

# Symbolic stochastic resonance in excitable systems

**Cristina Masoller**

Departament de Física

Universitat Politècnica de Catalunya

Campus Terrassa

[www.fisica.edu.uy/~cris](http://www.fisica.edu.uy/~cris)

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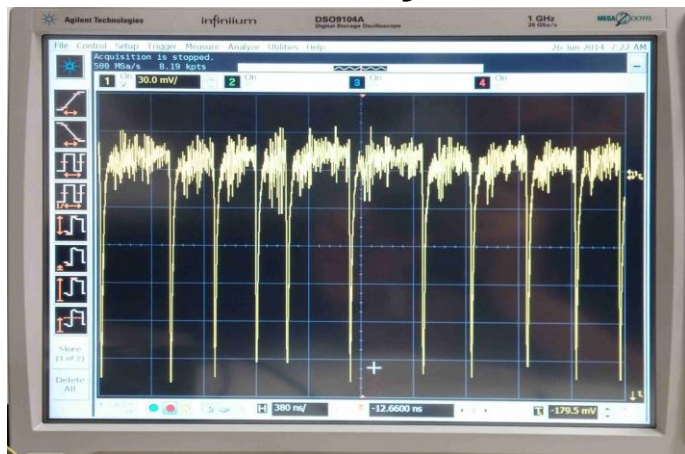
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XXVI Sitges Conference, May 2019

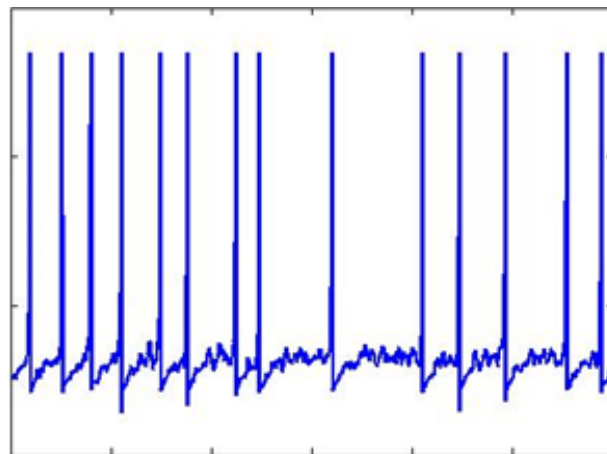
# Motivation: How excitable systems respond to **weak** external forcing in **noisy** environments?

## Laser intensity



Time  $10^{-6}$  s

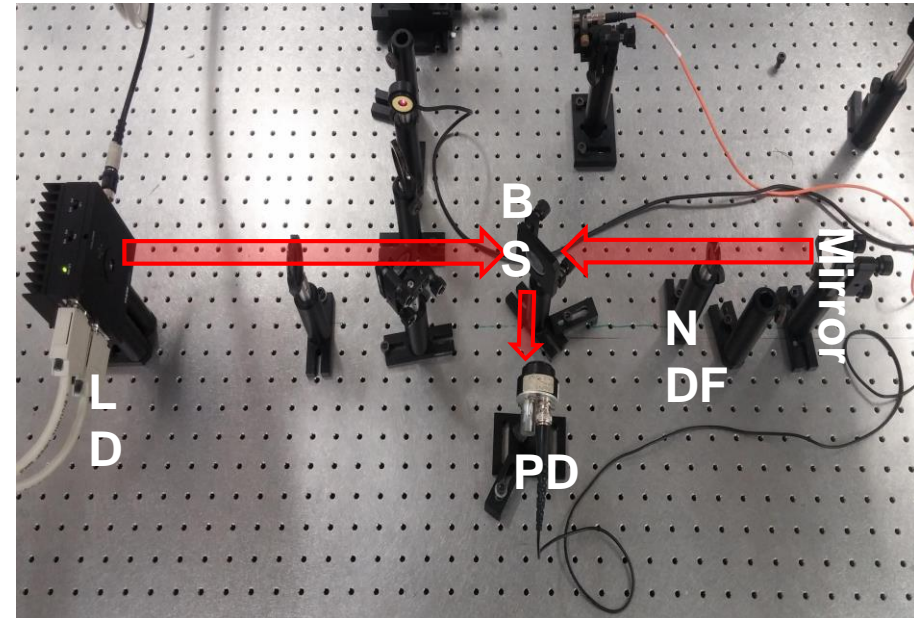
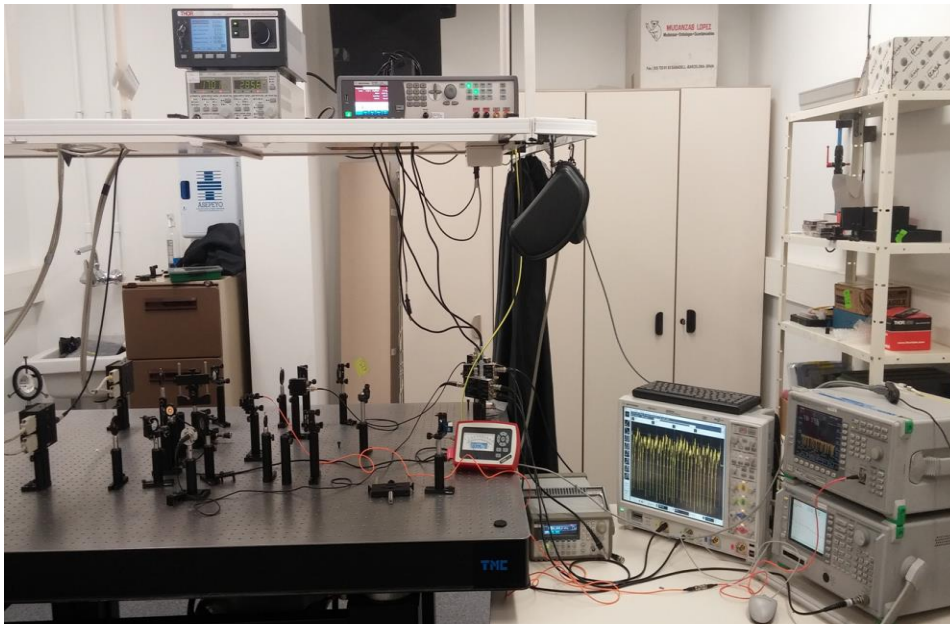
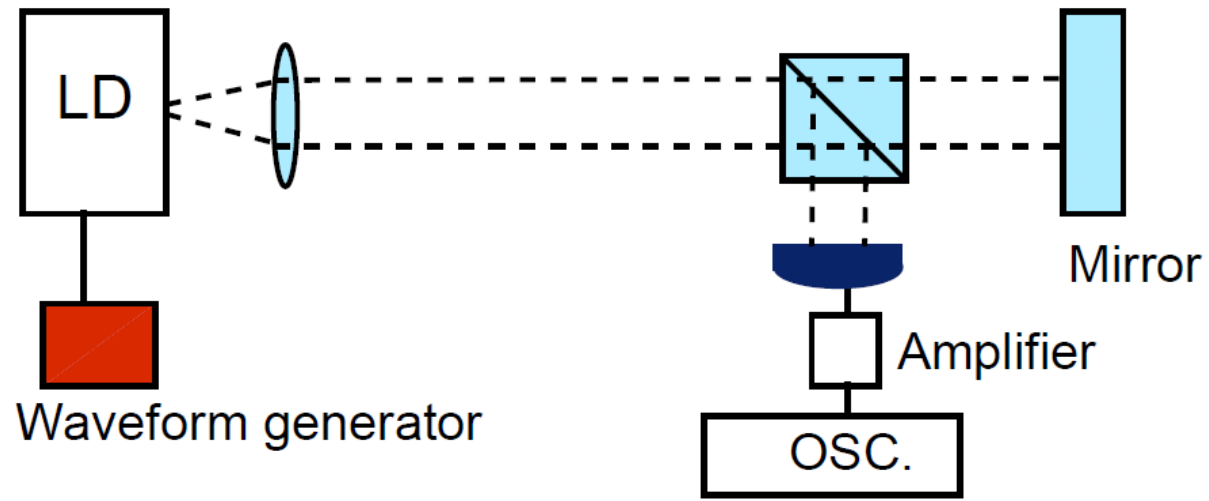
## Neuron model



