



# Low-Intensity Pulsed Ultrasound Stimulation Modulates the Nonlinear Dynamics of Local Field Potentials in Temporal Lobe Epilepsy

Xin Li<sup>1</sup>, Huifang Yang<sup>1</sup>, Jiaqing Yan<sup>2</sup>, Xingran Wang<sup>1</sup>, Xiaoli Li<sup>3\*</sup> and Yi Yuan<sup>1\*</sup>

<sup>1</sup> Institute of Electrical Engineering, Yanshan University, Qinhuangdao, China, <sup>2</sup> College of Electrical and Control Engineering, North China University of Technology, Beijing, China, <sup>3</sup> State Key Laboratory of Cognitive Neuroscience, Beijing Normal University, Beijing, China

## OPEN ACCESS

### Edited by:

Ioan Opris,  
University of Miami, United States

### Reviewed by:

Chengbo Meng,  
National Institute of Environmental  
Health Sciences (NIEHS),  
United States

Shucaai Wu,  
Beijing University of Technology,  
China

### \*Correspondence:

Xiaoli Li  
xiaoli@bnu.edu.cn  
Yi Yuan  
yuanyi513@ysu.edu.cn

### Specialty section:

This article was submitted to  
Neural Technology,  
a section of the journal  
Frontiers in Neuroscience

**Received:** 18 December 2018

**Accepted:** 11 March 2019

**Published:** 02 April 2019

### Citation:

Li X, Yang H, Yan J, Wang X, Li X  
and Yuan Y (2019) Low-Intensity  
Pulsed Ultrasound Stimulation  
Modulates the Nonlinear Dynamics  
of Local Field Potentials in Temporal  
Lobe Epilepsy.  
*Front. Neurosci.* 13:287.  
doi: 10.3389/fnins.2019.00287

Low-intensity pulsed ultrasound stimulation (LIPUS) can inhibit seizures associated with temporal lobe epilepsy (TLE), which is the most common epileptic syndrome in adults and accounts for more than half of the cases of intractable epilepsy. Electroencephalography (EEG) signal analysis is an important method for studying epilepsy. The nonlinear dynamics of epileptic EEG signals can be used as biomarkers for the prediction and diagnosis of epilepsy. However, how ultrasound modulates the nonlinear dynamic characteristics of EEG signals in TLE is still unclear. Here, we used low-intensity pulsed ultrasound to stimulate the CA3 region of kainite (KA)-induced TLE mice, simultaneously recorded local field potentials (LFP) in the stimulation regions before, during, and after LIPUS. The nonlinear characteristics, including complexity, approximate entropy of different frequency bands, and Lyapunov exponent of the LFP, were calculated. Compared with the control group, the experimental group showed that LIPUS inhibited TLE seizure and the complexity, approximate entropy of the delta (0.5–4 Hz) and theta (4–8 Hz) frequency bands, and Lyapunov exponent of the LFP significantly increased in response to ultrasound stimulation. The values before ultrasound stimulation were higher  $\sim 1.87$  (complexity),  $\sim 1.39$  (approximate entropy of delta frequency bands),  $\sim 1.13$  (approximate entropy of theta frequency bands) and  $\sim 1.46$  times (Lyapunov exponent) than that after ultrasound stimulation ( $p < 0.05$ ). The above results demonstrated that LIPUS can alter nonlinear dynamic characteristics and provide a basis for the application of ultrasound stimulation in the treatment of epilepsy.

**Keywords:** LIPUS, TLE, nonlinear dynamics, LFP, mice

## INTRODUCTION

Epilepsy is one of the most common neurological diseases in the world. Repeated seizures will cause anxiety, depression, accidental disability, and even death and greatly affect quality of life (Viale et al., 2015; Bielefeld et al., 2017; O'Connell et al., 2017). The reason for epileptic seizures is the sudden abnormal firing of neurons. During epileptic seizures, there are transient dysfunctions in the brain that trigger short-term obstacles in patient movement and consciousness (Bjørk et al., 2015; Robertson et al., 2015; Xu et al., 2016). Lesions in different parts of the brain cause different

types of epilepsy, and there is no one brain area associated with all types of epilepsy. Temporal lobe epilepsy (TLE) is the most common epileptic syndrome in adults and accounts for more than half of the cases of intractable epilepsy. Clinically, TLE has the characteristics of frequent seizures and difficult diagnosis and treatment (Monti and Meletti, 2015; Lévesque et al., 2016; Steinhäuser et al., 2016).

Electroencephalography (EEG) signals of epilepsy are an important tool for studying epilepsy (Mario et al., 2003; Noachtar and Rémi, 2009). Studies have shown that excessive repetitive firing of some neurons may cause their EEG time series to be in a lower dimensional chaotic motion state with epileptic seizure (Bosl et al., 2017). The characteristic parameters of EEG chaos can be used as biomarkers for the prediction and diagnosis of epilepsy (Kopczynska et al., 2018). Past studies have shown that complexity, approximate entropy and Lyapunov exponents were used as biomarkers of nonlinear characteristics of epileptic EEG signals (Varatharajah et al., 2018).

The complexity of EEG signals and functional status in the brain during epileptic seizures can be quantitatively evaluated by complexity (Mei et al., 2018). Previous studies have demonstrated that the complexity of EEG signals during epileptic seizures is lower than that in healthy humans (Wen-Chin et al., 2015; Noel et al., 2017). The complex characteristics and probability of chaotic systems can be reflected by the entropy. The epileptic seizures can be judged and measured by entropy as a biomarker (Li et al., 2018; Raghu et al., 2018). The sensitivity of the chaotic system to different initial values can be evaluated by the Lyapunov exponent. In the epileptic interval, the Lyapunov exponent is at a relatively low value; during the seizure, the Lyapunov exponent drops sharply and then gradually increases. After the seizure, the Lyapunov exponent remains elevated for a period of time and then returns to normal levels (Osowski et al., 2007). The above studies showed that the nonlinear characteristic parameters, including complexity, approximate entropy and Lyapunov exponent, changed significantly during epileptic seizures and can be used as biomarkers for the prediction and diagnosis of seizures.

