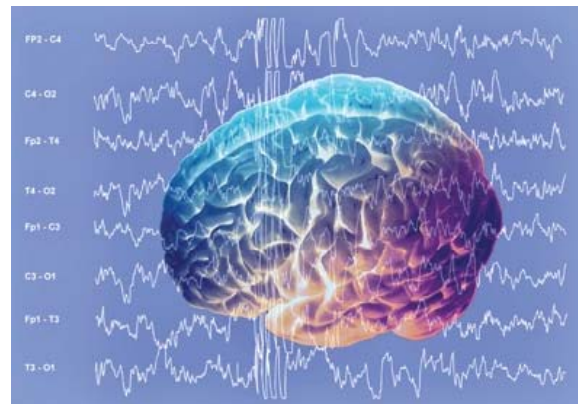


Synchrony in Neural Systems:

a very brief, biased, basic view

Tim Lewis
UC Davis

NIMBIOS Workshop on Synchrony
April 11, 2011

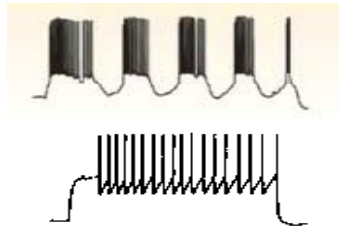
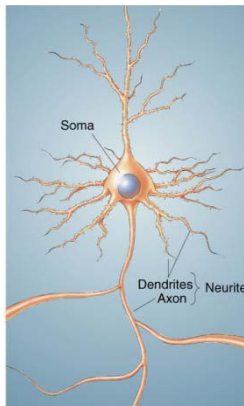


components of neuronal networks

neurons

cell type

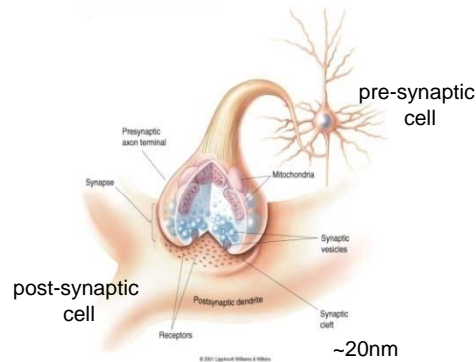
- intrinsic properties (densities of ionic channels, pumps, etc.)
- morphology (geometry)
- noisy, heterogeneous



synapses

synaptic dynamics

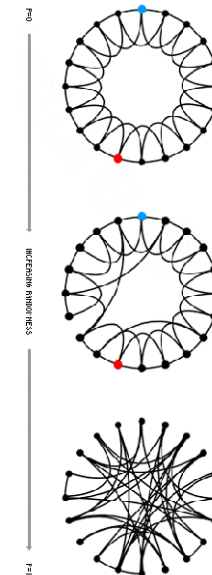
- excitatory/inhibitory; electrical
- fast/slow
- facilitating/depressing
- noisy, heterogeneous
- delays



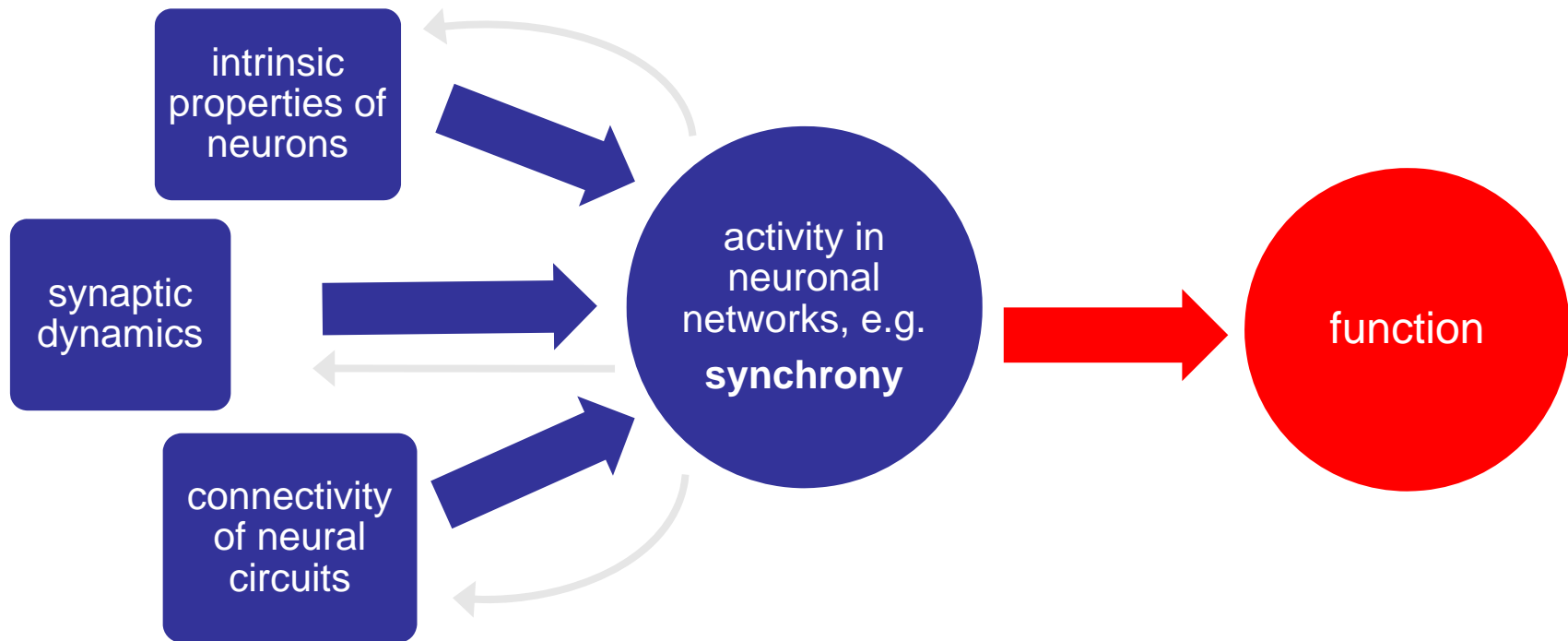
connectivity

network topology

- specific structure
- random; small world, local
- heterogeneous



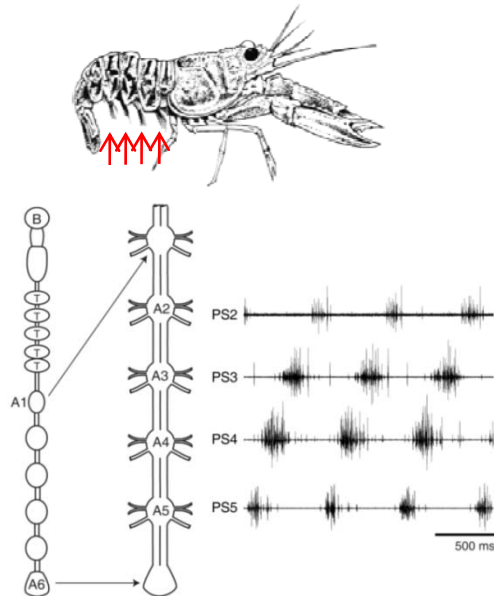
a fundamental challenge in neuroscience



why could “*synchrony*” be important for function in neural systems?

