Interictal spike connectivity in human epileptic neocortex

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Abstract

Objective: Interictal spikes are a biomarker of epilepsy, yet their precise roles are poorly understood. Using long-term neocortical recordings from epileptic patients, we investigated the spatial-temporal propagation patterns of interictal spiking.

Methods: Interictal spikes were detected in 10 epileptic patients. Short time direct directed transfer function was used to map the spatial-temporal patterns of interictal spike onset and propagation across different cortical topographies.

Results: Each patient had unique interictal spike propagation pattern that was highly consistent across times, regardless of the frequency band. High spiking brain regions were often not spike onset regions. We observed frequent spike propagations to shorter distances and that the central sulcus forms a strong barrier to spike propagation. Spike onset and seizure onset seemed to be distinct networks in most cases.

Conclusions: Patients in epilepsy have distinct and unique network of causal propagation pattern which are very consistent revealing the underlying epileptic network. Although spike are epileptic biomarkers, spike origin and seizure onset seems to be distinct in most cases.

Significance: Understanding patterns of interictal spike propagation could lead to the identification patient-specific epileptic networks amenable to surgical or other treatments.

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1. Introduction

Epilepsy is a common neurological disorder characterized by chronic and recurrent seizures. Seizures are abrupt, temporal and spatial patterns of highly synchronized, rhythmic electrical activity that are often followed by a long duration of an altered brain state (Karoly et al., 2016). Due to the clinical importance of seizures in epilepsy and its ramifications resulting in behavioral changes, extensive research has been performed on seizures, its network and the relationship of seizure onset with epileptic foci (Lee et al., 2000; Worrell et al., 2004; Wilke et al., 2010; Dai et al., 2012; Bandt et al., 2014; Miao et al., 2014a; Bartolomei et al., 2016). However seizures are not frequent events, which makes the identification of epileptic foci using seizure onset alone difficult. Evidence suggests that abnormal brain networks are not only confined to seizures but in fact, there are more frequent interictal events (events in between two consecutive seizures) produced by
this pathological network (Korzeniewska et al., 2014). Amongst these interictal events, interictal spikes are defined as brief electrographic transients of approximately 250 ms or less, consisting of a short sharp wave, followed by a lasting slow wave (de Curtis and Avanzini, 2001; Staley and Dudek, 2006; Staley et al., 2011). These interictal spikes were previously thought to be significant biomarkers of epileptic brain regions along with seizures (de Curtis and Avanzini, 2001). Although, vast literature exists on interictal spikes in the epileptic brain, little is known about the dynamic network of these events and their relationship to seizures. While spikes often originate from synchronously firing neurons at or near seizure onset zones (Hufnagel et al., 2000; Asano et al., 2003, 2004; Marsh et al., 2010), they are also observed in cortical regions far from seizure onset (Hufnagel et al., 2000; Kobayashi et al., 2001). Moreover, clinically, interictal spikes may not only be a biomarker of epileptic brain regions, and may also have a significant, yet independent impact on behavior as they are present in a wide variety of neuropsychiatric disorders in the absence of seizures (Baglietto et al., 2001; Holmes and Lenck-Santini, 2006; Nicolai and Kastelein-Nolst Trenité, 2011; Ebus et al., 2012).

Interictal spikes have also been postulated to play a role in the formation of epileptic foci that produce seizures (Staley et al., 2005; Staley and Dudek, 2006; White et al., 2010), and animal models have shown that spike generation often precedes the development of seizures (White et al., 2010). The resection of high-spiking cortical regions also has been associated with decreased seizure events and an improved post-surgical outcomes (Palmini et al., 1995; Asano et al., 2003, 2009; Tomlinson et al., 2016). Conversely, other studies challenge this concept, suggesting that increased spiking actually protects the brain against seizure activity (Lieb et al., 1978; Lange et al., 1983; de Curtis and Avanzini, 2001). Recent investigations of spike occurrence (frequency) and its correlation with seizure onset regions, have produced conflicting interpretations of whether spike occurrence is a reliable predictor of seizure onset zones (Asano et al., 2013). This suggests that spike occurrence alone is not a functionally useful measure, and further analysis of interictal spikes including their source activity, network pathways of propagation and network characterization are necessary to develop a better understanding of the role of interictal spikes in epilepsy.

