



From quasiperiodic partial synchronization to collective chaos in networks of inhibitory neurons

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Introduction

- Brain oscillations, Hans Berger 1929 (EEG)
- Display a broad range of frequencies
- Correlated with sleep stages & tasks
- The reflect some coordination of spike discharges in large ensembles of neurons
- Inhibition largely involved, particularly in "fast oscillations" (>30Hz)
- Computational models:
 - Inhibition + Synaptic Delays

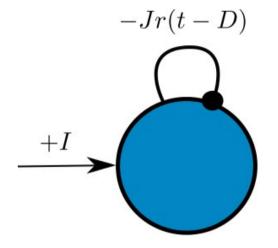
Fast oscillations in *Mean Field models* Heuristic Firing Rate Models

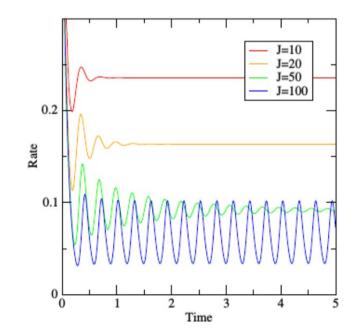
$$\tau \dot{r} = -r + \Phi(-Jr(t-D) + I)$$

- r(t): Firing rate (at time t)
- $\Phi(I)$: Transfer function (f-I curve)
- -Jr(t-D): Time delayed, inhibitory synaptic current
- *I*: External currents

Linear Stability analysis

- Fixed point $r^* = \Phi(Jr^* + I)$
- Characteristic equation $\tau \lambda = -1 + Je^{-\lambda D}$
- Hopf (supercritical): $\tan(\Omega_c D) = -(\tau/D)\Omega_c D$
- $T_c = \frac{2\pi}{\Omega_c} \in (2D, 4D)$
- $D \sim 5 \text{ms} \rightarrow T_c \in (10, 20) \text{ms}$: Fast Oscillations





Roxin, Brunel, Hansel, PRL (2005); Brunel, Hakim, Chaos (2008); Roxin, Montbrió, Phys D (2011)

Fast oscillations in spiking neuron models

• In many cases, neurons do not fire at the freq. of the mean field.

Dichotomy btw . Macroscopic & Microscopic dynamics

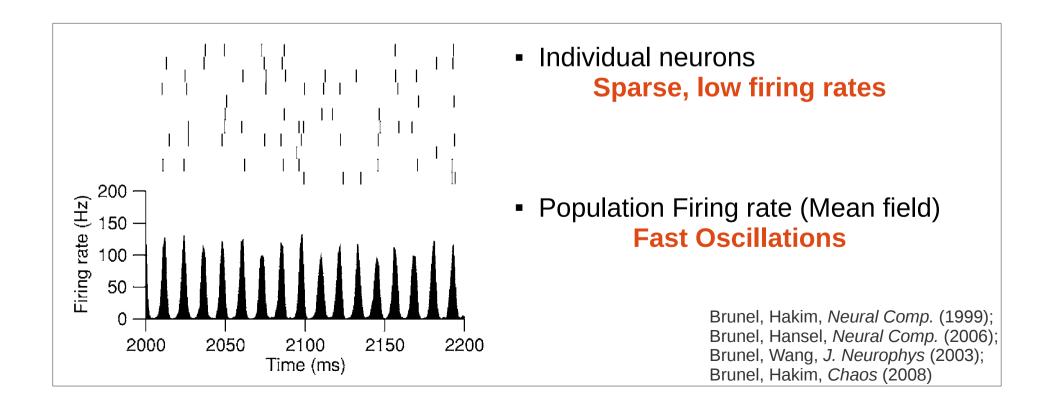
• Sharp contrast w. Collective Synchronization Winfree J.

Theor Biol. 1967, Kuramoto 1975, 1984.

• Different macroscopic, self-organized state: Sparse Synchronization Brunel & Hakim 1999

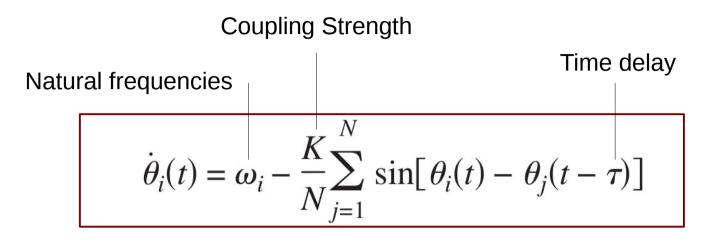
Sparse Synchronization

- Networks of **non-oscillatory**, spiking neurons
- Strongly driven by **noise**
- Inhibition
- Synaptic delays (fixed and/or synaptic kinetics)

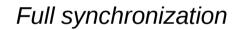


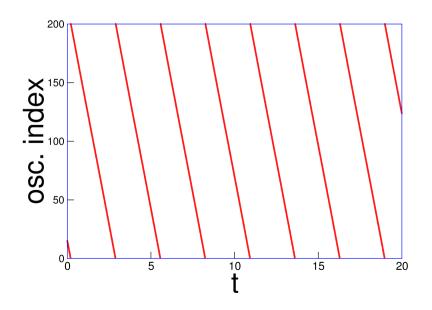
- Sparse Sync and Fast osc. in Heuristic FRM are assumed to be "the same state"
- HFRM are not derived from networks of spiking neurons though...
- Is there an exact link btw. Fast Oscillations in FRM (Inhib+Delay) and some state (not collective sync) in networks of oscillators?
- Let's look at the Kuramoto model with delay...

