



A faint, large-scale visualization of a complex network is positioned behind the title text. It features numerous small, light-colored nodes connected by a dense web of thin lines, resembling a brain's neural network or a social network graph.

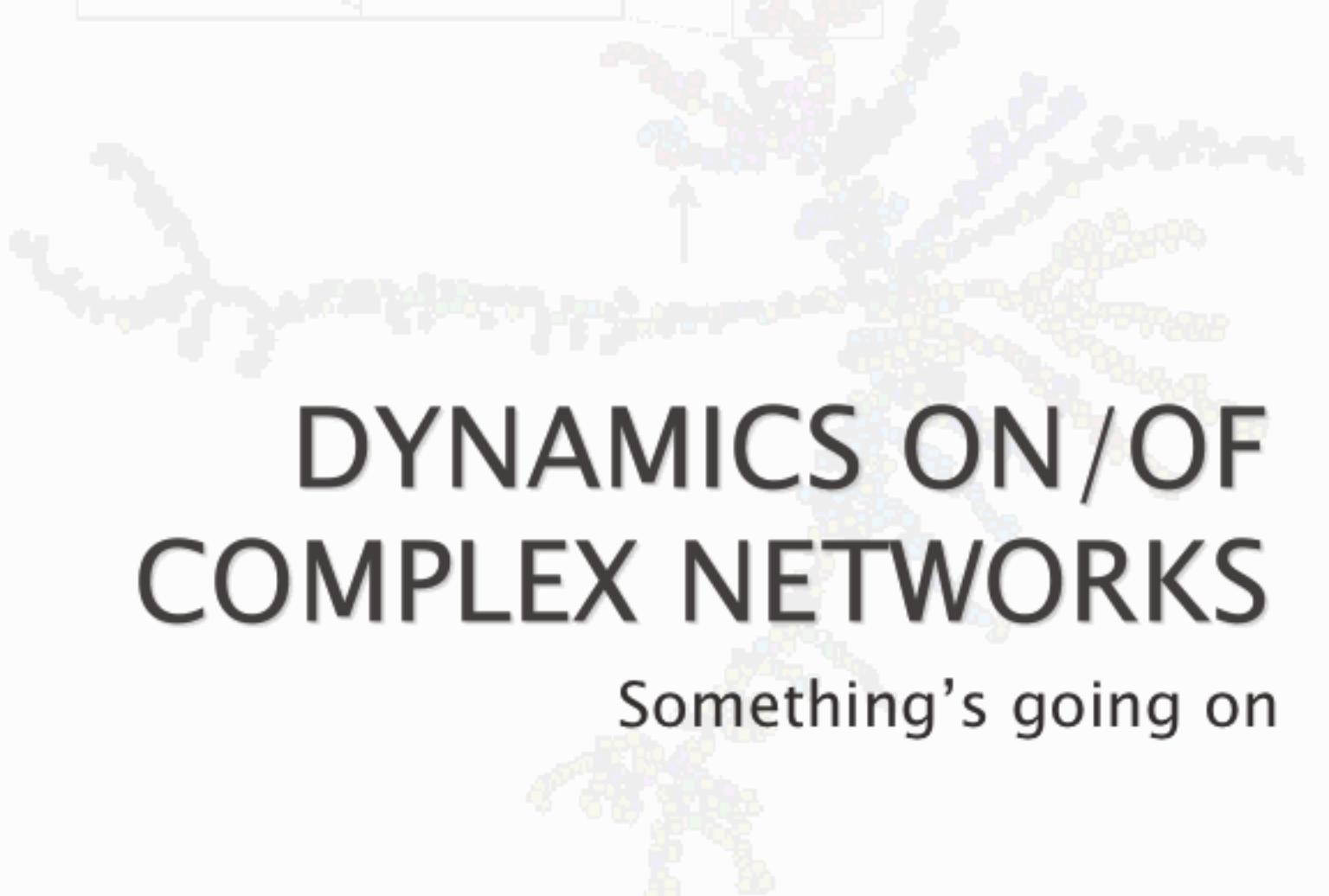
# Complex networks III

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@anduviera

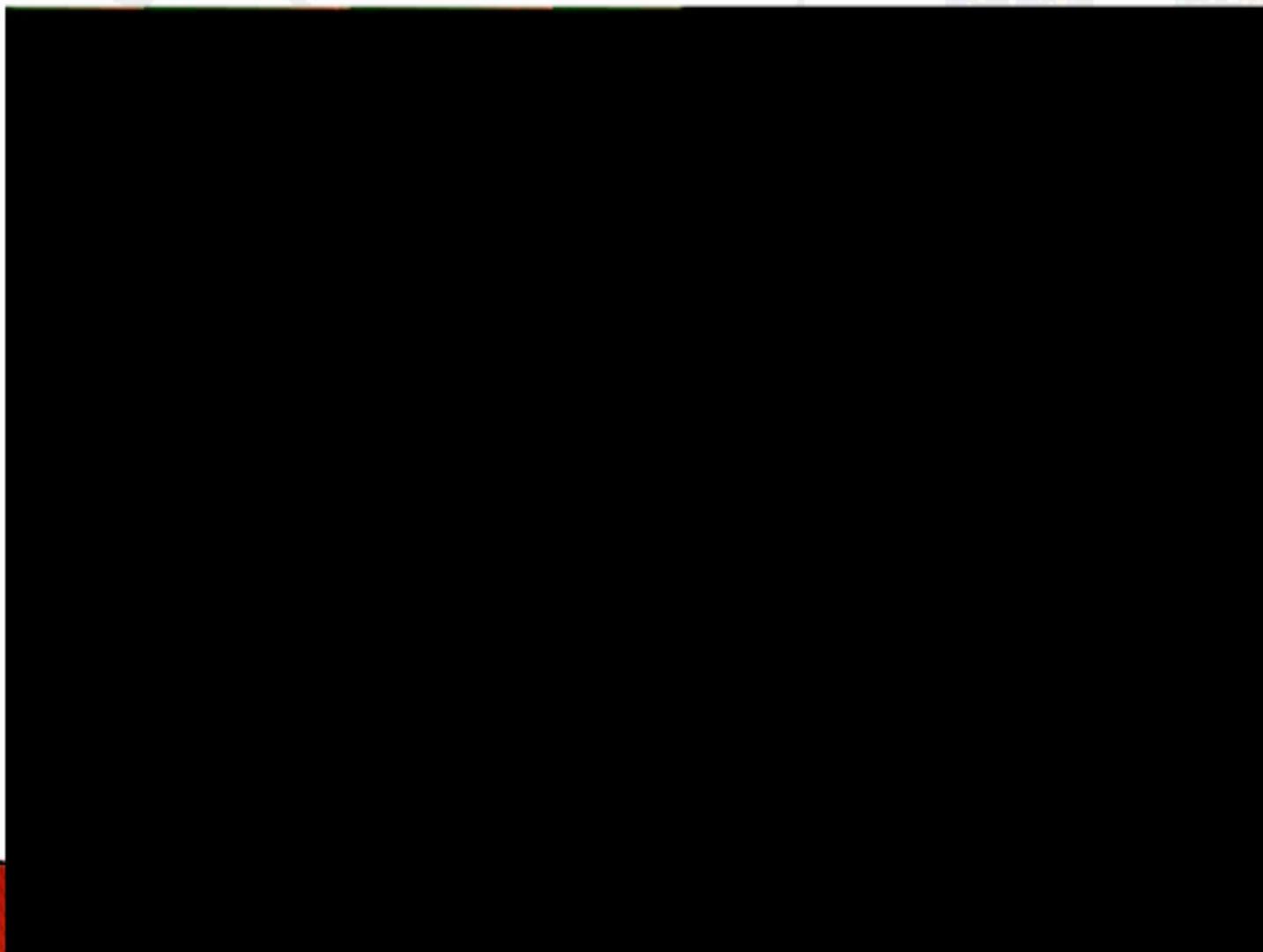
PHYSCOMP<sup>2</sup> Universitat de Barcelona



# DYNAMICS ON/OF COMPLEX NETWORKS

Something's going on

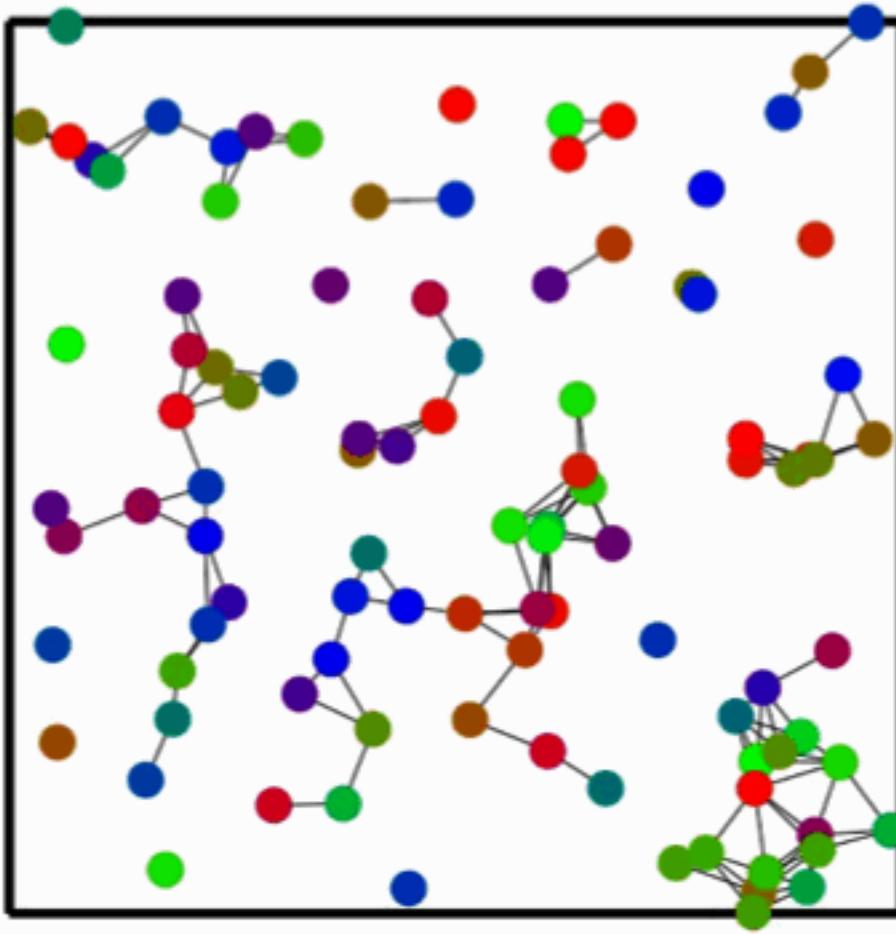
# Dynamics on networks



# Dynamics of networks



# Dynamics on dynamic networks



# Outline

- ▶ General issues
- ▶ Synchronization

- ▶ Network (link) dynamics:
  - global goal
  - local goal
- ▶ Flow in complex networks:
  - ideas
  - innovations
  - computer viruses
  - problems
  - people

# Global vs local optimization

- ▶ Design: the goal is to optimize global quantity (distance, clustering, density, ...)
- ▶ Evolution: decision taken at node level