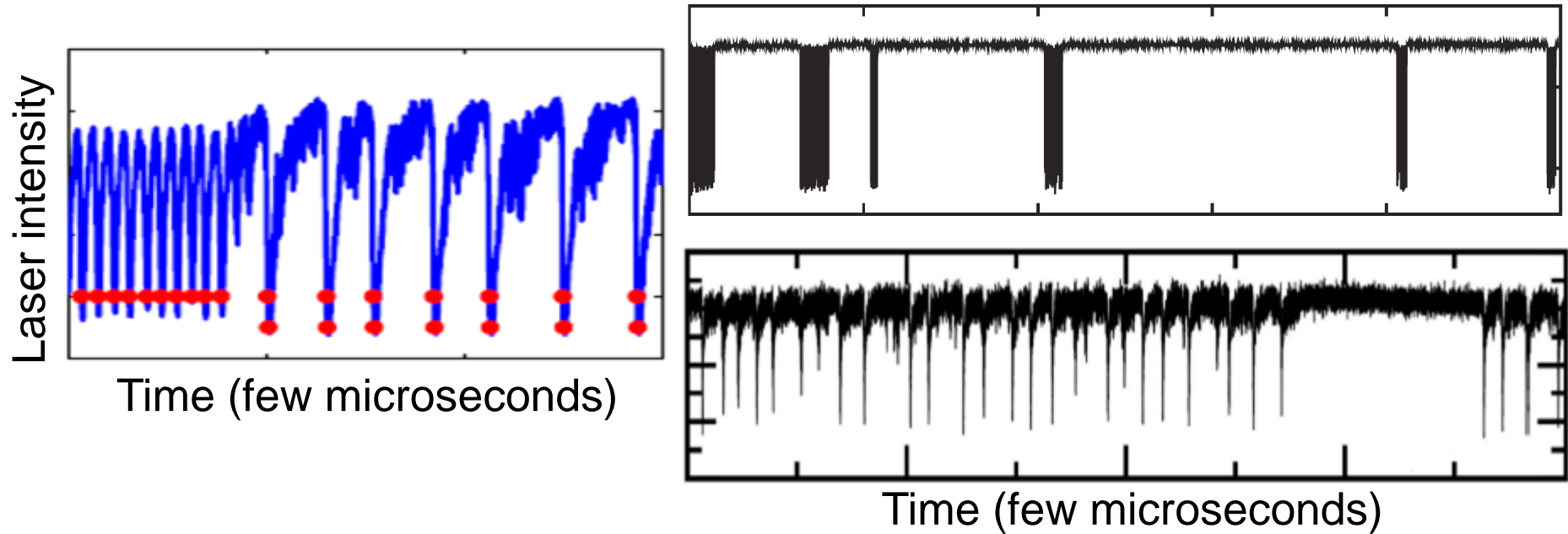
Time  $10^{-3}$  s

Can lasers mimic real neurons?

How can a neuron (or a laser) “encode”, in a sequence of spikes, the information of a weak signal, in the presence of noise?



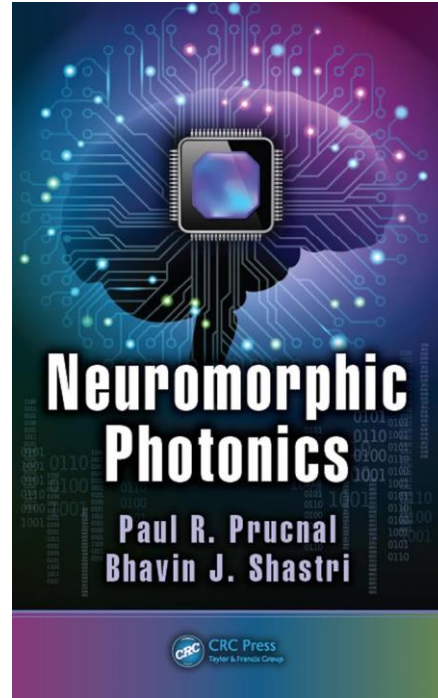
# The dynamics of a laser with feedback: excitability, tonic spikes and bursting. Similar to real neurons?



A. Aragoneses et al., *Sci. Rep.* **4**, 4696 (2014).

C. Quintero-Quiroz et al., *Sci. Rep.* **6** 37510 (2016).

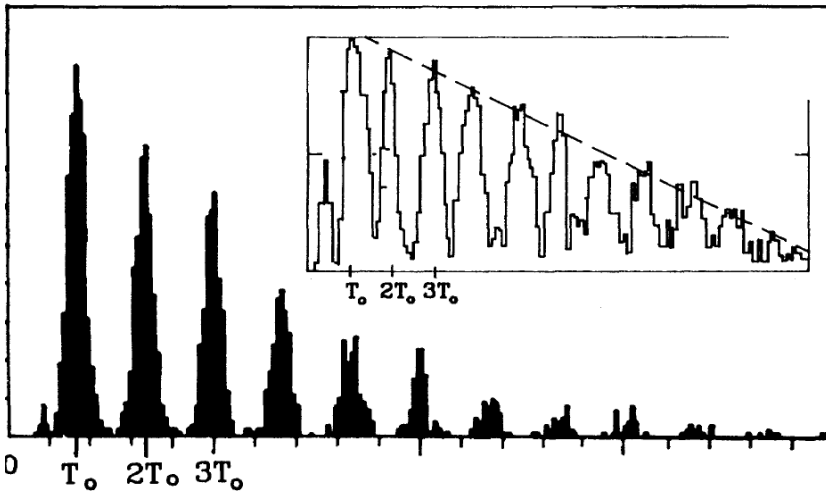
# Uncovering similarities can be interesting. But useful? Maybe...



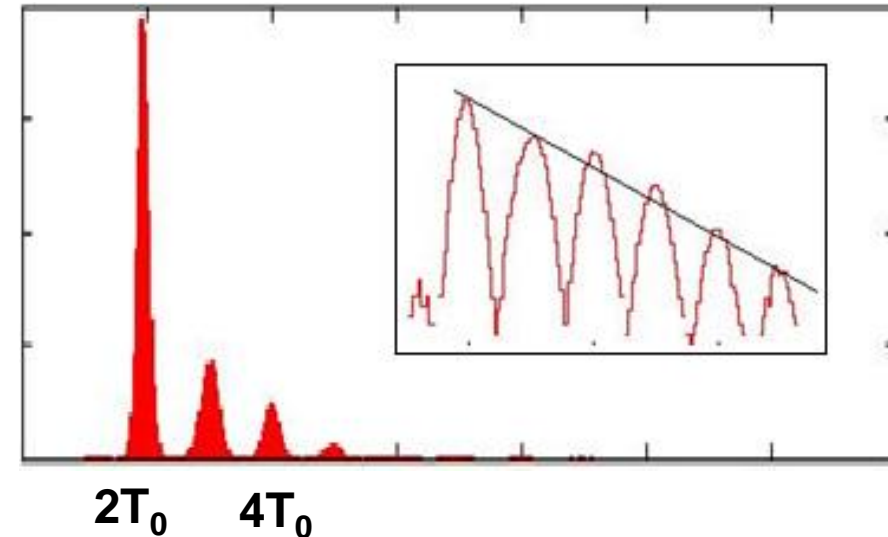
- Excitable lasers: building blocks of **photonic neurons**.
- Diode lasers: very very low cost, highly energy efficient.
- Very very fast.
- Main challenge: understand how lasers and neurons encode a **weak periodic** signal in the presence of **noise**.

# Inter-spike-interval (ISI) distribution with sinusoidal forcing

## Neuron empirical data



## Laser empirical data



Experimental data when the laser current is modulated with a sinusoidal signal of period  $T_0$ .

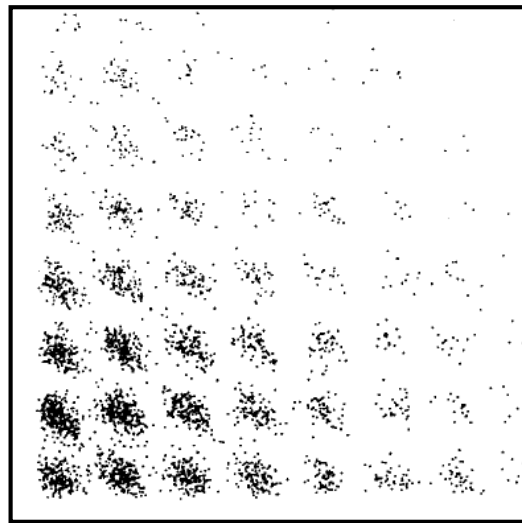
*Aragoneses et al. Opt. Express (2014)*

**Stochastic resonance:** noise-enhanced regularity.

*Gammaitoni, Hänggi et al, Rev. Mod. Phys. 70, 223 (1998)*

# Similar spike timing? Return maps of inter-spike-intervals

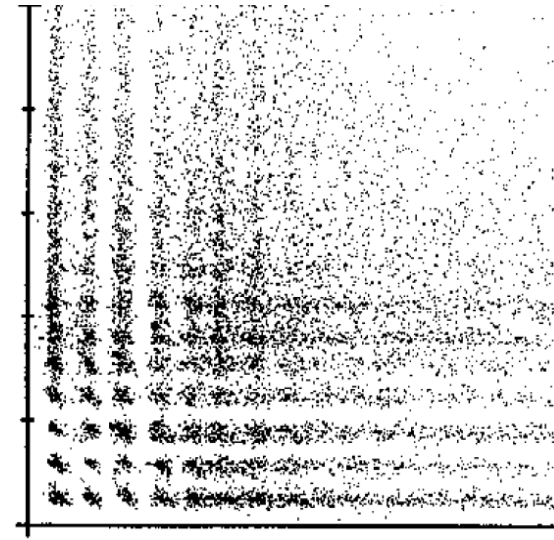
Neuronal ISIs



$\Delta T_i$

*A. Longtin, IJBC 1993*

Laser ISIs



*M. Giudici et al PRE 1997*

*A. Aragonese et al., Opt. Exp. 2014*

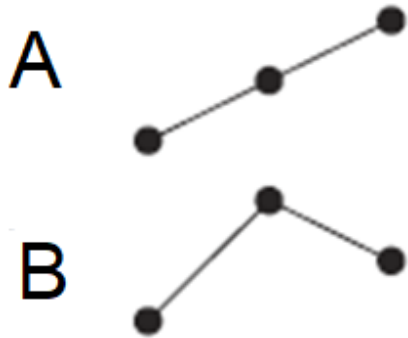
## HOW TO IDENTIFY SIMILAR TEMPORAL PATTERNS?

