Nonlinear tools to extract information from complex signals

examples from climate, brain, neurons & lasers)

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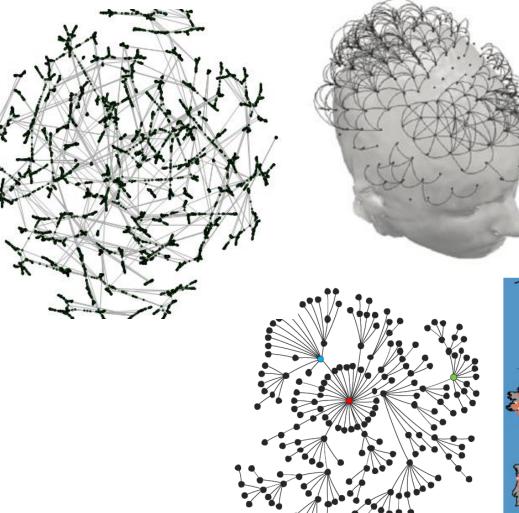


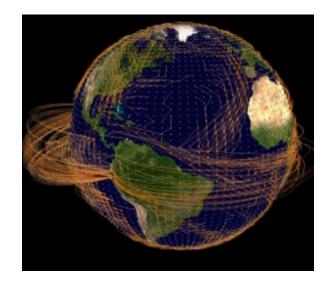
Presentation

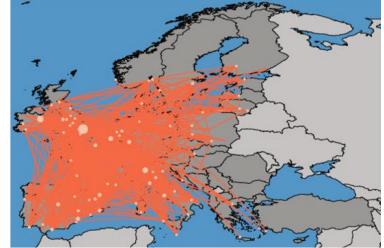
- Originally from Montevideo, Uruguay
- PhD in physics (lasers, Bryn Mawr College, USA)
- Since 2004 @ Universitat Politecnica de Catalunya.
- Member of the research group on *Dynamics, Nonlinear Optics and Lasers*.



