

# Control of synchronization bistability in oscillatory networks

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XXXIV Dynamics Days Europe, Bayreuth, 2014



# Outline

- Motivation
- Synchronization estimation
- Synchronization bistability
- Algorithms
- Results
- Conclusion



# Motivation

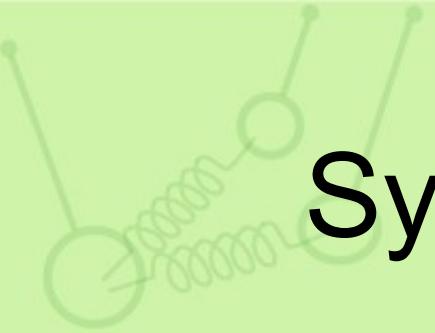
- Synchronization – widely observed phenomena
- Pathological synchronization – symptoms of neurological diseases
- Synchronized state – may be not uniquely stable
- Desynchronization methods:
  - I) open loop (e.g. coordinates reset, high frequency stimulation)
  - II) closed loop (e.g. PID, delayed feedback, act-and-wait)



# Synchronization estimation

If oscillator have uniquely predefined phase, then synchronization is estimated by the ***order parameter***:

$$r_1 = \frac{1}{N} \sum_{j=1}^N e^{i\theta_j}$$

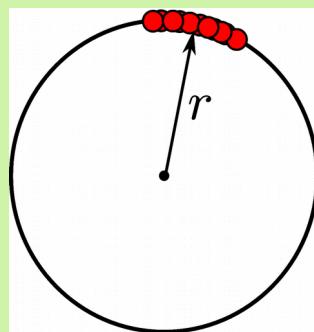


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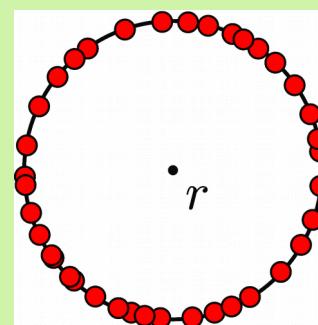
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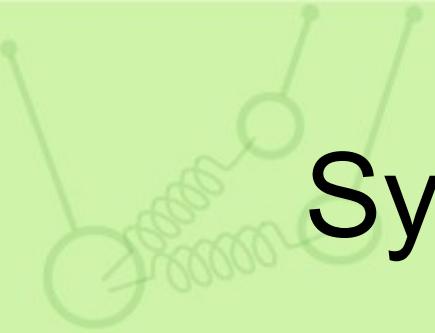
$$r_1 = \frac{1}{N} \sum_{j=1}^N e^{i\theta_j}$$

$|r_1| = 1$   
synchronized state



$|r_1| = 0$   
desynchronized state



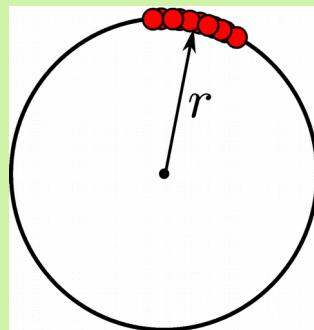


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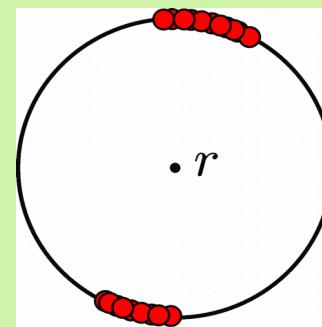
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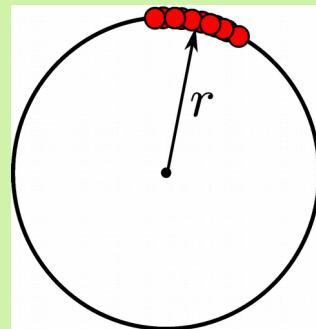
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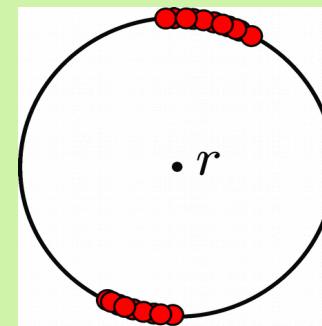
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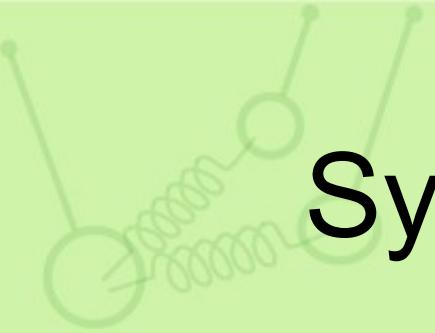
$$r_n = \frac{1}{N} \sum_{j=1}^N e^{in\theta_j}$$

