S.	-0.189180259	N.S.	N.S.	-61.28	1.50	8.4
	48	Planning- Production	NO	N.S.	N.S.	0.13964	0.2048724	-61.80	-3.43	8.1
	49	Planning	NO	N.S.	N.S.	0.33685	N.S.	-53.98	31.59	0.0
	50	Perception- Planning	NO	N.S.	0.32769173	0.37152	N.S.	-55.92	26.57	1.1
	51		YES	N.S.	0.171237471	0.29163	N.S.	-57.61	21.51	1.5
	52		YES	N.S.	N.S.	0.09562	N.S.	-58.98	16.43	1.9
	53	Planning	NO	N.S.	-0.384929302	0.29977	N.S.	-60.00	11.38	2.3
	54		YES	N.S.	N.S.	N.S.	N.S.	-60.66	6.39	2.7
	55		YES	N.S.	N.S.	N.S.	N.S.	-61.13	1.43	3.1
	56	Planning- Production	NO	0.137276056	-0.154542337	0.09038	0.0930603	-61.65	-3.50	3.5
	57	Planning	NO	N.S.	N.S.	0.3698	N.S.	-54.39	31.49	-4.9
	58		YES	N.S.	N.S.	0.11547	N.S.	-56.18	26.47	-4.4
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	Ele	ectrode	Rejected		GLM Weigh	nts (z)		MN	I Coordin	ates
Patient	Number	<u>GLM</u> Designation	by Jittering Analysis?	Trial	<b>Perception</b>	<u>Planning</u>	Production	X	¥	Z
	59		YES	N.S.	N.S.	N.S.	N.S.	-57.73	21.43	-3.
	60		YES	N.S.	N.S.	N.S.	N.S.	-58.98	16.38	-3.
	61		YES	N.S.	N.S.	0.04688	N.S.	-59.91	11.35	-3.
	62		YES	N.S.	N.S.	N.S.	N.S.	-60.55	6.36	-2.
	63		YES	N.S.	N.S.	0.04246	N.S.	-61.03	1.40	-2.
	64		YES	N.S.	N.S.	N.S.	N.S.	-61.56	-3.54	-1.
	1		NO	0.208118379	-0.199180401	N.S.	N.S.	-53.95	26.89	19.
	2	Planning	NO	0.686115686	-0.531954313	0.42777	-0.439542	-54.48	25.15	17.
	3		YES	N.S.	N.S.	N.S.	N.S.	-55.00	23.42	14.
	4		YES	N.S.	N.S.	N.S.	N.S.	-55.53	21.69	12.
	5	Excluded - Artifactual	N/A					-56.06	19.94	9.
	6	Excluded - Artifactual	N/A					-56.59	18.16	7.
	7		YES	N.S.	N.S.	0.17016	N.S.	-57.11	16.35	4.
494L	8		YES	N.S.	N.S.	N.S.	N.S.	-57.64	14.49	2.
	9		YES	0.28726197	N.S.	N.S.	N.S.	-58.16	12.58	-0.
	10		YES	0.473359244	-0.31610037	0.1646	-0.30333	-58.68	10.61	-3.
	11		YES	0.541709278	-0.363313985	0.22838	-0.333031	-59.20	8.61	-5.
	12		YES	0.400007469	N.S.	0.24115	N.S.	-59.71	6.58	-8.
	13		YES	N.S.	N.S.	N.S.	N.S.	-60.23	4.53	-11.
	14		YES	N.S.	N.S.	N.S.	N.S.	-60.75	2.47	-14.

	ient Electrode <u>Number GLM</u> <u>Designatio</u>		Rejected		GLM Weigh	nts (z)		MN	I Coordin	ates
Patient	<u>Number</u>	<u>GLM</u> Designation	by Jittering Analysis?	Trial	Perception	<u>Planning</u>	Production	X	Y	Z
	15		YES	N.S.	N.S.	0.25643	N.S.	-61.26	0.41	-16.8
	16		YES	N.S.	N.S.	-0.33144	-0.31285	-61.77	-1.63	-19.6
	17	Planning	NO	0.400029818	-0.482544483	0.52059	N.S.	-54.90	24.35	22.1
	18	Planning	NO	0.507576664	-0.484240639	0.37696	N.S.	-55.42	22.62	19.5
	19		YES	N.S.	N.S.	N.S.	N.S.	-55.95	20.90	17.0
	20		YES	N.S.	N.S.	0.34265	0.4062324	-56.47	19.18	14.5
	21		YES	N.S.	N.S.	0.40761	N.S.	-57.00	17.45	12.0
	22		YES	N.S.	N.S.	0.27841	N.S.	-57.52	15.69	9.5
	23		YES	N.S.	N.S.	N.S.	N.S.	-58.05	13.89	6.9
	24		YES	N.S.	N.S.	N.S.	N.S.	-58.57	12.04	4.3
	25		YES	N.S.	N.S.	0.34066	N.S.	-59.09	10.12	1.1
	26		YES	0.320965793	N.S.	0.21487	-0.398714	-59.61	8.15	-1.0
	27		YES	N.S.	N.S.	0.17307	-0.285865	-60.12	6.14	-3.7
	28		YES	N.S.	N.S.	0.23218	N.S.	-60.64	4.10	-6.5
	29		YES	N.S.	N.S.	N.S.	N.S.	-61.15	2.04	-9.3
	30		YES	N.S.	N.S.	N.S.	N.S.	-61.66	-0.03	-12.1
	31		YES	N.S.	N.S.	N.S.	N.S.	-62.17	-2.09	-14.9
	32		YES	N.S.	N.S.	N.S.	N.S.	-62.69	-4.14	-17.3

	Ele	ectrode	Rejected		GLM Weigh	nts (z)		MN	I Coordin	ates
Patient	<u>Number</u>	<u>GLM</u> Designation	by Jittering Analysis?	<u>Trial</u>	<b>Perception</b>	<u>Planning</u>	Production	X	Y	Z
	33	Planning	NO	0.595276233	-0.332447284	0.58862	N.S.	-55.84	21.80	24.4
	34	Planning	NO	0.482909313	-0.548620445	0.52244	N.S.	-56.37	20.08	21.8
	35		YES	N.S.	-0.387882246	N.S.	N.S.	-56.89	18.37	19.3
	36		YES	N.S.	N.S.	N.S.	N.S.	-57.41	16.65	16.8
	37	Planning	NO	N.S.	N.S.	0.41161	N.S.	-57.94	14.93	14.2
	38		YES	N.S.	N.S.	N.S.	N.S.	-58.46	13.19	11.3
	39		YES	N.S.	N.S.	N.S.	N.S.	-58.98	11.41	9.1
	40		YES	N.S.	N.S.	N.S.	N.S.	-59.50	9.57	6.5
	41		YES	N.S.	N.S.	0.40159	N.S.	-60.02	7.65	3.8
	42	Planning	NO	N.S.	N.S.	0.42644	-0.466477	-60.53	5.66	1.0
	43		YES	N.S.	N.S.	0.19285	N.S.	-61.04	3.64	-1.7
	44		YES	N.S.	N.S.	0.25574	N.S.	-61.55	1.60	-4.:
	45		YES	0.334387361	N.S.	N.S.	N.S.	-62.07	-0.47	-7.3
	46		YES	N.S.	N.S.	N.S.	N.S.	-62.58	-2.53	-10.2
	47		YES	N.S.	N.S.	0.23662	N.S.	-63.09	-4.59	-13.0
	48		YES	N.S.	N.S.	N.S.	N.S.	-63.59	-6.64	-15.9
	49		YES	N.S.	N.S.	0.25118	N.S.	-56.79	19.24	26.7
	50		YES	N.S.	N.S.	0.31929	N.S.	-57.31	17.52	24.1

Patient	Ele	ectrode	Rejected		GLM Weigh	nts (z)		MN	I Coordin	ates
Patient	<u>Number</u>	<u>GLM</u> Designation	by Jittering Analysis?	Trial	Perception	<u>Planning</u>	Production	X	Y	Z
	51	Planning	NO	0.659057171	-0.736069217	0.55545	N.S.	-57.83	15.81	21.5
	52	Planning	NO	0.359460492	-0.502344592	0.3998	N.S.	-58.35	14.10	19.(
	53		YES	N.S.	N.S.	0.1782	N.S.	-58.88	12.39	16.4
	54		YES	N.S.	N.S.	N.S.	N.S.	-59.40	10.65	13.8
	55		YES	N.S.	N.S.	0.15689	0.2936284	-59.91	8.88	11.2
	56		YES	N.S.	N.S.	0.13057	N.S.	-60.43	7.06	8.0
	57		YES	N.S.	N.S.	N.S.	N.S.	-60.94	5.13	5.8
	58		YES	N.S.	N.S.	N.S.	N.S.	-61.45	3.14	3.0
	59		YES	N.S.	N.S.	N.S.	N.S.	-61.96	1.12	0.2
	60		NO	0.610075759	N.S.	N.S.	-0.499255	-62.47	-0.93	-2.0
	61	Planning	NO	0.722563146	-0.333441997	0.25261	N.S.	-62.98	-2.99	-5.4
	62		YES	0.432883167	N.S.	0.44382	N.S.	-63.49	-5.05	-8.3
	63		YES	N.S.	N.S.	N.S.	N.S.	-63.99	-7.11	-11.2
	64		YES	N.S.	N.S.	N.S.	N.S.	-64.50	-9.16	-14.0
	65		YES	N.S.	N.S.	N.S.	N.S.	-57.74	16.66	28.9
	66		YES	N.S.	N.S.	0.39927	N.S.	-58.26	14.94	26.4
	67	Planning	NO	N.S.	-0.330036165	0.317	N.S.	-58.78	13.23	23.8
	68	Planning- Production	NO	N.S.	N.S.	0.20225	0.3570327	-59.30	11.52	21.2

	Ele	ectrode	Rejected		GLM Weigh	nts (z)		MN	I Coordin	ates
Patient	<u>Number</u>	<u>GLM</u> Designation	by Jittering Analysis?	Trial	Perception	<u>Planning</u>	Production	X	¥	Z
	69		YES	N.S.	N.S.	N.S.	0.3761229	-59.81	9.81	18.0
	70		YES	N.S.	N.S.	0.16391	0.4275066	-60.33	8.07	16.0
	71		YES	N.S.	N.S.	N.S.	0.4175961	-60.84	6.30	13.3
	72	Production	NO	N.S.	N.S.	N.S.	0.3403755	-61.36	4.46	10.0
	73		YES	N.S.	N.S.	0.15594	N.S.	-61.87	2.55	7.8
	74		YES	0.534607136	-0.337739136	0.19433	-0.31088	-62.38	0.58	5.(
	75	Planning	NO	0.676132778	N.S.	0.2744	-0.646717	-62.88	-1.44	2.1
	76		NO	0.811309793	N.S.	N.S.	-0.618533	-63.39	-3.48	-0.7
	77		YES	N.S.	N.S.	N.S.	N.S.	-63.89	-5.53	-3.6
	78		YES	0.2371267	N.S.	N.S.	N.S.	-64.40	-7.58	-6.5
	79		YES	N.S.	N.S.	0.16549	N.S.	-64.90	-9.64	-9.4
	80		YES	N.S.	N.S.	N.S.	N.S.	-65.41	-11.68	-12.2
	81		YES	N.S.	N.S.	0.32735	N.S.	-58.69	14.07	31.2
	82		YES	N.S.	N.S.	0.29691	N.S.	-59.20	12.35	28.0
	83		YES	N.S.	-0.324217746	N.S.	N.S.	-59.72	10.63	26.0
	84		YES	N.S.	N.S.	N.S.	N.S.	-60.23	8.92	23.3
	85	Planning- Production	NO	N.S.	N.S.	0.2685	0.4708179	-60.75	7.20	20.7