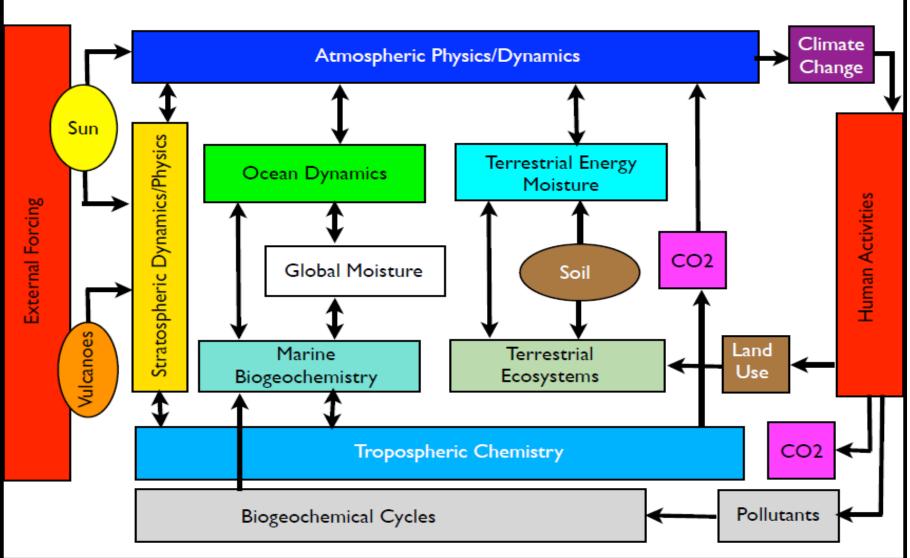
Complex systems



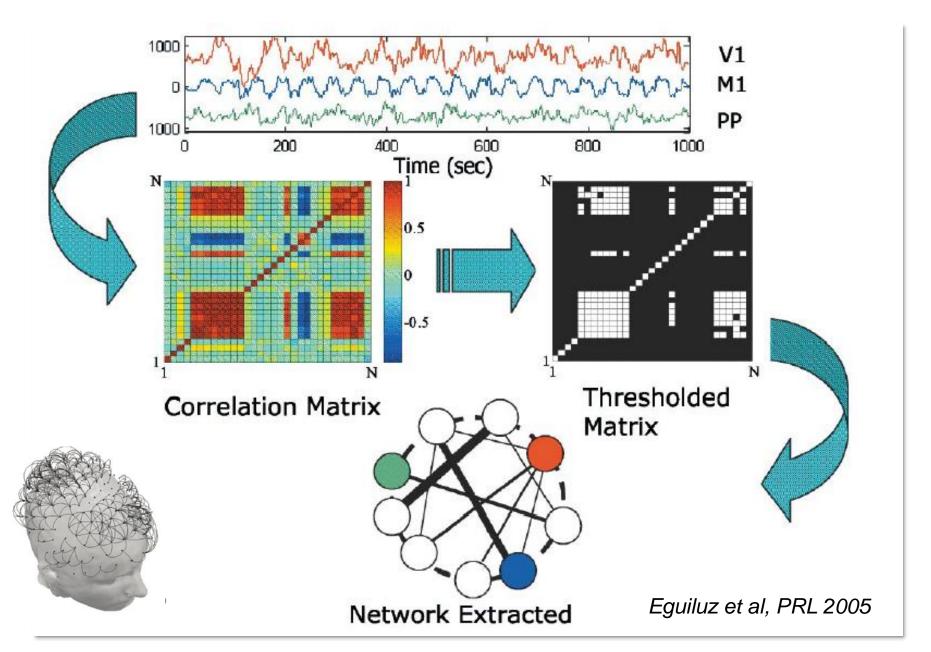




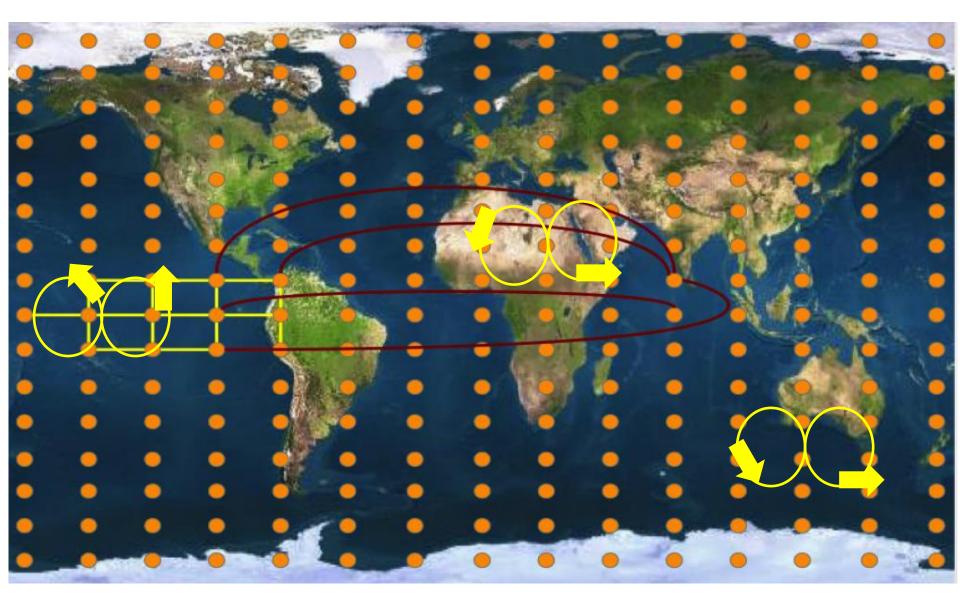
The Climate System

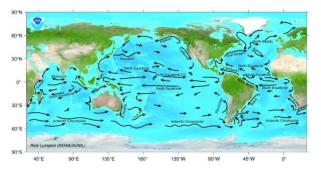


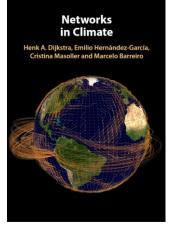
Functional brain network



We can use the same method to construct a climate network



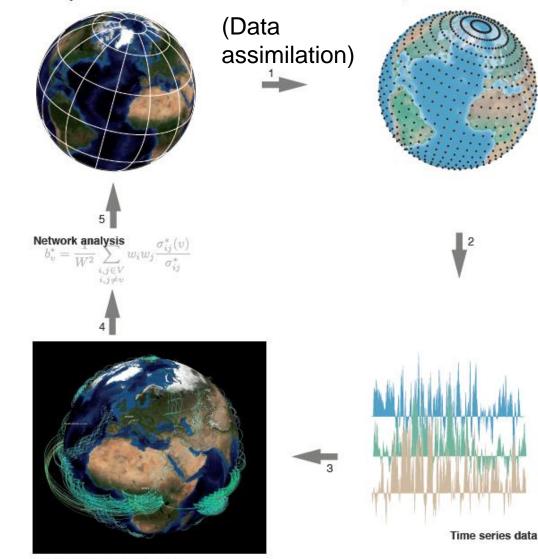




Cambridge University Press 2019

Earth system

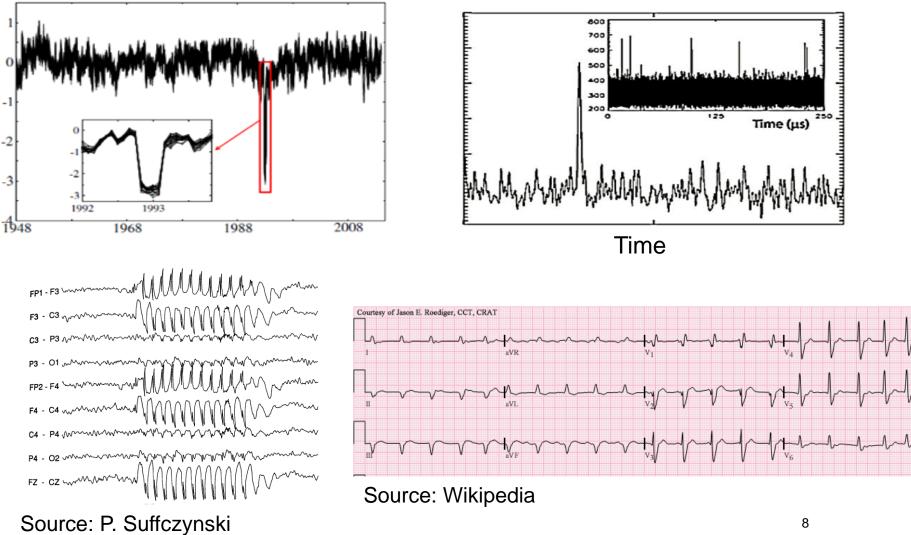
Grid points / observation sites



Functional climate network

Donges et al, Chaos 2015

Complex systems often display dynamical transitions and/or large fluctuations



1 SEC. 200 μV

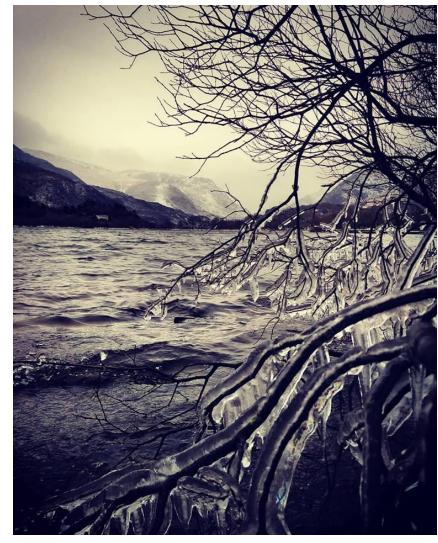
A main goal is to predict, and also, to try to understand, the dynamical behavior.