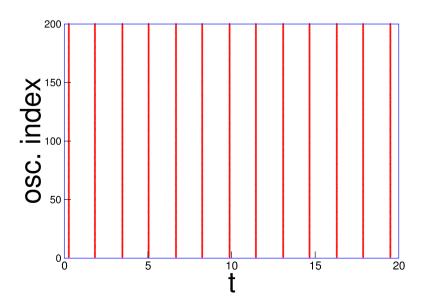
Kuramoto model with time delay



Incoherence Asynchronous (splay) state

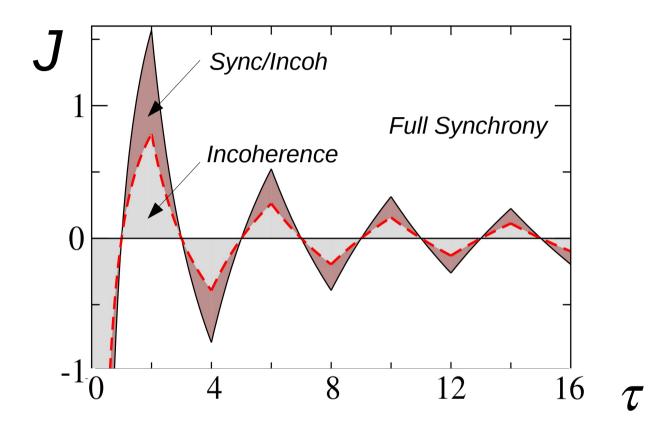






Yeung, Strogatz, Phys Rev Lett (1999)

Phase Diagram of the KM with delay



Not OK for modeling Fast Osc. in inhibitory networks:

- Only collective sync (also clustering)
- Same dynamics for **Excitation** and **Inhibition**

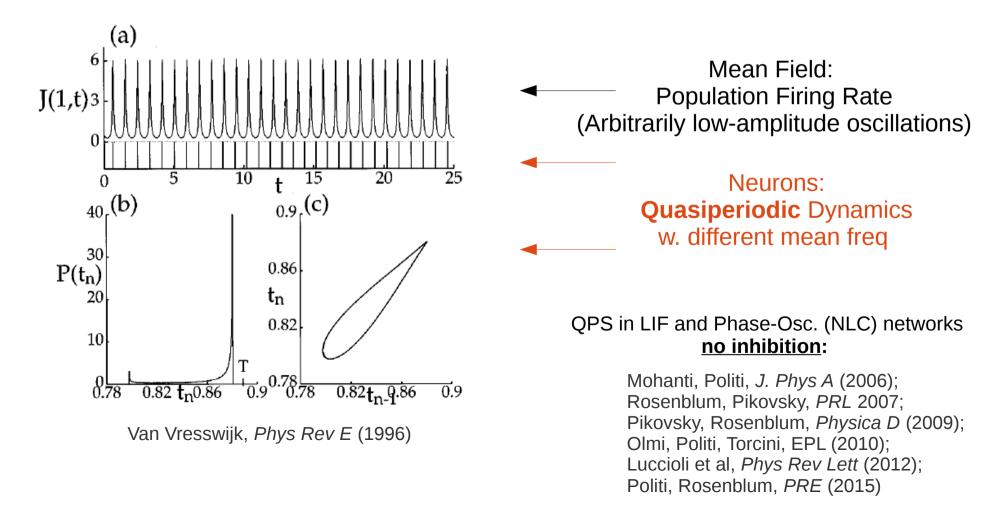
Potential candidates in oscillatory networks...

• Quasiperiodic Partial Synchronization

Collective Chaos

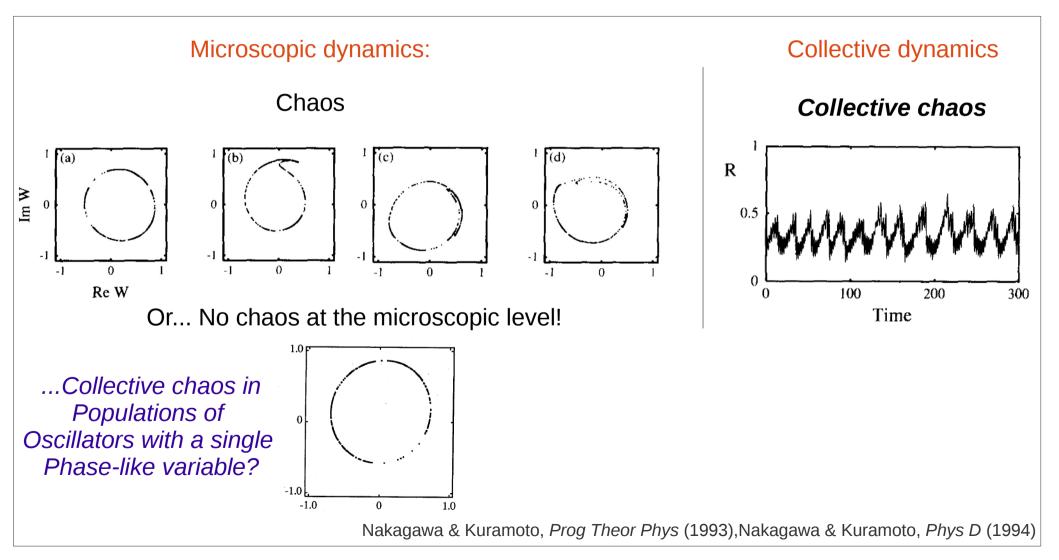
Quasiperiodic Partial Synchronization (QPS)

