

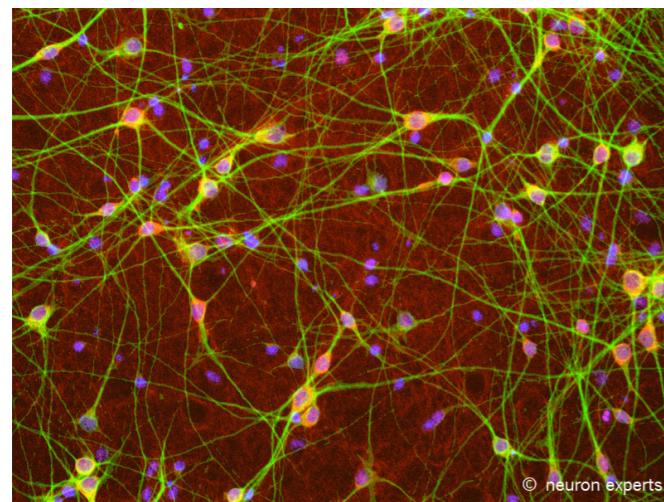


# Dynamics, Synchronization and Inverse problem in Neural Networks with synaptic plasticity

Matteo di Volo  
Group for Neural Theory  
ENS, Paris

From Microscopic to collective dynamics in Neural Circuits  
Dresden, 2016

# Neural Network



$Y(t)$  Electrical activity

Delta ( $\delta$ ) 0.5-4 Hz

Infants,  
sleeping adults

Spikes

Epilepsy -  
petit mal

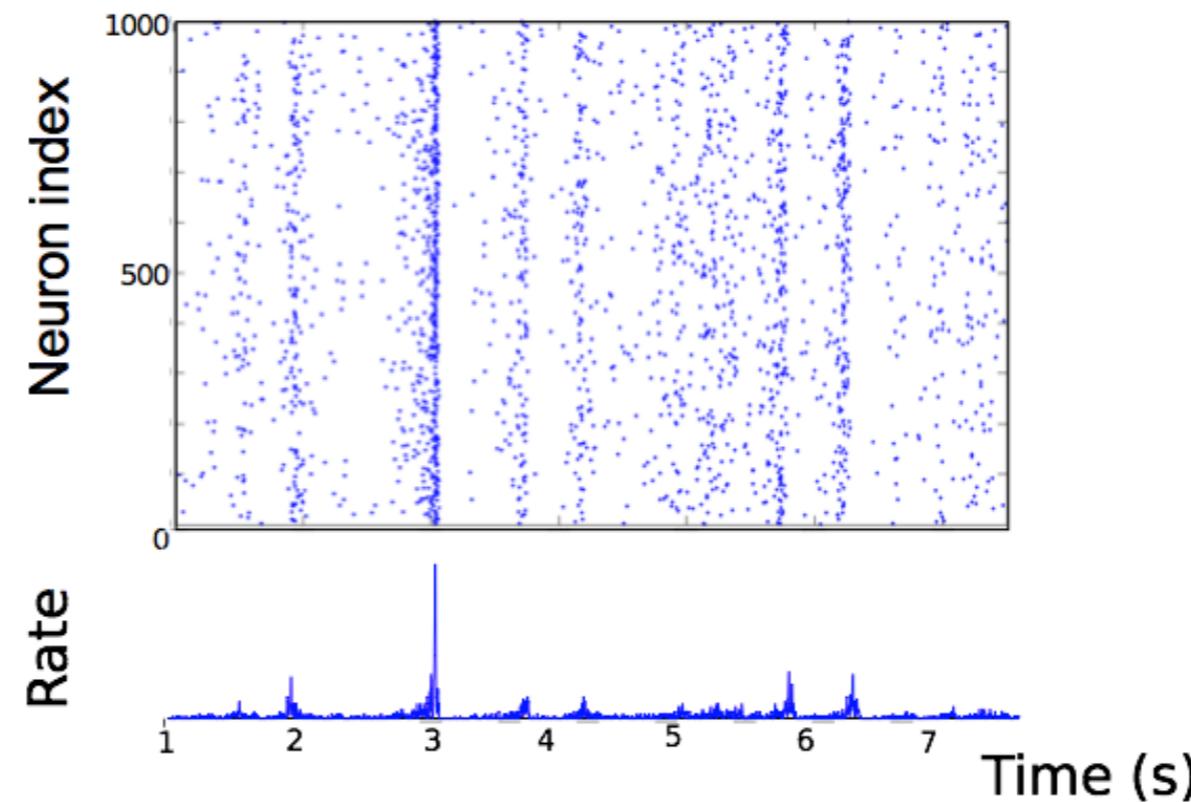
3 Hz

$V [\mu V]$

0

0 1 2 3 4 Time [s]

# Avalanches activity



# Outline

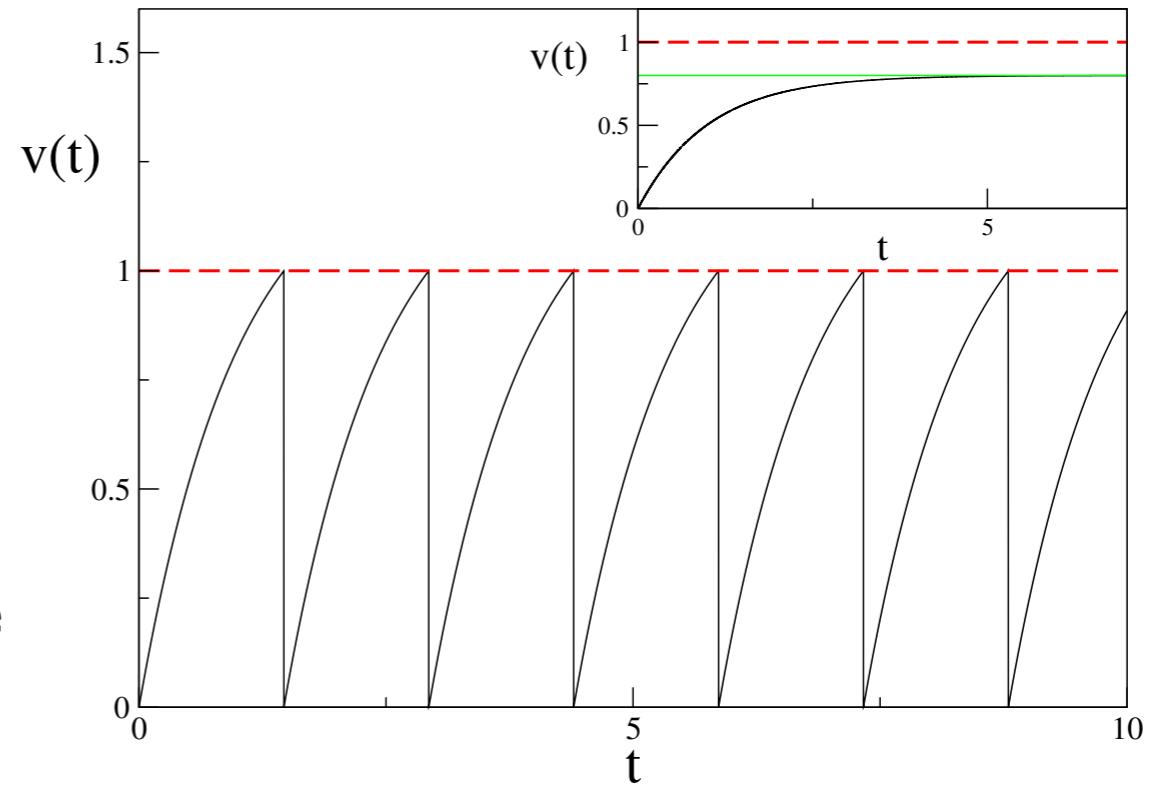
- LIF neurons with synaptic plasticity
- Heterogeneous Mean Field
- Partially synchronous and asynchronous regimes
- From collective activity to network structure