# Evolution

- ▶ Bornholdt & Rohlf: Global criticality from local dynamics
  - Network of interconnected binary elements
  - The dynamics reaches an attractor
  - Change the connectivity of a node according to its behavior during the attractor
- ▶ Gleiser & Zanette: rearrangements of links
  - Synchronization
  - Change of connectivity pattern to adjust to neighbors
  - Final state: community structure

# Optimization

- ▶ Global goal:
  - Distance: related to minimal cost in transportation
  - Number of connections: costly connections
  - A combination of parameters
- ▶ Initial configuration: random graph
- ▶ Change connections
- ▶ Accept if there is an improvement
- ▶ Stars vs trees

# Synchronization

A. Arenas, A.D.-G., C.J. Pérez-Vicente

- ▶ Community identification: maximization of modularity → very best partition
- ▶ Communities can be hierarchically organized
- ▶ Synchronization: dynamics at all scales from the innermost to the global scale

# Synchro in nature (I): hands clapping

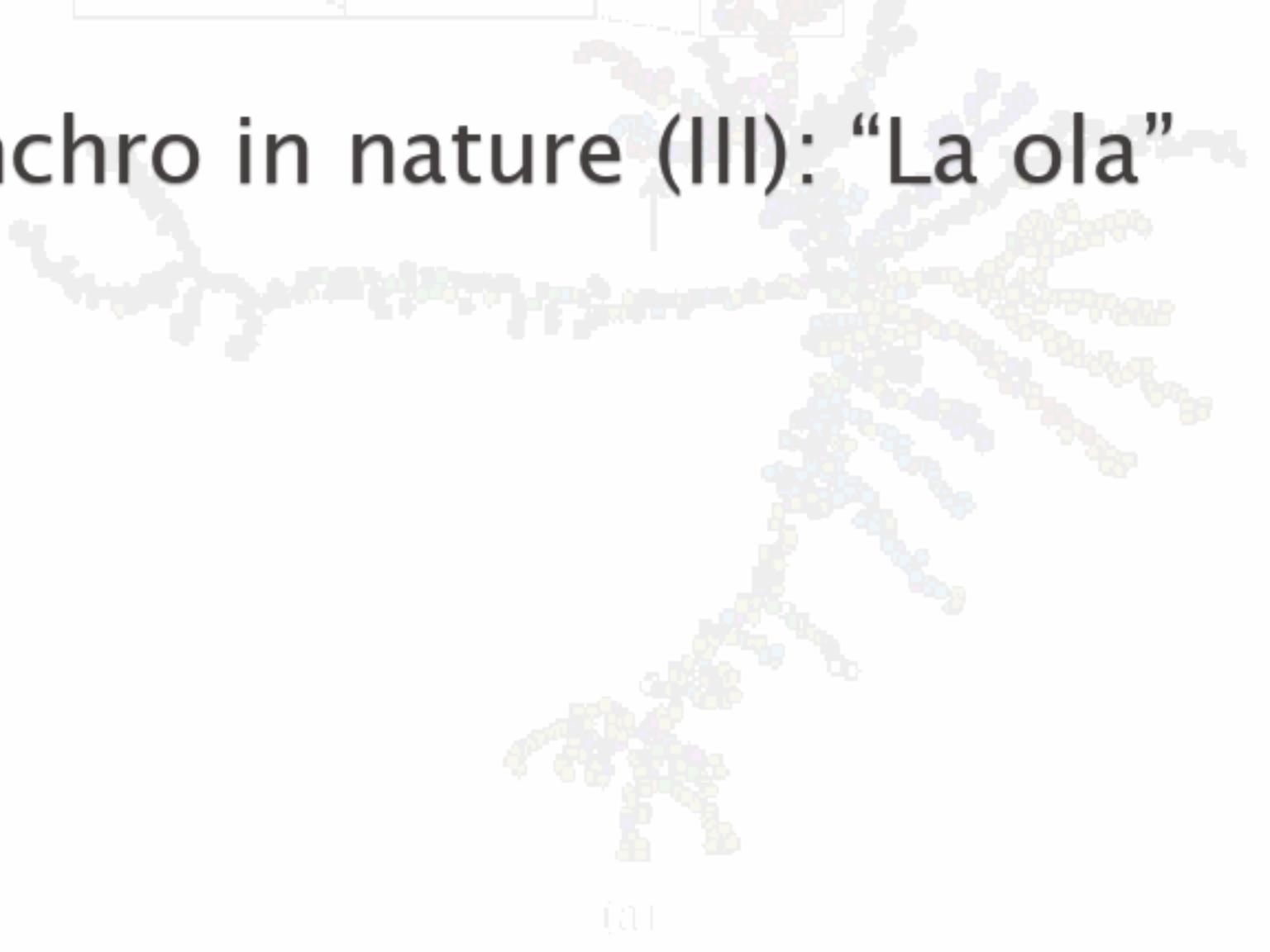
- ▶ Sound file (play)



# Synchro in nature (II): flashing fireflies

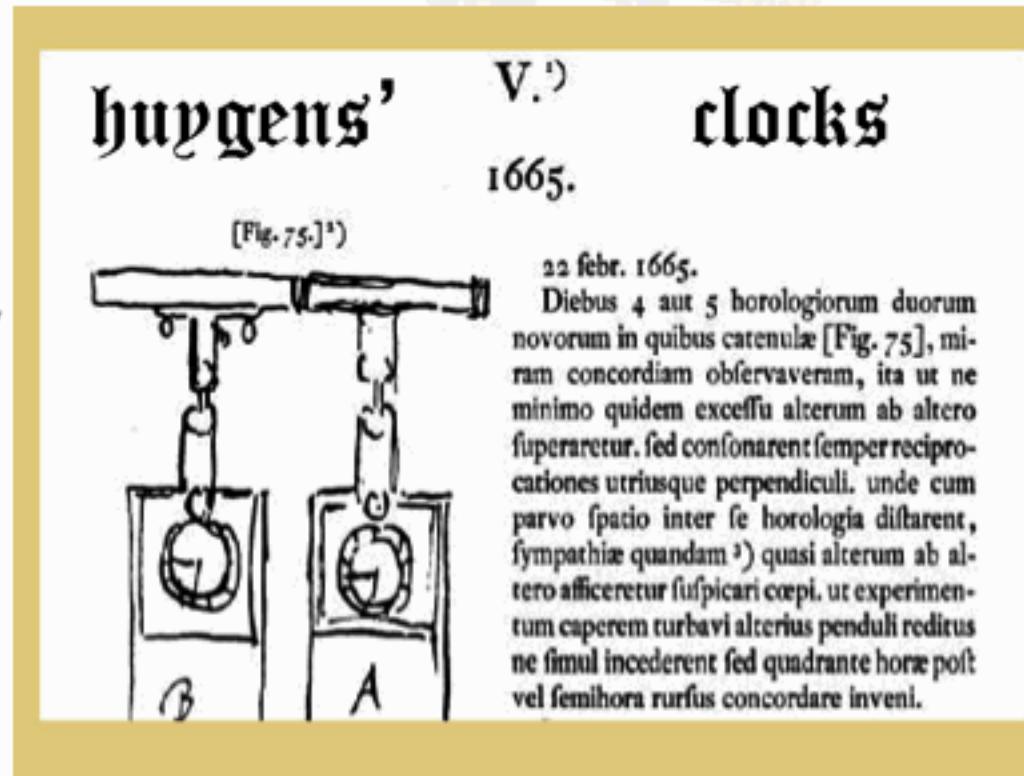


# Synchro in nature (III): “La ola”

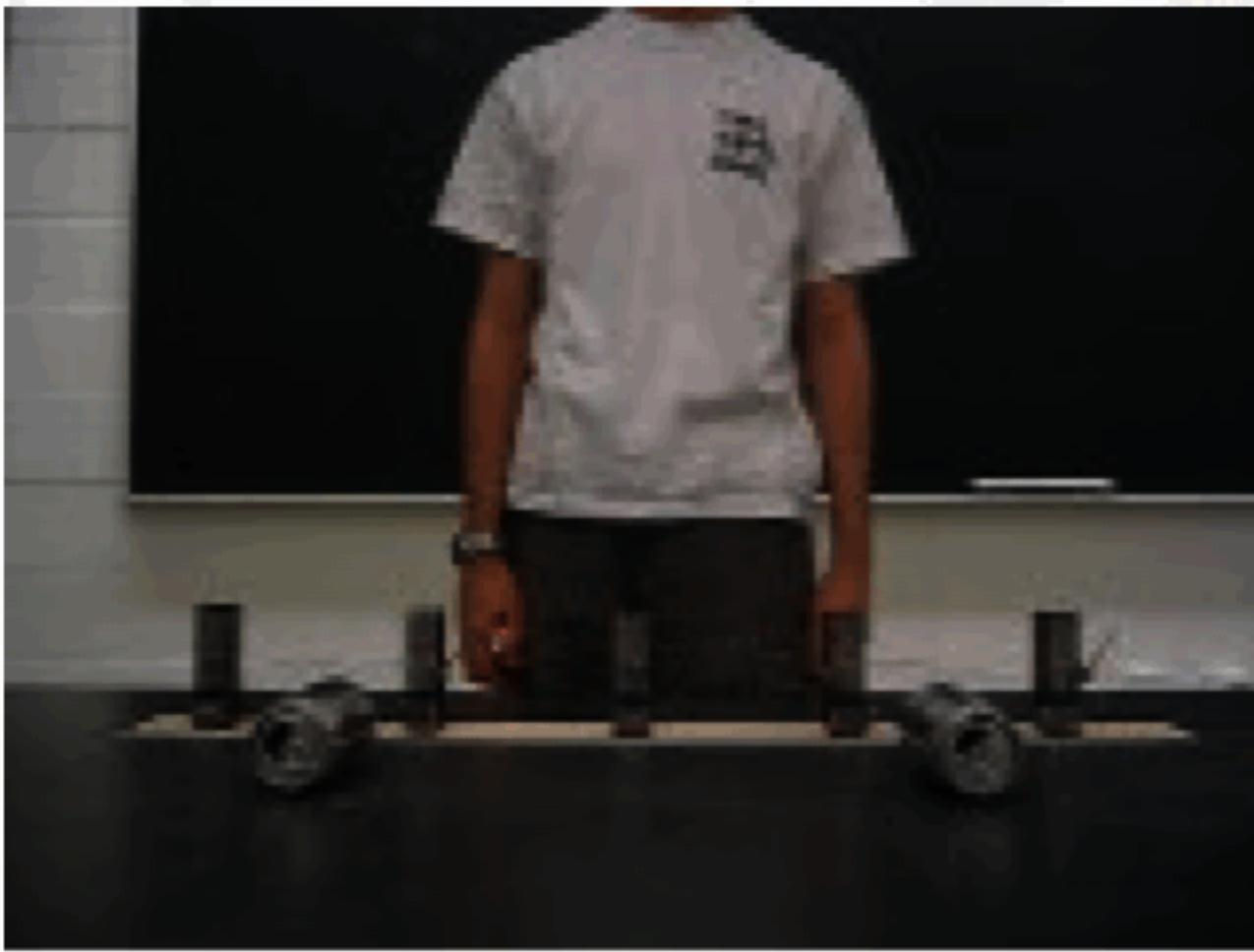


# Synchro in nature (IV)

- ▶ Hearts beats
- ▶ Cricket chirps
- ▶ Laser
- ▶ Superconductivity
- ▶ Menstrual synchrony  
in women living  
together
- ▶ .....