- Networks of Identical + Oscillatory + Excitatory LIF neurons
- Global coupling w. synaptic kinetics (alpha synapses)



Collective Chaos

Globally-coupled, Identical Limit-Cycle Oscillators (Landau-Stuart)



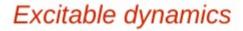
Hakim, Rappel *PRA* (1992); Takeuchi et al. *PRL* (2009,2011), Olmi, Politi, Torcini, *EPL* (2010); Ku, Girvan, Ott, *Chaos* 2015, Rosenblum, Pikovsky *PRE* (2015)

Derivation of a FRM with Inhibition + Delays (Not Heuristic)

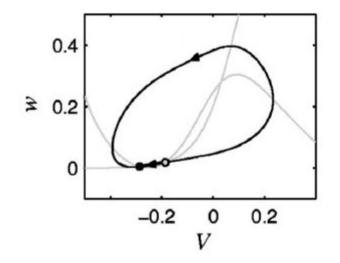
Montbrió, Pazó and Roxin, PRX 2015 Pazó, Montbrió PRL 2016

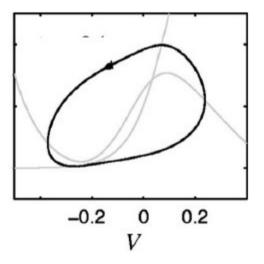
Spiking neurons Quadratic Integrate & Fire model (QIF)

The QIF model is the normal form of a SNIC bifurcation



Oscillatory dynamics



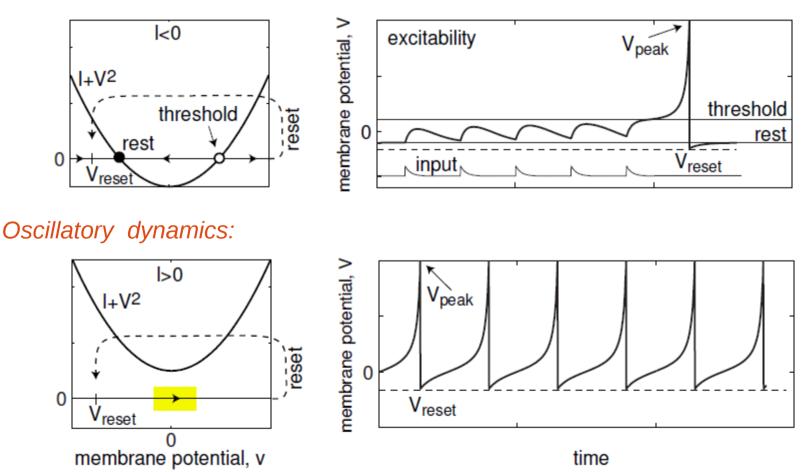


Ermentrout, Kopell, SIAM 1986; Latham et al. J Neurophys. 2000; Izhikevich, 2007

Dynamics of the QIF model

$$V = I + V^2$$
, if $V \ge V_{\text{peak}}$, then $V \leftarrow V_{\text{reset}}$

Excitable dynamics:



E. Izhikevich, "Dynamical Systems in Neuroscience", 2007

Ensemble of recurrently coupled QIF neurons with synaptic time delay

$$\tau \dot{V}_j = V_j^2 + I_j,$$
$$I_j = \eta_j + J s_D,$$

- Coupling: *J*>0: Excitation; *J*<0: Inhibition
- Mean synaptic activity ($s_D = s(t-D)$): $s_D = \frac{\tau}{N\tau_s} \sum_{i=1}^{N} \sum_{k} \int_{t-D-\tau_s}^{t-D} \delta(t'-t_j^k) dt'.$
- Fast synapses (τ_s ->0): $s_D = \tau r_D$

Time delayed, Population-Averaged Firing Rate

Thermodynamic limit Continuous formulation

$ ho(V \eta,t)dV$	Fraction of neurons with V between V and $V+dV$
	and parameter η at time <i>t</i>

 $g(\eta)$ PDF of the currents η

The Continuity Equation is

$$\partial_t \rho + \partial_V \left[(V^2 + \eta + Js + I)\rho \right] = 0$$

For each value of η !! Then the total density at time *t* is given by: $\int_{-\infty}^{\infty} \rho(V|\eta, t) g(\eta) d\eta$

Stationary solutions

 $-\partial_t \rho = \partial_V \left(\rho [V^2 + \eta] \right)$

• If
$$\eta > 0$$
: $\rho(V|\eta) = \frac{C(\eta)}{V^2 + \eta}$

• If $\eta \leq 0$: $\rho(V|\eta) = \delta(V - \tilde{C}(\eta))$

Lorentzian Ansatz

$$\rho(V|\eta) = \frac{1}{\pi} \frac{(x(\eta))}{(V-y(\eta))^2 + x(\eta)^2}$$

General solutions?