Are extreme fluctuations becoming more frequent? More extreme?







Credit: Richard Williams, North Wales, UK



Science & Environment

Rogue waves occurring less but 'becoming more extreme'

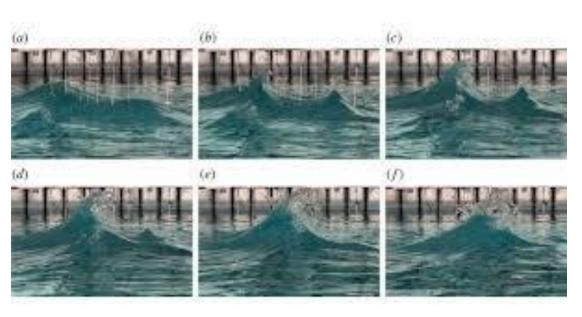
By Rebecca Morelle Science Correspondent, BBC News

() 21 March 2019





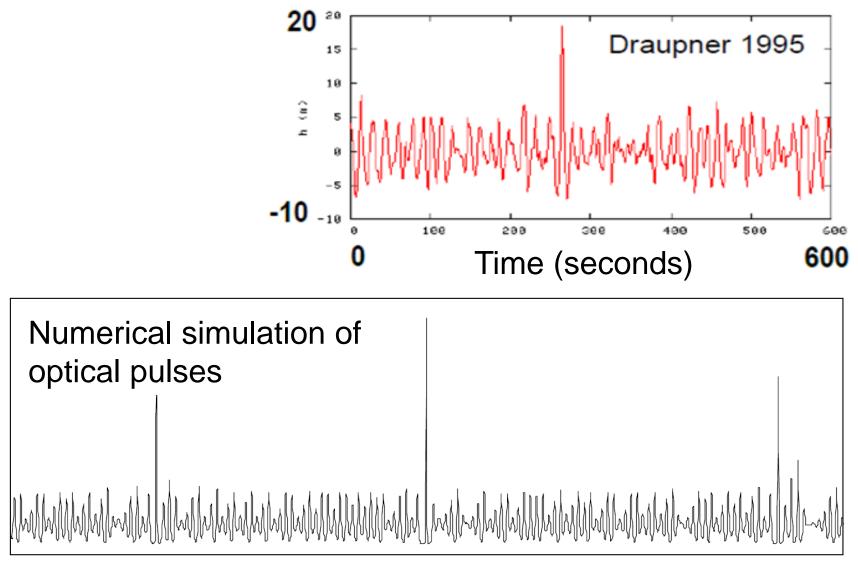




Rogue waves are a huge risk for the global shipping industry.

Are rogue waves in the ocean similar to those in a water tank, or in an optical system?

Ocean rogue waves and optical rogue waves



Time (some nanoseconds)

Time series analysis can unveil underlying statistical similarities in very different systems

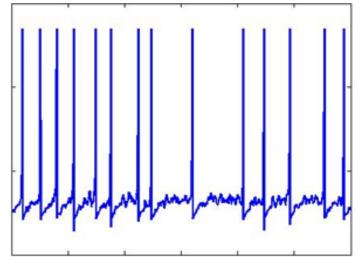


Laser intensity

Time

10⁻⁹ s

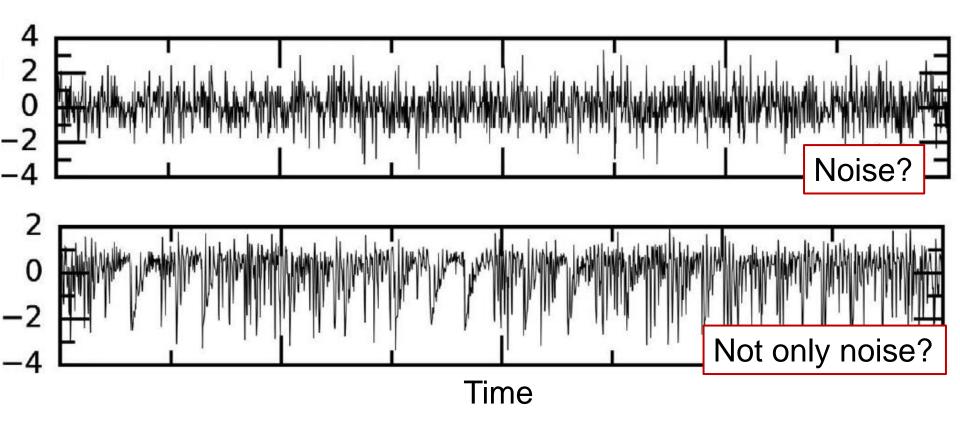
Neuron model



Time

10⁻³ s

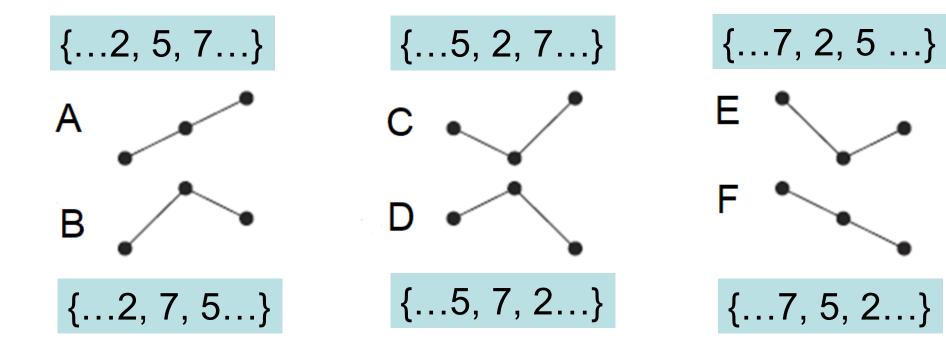
Methods of time-series analysis allow to detect and to quantify differences