# Synchro in the lab



# Synchronization dynamics

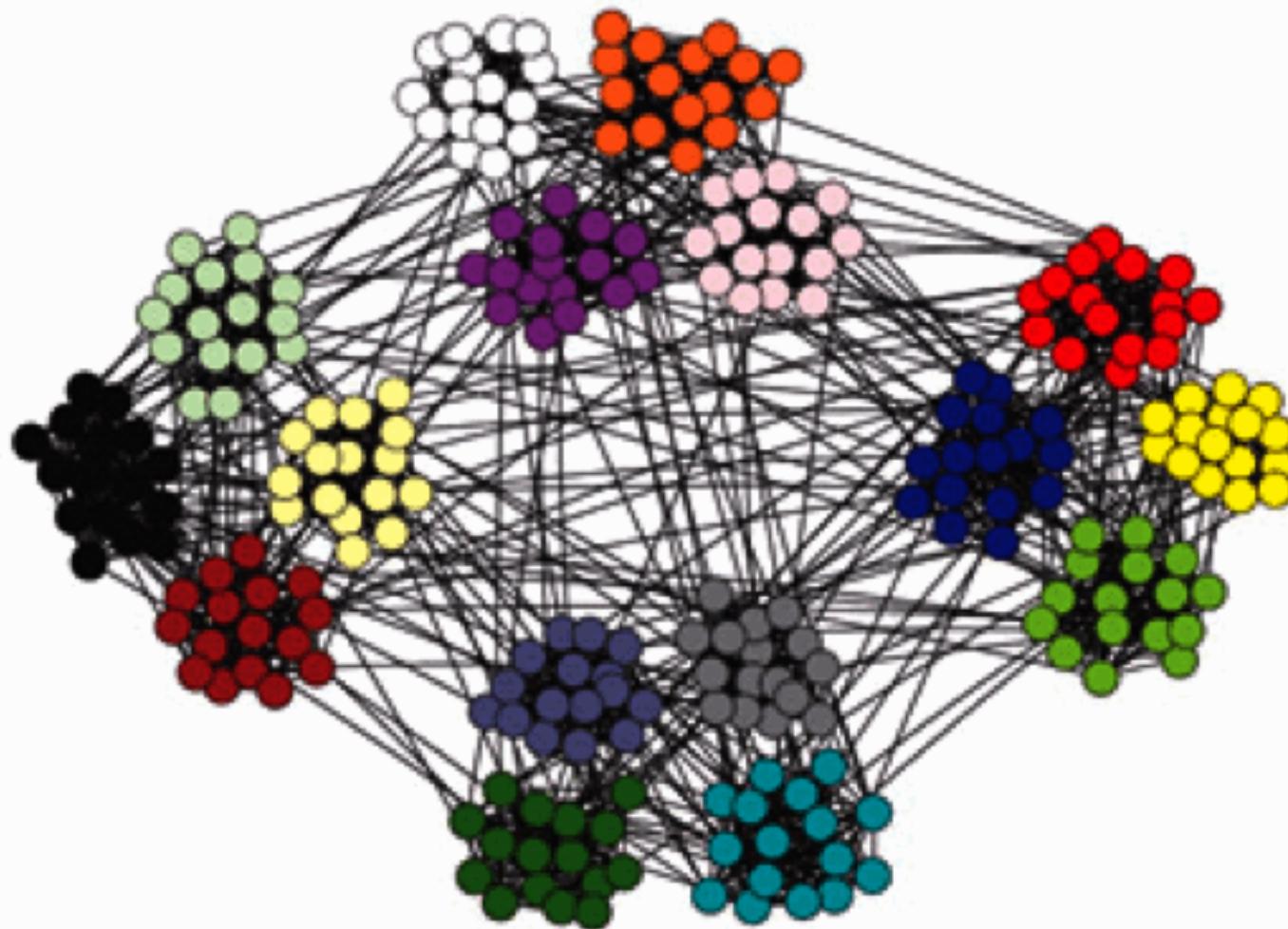
- ▶ Synchronization of Kuramoto oscillators

$$\frac{d\theta_i}{dt} = \omega_i + \sigma \sum_j A_{ij} \sin(\theta_j - \theta_i) \quad i = 1, \dots, N$$

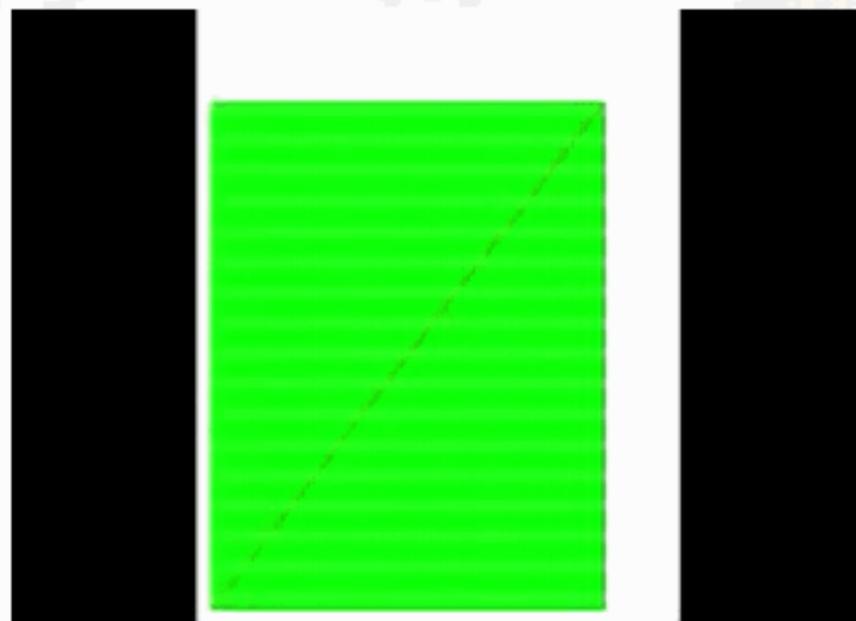
## Kuramoto (the Applet)

$$\frac{d\theta_i}{dt} = \sigma \sum_j A_{ij} \sin(\theta_j - \theta_i) \quad i = 1, \dots, N$$

# Hierarchical structure of communities



# Time evolution of the correlation matrix



$$\rho_{ij}(t) = \langle \cos(\theta_i(t) - \theta_j(t)) \rangle$$

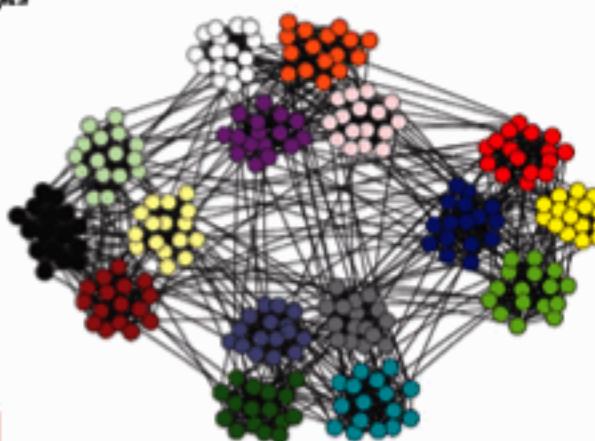
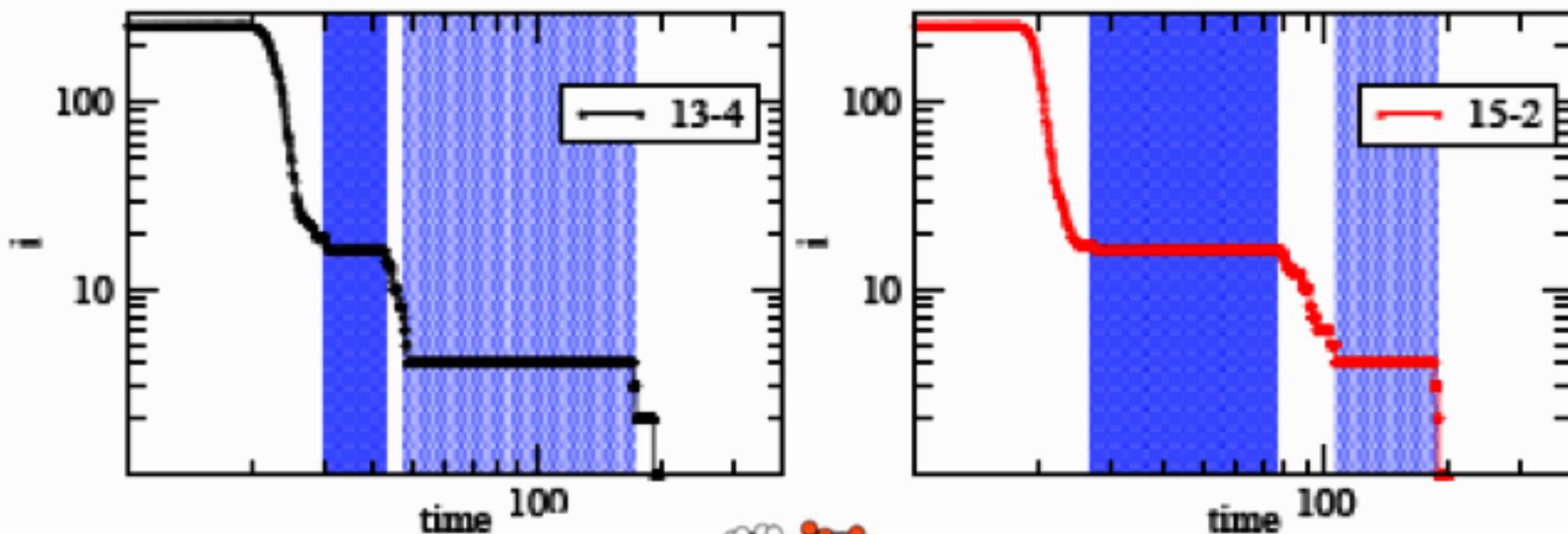
# New graphical representations

- ▶ Two nodes are connected if they are “synchronized”
- ▶ Dynamic connectivity matrix