Low-intensity pulsed ultrasound stimulation (LIPUS), as a new noninvasive neuromodulation technique, has received increasing attention (Landhuis, 2017). LIPUS has the advantage of high spatial resolution and high penetration depth (Alexander et al., 2011). The potential mechanisms of ultrasound stimulation for neuromodulation may be that ultrasound wave alter intramembrane cavitation, membrane conductance or mechanically sensitive ion channels to cause neuronal discharge (Tyler, 2011; Plaksin et al., 2014; Ye et al., 2018). Previous studies used LIPUS to stimulate different regions of the brain of rodent animals, monkeys, and humans. They found that LIPUS can modulate neuronal activity (Tyler et al., 2008; Tufail et al., 2010), neural network connections (Yu et al., 2016), cerebral hemodynamics (Yang et al., 2018; Yuan et al., 2018), and water molecular diffusion (Yuan et al., 2017). Previous studies have also shown that LIPUS can cause a reduction in the occurrence of epileptic EEG bursts and severity of epileptic behavior and can also result in fewer spontaneous recurrent seizures and improved performance in behavioral tasks assessing

sociability and depression in the chronic period of epilepsy (Min et al., 2011; Hakimova et al., 2015). However, how LIPUS affects the key biomarkers of the nonlinear characteristics of EEG signals in TLE is still unclear.

To address this issue, we used LIPUS to stimulate the CA3 region of KA-induced TLE mice, simultaneously recorded local field potentials (LFP) in the stimulation regions before, during and after ultrasound stimulation. The nonlinear characteristics, LFP were analyzed.

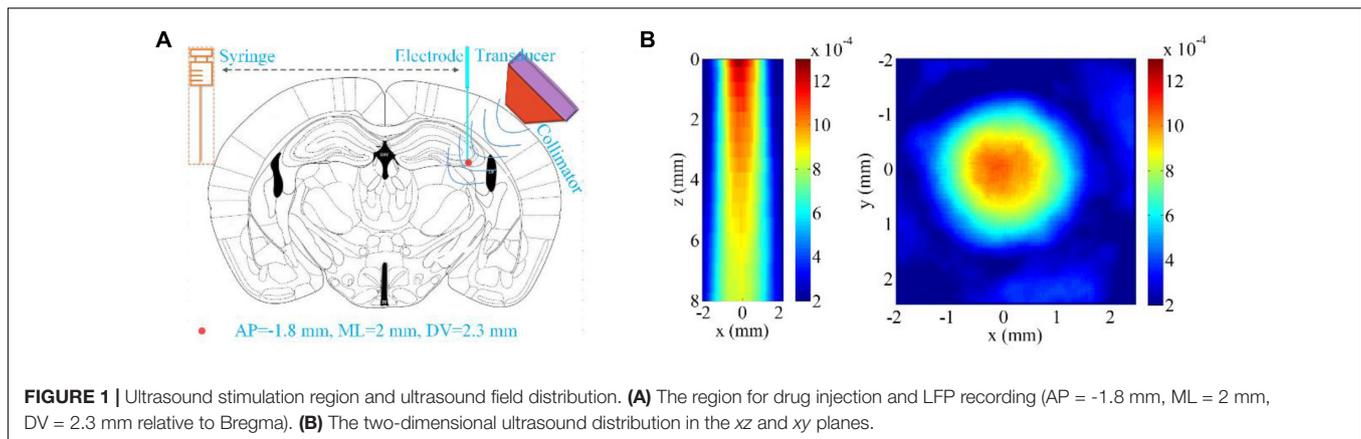
## MATERIALS AND METHODS

### Animals and Surgery

Fourteen C57BL/6 mice were randomly separated into two groups, the LIPUS group and the control group (all male, body weights 20–25 g, Beijing Vital River Laboratory Animal Technology Co., Ltd., China). Our study protocols were submitted to and approved by the Animal Ethics and Administrative Council of Yanshan University. All mice were anesthetized by 2% isoflurane (RWD Life Science Co., Shenzhen, China) in the experiment. A closed-loop animal temperature maintenance instrument (69002, Ruiwode, Shenzhen, China) was used to maintain body temperature at  $\sim 37^{\circ}\text{C}$  during all experiments. We referred to the previous work to make the TLE mouse model (Hakimova et al., 2015). As shown in **Figure 1A**, KA (1 mg/1 ml in saline, Tocris, United States) was unilaterally microinfused into the CA3 area (AP = -2.0 mm, ML = -2.3 mm, and DV = 2.0 mm from the bregma). Tungsten microelectrodes (WE50030.1B10, MicroProbe, United States) were inserted into the injection region 2 h after KA infusion.