	Ele	ectrode	Rejected		GLM Weigh	nts (z)		MN	I Coordin	ates
Patient	Number	<u>GLM</u> Designation	by Jittering Analysis?	Trial	Perception	<u>Planning</u>	Production	X	Y	Z
	86	Planning	NO	N.S.	N.S.	0.53598	N.S.	-61.26	5.45	18.0
	87	Planning- Production	NO	N.S.	N.S.	0.3821	0.807277	-61.77	3.67	15.3
	88	Planning- Production	NO	0.339863558	N.S.	0.483	0.6350564	-62.28	1.83	12.0
	89	Planning	NO	0.406196999	N.S.	0.2234	N.S.	-62.79	-0.07	9.8
	90	Planning	NO	0.613314132	N.S.	0.27609	-0.527217	-63.30	-2.03	7.0
	91	Planning	NO	0.809619678	N.S.	0.27393	-0.711245	-63.80	-4.03	4.3
	92	Planning	NO	0.796250048	N.S.	0.19069	-0.664806	-64.30	-6.05	1.1
	93		YES	N.S.	N.S.	N.S.	N.S.	-64.80	-8.09	-1.'
	94		YES	N.S.	N.S.	N.S.	0.2651268	-65.30	-10.14	-4.0
	95		YES	N.S.	N.S.	N.S.	N.S.	-65.81	-12.18	-7.:
	96		YES	N.S.	N.S.	0.14703	N.S.	-66.31	-14.22	-10.4
	97		YES	N.S.	N.S.	0.36278	N.S.	-59.63	11.47	33.4
	98		YES	N.S.	N.S.	0.20728	N.S.	-60.15	9.74	30.8
	99		YES	N.S.	-0.229651013	N.S.	N.S.	-60.66	8.02	28.2
	100	Production	NO	0.396756157	-0.43536151	N.S.	0.5928112	-61.17	6.30	25.5
	101	Production	NO	0.442640228	-0.559696993	N.S.	0.956519	-61.68	4.57	22.8
	102	Planning- Production	NO	0.317576577	N.S.	0.4956	1.2112041	-62.19	2.81	20.1

	Ele	ectrode	Rejected		GLM Weigh	nts (z)		MN	I Coordin	ates
Patient	<u>Number</u>	<u>GLM</u> Designation	Jittering Analysis?	<u>Trial</u>	Perception	<u>Planning</u>	Production	X	¥	Z
	103	Planning- Production	NO	N.S.	N.S.	0.56988	0.8823763	-62.70	1.02	17.4
	104	Planning	NO	0.571387409	-0.337086547	0.42918	N.S.	-63.21	-0.82	14.0
	105	Production	NO	0.320336712	-0.369097192	N.S.	0.7885149	-63.71	-2.71	11.3
	106	Perception- Planning	NO	0.466984061	0.277848136	0.29228	-0.312234	-64.22	-4.66	8.9
	107	Perception- Planning	NO	N.S.	0.54576096	0.4206	-0.347955	-64.72	-6.64	6.0
	108		YES	0.49289152	N.S.	N.S.	-0.387588	-65.22	-8.64	3.0
	109		YES	N.S.	N.S.	N.S.	N.S.	-65.72	-10.67	0.1
	110		YES	N.S.	N.S.	N.S.	N.S.	-66.21	-12.71	-2.8
	111		YES	N.S.	N.S.	N.S.	N.S.	-66.72	-14.74	-5.7
	112	Production	NO	N.S.	N.S.	N.S.	0.3636677	-67.22	-16.77	-8.0
	113		YES	-0.349168965	0.35204141	0.24921	0.5308946	-60.57	8.86	35.0
	114		YES	N.S.	N.S.	0.14352	N.S.	-61.09	7.13	33.0
	115		YES	N.S.	N.S.	N.S.	N.S.	-61.60	5.40	30.3
	116	Production	NO	0.47725241	-0.419504301	N.S.	0.8882201	-62.11	3.67	27.6
	117	Planning- Production	NO	0.511615973	N.S.	0.68658	0.3154233	-62.61	1.93	24.9
	118	Planning- Production	NO	0.54340936	-0.241996841	0.39515	0.9863127	-63.12	0.16	22.1
	119	Production	NO	N.S.	N.S.	N.S.	0.8051178	-63.63	-1.64	19.4

	Ele	ectrode	Rejected		GLM Weigh	nts (z)		MN	I Coordin	ates
Patient	<u>Number</u>	<u>GLM</u> Designation	Jittering Analysis?	<u>Trial</u>	Perception	<u>Planning</u>	Production	X	¥	Z
	120	Production	NO	0.569992103	-0.831687471	N.S.	0.8015205	-64.13	-3.48	16.5
	121	Production	NO	N.S.	-0.460260109	N.S.	0.4765639	-64.63	-5.37	13.7
	122	Production	NO	N.S.	N.S.	N.S.	0.5489617	-65.13	-7.30	10.8
	123	Planning	NO	0.4876166	N.S.	0.23818	N.S.	-65.63	-9.26	7.9
	124		YES	0.403989498	N.S.	0.19335	N.S.	-66.13	-11.25	4.9
	125	Planning	NO	0.461298078	N.S.	0.17854	N.S.	-66.63	-13.26	2.0
	126		NO	0.660195039	N.S.	N.S.	-0.570145	-67.13	-15.29	-0.9
	127	Perception- Planning	NO	N.S.	0.395354754	0.19466	N.S.	-67.63	-17.31	-3.8
	128	Production	NO	N.S.	N.S.	N.S.	0.3904261	-68.13	-19.33	-6.8
	129		YES	N.S.	-0.290047821	0.32529	N.S.	-28.03	58.45	31.7
	130	Planning	NO	N.S.	-0.478665538	0.4469	N.S.	-31.02	55.46	31.8
	131		YES	N.S.	-0.472549924	0.44554	N.S.	-34.06	52.48	31.9
	132		YES	N.S.	N.S.	N.S.	N.S.	-45.21	37.44	32.8
	133		YES	N.S.	N.S.	0.41443	N.S.	-46.83	34.39	33.1
	134	Planning- Production	NO	N.S.	N.S.	0.55915	0.4932839	-48.37	31.31	33.4
	135		YES	N.S.	N.S.	0.19102	N.S.	-30.35	58.25	28.2
	136		YES	N.S.	N.S.	0.44957	N.S.	-33.35	55.26	28.3

	Ele	ectrode	Rejected	Rejected by Jittering Analysis?     GLM Weights (z)       Perception     Planning     Production					I Coordin	ates
Patient	Number	<u>GLM</u> Designation	by Jittering Analysis?	Trial	Perception	<u>Planning</u>	Production	X	Y	Z
	137		YES	N.S.	N.S.	0.30719	N.S.	-36.39	52.27	28.4
	138		YES	N.S.	N.S.	N.S.	N.S.	-39.18	49.27	28.0
	139		YES	N.S.	N.S.	0.23668	N.S.	-41.53	46.27	28.
	140		YES	N.S.	N.S.	N.S.	N.S.	-43.49	43.26	28.9
	141		YES	N.S.	N.S.	-0.22107	N.S.	-45.20	40.23	29.2
	142		YES	N.S.	N.S.	N.S.	N.S.	-46.76	37.18	29.:
	143	Planning	NO	N.S.	-0.359056089	0.48131	N.S.	-48.24	34.10	29.8
	144	Planning	NO	0.415476798	-0.531732573	0.43211	N.S.	-49.66	31.01	30.1
	145		YES	-0.480200486	N.S.	0.24176	0.4394925	-32.67	58.07	24.8
	146		YES	-0.358477925	N.S.	N.S.	N.S.	-35.65	55.08	24.9
	147		YES	N.S.	N.S.	N.S.	N.S.	-38.66	52.08	25.0
	148		YES	N.S.	N.S.	N.S.	N.S.	-41.38	49.08	25.
	149		YES	N.S.	N.S.	N.S.	N.S.	-43.49	46.07	25.3
	150		YES	N.S.	N.S.	-0.22454	N.S.	-45.19	43.04	25.0
	151		YES	N.S.	N.S.	-0.16513	N.S.	-46.69	39.99	25.9
	152		YES	N.S.	N.S.	N.S.	N.S.	-48.08	36.91	26.2

	Ele	ectrode	Rejected	d GLM Weights (z) g S? Trial Perception Planning Production					I Coordin	ates
Patient	Number	<u>GLM</u> Designation	by Jittering Analysis?	Trial	Perception	<u>Planning</u>	Production	X	¥	Z
	153	Planning	NO	N.S.	N.S.	0.46842	N.S.	-49.44	33.81	26.:
	154		YES	N.S.	N.S.	0.42377	N.S.	-50.78	30.69	26.
	155		YES	N.S.	N.S.	0.23393	N.S.	-34.74	57.90	21.
	156		YES	-0.418091248	N.S.	N.S.	N.S.	-37.62	54.92	21.
	157		YES	N.S.	N.S.	N.S.	N.S.	-40.53	51.92	21.
	158		YES	N.S.	N.S.	0.22088	N.S.	-43.17	48.90	21.
	159	Planning	NO	0.431468356	-0.466100735	0.52114	N.S.	-45.00	45.88	22.
	160		YES	N.S.	N.S.	N.S.	N.S.	-46.49	42.83	22.
	161		NO	N.S.	N.S.	-0.1945	N.S.	-47.83	39.75	22.
	162		YES	N.S.	-0.270173192	N.S.	N.S.	-49.13	36.64	22.
	163		YES	N.S.	N.S.	N.S.	N.S.	-50.41	33.51	23.
	164		YES	N.S.	N.S.	0.45142	N.S.	-51.70	30.36	23.
	165		YES	N.S.	N.S.	N.S.	0.395963	-36.46	57.73	18.
	166		YES	N.S.	N.S.	N.S.	N.S.	-39.10	54.77	18.
	167		YES	N.S.	N.S.	N.S.	N.S.	-41.70	51.78	18.
	168	Planning	NO	N.S.	-0.364397159	0.28151	N.S.	-44.01	48.77	18.
	169	Planning	NO	N.S.	-0.40258237	0.41346	N.S.	-45.81	45.72	18.