# Symbolic analysis of the sequence of time intervals: ordinal analysis

$$\{\dots x_i, x_{i+1}, x_{i+2}, \dots\}$$

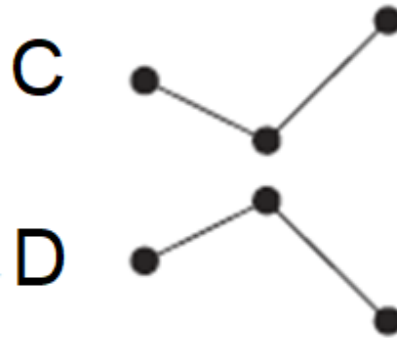
Possible order relations among three numbers (e.g., 2, 5, 7)

$\{\dots 2, 5, 7 \dots\}$



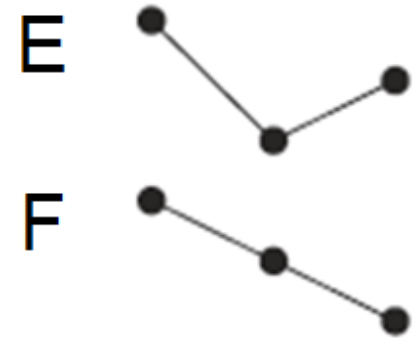
$\{\dots 2, 7, 5 \dots\}$

$\{\dots 5, 2, 7 \dots\}$



$\{\dots 5, 7, 2 \dots\}$

$\{\dots 7, 2, 5 \dots\}$

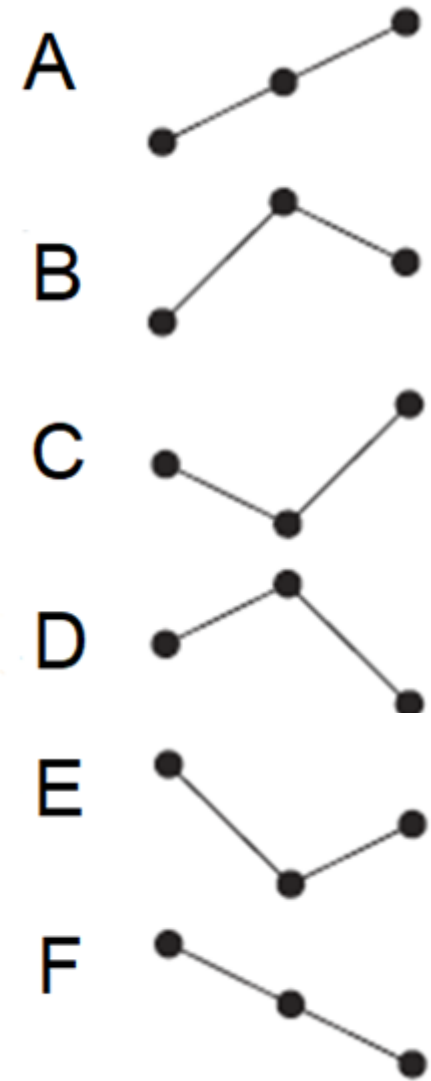
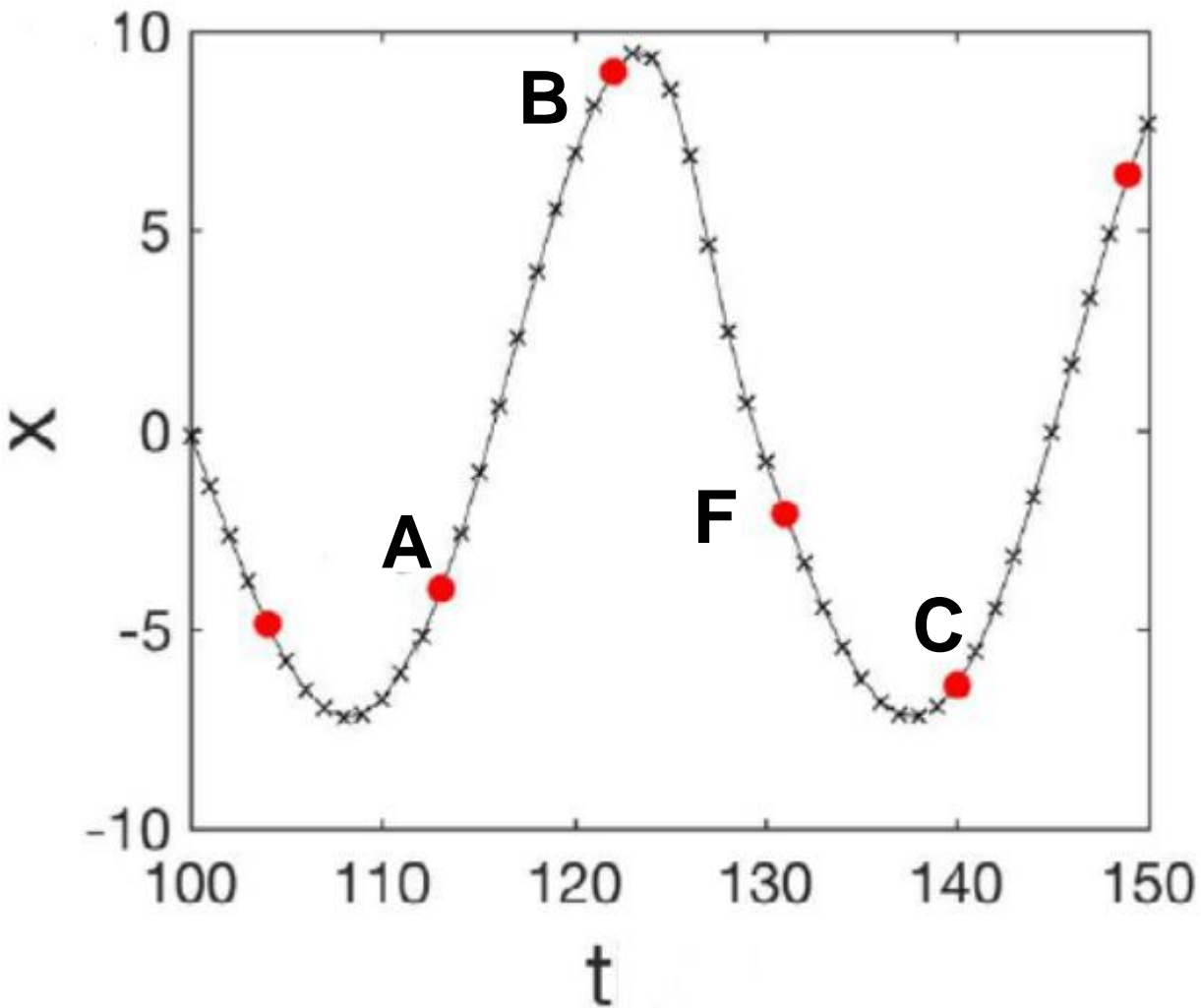


$\{\dots 7, 5, 2 \dots\}$

Drawback: information about the actual values is lost.

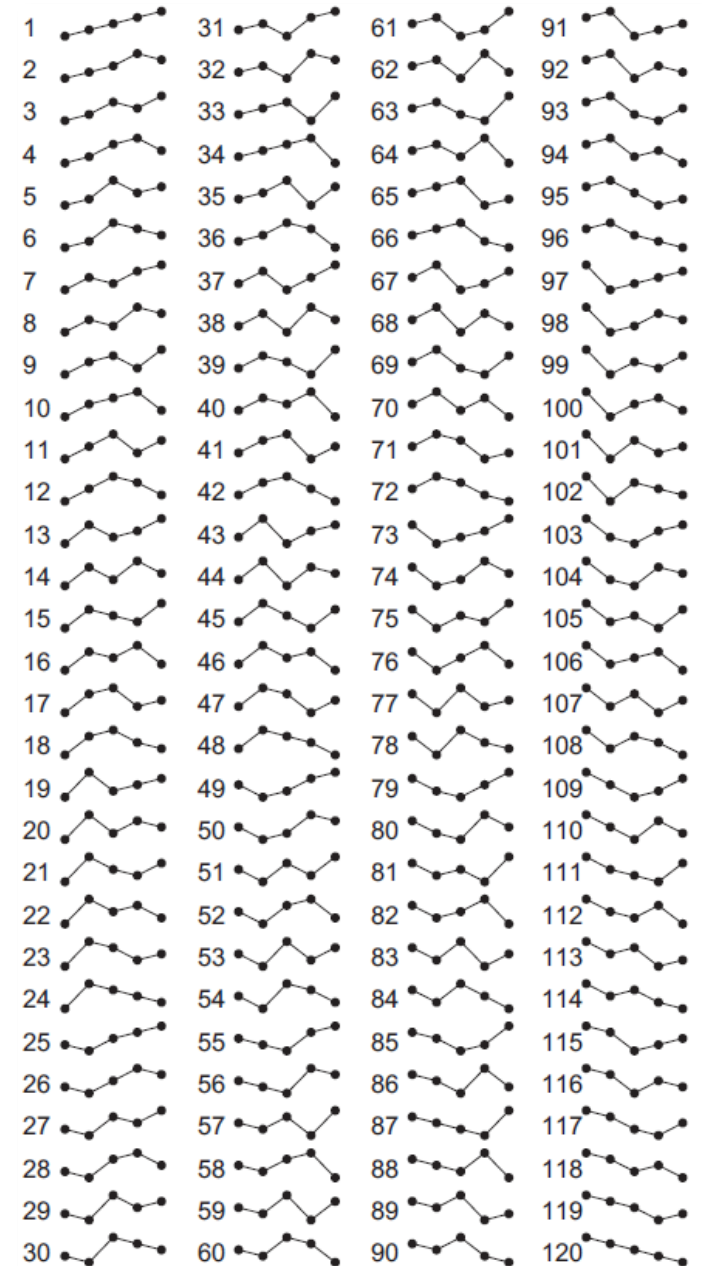
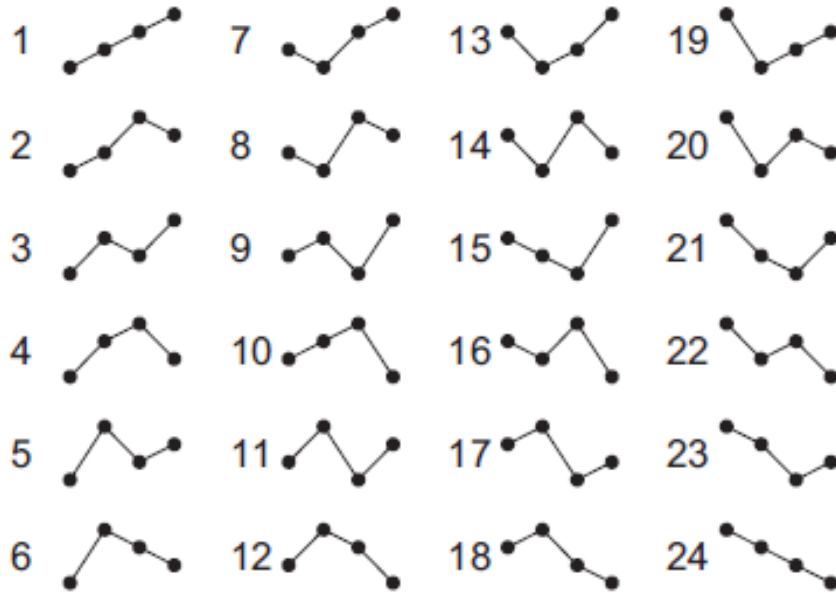