First analysis method: ordinal analysis

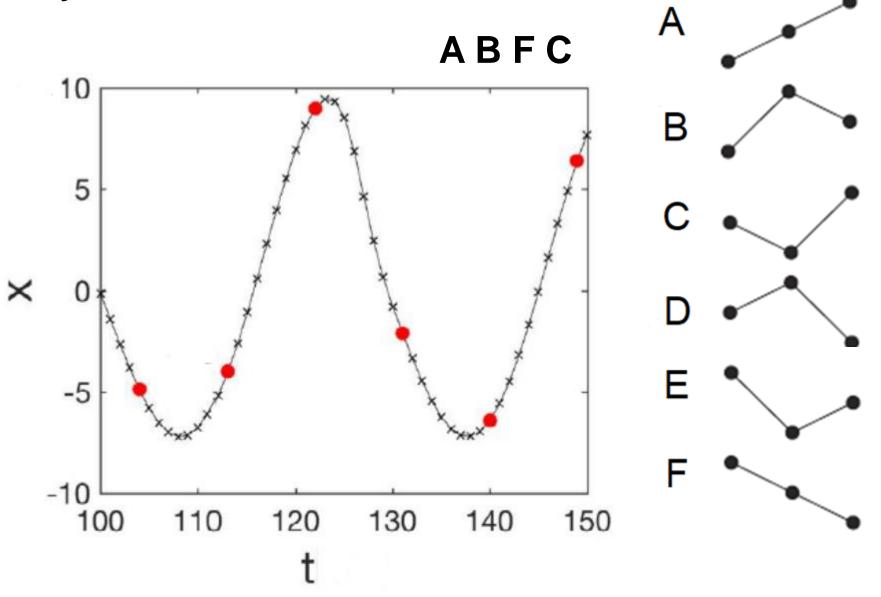
$$\{\dots X_i, X_{i+1}, X_{i+2}, \dots\}$$

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



Bandt and Pompe: Phys. Rev. Lett. 2002

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

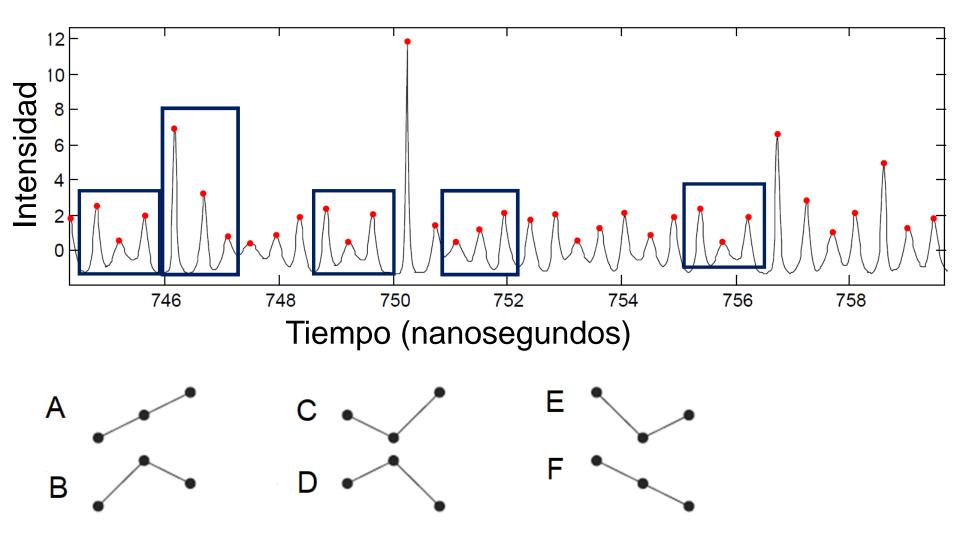


Possible number of patterns increases as D! (D=size of pattern)

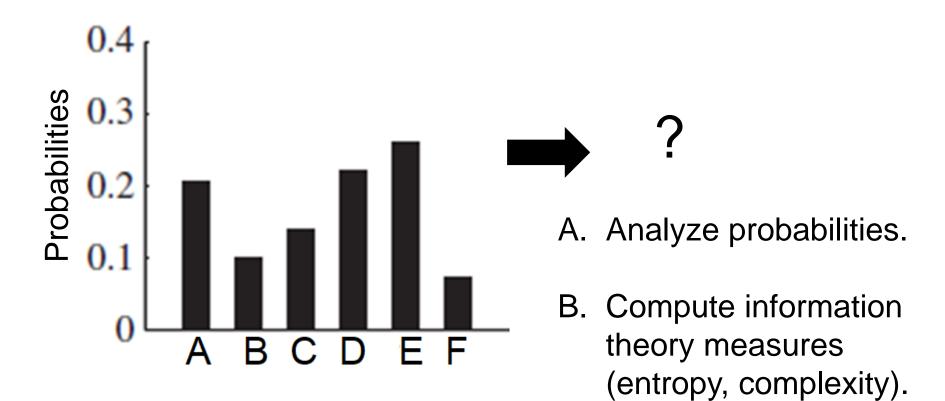
 $\{\ldots X_{i}, X_{i+1}, X_{i+2}, X_{i+3} \ldots\}$ 1 7 7 13 19 2 8 14 20 3 / 9 / 15 / 21 / 4 10 16 22 5 11 17 23 6 12 18 24