1. coordination of overt behavior: locomotion, breathing, chewing, etc.

example: crayfish swimming



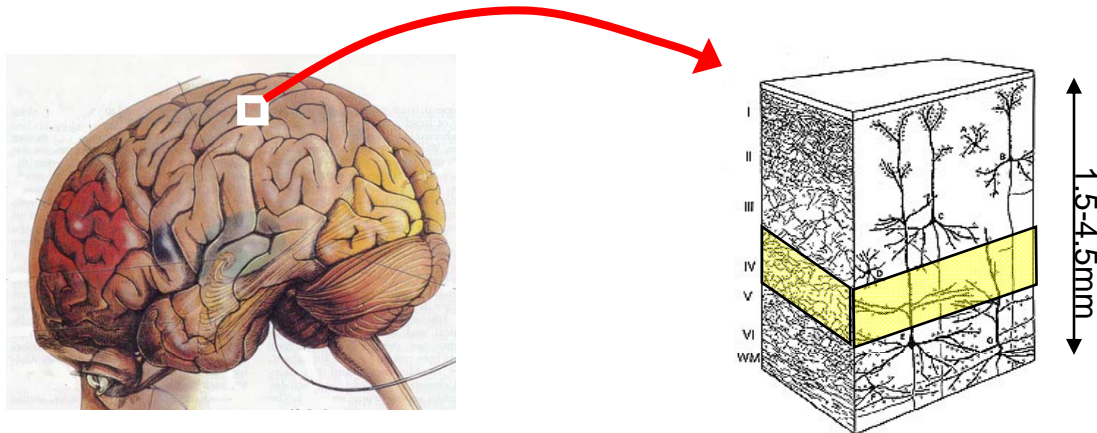
shrimp swimming



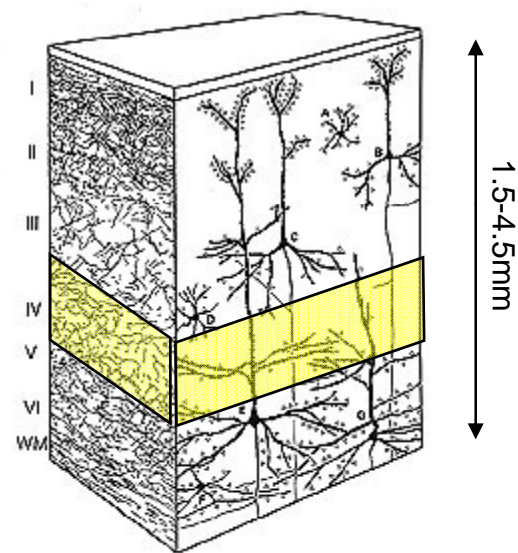
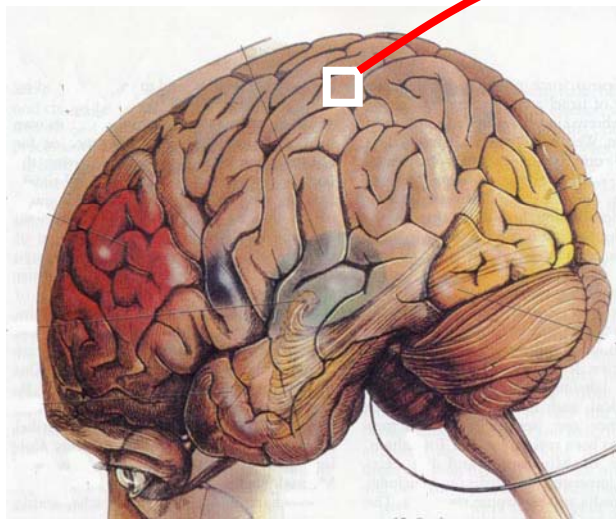
B. Mulloney et al

why could “*synchrony*” be important for function in neural systems?

1. **coordination of overt behavior:** locomotion, breathing, chewing, etc.
2. **cognition, information processing (e.g. in the cortex) ...?**



Should we expect synchrony in the cortex?



Should we expect synchrony in the cortex?

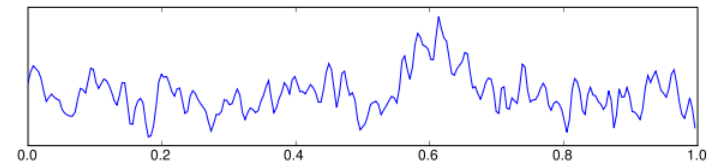
EEG: “brain waves”: **behavioral correlates, function/dysfunction**



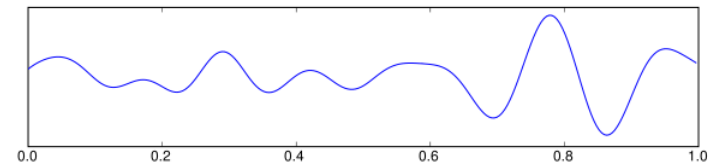
Figure 1: A geodesic net with 128 electrodes making scalp contact with a salinated sponge material is shown (Courtesy Electrical Geodesics, Inc). This is one of several kinds of EEG recording methods. Reproduced from Nunez (2002).

EEG recordings

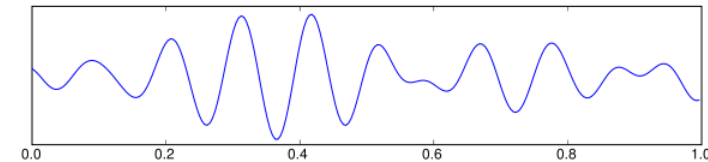
“raw”



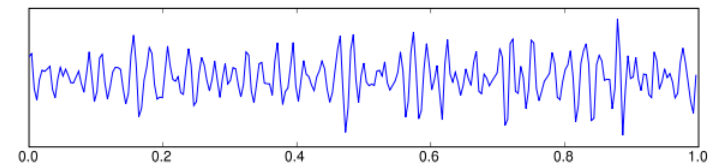
θ



α

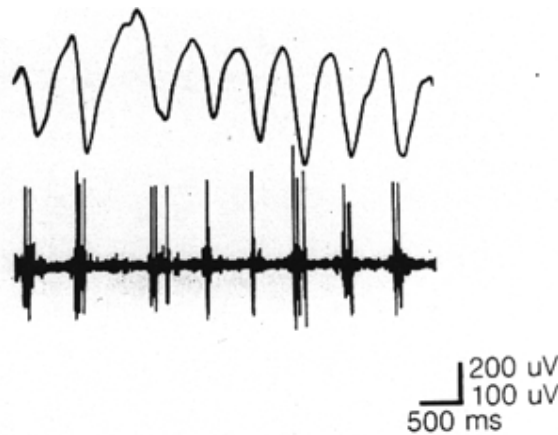


γ



time (sec)

large-scale cortical oscillations arise
from synchronous activity in neuronal networks



Gray and Singer (1989) Proc. Nat. Acad. Sci. 86: 1698-1702.