Which is the sequence of “letters” (patterns) defined by the red dots?



# Number of patterns increases as D! (D=size of pattern)

$\{\dots X_i, X_{i+1}, X_{i+2}, X_{i+3} \dots\}$



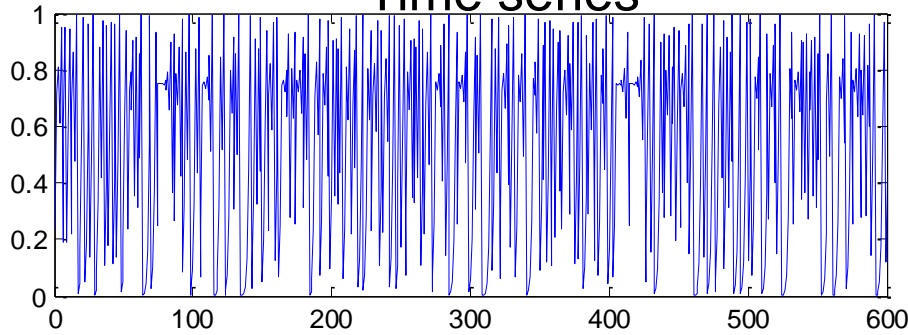
*U. Parlitz et al. / Computers in Biology and Medicine 42 (2012) 319-327*