 $\{\dots X_{i}, X_{i+1}, X_{i+2}, X_{i+3}, X_{i+4} \dots\}$ 2 32 5 62 5 92 5 5 3 3 63 63 93 93 4 34 64 64 94 5 35 - 65 - 95 - 95 6 36 66 66 96 96 10 40 70 70 100 11 - 4 - 41 - 41 - 71 - 71 - 101 12 42 72 102 13 43 73 103 14 44 74 74 104 15 45 75 105 16 46 76 106 17 47 77 77 107 18 48 78 78 108 19 49 49 79 109 20 50 50 80 110 110 21 _____ 51 ____ 81 ____ 111 ____ 22 52 82 112 23 - 53 - 83 - 113 - 113 24 54 54 84 114 25 55 55 85 115 26 56 56 86 116 116 27 57 57 87 117 28 58 58 88 118 29 59 59 89 119 30 60 90 120

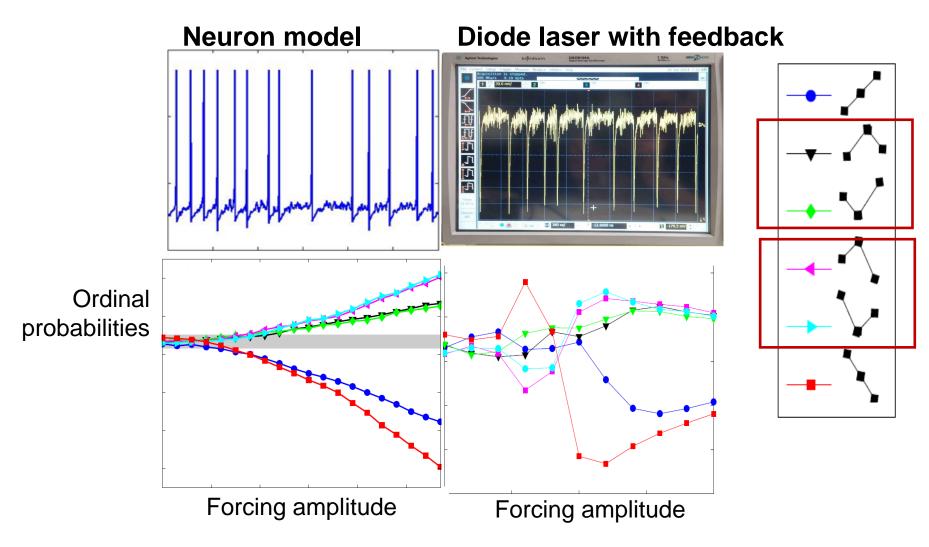
Which pattern occurs before high pulses?



For each time series we compute the probabilities of the different patterns



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



J. M. Aparicio-Reinoso et al PRE 94, 032218 (2016) A. Aragoneses et al, Sci. Rep. 4, 4696 (2014)

Interesting but useful? Maybe...

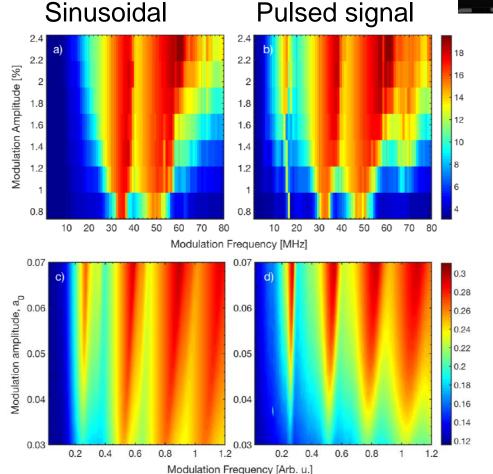
- Photonic neurons.
- Diode lasers: very low cost, highly energy efficient.
- Very very fast.
- Main challenge: understand how neurons receive, encode and process information.

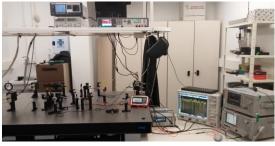




How a diode laser and a neuron encode a weak periodic signal? spike rate code?

Diode laser with optical feedback (experiments modulating the laser current)





Spike

rate in

color

code

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

FitzHugh-Nagumo neuron model (with the same input signal)

Temporal code?

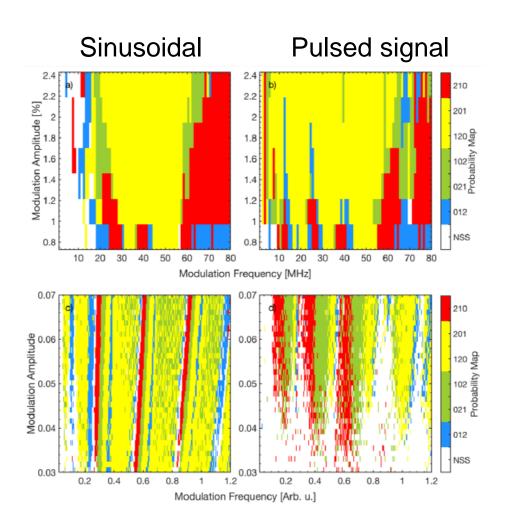
Ordinal analysis unveils some differences in spike timing.



FitzHugh-

Nagumo

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).

B. By computing information theory measures (entropy, complexity) we can characterize dynamical transitions

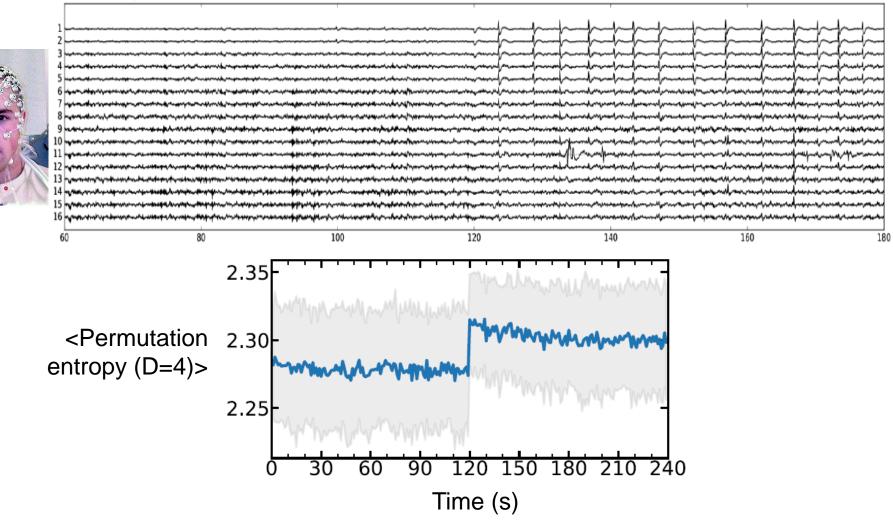
The time-series is described by a set of probabilities $\sum_{i=1}^{n} p_i = 1$