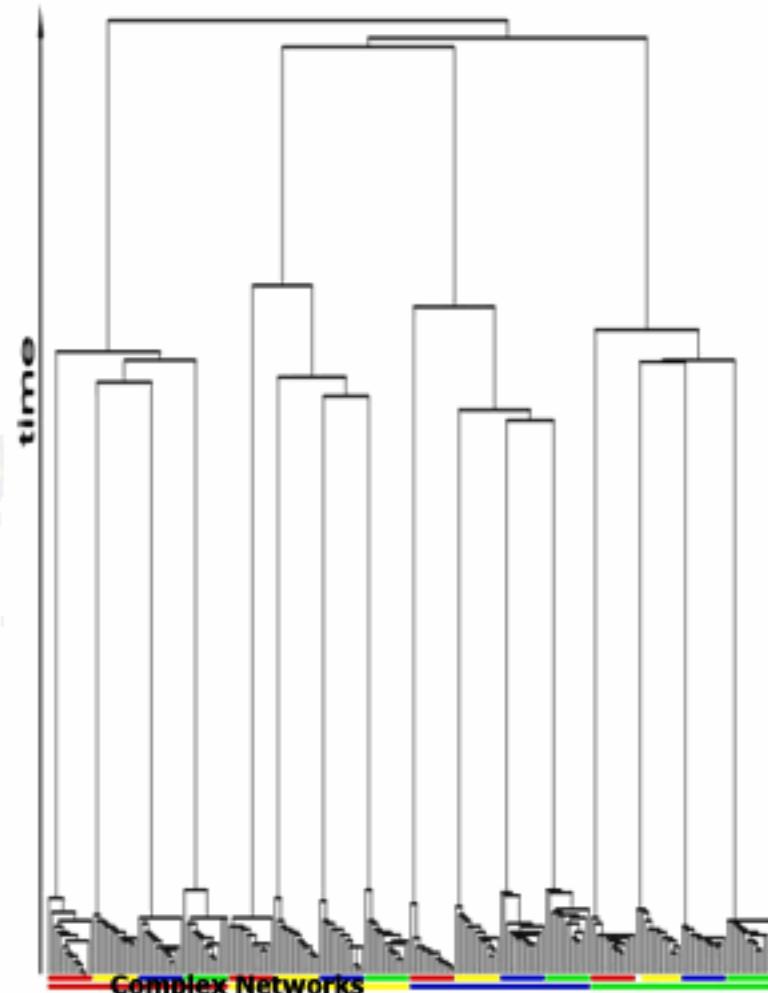
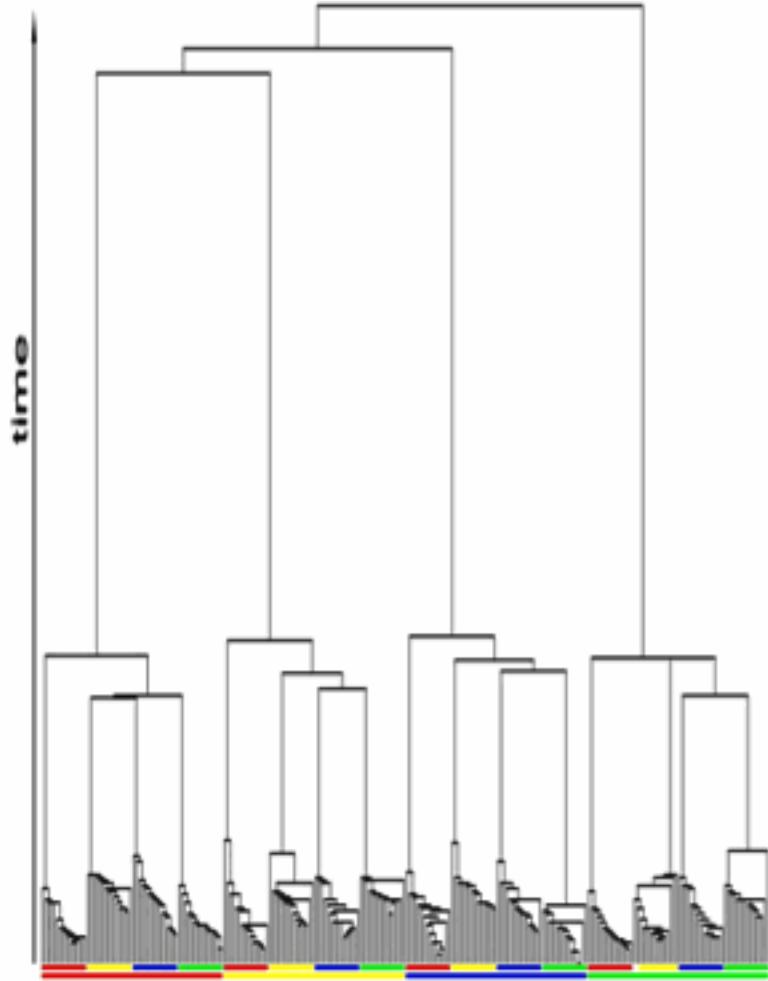
$$\mathcal{D}_t(T)_{ij} = \begin{cases} 1 & \text{if } \rho_{ij}(t) > T \\ 0 & \text{if } \rho_{ij}(t) < T \end{cases}$$

- ▶ Fixed time – moving threshold
- ▶ Fixed threshold – time evolution of the network
- ▶ Structure at different time scales

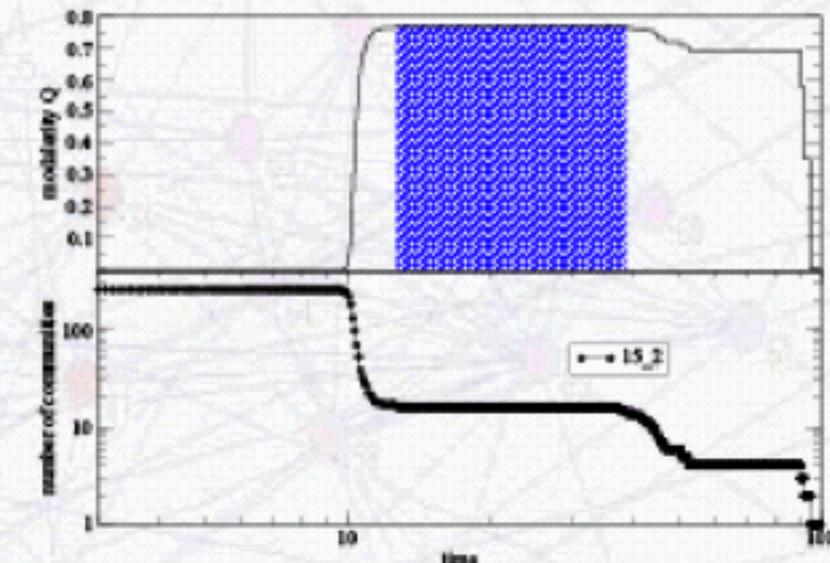
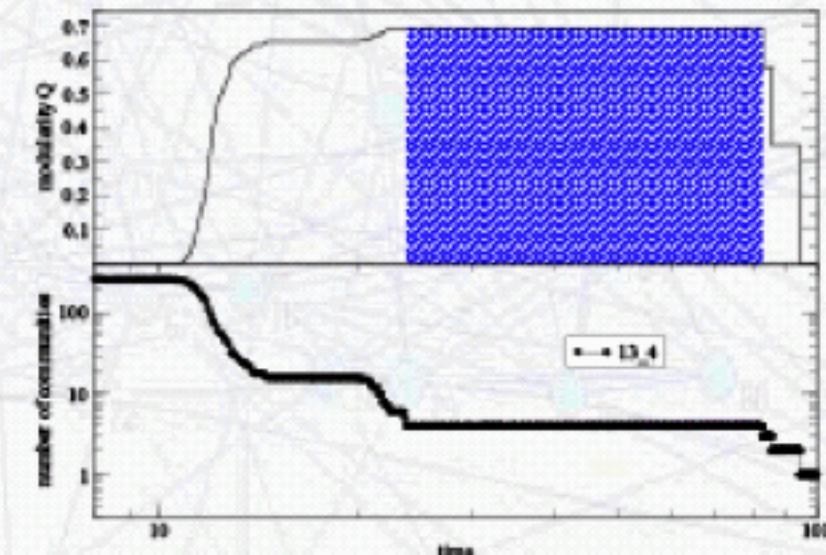
# Number of connected components



# Dendogram of connections



# Number of communities and modularity

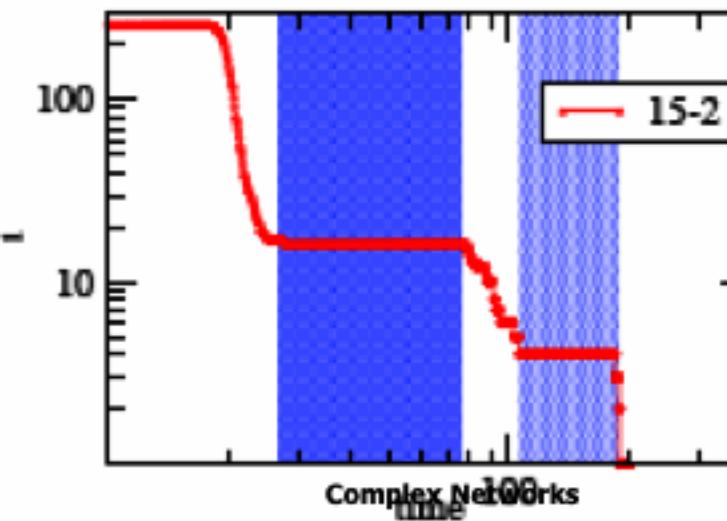
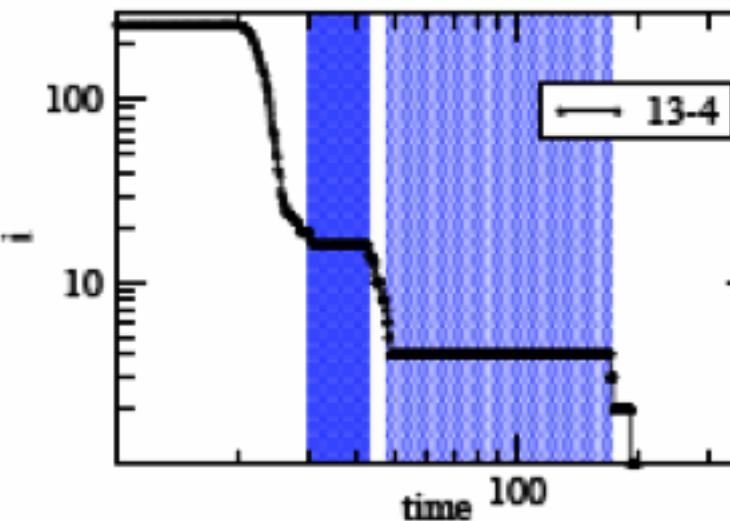
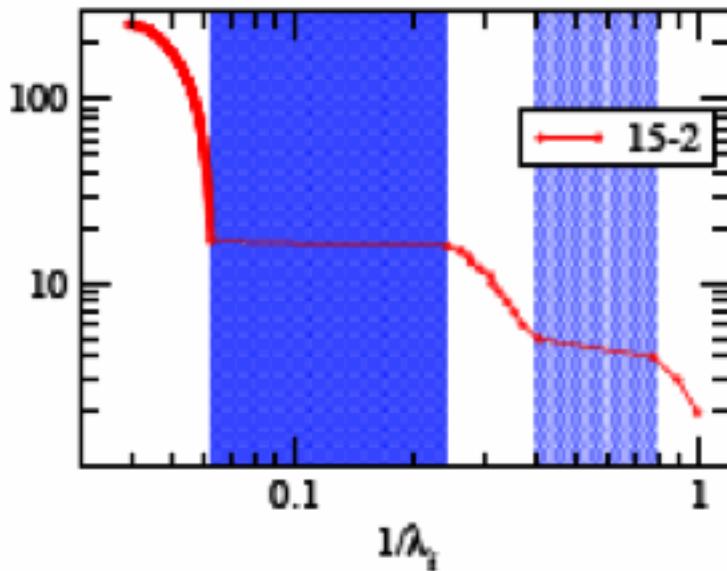
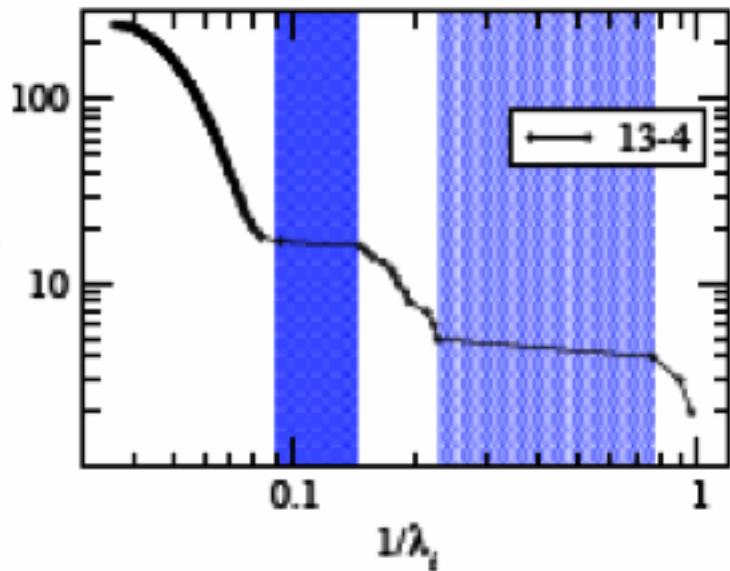