**Permutation entropy:**  $S_p = -\sum p_i \log p_i$

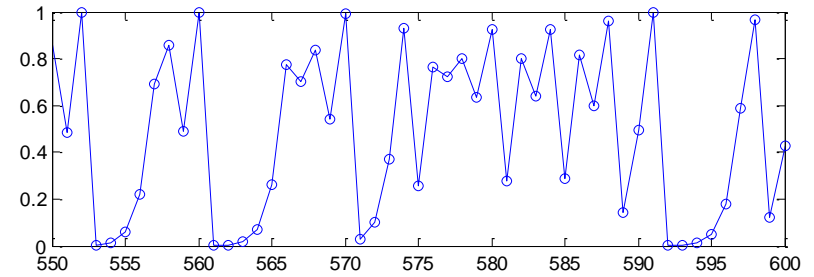
# Example: Logistic map, D=3

$$x(i+1) = r x(i)[1 - x(i)]$$

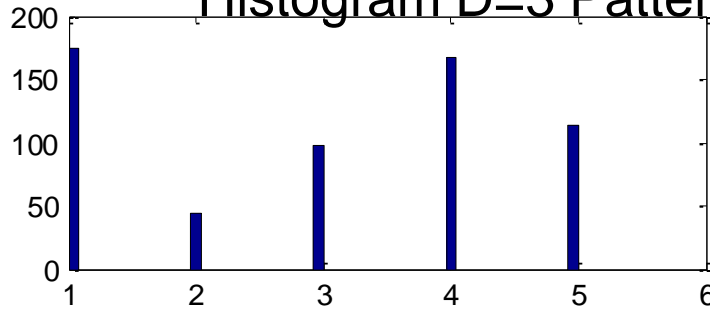
Time series



Detail



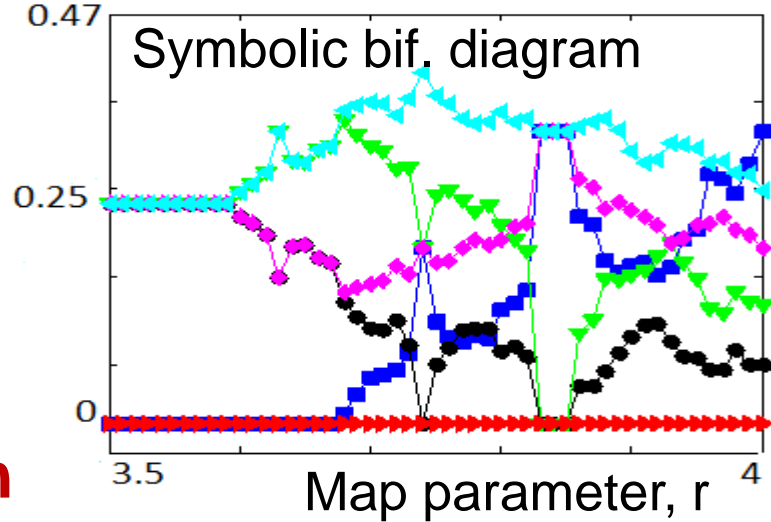
Histogram D=3 Patterns



**forbidden**



Probabilities



**012**

**021**

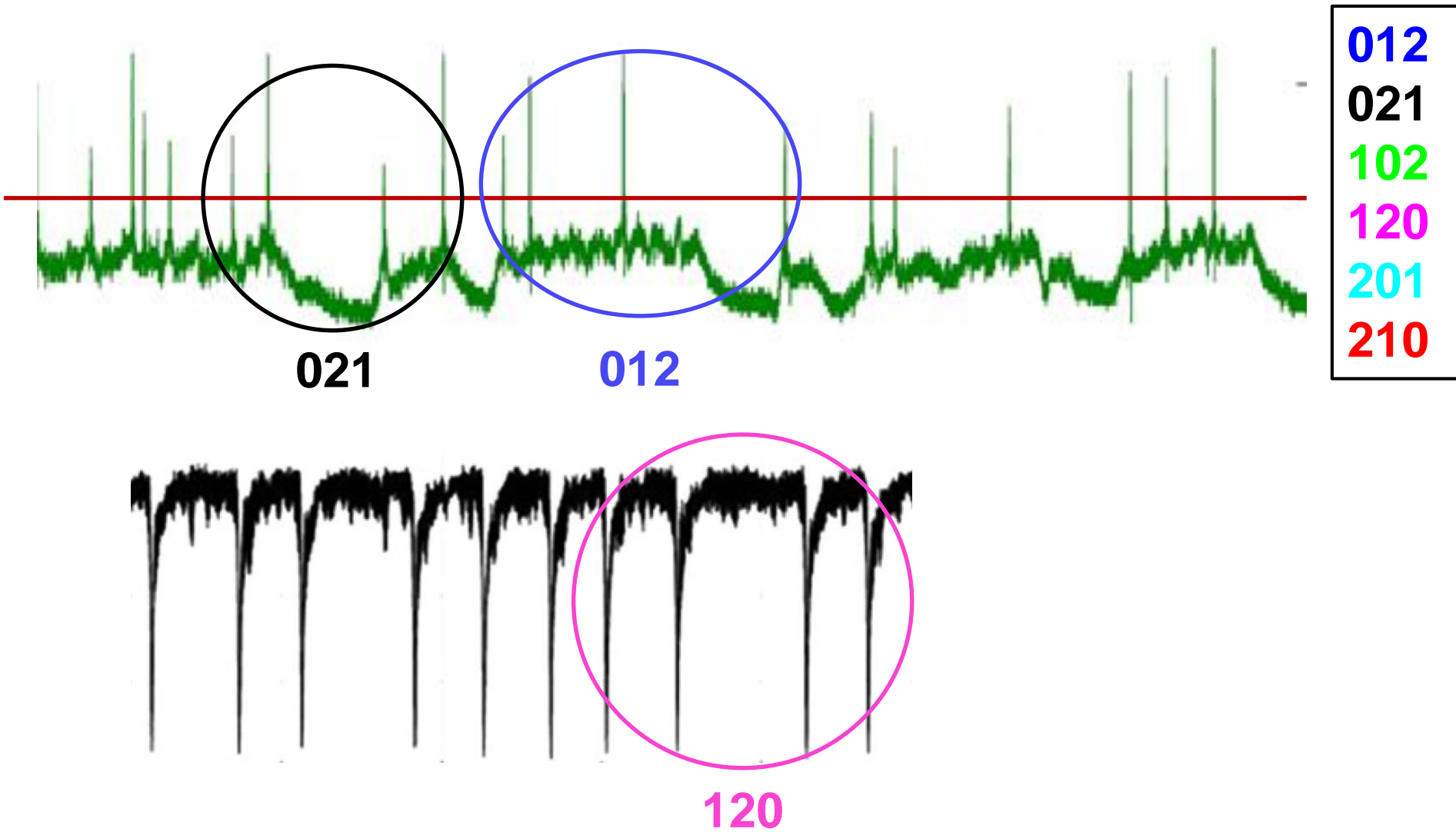
**102**

**120**

**201**

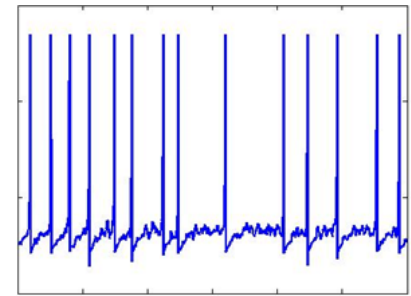
**210**

# Analysis of D=3 patterns in spike sequences



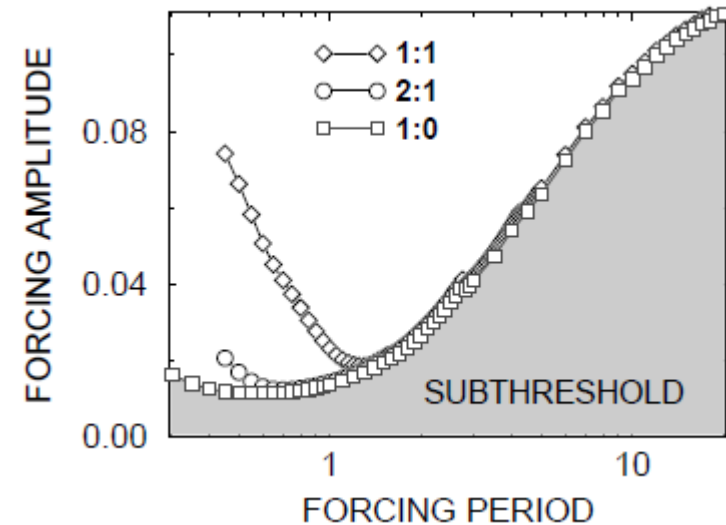
# Modeling neuronal spikes with FitzHugh-Nagumo model

$$\epsilon \frac{dx}{dt} = x - \frac{x^3}{3} - y,$$
$$\frac{dy}{dt} = x + a + a_0 \cos(2\pi t/T) + D\xi(t),$$



Time

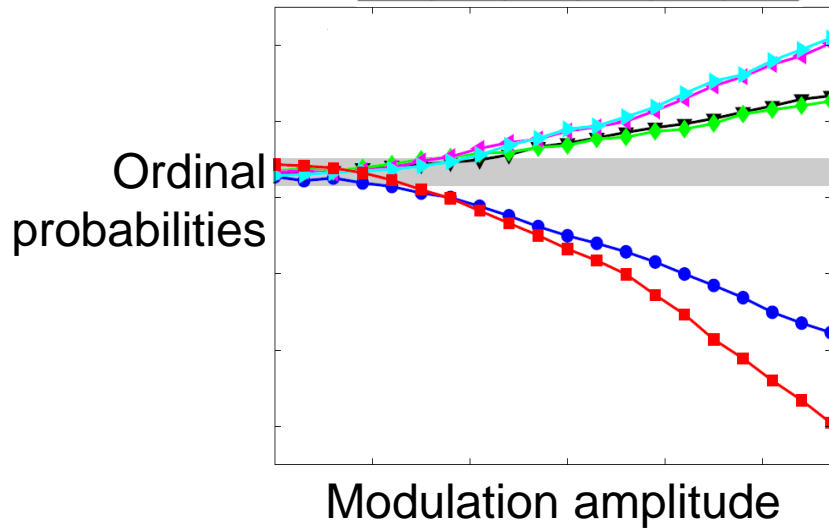
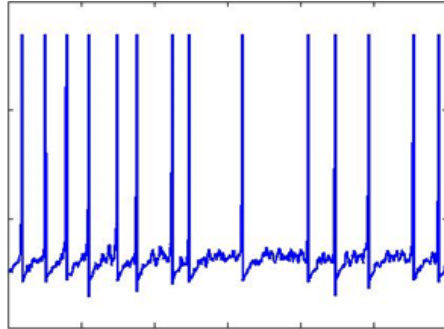
- Gaussian white noise and subthreshold signal:  $a_0$  and  $T$  such that spikes are **noise-induced**.
- Time series with  $M=100,000$  spikes simulated ( $a=1.05$ ,  $\epsilon=0.01$ ).



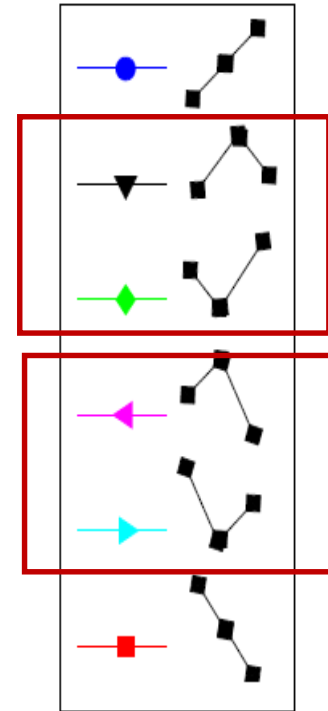
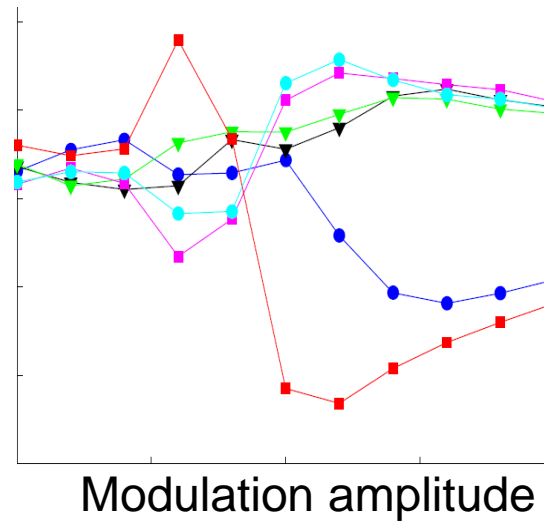
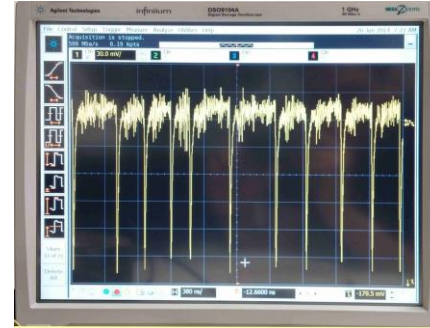
*Longtin and Chialvo, PRL 1998*