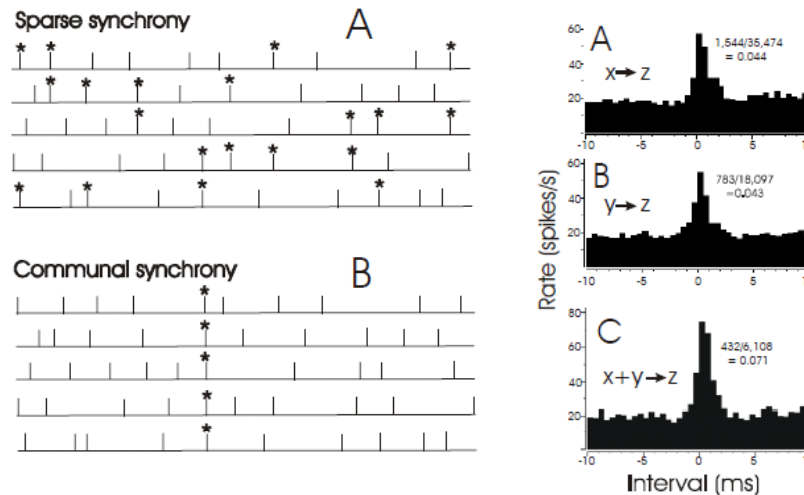
in vivo γ -band (30-70 Hz) cortical oscillations.

how can we gain insight into the functions and dysfunctions related to neuronal synchrony?

1. Develop appropriate/meaningful ways of measuring/quantifying synchrony.
2. Identify mechanisms underlying synchronization – both from the dynamical and biophysical standpoints.

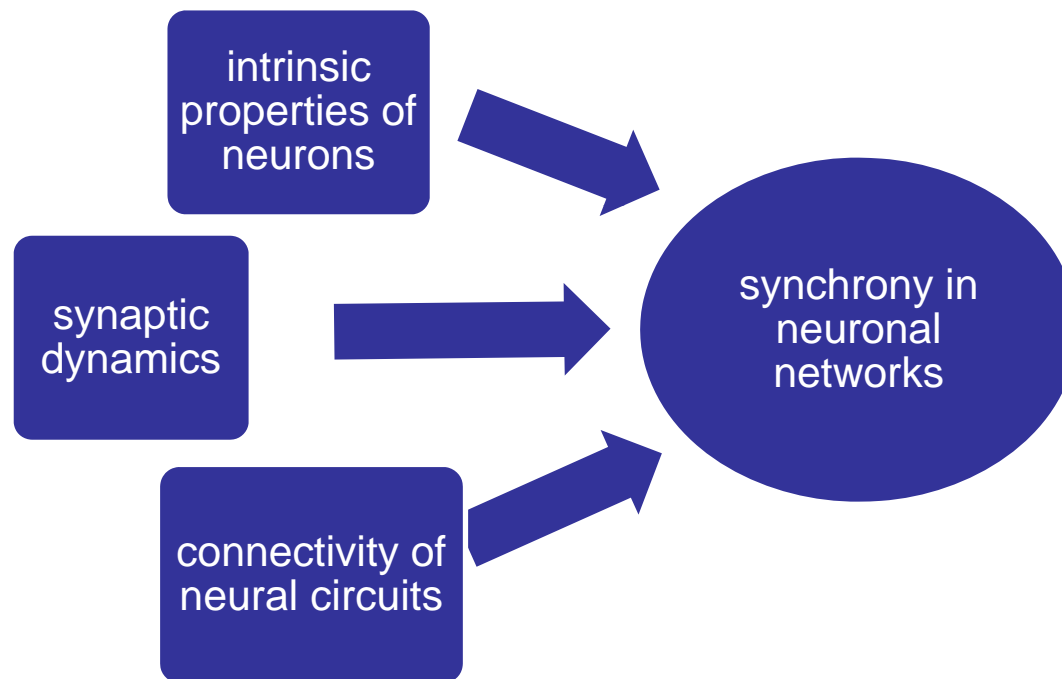
1. measuring correlations/levels of synchrony

- i. limited spatio-temporal data.
- ii. measuring phase
- iii. spike-train data (embeds “discrete” spikes in continuous time)
- iv. appropriate assessment of chance correlations.
- v. higher order correlations (temporally and spatially)



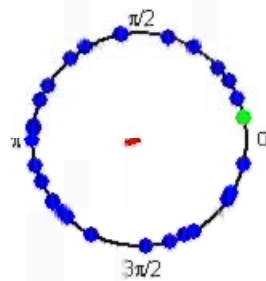
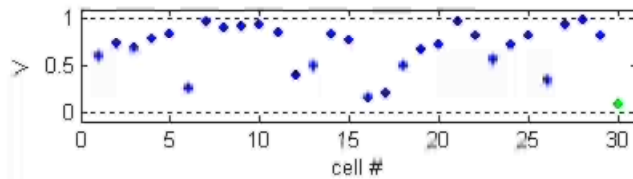
Swadlow, et al.

2. identify mechanisms underlying synchrony



[mechanism A] synchrony in *networks of neuronal oscillators*

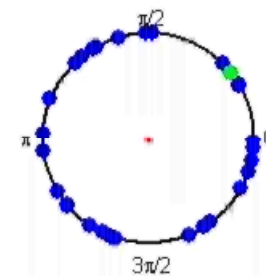
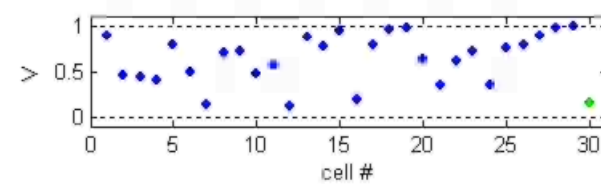
“fast” excitatory synapses



synchrony

**movie:
LIFfastEsynch.avi**

“slow” excitatory synapse



asynchrony

**movie:
LIFslowEasynch.avi**

[mechanism A] synchrony in networks of neuronal oscillators

Some basic mathematical frameworks:

1. phase models (e.g. Kuramoto model)

$$\begin{aligned}\frac{d\theta_j}{dt} &= \omega_j + H_j(\theta_1, \dots, \theta_N), \quad j = 1, \dots, N \\ &= \omega_j + \sum_{k=1}^N w_{kj} H(\theta_k - \theta_j) \quad \theta_j \in [0, 1)\end{aligned}$$

[mechanism A] synchrony in networks of neuronal oscillators

Some basic mathematical frameworks:

2. theory of weak coupling (Malkin, Neu, Kuramoto, Ermentrout-Kopell, ...)