Here, we use an effective connectivity measure to map the causal network behavior of interictal spikes across different cortical regions in the human epileptic brain. Causal connections were computed using direct-directed transfer function (dDTF). We analyzed the dynamic nature of interictal spike networks over time in different patients across a wide range of frequency bands. We further assessed the effects of deep brain sulci, such as the central sulcus, on propagation patterns. Finally, we performed an initial evaluation on the relationships between interictal spike networks and physician identified seizure onset regions.

### 2. Materials and methods

#### 2.1. Patient selection and data collection

Ten pediatric epileptic patients (4F/6M, age: 2–15 years) with drug-resistant focal seizures were selected under a protocol approved by the institutional review board of Wayne State University and the University of Illinois at Chicago (Table 1). All the patients were identified as having intractable epilepsy and underwent surgical implantation of subdural recording electrodes for long-term seizure monitoring at the Children's Hospital of Michigan, Detroit, Michigan.

The electrocorticography (ECoG) recordings were obtained using a 192 channel Nihon Kohden Digital System (Nihon Kohden America Inc, Foothill Ranch, CA, USA) with 1000 Hz sampling rate from subdural electrodes (4 mm in diameter and spaced 10 mm apart, PMT platinum electrodes, PMT Corporation, Chanhassen, MN, USA). An experienced electroencephalographer help identify three 10-minute interictal ECoG recording segments for each patient. These segments patients were all awake but in quiet restfulness in the resting state (closed eyes, wakeful, and at least 6 h apart from any seizure activity). We obtained MRI scans for constructing 3D models of each patient’s brain using Freesurfer software (Dale et al., 1999; Fischl et al., 1999; Fischl and Dale, 2000). We localized and co-registered Electrodes on the 3D brain model using post electrode implantation CT images, and verified electrode locations with the help of intraoperative photographs taken at the time of grid removal (to ensure they did not move (Nakai et al., 2017). We then computed geodesic distances (tracking along the gray-white junction) between electrode positions and cortical thickness measures were computed (Zou et al., 2006; Wang et al., 2014).

Seizure onset zones (SOZ), as previously defined (Asano et al., 2009) were identified from intracranial recordings by an experienced epileptologist (EA). Each patient presented focal clinical or electrographic seizures along with complex epileptic behaviors, except for two patients where only subtle/minimal clinical symptoms were noted during the seizure events. Moreover, five out of ten patients also had epileptic spasms (Asano et al., 2005; Nariai et al., 2011). All patients had neocortical epilepsy often with mild pathological gliosis except for an individual with tuberous sclerosis and another with polymicrogyria.

#### 2.2. Signal analysis

For the purpose of the present study, we focused only on a subset of the electrodes, specifically a 64 channel (8 x 8) grid placed on the frontal-parietal lobe overlying the central sulcus, to understand the interictal spike behavior with a common reference and the role of the central sulcus in the propagation of epileptic spikes.

