

# Chimera States and Seizures in a Mouse Neuronal Model

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# Chimera States

A chimera state is **the coexistence of synchronous and asynchronous populations in a network of nonlocally-coupled oscillators [1, 2]**.

A mechanical example of a chimera state

# Measures

The **chimera-like index**  $\chi$  is a useful measure for how chimeric a system is.

$$\chi = \langle \sigma_{\text{chi}} \rangle_T \quad \text{where} \quad \sigma_{\text{chi}}(t) = \frac{1}{M-1} \sum_{c \in C} (r_c(t) - \langle r_c \rangle_C)^2 \quad (1)$$

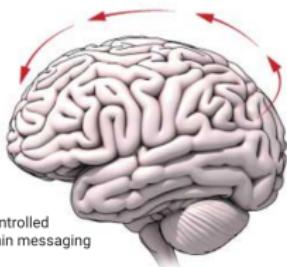
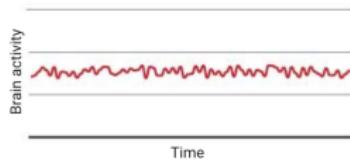
and  $r_c$  is the order parameter of community  $c$ . It is the **time average of the variance between communities of the order parameter**. Its maximum value is  $1/7$ , and will be normalized as such.

# Seizures

Seizures are **excessive(ly) synchronous neural activity** [6]. Focal seizures occur when parts of the brain seize while the rest behaves normally.

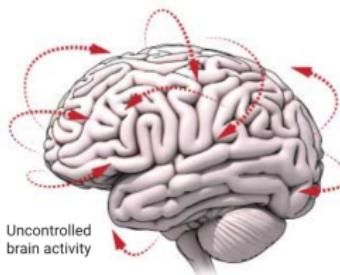
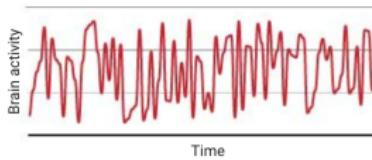
## Normal firing

Neurons relay messages to one another when electrical impulses trigger neurotransmitter release. Neurotransmitters encourage other neurons to fire or hinder their firing, keeping brain activity in check.



## Misfiring

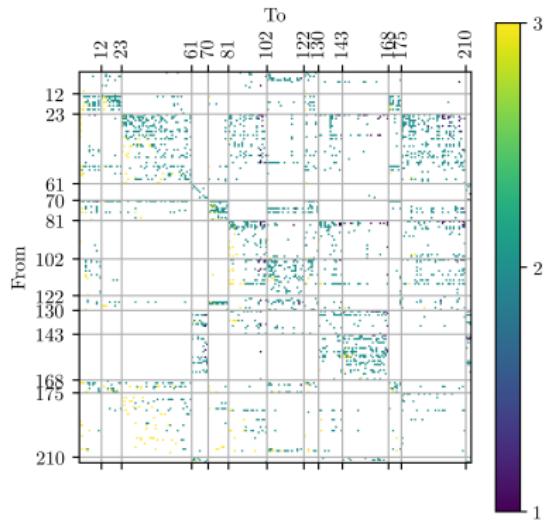
An imbalance in neurotransmitter signaling causes neurons to become overexcited or lose their inhibitory control, causing abnormal bursts of electrical activity, resulting in a seizure.



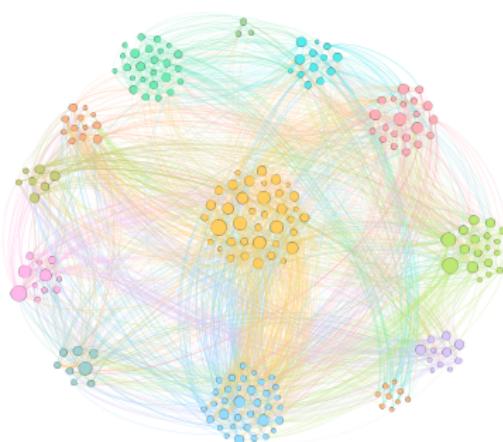
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# The Mouse Connectome

The oscillator network was modeled after the mouse connectome [4].