• Lorentzian Ansatz:
$$\rho = \frac{1}{\pi} \frac{1}{(V-y)^2 + x^2}$$

• Continuity Eq: $-\partial_t \rho = \partial_V \left(\rho [V^2 + \eta] \right)$

We substitute the LA into the continuity eq

•
$$\partial_t \rho = \frac{1}{\pi} \frac{1}{((V-y)^2 + x^2)^2} \left(\dot{x} [(V-y)^2 + x^2] - x [2x\dot{x} - 2\dot{y}(V-y)] \right)$$

•
$$\partial_V \left(\rho [V^2 + \eta] \right) = \frac{-2(V-y)x}{\pi ((V-y)^2 + x^2)^2} [V^2 + \eta] + \frac{2Vx}{\pi ((V-y)^2 + x^2)}$$

Equating the expressions

$$-\dot{x}\left((V-y)^2 + x^2\right) + 2x\left(\dot{x} - \dot{y}(V-y)\right) = -2(V-y)x[V^2+\eta] + 2Vx\left[(V-y)^2 + x^2\right]$$

The identity must hold at all orders!!

•
$$O(V^2)$$
: $\dot{x} = 2xy$

•
$$O(V): \dot{y} = y^2 - x^2 + \eta$$

• O(1): Linear combination of previous equations

Dynamics in the Lorentzian manifold

$$\rho(V|\eta,t) = \frac{1}{\pi} \frac{x(\eta,t)}{[V-y(\eta,t)]^2 + x(\eta,t)^2} \qquad \partial_t \rho + \partial_V \left[(V^2 + \eta + Js + I)\rho \right] = 0$$
Lorentzian ansatz
$$Continuity equation$$

$$w(\eta,t) \equiv x(\eta,t) + iy(\eta,t)$$

$$\partial_t w(\eta,t) = i \left[\eta + Js(t) - w(\eta,t)^2 + I(t) \right]$$

$$s(t) = r(t) : \text{Fast Synapses}$$

Closing this equation requires to express w as a function of r and some other meaningful macroscopic observables

Lorentzian Ansatz Firing Rate & Mean Membrane potential

Firing Rate = Prob flux at threshold: $r(\eta, t) = \rho(V \to \infty | \eta, t) \dot{V}(V \to \infty | \eta, t)$

Firing Rate

$$x(\eta, t) = \pi r(\eta, t) \qquad r(t) = \frac{1}{\pi} \int_{-\infty}^{\infty} x(\eta, t) g(\eta) d\eta$$

Mean Membrane potential

$$y(\eta, t) = \text{P.V.} \int_{-\infty}^{\infty} \rho(V|\eta, t) V \, dV$$

$$v(t) = \int_{-\infty}^{\infty} y(\eta, t) g(\eta) d\eta$$

Firing Rate Model

Lorentzian distribution of currents

$$g(\eta) = \frac{1}{\pi} \frac{\Delta}{(\eta - \bar{\eta})^2 + \Delta^2}$$

Cauchy Residue's theorem to solve

$$r(t) = \frac{1}{\pi} \int_{-\infty}^{\infty} x(\eta, t) g(\eta) d\eta$$

$$egin{aligned} & au \dot{r} = rac{\Delta}{\pi au} + 2rv, \ & au \dot{v} = v^2 + ar{\eta} + J au r_D - au^2 \pi^2 r^2, \end{aligned}$$

Montbrió, Pazó and Roxin, *PRX 2015* Pazó, Montbrió *PRL 2016* Ott, Antonsen, *Chaos 2008*

Linear Stability Analysis of Incoherence

$$egin{aligned} & au \dot{r} = rac{\Delta}{\pi au} + 2rv, \ & au \dot{v} = v^2 + ar{\eta} + J au r_D - au^2 \pi^2 r^2, \end{aligned}$$

 $\tau=\eta=1$, without loss of generality

For identical neurons, the only fixed point is:

$$\left((J+\sqrt{J^2+4\pi^2})/(2\pi^2),0\right)$$

Incoherent state (splay state)