### Table 1

<table>
<thead>
<tr>
<th>patient</th>
<th>age</th>
<th>sex</th>
<th>diagnosis</th>
<th>total spikes in 10 mins</th>
<th>seizure focus</th>
</tr>
</thead>
<tbody>
<tr>
<td>Ep162</td>
<td>1</td>
<td>F</td>
<td>Polymicrogyria</td>
<td>6927</td>
<td>Left hemispheric</td>
</tr>
<tr>
<td>Ep164</td>
<td>3</td>
<td>F</td>
<td>White matter gliosis, superficial heterotopia</td>
<td>4274</td>
<td>Left temporal occipital</td>
</tr>
<tr>
<td>Ep165</td>
<td>3</td>
<td>F</td>
<td>Mild gliosis</td>
<td>4475</td>
<td>Left hemispheric</td>
</tr>
<tr>
<td>Ep169</td>
<td>8</td>
<td>M</td>
<td>Mild gliosis</td>
<td>3731</td>
<td>Right occipital-temporal</td>
</tr>
<tr>
<td>Ep181</td>
<td>13</td>
<td>M</td>
<td>Subcortical band heterotopia.</td>
<td>5664</td>
<td>Left temporal</td>
</tr>
<tr>
<td>Ep187</td>
<td>5</td>
<td>M</td>
<td>Tuberous sclerosis</td>
<td>4202</td>
<td>Left frontal</td>
</tr>
<tr>
<td>Ep196</td>
<td>2</td>
<td>M</td>
<td>Patchy subcortical myelinated fiber loss and gliosis</td>
<td>1427</td>
<td>Left frontal-temporal-parietal</td>
</tr>
<tr>
<td>Ep198</td>
<td>3</td>
<td>F</td>
<td>Gliosis</td>
<td>631</td>
<td>Left frontal-parietal</td>
</tr>
<tr>
<td>Ep199</td>
<td>2</td>
<td>M</td>
<td>Gliosis and patchy foci of decreased myelin</td>
<td>1980</td>
<td>Right temporal</td>
</tr>
<tr>
<td>Ep202</td>
<td>3</td>
<td>M</td>
<td>Tuberous sclerosis, and mild patchy cortical gliosis</td>
<td>6467</td>
<td>Right temporal</td>
</tr>
</tbody>
</table>
For our causal propagation analysis, we did not include other electrode locations that, in some cases, limited our ability to include all seizure onset and spiking areas for this analysis.

We processed these 64 ECoG signals through a multistage algorithm to evaluate the propagation of interictal spikes. We first identified interictal spikes using an established spike detection algorithm and manually validated these results with a trained epileptologist (JAL) and a panel of human reviewers (Barkmeier et al., 2012). Then, we isolated synchronous interictal spikes (peaks within 50 ms). We identified time blocks containing these synchronous spikes from 100 ms prior to and 250 ms after all spike peaks, excluding channels without interictal spikes.

2.3. Discrete short-time direct directed transfer function (dDTF) analysis and statistical validation

Post extraction, the time blocks were then processed for multivariate causality analysis using discrete short time direct directed transfer function (dDTF). dDTF is a Granger causality based model designed for quasi-stationary signals (Kamiński and Blinowska, 1991; Kus et al., 2004). This method predicts the direct flow of information between a pair of signals in the presence of all other signals in a multivariate environment, not confounded by any indirect causal interaction from other signals for a short time period. Prior to the signal processing for causality evaluation, the extracted spike epochs were tested for stationary nature using Phillips–Perron test (λ = 0.965% of spikes found stationary with p < 0.3) and Kwiatkowski–Phillips–Schmidt–Shin (KPPS) test (λ = 90% of spikes found stationary with p < 0.1). The isolated spike epochs were then processed for the dDTF evaluation following the recently described method by our group (Maharathi et al., 2016). Briefly, each spike epoch was fitted to a multivariate autoregressive model (MVAR) using modified covariance method to the form $X(t) = \sum_{i=1}^{m_{ord}} A(i)X(t-i) + E(t)$, where $X(t)$ is the data vector, $A(i)$ is the model coefficient, $E(t)$ is the zero mean uncorrelated white noise, and $m_{ord}$ is the order of the model fit. The model order was determined using Akaike information criterion (AIC), where model order of the fit corresponds to the model order with the minimum AIC value (max model order was fixed to 20, however the median model order across all patients was 8 ± 5).

Then the data model is transformed to frequency domain using the Z-transform for a specific frequency band of interest and the power spectra of the signal is calculated as $S(f) = H(f)VH^H(f)$, where $V$ is the variance of the Noise matrix and $H(f)$ is the transfer matrix. Following this, we computed two parameters, partial coherence as $\chi^2(f) = \frac{\hat{r}^2_{ij}(f)}{\sum_{i,j=1}^{m_{ord}} \hat{r}^2_{ii}(f)}$, where $m_{ord}$ is the minor of the spectral matrix after removing the $i$th row and $j$th column; and full frequency normalized direct transfer function (fDTF) as $n^2_{ij}(f) = \frac{|\hat{H}_{ij}(f)|^2}{\sum_{i,j=1}^{m_{ord}} |\hat{H}_{ij}(f)|^2}$. The direct directed transfer function is evaluated as the linear product of partial coherence and fDTF. It provides direct causal influence of $i$th channel on $j$th channel in the presence of all other channels and has values between 0 and 1. The above process was repeated for all such epochs of synchronous interictal spikes identified in the previous step. To understand the causal propagation of interictal spikes at each frequency band, the isolated spike epochs were processed through the dDTF algorithm for a specific frequency band, including low frequencies (1–50 Hz), delta (1–4 Hz), theta (4–8 Hz), alpha (8–13 Hz), beta (13–30 Hz), gamma (30–50 Hz) and high-frequency oscillations (HFO, 80–250 Hz).