Shannon entropy:
$$H = -\sum_{i} p_i \log_2 p_i$$

- Interpretation: "quantity of surprise one should feel upon reading the result of a measurement" K. Hlavackova-Schindler et al, Physics Reports 441 (2007)
- The entropy computed from the ordinal probabilities is known as permutation entropy.

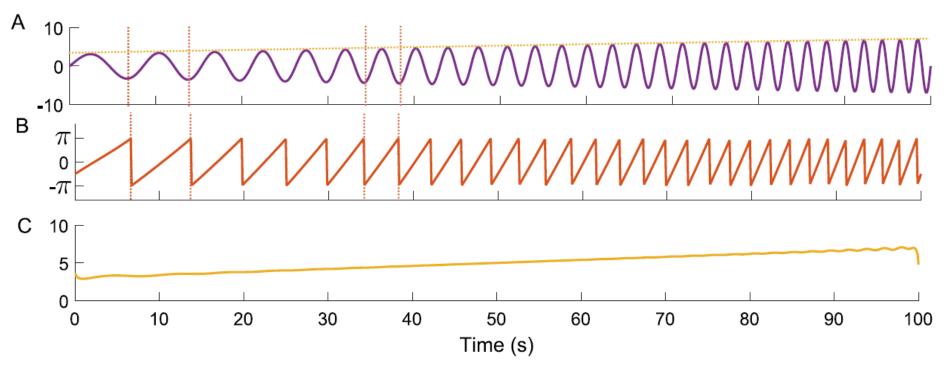


Eyes open



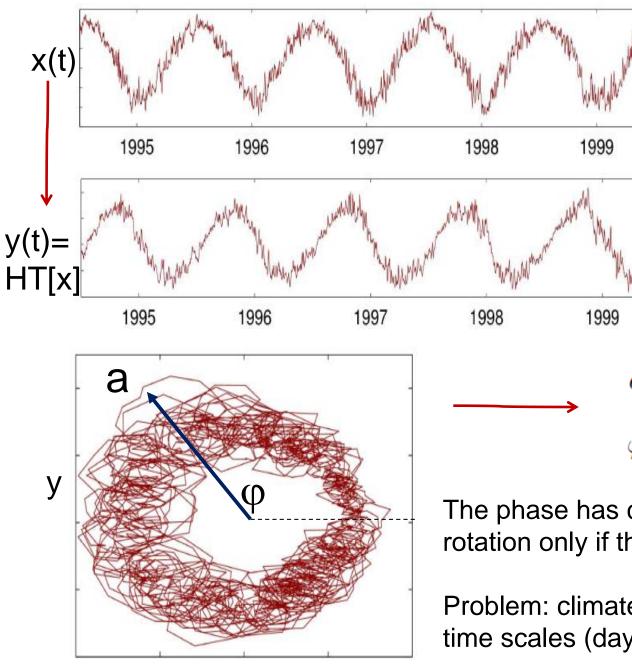
C. Quintero-Quiroz et al, "*Differentiating resting brain states using ordinal symbolic analysis*", Chaos 28, 106307 (2018).

Second analysis method: The Hilbert transform



- (A) The original signal. (B) The instantaneous phase extracted using the Hilbert transform. (C) The instantaneous amplitude.
 A = C cos(B).
- How to calculate the instantaneous amplitude and phase?

G. Lancaster et al, Physics Reports 748 (2018) 1-60



Х

Surface air temperature

$$a_j(t) = \sqrt{x_j^2(t) + y_j^2(t)}$$

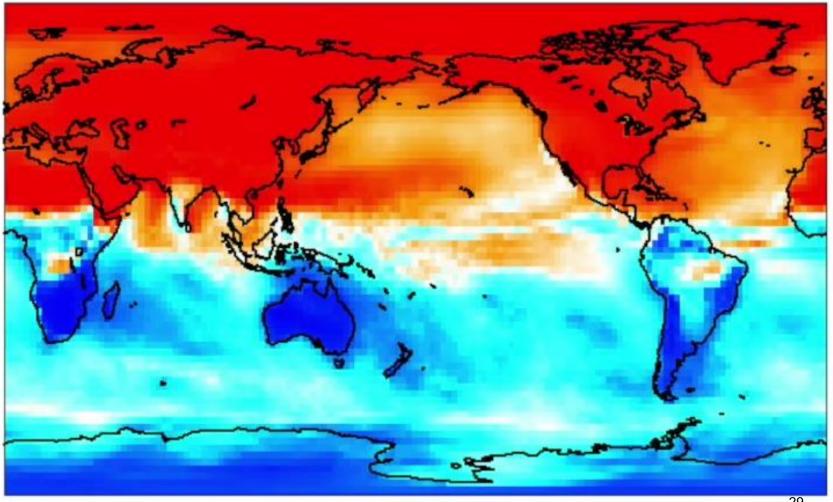
 $\varphi_j(t) = \arctan[y_j(t)/x_j(t)]$

The phase has clear physical meaning of rotation only if the signal is "narrow band".