### LIPUS Protocol and LFP Signal Recording

The LIPUS system was similar to that used in our previous paper (Yuan et al., 2018). In the LIPUS group, the fundamental frequency (FF), pulsed repetition frequency (PRF), duty cycle (DC), and stimulation duration (SD) of the pulsed ultrasound were 500 kHz, 500 Hz, 50% and 30 s, respectively. In the control group, the parameter settings were the same as those in the LIPUS group, excepted that we turned off the power amplifier. The unfocused ultrasound transducer (V301-SU, Olympus, United States) was connected to the mouse skull by a 3D printed conical collimator filled with ultrasound coupling gel. The diameter of the hole at the bottom of the conical collimator was 2 mm. The two-dimensional ultrasound distribution under *ex vivo* mouse skull in the *xz* and *xy* planes shown in **Figure 1B** was measured by a calibrated needle-type hydrophone (HNR500; Onda, United States) that was moved by a two-dimensional electric translation platform. The ultrasound pressure of both pulsed and continuous ultrasound was 0.26 MPa. A microelectrode inserted into the CA3 region was used for recording the LFP signals before, during and after LIPUS. The amplification and acquisition of LFP signals was the same as in our previous study (Yuan et al., 2016). Raw LFP signals were acquired



at a sampling frequency of 2 kHz, and a low-pass filter with a 1000-Hz cutoff frequency was set in the differential preamplifier system.

The seizures were defined by the LFP according to the previous literature (amplitude:  $3 \times$  background, duration:  $> 10$  s) (Hakimova et al., 2015). When seizures were confirmed, the LFPs were continuously recorded for 90 s in the control group. When the seizures lasted more than 30 s in the LIPUS group, we applied ultrasound stimulation for 30 s and recorded for a total duration of 90 s.

## Nonlinear Dynamics Analysis of LFP Signals and Statistical Analysis

### Lemple-Zie Complexity

We calculated the Lemple-Zie complexity of the LFP according to previous literature (Sabeti, 2009). The normalized Lemple-Zie complexity can be expressed as

$$LZC = C(n)/b(n) \quad (1)$$

where  $C(n)$  is the complexity of the LFP sequence,  $b(n) = N/\log_2(n)$ . When  $N$  tends to infinity,  $b(n)$  provides an asymptotic value of  $C(n)$ .

### Approximate Entropy

We extracted the signals of different frequency bands (delta [0.5–4 Hz], theta [4–8 Hz], alpha [8–13 Hz], beta [13–30 Hz], gamma [30–45 Hz]) from the LFP and then calculated the approximate entropy of the signals in different frequency bands based on the literature (Jing et al., 2006; Mateo et al., 2006). The approximate entropy can be expressed by the following equation:

$$ApEn(m, r, N) = \phi^m(r) - \phi^{m+1}(r) \quad (2)$$

where  $N$  is the length of data,  $m$  is the vector dimension, and  $r$  is the tolerance.

### Largest Lyapunov Exponent

The largest Lyapunov exponent was calculated by the method mentioned in a previous study (Osowski et al., 2007). The largest

Lyapunov exponent can be expressed as

$$L = \frac{1}{K\Delta t} \sum_{j=1}^K \log_2 \frac{d_j(\Delta t)}{d_j(0)} \quad (3)$$

The detailed parameters of the above equation can be found in the literature (Osowski et al., 2007).

### Statistical Analysis

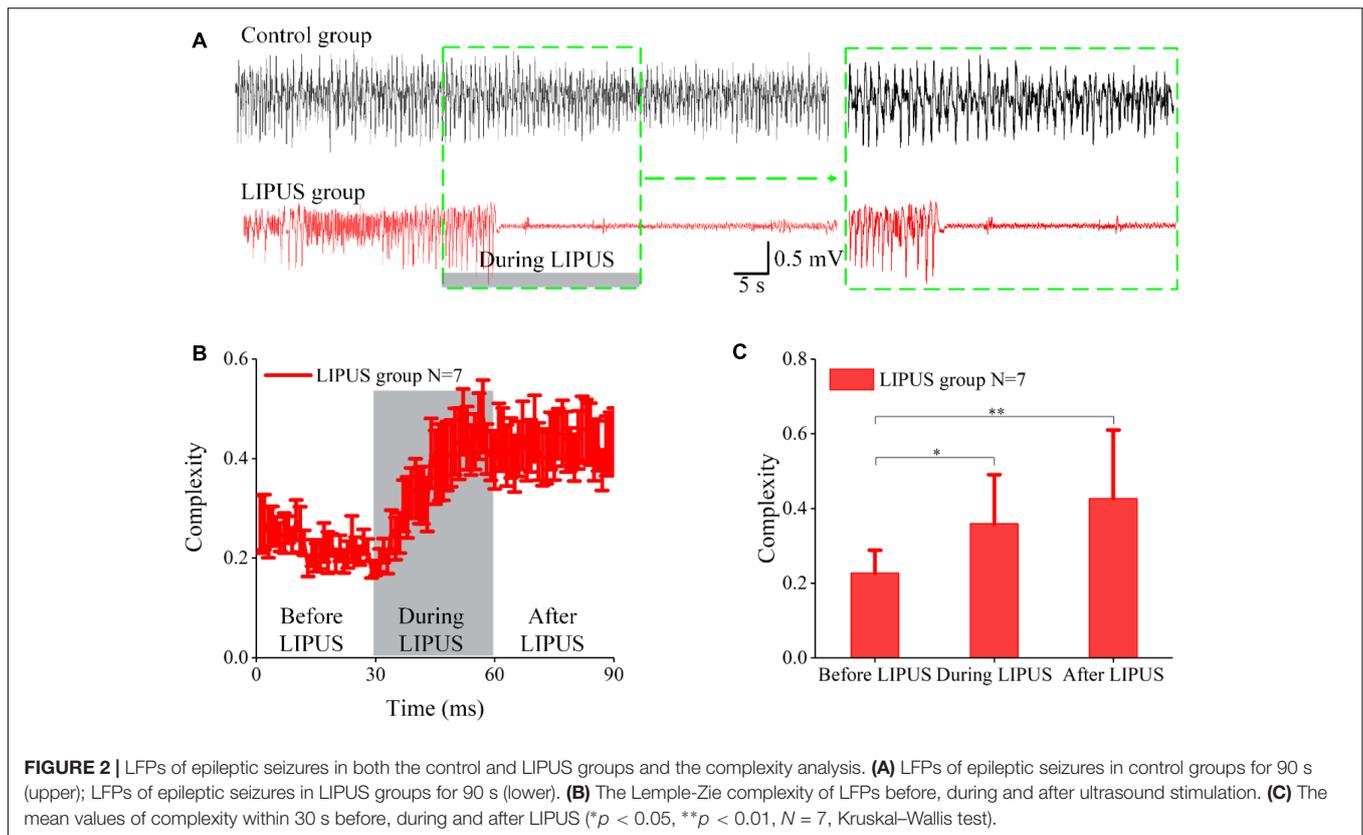
The results were analyzed by the Kruskal–Wallis test, and a difference was deemed significant if  $p < 0.05$ . Statistical analysis was performed using MATLAB software.