We used surrogate data validation to determine the significance of each causal connection (Pritchard et al., 1995). Briefly, time blocks were transformed to the frequency domain using Fourier transform and the phase values were randomized. Post randomization, the data was transformed back to the time domain and processed for dDTF evaluation as mentioned above using same model parameters. The surrogate data method was repeated 100 times. The surrogate data dDTF estimation serves “false positive” level where no causal relationships among different electrode positions for the given dataset. If dDTF values were greater than the 95th percentile of these surrogate dDTF simulations, we declared this a causal propagation between electrodes (Maharathi et al., 2016). To understand how each patient’s data was consistent across time, we also determined similarity amongst their three 10-minute epochs (pairwise using a Pearson correlation coefficient). These values were normalized to a maximum of causal propagations to further evaluate the network activity of interictal spiking. An alpha level of 0.05 established statistical significance.

3. Results

3.1. Interictal spike propagation was highly reproducible

Using three 10-minute segments selected from each of 10 patients with neocortical epilepsy, we used an established spike detection algorithm to detect all spikes in all channels independently under an 8 × 8 subcortical grid centered over the central sulcus (Barkmeier et al., 2012). A trained electroencephalographer validated the accuracy of this algorithm for each patient. Next, clusters of interictal spikes within defined time periods were evaluated for synchrony; at least two peaks falling within 50 milliseconds of each other was considered a synchronous event. The dDTF was then used to identify causal propagations (Fig. 1).

This approach was robust, reproducible and revealed highly consistent causal propagation patterns of interictal spiking within each patient when evaluated at three different time segments (Fig. 2). Interestingly, along with being conserved, these dynamic network patterns were unique to each patient. Pearson correlation coefficients were used to measure the degree of correlation of the propagation patterns within each patient. For this, the number of onset propagations were summarized for each electrode within the selected electrode grid for individual 10 min ECoG segments. For each patient, the Pearson correlation coefficient was calculated for the corresponding electrode between each pair of 10 min ECoG files. In 9 out of 10 patients, a patient-wise high correlation was observed in onset patterns across the multiple ECoG segments (mean 0.83 ± 0.09) (Fig. 2B).

3.2. Propagation patterns were consistent across all frequency domains

An important area of research involves the role of epileptic activities in different frequency domains, where some studies suggest that higher frequency oscillations have good predictive values for epileptic brain foci (Burnos et al., 2014; Miao et al., 2014b; Höller et al., 2015; Frauscher et al., 2017). Here we examined the propagation pattern for each patient focusing on specific frequencies and found that each frequency band measured had the same propagation pattern (Fig. 3). Using the same three 10 min ECoG samples, we summarized the number of onset propagations from each electrode in the selected grid. Once evaluated, we calculated the multiple correlation between onset patterns of different frequency bands for corresponding electrodes in each 10 min ECoG segment for each patient. We found high onset correlation values for each frequency domain (delta = 0.92 ± 0.03, theta = 0.97 ± 0.01, alpha = 0.98 ± 0.01, beta = 0.98 ± 0.01, gamma = 0.96 ± 0.02, HFO = 0.94 ± 0.03 in mean ± SD) (Fig. 3A). Counting propagations in each frequency band (delta = 1430.3 ± 1697.9, theta = 3131 ± 3322.5, alpha =
3846.8 ± 3051.3, beta = 4389.1 ± 3061.5, gamma = 4707.4 ± 4221.3,
HFO = 3990.2 ± 3888.9 count/10 min) suggests that beta was most
''spike propagation active'' followed by the gamma and HFO
frequency bands. The total number of propagations in each
frequency band across patients varied with patient EP198 having
fewest propagations (delta = 73, theta = 207, alpha = 219,
beta = 163, gamma = 49.6, HFO = 72).