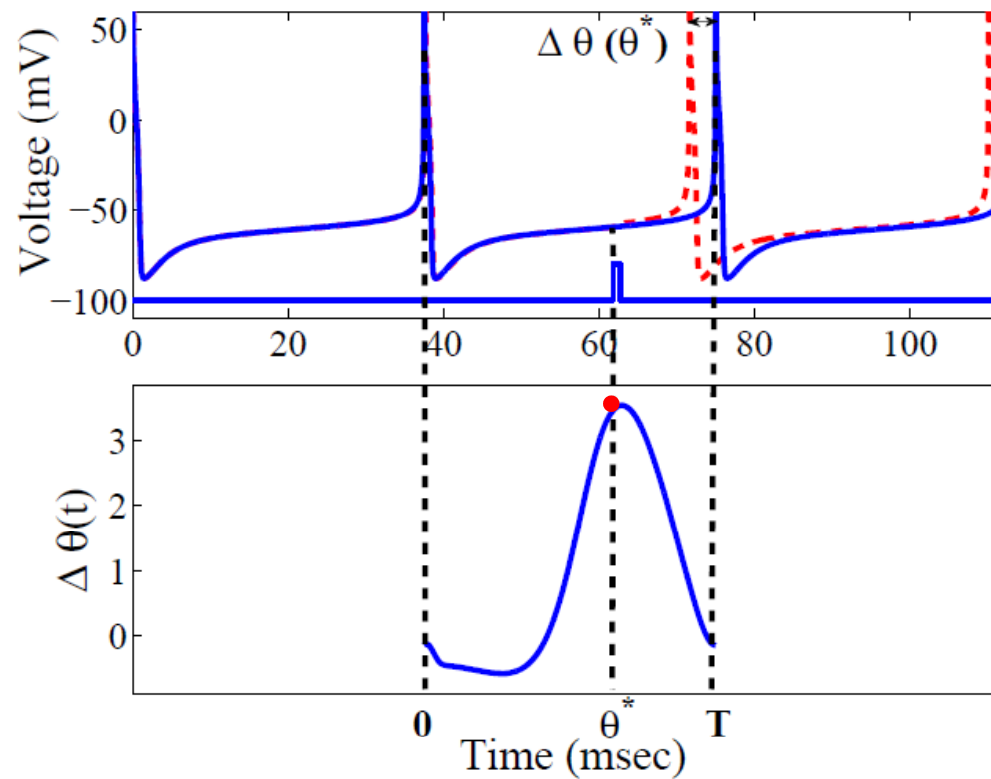
$$\begin{aligned}\frac{d\theta_j}{dt} &= \omega_j + \sum_{k=1}^N w_{kj} \frac{1}{T} \int_0^T Z(\tilde{t} + \phi_j T) \tilde{I}_{syn}(V_o(\tilde{t} + \phi_k T)) d\tilde{t}, \quad j = 1, \dots, N \\ &= \omega_j + \sum_{k=1}^N w_{kj} H(\theta_k - \theta_j)\end{aligned}$$

iPRC

synaptic
current

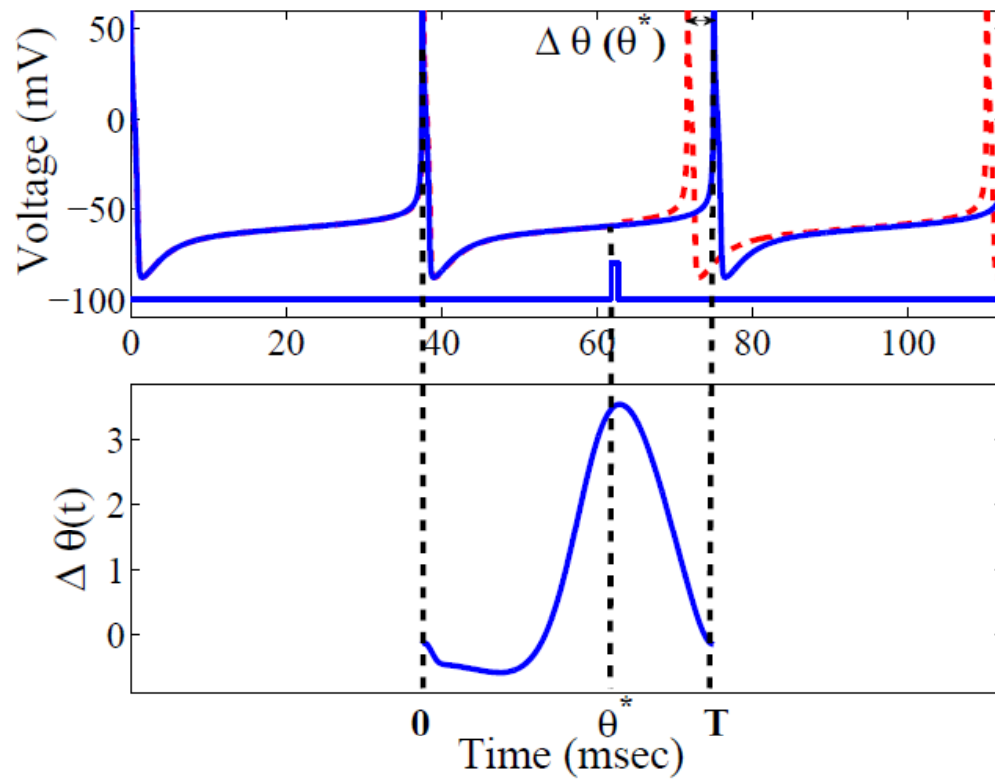
phase response curve (PRC) $\Delta\theta(\theta)$

quantifies the phase shifts in response to **small, brief** (δ -function) input at different phases in the oscillation.



infinitesimal phase response curve (iPRC) $Z(\theta)$

the PRC normalized by the stimulus “amplitude” (i.e. total charge delivered).