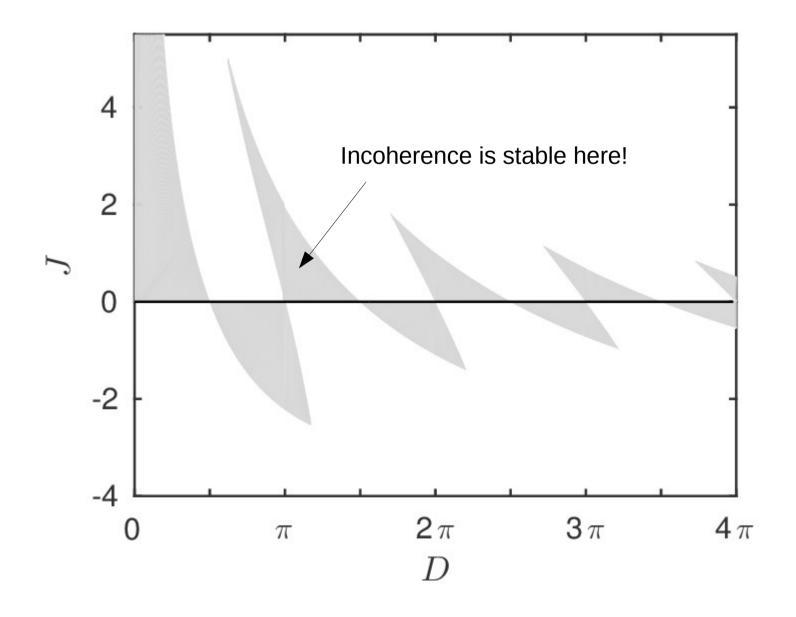
Linearizing around the f.p. and imposing the cond. of marginal stab: λ = i Ω

Hopf boundaries:

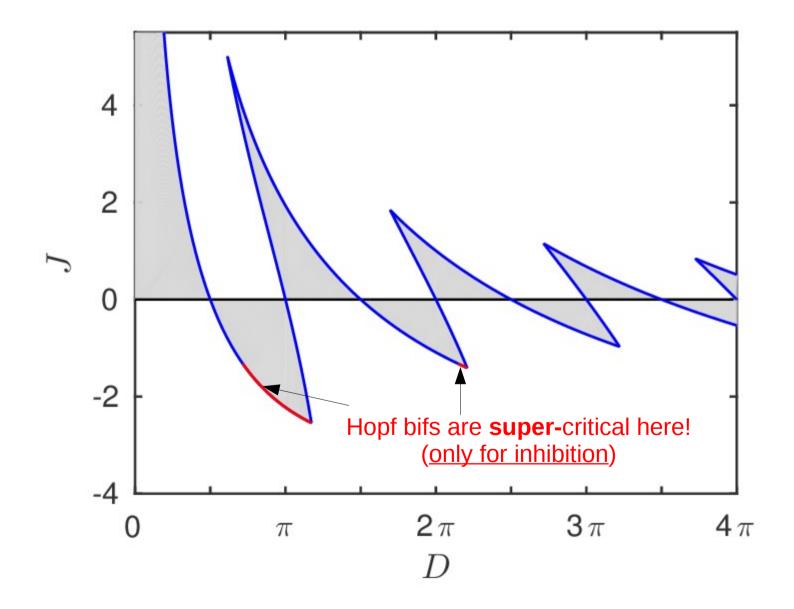
$$\Omega_n = n\pi/D.$$

$$J_{H}^{(n)} = \pi (\Omega_{n}^{2} - 4) \times \begin{cases} (6\Omega_{n}^{2} + 12)^{-1/2} & \text{for odd } n \\ (2\Omega_{n}^{2} - 4)^{-1/2} & \text{for even } n \end{cases}$$

Incoherence (Hopf) boundaries



Weakly nonlinear analysis (two timing)



Stability of Sync QIF ↔ Phase models

$$\tau \dot{V}_j = V_j^2 + \eta_j + J\tau r_D$$

When: $V_{\text{peak}} = -V_{\text{reset}} \rightarrow \text{infty}$:

- Inter-spike Interval self-oscillatory neurons ($\eta > 0, J=0$): ISI = $\pi \tau / \sqrt{\eta_j}$.
- Winfree Model (identical, self-oscillatory neurons): $V_j = \sqrt{\eta} \tan\left(\frac{\psi_j}{2}\right)$,

$$\tau \dot{\psi}_j = 2\sqrt{\eta} + (1 + \cos \psi_j) \frac{J}{\sqrt{\eta}} \tau r_D.$$

• Theta-Neurons: $V_j= an(heta_j/2)$ Ermentrout and Kopell,SIAM J Appl Math 1986

$$\tau \dot{\theta}_j = (1 - \cos \theta_j) + (1 + \cos \theta_j) \left[J \tau r_D + \eta_j \right]$$

Linear stability analysis of Sync

Winfree (QIF) model:

$$\tau \dot{\psi}_j = 2\sqrt{\eta} + (1 + \cos\psi_j) \frac{J}{\sqrt{\eta}} \tau r_D.$$