	Ele	ectrode	Rejected		GLM Weigh	nts (z)		MN	I Coordin	ates
Patient	Number	<u>GLM</u> Designation	by Jittering Analysis?	Trial	Perception	<u>Planning</u>	Production	X	Y	Z
	170		YES	N.S.	-0.289334239	N.S.	N.S.	-47.28	42.63	18.9
	171		YES	N.S.	N.S.	N.S.	0.3799264	-48.60	39.51	19.1
	172		YES	N.S.	N.S.	-0.1498	N.S.	-49.88	36.36	19.′
	173		YES	N.S.	N.S.	0.18761	N.S.	-51.14	33.19	20.
	174	Planning	NO	0.513333262	-0.361443944	0.5672	N.S.	-52.43	30.02	20.:
	175		YES	N.S.	N.S.	0.24259	N.S.	-37.87	57.54	14.
	176		NO	N.S.	N.S.	N.S.	N.S.	-40.19	54.60	14.
	177		YES	N.S.	N.S.	N.S.	N.S.	-42.40	51.65	14.9
	178		YES	N.S.	N.S.	N.S.	N.S.	-44.41	48.64	15.
	179		YES	0.566479099	-0.532987781	0.21628	N.S.	-46.16	45.56	15.:
	180	Planning	NO	0.719029168	-0.851926431	0.28542	N.S.	-47.69	42.43	15.
	181		YES	N.S.	N.S.	N.S.	N.S.	-49.08	39.25	16.
	182		YES	N.S.	-0.490610546	0.22024	N.S.	-50.39	36.06	16.
	183		YES	0.327363325	-0.333835397	N.S.	N.S.	-51.69	32.86	16.9
	184		YES	0.386698715	N.S.	0.21871	N.S.	-53.00	29.66	17.4

	Ele	ectrode	Rejected		GLM Weigh	nts (z)		MN	I Coordin	ates
Patient	Number	<u>GLM</u> Designation	by Jittering Analysis?	Trial	Perception	<u>Planning</u>	Production	X	Y	Z
	185		YES	N.S.	N.S.	N.S.	N.S.	-39.09	57.31	11.4
	186		YES	N.S.	N.S.	N.S.	N.S.	-41.09	54.40	11.5
	187		YES	N.S.	N.S.	N.S.	N.S.	-42.93	51.47	11.0
	188		YES	N.S.	N.S.	N.S.	N.S.	-44.65	48.48	11.8
	189		YES	N.S.	N.S.	N.S.	N.S.	-46.34	45.37	12.1
	190		YES	N.S.	N.S.	0.20705	N.S.	-47.94	42.19	12.5
	191	Planning	NO	N.S.	-0.397844856	0.34621	N.S.	-49.40	38.97	12.9
	192	Excluded - Artifactual	N/A					-50.78	35.73	13.3
	193		YES	N.S.	-0.350056308	N.S.	N.S.	-52.12	32.50	13.8
	194		YES	0.422534773	-0.41124449	0.2455	N.S.	-53.48	29.28	14.2
	195		YES	-0.361720879	N.S.	N.S.	0.3754501	-40.27	57.03	8.1
	196		YES	N.S.	N.S.	0.34366	N.S.	-42.00	54.13	8.2
	197		YES	N.S.	N.S.	N.S.	N.S.	-43.54	51.22	8.4
	198		YES	0.550820192	-0.62394099	N.S.	N.S.	-44.99	48.24	8.0
	199		YES	N.S.	N.S.	0.37912	N.S.	-46.64	45.10	8.9
	200	Planning- Production	NO	N.S.	N.S.	0.38229	0.4774449	-48.24	41.89	9.3
	201	Planning	NO	N.S.	-0.360895411	0.26511	N.S.	-49.74	38.64	9.3
	202		YES	N.S.	N.S.	N.S.	N.S.	-51.14	35.37	10.2
	203		YES	0.596919851	-0.590250453	0.42056	N.S.	-52.52	32.11	10.7
	204	Planning	NO	0.583516417	-0.485800061	0.4441	N.S.	-53.91	28.88	11.1

	Ele	ectrode	Rejected		GLM Weigh		MN	I Coordin	ates	
Patient	<u>Number</u>	<u>GLM</u> Designation	by Jittering Analysis?	Trial	<b>Perception</b>	<u>Planning</u>	Production	X	Y	Z
	205	Planning- Production	NO	-0.525620905	N.S.	0.33882	0.7154558	-41.47	56.69	4.9
	206		YES	-0.518968937	N.S.	N.S.	0.3371256	-43.02	53.78	5.1
	207		YES	N.S.	-0.442286298	0.2905	N.S.	-44.40	50.85	5.2
	208		YES	N.S.	N.S.	N.S.	N.S.	-45.76	47.85	5.5
	209		YES	N.S.	N.S.	N.S.	N.S.	-47.23	44.73	5.8
	210	Planning- Production	NO	N.S.	N.S.	0.40709	0.420953	-48.74	41.52	6.2
	211		YES	N.S.	N.S.	N.S.	N.S.	-50.20	38.26	6.0
	212	Planning- Production	NO	N.S.	-0.417270935	0.34341	0.4051321	-51.59	34.98	7.1
	213	Planning	NO	N.S.	-0.887215574	0.72968	N.S.	-52.97	31.70	7.0
	214	Planning	NO	0.984719301	-0.857115685	0.5447	-0.670937	-54.38	28.46	8.0
	215	Planning- Production	NO	-0.395708892	N.S.	0.28817	0.5820646	-42.65	56.30	1.8
	216		YES	N.S.	N.S.	N.S.	0.3792047	-44.09	53.37	1.9
	217		YES	N.S.	-0.353313752	N.S.	N.S.	-45.40	50.42	2.1
	218		YES	N.S.	N.S.	N.S.	N.S.	-46.69	47.40	2.4
	219		YES	N.S.	N.S.	N.S.	N.S.	-48.04	44.28	2.7
	220		YES	N.S.	N.S.	N.S.	N.S.	-49.44	41.09	3.1
	221		YES	N.S.	N.S.	0.28888	N.S.	-50.83	37.84	3.5
	222		YES	N.S.	N.S.	N.S.	N.S.	-52.20	34.57	4.0
	223		YES	N.S.	N.S.	N.S.	N.S.	-53.56	31.29	4.5
	224	Planning	NO	0.789961341	-0.813582119	0.54161	-0.419804	-54.97	28.04	4.9

	Ele	ectrode	Rejected		GLM Weigl	hts (z)		MN	I Coordin	ates
Patient	Number	<u>GLM</u> Designation	by Jittering Analysis?	Trial	<b>Perception</b>	<u>Planning</u>	Production	x	¥	Z
	1		YES	N.S.	N.S.	N.S.	N.S.	-24.36	8.56	-14.
	2		YES	N.S.	N.S.	N.S.	N.S.	-34.51	6.83	-12.
	3		YES	N.S.	0.083650613	N.S.	N.S.	-44.92	6.10	-10.
	4		YES	N.S.	N.S.	N.S.	N.S.	-55.31	4.87	-9.
	15	Excluded - In white matter	N/A					-28.89	-39.57	13.
	16	Excluded - In white matter	N/A					-31.61	-35.42	11.
	17	Excluded - In white matter	N/A					-34.46	-31.23	9.
	18	Perception- Production	NO	N.S.	0.58311654	-0.14119	0.490018	-37.48	-27.08	7.
442L Chronic	19	Perception- Production	NO	N.S.	0.784240992	-0.09455	0.4448385	-40.68	-23.02	5.
Chiome	20	Perception- Production	NO	N.S.	0.508104183	-0.17185	0.1970076	-43.99	-19.00	4.
	21	Perception- Production	NO	N.S.	0.526808366	-0.10338	0.1661044	-47.33	-14.98	2.
	22	Perception	NO	N.S.	0.370590562	-0.11896	N.S.	-50.64	-10.89	1.
	23		YES	N.S.	0.270722266	N.S.	N.S.	-53.85	-6.68	-0.
	24		YES	N.S.	0.266031243	N.S.	N.S.	-56.93	-2.40	-1.
	25		YES	N.S.	0.338945069	N.S.	0.1845125	-59.87	1.90	-2.
	26	Excluded - Outside of brain	N/A					-62.72	6.20	-4.

	Electrode           Electrode           Number         GLM           Designation		Rejected		GLM Weigh	nts (z)		MN	I Coordin	ates
Patient	Number	<u>GLM</u> Designation	by Jittering Analysis?	Trial	<b>Perception</b>	<u>Planning</u>	<b>Production</b>	X	¥	Z
	53	Perception	NO	N.S.	0.289494235	-0.11427	N.S.	-32.29	-15.55	2.1
	54	Perception	NO	N.S.	0.35177898	-0.19546	N.S.	-37.39	-16.58	3.'
	55	Perception	NO	N.S.	0.96298254	N.S.	N.S.	-42.17	-17.82	5.0
	56	Perception- Production	NO	N.S.	0.76026345	-0.16449	0.1437885	-46.74	-19.06	6.2
	57	Perception	NO	N.S.	0.359656568	-0.09559	N.S.	-51.32	-19.98	7.4
	58	Perception	NO	0.248615927	0.232630454	-0.30897	N.S.	-55.99	-20.33	9.1
	59		YES	N.S.	0.255673852	N.S.	N.S.	-60.38	-20.31	11.3
	60	Perception	NO	N.S.	0.442240776	N.S.	N.S.	-64.70	-20.26	14.3
	61	Excluded - Seizure focus	N/A					-30.24	-29.45	-19.9
	62	Excluded - Seizure focus	N/A					-37.81	-35.77	-22.7
	63	Excluded - Seizure focus	N/A					-46.43	-41.41	-23.9
	64		YES	N.S.	N.S.	-0.12941	N.S.	-54.78	-46.10	-20.7
	65	Excluded - Seizure focus	N/A					-27.91	-46.32	-13.
	66	Excluded - Seizure focus	N/A					-25.77	-36.99	-15.7
	67	Excluded - Seizure focus	N/A					-23.51	-27.35	-20.0
	68	Excluded - Seizure focus	N/A					-21.36	-17.70	-25.3
	69	Excluded - Seizure focus	N/A					-19.67	-8.28	-30.1