## RESULTS

### Effect of LIPUS on LFP Complexity in Acute TLE

We first proved that LIPUS can inhibit TLE seizures. **Figure 2A**(upper) shows continuous seizure for 90 s in the epilepsy group without LIPUS. In the LIPUS group [**Figure 2A**(lower)], continuous seizure lasted for 30 s. When ultrasound stimulation began, the epileptic seizure was not immediately suppressed, but was suppressed as the ultrasound stimulation continued and stopped after LIPUS. This result was consistent with that in the previous literature (Hakimova et al., 2015).

To evaluate the effect of LIPUS on LFP complexity in acute TLE, we calculated the Lemple-Zie complexity of LFP before, during and after ultrasound stimulation. As shown in **Figure 2B**, the complexity was at a low level and remained the same as that observed before LIPUS. When ultrasound stimulation was applied, the complexity gradually increased during LIPUS and was then elevated after LIPUS. We calculated the mean values of complexity within 30 s before, during and after LIPUS, and the results are shown in **Figure 2C**. The complexity after LIPUS was  $\sim 1.87$  times higher than that before LIPUS, representing a significant difference ( $**p < 0.01$ ,  $N = 7$ , Kruskal–Wallis test). In the control group, as shown in **Supplementary Figures S1A,B**, there was no significant change in complexity during



the 90-s period. The above results indicated that ultrasound stimulation can inhibit TLE seizures and alter the complexity of epileptic LFP signals.