$|r_1| = 1$   
synchronized state



$|r_1| = 0$   
desynchronized state





# Synchronization estimation

In real life to define separate oscillators phase in coupled network is impossible.

Neuron network models have shape

Oscill. Eqs.

$$\begin{aligned}\frac{dx_i}{dt} &= F(x_i, \mathbf{y}_i) + W(x_1, \dots, x_N), \\ \frac{d\mathbf{y}_i}{dt} &= \mathbf{G}(x_i, \mathbf{y}_i).\end{aligned}$$

Coupling

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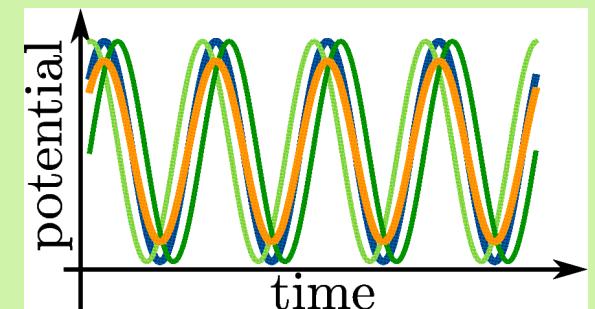
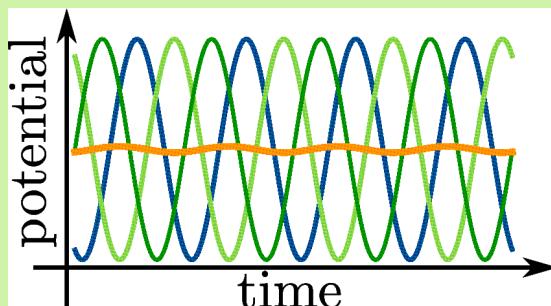
Coupling

Synchronization may be estimated by the **mean potential field variation**:

$S \approx 0$   
desynchronized state

$$S = \text{Var} \left[ \frac{1}{N} \sum_{i=1}^N x_i \right]$$

$S \approx \text{Var}[x_i]$   
synchronized state





# Synchronization bistability

Incoherent and partially synchronized stable states coexist at a particular range of parameters.



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## Example:

“Kuramoto Model of Coupled Oscillators with Positive and Negative Coupling Parameters: An Example of Conformist and Contrarian Oscillators” by H. Hong and S. H. Strogatz (PRL, 2011)

$$\dot{\theta}_j = \omega_j + \frac{K_1}{N} \sum_{k=1}^{N_1} \sin(\theta_k - \theta_j)$$