[mechanism A] synchrony in networks of neuronal oscillators

Some mathematical frameworks:

2. spike-time response curve (STRC) maps

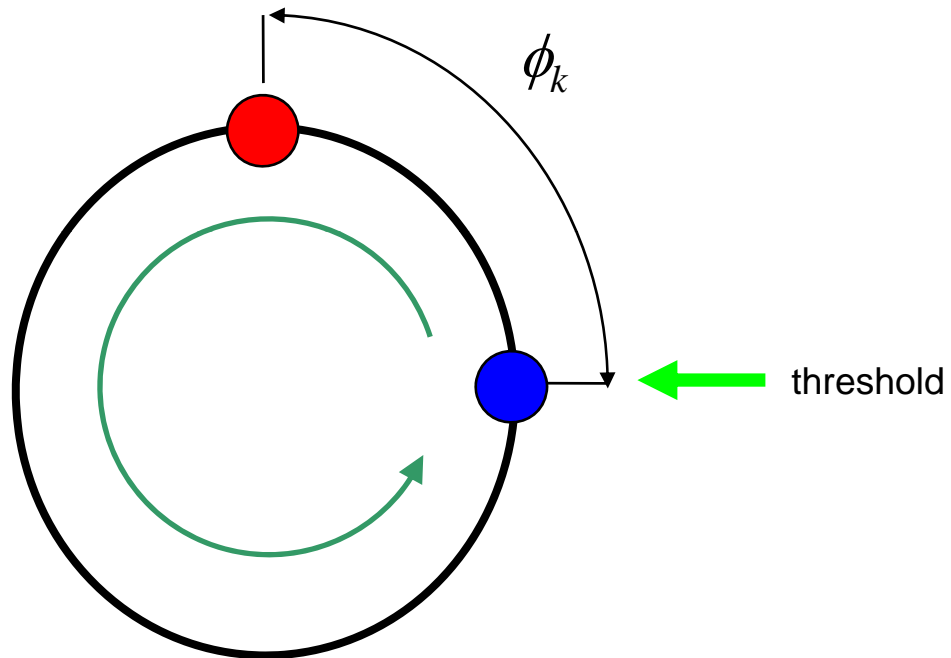
$$\phi_{k+1} = \phi_k + \Delta\theta(1 - \phi_k) + \Delta\theta(\phi_k + \Delta\theta(1 - \phi_k))$$

ϕ_k = phase difference between pair of coupled neurons when neuron 1 fires for the k^{th} time.

$\Delta\theta(\theta)$ = phase response curve for neuron for a given stimulus..

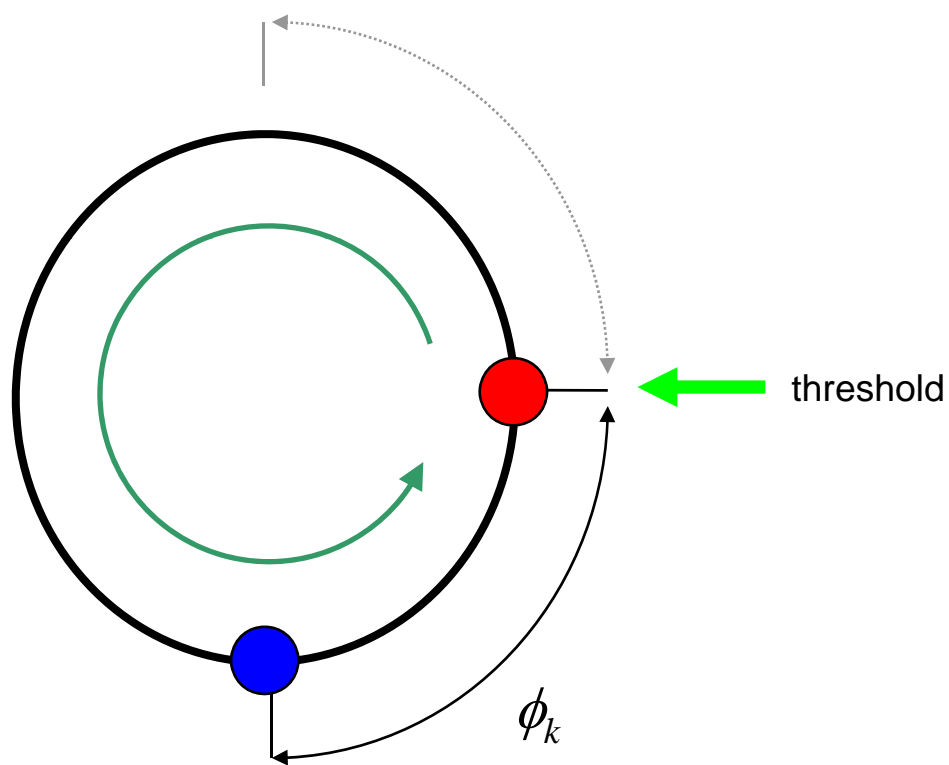
pair of coupled cells: reduction to 1-D map

blue cell has just been reset after crossing threshold.



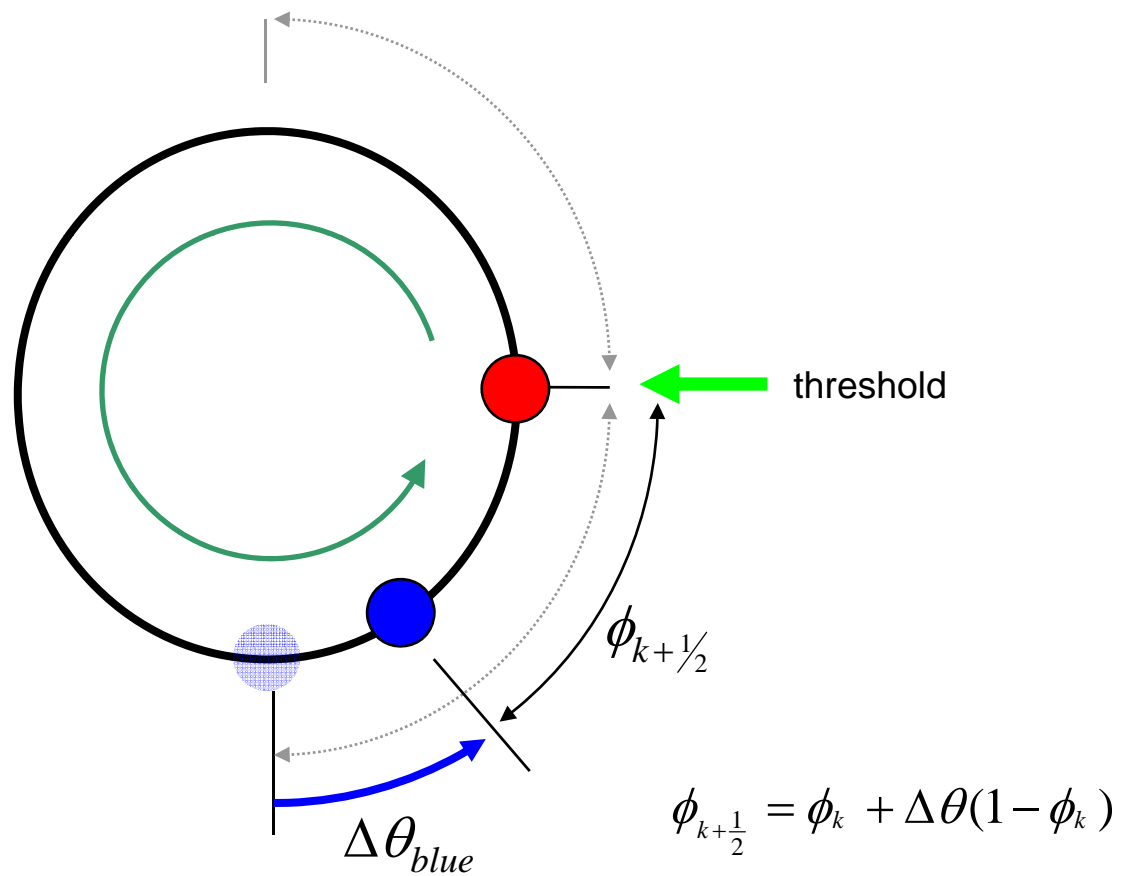
pair of coupled cells: reduction to 1-D map

red cell hits threshold.