# By analyzing the ordinal probabilities (from inter-spike intervals, ISIs) we uncover laser-neuron similarities

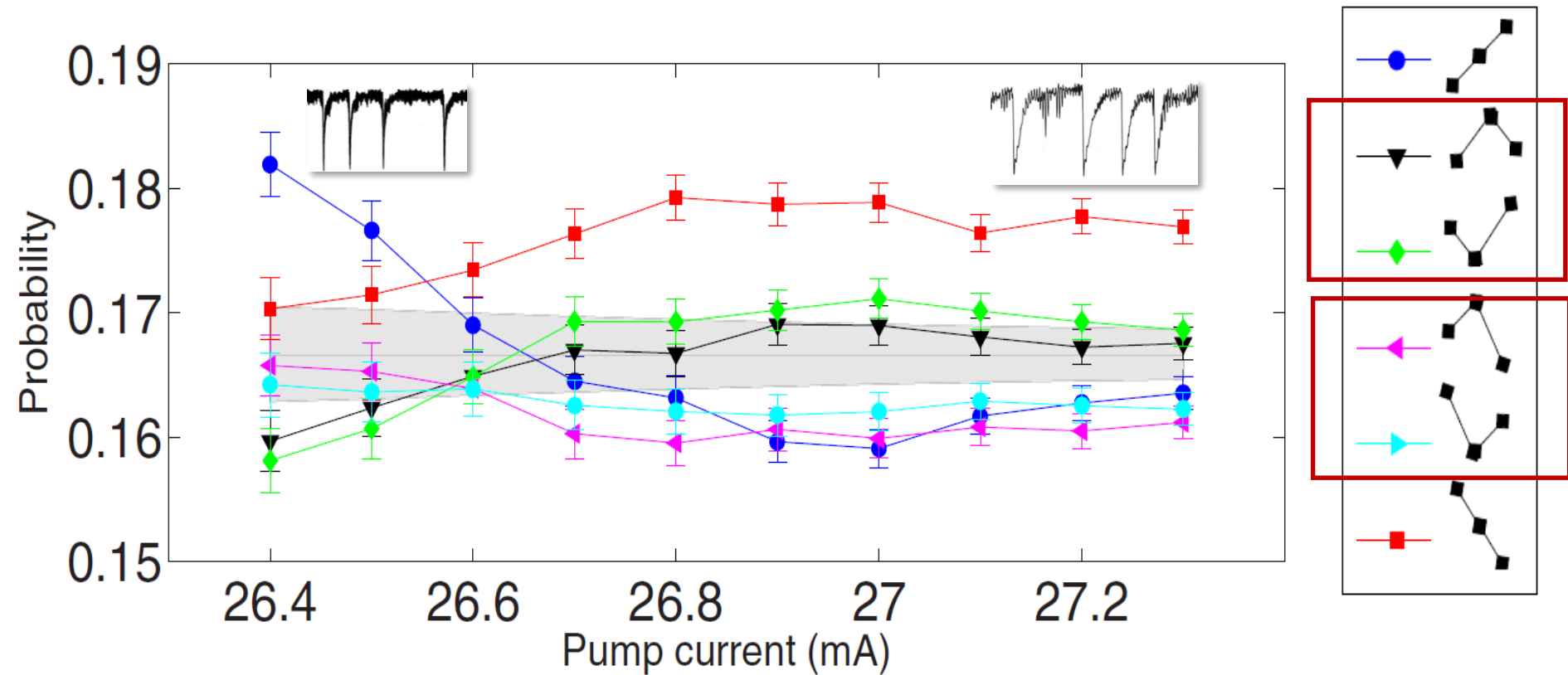
## Neuron model



## Laser with feedback



# Role of experimental control parameter: the laser current



# Minimal model? A modified circle map

$$\varphi_{i+1} = \varphi_i + \rho + \frac{K}{2\pi} \left[ \sin(2\pi\varphi_i) + \alpha_c \sin(4\pi\varphi_i) \right] + D\zeta$$

$$X_i = \varphi_{i+1} - \varphi_i$$

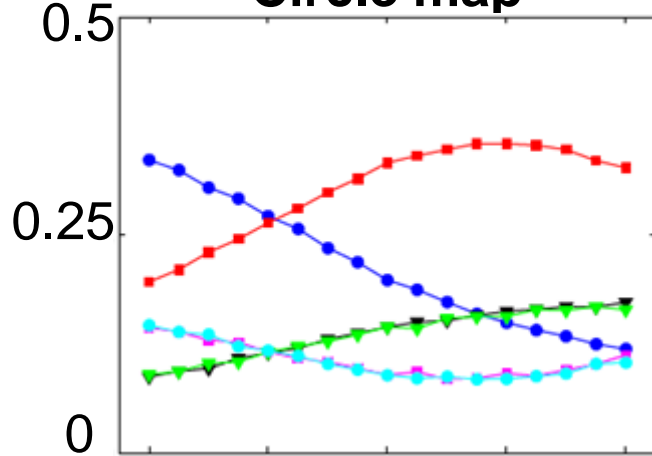
$\rho$  =  $\frac{\text{natural frequency}}{\text{forcing frequency}}$

K = forcing amplitude

D = noise strength

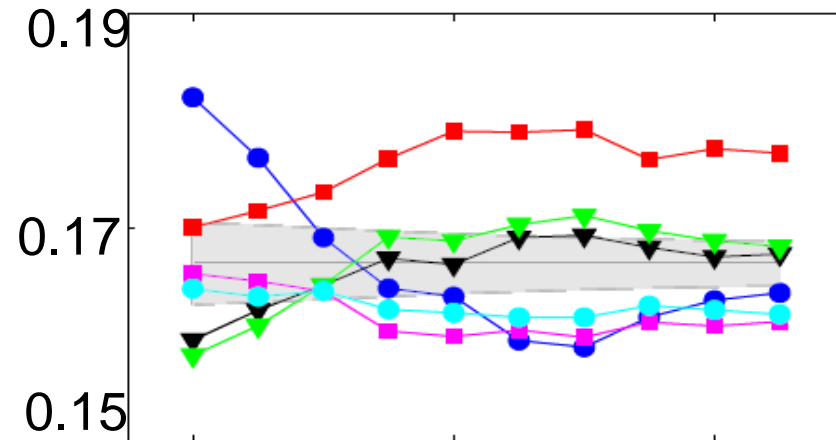
Prob

Circle map



Map parameter  $\alpha_c$

Laser spikes



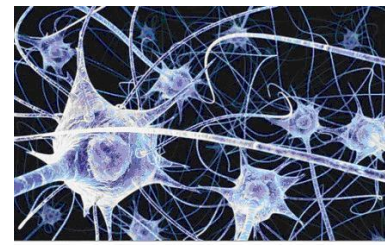
Laser current

Same “clusters” & same hierarchical structure

*A. Aragonese et al, Sci. Rep. 4, 4696 (2014)*

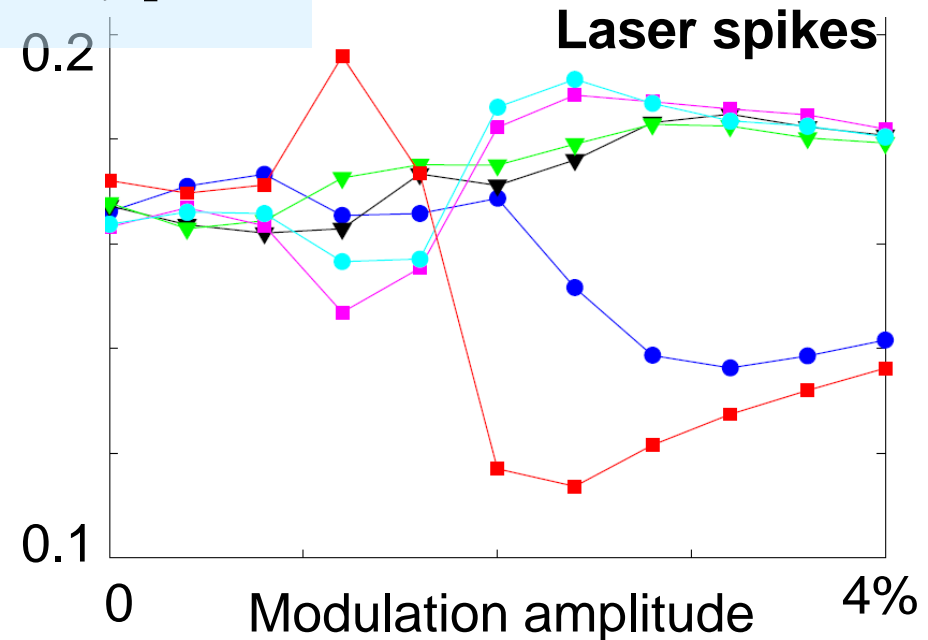
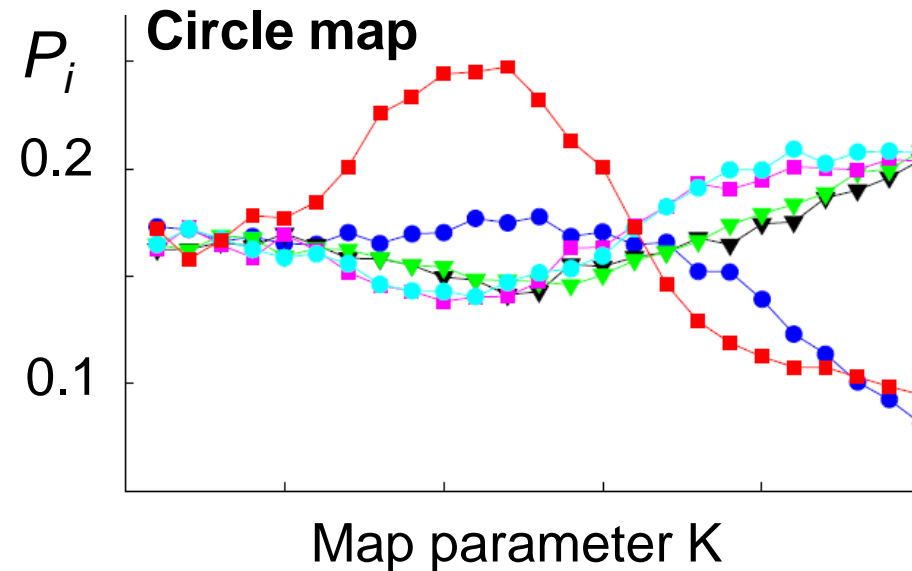