# LIF Neuronal Model

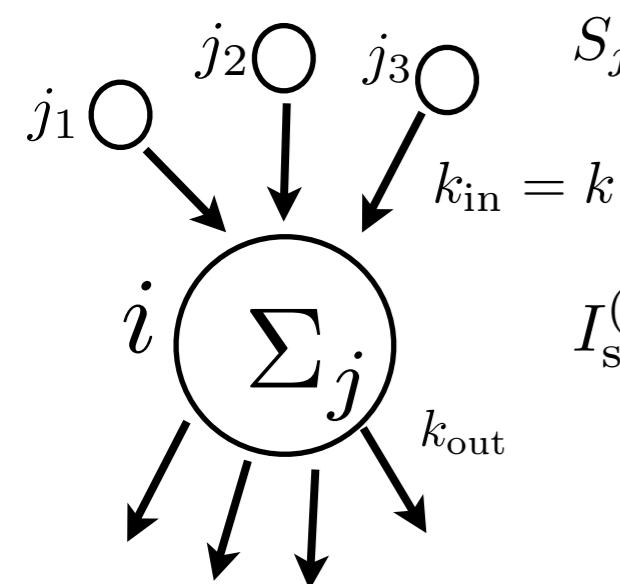
$$\dot{v} = a - v + I_{\text{syn}}$$

$$v > v_{\text{th}} = 1 \quad \begin{cases} \text{spike} \\ v = 0 \end{cases}$$

$a > 1$   
 ↓  
**Spiking  
Regime**



**Short term plasticity:  
TUM model for excitatory neurons**



$$S_j(t) = \sum_{n|t_n < t} \delta(t - t_n(j))$$

$$I_{\text{syn}}^{(i)}(t) = \tilde{g} \sum_{j \in \text{presyn.}i} y_j(t)$$

$$\dot{y}_j = -\frac{y_j}{\tau_{\text{in}}} + ux_j S_j$$

$$\dot{x}_j = \frac{z_j}{\tau_r} - ux_j S_j$$

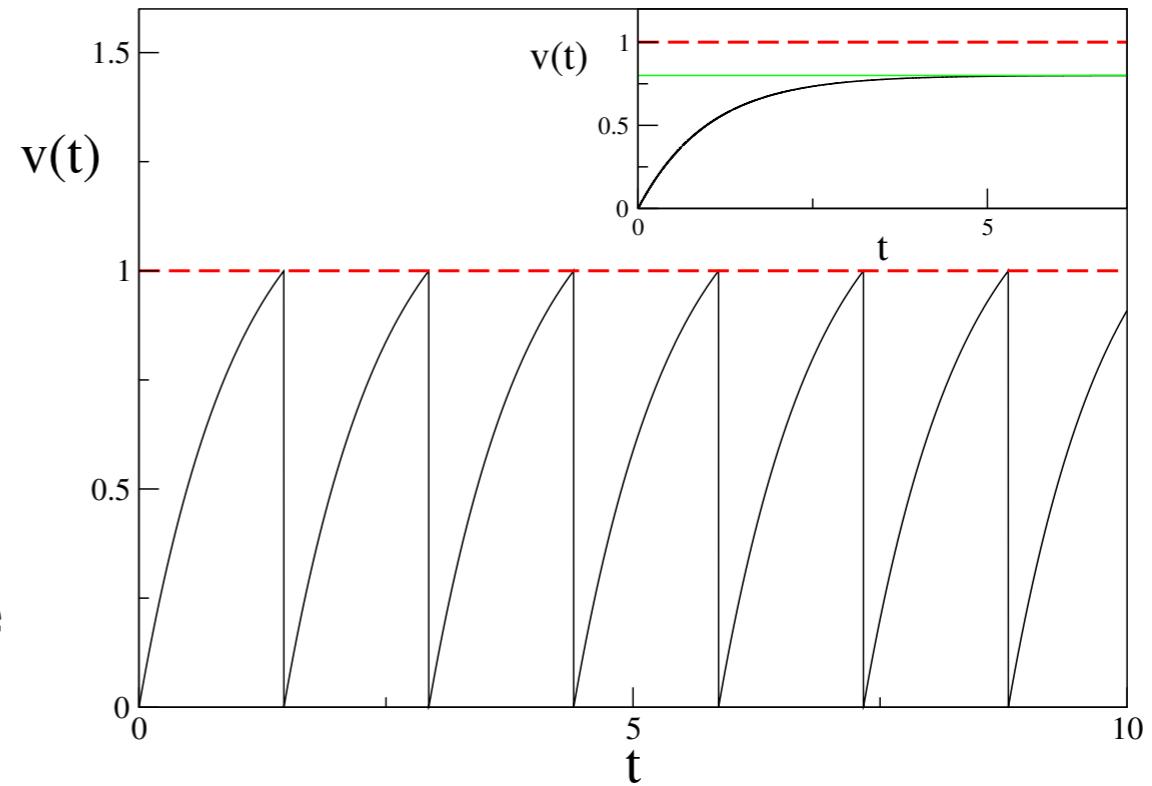
$$x_j + y_j + z_j = 1$$

# LIF Neuronal Model

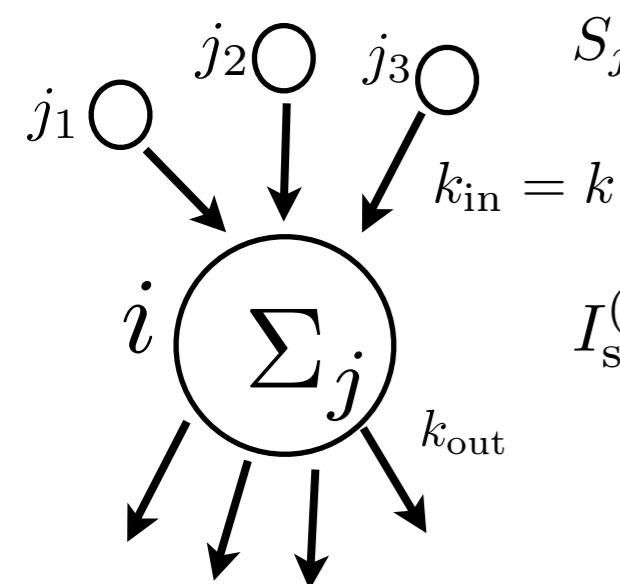
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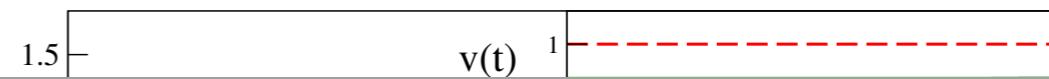
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$$x_j + y_j + z_j = 1$$

# LIF Neuronal Model

$$\dot{v} = a - v + I_{\text{syn}}$$



*Finite size network of  $N$  neurons*

$$\dot{v}_i = a - v_i + \frac{g}{N} \sum_{j \neq i} g_{ij} y_j$$

$$\tau_{\text{in}} = 0.2 ; \tau_r = 26.6$$

$$u = 0.5 ; a = 1.3$$

$$g = 30$$

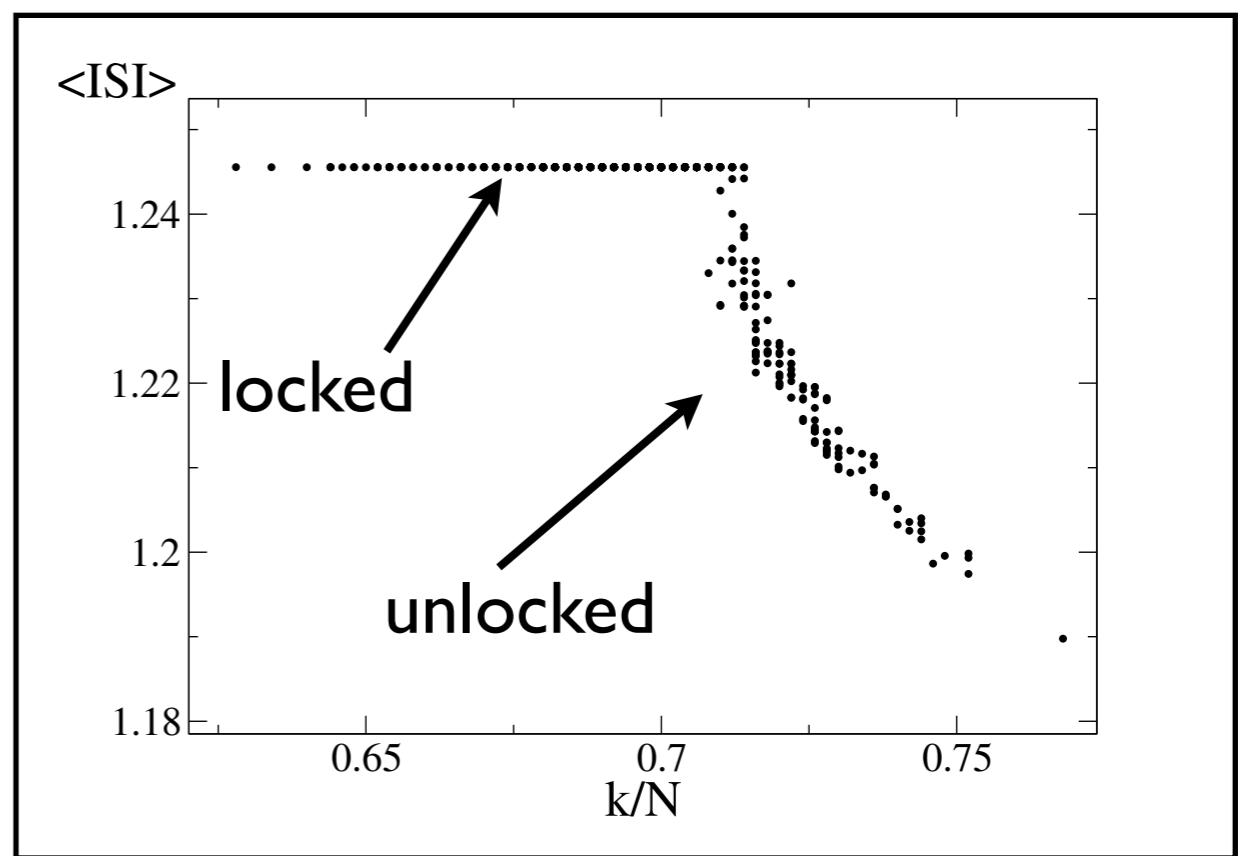
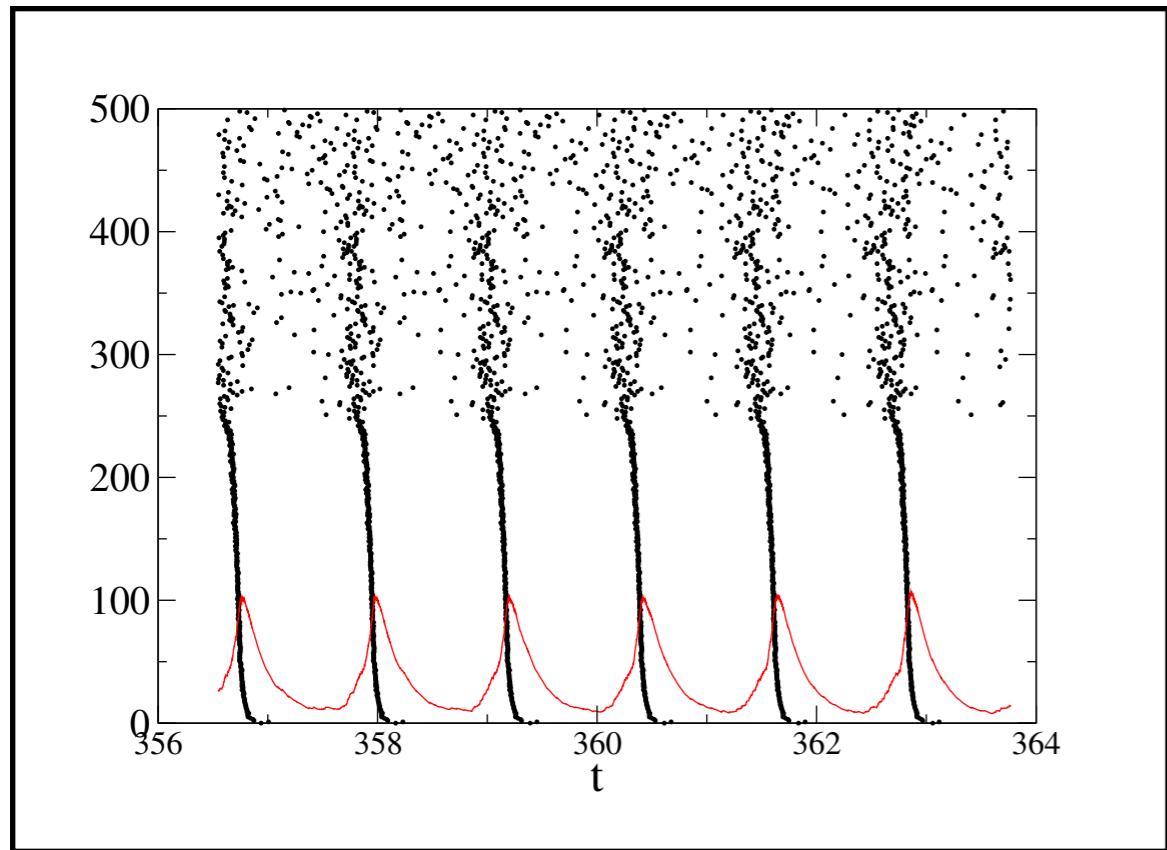
$$\dot{y}_i = -\frac{y_i}{\tau_{\text{in}}} + ux_i S_i$$