conformists

$$- \frac{K_2}{N} \sum_{k=N_1+1}^N \sin(\theta_k - \theta_j)$$

contrarians

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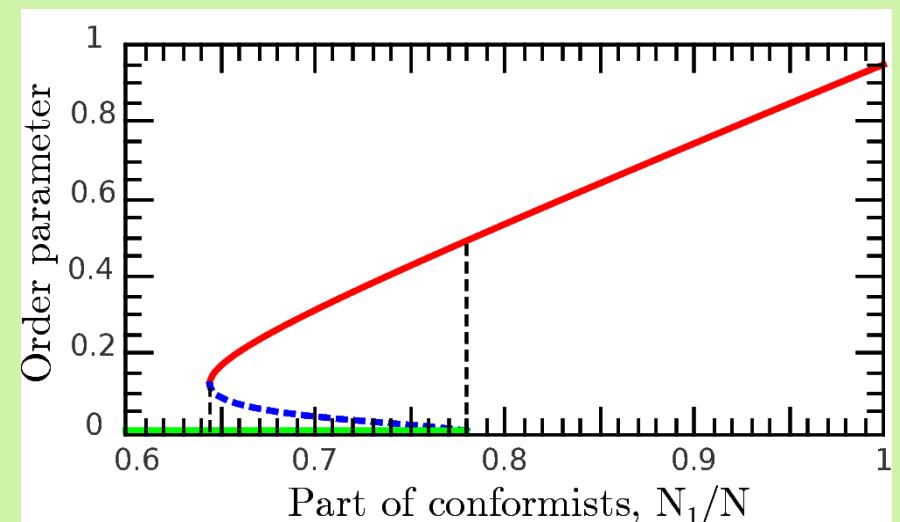
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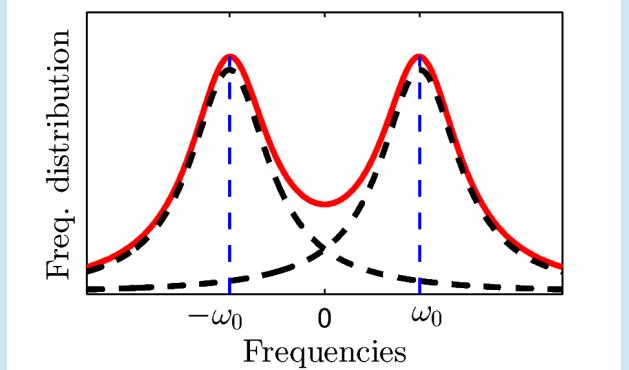
# Synchronization bistability

Other examples for phase oscillators:

- Bimodal frequencies distributions<sup>1</sup>
- Kuramoto-Sakaguchi model<sup>2</sup>
- Scale free network<sup>3</sup>

$$\dot{\theta}_j = \omega_j + \frac{K}{N} \sum_{k=1}^N \sin(\theta_k - \theta_j)$$

Frequencies distributed by the sum of two Lorentz distr.



[1] E. Martens et al. , “Exact results for the Kuramoto model with a bimodal frequency distribution”, Phys. Rev. E, (2009)

[2] O. E. Omel'chenko and M. Wolfrum, "Nonuniversal transitions to synchrony in the Sakaguchi-Kuramoto model", Phys. Rev. Lett., (2012)

[3] J. G. Gardenes et al., “Explosive Synchronization Transitions in Scale-Free Networks”, Phys. Rev. Lett. (2011)



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- Scale free network<sup>3</sup>

$$\dot{\theta}_j = \omega_j + \frac{K}{N} \sum_{k=1}^N \sin(\theta_k - \theta_j + \alpha)$$

Some specific unimodal frequencies distributions

[1] E. Martens et al. , “Exact results for the Kuramoto model with a bimodal frequency distribution”, Phys. Rev. E, (2009)

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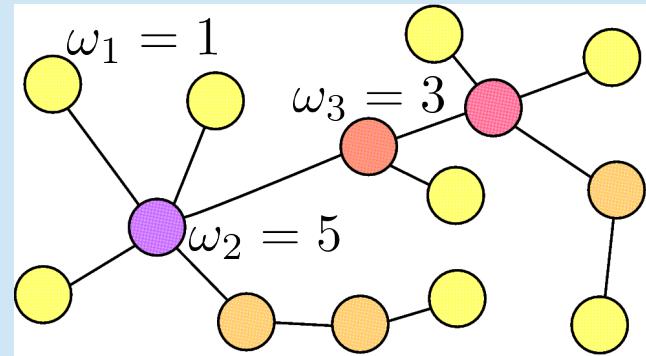
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Probability to be attached to  $k$  oscillators  $\sim k^{-\gamma}$



$$\dot{\theta}_j = \omega_j + K \sum_{k=1}^N A_{ij} \sin(\theta_k - \theta_j)$$

$\omega_j \sim \#$  attached oscillators

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# Synchronization bistability