3.3. Regions of spike onset did not correspond to regions with
the highest spike occurrence

Most studies to date have focused on rate of spike occurrence
(frequency) as a measure of the epileptogenicity of a given brain
region, with regions exhibiting the highest spike occurrence consid-
ered to be more epileptic. Along with spike occurrence, we analyzed
the spike onset regions. Spike onset regions are defined as brain regions which have the greatest number of outgoing spike propagations above a defined maximum threshold within the selected grid. Interestingly, we did not find interictal spike onset regions to have the highest occurrence of spikes (Fig. 4A). In fact, electrodes with the highest spike occurrence were more likely to receive spikes from multiple brain locations. Out of all interictal spikes investigated, only 71.3 ± 17.7% of spikes were found present in the selected synchronous spike epochs, and not all synchronous spikes propagated. We isolated spike propagation events in each frequency band, and a small proportion propagated while also shifting frequency (delta = 25 ± 12.8%, theta = 45.1 ± 16.5%, alpha = 51 ± 16.5%, beta = 51.9 ± 18.7%, gamma = 44.2 ± 22.7%, HFO = 47.2 ± 23.6%). Only a small percentage of total locations acted as onset zones (delta = 7.9 ± 3, theta = 10.9 ± 3, alpha = 10.7 ± 3.4, beta = 11.5 ± 3.6, gamma = 12.2 ± 5.6, HFO = 8.6 ± 3.2). Many spikes participated in spread of interictal spike patterns (delta = 13.4 ± 7.4, theta = 20.3 ± 8.8, alpha = 17.5 ± 7.4, beta = 14.2 ± 6, gamma = 9.7 ± 4.2, HFO = 16.9 ± 9.5), and hence were “intermediaries” of propagation, having incoming and outgoing events (delta = 3.7 ± 3, theta = 14 ± 6.5, alpha = 22.8 ± 8.8, beta = 26.2 ± 11.2, gamma = 22.3 ± 14.7, HFO = 21.7 ± 15.6). In summary, dDTF analysis suggests that a majority of cortical epileptic spiking brain regions are receiving rather than generating interictal spikes. Furthermore, only a small proportion of spike onset regions brain areas truly initiate spiking, as most act as intermediary regions that both receive and spread the spike to other areas (p < 0.05 across all frequency bands, pairwise t-test).

Based on these findings, we developed spatial models of outgoing (spike onset) and incoming (spike spread) epileptic activities as shown in Fig. 4B. Isolating individual electrodes on the grid shows highly consistent patterns of both outgoing and incoming activity, providing a unique look at the epileptic connectivity of these brain regions. dDTF analysis therefore allows a deeper understanding of the cortico-cortical networks that underlie interictal spiking which could have important downstream value in improving surgical outcomes in patients with epilepsy.

3.4. Spatial patterns of interictal spike propagation related to cortical structures

We next combined the temporal and directional patterns of interictal spikes using dDTF with 3D brain models for each patient as a means to understand how brain structure influences spike propagation. Fig. 5 shows that while approximately 50% of outgoing spike propagations travel to nearby, often adjacent electrodes (~50 mm geodesic distance), the other half of spikes propagate over longer distances. Surprisingly, approximately 22% of spikes propagated over 10 cm (geodesic distance) from their site of onset. In addition, the directionality of spike propagation was nonrandom and affected by cortical topography (Fig. 6). Since all patients had an electrode grid overlaying the central sulcus, we determined the relative number of spike propagations that crossed the central sulcus versus electrode locations on the same side of central sulcus. Surprisingly, a majority (76 ± 9.6%) of spike propagations did not cross the central sulcus, in either frontal or parietal lobes. This result was also independent from ECOG frequency bands.