$$\dot{x}_i = \frac{1 - x_i - y_i}{\tau_r} - ux_i S_i$$

# Erdös–Renyi random Network

each link connected with probability  $p$

$$\text{large } N : P_N(k) = G(Np, Np(1 - p))$$



$$Y(t) = \frac{1}{N} \sum_i y_i(t)$$

$$\tilde{k} = \frac{k}{N}$$

# Thermodynamic limit

Erdös–Renyi:  $P(\tilde{k}) = G(p, p(1 - p)/N)$

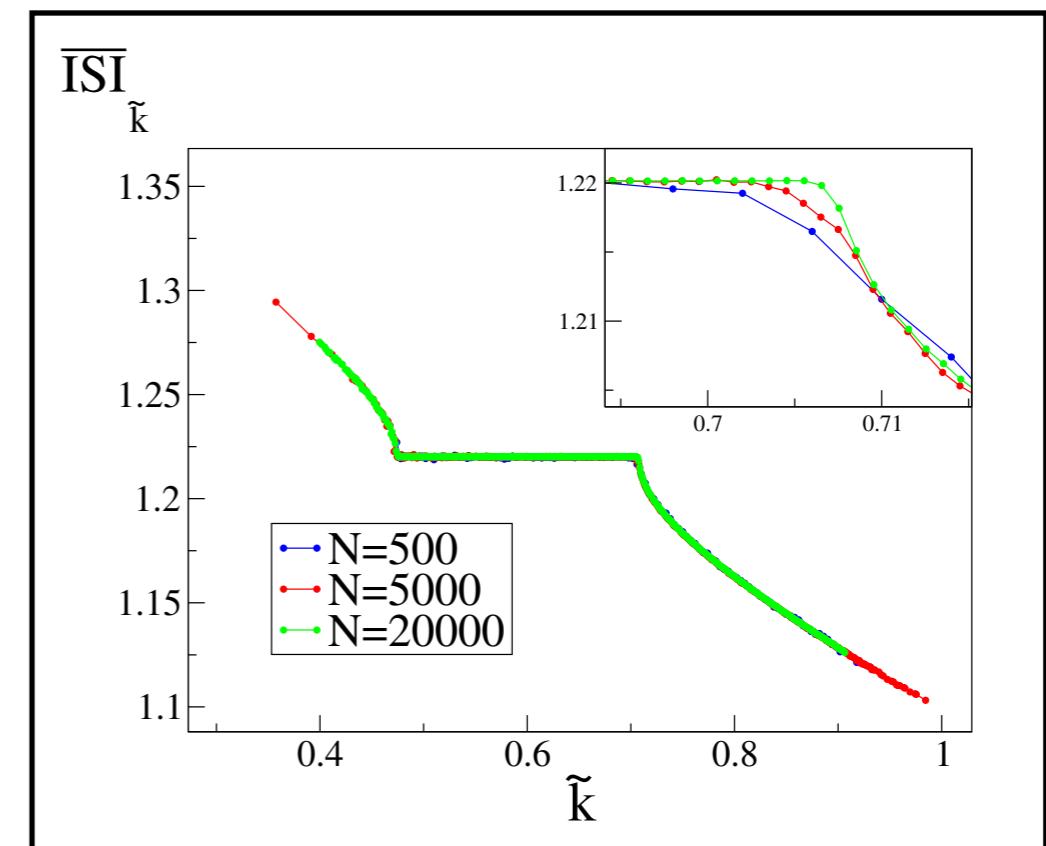
fluctuations  $\sigma_{\tilde{k}} \sim 1/\sqrt{N}$

Dynamics ( $N \rightarrow \infty$ )  $\neq$  Dynamics (finite  $N$ )

## New Network construction

$P(\tilde{k})$  fixed

extract  $\tilde{k}_i$  from  $P(\tilde{k})$   
&  
assign randomly  $N\tilde{k}_i$  inputs



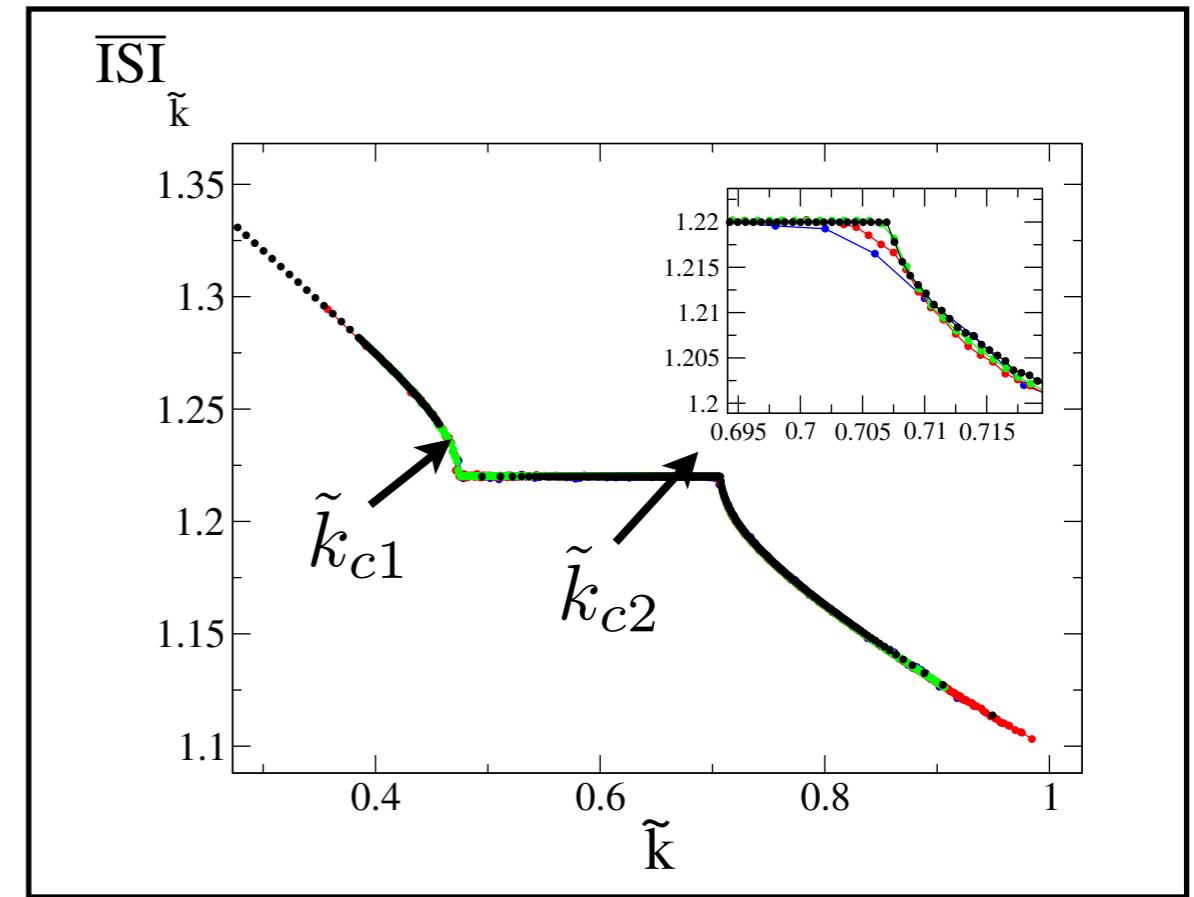
Gaussian  $P(\tilde{k})$ :  $\langle \tilde{k} \rangle = 0.7$   $\sigma_{\tilde{k}} = 0.06$