# Spectral properties

- ▶ Spectrum of the Laplacian matrix
- ▶ We order the eigenvalues

$$0 = \lambda_1 \leq \lambda_2 \leq \dots \leq \lambda_N$$

# Spectral versus dynamics



# Other (small) networks

- ▶ See the URL:

<http://www.ffn.ub.es/albert/synchro.html>

# Spectral properties

- ▶ Laplacian matrix
- ▶ Adjacency (connectivity) matrix

# Laplacian matrix

- Diffusion equation. Laplacian operator

$$\frac{dn(x, t)}{dt} = \nabla^2 n(x, t)$$

- Discrete Laplacian operator (1d)

$$\frac{dn_i(t)}{dt} = (n_{i+1}(t) - n_i(t)) - (n_i(t) - n_{i-1}(t))$$

$$\frac{dn_i(t)}{dt} = n_{i+1}(t) + n_{i-1}(t) - 2 \cdot n_i(t)$$

- Discrete Laplacian operator (2d)

$$\frac{dn_{i,j}(t)}{dt} = n_{i+1,j}(t) + n_{i-1,j}(t) + n_{i,j+1}(t) + n_{i,j-1}(t) - 4 \cdot n_{i,j}(t)$$

- In general

$$\frac{dn_i(t)}{dt} = \sum_i^N a_{i,j}n_j(t) - k_i \cdot n_i(t) = -L_{ij}n_j(t)$$

- Where we have introduced the Laplacian matrix

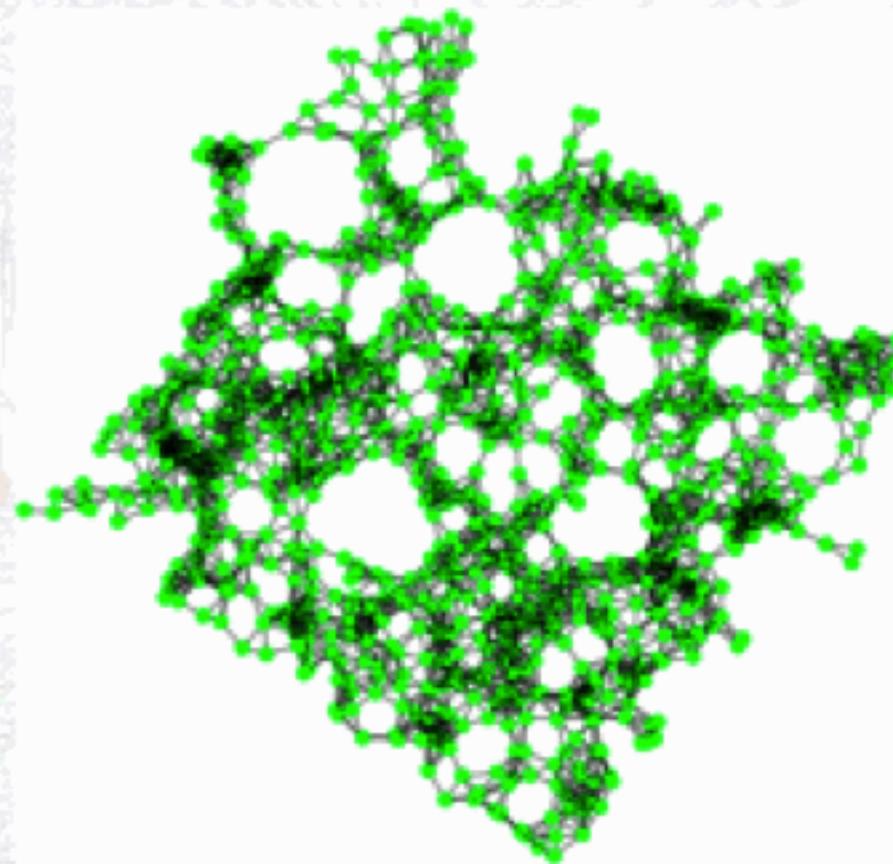
$$L_{ij} = k_j \delta_{ij} - a_{ij}$$

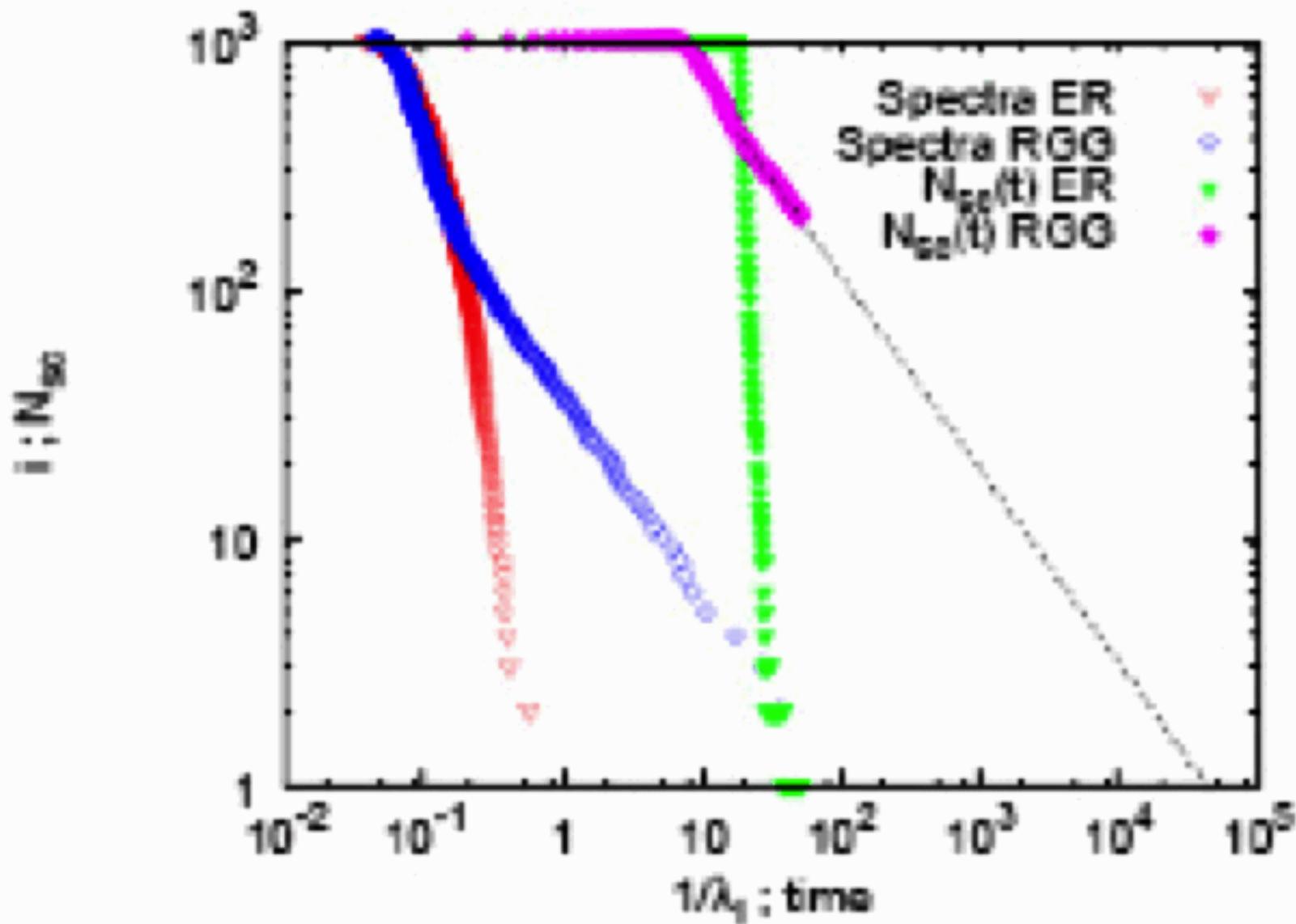
# Properties of Laplacian matrix

- ▶ Important on dynamical properties
- ▶ Discrete spectrum: eigenvalues and eigenvectors
- ▶ Ordered  $0 = \lambda_1 \leq \lambda_2 \leq \dots \leq \lambda_N$
- ▶ Number of 0 eigenvalues is equal to the number of (dis)connected components
- ▶  $\lambda_2$  is related to the time the system needs to be synchronized. Intuitively, when it is zero there are at least two disconnected components and the system will never synchronize

# Networks without community structure

Random Geographic Graph (RGG)