### Effect of LIPUS on the Approximate Entropy of the LFP in Acute TLE

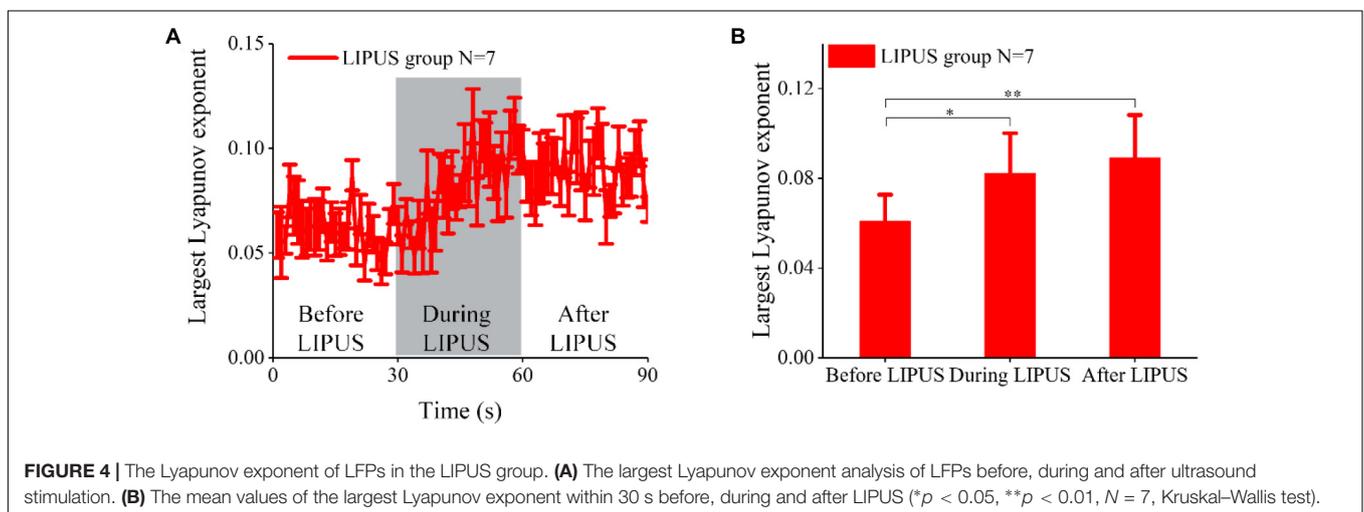
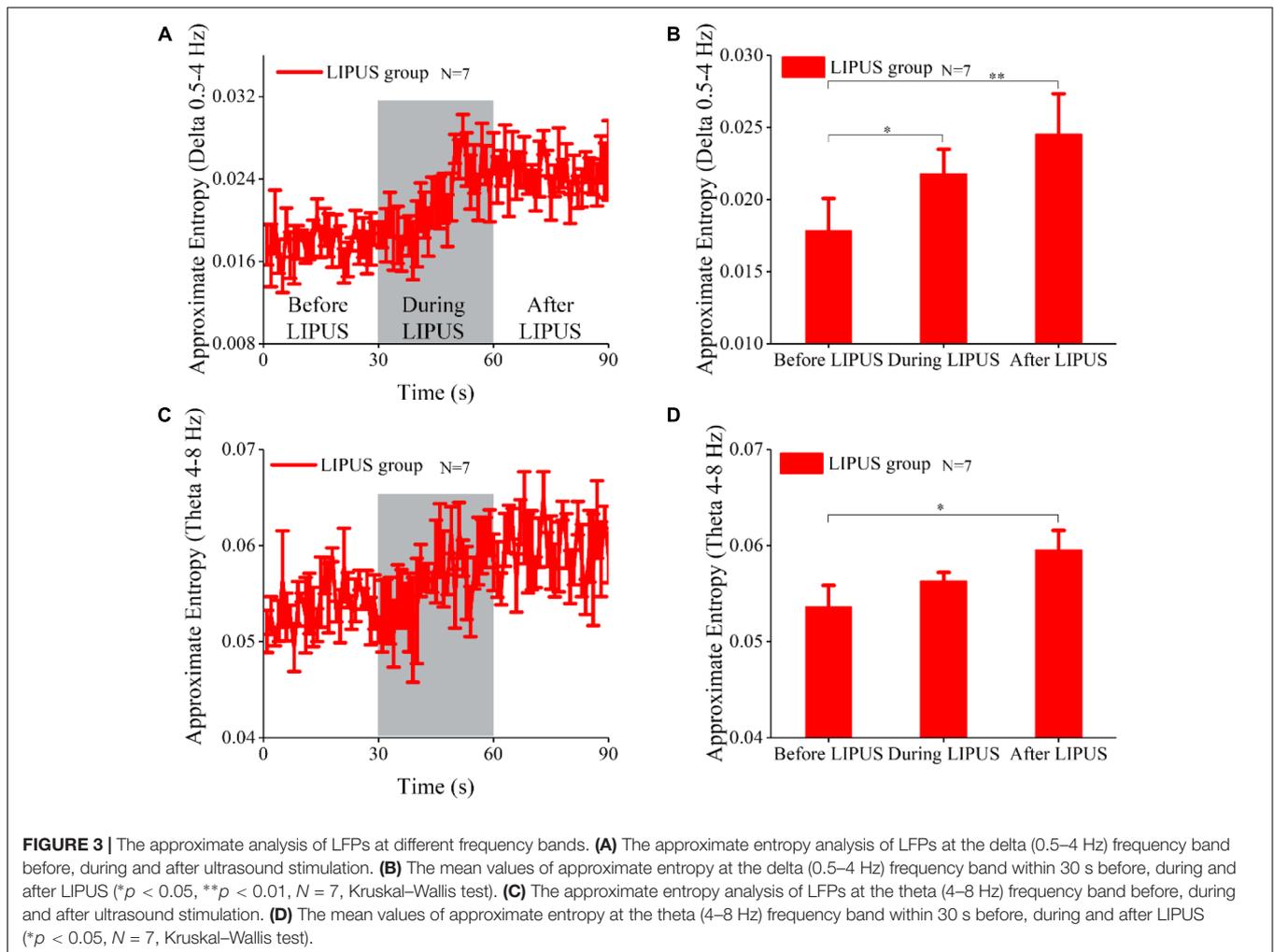
The approximate entropy of different frequency bands is often used to evaluate changes in EEG signals. Here, we analyzed the approximate entropy of LFP signals at different frequency bands in the LIPUS group. As shown in **Figure 3**, the approximate entropy at the delta (0.5–4 Hz) and theta (4–8 Hz) frequency bands was significantly modulated by ultrasound. The trends of the changes in approximate entropy at the delta (0.5–4 Hz) and theta (4–8 Hz) frequency bands were similar to that of complexity. We also calculated the mean value of approximate entropy within 30 s before, during and after LIPUS (**Figures 3B,D**). The approximate entropy at delta and theta after LIPUS was  $\sim 1.39$  (delta) and 1.13 (theta) times higher than that before LIPUS ( $N = 7$ , Kruskal–Wallis test, \* $p < 0.05$ ). As shown in **Supplementary Figure S2**, there was no significant change in approximate entropy at the frequency bands of alpha (8–13 Hz), beta (13–30 Hz), or gamma (30–45 Hz) before, during and after LIPUS. The approximate entropy remained at a stable level with epileptic seizures during the 90-s period in the control group, as shown in **Supplementary Figure S3**. The above results indicated that ultrasound stimulation can only modulate the approximate entropy of the LFP at the delta and theta frequency bands.

### Effect of LIPUS on the Largest Lyapunov Exponent of the LFP in Acute TLE

Here, we analyzed the maximum Lyapunov exponent of the LFP in both the control and LIPUS groups. The experimental results of the LIPUS group were shown in **Figure 4**. The largest Lyapunov exponent did not change significantly with epileptic seizure and gradually increased with ultrasound stimulation, finally reaching a higher level after LIPUS. Statistical analysis indicated that the largest Lyapunov exponent after LIPUS was  $\sim 1.46$  times higher than that before LIPUS ( $N = 7$ , Kruskal–Wallis test, \* $p < 0.05$ ). The Lyapunov exponent in the control group showed no significant change, as shown in **Supplementary Figure S4**. The above results verified that ultrasound stimulation can significantly modulate the maximum Lyapunov exponent of the LFP in acute TLE.

## DISCUSSION

Electroencephalography signals contain a large amount of physiological and pathological information, and analysis of EEG signals is an important and powerful tool for studying epilepsy (Acharya et al., 2011). In past studies, researchers have proposed many methods to evaluate the diagnostic and therapeutic effects of epilepsy by analyzing epileptic EEG signals (Giannakakis et al., 2014). In these identification methods for epilepsy, the nonlinear characteristic parameters of EEG, including entropy, complexity and Lyapunov exponent, are often used as important biomarkers.