# Heterogeneous Mean Field

$$\dot{v}_i = a - v_i + \frac{g}{N} \sum_j \epsilon_{ij} y_j$$

$$\frac{1}{k_i} \sum_j g_{ij} y_j(t) \underset{\text{MF}}{\simeq} \frac{1}{N} \sum_j y_j(t) = Y(t) \implies F_i(t) = \frac{g}{N} \sum_j g_{ij} y_j(t) \rightarrow g \tilde{k}_i Y(t)$$

$$\begin{aligned}\dot{v}_{\tilde{k}}(t) &= a - v_{\tilde{k}}(t) + g \tilde{k} Y(t) \\ \dot{y}_{\tilde{k}}(t) &= -\frac{y_{\tilde{k}}(t)}{\tau_{\text{in}}} + u x_{\tilde{k}}(t) S_{\tilde{k}}(t) \\ \dot{x}_{\tilde{k}}(t) &= \frac{(1 - y_{\tilde{k}}(t) - x_{\tilde{k}}(t))}{\tau_r} - u x_{\tilde{k}}(t) S_{\tilde{k}}(t) \\ Y(t) &= \int_0^1 P(\tilde{k}) y_{\tilde{k}}(t) d\tilde{k}\end{aligned}$$



Gaussian  $P(\tilde{k})$ :  $\langle \tilde{k} \rangle = 0.7$   $\sigma_{\tilde{k}} = 0.06$

# Stability Analysis

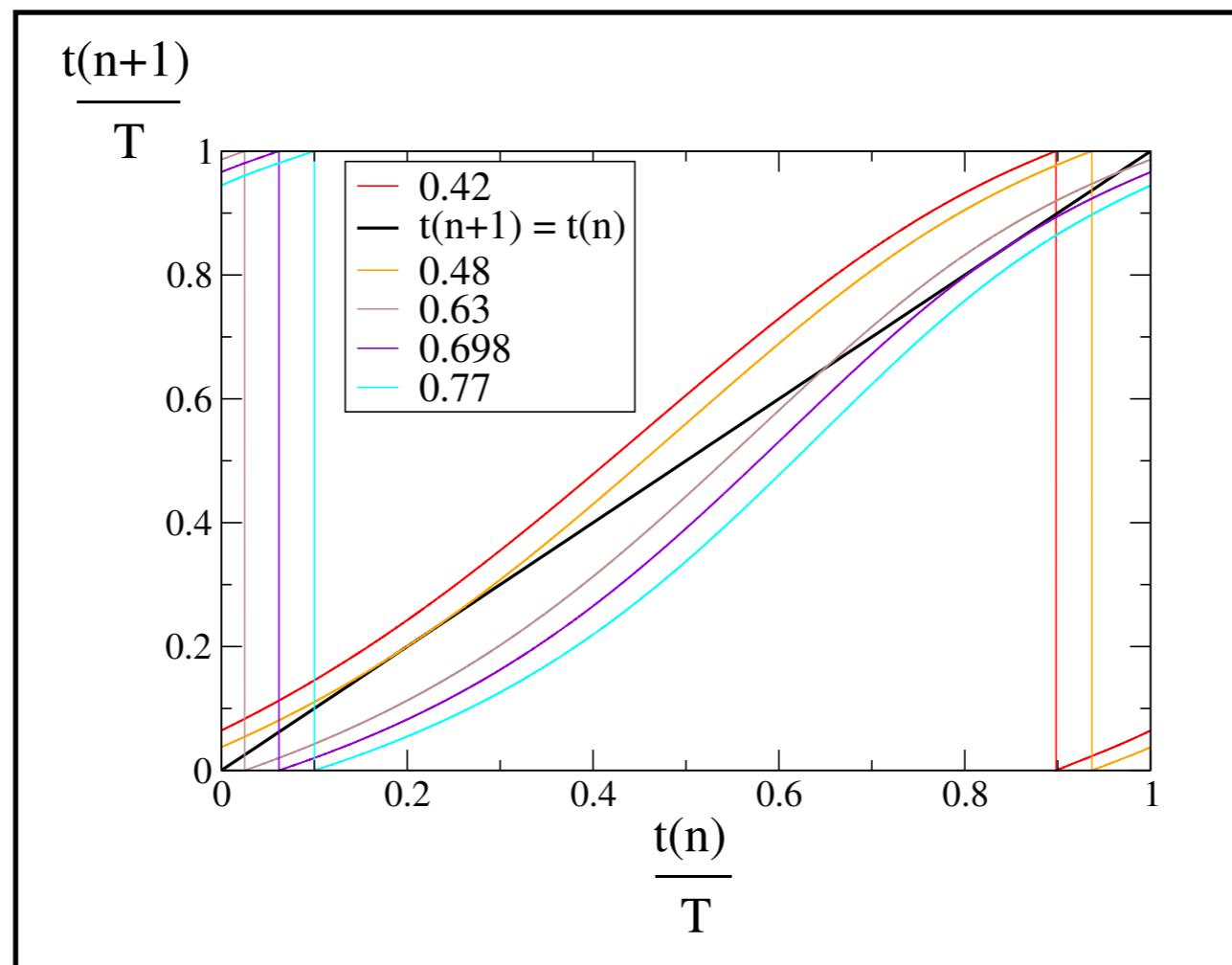
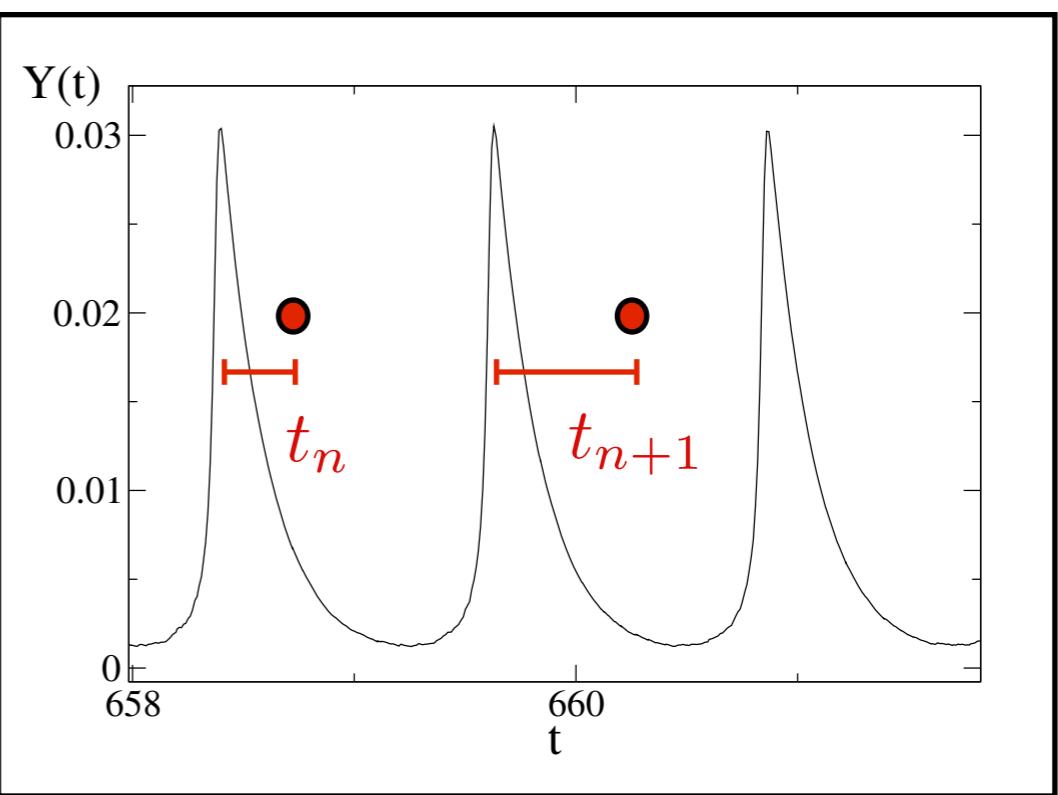
Take  $Y(t)$  from HMF simulation

$Y(t)$  periodic of period  $T$

$$\dot{v}_{\tilde{k}}(t) = a - v_{\tilde{k}}(t) + g \tilde{k} Y(t)$$

Obtain a map:

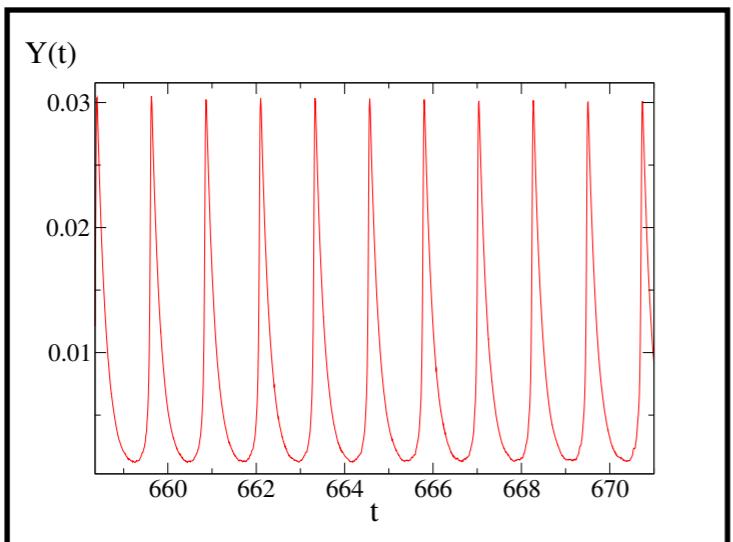
$$t_{n+1}(\tilde{k}) = M_{\tilde{k}} t_n(\tilde{k})$$



# *From partial synchrony to asynchronous phase: the role of degree disorder*

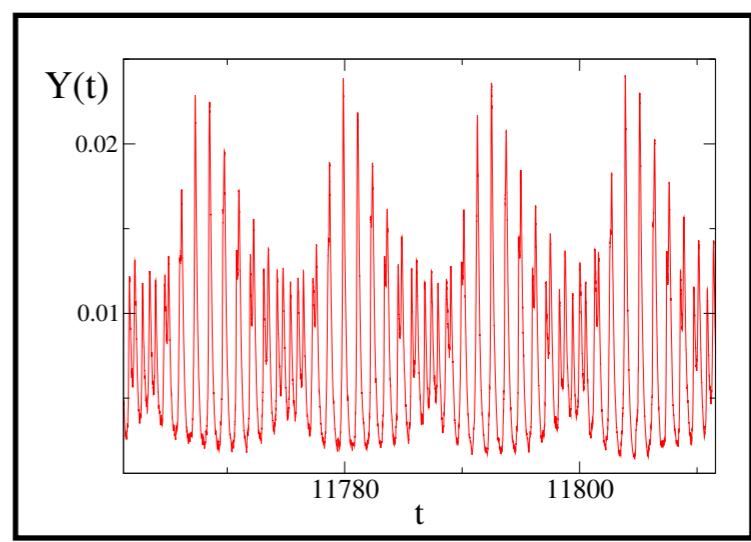
single peak  $P(\tilde{k})$

$\Omega$



double peak  $P(\tilde{k})$

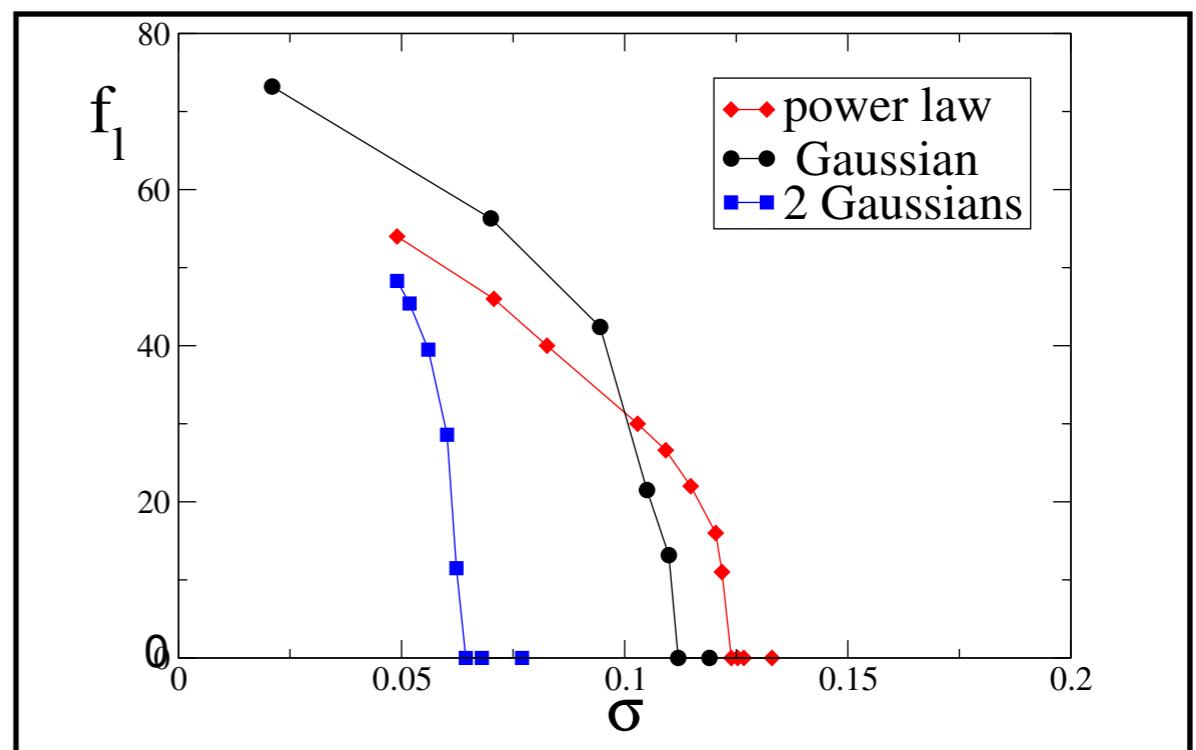
$(\Omega_1, \Omega_2)$



## Transition to asynchrony

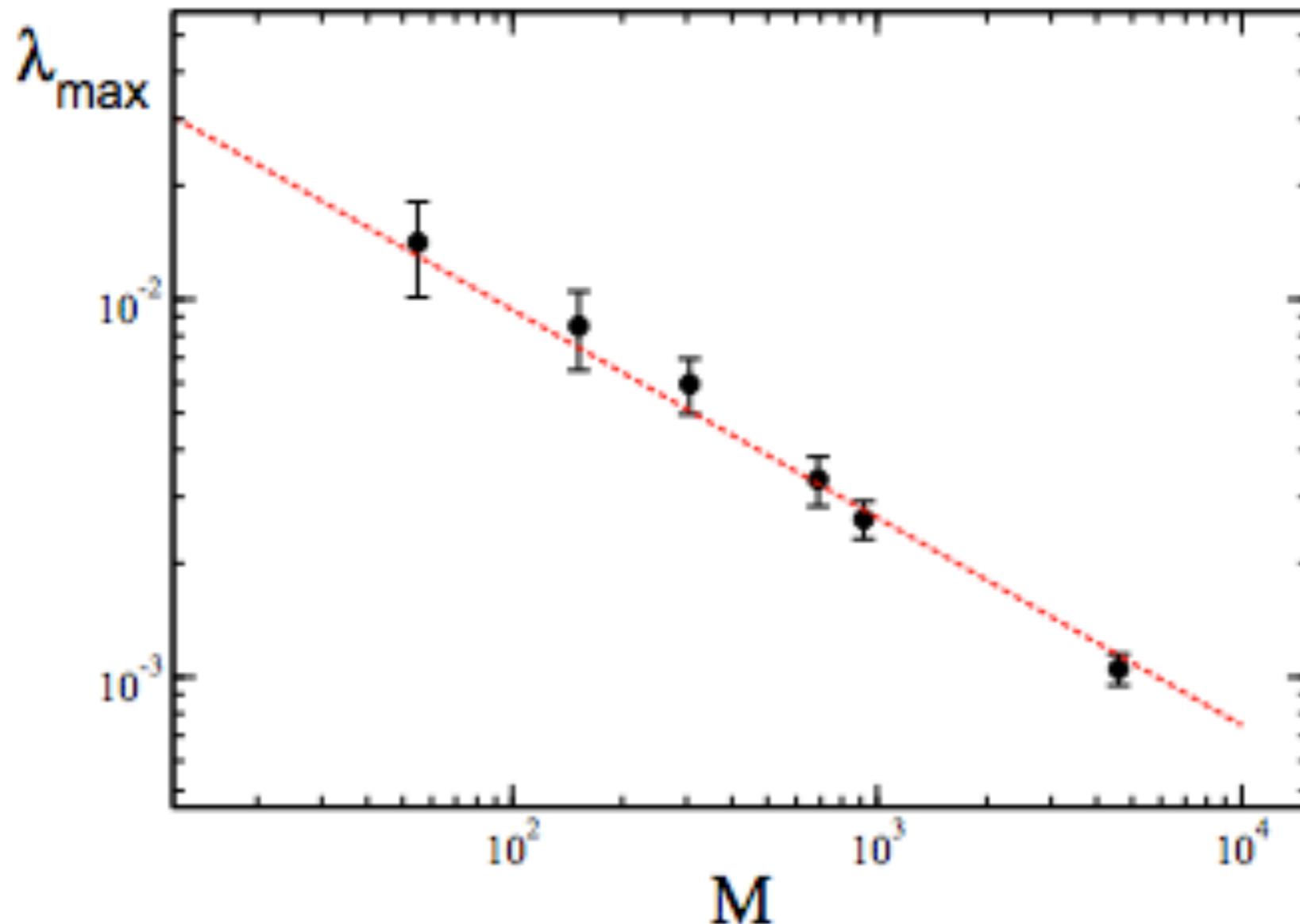
$$R = \left\langle \frac{1}{N} \left| \sum_{j=1}^N e^{i\phi_j(t)} \right| \right\rangle$$