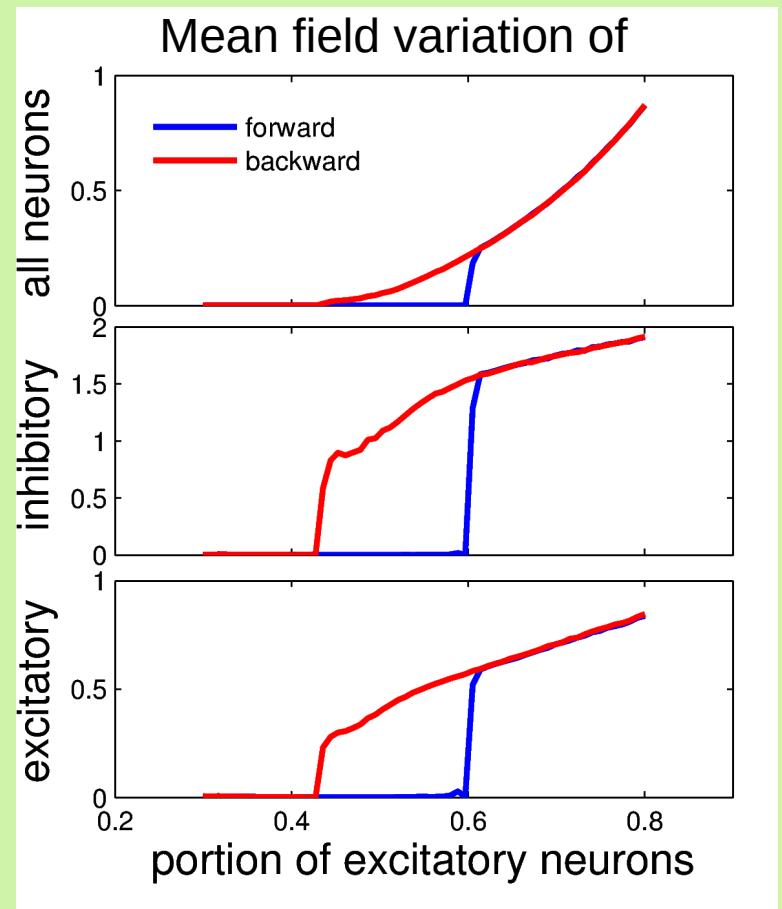
FitzHugh-Nagumo synaptically coupled neurons [“conformists-contrarian” analogue]:

$$\begin{aligned}\dot{v}_j &= v_j(1 - v_j^2/3) - u_j + I + I_j^{(syn)}, \\ \dot{u}_j &= \varepsilon_j(b_0 + b_1 v_j - u_j),\end{aligned}$$

$$I_j^{(syn)} = (v_j^{(0)} - v_j) \frac{K_j}{N} \sum_k \frac{1}{1 + \exp(-(v_k - v_T)/\Delta)}$$

$k$ 'th neuron acts on  $j$ 'th only when exceeds some threshold.

	Excitatory	Inhibitory
$K_j$	0.425	1.275
$v_j^{(0)}$	2.5	-2.5

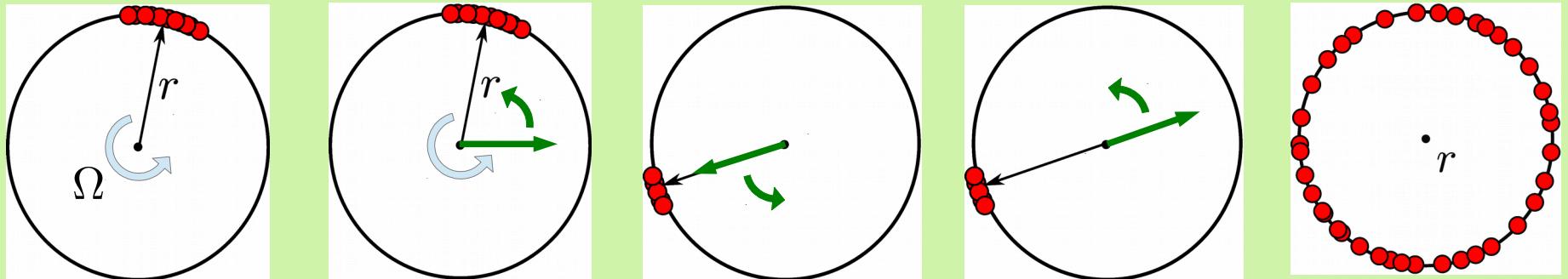




# Algorithms

- Systems synchronization and phase reversion by  $\pi$  with external periodic force