# Connection with neurons: the circle map describes many excitable systems



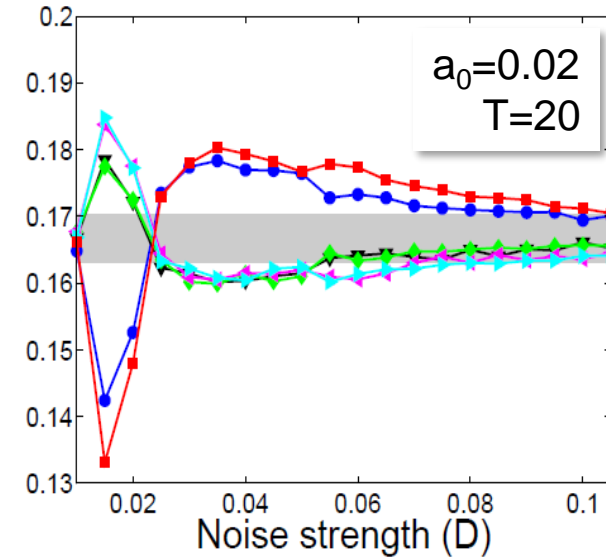
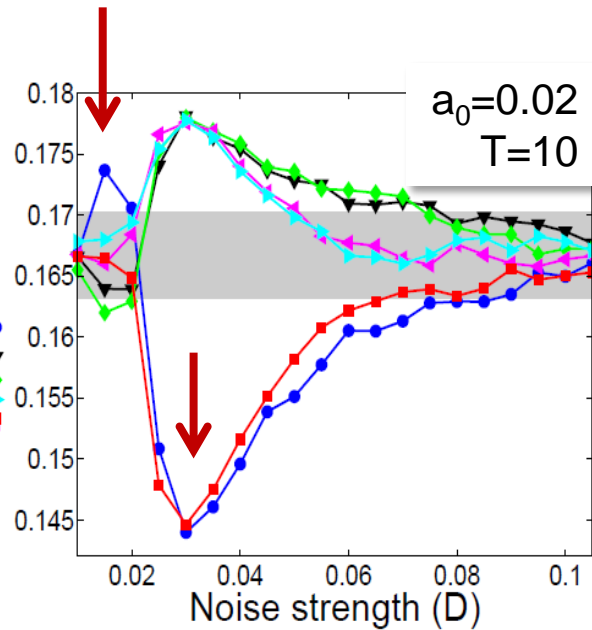
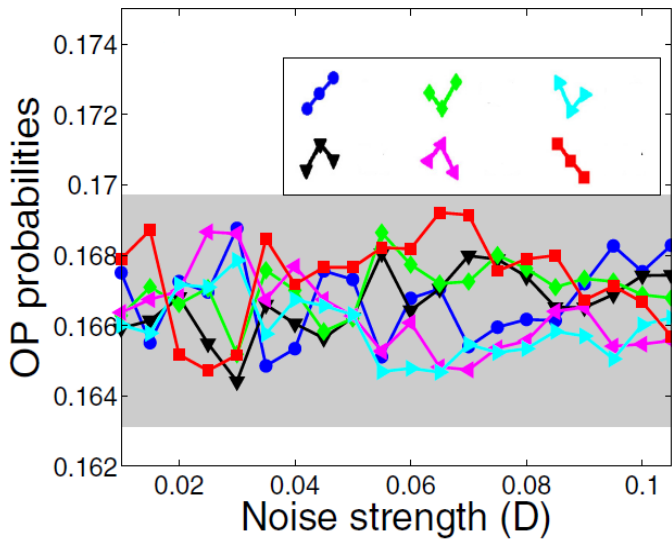
- Spike correlations in sensory neurons (Neiman and Russell, PRE 2005)
- Can we test its validity as a minimal model for the laser spikes?

$$\varphi_{i+1} = \varphi_i + \rho + \frac{K}{2\pi} [\sin(2\pi\varphi_i) + \alpha_c \sin(4\pi\varphi_i)] + D\zeta$$



# FHN neuron model: symbolic stochastic resonance

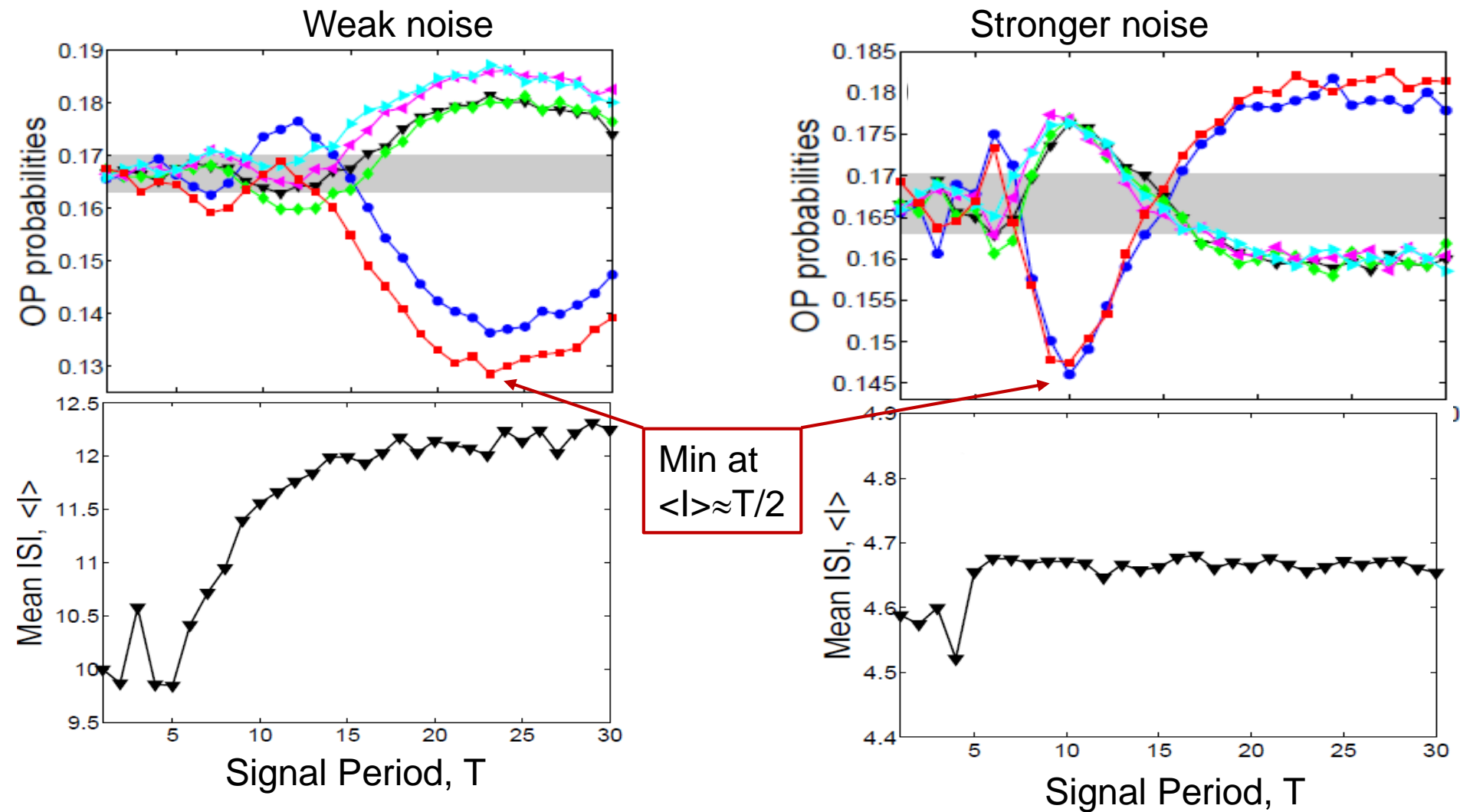
No external forcing ( $a_0=0$ )



- The signal induces preferred and infrequent patterns.
- They depend on the period and on the noise strength.
- Resonant-like behavior.

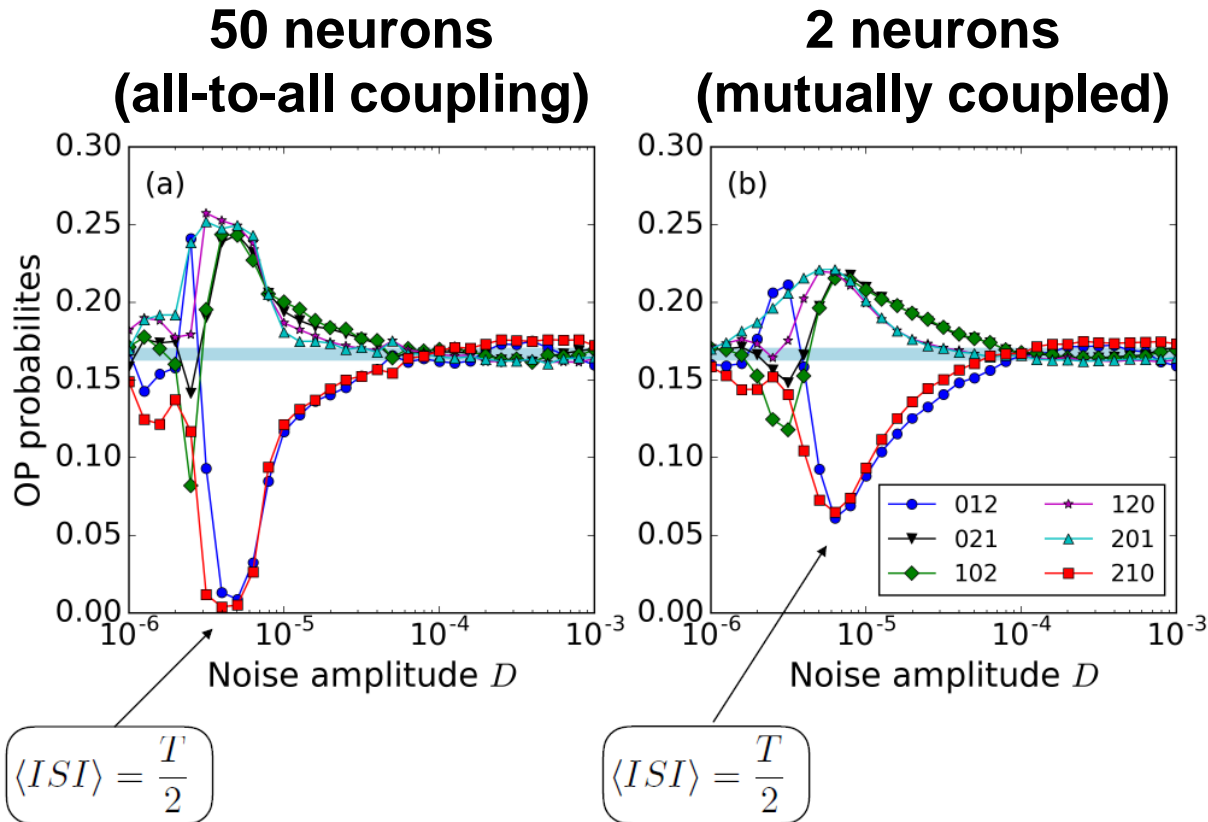
*J. M. Aparicio-Reinoso, M. C. Torrent and C. Masoller, PRE 94, 032218 (2016)*