In this study, low-intensity ultrasound was used to stimulate the CA3 region for acute TLE, and LFPs of the CA3 region were simultaneously recorded. Our results demonstrated that low-intensity ultrasound can effectively modulate nonlinear dynamics

in TLE. The difference in the nonlinear characteristics of epileptic EEGs is closely related to the abnormal synchronous firing of brain neurons during seizures (Lehnertz, 2008). Ultrasound stimulation modulates the abnormal synchronous firing of

brain neurons and inhibits epileptic seizures, thus causing a significant change in the nonlinear dynamic characteristics of epileptic EEGs. To our knowledge, this is the first report on the nonlinear dynamic characteristic of ultrasound modulation of seizures in acute TLE.

Epilepsy EEG analysis has an important guiding role in the treatment of epilepsy by LIPUS. In this study, we demonstrated that LIPUS can modulate the nonlinear dynamic characteristics of epilepsy EEG. Therefore, when LIPUS is used to intervene in the electrophysiological activity of the brain, if the nonlinear dynamic characteristics of the EEG signal can be monitored, the different states of epilepsy can be correctly identified. Furthermore, this technique can guide the selection of appropriate ultrasound parameters, such as ultrasound intensity, pulsed repetition frequency, stimulation duration or duty cycles.

In the previous study, Hakimova et al. used low-intensity ultrasound to stimulate mice with TLE and found that LIPUS can effectively inhibit seizure of TLE. There are some similarities and differences between our and Hakimova et al's studies. (1) Similarities: KA was used to make mouse model of TLE and the location of the KA injection and recording electrode was the same. The duration of ultrasound stimulation was 30 s in both studies. (2) Differences: In Hakimova et al's study, they investigated behavior, number of seizures, latency of seizure and seizure spikes of mice with TLE under LIPUS. In our study, we investigated the nonlinear characteristics including complexity, approximate entropy of different frequency bands, and Lyapunov exponent of the LFPs from CA3 region of mice with TLE under ultrasound stimulation.

In humans, there is significant pyramidal cell loss in the hippocampal subfields CA1 and CA3 (DeGiorgio et al., 1992). Because ultrasound stimulation with single parameter has no specific selectivity to neurons, we speculate that it can also cause changes of nonlinear characteristics of LFPs in CA1 region of TLE mice.

There are two limitations and challenges to our study: (i) ultrasound stimulation can inhibit acute TLE seizures and modulate the nonlinear dynamics characteristic of LFPs. However, the accurate molecular mechanism by which ultrasound stimulation inhibits epileptic seizures is still

unknown. Low-intensity ultrasound stimulation can enhance the expression of brain-derived neurotrophic factors in astrocytes through activation of TrkB-Akt and Calcium-CaMK signaling pathways (Liu et al., 2017). We hypothesize that ultrasound modulates the expression of epileptic seizure-associated proteins, which may be the molecular mechanism by which ultrasound inhibits epileptic seizures, but this hypothesis needs to be verified in future work. (ii) In addition, previous studies have shown that ultrasonic parameters can cause different modulatory effects on neural activity (King et al., 2013; Yuan et al., 2016). We only studied the inhibition of single parameters in acute TLE. To find the optimal parameters for ultrasound inhibition of TLE, we will conduct multiparameter studies and identify the ultrasound parameters with the best therapeutic effect.

In summary, our study indicates that ultrasound stimulation can modulate the nonlinear dynamic characteristics of the LFP in acute TLE.

## AUTHOR CONTRIBUTIONS

YY designed and coordinated the study. All authors carried out experiment and data process, drafted the manuscript, and gave final approval for publication.

## FUNDING

This work was supported by the Key Project of Natural Science Foundation of Hebei Province (F2018203256), Fund for Scientific Instruments of the National Natural Science Foundation of China (61827811), and National Natural Science Foundation of China (61673336, 61503321).

## SUPPLEMENTARY MATERIAL

The Supplementary Material for this article can be found online at: <https://www.frontiersin.org/articles/10.3389/fnins.2019.00287/full#supplementary-material>

## REFERENCES

- Acharya, U. R., Sree, S. V., Chattopadhyay, S., Yu, W., and Ang, P. C. A. (2011). Application of recurrence quantification analysis for the automated identification of epileptic EEG signals. *Int. J. Neural Syst.* 21, 199–211. doi: 10.1142/S0129065711002808
- Alexander, B., Korb, A. S., Douglas, P. K., Cohen, M. S., Melega, W. P., Mulgaonkar, A. P., et al. (2011). A review of low-intensity focused ultrasound pulsation. *Brain Stimul.* 4, 125–136. doi: 10.1016/j.brs.2011.03.007
- Bielefeld, P., Sierra, A., Encinas, J. M., Maleticsavatic, M., Anderson, A., and Fitzsimons, C. P. (2017). A standardized protocol for stereotaxic intrahippocampal administration of Kainic acid combined with electroencephalographic seizure monitoring in mice. *Front. Neurosci.* 11:160. doi: 10.3389/fnins.2017.00160
- Bjork, M. H., Veiby, G., Engelsen, B. A., and Gilhus, N. E. (2015). Depression and anxiety during pregnancy and the postpartum period in women with epilepsy: a review of frequency, risks and recommendations for treatment. *Seizure* 28, 39–45. doi: 10.1016/j.seizure.2015.02.016
- Bos, W. J., Loddenkemper, T., and Nelson, C. A. (2017). Nonlinear EEG biomarker profiles for autism and absence epilepsy. *Neuropsychiatr. Electrophysiol.* 3:1. doi: 10.1186/s40810-017-0023-x
- DeGiorgio, C. M., Tomiyasu, U., Gott, P. S., and Treiman, D. M. (1992). Hippocampal pyramidal cell loss in human status epilepticus. *Epilepsia* 33, 23–27. doi: 10.1111/j.1528-1157.1992.tb02278.x
- Giannakakis, G., Sakkalis, V., Pediaditis, M., and Tsiknakis, M. (2014). Methods for seizure detection and prediction: an overview. *Neuromethods* 91, 131–157. doi: 10.1007/7657\_2014\_68
- Hakimova, H., Kim, S., Chu, K., Sang, K. L., Jeong, B., and Jeon, D. (2015). Ultrasound stimulation inhibits recurrent seizures and improves behavioral outcome in an experimental model of mesial temporal lobe epilepsy. *Epilepsy Behav.* 49, 26–32. doi: 10.1016/j.yebeh.2015.04.008
- Jing, H., Jianbo, G., and Principe, J. C. (2006). Analysis of biomedical signals by the lempel-Ziv complexity: the effect of finite data size. *IEEE Trans. Biomed. Eng.* 53, 2606–2609. doi: 10.1109/TBME.2006.883825