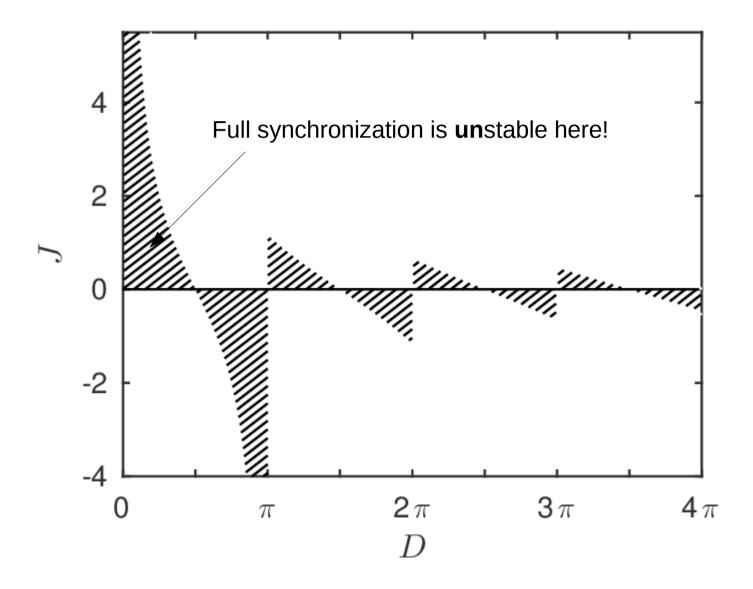
We find the boundaries:

$$J_c^{(m)} = 2 \cot\left(\frac{D}{m}\right), \quad \text{with } m = 1, 3, 5, \dots$$

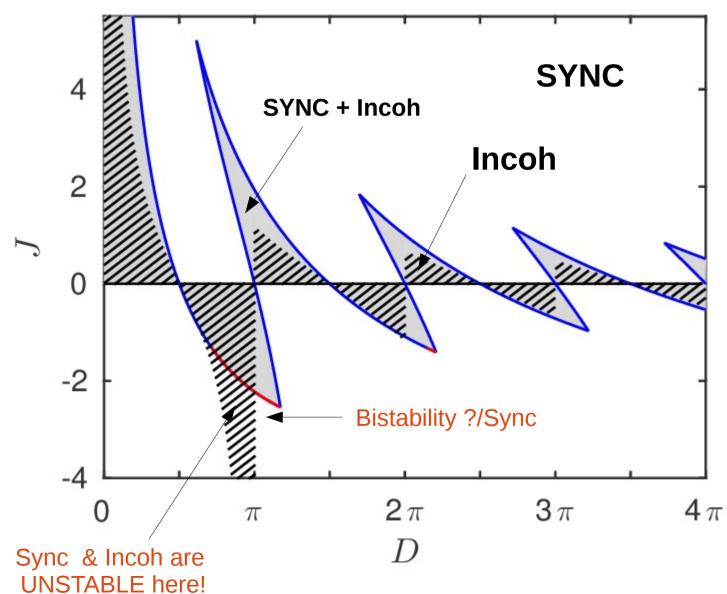
And:

$$D = n\pi$$
 ($n = 1, 2...$).