# Mobility creates complex time-dependent networks

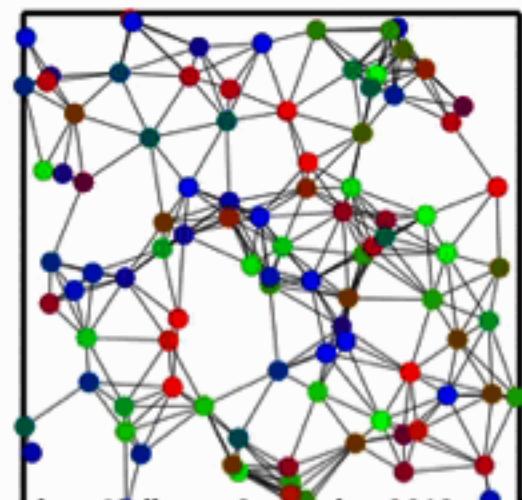
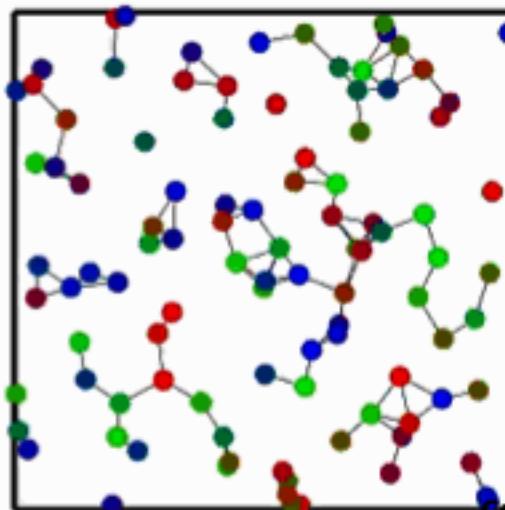
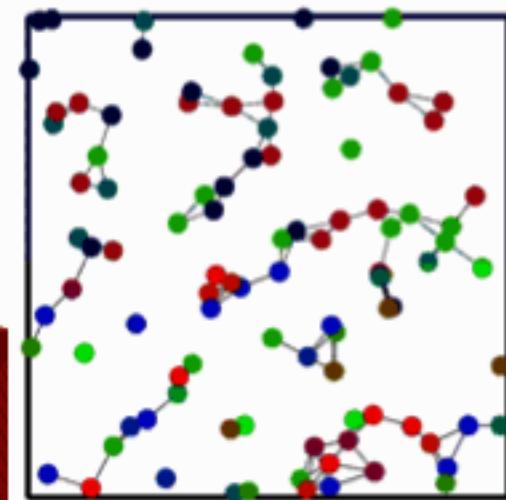
- ▶ Agents that move create contact networks on its way
- ▶ Interact with nearest neighbors:
  - range of interaction
  - pre-define a number of interacting

# Movies



kuramoto oscillators

Search



# Synchronization in complex nets

- ▶ **Synchronization in complex network:**

Interplay between topology and dynamics

A. Arenas, A.D.-G., J. Kurths, Y. Moreno, C. Zhou, Phys. Rep. 469, 93 (2008)

- ▶ **Spectral analysis:**

- ▶ **In most of the previous studies, network topology is fixed**

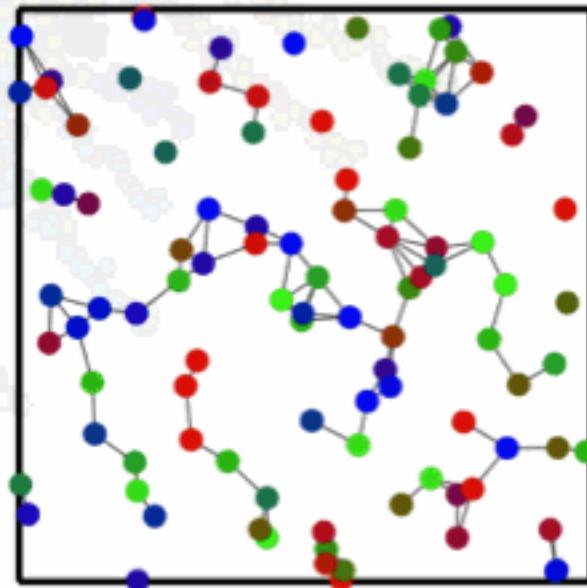
What happens if topology changes in time?  
Is spectral approach possible?

# Kuramoto Model

N. Fujiwara, J. Kurths, A.D-G, PRE (2011)

$$\varphi_i(t + \tau_P) = \varphi_i(t) + \sum_{j=1}^N \sigma(d_{ij}) \sin(\varphi_j(t) - \varphi_i(t))$$

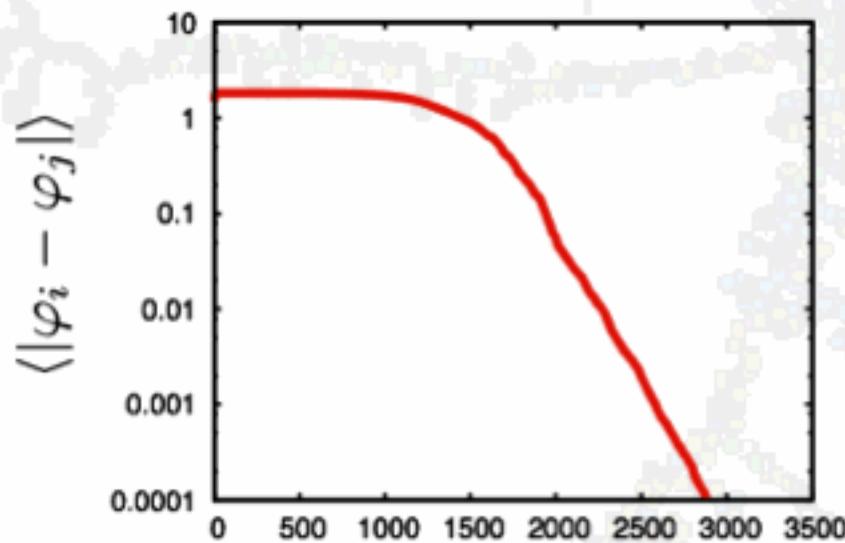
$$\sigma(d_{ij}) = \begin{cases} \sigma & (d_{ij} < d) \\ 0 & (d_{ij} > d) \end{cases}$$



# Applet

- ▶ Java applet simulation
- ▶ [http://complex.ffn.ub.es/~albert/mobile/  
Kuramoto.html](http://complex.ffn.ub.es/~albert/mobile/Kuramoto.html)

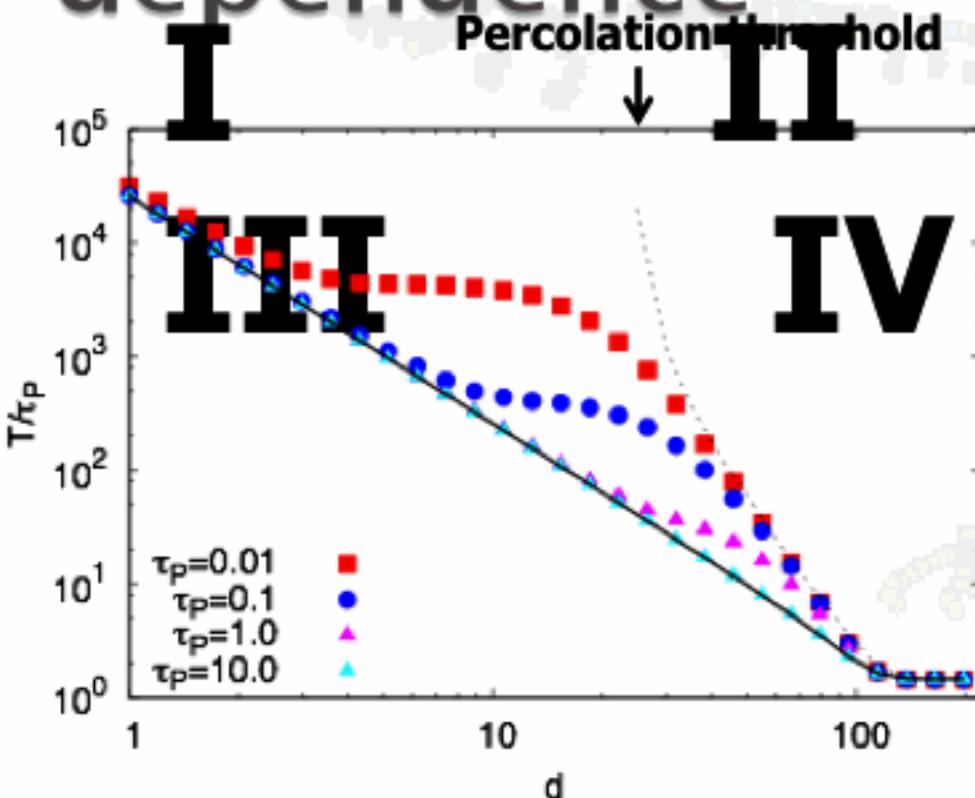
# Time evolution



time

$$T/\tau_P$$

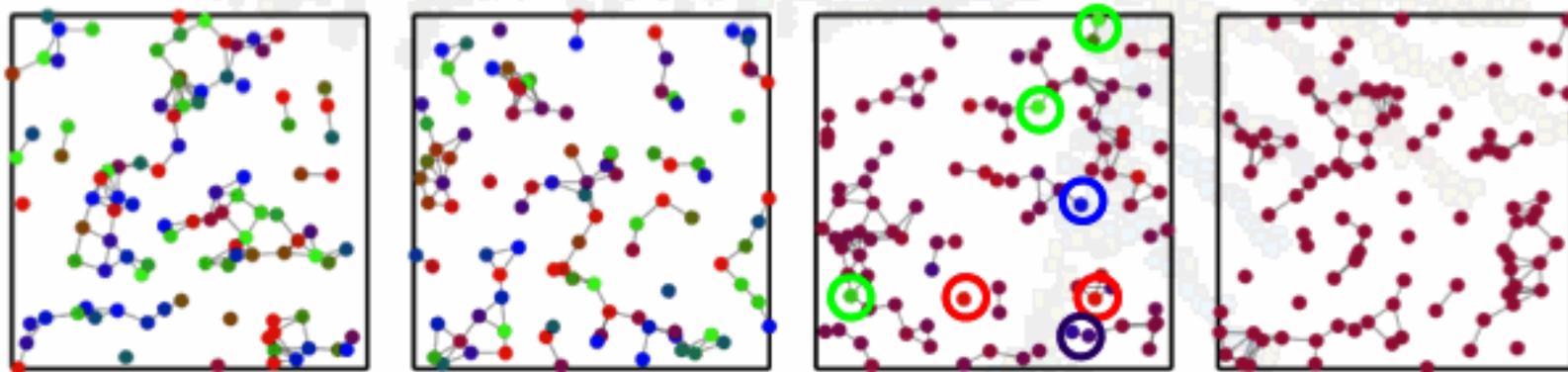
# $d$ (interaction range) dependence



$N = 100, L = 200, v = 10, \tau_M = 1.0, \sigma = 0.005$

- ▶ I: fast switching
- ▶ II: multi cluster
  - local synchronization
  - slow topology change
- ▶ III: single cluster
  - local synchronization
- ▶ IV: complete graph

