

# Second Assignment

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## Hindmarsh-Rose model, the $e$ parameter

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This set of differential equations is a neural model with different parameters that allow to change its behaviour. In this scope, it is important to mention the  $e$  parameter, which is used to change the regime of the model from regular (Figure 1.A) to chaotic (Figure 1.B). The value selected for the regular simulation is  $e = 3.0$  and for the chaotic one is  $e = 3.281$ .

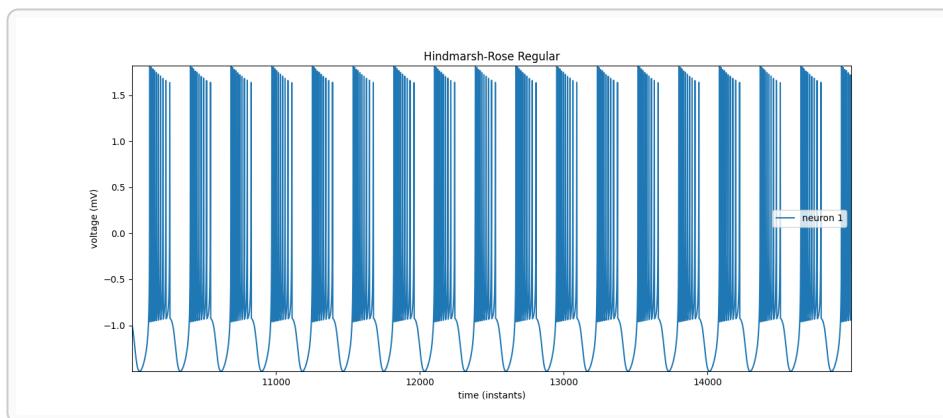


Figure 1.A: Hindmarsh-Rose model in regular regime, with  $e$  being 3.0

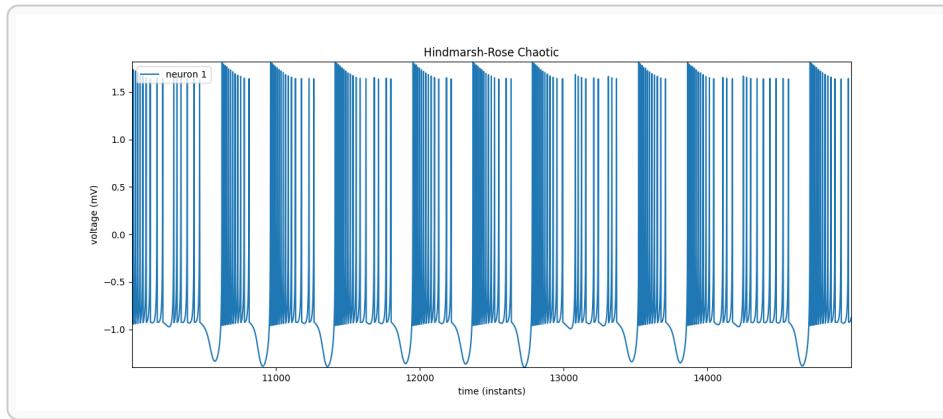


Figure 1.B: Hindmarsh-Rose model in chaotic regime, with  $e$  being 3.281

For executing this simulations, you can run this commands:

```
cargo run --release -p hindmarsh-rose-rs -- --downsample-rate 50 --goal  
20000 --runge-kutta --filename hindmarsh-rose-regular.csv  
analysis  
cargo run --release -p hindmarsh-rose-rs -- --e 3.281 --downsample-rate  
50 --goal 20000 --runge-kutta --filename hindmarsh-rose-  
chaotic.csv analysis
```

Both executions will be stored under `data/` directory inside the specified directory.

## Analyzing the chaotic regime

Because of the nature of `e` parameter, one question rises, What is the maximum value for `e` that we can assign preserving the regular regime. In other words, what is the asymptotic value that separates the chaotic regime to the regular one for `e`? The value calculated is  $e = 3.221$  and for achieving this, two methods can be used. The first one, is giving values to `e` from  $3.281$  (or the specified one in the model) to the limit value, observing if the uniform pattern is present. The other one is using the standard deviation of the ISIs (Inter Spikes Intervals) duration, assigning a tolerance that when reached, will stop the script (this one is preferred) (Figure 2).