	Ele	ectrode	Rejected			MN	I Coordin	ates		
Patient	<u>Number</u>	<u>GLM</u> Designation	by Jittering Analysis?	<u>Trial</u>	<b>Perception</b>	<u>Planning</u>	Production	<u>X</u>	Y	Z
	70	Excluded - Seizure focus	N/A					-19.26	0.57	-34.3
	71		YES	N.S.	N.S.	N.S.	N.S.	-21.82	8.59	-40.3
	72		YES	N.S.	N.S.	N.S.	N.S.	-25.77	17.82	-39.8
	73		YES	N.S.	N.S.	N.S.	N.S.	-32.46	23.63	-33.7
	74		YES	N.S.	N.S.	-0.09122	N.S.	-40.97	24.79	-27.9
	75		YES	N.S.	N.S.	N.S.	N.S.	-51.44	35.91	7.3
	76		YES	N.S.	N.S.	0.28314	N.S.	-52.92	32.42	10.5
	77		YES	N.S.	N.S.	0.09471	0.1694217	-54.23	28.83	13.7
	78	Planning	NO	N.S.	-0.231101435	0.25731	N.S.	-55.43	25.22	17.0
	79		YES	N.S.	N.S.	N.S.	N.S.	-56.67	21.67	20.5
	80		YES	N.S.	-0.212510613	0.26971	N.S.	-46.43	42.01	-6.2
	81	Planning- Production	NO	N.S.	-0.257939163	0.34577	0.2372587	-48.69	38.87	-2.0
	82		YES	N.S.	N.S.	0.33808	N.S.	-50.82	35.76	0.4
	83		YES	0.198960758	-0.202421681	0.2593	N.S.	-52.74	32.56	3.3
	84		YES	N.S.	N.S.	0.24347	N.S.	-54.38	29.18	6.4
	85		YES	N.S.	N.S.	0.16327	N.S.	-55.80	25.67	9.3
	86		YES	N.S.	-0.104057039	0.09603	N.S.	-57.08	22.13	13.0
	87		YES	N.S.	N.S.	N.S.	N.S.	-58.38	18.62	16.5
	88		YES	N.S.	N.S.	0.17666	N.S.	-46.58	38.30	-10.2
	89		YES	N.S.	N.S.	0.18563	N.S.	-49.15	35.24	-6.8
	90		YES	N.S.	N.S.	0.1006	N.S.	-51.74	32.23	-3.8

	Ele	ectrode	Rejected		GLM Weigh	nts (z)		MN	I Coordin	ates
Patient	Number	<u>GLM</u> Designation	by Jittering Analysis?	Trial	<b>Perception</b>	<u>Planning</u>	<b>Production</b>	X	Y	Z
	91		YES	N.S.	N.S.	0.14279	N.S.	-54.03	29.12	-0.8
	92		YES	N.S.	N.S.	0.15158	N.S.	-55.87	25.83	2.3
	93	Perception- Planning	NO	N.S.	0.227715842	0.25491	N.S.	-57.39	22.42	5.0
	94	Planning- Production	NO	N.S.	-0.256299395	0.37672	0.2444724	-58.79	18.96	9.1
	95	Planning	NO	N.S.	-0.305080447	0.41982	N.S.	-60.16	15.50	12.0
	96		YES	N.S.	N.S.	0.15559	N.S.	-46.86	34.61	-14.5
	120		YES	N.S.	N.S.	0.29606	N.S.	-49.71	31.61	-11.2
	121		YES	N.S.	N.S.	0.23189	N.S.	-52.62	28.65	-8.1
	122		YES	N.S.	-0.248406893	N.S.	N.S.	-55.23	25.62	-5.1
	123		YES	0.24249816	-0.165329408	0.12059	N.S.	-57.19	22.38	-1.8
	124		YES	N.S.	-0.113972627	N.S.	N.S.	-58.84	19.04	1.5
	125	Planning	NO	0.387885157	-0.618726533	0.45789	N.S.	-60.36	15.65	5.1
	126		YES	N.S.	-0.208991946	0.10942	0.1889376	-61.82	12.24	8.3
	97		YES	N.S.	N.S.	N.S.	N.S.	-29.12	6.85	-43.1
	98	Excluded - Seizure focus	N/A					-38.66	12.39	-40.0
	99	Excluded - Seizure focus	N/A					-46.81	16.32	-33.9
	100	Excluded - Seizure focus	N/A					-49.74	15.74	-24.4
	101	Excluded - Seizure focus	N/A					-53.86	16.55	-15.1
	102		YES	N.S.	N.S.	N.S.	N.S.	-58.09	18.08	-5.8
	103		YES	N.S.	N.S.	N.S.	0.1354987	-31.64	1.33	-42.8
	104		YES	N.S.	N.S.	N.S.	0.236212	-42.09	3.81	-43.1

	Ele	ectrode	Rejected		GLM Weig	hts (z)		MN	II Coordin	ates
Patient	Number	<u>GLM</u> Designation	by Jittering Analysis?	<u>Trial</u>	Perception	<u>Planning</u>	Production	X	Y	Z
	105	Excluded - Seizure focus	N/A					-49.87	6.63	-35.
	106	Excluded - Seizure focus	N/A					-54.41	7.31	-26.
	107		YES	N.S.	N.S.	N.S.	N.S.	-58.75	7.74	-16.
	108		YES	N.S.	N.S.	N.S.	N.S.	-31.92	-7.05	-41.
	109	Excluded - Seizure focus	N/A					-42.07	-6.25	-40.
	110		YES	N.S.	N.S.	N.S.	N.S.	-51.18	-4.11	-35.
	111		YES	N.S.	N.S.	N.S.	N.S.	-58.01	-2.19	-27.
	112		YES	N.S.	N.S.	N.S.	0.2760766	-32.36	-10.69	-40.
	113		YES	N.S.	N.S.	N.S.	N.S.	-43.11	-12.49	-38.
	114		YES	N.S.	N.S.	N.S.	N.S.	-53.23	-13.15	-34.
	115	Excluded - Seizure focus	N/A					-61.75	-12.85	-28.
	116	Excluded - Seizure focus	N/A					-30.91	-22.23	-23.
	117	Excluded - Seizure focus	N/A					-41.94	-21.60	-24.
	118	Excluded - Seizure focus	N/A					-52.34	-23.32	-26.
	119	Excluded - Seizure focus	N/A					-62.22	-22.27	-25.
	129		YES	N.S.	N.S.	N.S.	N.S.	-69.23	-16.49	-17.
	130		YES	N.S.	N.S.	0.1037	N.S.	-68.45	-13.96	-13.

	Ele	ectrode	Rejected	ected GLM Weights (z) Pring Pysis? <u>Trial Perception Planning Producti</u>					I Coordin	ates
Patient	<u>Number</u>	<u>GLM</u> Designation	by Jittering Analysis?	Trial	Perception	<u>Planning</u>	Production	X	¥	Z
	131		YES	N.S.	N.S.	N.S.	N.S.	-67.65	-11.41	-9.4
	132		YES	N.S.	N.S.	N.S.	N.S.	-66.83	-8.83	-5.1
	133		YES	N.S.	N.S.	N.S.	N.S.	-66.01	-6.20	-0.6
	134	Perception	NO	N.S.	0.392321059	N.S.	N.S.	-65.24	-3.48	4.3
	135		YES	N.S.	0.205796258	N.S.	N.S.	-64.57	-0.73	9.4
	136		YES	N.S.	N.S.	N.S.	0.1983028	-64.02	2.02	14.7
	137		YES	N.S.	N.S.	N.S.	N.S.	-70.55	-20.94	-15.5
	138		YES	N.S.	-0.123781033	0.20156	N.S.	-69.73	-18.41	-11.2
	139		YES	N.S.	N.S.	0.25577	N.S.	-68.88	-15.84	-7.0
	140		YES	N.S.	N.S.	N.S.	N.S.	-67.99	-13.23	-2.7
	141		YES	N.S.	0.213025785	N.S.	N.S.	-67.07	-10.58	1.0
	142		YES	N.S.	0.437730403	0.24572	0.4487851	-66.12	-7.90	6.3
	143	Perception	NO	N.S.	0.522819238	N.S.	N.S.	-65.17	-5.20	11.4
	144	Production	NO	N.S.	N.S.	N.S.	0.5868092	-64.36	-2.50	16.5
	145		YES	N.S.	N.S.	N.S.	N.S.	-71.90	-25.41	-13.1
	146		YES	N.S.	N.S.	0.11675	N.S.	-71.05	-22.86	-8.8
	147		YES	N.S.	N.S.	0.24004	N.S.	-70.17	-20.28	-4.5

	Ele	ectrode	Rejected			MN	I Coordin	ates		
Patient	<u>Number</u>	<u>GLM</u> Designation	by Jittering Analysis?	Trial	Perception	<u>Planning</u>	Production	X	Y	Z
	148		YES	N.S.	N.S.	0.1731	N.S.	-69.25	-17.64	-0.3
	149		YES	N.S.	0.349182075	N.S.	0.1443428	-68.27	-14.97	4.0
	150		YES	N.S.	0.312134873	N.S.	N.S.	-67.24	-12.31	8.0
	151	Perception	NO	N.S.	0.608022411	N.S.	N.S.	-66.13	-9.66	13.4
	152	Production	NO	N.S.	-0.153422794	N.S.	0.4770204	-65.09	-7.02	18.4
	153		YES	0.137525003	-0.125487377	0.09863	N.S.	-73.25	-29.88	-10.7
	154		YES	N.S.	N.S.	0.10122	N.S.	-72.39	-27.33	-6.3
	155		YES	N.S.	-0.107251031	0.14241	N.S.	-71.50	-24.74	-2.0
	156		YES	N.S.	N.S.	N.S.	N.S.	-70.59	-22.09	2.2
	157		YES	N.S.	N.S.	N.S.	N.S.	-69.63	-19.41	6.5
	158		YES	N.S.	N.S.	N.S.	N.S.	-68.66	-16.75	11.(
	159		YES	N.S.	N.S.	N.S.	N.S.	-67.77	-14.09	15.3
	160		YES	N.S.	N.S.	N.S.	0.2152216	-67.08	-11.43	20.5
	161		YES	N.S.	N.S.	N.S.	N.S.	-68.74	-42.35	-10.9
	162		YES	N.S.	N.S.	N.S.	N.S.	-70.69	-40.01	-6.4
	163		YES	N.S.	N.S.	N.S.	N.S.	-72.15	-37.50	-1.9
	164		YES	N.S.	N.S.	N.S.	N.S.	-72.10	-34.83	2.0
	Ele	ectrode	Rejected			MN	I Coordin	ates		
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Patient	<u>Number</u>	<u>GLM</u> Designation	by Jittering Analysis?	Trial	Perception	<u>Planning</u>	Production	X	Y	Z
	165		YES	N.S.	N.S.	N.S.	N.S.	-71.72	-32.09	6.3
	166		YES	N.S.	0.149587084	N.S.	N.S.	-71.12	-29.25	10.7
	167		YES	N.S.	N.S.	N.S.	N.S.	-70.32	-26.28	15.1
	168		YES	N.S.	0.141415445	N.S.	N.S.	-69.69	-23.33	19.5
	169		YES	N.S.	N.S.	N.S.	N.S.	-67.28	-46.72	-8.2
	170		YES	N.S.	N.S.	N.S.	N.S.	-68.97	-44.42	-3.7
	171		YES	N.S.	-0.17421237	0.28608	N.S.	-70.15	-41.94	0.1
	172		YES	N.S.	N.S.	N.S.	N.S.	-70.35	-39.32	4.9
	173		YES	N.S.	N.S.	N.S.	N.S.	-70.00	-36.60	9.3
	174		YES	N.S.	N.S.	N.S.	N.S.	-69.39	-33.78	13.7
	175	Excluded - Seizure focus	N/A					-68.62	-30.85	18.1
	176		YES	N.S.	0.115340733	N.S.	N.S.	-68.07	-27.93	22.0
	177		YES	N.S.	N.S.	N.S.	N.S.	-65.60	-50.91	-5.5
	178		YES	N.S.	N.S.	0.11665	N.S.	-67.01	-48.67	-1.0
	179	Excluded - Seizure focus	N/A					-67.98	-46.25	3.4
	180	Excluded - Seizure focus	N/A					-68.34	-43.71	7.8
	181		YES	N.S.	N.S.	N.S.	N.S.	-68.20	-41.05	12.3