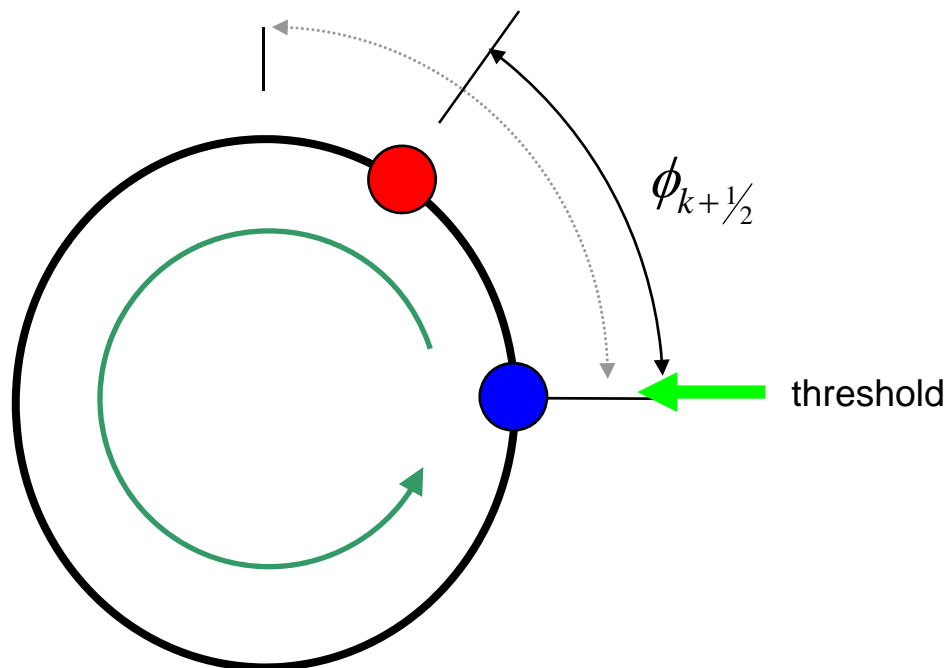
pair of coupled cells: reduction to 1-D map

blue cell is phase advanced by synaptic input from red cell; red cell is reset.



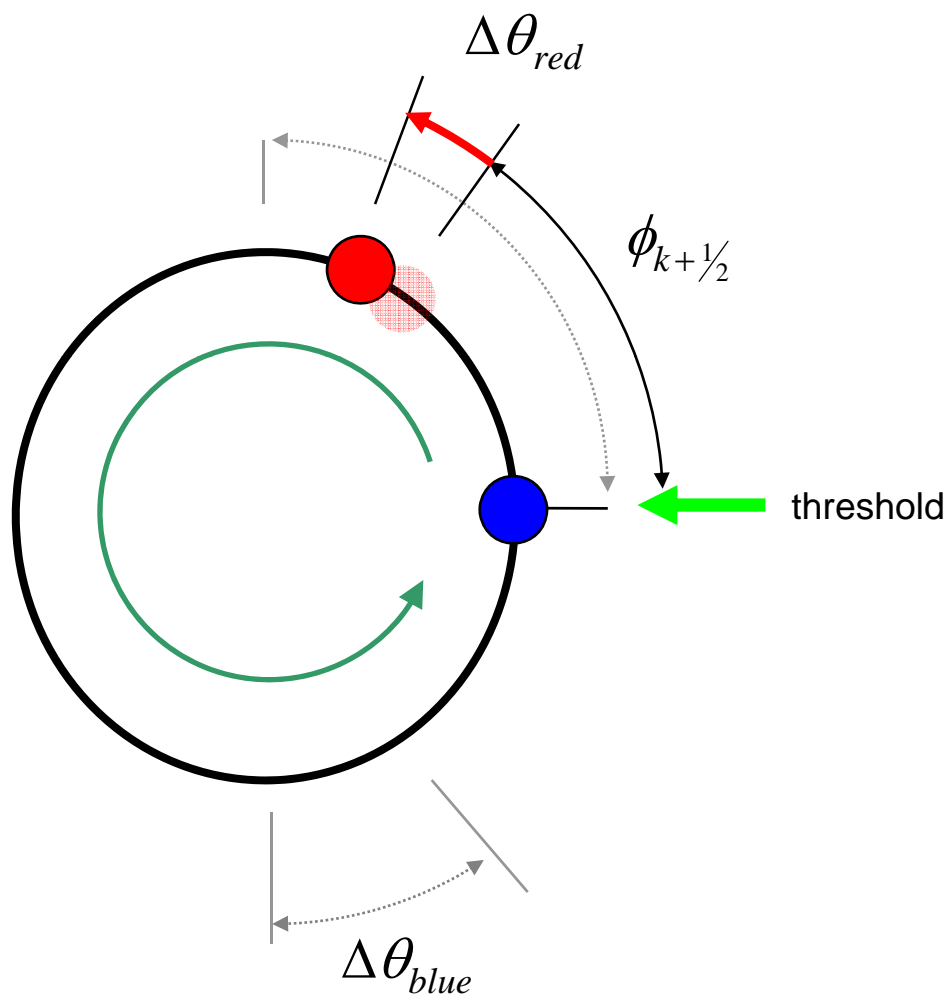
pair of coupled cells: reduction to 1-D map

blue cell hits threshold.