Synchronization boundaries

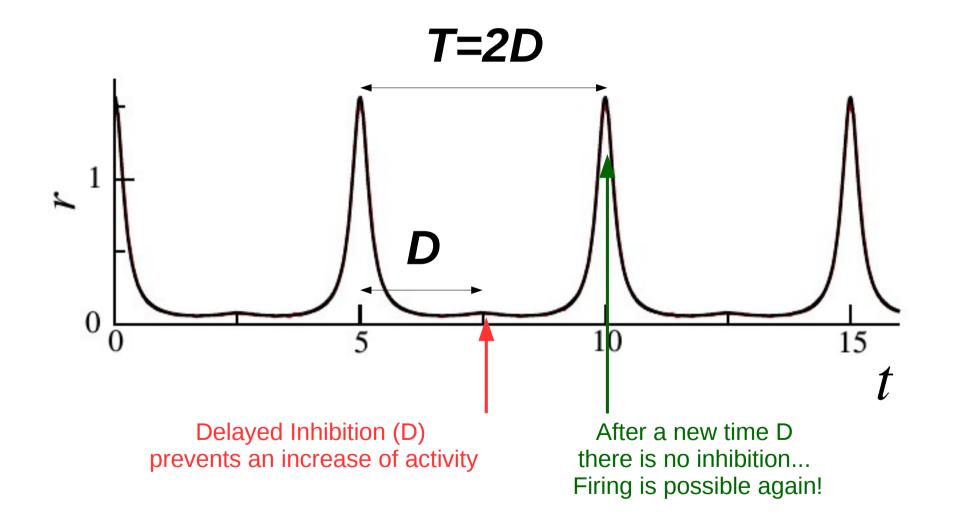


Phase diagram

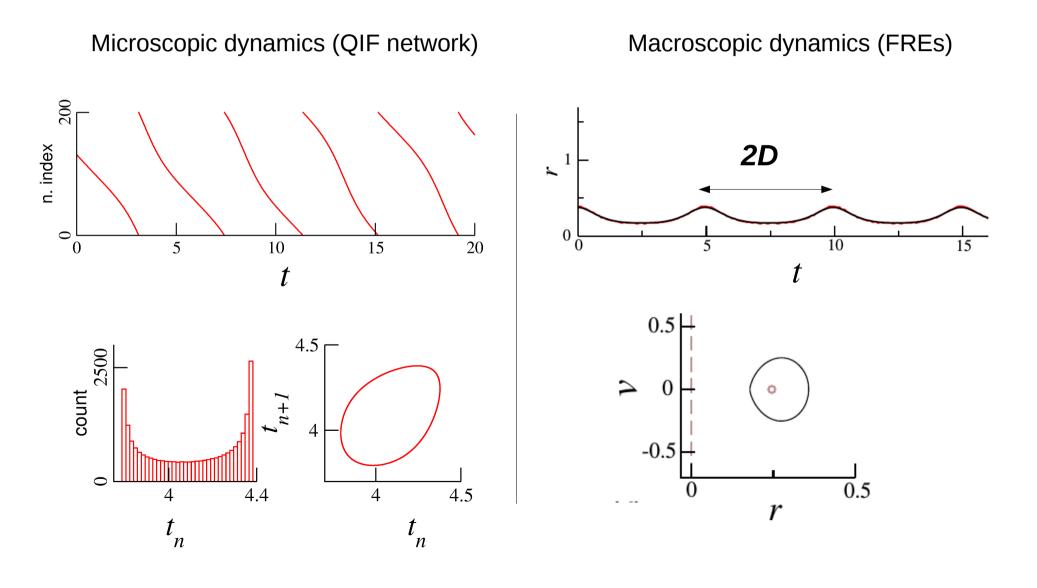


Pazó, Montbrió, PRL (2016)

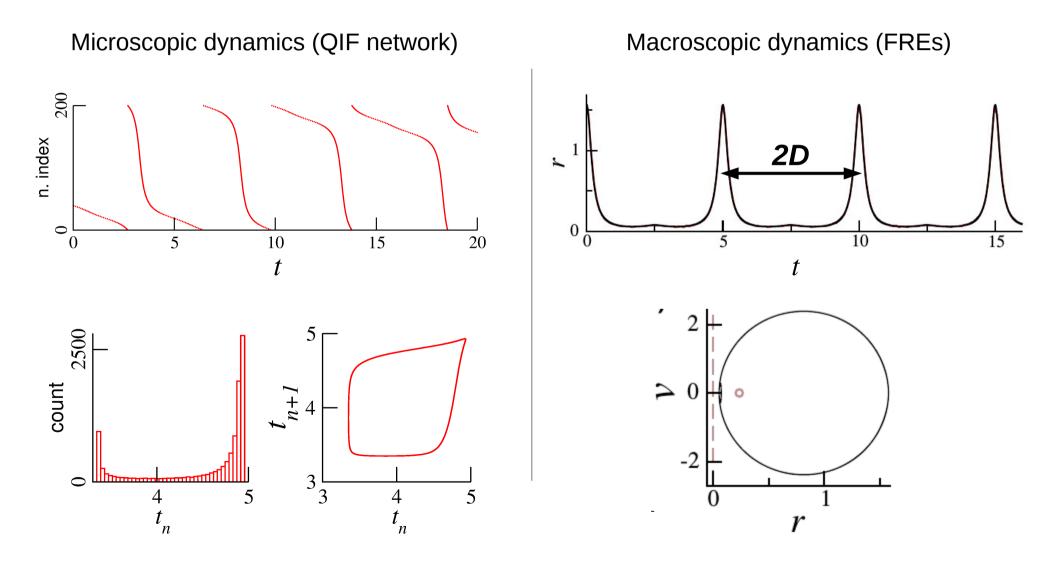
Fast oscillations as in Heuristic FRM



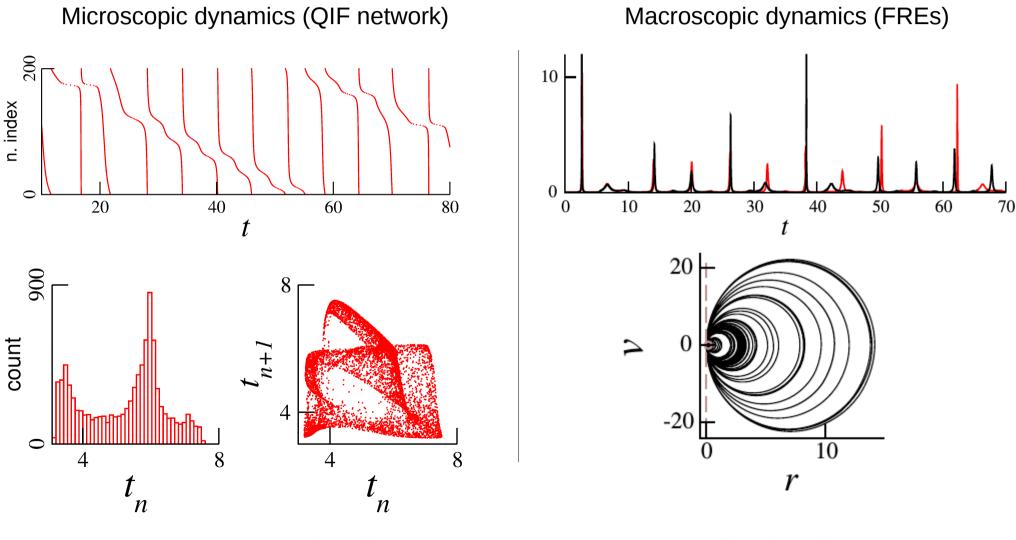
Micro vs. Macro dynamics: Fast osc. in Inhibitory networks & Quasiperiodic Partial Sync



