

Nonlocal Coupling Effects on a Network of Neurons

¹Ricardo Del Rio, ²Shravani Deo, ³Gabriela Jaramillo ¹Department of Chemistry, University of Houston

Objective

The objective of this research was to study how nonlocal coupling affects the formation of patterns and chimera states in networks of neurons.

Background

- Systems of neurons can exhibit excitatory and oscillatory behavior. An example of excitatory behavior is a neuron's action potential, and an example of oscillatory behavior is the firing of neurons in a periodic manner.
- We use Wiener-Rosenblueth (WR) and FitzHugh-Nagumo (FHN) models to study how nonlocal coupling (the connection between neurons) affects pattern formation in excitatory and oscillatory systems, respectively.
- We model a network of neurons with a square grid, and apply WR and FHN equations to it.
- In excitatory systems, we will explore how nonlocal coupling affects spiral wave formation. In oscillatory systems, we will determine how the Levy walk model and a two layer network affect the formation of chimeras.



Left: What is an action potential?, (Molecular Devices), https://www.moleculardevices.com/applications/patch-clampelectrophysiology/what-action-potential#gref Right: Neural oscillation, (Wikipedia), https://en.wikipedia.org/wiki/Neural oscillation

Motivation

- Experiments have found that in the brain there are mixed regions of synchronous and asynchronous neural activity. These states, called chimeras, have been found to increase efficiency in the brain [4].
- Research suggests that epileptic seizures occur when chimera states form in synchronous regions and then collapse [1].
- It is believed that stabilizing chimera states in the brain can reduce epileptic seizures.
- Here we study simple models describing network of neurons that are capable of reproducing chimera states. We explore the role of nonlocal coupling in stabilizing these patterns, with the hope that our results can guide research into epileptic seizures.

FitzHugh-Nagumo Model

- The FHN model is a simplified, dynamical, two-dimensional version of the Hodgkin-Huxley model, which models neuron behavior as an electrical circuit.
- FHN consists of a voltage-like variable, u, that exhibits excitatory behavior and a recovery variable, v, that provides a negative feedback.
- Based on the parameters, the FHN model can exhibit excitatory or oscillatory behavior. However, we will only look at the oscillatory case. This occurs when the parameter *a* in the equations below is less than 1.
- Using the Poincaré-Bendixson Theorem, we can prove that when *a* < 1, the system has a limit cycle and therefore displays oscillatory behavior.



 $\frac{dv_{ij}}{dt} = u_{ij} + a + \frac{\sigma}{N_r - 1} \sum_{(mn) \in B_r(i,j)} byx(u_{ij} - u_{mn}) + byy(v_{ij} - v_{mn})$

Schmidt, Kasimatis, Hizanidis, Provata, and Hovel, Chimera patterns in two-dimensional networks of *coupled neurons*, (2017)



Levy-Type Coupling • We will use a Levy distribution to determine the coupling strength between different neurons. • Levy distributions are heavy-tailed probability distributions use to describe the step lengths of a particle undergoing a Levy flight.



Grid on the left: nonlocal coupling in the original FHN system; grid on the right: FHN model with Levy-type coupling. For both models, N = 50, a = 0.5, $\sigma = 0.1$, $\phi = \pi/2 - 0.15$, r = 19, dt = 0.02.

²Department of Chemical Engineering, University of Houston ³Department of Mathematics, University of Houston

$\Phi_{ij}^{n+1} = \begin{cases} \Phi_{ij}^n + 1, & \text{if } 0 < \Phi_{ij}^n < \tau_e \\ 0, & \text{if } \Phi_{ij}^n = \tau_e + \tau_r \\ 0, & \text{if } \Phi_{ij}^n = 0 \text{ and } \\ 1, & \text{if } \Phi_{ij}^n = 0 \text{ and } \end{cases}$	$ \begin{aligned} &+ \tau_r \\ u_{ij}^{n+1} < h \\ u_{ij}^{n+1} &\geq h \end{aligned} I_{ij}^n = \begin{cases} 1, & \text{if } 0 < \Phi_{ij}^n < \tau_e \\ 0, & \text{if } \tau_e < \Phi_{ij}^n \leq \tau_e + \tau_r \text{ or } \Phi_{ij}^n = 0; \\ 0, & \text{if } \tau_e < \Phi_{ij}^n \leq \tau_e + \tau_r \text{ or } \Phi_{ij}^n = 0; \end{cases} $
$u_{ij}^{n+1} = gu_i^n j + \sum_{k,l} C(k,l) I_{i+k,j+l}^n$	$C(k,l) = \begin{cases} 1, & \text{if } k \le 1 \text{ and } l \le 1\\ 0, & \text{otherwise} \end{cases}$

• Early research found that people's brains exhibited spiral wave like patterns when hallucinating [8].

affect pattern formation.

coupling.



те = 5, тг = 7, g = 0, h = 3, R = 1





do not influence it as much as neighbors that are closer to it.





UNIVERSITY of FOUSTON

- [7] Klafter, Sokolov, Anomalous diffusion spreads its wings, (Physics World, 2005)
- [8] Ermentrout, Bard. "Neural networks as spatio-temporal pattern-forming systems." Reports on progress in physics 61.4 (1998): 353.