- King, R. L., Brown, J. R., Newsome, W. T., and Pauly, K. B. (2013). Effective parameters for ultrasound-induced in vivo neurostimulation. *Ultrasound Med. Biol.* 39, 312–331. doi: 10.1016/j.ultrasmedbio.2012.09.009
- Kopczynska, M., Zelek, W. M., Vespa, S., Touchard, S., Wardle, M., Loveless, S., et al. (2018). Complement system biomarkers in epilepsy. *Seizure* 60:1. doi: 10.1016/j.seizure.2018.05.016
- Landhuis, E. (2017). Ultrasound for the brain. *Nature* 551:257. doi: 10.1038/d41586-017-05479-7
- Lehnertz, K. (2008). Epilepsy and nonlinear dynamics. *J. Biol. Phys.* 34, 253–266. doi: 10.1007/s10867-008-9090-3
- Lévesque, M., Avoli, M., and Bernard, C. (2016). Animal models of temporal lobe epilepsy following systemic chemoconvulsant administration. *J. Neurosci. Methods* 260, 45–52. doi: 10.1016/j.jneumeth.2015.03.009
- Li, P., Karmakar, C., Yearwood, J., Venkatesh, S., Palaniswami, M., and Liu, C. (2018). Detection of epileptic seizure based on entropy analysis of short-term EEG. *PLoS One* 13:e0193691. doi: 10.1371/journal.pone.0193691
- Liu, S. H., Lai, Y. L., Chen, B. L., and Yang, F. Y. (2017). Ultrasound enhances the expression of brain-derived neurotrophic factor in astrocyte through activation of TrkB-Akt and calcium-CaMK signaling pathways. *Cereb. Cortex* 27, 3152–3160. doi: 10.1093/cercor/bhw169
- Mario, C., Jacques, M., and Michel, L. V. Q. (2003). Statistical assessment of nonlinear causality: application to epileptic EEG signals. *J. Neurosci. Methods* 124, 113–128. doi: 10.1016/S0165-0270(02)00367-9
- Mateo, A., Roberto, H., Daniel, A., and Daniel, A. (2006). Interpretation of the Lempel-Ziv complexity measure in the context of biomedical signal analysis. *IEEE Trans. Biomed. Eng.* 53, 2282–2288. doi: 10.1109/TBME.2006.883696
- Mei, Z., Zhao, X., Chen, H., and Chen, W. (2018). Bio-signal complexity analysis in epileptic seizure monitoring: a topic review. *Sensors* 18:E1720. doi: 10.3390/s18061720
- Min, B. K., Bystritsky, A., Jung, K. I., Fischer, K., Zhang, Y., Maeng, L. S., et al. (2011). Focused ultrasound-mediated suppression of chemically-induced acute epileptic EEG activity. *BMC Neurosci.* 12:23. doi: 10.1186/1471-2202-12-23
- Monti, G., and Meletti, S. (2015). Emotion recognition in temporal lobe epilepsy: a systematic review. *Neurosci. Biobehav. Rev.* 55, 280–293. doi: 10.1016/j.neubiorev.2015.05.009
- Noachtar, S., and Rémi, J. (2009). The role of EEG in epilepsy: a critical review. *Epilepsy Behav.* 15, 22–33. doi: 10.1016/j.yebeh.2009.02.035
- Noel, J. P., Kurela, L. A., Baum, S. H., Yu, H., Neimat, J. S., Gallagher, M. J., et al. (2017). Multisensory temporal function and EEG complexity in patients with epilepsy and psychogenic nonepileptic events. *Epilepsy Behav.* 70, 166–172. doi: 10.1016/j.yebeh.2017.02.018
- O'Connell, B. K., Gloss, D., and Devinsky, O. (2017). Cannabinoids in treatment-resistant epilepsy: a review. *Epilepsy Behav.* 70, 341–348. doi: 10.1016/j.yebeh.2016.11.012
- Osowski, S., Swiderski, B., Cichocki, A., and Rysz, A. (2007). Epileptic seizure characterization by Lyapunov exponent of EEG signal. *COMPEL Int. J. Comput. Math. Electr. Eng.* 26, 1276–1287. doi: 10.1108/03321640710823019
- Plaksin, M., Shoham, S., and Kimmel, E. (2014). Intramembrane cavitation as a predictive bio-piezoelectric mechanism for ultrasonic brain stimulation. *Phys. Rev. X* 4:011004. doi: 10.1103/PhysRevX.4.011004
- Raghu, S., Sriraam, N., Kumar, G. P., and Hegde, A. S. (2018). A novel approach for real time recognition of epileptic seizures using minimum variance modified fuzzy entropy. *IEEE Trans. Biomed. Eng.* 65, 2612–2621. doi: 10.1109/TBME.2018.2810942
- Robertson, J., Hatton, C., Emerson, E., and Baines, S. (2015). Prevalence of epilepsy among people with intellectual disabilities: a systematic review. *Seiz. Eur. J. Epilepsy* 29, 46–62. doi: 10.1016/j.seizure.2015.03.016
- Sabeti, M. (2009). Entropy and complexity measures for EEG signal classification of schizophrenic and control participants. *Artif. Intell. Med.* 47, 263–274. doi: 10.1016/j.artmed.2009.03.003