$$\phi_i(t, m) = 2\pi \frac{t - t_i(m)}{t_i(m+1) - t_i(m)}$$

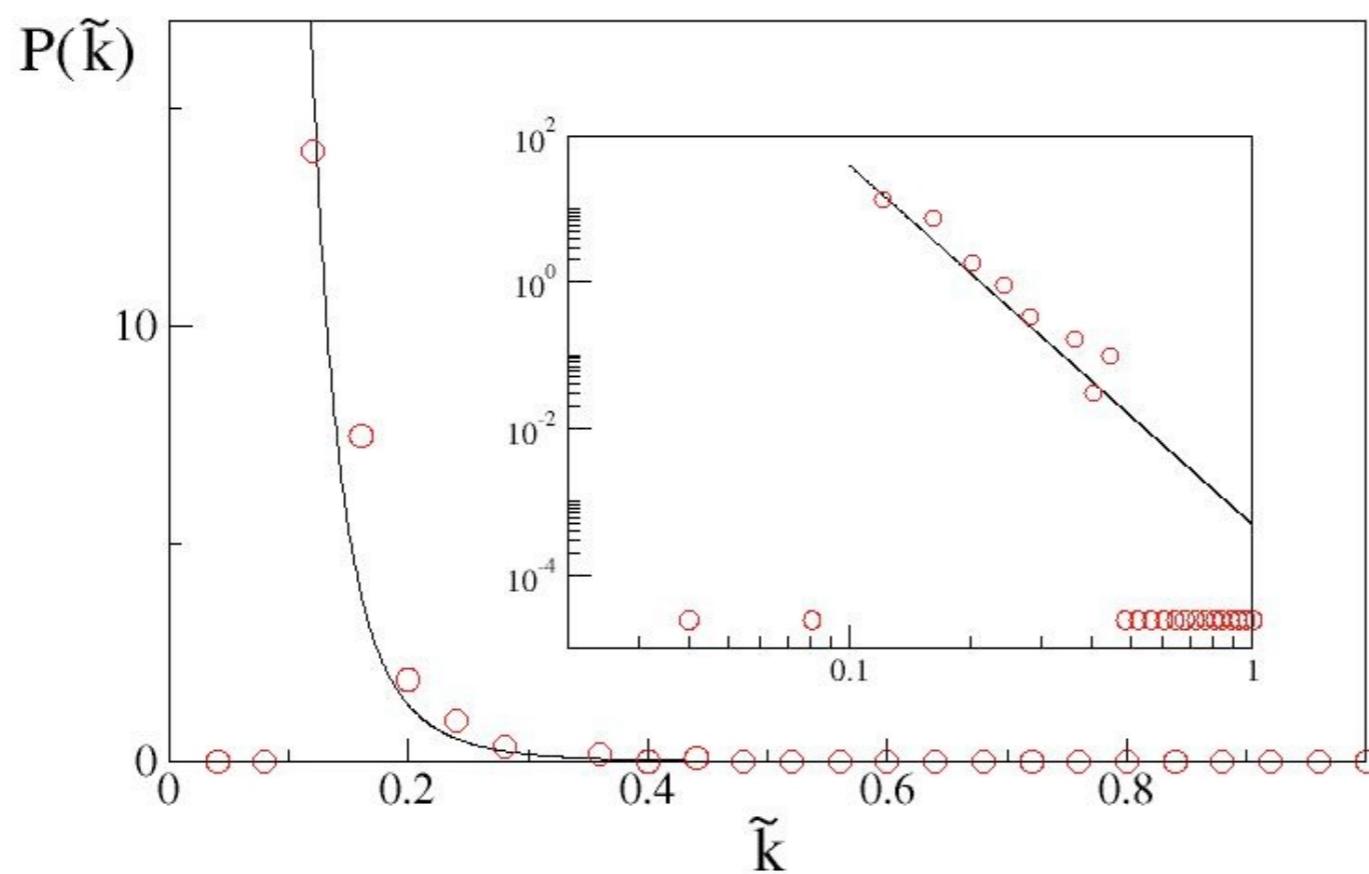
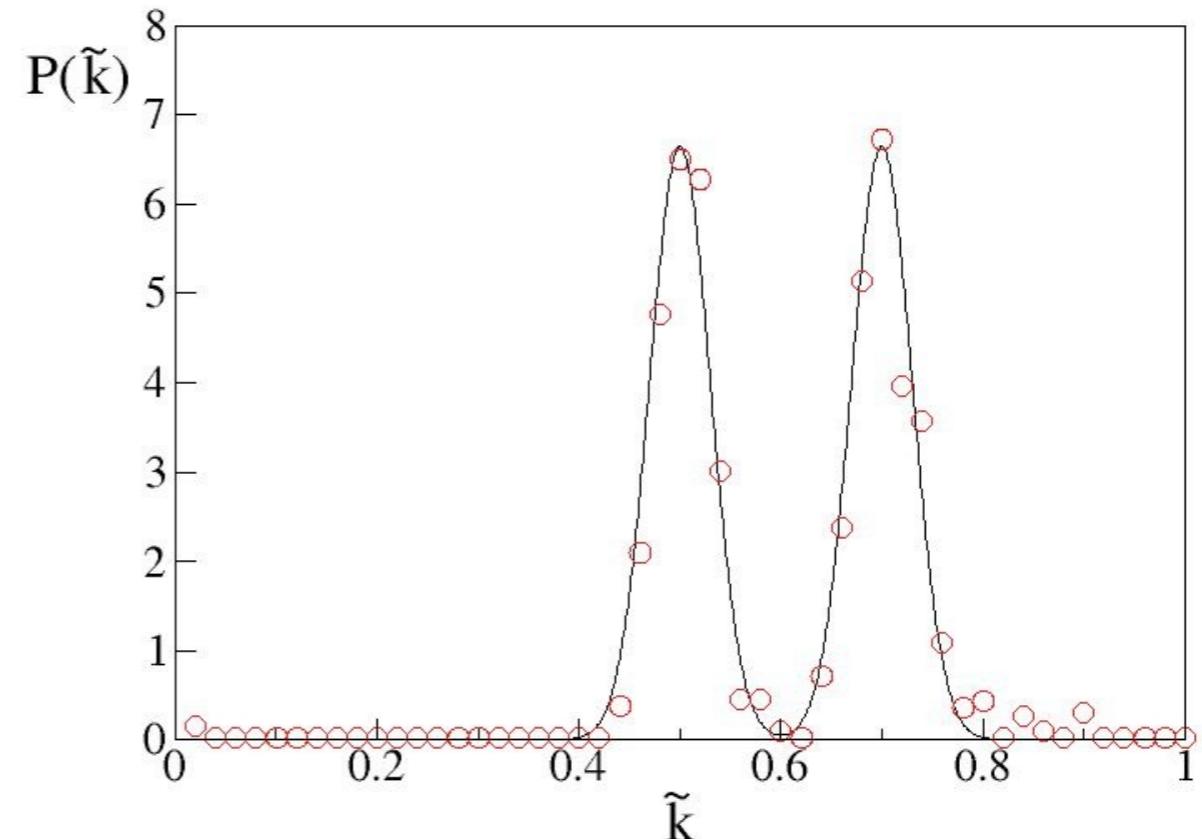
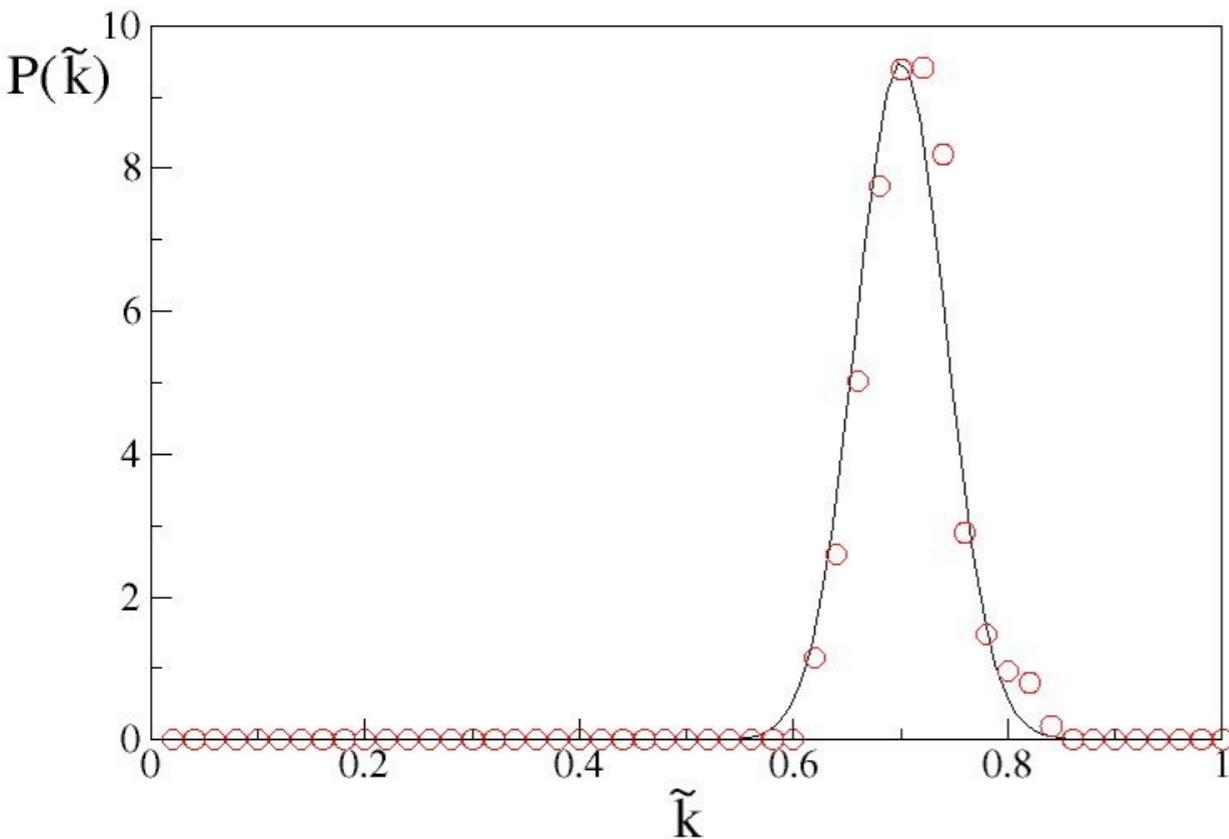