Decreasing *J*, **period of QPS remains constant** (symmetry of l.c. $V \rightarrow -V$)



Decreasing J,D: Macroscopic Chaos

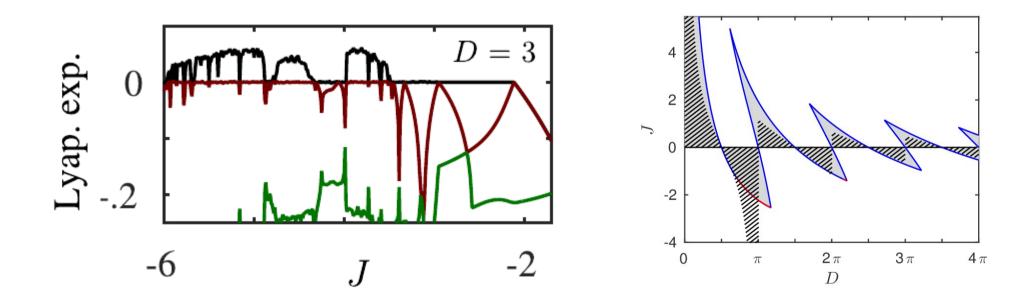


Neurons are not chaotic!

Collective chaos

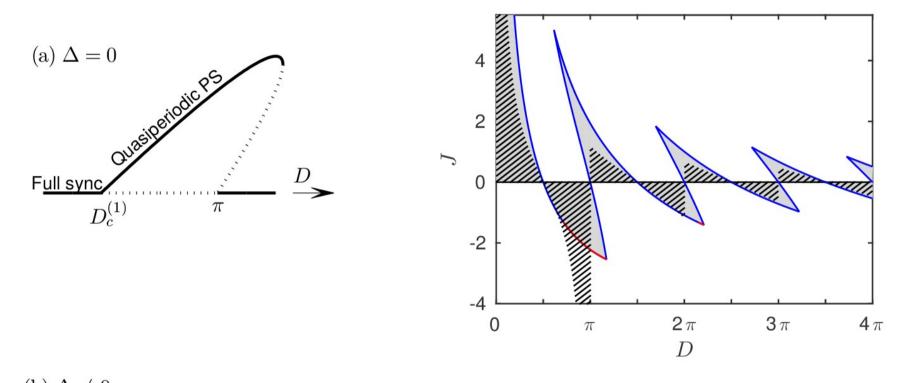
Transition from QPS to Collective Chaos period-doubling cascade

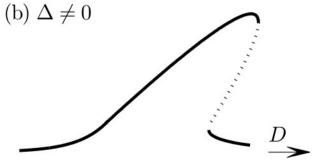
Using the FREs we find:



In the thermodynamic limit the system shows genuine Collective Chaos

Onset of QPS and (weak) heterogeneity

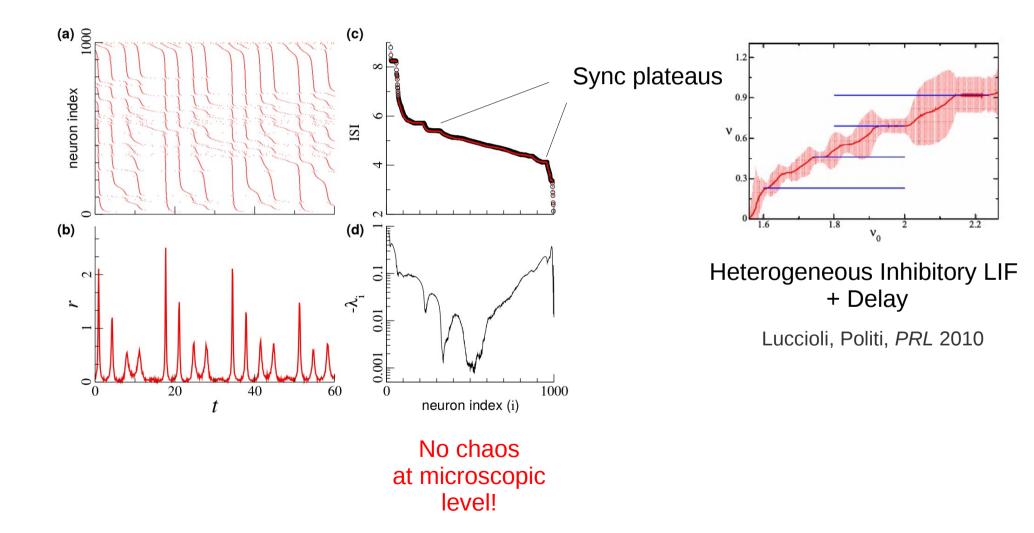




TC bifs. are not robust

bistability remains though!

Macroscopic chaos in heterogeneous networks



Summary

- Using an exact FRM we related QPS and CC to Fast Oscillations in Inhibitory Networks
- Same Collective Oscillations arise due to distinct Microscopic dynamics: sparse sync, QPS, CC...
- Transition from QPS to CC via period doubling cascade
- CC and QPS are also present in (weakly) Heterogeneous networks

Thanks!

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