	Ele	ectrode	Rejected		GLM Weigl		MN	II Coordin	ates	
Patient	Number	<u>GLM</u> Designation	by Jittering Analysis?	<u>Trial</u>	Perception	<u>Planning</u>	Production	X	Y	Z
	182	Excluded - Seizure focus	N/A					-67.73	-38.28	16.
	183	Excluded - Seizure focus	N/A					-67.11	-35.42	21.
	184	Excluded - Seizure focus	N/A					-66.65	-32.53	25.
	185	Excluded - Seizure focus	N/A					-63.55	-54.82	-2.
	186	Excluded - Seizure focus	N/A					-64.79	-52.67	1.
	187		YES	N.S.	N.S.	N.S.	N.S.	-65.70	-50.37	6.
	188		YES	N.S.	N.S.	N.S.	N.S.	-66.23	-47.94	10.
	189		YES	N.S.	N.S.	N.S.	N.S.	-66.38	-45.38	15.
	190	Excluded - Seizure focus	N/A					-66.26	-42.71	19.
	191	Excluded - Seizure focus	N/A					-65.94	-39.93	24.
	192	Excluded - Seizure focus	N/A					-65.59	-37.08	29.
	193	Excluded - Seizure focus	N/A					-60.97	-58.38	-0.
	194	Excluded - Seizure focus	N/A					-62.24	-56.39	4.
	195	Excluded - Seizure focus	N/A					-63.25	-54.26	9.
	196	Excluded - Seizure focus	N/A					-64.01	-52.00	13.
	197		YES	-0.155973803	N.S.	N.S.	N.S.	-64.53	-49.57	18.

	Ele	ectrode	Rejected		GLM Weigh	ts (z)		MN	I Coordin	ates
Patient	<u>Number</u>	<u>GLM</u> Designation	by Jittering Analysis?	Trial	<b>Perception</b>	<u>Planning</u>	Production	X	Y	Z
	198		YES	N.S.	N.S.	N.S.	N.S.	-64.87	-46.99	22.9
	199		YES	N.S.	N.S.	N.S.	N.S.	-65.13	-44.31	27.0
	200		YES	N.S.	N.S.	N.S.	-0.147215	-65.27	-41.53	32.2
	201		YES	N.S.	N.S.	N.S.	N.S.	-58.10	-61.72	2.5
	202		YES	-0.158471267	0.095273408	N.S.	N.S.	-59.43	-59.89	7.2
	203		YES	N.S.	N.S.	N.S.	N.S.	-60.62	-57.98	11.9
	204		YES	N.S.	N.S.	N.S.	N.S.	-61.63	-55.90	16.0
	205		YES	N.S.	N.S.	N.S.	N.S.	-62.46	-53.59	21.2
	206		YES	N.S.	N.S.	N.S.	N.S.	-63.19	-51.09	25.9
	207		YES	N.S.	-0.110612424	N.S.	N.S.	-63.92	-48.46	30.0
	208		YES	N.S.	-0.136739472	N.S.	N.S.	-64.55	-45.75	35.3
	209		YES	N.S.	N.S.	N.S.	N.S.	-55.10	-64.94	5.1
	210		YES	N.S.	N.S.	N.S.	N.S.	-56.50	-63.27	9.9
	211		YES	N.S.	N.S.	N.S.	N.S.	-57.87	-61.58	14.7
	212		YES	N.S.	N.S.	N.S.	N.S.	-59.08	-59.67	19.4
	213		YES	N.S.	N.S.	N.S.	-0.141312	-60.08	-57.44	24.2
	214		YES	N.S.	N.S.	N.S.	N.S.	-60.93	-54.95	28.8

	Ele	ectrode	Rejected		GLM Weigh	nts (z)		MN	I Coordin	ates
Patient	<u>Number</u>	<u>GLM</u> Designation	by Jittering Analysis?	<u>Trial</u>	Perception	<u>Planning</u>	Production	X	Y	Z
	215		YES	N.S.	N.S.	N.S.	N.S.	-61.67	-52.31	33.5
	216		YES	N.S.	-0.19565944	N.S.	N.S.	-62.19	-49.57	38.1
	217	Excluded - Seizure focus	N/A					-52.14	-68.18	7.8
	218	Excluded - Seizure focus	N/A					-53.61	-66.63	12.6
	219		YES	N.S.	N.S.	N.S.	N.S.	-55.08	-65.08	17.5
	220		YES	N.S.	N.S.	N.S.	N.S.	-56.40	-63.31	22.3
	221		YES	N.S.	N.S.	N.S.	N.S.	-57.44	-61.10	27.0
	222		YES	N.S.	N.S.	N.S.	N.S.	-58.27	-58.61	31.7
	223		YES	N.S.	N.S.	N.S.	N.S.	-58.88	-55.96	36.3
	224		YES	-0.08661985	0.06167594	0.05387	0.0843445	-59.28	-53.21	41.0
	225		YES	N.S.	N.S.	N.S.	-0.191236	-50.33	-46.07	57.(
	226		YES	N.S.	N.S.	N.S.	N.S.	-52.69	-36.18	54.(
	227		YES	N.S.	N.S.	N.S.	N.S.	-54.98	-25.64	50.3
	228		YES	N.S.	-0.112604223	N.S.	N.S.	-57.68	-15.58	46.0
	229	Production	NO	N.S.	-0.207951847	N.S.	0.2739099	-59.38	-6.01	42.5

	Ele	ectrode	Rejected	ted GLM Weights (z) M						ates
Patient	<u>Number</u>	<u>GLM</u> Designation	by Jittering Analysis?	<u>Trial</u>	Perception	<u>Planning</u>	Production	X	Y	Z
	230		YES	N.S.	N.S.	N.S.	N.S.	-58.31	3.13	37.3
	231		YES	0.193135157	N.S.	N.S.	N.S.	-56.09	12.02	32.7
	232		YES	N.S.	-0.351506755	0.11044	-0.218381	-53.57	20.66	28.2
	233		YES	N.S.	N.S.	N.S.	N.S.	-44.47	-43.79	63.2
	234		YES	N.S.	N.S.	N.S.	N.S.	-47.06	-33.80	61.4
	235		YES	N.S.	N.S.	N.S.	N.S.	-49.38	-22.82	58.9
	236		YES	N.S.	-0.138208913	-0.14485	N.S.	-51.42	-12.45	55.4
	237		YES	N.S.	-0.147250548	-0.20659	-0.17116	-52.42	-3.02	51.0
	238		YES	N.S.	-0.183942603	N.S.	N.S.	-51.81	5.96	45.9
	239		YES	N.S.	N.S.	N.S.	N.S.	-50.44	14.71	40.8
	240		YES	N.S.	-0.155540289	N.S.	-0.135277	-49.11	22.98	36.3
	241		YES	N.S.	N.S.	N.S.	N.S.	-37.65	-40.41	68.8
	242		YES	N.S.	N.S.	N.S.	N.S.	-40.63	-30.74	68.0
	243		YES	0.131789619	N.S.	N.S.	N.S.	-43.06	-19.27	67.6
	244	Production	NO	0.19358143	-0.140244608	-0.11378	0.4502113	-44.36	-9.02	64.2
	245		YES	0.176286436	-0.169108276	0.14281	N.S.	-44.59	-0.20	59.4
	246		YES	N.S.	N.S.	0.18695	N.S.	-44.11	8.41	54.2

	Ele	ectrode	Rejected		GLM Weigh	nts (z)		MN	I Coordin	ates
Patient	Number	<u>GLM</u> Designation	by Jittering Analysis?	Trial	Perception	<u>Planning</u>	Production	X	¥	Z
	247		YES	N.S.	N.S.	0.14006	N.S.	-43.25	17.26	49.
	248	Planning	NO	N.S.	-0.342254315	0.20873	N.S.	-42.04	26.55	44.
	249		YES	N.S.	N.S.	N.S.	N.S.	-28.41	-33.81	74.
	250		YES	N.S.	N.S.	N.S.	N.S.	-32.78	-26.62	75.
	251		YES	N.S.	N.S.	N.S.	N.S.	-35.22	-15.26	75.
	252		YES	0.173277009	N.S.	N.S.	-0.171999	-36.15	-5.91	72.4
	253		YES	N.S.	0.347646325	0.18108	N.S.	-36.19	2.35	67.:
	254		YES	0.249849017	N.S.	0.1504	N.S.	-35.83	10.58	62.
	255		YES	N.S.	N.S.	N.S.	N.S.	-35.32	18.95	57.
	256		YES	N.S.	-0.211392421	0.18779	N.S.	-34.49	27.22	53.
	1	Production	NO	0.250784621	-0.185148907	N.S.	0.5861435	-32.02	-10.14	-47.
	2		YES	N.S.	N.S.	N.S.	N.S.	-41.78	-15.03	-45.
	3		YES	N.S.	N.S.	N.S.	N.S.	-51.70	-17.99	-39.
477L Chronic	4	Production	NO	0.192196317	-0.213378598	N.S.	0.5117164	-59.31	-21.42	-34.
	9		NO	0.484883826	-0.629336107	N.S.	N.S.	0.14	-72.02	-7.
	10		NO	0.390912908	-0.50682255	N.S.	N.S.	-5.87	-76.21	-19.
	11		NO	0.420299074	-0.560083719	N.S.	N.S.	-14.56	-81.86	-24.