# HMF model is non-chaotic

$$\lambda_{max} \sim 1/\sqrt{M}$$



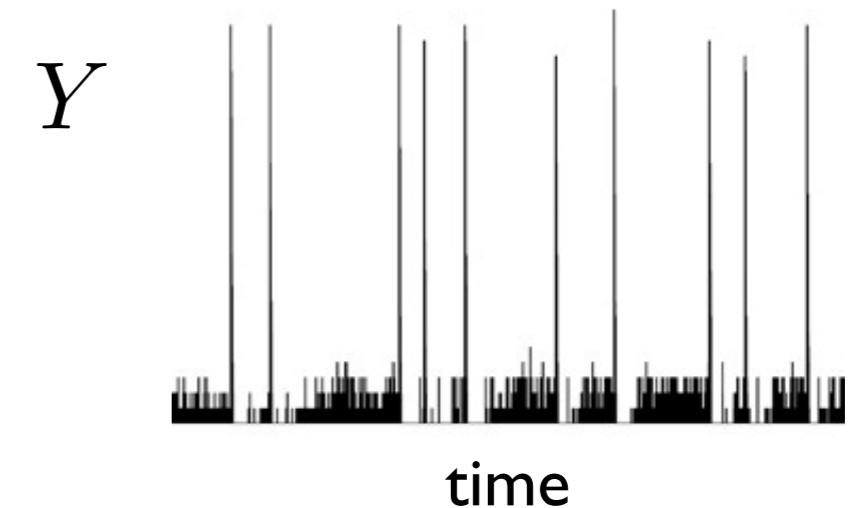
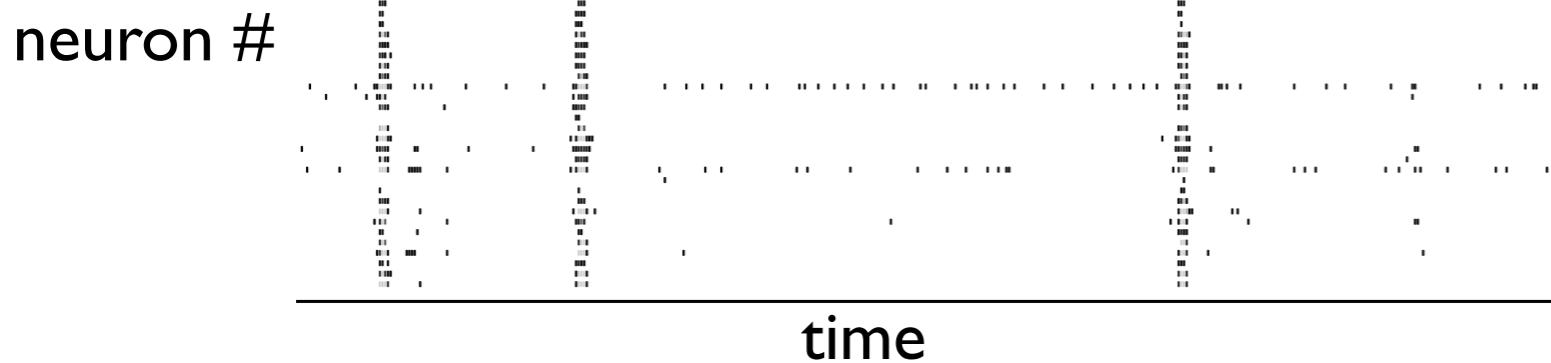
# Global Inverse Problem: $Y(t) \rightarrow P(\tilde{k})$



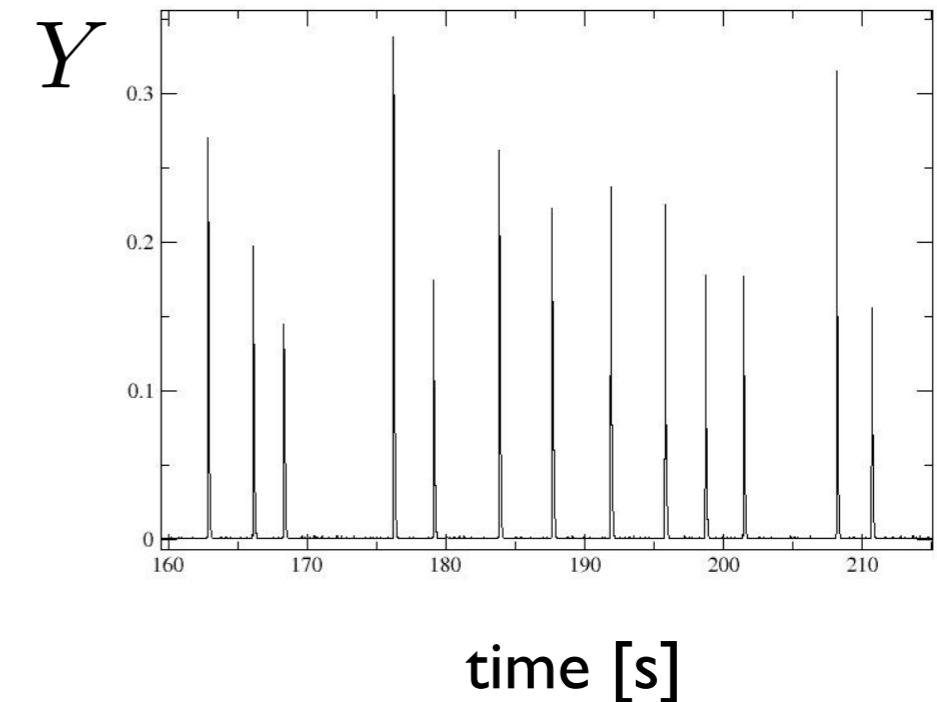
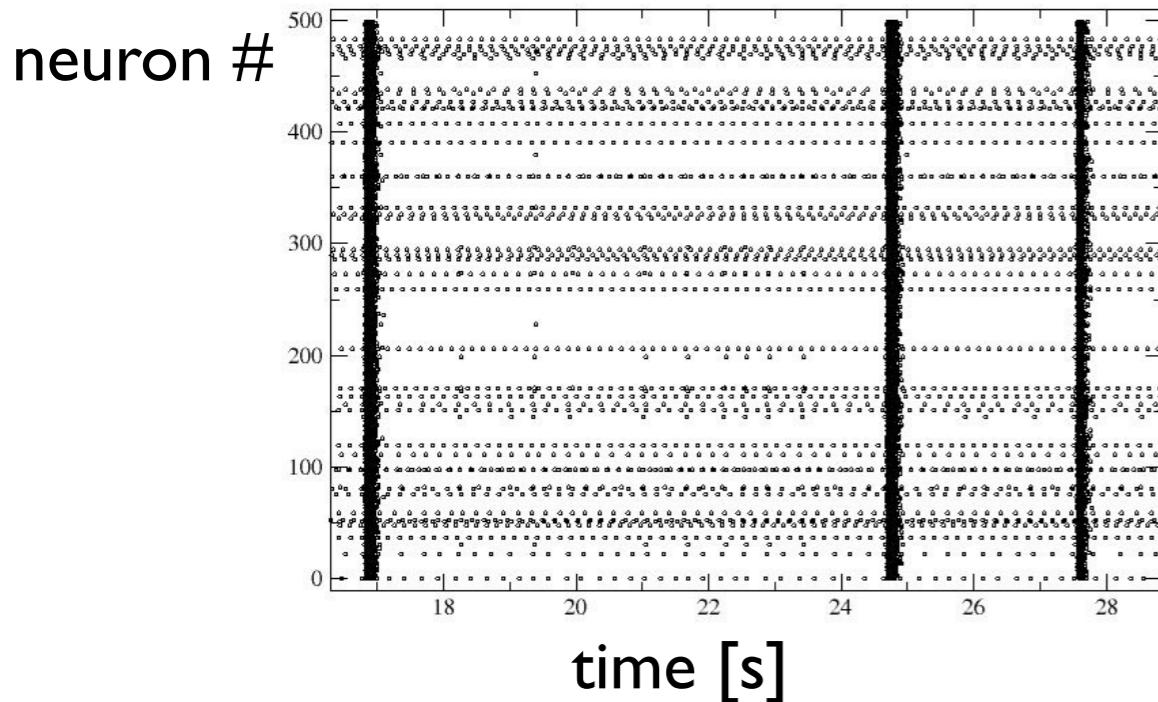
# Different setups: the disorder in neurons excitability

in vitro experiments

Robinette et al., Front. Neuroeng. (2011)