3.5. Spike and seizure onsets regions were not often interrelated

Even though interictal spikes are considered prominent biomarkers of epileptic brain regions that also produce seizures (Palmini et al., 1995), their exact spatial and temporal associations with seizures are not well understood. We investigated whether seizure onset regions had any association with the onset or spread of interictal spiking brain regions. We first considered electrodes with activity exceeding 80% of maximum of the entire grid in at least one of the three files. Fisher’s exact test along with Chi-square tests of independence was used to test for any relation between seizure onset regions and (1) interictal spiking regions (2) spike onset regions and (3) spike spread regions. In patient Ep164, there was a strong relationship between seizure onset and interictal spiking regions, spike onset regions, and spike spread regions that were all significant (p < 0.05). However, in the remaining 7 patients, the spike network had a more complex relationships, if any, with seizure onset regions (Fig. 7). For example, in patient Ep169, seizure onset overlapped more with high spiking regions compared to spike onset and spread regions whereas in patient Ep199, seizure onset regions were well associated with spike onset and spread regions (p < 0.05). In patients Ep165 and Ep196, who also had infantile spasms, seizure onset regions were diffuse an encompassed a large portion of the grid. Further analysis will be required to understand fully the relationships between spike onset and seizure onset, especially in patient with infantile
spasms who often have both diffuse as well as focal seizure onset regions. Two of the patients (Ep181 and Ep187), had their seizure active zones outside of the measurement grids and were therefore removed from consideration. However, even in these patients, the spike onset and spread regions were located on the electrodes in close proximity to seizure onset regions.

4. Discussion

We performed a novel effective connectivity analysis using dDTF to better understand the spatial-temporal patterns of interictal spike propagation in the human epileptic neocortex. We observed consistent pattern of interictal spike propagation which are individualized for each patient and are highly reproducible across multiple days of recording for the same patient. We also found that spikes propagate in a similar fashion across a wide range of frequency bands, including HFOs. Our findings shed new light on the relationships between cortical anatomy and spike propagation. Specifically, propagations often travelled locally, also across great distances but rarely and the central sulcus appears to be a significant barrier for interictal spike propagation which is in congruent with previous studies reporting that signal propagation often skips the motor area (Iimura et al., 2017). Surprisingly, brain regions with the highest spike occurrences most often did not initiate spiking. While our study was primarily focused on spike onset, we found no clear relationships between interictal spike onset regions and seizure onset regions, suggesting that spikes and seizures might be having different networks.

Interictal spikes are known to be highly associated with the epileptic state (Staley and Dudek, 2006; Rodin et al., 2009) and are thought to result from synchronous cortical hyper-excitation of neural assemblies (Pillai and Sperling, 2006). Previous studies have found that spikes have consistent frequencies and locations for each patient's interictal spike network and interictal spikes have been used to confirm the clinical diagnosis of epilepsy and to assist in the planning of drug management and epilepsy surgery (Pillai and Sperling, 2006; Rodin et al., 2009). The spatially conserved pattern of spike propagation within each patient signifies an invariability of the epileptic network over time. This suggests that once the wiring of neurons turns pathologic, the underlying network of interictal epileptic activity remains quite consistent.
A major question in epilepsy is whether interictal spikes are simply singlet spikes from a repetitively spiking seizure focus or represent an entirely different network. Previous research has suggested that interictal spikes proceed the development of seizures, but when increased, correlate with spontaneous seizure events (White et al., 2010). It might therefore be possible to predict seizure onset regions based on occurrence of interictal spiking. However, there is widely discordant data for this hypothesis as it is well known that regions of high interictal spiking are not always seizure onset regions (Hufnagel et al., 2000; Asano et al., 2004; Jacobs et al., 2008; Marsh et al., 2010). There is conflicting data to support the removal of high spiking regions for a positive surgical outcome (Alarcon et al., 1997; Bautista et al., 1999; Asano et al., 2009; Mégevand et al., 2014). An important new observation from our
study that may help resolve this discordance is that most regions with the highest interictal spiking frequency are not areas of spike onset, but areas that receive the spread of spikes.

All of these findings taken together favor the hypothesis that interictal spiking and seizure networks are distinct with different mechanisms of origin (Karoly et al., 2016). Surprisingly, except in one patient, we did not find a tangible relationship between spike onset and seizure onset. However, in the present study we did not use data from all brain regions covered by electrodes that could also be involved, nor did we focus on less frequent spiking regions in our dDTF analysis that focuses on the most common patterns of spread. Clearly, more research will be needed to explore the complex relationships between the interictal spike and seizure networks. Further studies will need to look at not only the spatial, but also the temporal relationships between interictal spike onset regions and seizure onset regions.

