

Nonlinear time-series analysis and complex network approach for identifying and characterizing regime transitions

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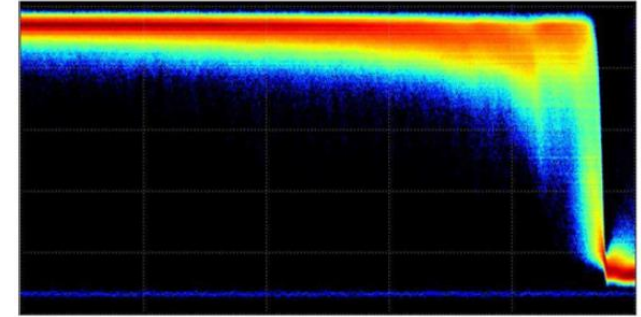
TransNet 2016
Amsterdam



Dynamical regime transitions in optical systems

■ Polarization switching

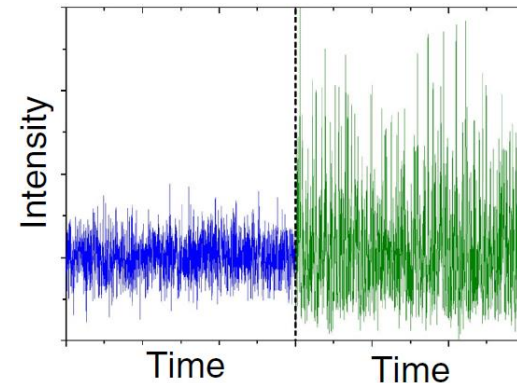
Semiconductor laser output as the pump current increases



Time

■ Transition to turbulence

Raman fiber laser output as the pump power increases



Scope of this presentation:

- Data analysis tools based on symbolic analysis and network representation of time-series provide new insight into these phenomena.
- Optical data can be useful for testing novel analysis tools.

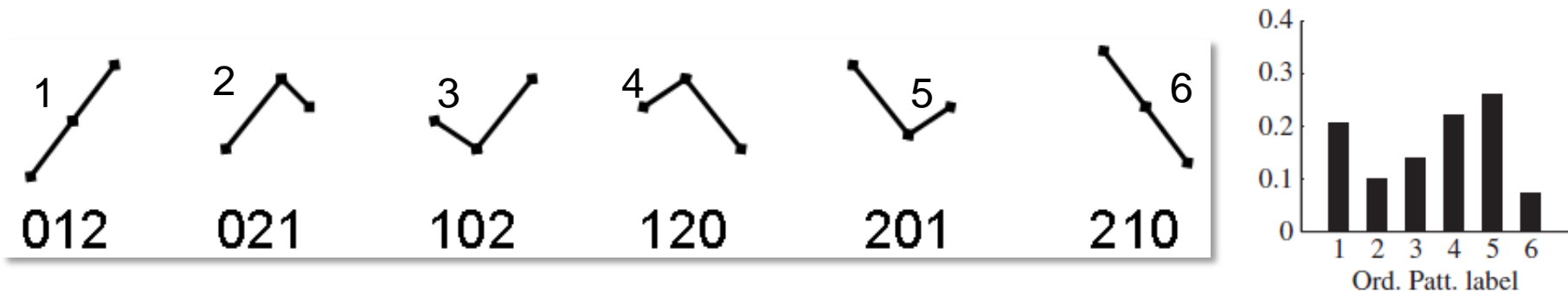


- Tools to analyze the data
- Early-warning signs of upcoming PS
- Temporal correlations during laminar-turbulence transition
- Summary

Method of nonlinear **symbolic** time-series analysis: ordinal patterns

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

Brandt & Pompe, PRL 88, 174102 (2002)