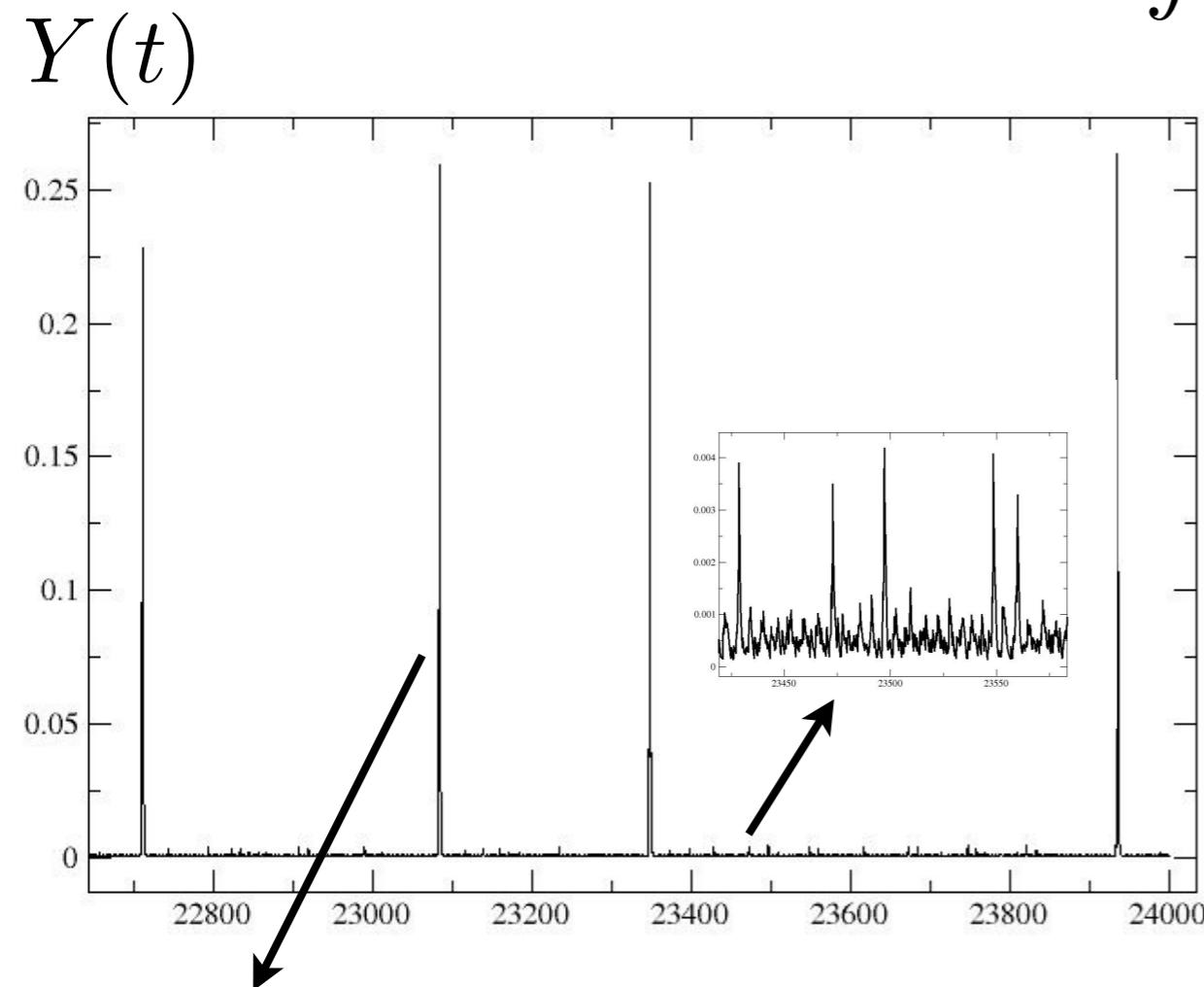
Model: disorder on  $a_i$  around  $a_c = 1$



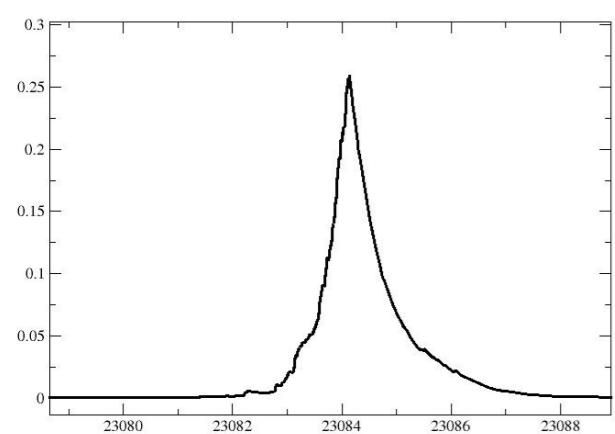
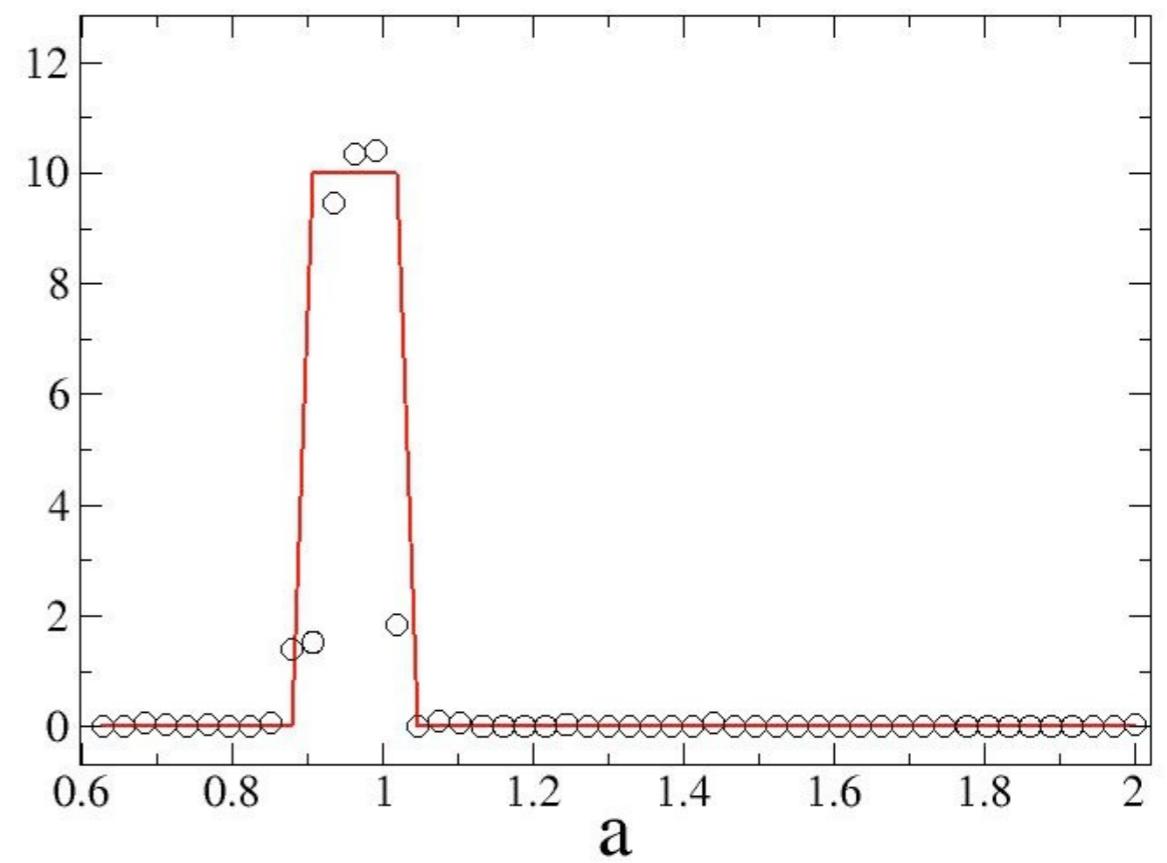
# Different setups: the inversion procedure

Uniform distribution  $P(a)$  around threshold, All-to-All network

$$Y(t) = \int P(a)y_a(t)da$$



$P(a)$



# Conclusions

- Heterogeneous Mean Field reproduces finite size dynamics
- Rich dynamical phase
- Connectivity distribution from global signals

## Collaborators:

*University of Parma:* R. Burioni, , M. Casartelli  
A. Vezzani

*University of Florence:* R. Livi