Another physiological aspect of epilepsy that has gained significant interest is in the area of HFOs. Some have found HFOs to highly correlate with seizure onset regions (Jirsch et al., 2006; Ochi et al., 2007; Jacobs et al., 2008; Salami et al., 2014; Modur and Miocinovic, 2015). We looked at the propagation patterns embedded within interictal spikes and found that all frequencies, from lower frequency bands to HFOs propagate in a similar fashion. This may not be too surprising as it has previously been shown that high frequency oscillations are often coexistent with interictal spikes (Jacobs et al., 2008, 2016; Rodin et al., 2009) and in the present study we focused entirely on interictal spiking.

By co-registering electrode locations with 3D MRI reconstructions, we examined the spatial aspects of spike propagation with respect to cortical structures. We found that the distance interictal spikes propagate is generally quite short, most often to adjacent electrodes, spikes can also traverse larger distances, likely through subcortical pathways, but the chances are few. Perhaps the most salient observation was that the central sulcus acts as a significant anatomic barrier to spike propagation in all patients we examined. The central sulcus divides primary sensory from motor cortex and is one of the deepest sulci in the brain that has been extensively studied for its functional localization and cortical plasticity. Identifying the location of the central sulcus is critical in surgical planning to reduce unwanted motor and sensory deficits (Towle et al., 2003). Previous studies have found that spikes originating from sulcal and gyral cortices on either side of central sulcus propagate across central sulcus (Jung et al., 2003). In contrast, when looking at all spiking events statistically, we observed that the central sulcus is a strong barrier to the spread of spikes. A recent study on human epileptic tissues that defined tissue markers of spiking found that molecular markers of epileptic activity in the MAPK/CREB pathway also failed to cross deep gyri, suggesting an underlying molecular and cytoarchitectonic underpinning that creates anatomical barriers across the cortex (Beaumont et al., 2012).

Although from the current study it is difficult to point out the exact mechanism of how central sulcus acts as a barrier, a probable reason might be the large cortical surface coverage of the deep sulci and thinner cortex. More studies will be needed to connect structural and functional information together using a variety of connectivity measures including cortico-cortical evoked potentials or fiber tractography. This will allow further exploration of the effects of deeper sulci and other anatomical abnormalities on spike prop-
agation networks across different brain regions. Because our study only involved surface grid electrodes, examination of spike propagation patterns from patients with stereo EEG could also help in this endeavor. Finally, propagation patterns may be age dependent and examining patients within different age groups would be helpful in learning how these networks may change as a function of brain development.

Several signal analysis tools have been proposed to quantify and understand interictal spike propagations. Granger causality based functional measures (Granger, 1969), such as direct transfer function (DTF) (Bressler and Seth, 2011) and partial direct coherence (PDC) (Baccalá and Sameshima, 2001), evaluate the causal flow of information between signals in a multivariate environment. They provide statistical estimates of connection strength along with the direction that information flows. Since these evaluations compute the causality based on phase differences, they are robust to volume conduction as compared to other bivariate analysis. With simulated and experimental data, it has been shown that an enhancement of the DTF algorithm, direct DTF (dDTF), is more efficient in identifying causal connections (statistically) with fewer false detections. It also has a lower fictitious causal density and high spectral selectivity (Astolfi et al., 2005, 2007; Bressler and Seth, 2011; Fasoula et al., 2013). Such approaches enable robust understanding of path and timing of network signal propagations. More recently, the direct-Directed dDTF has been successfully implemented in previous studies evaluating seizure propagation and epileptic source localization Manuscript– Spike Propagation—clinical neurophysiology_v4.

The clinical implications for this work are still not fully developed. Although there are several limitations in the current study, and primarily we focused on evaluating the interictal spike propagation patterns, taken together our findings suggest that interictal spikes and seizures develop nearby, but from different networks. Whether removal of spike onset regions would produce a better surgical outcome remains an open question not answerable by our current retrospective data. However, application of this analysis prospectively could allow a greater understanding on whether disrupting or removing interictal spiking networks on surgical outcome.

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Conflict of Interest

None of the authors report any conflict of interest.

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