	Ele	ectrode	Rejected			MN	I Coordin	ates		
Patient	<u>Number</u>	<u>GLM</u> Designation	by Jittering Analysis?	Trial	<b>Perception</b>	<u>Planning</u>	<b>Production</b>	X	Y	Z
	12		YES	0.299435208	-0.191865191	0.12348	-0.253442	-23.37	-87.97	-24.6
	13		YES	N.S.	N.S.	N.S.	N.S.	-26.18	-41.64	-22.5
	14		YES	N.S.	N.S.	N.S.	-0.322284	-36.57	-38.55	-27.9
	15	Perception	NO	N.S.	0.358448236	N.S.	-0.178556	-46.03	-37.28	-31.4
	16		YES	N.S.	N.S.	N.S.	N.S.	-54.89	-35.82	-32.5
	41		YES	N.S.	N.S.	N.S.	N.S.	-20.61	-55.67	-17.2
	42		YES	N.S.	-0.368949606	-0.11748	N.S.	-30.92	-58.86	-21.3
	43		YES	N.S.	-0.309745509	N.S.	N.S.	-40.07	-63.79	-22.0
	44		YES	0.556917684	-0.506005045	N.S.	-0.505072	-48.97	-66.83	-19.8
	97		YES	N.S.	N.S.	N.S.	N.S.	-42.32	47.25	21.2
	98		YES	N.S.	N.S.	0.2458	N.S.	-47.03	37.42	26.5
	99	Planning- Production	NO	N.S.	N.S.	0.51527	0.2413405	-51.23	27.57	31.2
	100	Planning	NO	N.S.	-0.264541483	0.29671	N.S.	-53.11	17.39	34.(
	101	Planning	NO	N.S.	-0.192696698	0.46695	N.S.	-54.22	7.33	36.5
	102	Planning- Production	NO	0.430315217	N.S.	0.37904	0.4173668	-55.74	-2.37	40.5
	103	Planning	NO	0.634872202	N.S.	0.33053	N.S.	-57.40	-11.86	44.3
	104		YES	N.S.	N.S.	N.S.	N.S.	-59.59	-21.01	46.0
	105	Planning- Production	NO	0.326693518	-0.311377292	0.20608	0.4306377	-60.84	-30.87	47.9

	Ele	ectrode	Rejected		GLM Weigh	GLM Weights (z) Perception Planning Produ		MN	I Coordin	ates
Patient	Number	<u>GLM</u> Designation	by Jittering Analysis?	Trial	Perception	<u>Planning</u>	Production	X	<u>¥</u>	Z
	106	Planning- Production	NO	N.S.	N.S.	0.26215	0.3723763	-59.30	-42.66	48.5
	129		YES	0.362128673	-0.158611686	N.S.	-0.174346	-21.13	-108.8 7	9.1
	130		NO	N.S.	N.S.	N.S.	-0.50015	-30.68	-102.3 9	6.9
	131	Excluded - Seizure focus	N/A					-39.42	-95.63	4.2
	132	Perception	NO	0.282296045	0.20794382	N.S.	-0.375022	-46.89	-88.46	2.3
	133		YES	N.S.	N.S.	N.S.	N.S.	-53.38	-80.35	0.0
	134	Perception	NO	N.S.	0.580043992	N.S.	N.S.	-58.67	-71.14	-0.5
	135		YES	N.S.	0.207679782	0.20393	N.S.	-62.79	-60.97	-1.9
	136		YES	N.S.	N.S.	N.S.	N.S.	-66.33	-50.12	-3.4
	137		YES	0.398540472	-0.22023389	N.S.	-0.2324	-21.57	-106.5 3	16.0
	138	Perception	NO	-0.289678029	0.601936213	N.S.	N.S.	-31.35	-100.0 6	15.0
	139		YES	0.412649848	-0.343000963	N.S.	-0.346572	-40.13	-93.07	14.8
	140		YES	0.231084149	-0.309147965	N.S.	-0.262653	-47.62	-86.27	13.9
	141		YES	N.S.	N.S.	-0.1891	N.S.	-54.39	-78.70	12.8
	142	Perception	NO	-0.270866836	0.412288465	N.S.	N.S.	-60.31	-69.91	11.9
	143		YES	-0.167684183	0.206101397	N.S.	N.S.	-64.85	-59.90	10.7

	Ele	ectrode	Rejected			MN	I Coordin	ates		
Patient	<u>Number</u>	<u>GLM</u> Designation	by Jittering Analysis?	Trial	<b>Perception</b>	<u>Planning</u>	Production	X	Y	Z
	144	Perception	NO	0.221536403	0.393187376	N.S.	N.S.	-67.92	-49.17	8.9
	145		YES	N.S.	N.S.	-0.02561	N.S.	-20.61	-103.4 4	23.3
	146		YES	N.S.	-0.121312545	-0.09938	N.S.	-30.57	-97.43	25.1
	147		YES	0.225848462	-0.222320852	N.S.	N.S.	-39.12	-90.92	26.4
	148		YES	N.S.	N.S.	N.S.	-0.243976	-46.84	-84.18	26.4
	149		YES	N.S.	N.S.	0.26223	N.S.	-53.93	-76.89	25.7
	150		YES	N.S.	N.S.	N.S.	0.1765627	-60.41	-68.47	24.7
	151		YES	N.S.	0.130518643	N.S.	N.S.	-65.76	-58.70	23.6
	152		YES	N.S.	N.S.	N.S.	0.2066385	-68.21	-47.98	21.4
	153		NO	0.422759481	-0.665496903	N.S.	N.S.	-17.32	-98.69	33.5
	154	Excluded - Seizure focus	N/A					-26.81	-94.66	36.8
	155		YES	0.155334661	N.S.	N.S.	N.S.	-36.15	-88.63	38.5
	156		YES	N.S.	0.145119769	N.S.	N.S.	-44.49	-81.89	39.(
	157		YES	N.S.	0.155112344	N.S.	N.S.	-51.74	-74.71	38.1
	158		YES	N.S.	N.S.	N.S.	N.S.	-58.01	-66.51	36.8
	159		YES	N.S.	N.S.	N.S.	0.1934886	-62.99	-56.72	35.3
	160	Production	NO	-0.211418842	N.S.	-0.1671	0.205268	-66.12	-46.18	33.2

	Ele	ectrode	Rejected			MN	I Coordin	ates		
Patient	Number	<u>GLM</u> Designation	by Jittering Analysis?	<u>Trial</u>	<b>Perception</b>	<u>Planning</u>	<b>Production</b>	X	¥	Z
	161	Excluded - Seizure focus	N/A					-33.02	-81.57	45.3
	162	Excluded - Seizure focus	N/A					-37.64	-73.08	51.4
	163		YES	N.S.	N.S.	N.S.	N.S.	-42.29	-63.68	56.0
	164		YES	N.S.	N.S.	0.18591	N.S.	-45.72	-53.02	60.4
	165		YES	N.S.	N.S.	0.15576	0.2083636	-47.76	-42.62	61.0
	166		YES	-0.286641893	N.S.	0.17842	0.4369122	-49.54	-31.27	60.1
	167		YES	N.S.	N.S.	0.1214	0.1490372	-50.52	-20.40	56.9
	168	Production	NO	0.326546674	N.S.	N.S.	0.8042652	-50.98	-10.23	53.0
	169	Excluded - Seizure focus	N/A					-26.45	-79.69	55.
	170		YES	N.S.	N.S.	0.20875	0.2170109	-30.24	-71.46	59.0
	171		YES	N.S.	N.S.	N.S.	0.1899193	-33.99	-61.77	63.′
	172		YES	N.S.	N.S.	N.S.	N.S.	-38.00	-50.25	67.9
	173		YES	-0.173433129	N.S.	0.1086	0.436405	-40.28	-39.73	69.0
	174		YES	-0.157188337	N.S.	N.S.	0.3799243	-42.02	-28.66	67.4
	175		YES	N.S.	N.S.	0.20203	0.2065716	-42.71	-18.48	63.8
	176		YES	N.S.	N.S.	N.S.	0.3066738	-43.24	-8.53	60.3
	177	Excluded - Seizure focus	N/A					-18.81	-77.69	55.3

	Ele	ectrode	Rejected		GLM Weigh		MN	I Coordin	ates	
Patient	<u>Number</u>	<u>GLM</u> Designation	by Jittering Analysis?	<u>Trial</u>	<b>Perception</b>	<u>Planning</u>	<b>Production</b>	X	Y	Z
	178	Excluded - Seizure focus	N/A					-22.04	-70.06	61.0
	179	Production	NO	N.S.	N.S.	N.S.	0.4920054	-23.98	-61.01	67.1
	180		YES	N.S.	N.S.	N.S.	0.2789078	-27.81	-50.11	72.4
	181		YES	N.S.	N.S.	N.S.	N.S.	-30.65	-38.93	74.2
	182	Production	NO	N.S.	N.S.	N.S.	0.5686038	-32.37	-27.70	72.7
	183	Planning- Production	NO	N.S.	N.S.	0.24375	0.5663252	-33.39	-17.54	69.3
	184	Planning- Production	NO	N.S.	N.S.	0.14376	0.3828541	-33.89	-8.42	66.8
	185	Excluded - Seizure focus	N/A					-10.09	-75.08	53.1
	186	Excluded - Seizure focus	N/A					-12.75	-67.51	60.4
	187		YES	N.S.	N.S.	-0.13232	N.S.	-15.00	-59.16	68.2
	188		YES	N.S.	N.S.	N.S.	N.S.	-17.87	-49.43	74.5
	189		YES	N.S.	0.190913917	N.S.	N.S.	-20.41	-38.64	77.4
	190		YES	N.S.	N.S.	N.S.	0.2756657	-22.30	-26.70	76.3
	191	Planning- Production	NO	N.S.	N.S.	0.15417	0.5477178	-23.67	-15.68	74.1
	192		YES	N.S.	0.116525491	N.S.	N.S.	-24.62	-5.14	71.6
	193		YES	N.S.	N.S.	N.S.	0.2251305	-48.52	49.54	-3.0
	194	Planning- Production	NO	N.S.	-0.356131656	0.37344	0.3076047	-50.90	40.27	1.1
	195		YES	N.S.	N.S.	N.S.	N.S.	-53.53	31.02	5.2
	196		YES	N.S.	N.S.	-0.13817	N.S.	-56.46	21.53	9.1

	Ele	ectrode	Rejected		GLM Weigh	nts (z)		MN	I Coordin	ates
Patient	Number	<u>GLM</u> Designation	by Jittering Analysis?	Trial	Perception	<u>Planning</u>	Production	X	Y	Z
	197	Perception- Planning	NO	0.237520638	0.244929423	0.18556	N.S.	-59.47	11.93	13.0
	198	Production	NO	N.S.	-0.249380075	N.S.	0.3943139	-62.36	2.23	17.1
	199	Planning- Production	NO	0.292747754	-0.350743936	0.19146	0.5363179	-64.69	-7.37	21.:
	200		YES	N.S.	N.S.	N.S.	N.S.	-66.27	-16.63	26.0
	201		YES	-0.276541896	N.S.	N.S.	0.2300405	-66.82	-26.87	31.8
	202	Production	NO	N.S.	-0.141298012	-0.22988	0.1789325	-65.46	-39.88	36.3
	203		YES	N.S.	N.S.	0.10268	0.172734	-48.44	18.70	-21.0
	204		YES	N.S.	0.322106128	N.S.	N.S.	-53.68	10.87	-15.3
	205	Excluded - Seizure focus	N/A					-58.36	3.56	-9.0
	206	Excluded - Seizure focus	N/A					-61.41	-3.31	-3.'
	207	Excluded - Seizure focus	N/A					-66.37	-11.93	3.
	208	Perception- Planning	NO	0.167387669	0.914867258	N.S.	0.2850466	-68.42	-20.47	7.9
	209		YES	N.S.	0.326878501	N.S.	N.S.	-69.04	-29.70	10.9
	210	Perception- Planning	NO	N.S.	0.235113453	N.S.	0.4693901	-69.25	-40.69	12.4
	211		YES	N.S.	N.S.	-0.25645	N.S.	-66.25	-11.92	-19.7
	212		YES	N.S.	N.S.	N.S.	N.S.	-66.72	-19.40	-15.7
	213		YES	N.S.	N.S.	0.16958	N.S.	-70.22	-28.22	-9.9

Patient	Electrode		Rejected	GLM Weights (z)				MNI Coordinates		
	<u>Number</u>	<u>GLM</u> Designation	Jittering Analysis?	<u>Trial</u>	Perception	<u>Planning</u>	Production	X	¥	Z
	214		YES	N.S.	N.S.	-0.17149	N.S.	-70.23	-37.80	-3.9

#### **Extended Data Table 3b**

DKT parcellation of all CI task-responsive electrodes

DKT Label	# Perception Sites	# Production Sites	# Planning Sites	All <sup>**</sup>
Left superior frontal gyrus	0	0	1	11
Left rostral middle frontal gyrus $^*$	0	3	20	91
Left caudal middle frontal gyrus $^*$	0	5	19	37
Left pars opercularis*	4	14	43	109
Left pars triangularis *	3	6	27	89
Left pars orbitalis	0	1	1	14
Left precentral gyrus *	1	53	29	128
Left postcentral gyrus	1	16	3	46
Left superior parietal lobule	0	1	0	12
Left supramarginal gyrus	0	4	2	17
Left Inferior parietal lobule	1	0	0	20
Left superior temporal gyrus *	11	5	14	103
Left middle temporal gyrus	0	6	1	53
Left Inferior temporal gyrus	2	2	0	12
Left transverse temporal gyrus	9	5	0	12
Left lateral occipital cortex	2	0	0	11
Left insular cortex	2	0	0	4
Total	36	121	160	769

Regions which combine to include 152 of 160 (95.0%) of all planning electrodes (inferior and middle frontal gyri, superior temporal gyrus, and precentral gyrus).