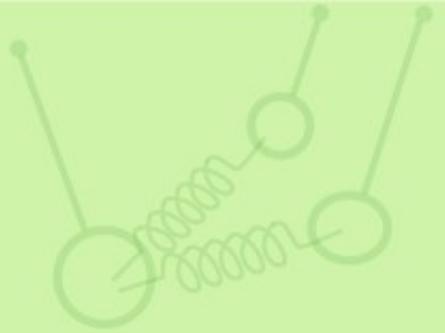
$$F(t) = a \cos(\Omega t + \phi(t))$$



The moment of external force disconnection should be determined empirically.

- Stimulate system with high frequency periodic signal with decaying amplitude

$$F(t) = a \exp(-t/\tau) \cos(\bar{\Omega}t)$$



# Results

“Conformists-contrarians” model [phase change]:

$$\dot{\theta}_j = \omega_j + \frac{1}{N} \sum_{k=1}^N K_k \sin(\theta_k - \theta_j) + a \sin(\Omega t - \theta_j)$$

System:

Coupling  $K_k \in \{-3, 1\}$

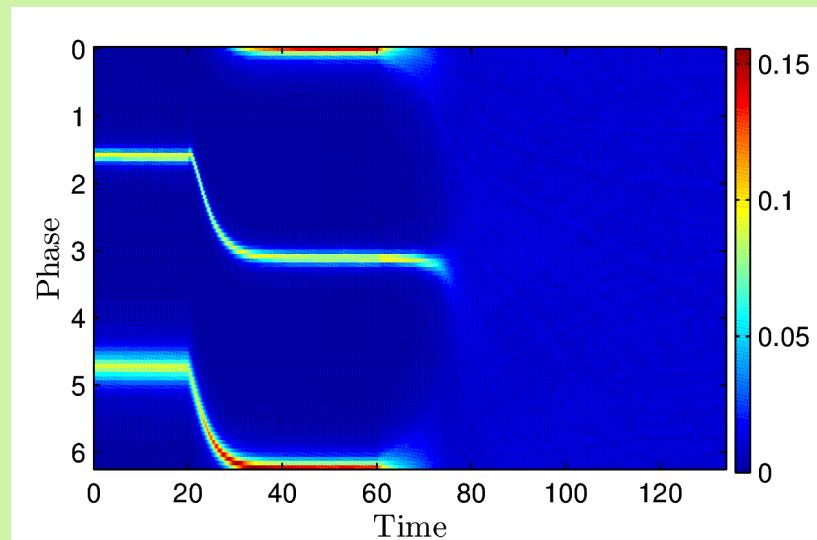
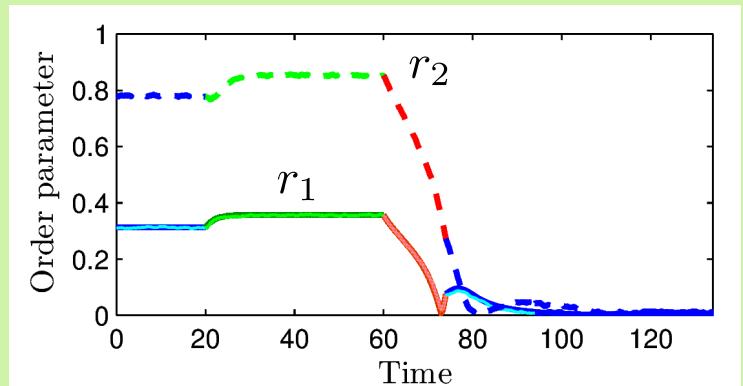
Number of oscillators  $N = 10000$

Central frequency  $\omega_0 = 0$

Force parameters:

Amplitude  $a = 0.2$

Frequency  $\Omega = 0$



# Results

“Conformists-contrarians” model:

$$\begin{aligned}\dot{\theta}_j = & \omega_j + \frac{1}{N} \sum_{k=1}^N K_k \sin(\theta_k - \theta_j) \\ & + a \exp(-t/\tau) \sin(\Omega t - \theta_j)\end{aligned}$$