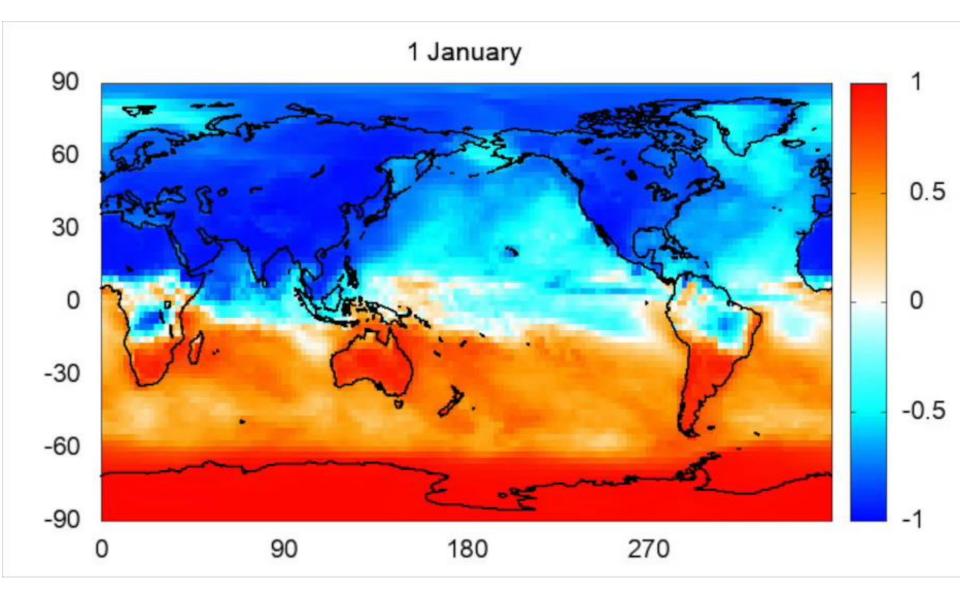
Problem: climate dynamics involves many time scales (days – decades). 28

Cosine of the Hilbert phase in color code. Where the data comes from? Daily reanalysis, 1979 to 2017

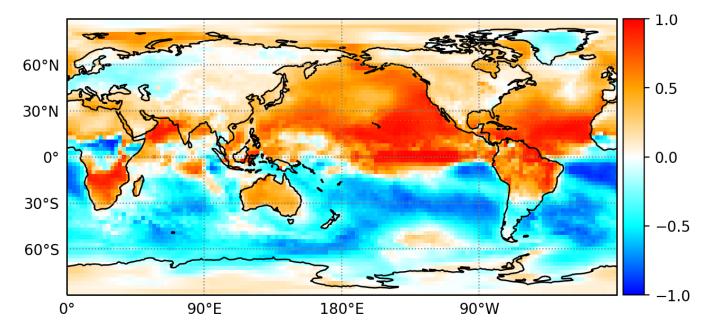
1 July

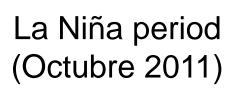


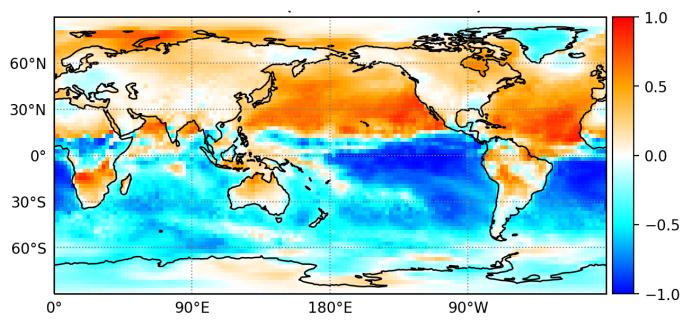
Hilbert visualization of the seasons: temporal evolution of the cosine of the phase



El Niño period (October 2015)







How to detect significant changes in the last 30 years?

$$\Delta a = \langle a \rangle_{2016-2007} - \langle a \rangle_{1988-1979}$$
$$\frac{\Delta a}{\langle a \rangle_{2016-1979}}$$