(a)



(b)

# Model Parameters

Symbol	Value	Meaning
$x_j$	—	<b>Membrane potential of the <math>j</math>th neural mass</b>
$y_j$	—	<b>Associated with fast intra-neural processes</b>
$z_j$	—	<b>Associated with slow intra-neural processes</b>
$b$	3.2	Tunes the spiking frequency
$I_j$	4.4	External input current
$x_{\text{rev}}$	2	Ambient reverse potential
$\lambda$	10	Sigmoidal activation function parameter
$\theta$	-0.25	Sigmoidal activation function parameter
$\mu$	0.01	Time scale for variation of $z$
$s$	4	Governs adaptation
$x_{\text{rest}}$	-1.6	Resting/equilibrium potential
$\alpha$	Varied	<b>Connection strength within cortices</b>
$n'_j$	Figure 1a	<b>Number of connections within a cortex from the <math>j</math>th neuron</b>
$G'$	Figure 1a	<b>Intra-cortical connection matrix</b>
$\beta$	Varied	<b>Connection strength between cortices</b>
$n''_j$	Figure 1a	<b>Number of connections between cortices from the <math>j</math>th neuron</b>
$G''$	Figure 1a	<b>Inter-cortical connection matrix</b>

**Table:** The list of parameters used in modeling the Hindmarsh-Rose network.

# The Model

We used a network of modified Hindmarsh-Rose neurons [5]:

$$\begin{aligned} \dot{x}_j &= y_j - x_j^3 + bx_j^2 + I_j - z_j \\ &\quad - \frac{\alpha}{n'_j} \sum_{k=1}^N G'_{jk} \Theta_j(x_k) - \frac{\beta}{n''_j} \sum_{k=1}^N G''_{jk} \Theta_j(x_k) \end{aligned} \quad (2)$$

$$\dot{y}_j = 1 - 5x_j^2 - y_j \quad (3)$$

$$\dot{z}_j = \mu(s[x_j - x_{\text{rest}}] - z_j) \quad (4)$$

where

$$\Theta_j(x_k) = \frac{x_j - x_{\text{rev}}}{1 + e^{-\lambda(x_k - \theta)}} \quad (5)$$

This sigmoidal activation function helps make the model apply to groups of neurons (subcortices) instead of single neurons.

# Simulated EEG Output

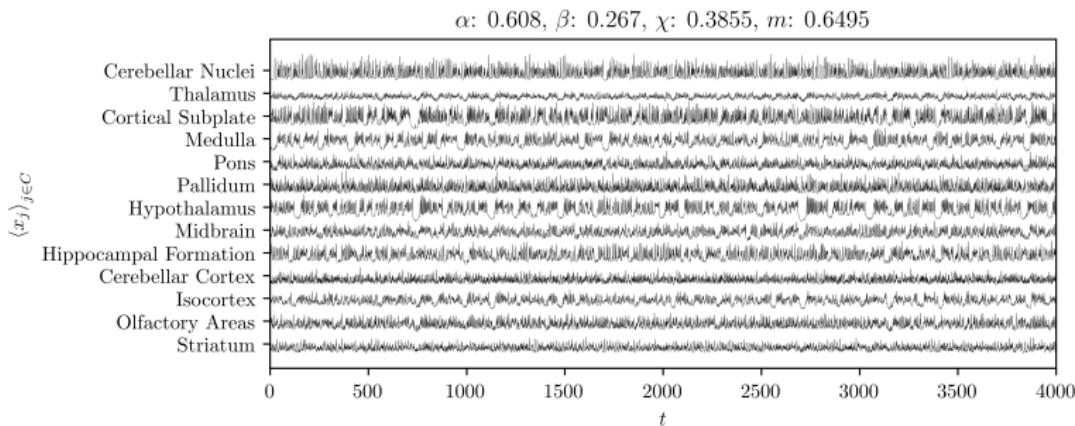


Figure: The mean membrane potential within each cortex.

Thalamus, Pons, Striatum — wild type

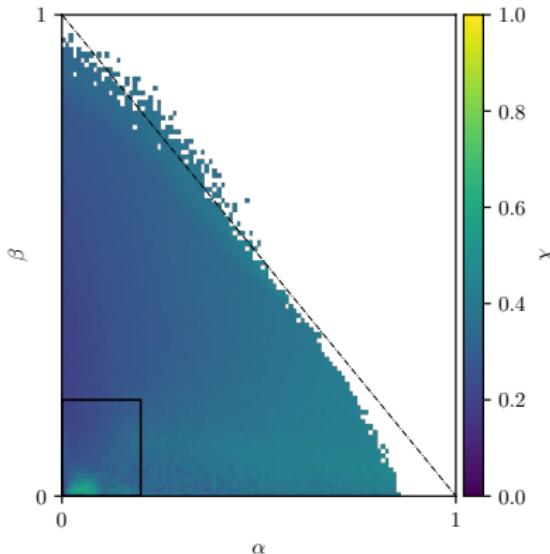
Cerebellar cortex — spiking behavior, spike runs