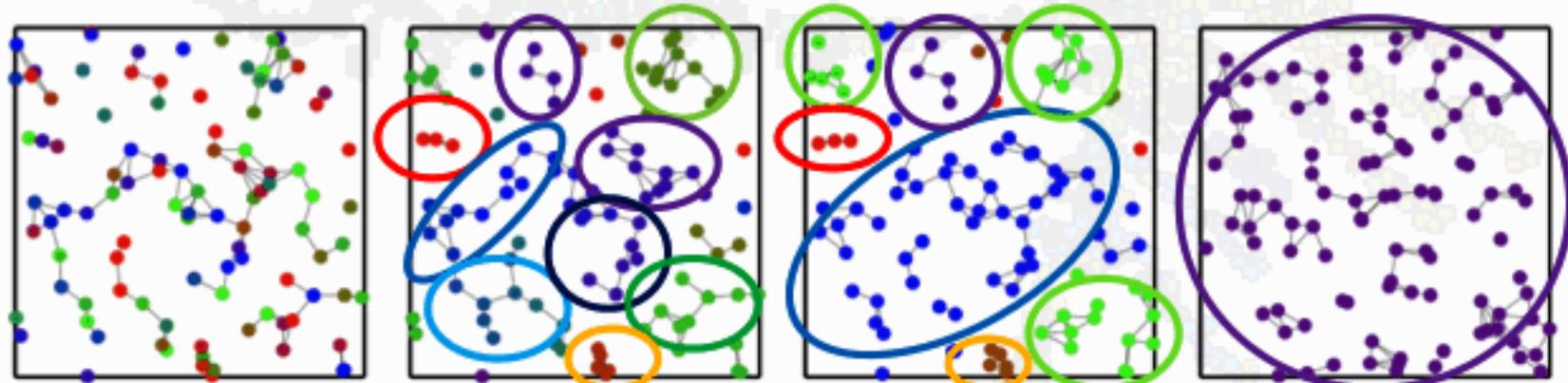
# Region I: global synchronization



time →

- ▶ Global synchronization
- ▶ Non-synchronized oscillators are isolated
- ▶ Fast switching approximation describes the synchronization dynamics
- ▶ Effect of agent dynamics disappear

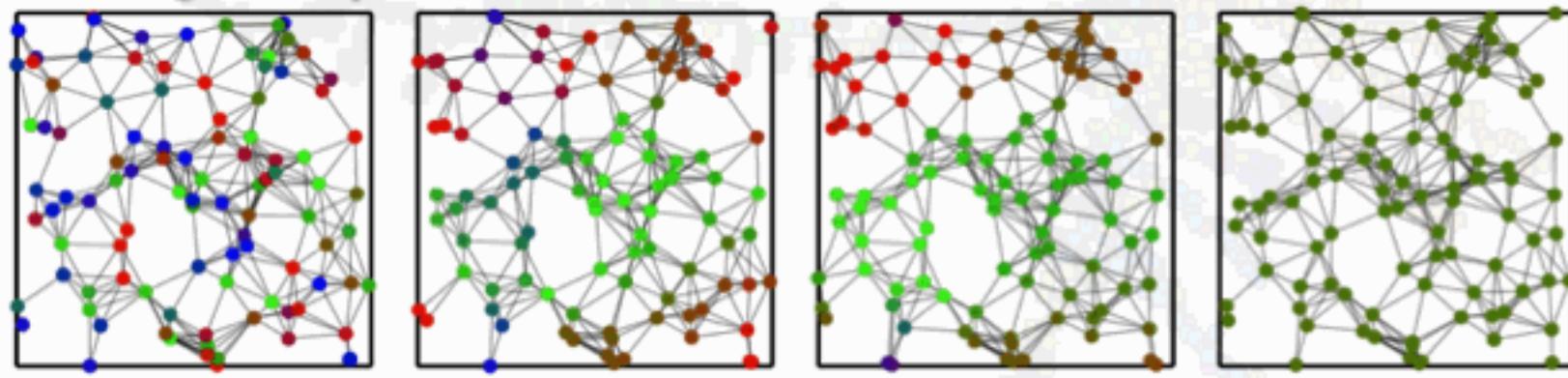
## Region II: local synchronization multiple cluster



- ▶ Below percolation threshold
- ▶ Synchronization inside cluster takes place at first
- ▶ Global synchronization is achieved due to the motion of agents
- ▶ Characteristic time is dominated by cluster interaction (agent motion)
- ▶ Slower than fast switching approximation

tir

# Region III: single cluster small jump

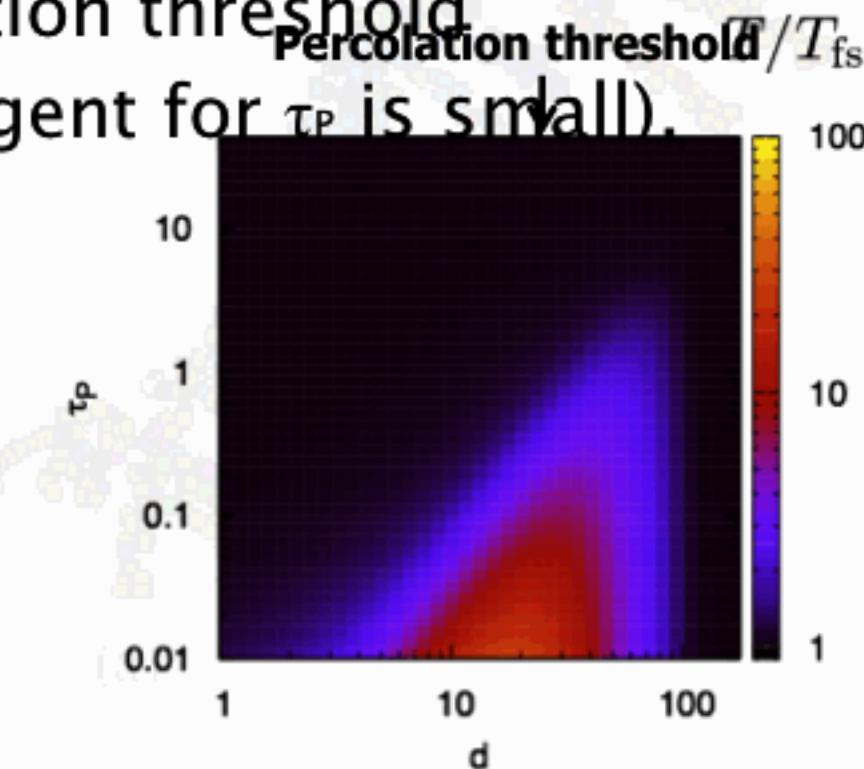


- ▶ Above percolation threshold, whole network is connected
- ▶ Local synchronization takes place at first
- ▶ Second smallest eigenvalues of instantaneous Laplacian matrices  $\lambda_2(t)$  dominate the synchronization dynamics
- ▶  $T \sim 1/(\sigma \langle \lambda_2 \rangle)$
- ▶ Slower than fast switching approximation

tir

# Deviation from fast switching approximation in d- $\tau_P$ plane

- ▶  $T$  deviates from  $T_{fs}$  for  $d$  value near percolation threshold
- ▶ Smaller  $\tau_P$  (jump of agent for  $\tau_P$  is small), larger  $T/T_{fs}$



# Dynamic transition: local to global synchronization

- Number of steps for a cluster to internally synchronize

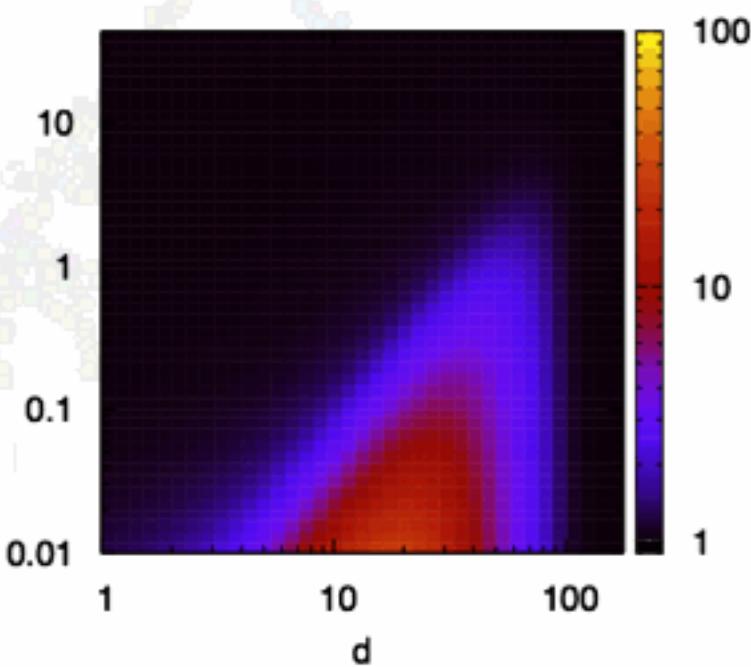
$$n_s = \frac{1}{\sigma \lambda_2^S(d)},$$

- Number of steps for an agent to leave a cluster**

$$n_m = \frac{\xi^2(d)}{v^2 \tau_M \tau_P}.$$

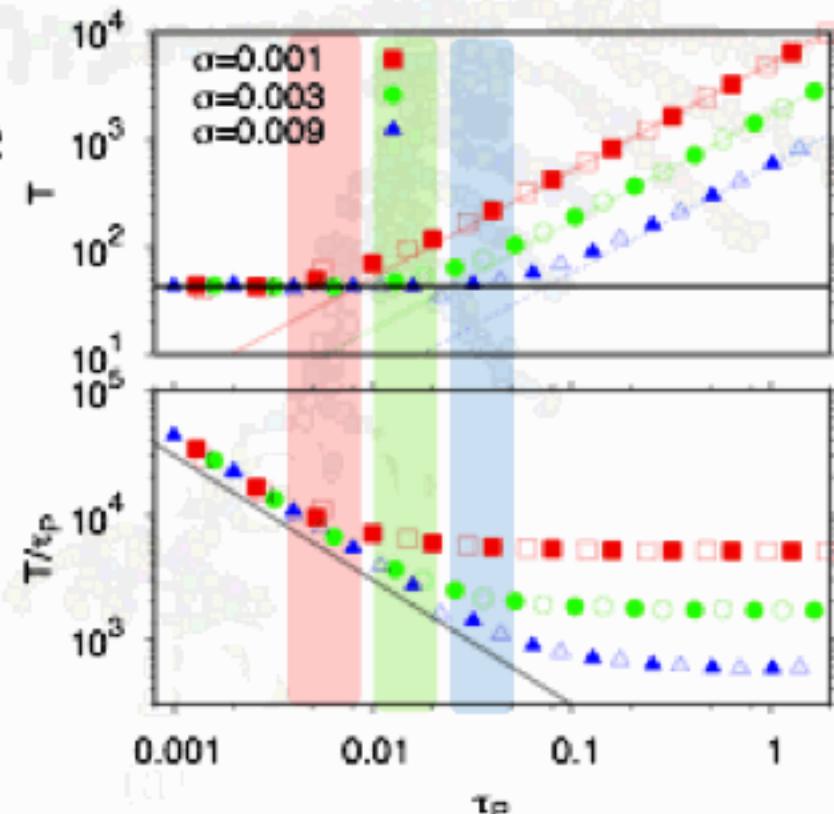
# Transition

$$\eta = \frac{n_m}{n_s} = \frac{\sigma f(d)}{v^2 \tau_M \tau_P}.$$



# Dependence of characteristic time on signal interval (below percolation)

- For large  $\tau_p$ ,  $T$  is large (slow synchronization achievement)
- For small  $\tau_p$ , more number of signals are required for synchronization (low efficiency)



which both fast synchronization and

# Matrix product for linearized equation

- When the phase difference is small, the linearized equation describes the synchronization dynamics

$$\varphi_i(t + \tau_P) = \varphi_i(t) - \sigma \sum_{j=1}^N L_{ij}(t) \varphi_j(t),$$