0

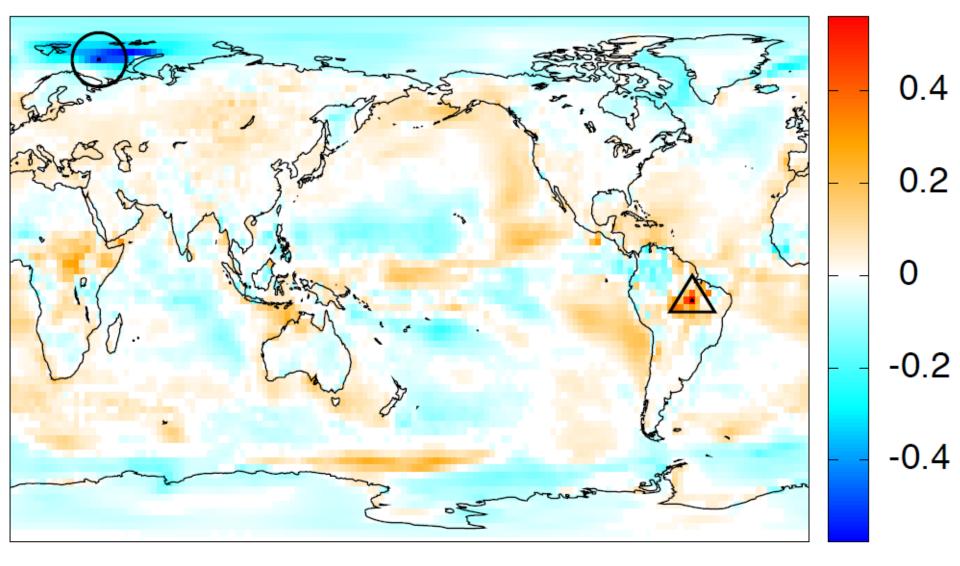
Significant if

$$\frac{\Delta a}{\langle a \rangle} \geq \langle . \rangle_s + 2\sigma_s$$

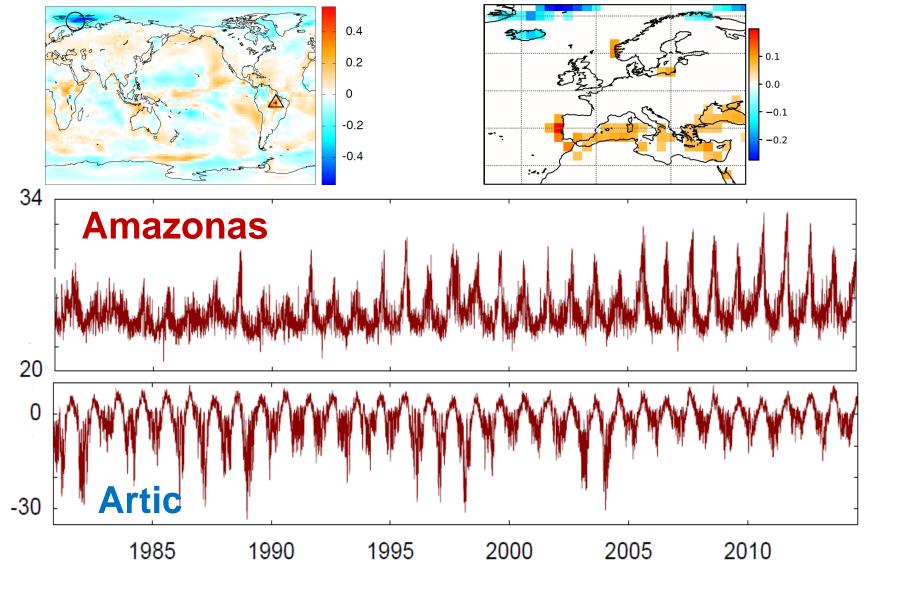
$$\frac{\Delta a}{\langle a \rangle} \leq \langle . \rangle_s - 2\sigma_s$$

Average over 100 "surrogates"





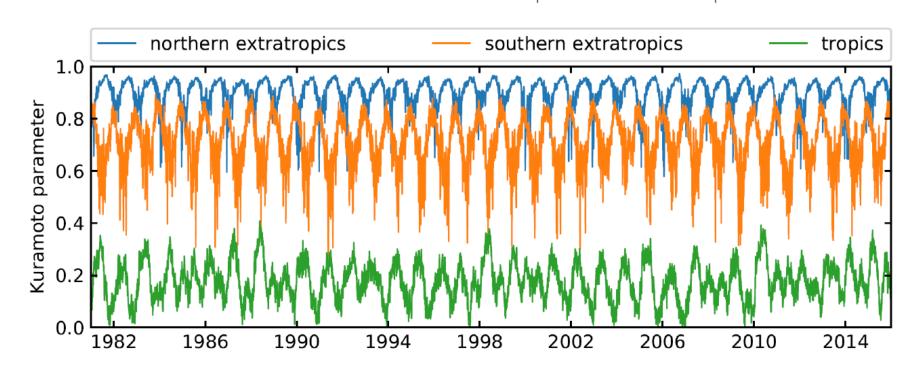
D. A. Zappala, M. Barreiro, and C. Masoller, "*Quantifying changes in spatial patterns of surface air temperature dynamics over several decades*", Earth Syst. Dynam. **9**, 383–391 (2018).



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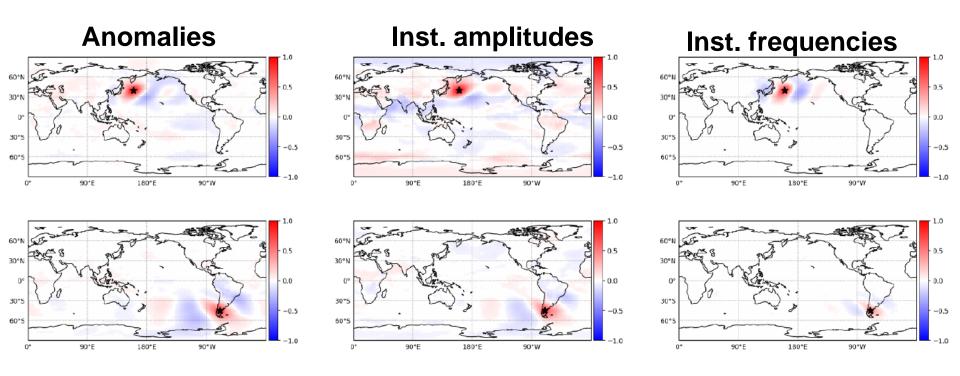
Quantification of phase synchronization

area-weighted Kuramoto parameter: $r(t) = \left| \frac{\sum_{j \in S} w_j \exp[i\varphi_j(t)]}{\sum_{i \in S} w_i} \right|$



D. A. Zappala, M. Barreiro and C. Masoller, "Quantifying phase synchronization and unveiling Rossby wave patterns in surface air temperature dynamics", in preparation.

Identification of Rossby waves



Crosscorrelation in color code.

D. A. Zappala, M. Barreiro and C. Masoller, "Quantifying phase synchronization and unveiling Rossby wave patterns in surface air temperature dynamics", in preparation.

Take home message

- Time series analysis allows to understand, predict, or classify dynamical behaviors of complex systems.
- Even when the data does not meet the mathematical or algorithmic requirements, the results can give useful insights.
- Many interdisciplinary applications.



Thanks to: Jordi Tiana (UPC) Carlos Quintero (UPC) Dario Zappala (UPC) Marcelo Barreiro (Uruguay)

Thank you for your attention!

CRFA

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