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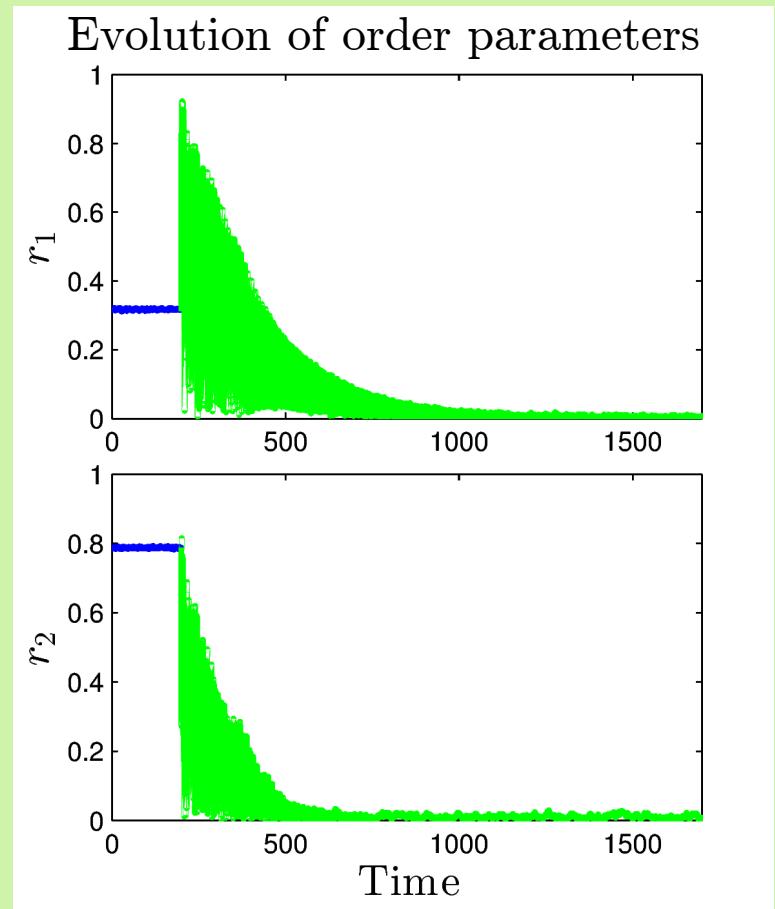
Central frequency  $\omega_0 = 0.3$

Force parameters:

Amplitude  $a = 4$

Frequency  $\Omega = 2$

Decay time  $\tau = 200$





# Results

FitzHugh-Nagumo “conformists-contrarian” model:

Number of neurons:  $N = 20000$

Mean field frequency:  $\omega_0 \approx 2\pi/75$

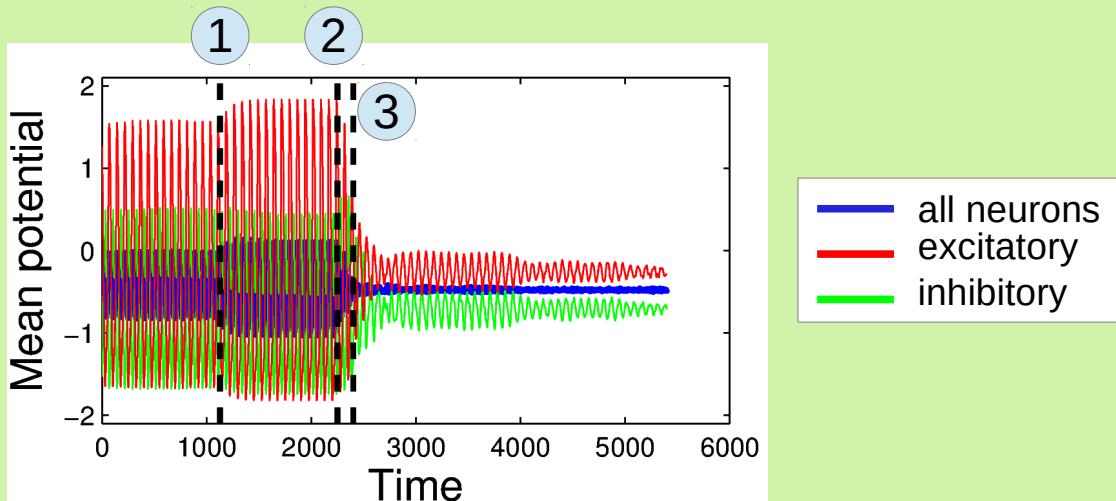
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FitzHugh-Nagumo “conformists-contrarian” model:

Number of neurons:  $N = 20000$

Mean field frequency:  $\omega_0 \approx 2\pi/75$

Phase change:  $F(t) = a \cos(\Omega t + \phi(t))$



External force parameters:

$$a = 0.05, \Omega = 2\pi/75$$

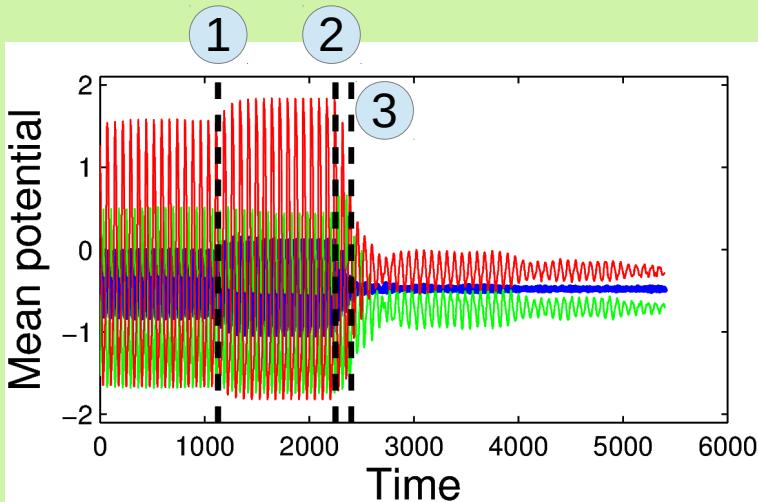
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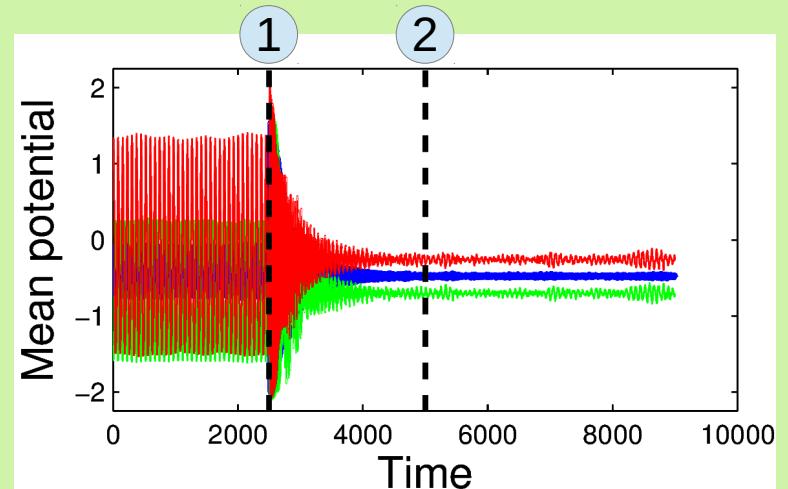
Number of neurons:  $N = 20000$

Mean field frequency:  $\omega_0 \approx 2\pi/75$

Phase change:  $F(t) = a \cos(\Omega t + \phi(t))$



Decaying periodic force:  $F(t) = a \exp(-t/\tau) \cos(\Omega t)$



External force parameters:

$$a = 0.05, \Omega = 2\pi/75$$

External force parameters:

$$a = 1.5, \Omega = 2\pi/10, \tau = 500$$



# Conclusion

- Proposed algorithms are able to drive synchronized bistable systems to desynchronized state

Further work and unanswered questions:

- How frequent is investigated systems in the nature?
- How to improve algorithms stability? (dependence on number of oscillators, noise, ...)
- How to estimate control parameters?
- etc...

Thank you for attention!

## Acknowledgments

This research was funded by the European Social Fund under the Global Grant measure (grant No. VP1-3.1-SMM-07-K-01-025)



# The end