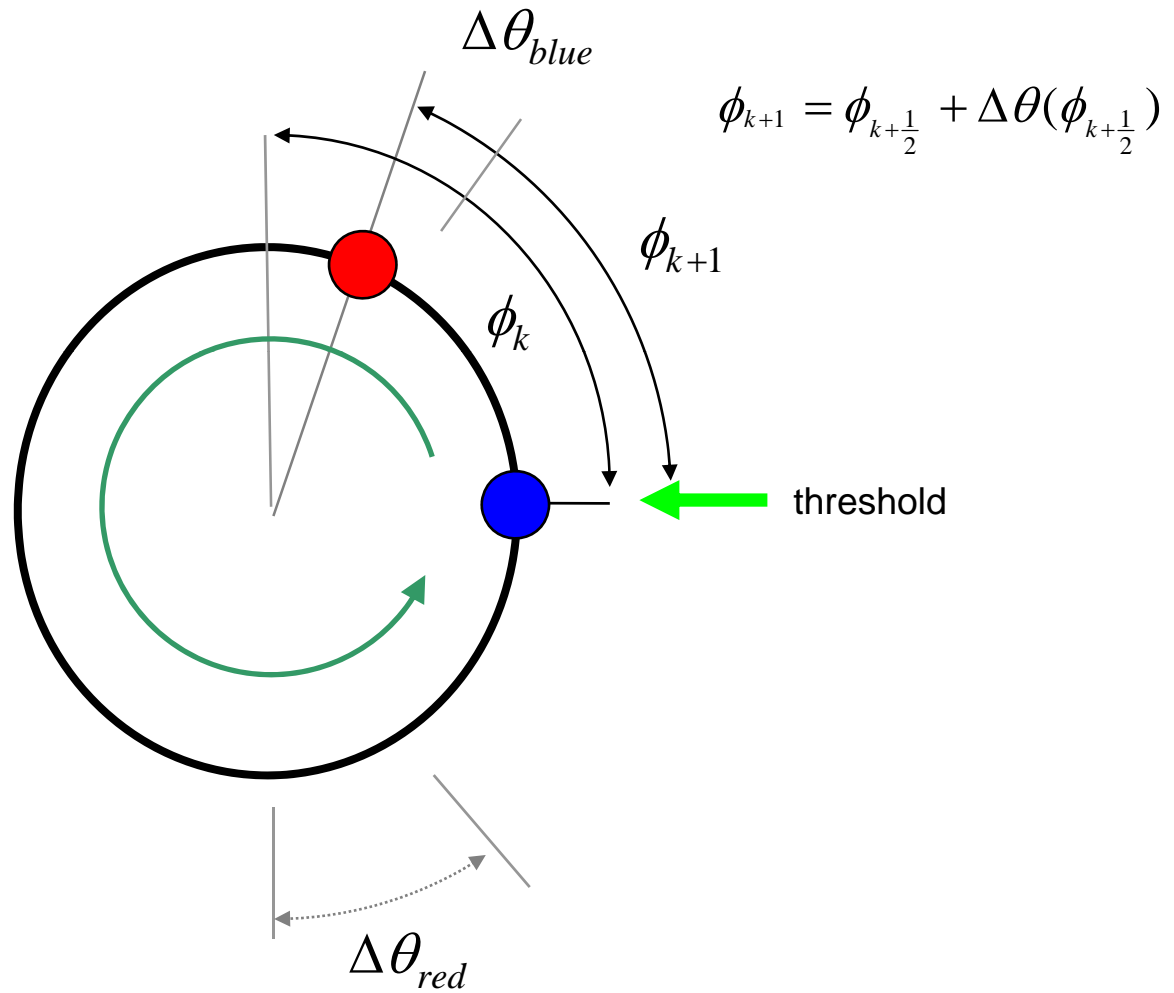


pair of coupled cells: reduction to 1-D map

red cell is phase advanced by synaptic input from blue cell; blue cell is reset.



pair of coupled cells: reduction to 1-D map



pair of coupled cells: reduction to 1-D map

(similar to Strogatz and Mirillo, 1990)

$$\phi_{k+\frac{1}{2}} = \phi_k + \Delta\theta(1 - \phi_k)$$

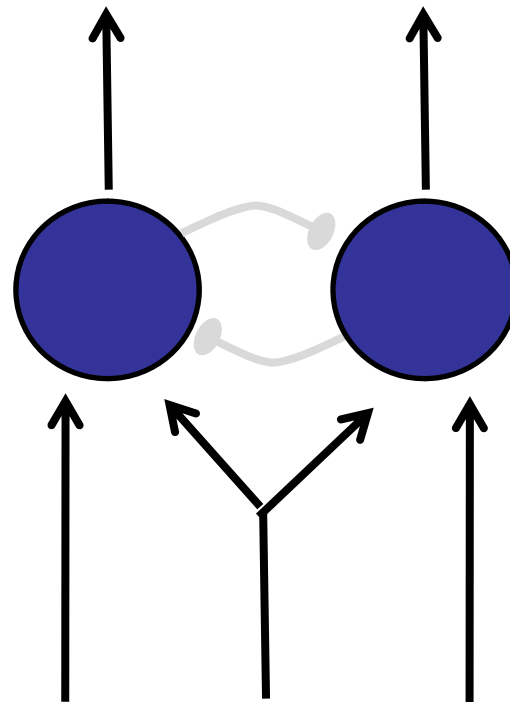
$$\phi_{k+1} = \phi_{k+\frac{1}{2}} + \Delta\theta(\phi_{k+\frac{1}{2}})$$



$$\phi_{k+1} = \phi_k + \Delta\theta(1 - \phi_k) + \Delta\theta(\phi_k + \Delta\theta(1 - \phi_k))$$

* shape of Z determines phase-locking dynamics

[mechanism B] correlated/common input into oscillating or excitable cells (i.e., the neural Moran effect)

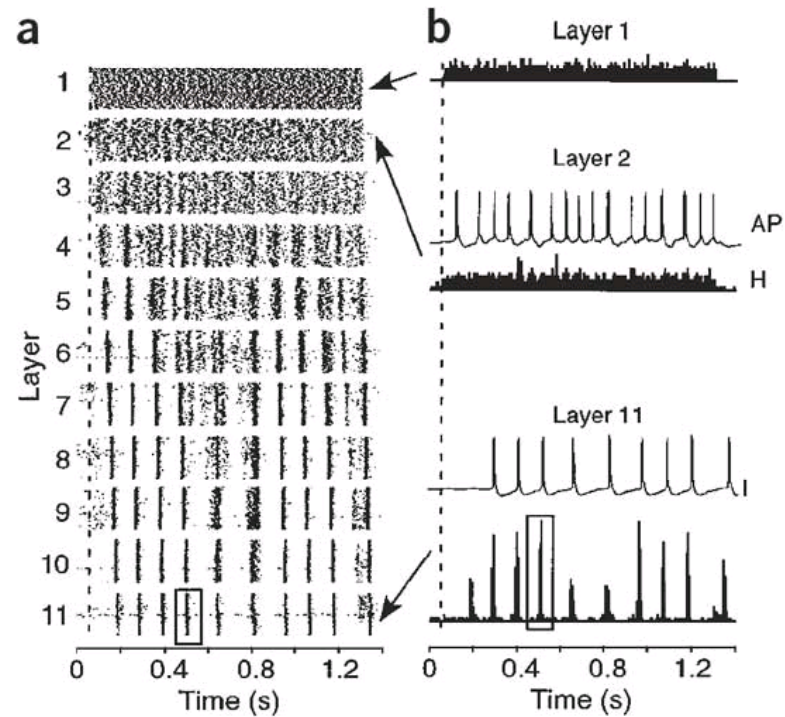
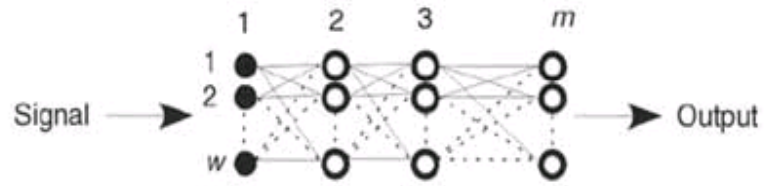


output

... possible network coupling too

input

[mechanism C] self-organized activity in networks of excitable neurons: feed-forward networks