\*\* Total does not include electrodes which are located outside of the anatomical structures listed in this table.

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# REFERENCES

- Sacks H, Schegloff EA & Jefferson G A Simplest Systematics for the Organization of Turn-Taking for Conversation. Language 50, 696–735 (1974).
- Levinson SC & Torreira F Timing in turn-taking and its implications for processing models of language. Front Psychol 6, 731 (2015). [PubMed: 26124727]
- 3. Heldner M & Edlund J Pauses, gaps and overlaps in conversations. Journal of Phonetics 38 (2010).
- 4. Stivers T et al. Universals and cultural variation in turn-taking in conversation. Proc Natl Acad Sci U S A 106, 10587–10592 (2009). [PubMed: 19553212]
- Schiffman HR Sensation and Perception. An Integrated Approach. (John Wiley and Sons, Inc., New York; 2001).
- 6. Bates E et al. Timed picture naming in seven languages. Psychon Bull Rev. 10, 344–380 (2003). [PubMed: 12921412]
- Flinker A et al. Redefining the role of Broca's area in speech. Proc Natl Acad Sci U S A 112, 2871–2875 (2015). [PubMed: 25730850]
- Bohland JW, Bullock D & Guenther FH Neural representations and mechanisms for the performance of simple speech sequences. J Cogn Neurosci 22, 1504–1529 (2010). [PubMed: 19583476]
- Basilakos A, Smith KG, Fillmore P, Fridriksson J & Fedorenko E Functional Characterization of the Human Speech Articulation Network. Cereb Cortex 28, 1816–1830 (2018). [PubMed: 28453613]
- Mirman D, Kraft AE, Harvey DY, Brecher AR & Schwartz MF Mapping articulatory and grammatical subcomponents of fluency deficits in post-stroke aphasia. Cogn Affect Behav Neurosci 19, 1286–1298 (2019). [PubMed: 31240565]
- 11. Papoutsi M et al. From phonemes to articulatory codes: an fMRI study of the role of Broca's area in speech production. Cereb Cortex 19, 2156–2165 (2009). [PubMed: 19181696]
- 12. Guenther FH Neural Control of Speech. (MIT Press, 2016).
- Sahin NT, Pinker S, Cash SS, Schomer D & Halgren E Sequential processing of lexical, grammatical, and phonological information within Broca's area. Science 326, 445–449 (2009). [PubMed: 19833971]
- Dronkers NF A new brain region for coordinating speech articulation. Nature 384, 159–161 (1996). [PubMed: 8906789]
- 15. Broca P Remarques sur le siege de la faculté du langage articulé, suivies d'une observation d'aphémie (perte de la parole). Bulletins et mémoires de la Société Anatomique de Paris 36, 330–356 (1861).
- Trupe LA et al. Chronic apraxia of speech and Broca's area. Stroke 44, 740–744 (2013). [PubMed: 23362082]
- Chang EF et al. Pure Apraxia of Speech After Resection Based in the Posterior Middle Frontal Gyrus. Neurosurgery 87, E383–E389 (2020). [PubMed: 32097489]
- Buchsbaum BR, Greer S, Chang WL & Berman KF Meta-analysis of neuroimaging studies of the Wisconsin card-sorting task and component processes. Hum Brain Mapp 25, 35–45 (2005). [PubMed: 15846821]
- Brass M & von Cramon DY The role of the frontal cortex in task preparation. Cereb Cortex 12, 908–914 (2002). [PubMed: 12183390]
- 20. Sierpowska J et al. Involvement of the middle frontal gyrus in language switching as revealed by electrical stimulation mapping and functional magnetic resonance imaging in bilingual brain tumor patients. Cortex 99, 78–92 (2018). [PubMed: 29197227]
- Levinson SC Turn-taking in Human Communication--Origins and Implications for Language Processing. Trends Cogn Sci 20, 6–14 (2016). [PubMed: 26651245]
- 22. Indefrey P The spatial and temporal signatures of word production components: a critical update. Front Psychol 2, 255 (2011). [PubMed: 22016740]
- Levelt WJ, Roelofs A & Meyer AS A theory of lexical access in speech production. Behav Brain Sci 22, 1–38; discussion 38–75 (1999). [PubMed: 11301520]

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- 24. Fuchs S, Petrone C, Krivokapi J & Hoole P Acoustic and respiratory evidence for utterance planning in German. . Journal of Phonetics 41, 29–47 (2013).
- Redford MA Unifying Speech and Language in a Developmentally Sensitive Model of Production. J Phon 53, 141–152 (2015). [PubMed: 26688597]
- 26. Turk A & Shattuck-Hufnagel S Timing Evidence for Symbolic Phonological Representations and Phonology-Extrinsic Timing in Speech Production. Front Psychol 10, 2952 (2019). [PubMed: 32038364]
- 27. Schuhmann T, Schiller NO, Goebel R & Sack AT The temporal characteristics of functional activation in Broca's area during overt picture naming. Cortex 45, 1111–1116 (2009). [PubMed: 19111289]
- Alario FX, Chainay H, Lehericy S & Cohen L The role of the supplementary motor area (SMA) in word production. Brain Res 1076, 129–143 (2006). [PubMed: 16480694]
- 29. Giahi Saravani A, Forseth KJ, Tandon N & Pitkow X Dynamic Brain Interactions during Picture Naming. eNeuro 6 (2019).
- 30. Ferpozzi V et al. Broca's Area as a Pre-articulatory Phonetic Encoder: Gating the Motor Program. Front Hum Neurosci 12, 64 (2018). [PubMed: 29520225]
- 31. Tilsen S et al. Anticipatory Posturing of the Vocal Tract Reveals Dissociation of Speech Movement Plans from Linguistic Units. PLoS One 11, e0146813 (2016). [PubMed: 26760511]
- Ramanarayanan V, Goldstein L, Byrd D & Narayanan SS An investigation of articulatory setting using real-time magnetic resonance imaging. J Acoust Soc Am 134, 510–519 (2013). [PubMed: 23862826]
- 33. Bogels S, Magyari L & Levinson SC Neural signatures of response planning occur midway through an incoming question in conversation. Sci Rep 5, 12881 (2015). [PubMed: 26242909]
- Ferreira F & Swets B How Incremental Is Language Production? Evidence from the Production of Utterances Requiring the Computation of Arithmetic Sums. Journal of Memory and Language 46, 57–84 (2002).
- 35. Wagner V, Jescheniak JD & Schriefers H On the flexibility of grammatical advance planning during sentence production: Effects of cognitive load on multiple lexical access. J Exp Psychol Learn Mem Cogn 36, 423–440 (2010). [PubMed: 20192540]
- Bogels S, Casillas M & Levinson SC Planning versus comprehension in turn-taking: Fast responders show reduced anticipatory processing of the question. Neuropsychologia 109, 295–310 (2018). [PubMed: 29269305]
- Dubey A & Ray S Cortical Electrocorticogram (ECoG) Is a Local Signal. J Neurosci 39, 4299– 4311 (2019). [PubMed: 30914446]
- Buzsaki G, Anastassiou CA & Koch C The origin of extracellular fields and currents--EEG, ECoG, LFP and spikes. Nat Rev Neurosci 13, 407–420 (2012). [PubMed: 22595786]
- 39. Cheung C, Hamiton LS, Johnson K & Chang EF The auditory representation of speech sounds in human motor cortex. Elife 5 (2016).
- Wilson SM, Saygin AP, Sereno MI & Iacoboni M Listening to speech activates motor areas involved in speech production. Nat Neurosci 7, 701–702 (2004). [PubMed: 15184903]
- 41. Glanz Iljina O et al. Real-life speech production and perception have a shared premotor-cortical substrate. Sci Rep 8, 8898 (2018). [PubMed: 29891885]
- 42. Cisek P & Kalaska JF Neural mechanisms for interacting with a world full of action choices. Annu Rev Neurosci 33, 269–298 (2010). [PubMed: 20345247]
- 43. Ray S & Maunsell JH Different origins of gamma rhythm and high-gamma activity in macaque visual cortex. PLoS Biol 9, e1000610 (2011). [PubMed: 21532743]
- 44. Leszczynski M et al. Dissociation of broadband high-frequency activity and neuronal firing in the neocortex. Sci Adv 6, eabb0977 (2020). [PubMed: 32851172]
- 45. Mukamel R et al. Invariance of firing rate and field potential dynamics to stimulus modulation rate in human auditory cortex. Hum Brain Mapp 32, 1181–1193 (2011). [PubMed: 20665720]
- 46. Steinschneider M, Volkov IO, Noh MD, Garell PC & Howard MA 3rd Temporal encoding of the voice onset time phonetic parameter by field potentials recorded directly from human auditory cortex. J Neurophysiol 82, 2346–2357 (1999). [PubMed: 10561410]