- Steinhäuser, C., Grunnet, M., and Carmignoto, G. (2016). Crucial role of astrocytes in temporal lobe epilepsy. *Neuroscience* 323, 157–169. doi: 10.1016/j.neuroscience.2014.12.047
- Tufail, Y., Matyushov, A., Baldwin, N., Tauchmann, M. L., Georges, J., Yoshihiro, A., et al. (2010). Transcranial pulsed ultrasound stimulates intact brain circuits. *Neuron* 66, 681–694. doi: 10.1016/j.neuron.2010.05.008
- Tyler, W. J. (2011). Noninvasive neuromodulation with ultrasound? A continuum mechanics hypothesis. *Neuroscientist* 17, 25–36. doi: 10.1177/1073858409348066
- Tyler, W. J., Yusuf, T., Michael, F., Tauchmann, M. L., Olson, E. J., and Cassondra, M. (2008). Remote excitation of neuronal circuits using low-intensity, low-frequency ultrasound. *PLoS One* 3:e3511. doi: 10.1371/journal.pone.0003511
- Varatharajah, Y., Berry, B., Cimbalknik, J., Kremen, V., Van Gompel, J. J., Smm, S., et al. (2018). Integrating artificial intelligence with real-time intracranial EEG monitoring to automate interictal identification of seizure onset zones in focal epilepsy. *J. Neural Eng.* 15:046035. doi: 10.1088/1741-2552/aac960
- Viale, L., Allotey, J., Cheong-See, F., Arroyo-Manzano, D., Mccorrey, D., Bagary, M., et al. (2015). Epilepsy in pregnancy and reproductive outcomes: a systematic review and meta-analysis. *Lancet* 386, 1845–1852. doi: 10.1016/S0140-6736(15)00045-8
- Wen-Chin, W., Jiang, G. J. A., Chi-Feng, C., Wen-Yu, L., Chun-Yen, L., Wang-Tso, L., et al. (2015). Complexity of multi-channel electroencephalogram signal analysis in childhood absence epilepsy. *PLoS One* 10:e0134083. doi: 10.1371/journal.pone.0134083
- Xu, Y., Nguyen, D., Mohamed, A., Carcel, C., Li, Q., Kutlubaev, M., et al. (2016). Frequency of a false positive diagnosis of epilepsy: a systematic review of observational studies. *Seizure* 41, 167–174. doi: 10.1016/j.seizure.2016.08.005
- Yang, P. F., Phipps, M. A., Newton, A. T., Chaplin, V., Gore, J. C., Caskey, C. F., et al. (2018). Neuromodulation of sensory networks in monkey brain by focused ultrasound with MRI guidance and detection. *Sci. Rep.* 8:7993. doi: 10.1038/s41598-018-26287-7
- Ye, J., Tang, S. Y., Meng, L., Li, X., Wen, X. X., Chen, S. H., et al. (2018). Ultrasonic control of neural activity through activation of the mechanosensitive channel MscL. *Nano Letter* 18, 4148–4155. doi: 10.1021/acs.nanolett.8b00935
- Yu, K., Sohrabpour, A., and He, B. (2016). Electrophysiological source imaging of brain networks perturbed by low-intensity transcranial focused ultrasound. *IEEE Trans. Biomed. Eng.* 63, 1787–1794. doi: 10.1109/TBME.2016.2591924
- Yuan, Y., Dong, Y., Hu, S., Zheng, T., Du, D., Du, J., et al. (2017). Reduced apparent diffusion coefficient in various brain areas following low-intensity transcranial ultrasound stimulation. *Front. Neurosci.* 11:562. doi: 10.3389/fnins.2017.00562
- Yuan, Y., Yan, J., Ma, Z., and Li, X. (2016). Noninvasive focused ultrasound stimulation can modulate phase-amplitude coupling between neuronal oscillations in the rat hippocampus. *Front. Neurosci.* 10:348. doi: 10.3389/fnins.2016.00348
- Yuan, Y., Zhao, Y., Jia, H., Liu, M., Hu, S., Li, Y., et al. (2018). Cortical hemodynamic responses under focused ultrasound stimulation using real-time laser speckle contrast imaging. *Front. Neurosci.* 12:269. doi: 10.3389/fnins.2018.00269

**Conflict of Interest Statement:** The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

Copyright © 2019 Li, Yang, Yan, Wang, Li and Yuan. This is an open-access article distributed under the terms of the Creative Commons Attribution License (CC BY). The use, distribution or reproduction in other forums is permitted, provided the original author(s) and the copyright owner(s) are credited and that the original publication in this journal is cited, in accordance with accepted academic practice. No use, distribution or reproduction is permitted which does not comply with these terms.