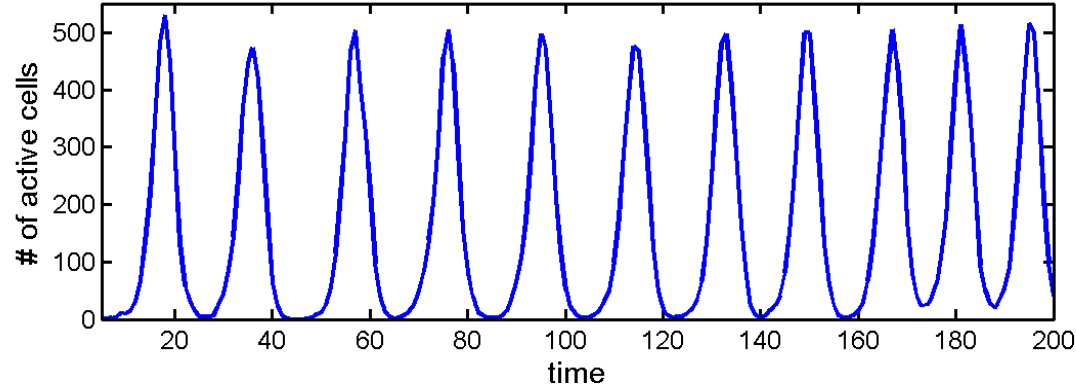


e.g. AD Reyes, *Nature Neurosci* 2003

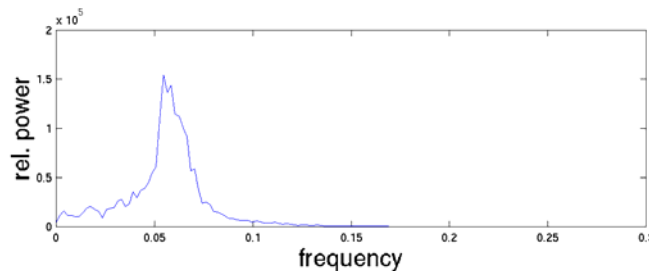
[mechanism C] self-organized activity in networks of excitable neurons: random networks

(i) Topological target patterns

units: excitable dynamics with low level of random spontaneous activation (Poisson process);
network connectivity: sparse (Erdos-Renyi) random network; strong bidirectional coupling.



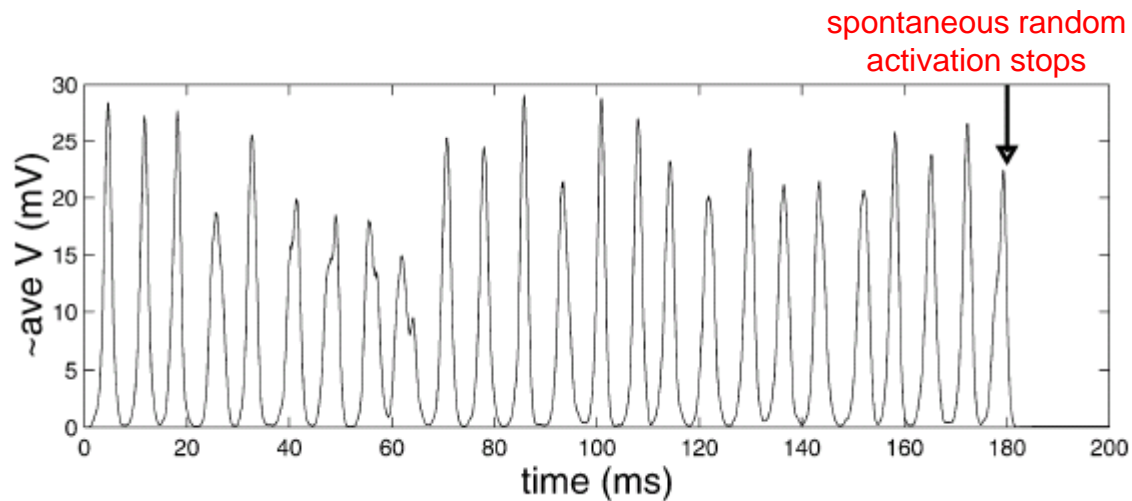
$\lambda=0.0001$, 75×50 network, $r_c=50$, $c=0.37$



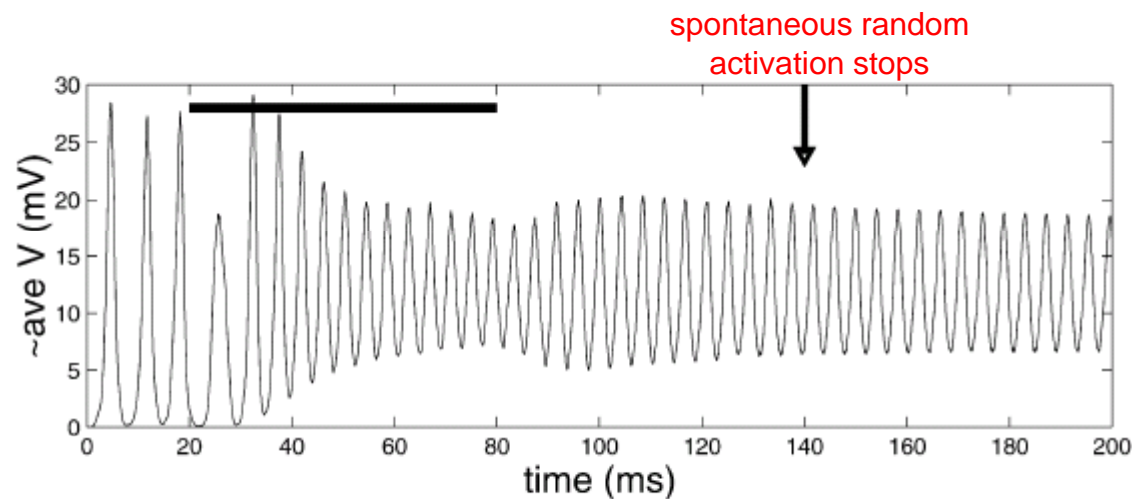
movie:
topotargetoscill.avi

e.g. Lewis & Rinzel, *Network: Comput. Neural Syst.* 2000

[mechanism C] self-organized rhythms in networks of excitable neurons



(i) *topological target patterns waves*



(ii) *reentrant waves*

e.g. Lewis & Rinzel, *Neurocomput.* 2001

Conclusions / Discussions / Open questions