- Mesgarani N, Cheung C, Johnson K & Chang EF Phonetic feature encoding in human superior temporal gyrus. Science 343, 1006–1010 (2014). [PubMed: 24482117]
- 48. Flinker A, Chang EF, Barbaro NM, Berger MS & Knight RT Sub-centimeter language organization in the human temporal lobe. Brain Lang 117, 103–109 (2011). [PubMed: 20961611]
- Hamilton LS, Edwards E & Chang EF A Spatial Map of Onset and Sustained Responses to Speech in the Human Superior Temporal Gyrus. Curr Biol 28, 1860–1871 e1864 (2018). [PubMed: 29861132]
- Bouchard KE, Mesgarani N, Johnson K & Chang EF Functional organization of human sensorimotor cortex for speech articulation. Nature 495, 327–332 (2013). [PubMed: 23426266]
- 51. Mugler EM et al. Differential Representation of Articulatory Gestures and Phonemes in Precentral and Inferior Frontal Gyri. J Neurosci 38, 9803–9813 (2018). [PubMed: 30257858]
- 52. Cogan GB et al. Sensory-motor transformations for speech occur bilaterally. Nature 507, 94–98 (2014). [PubMed: 24429520]
- Kotz SA et al. Lexicality drives audio-motor transformations in Broca's area. Brain Lang 112, 3–11 (2010). [PubMed: 19698980]
- 54. Gerardin E et al. Partially overlapping neural networks for real and imagined hand movements. Cereb Cortex 10, 1093–1104 (2000). [PubMed: 11053230]
- Fadiga L & Craighero L Hand actions and speech representation in Broca's area. Cortex 42, 486–490 (2006). [PubMed: 16881255]
- 56. Papitto G, Friederici AD & Zaccarella E The topographical organization of motor processing: An ALE meta-analysis on six action domains and the relevance of Broca's region. Neuroimage 206, 116321 (2020). [PubMed: 31678500]
- 57. Lange VM & Laganaro M Inter-subject variability modulates phonological advance planning in the production of adjective-noun phrases. Front Psychol 5, 1–12 (2014). [PubMed: 24474945]
- Torreira F, Bogels S & Levinson SC Breathing for answering: the time course of response planning in conversation. Front Psychol 6, 284 (2015). [PubMed: 25814976]
- Knudsen B, Creemers A & Meyer AS Forgotten Little Words: How Backchannels and Particles May Facilitate Speech Planning in Conversation? Front Psychol 11, 593671 (2020). [PubMed: 33240183]
- 60. Dichter BK, Breshears JD, Leonard MK & Chang EF The Control of Vocal Pitch in Human Laryngeal Motor Cortex. Cell 174, 21–31 e29 (2018). [PubMed: 29958109]
- Chang EF et al. Stereotactic probability and variability of speech arrest and anomia sites during stimulation mapping of the language dominant hemisphere. J Neurosurg 126, 114–121 (2017). [PubMed: 26894457]
- 62. Long MA et al. Functional Segregation of Cortical Regions Underlying Speech Timing and Articulation. Neuron 89, 1187–1193 (2016). [PubMed: 26924439]
- Klaus J & Hartwigsen G Dissociating semantic and phonological contributions of the left inferior frontal gyrus to language production. Hum Brain Mapp 40, 3279–3287 (2019). [PubMed: 30969004]
- Tate MC, Herbet G, Moritz-Gasser S, Tate JE & Duffau H Probabilistic map of critical functional regions of the human cerebral cortex: Broca's area revisited. Brain 137, 2773–2782 (2014). [PubMed: 24970097]
- Nottebohm F, Stokes TM & Leonard CM Central control of song in the canary, Serinus canarius. J Comp Neurol 165, 457–486 (1976). [PubMed: 1262540]
- 66. Long MA & Fee MS Using temperature to analyse temporal dynamics in the songbird motor pathway. Nature 456, 189–194 (2008). [PubMed: 19005546]
- Okobi DE Jr., Banerjee A, Matheson AMM, Phelps SM & Long MA Motor cortical control of vocal interaction in neotropical singing mice. Science 363, 983–988 (2019). [PubMed: 30819963]
- Tremblay P & Dick AS Broca and Wernicke are dead, or moving past the classic model of language neurobiology. Brain Lang 162, 60–71 (2016). [PubMed: 27584714]
- Hosman T et al. Auditory cues reveal intended movement information in middle frontal gyrus neuronal ensemble activity of a person with tetraplegia. Sci Rep 11, 98 (2021). [PubMed: 33431994]

- 70. Catani M et al. Short frontal lobe connections of the human brain. Cortex 48, 273–291 (2012). [PubMed: 22209688]
- Glasser MF et al. A multi-modal parcellation of human cerebral cortex. Nature 536, 171–178 (2016). [PubMed: 27437579]
- Mathis A et al. DeepLabCut: markerless pose estimation of user-defined body parts with deep learning. Nat Neurosci 21, 1281–1289 (2018). [PubMed: 30127430]
- 73. Deger K & Ziegler W Speech motor programming in apraxia of speech. Journal of Phonetics 30, 321–335 (2002).
- 74. Neef NE, Hoang TN, Neef A, Paulus W & Sommer M Speech dynamics are coded in the left motor cortex in fluent speakers but not in adults who stutter. Brain 138, 712–725 (2015). [PubMed: 25595146]
- 75. Jackson ES et al. A fNIRS Investigation of Speech Planning and Execution in Adults Who Stutter. Neuroscience 406, 73–85 (2019). [PubMed: 30851356]
- Dale AM, Fischl B & Sereno MI Cortical surface-based analysis. I. Segmentation and surface reconstruction. Neuroimage 9, 179–194 (1999). [PubMed: 9931268]
- 77. Fischl B et al. Automatically parcellating the human cerebral cortex. Cereb Cortex 14, 11–22 (2004). [PubMed: 14654453]
- Klein A & Tourville J 101 labeled brain images and a consistent human cortical labeling protocol. Front Neurosci 6, 171 (2012). [PubMed: 23227001]
- 79. Desikan RS et al. An automated labeling system for subdividing the human cerebral cortex on MRI scans into gyral based regions of interest. Neuroimage 31, 968–980 (2006). [PubMed: 16530430]
- Avants BB et al. A reproducible evaluation of ANTs similarity metric performance in brain image registration. Neuroimage 54, 2033–2044 (2011). [PubMed: 20851191]
- Tyszka JM & Pauli WM In vivo delineation of subdivisions of the human amygdaloid complex in a high-resolution group template. Hum Brain Mapp 37, 3979–3998 (2016). [PubMed: 27354150]
- Kovach CK & Gander PE The demodulated band transform. J Neurosci Methods 261, 135–154 (2016). [PubMed: 26711370]
- 83. Liu Y, Coon WG, Pesters A, de B, P. & Schalk, G. The effects of spatial filtering and artifacts on electrocorticographic signals. J. Neural Eng 12, 056008 (2015). [PubMed: 26268446]
- Friston KJ et al. Statistical Parametric Maps in functional imaging: A general linear approach. Human Brain Mapping 2, 189–210 (1995).
- 85. Qian T, Wu W, Zhou W, Gao S & Hong B in Annual International Conference of the IEEE Engineering in Medicine and Biology Society 2347–2350 (Boston, MA; 2011).

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#### Fig. 1 |.

Cortical responses during interactive speech. a, Example interaction between the experimenter and participant 436 during conversation (top) with a diagram depicting the three behavioural phases of turn-taking (bottom). b, A schematic of all turn-taking interactions with participant 436 (top) and the associated distribution of inter-turn gaps (bottom). The gold arrow indicates an example interaction from a. c–f, Example early (c, e) and late (d, f) CI trials with high gamma activity from participants 442 (c, d) and 436 (e, f). The timing of the experimenter question, the CI and the participant answer are indicated

with grey, light blue and black bars, respectively (coloured bars denote GLM regressor timing). g–j, Activity of the four example electrodes in c–f across all trials aligned to question (top) or CI (bottom): e55 (g), e125 (h), e26 (i) and e57 (j). The white ticks denote the onset and offset of the task epochs indicated with coloured arrows in c–j. In c–j, the waveforms are smoothed for display (500-ms mean boxcar filter). k–n, Perception, planning and production GLM weights for the example electrodes in g–j, respectively. Significant weights are indicated with asterisks. o, The distribution of all task-responsive electrodes in GLM weight space. The cluster membership is indicated by the greyscale colour. p, q, The distribution of electrodes displaying responses in one window of the CI task (p) or multiple windows (q). The response class is indicated by colour. In q, unmixed electrodes from p are denoted by small black points.



### Fig. 2 |.

Spatial organization of speech planning responses. a, The average median high gamma activity across trials from electrodes on the left lateral cortical surface and superior temporal plane at different phases of the CI task. The electrode locations are shown on the left. Median values from time windows <50 ms were not included for analysis. b, GLM classification of the responses depicted in a. c–e, Canonical cortical surfaces displaying electrodes with responses related to speech perception (c), production (d) and planning (e) across all participants. Electrodes that did not display a significant response for a given process are indicated with small white circles. f, The proportion of electrodes displaying planning responses (1-cm-diameter spatial smoothing) (top) and regions displaying consistent planning responses in the CI task labelled on a magnified cortical surface (bottom).



## Fig. 3 |.

Linguistic selectivity of planning responses. a–d, High gamma activity in single trials of each CR task: nonspeech hand (CR1; a) and orofacial (CR2; b) movements, speech repetition (CR3; c) and pluralization (CR4; d). The timing of experimenter question, CI and participant response are indicated with grey, coloured and black bars, respectively. The waveforms are smoothed for display (250-ms mean boxcar filter). e–h, The planning weights for each CR task plotted against planning weights from the CI task. The inset pie charts indicate the proportion of planning electrodes that were significantly responsive (coloured), not active (grey) or suppressed (black) for each CR task. i, The distribution of planning weights for all planning-related electrodes in the CR and CI tasks. The asterisks indicate

significant differences between tasks (P < 0.005, Friedman test with Dunn–Šidák post hoc tests). j, Median reaction times (RT) in late trials of the CR3, CR4 and CI tasks did not differ significantly (P = 0.34, Friedman test; NS, not significant). k–n, Canonical cortical surfaces indicating electrodes with planning responses for each CR task. Electrodes that did not produce a significant response for a given task are denoted by small white circles. o–r, The proportion of planning electrodes showing significantly elevated responses in each CR task (1-cm-diameter spatial smoothing). The dashed lines indicate approximate boundaries of the cMFG, the PreCG and the cIFG. s, The percentage of electrodes located within each structure that displayed planning responses for each CR task. The distribution of percentages generated with shuffled data is indicated in grey.



### Fig. 4 |.

Planning activity during natural conversation. a, High gamma activity from an example planning electrode in participant 472 (left) during the CI task (middle) and unconstrained conversation (right). The waveforms are smoothed for display (250-ms mean boxcar filter). b, c, For all CI task-responsive sites in participant 442, the electrode activity distribution in PC coefficient space during the CI task (b) and conversation (c). The dashed outlines indicate clusters and the numbers denote electrode identity. d, The PC coefficients for electrodes in each cluster from c. The individual electrode coefficients are depicted as points, the averages are indicated with bars and functional PC categorization is designated by bar colour. e, f, The average activity of perception and planning PCs aligned to the

offset of experimenter speech (e) and planning and production PCs aligned to the onset of participant speech (f) in the CI task (left) and conversation (right). The thin and thick lines indicate responses from individual participants and the average across participants, respectively. Periods of significantly elevated activity for each participant are denoted with bars under plots (P < 0.05, permutation test). g, Example production and planning PC activity during conversation with participant 436. The activity is displayed both as a waveform and by events whose thickness is scaled to PC amplitude (75th to 95th percentile). h, i, Timing of production (h) and planning (i) events for all interactions during conversation with participant turn onset and ordered by gap duration. The grey boxes represent experimenter and participant turn timing and unfilled pauses during participant turns left as whitespace. j–l, Empirical cumulative distribution functions of median perception, planning and production PC activity in experimenter turns (j), inter-turn gaps (k) and participant turns (l). Each line represents data from an individual participant. The inset plots depict the median value for each distribution as dots and the average median value across PCs as bars.