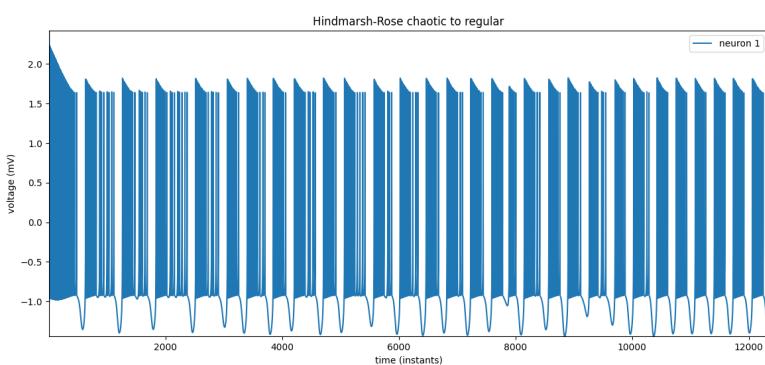


Figure 2: Hindmarsh-Rose model from chaotic to regular

For executing this simulation, you can run this commands (You will need to run each one in a terminal):

In one terminal the model:

```
cargo run --release -p hindmarsh-rose-rs -- --e 3.281 --downsample-rate  
50 --eternal-loop --write-on-pipe --runge-kutta analysis
```

In other one the analyzer, as stated previously, you can execute by setting a maximum value for `e` or using the standard deviation tolerance as stopping parameter:

```
# With max e
cargo run --release -p hindmarsh-rose-analyzer-rs -- --max-e 3.26
# With std tolerance
cargo run --release -p hindmarsh-rose-analyzer-rs -- --stop-with-
    standard-deviation -sdt 10
```

The execution will be stored in `data/hindmarsh-rose-analysis.csv`, there is also a file that stores the standard deviations (`data/hindmarsh-rose-analysis-stds.csv`) and another that keeps the times of the spike of the first depolarization and the last repolarization (`data/hindmarsh-rose-analysis-intrabursts.csv`).

## Closed loop of a neural circuit

This bidirectional synapse connection synchronizes two Hindmarsh-Rose models with chaotic activity. The synaptic models used for this interaction are both chemical and inhibitory, one fast and the other slow. The interaction shows that both synapses, despite being in a chaotic regime, regularize their activity to synchronize themselves in antiphase (Figure 3). Nevertheless, it shows overlap in some spikes due to some loss of synchronization that gets corrected with the evolution of the interaction (because of the synapses).

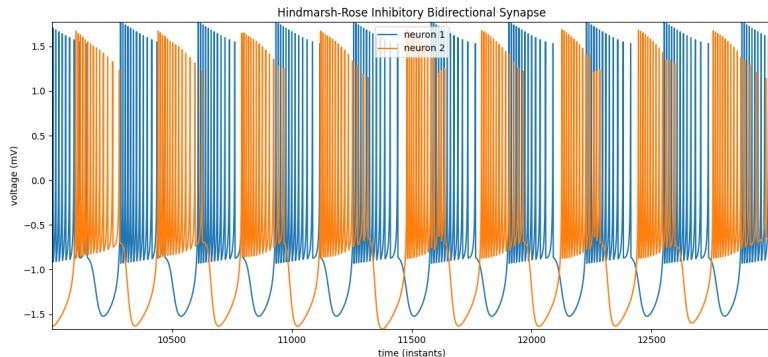


Figure 3: Two Hindmarsh-Rose models connected via two chemical and inhibitory synapse models

For executing this simulation, it is only recommended to execute the main program, because it requires a lot of pipes and it can cause some desynchronizing since they introduce some latency to the interaction. For this, execute:

```
cargo run --release
```

The execution will be stored in `data/hindmarsh-rose-bidirectional-in-chem-syn.csv`, bear in mind that the neurons names on the header are `x_pos` and `x_pre` for differencing purposes, but both neurons are presynaptic and postsynaptic (depending of the synapse connection).