In our case **Laplacian matrix depends on time**

- consider the transformation of the normal modes (eigenmode of L)

$$\varphi_j(t) = \sum_{k=1}^N U_{jk}(t) \theta_k(t), \quad \sum_{k=1}^N L_{jk}(t) \theta_k(t) = \lambda_j(t) \theta_j(t)$$

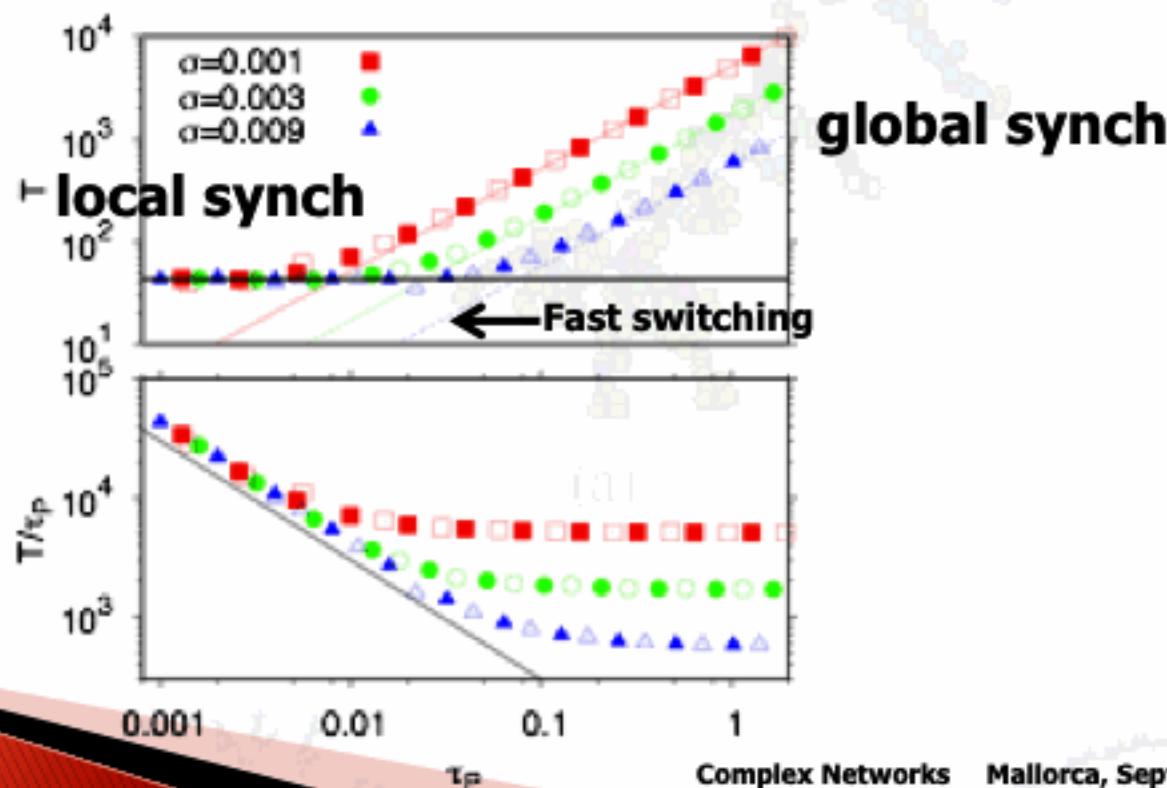
- we get the time evolution of the normal modes as

$$\begin{aligned}\theta_l(t + \tau_P) &= \sum_{i,k} U_{li}^T(t + \tau_P) U_{ik}(t) [1 - \sigma \lambda_k(t)] \theta_k(t) \\ &\equiv \sum_k \textcolor{red}{O_{lk}(t)} \textcolor{blue}{[1 - \sigma \lambda_k(t)]} \theta_k(t)\end{aligned}$$

**agent mobility oscillator dynamics**

# Matrix product for linearized equation

- Finally, we get  $\theta_{k_n}(t + n\tau_P) = \prod_{q=0}^{n-1} \left[ \sum_{k_q=1}^N O_{k_{q+1} k_q} (1 - \sigma \lambda_{k_q}) \right] \theta_{k_0}(t)$
- Compare empirical T with second smallest eigenvalue of the product of matrices (independent way), and they coincide for any value of the parameters even when fast switching approximation does not work



# Derivation of fast switching approximation

$$\theta_{k_n}(t + n\tau_P) = \prod_{q=0}^{n-1} \left[ \sum_{k_q=1}^N O_{k_{q+1} k_q} (1 - \sigma \lambda_{k_q}) \right] \theta_{k_0}(t)$$

$$\prod_{q=1}^n (1 - \sigma \lambda_{l_q}) \approx e^{n \langle \log(1 - \sigma \lambda) \rangle}$$

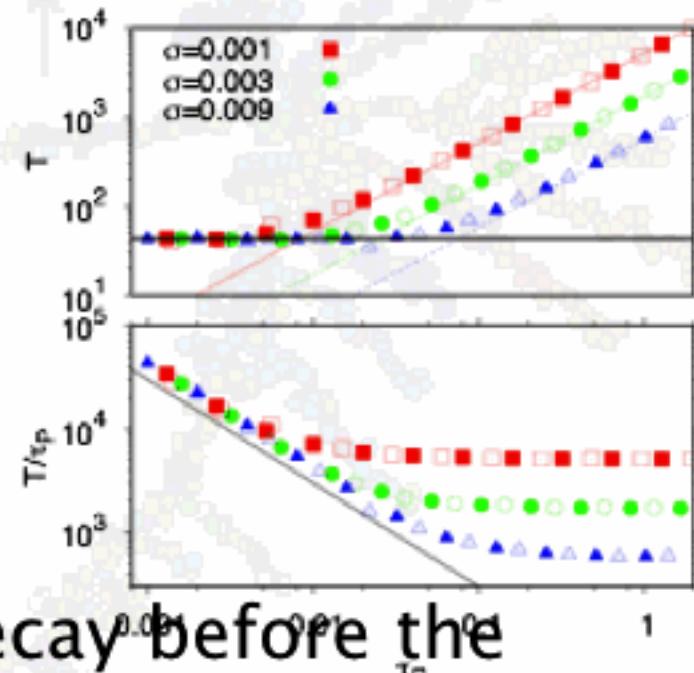
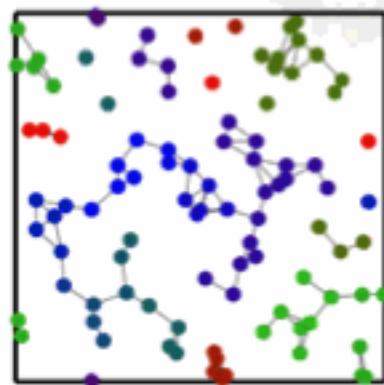
$$T = -\tau_P / \langle \log(1 - \sigma \lambda) \rangle$$

$$\frac{\tau_P}{T} = \sigma \langle \lambda \rangle$$

$$\frac{\tau_P}{T} = (N-1)\sigma\rho$$

$$\rho = \begin{cases} \frac{\pi d^2}{L^2} & d < \frac{L}{2} \\ L\sqrt{4d^2 - L^2} + d^2[\pi - 4\cos^{-1}(\frac{L}{2d})] & \frac{L}{2} < d < \frac{L}{\sqrt{2}} \\ 1 & d > \frac{L}{\sqrt{2}} \end{cases}$$

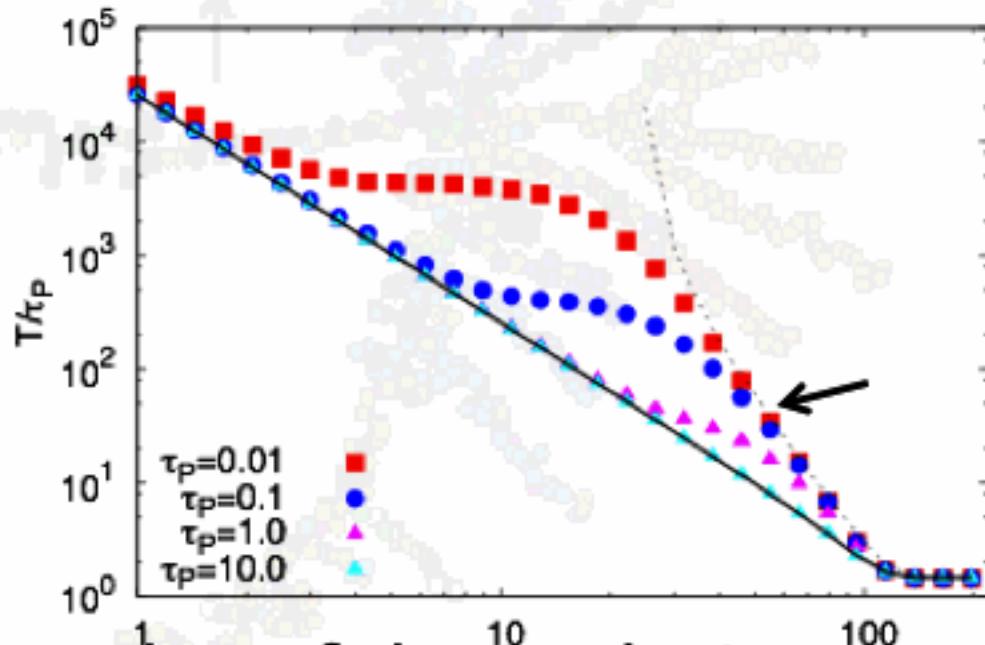
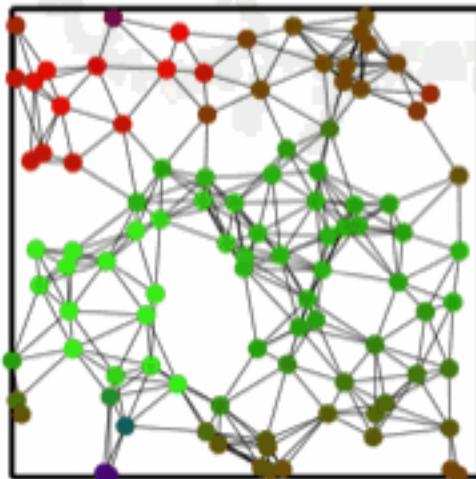
# Multiple cluster local synchronization



- Non-zero eigenmodes decay before the topology changes. Zero eigenmodes dominates global dynamics
- Matrix product approximation suggests that efficiency scales as

$$\frac{T}{\tau_P} \propto (v^2 \tau_M \tau_P)^{-1}$$

# Local synchronization single cluster



- ▶ Second smallest eigenvalue of the Laplacian matrix dominates the instantaneous decay rate
- ▶ When topology change is slow, dynamics is governed by the average of second smallest eigenvalue (algebraic connectivity)