Medulla, Hypothalamus — repetitive spike runs

Cortical subplate — seizure-like behavior

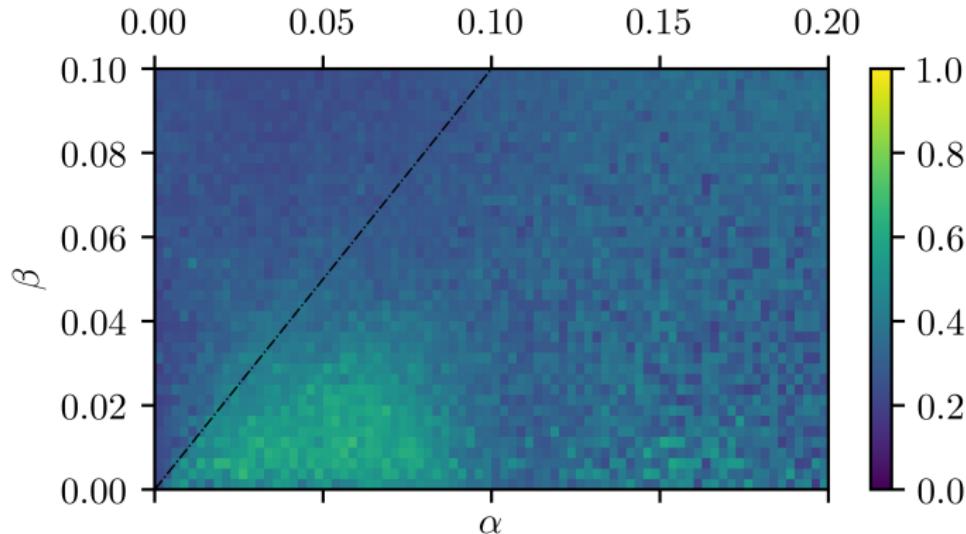
# Model Physicality

The neurons only collectively fired for certain values of  $\alpha$  (intra-cortical coupling) and  $\beta$  (inter-cortical coupling) [3].



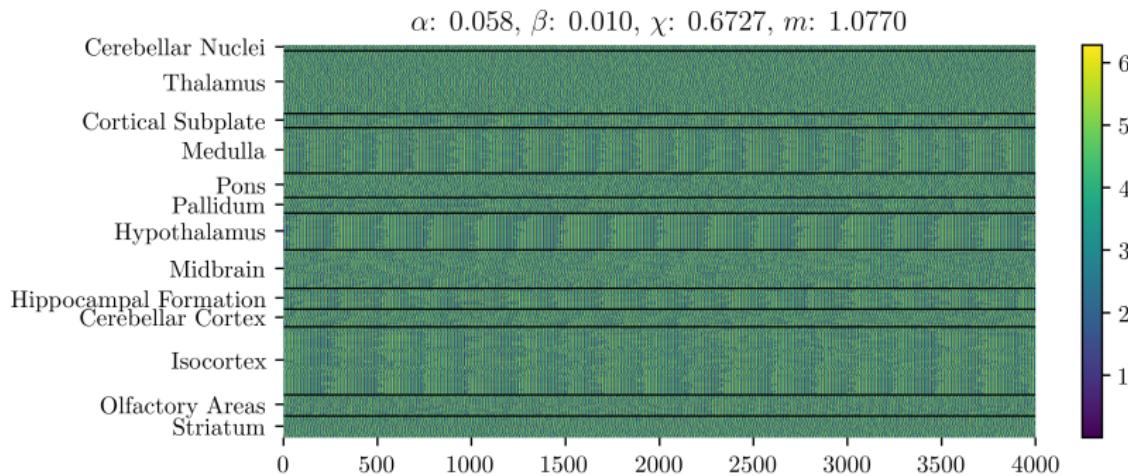
**Figure:** The chimera-like landscape of parameter space on  $(\alpha, \beta) \in (0, 0.9) \times (0, 0.9)$ . The unphysical region of the model is shown in white.

# That Highly Chimeric Patch



**Figure:** The chimera index of runs with  $(\alpha, \beta) \in (0, 0.2) \times (0, 0.1)$ . As before, the chimera-like index is normalized to  $\frac{1}{7}$ . The dashed line shows  $\beta = \alpha$ .

# Yes, It's Real



**Figure:** A run of the Hindmarsh-Rose simulation in the chimeric island. Synchronization is most consistently evident in the medulla, the hypothalamus, and the isocortex.

Chimera states...  
○○

...And Seizures...  
○

...In a Mouse Neuronal Model  
○○○

Results  
○○○●

References

# Another Viewpoint

An animation of the phase of the simulation.

# References |

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