# Role of the signal period



*J. M. Aparicio-Reinoso, M. C. Torrent and C. Masoller, PRE 94, 032218 (2016)*

# The resonance is enhanced by neuronal coupling

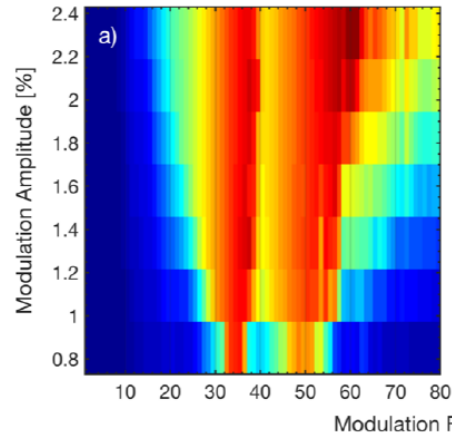


*M. Masoliver and C. Masoller, "Neuronal coupling benefits the encoding of weak periodic signals in symbolic spike patterns", arXiv:1905.01933 (2019)*

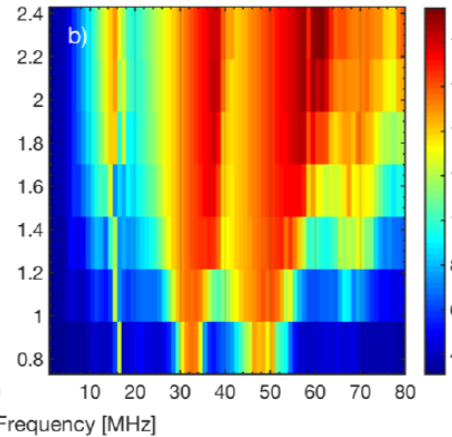
# Comparison: how a diode laser and a neuron encode a weak periodic signal? spike rate code?

Laser with optical feedback (experiments modulating the laser current)

Sinusoidal

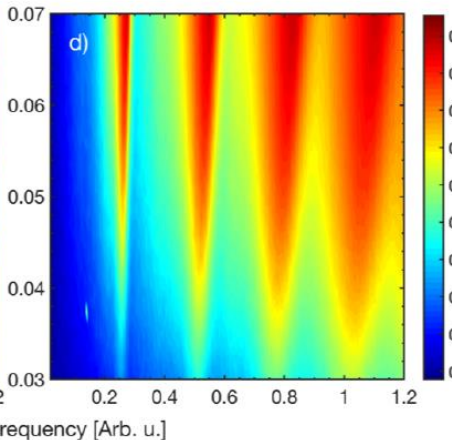
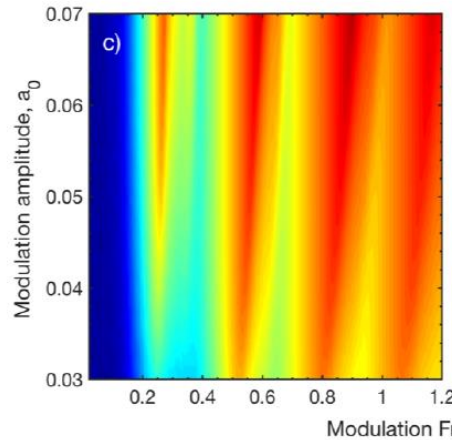


Pulsed signal



**Spike rate in color code**

Neuron model (with the same input signal)

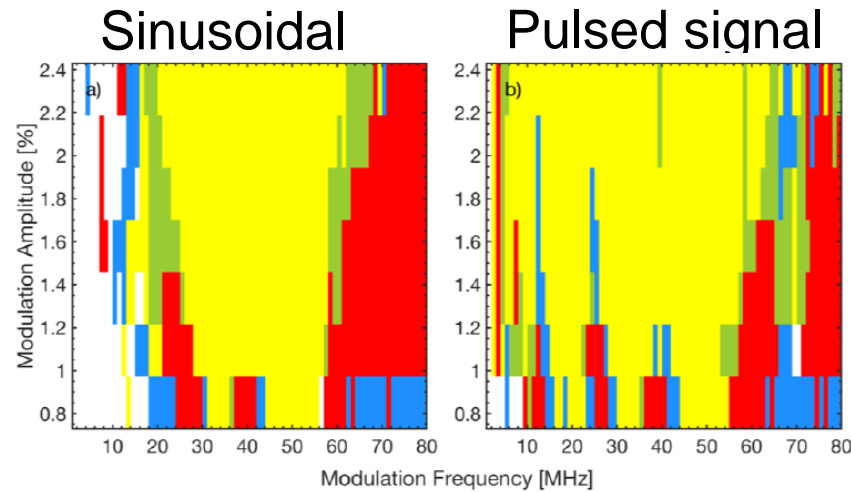


*J. Tiana-Alsina, C. Quintero-Quiroz and C. Masoller, "Comparing the dynamics of periodically forced lasers and neurons", submitted (2019)*

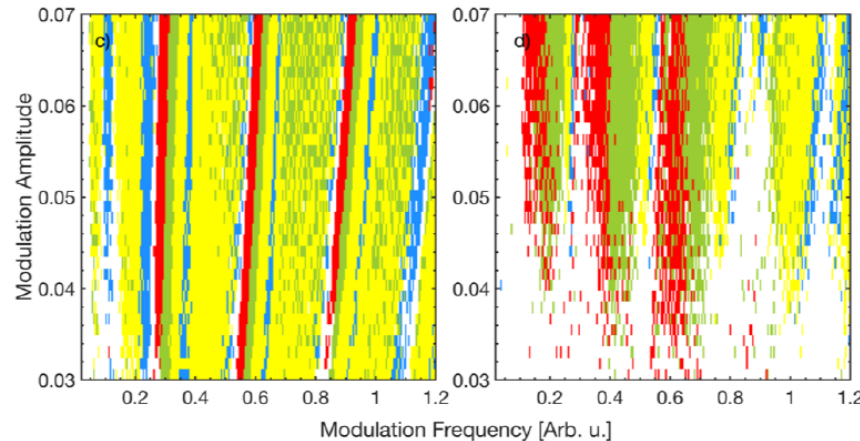
# Temporal code?

## Ordinal analysis unveils differences in spike timing

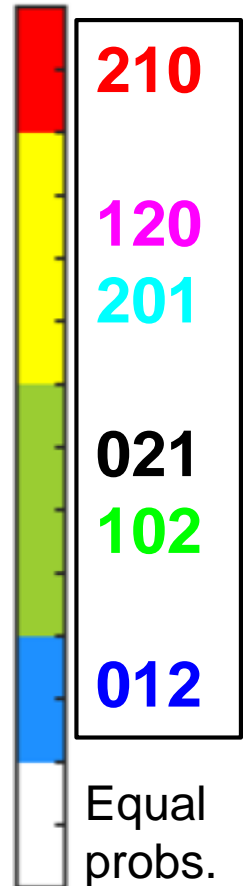
Laser with  
optical  
feedback



Neuron  
model



Most  
probable  
pattern in  
color  
code



*J. Tiana-Alsina, C. Quintero-Quiroz and C. Masoller, "Comparing the dynamics of periodically forced lasers and neurons", submitted (2019)*

# Take home message

- We have studied how neurons and lasers encode a weak periodic signal in their output spike sequences.
- Using symbolic analysis applied to the inter-spike-intervals we uncovered preferred and infrequent spike patterns.
- They depend on the amplitude and the period of the signal, and the noise strength.
- Symbolic resonance: the probabilities of the “trend” patterns take minimum values when  $\langle \text{ISI} \rangle \approx T/2$ .
- We found similarities between neurons and lasers, but also some differences.
- We found a minimal model (a modified circle map) that describes the variation of the ordinal probabilities with experimental parameters.



M. Masoliver

C. Quintero

J. Tiana

A. Aragoneses et al,  
Sci. Rep. 4, 4696 (2014).

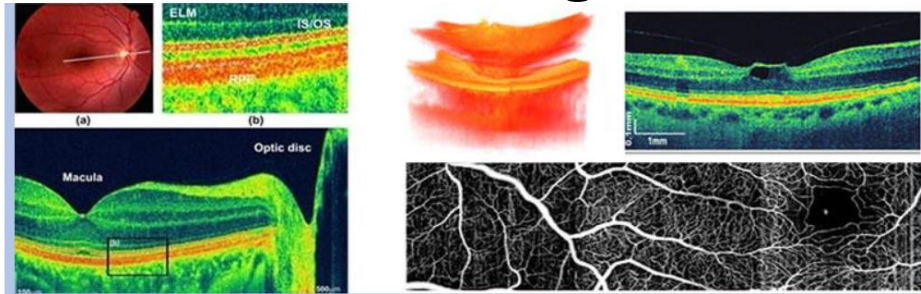
J. A. Reinoso et al,  
PRE 94, 032218 (2016).

J. Tiana et al,  
PRE 99, 022207 (2019).

M. Masoliver & C. Masoller,  
arXiv:1905.01933 (2019).

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- J. Tiana (UPC, Barcelona)
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