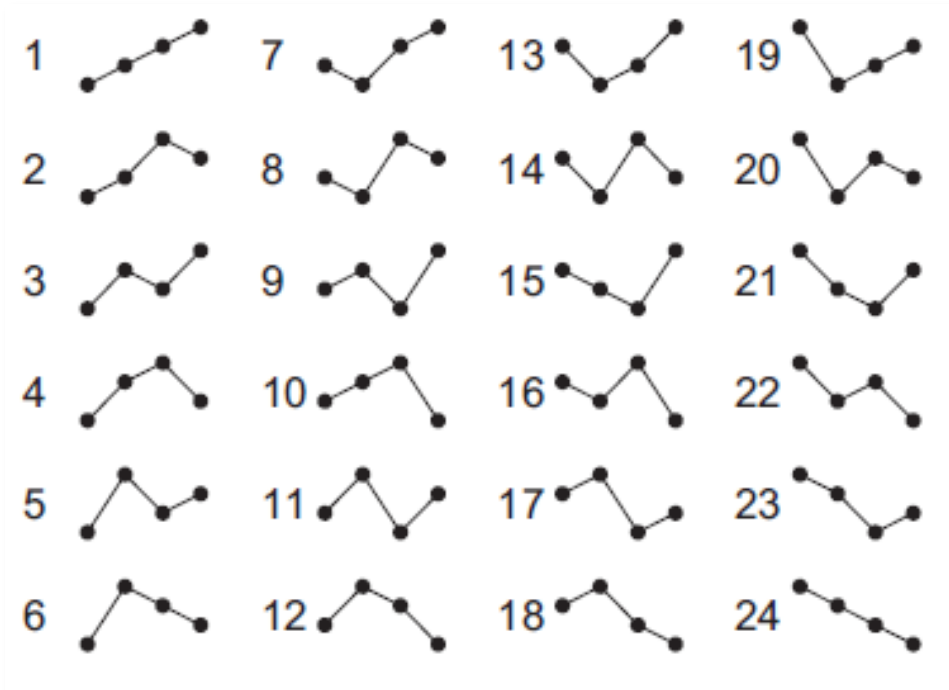


The OP probabilities allow identifying more expressed and/or infrequent patterns in the order of the sequence of data values.

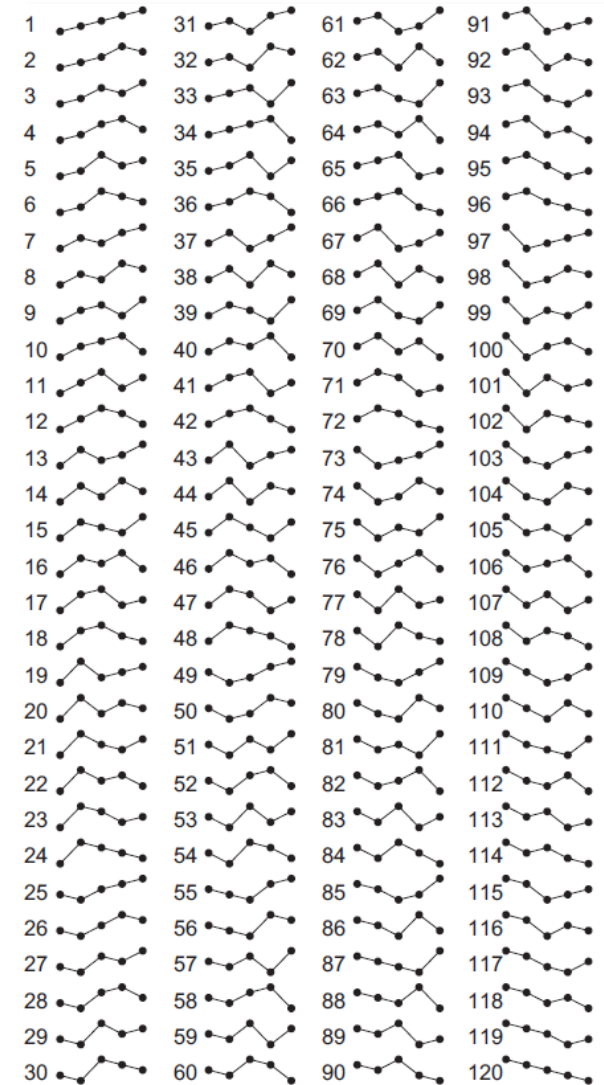
Random data?
(OPs equally probable)

- Advantage: the probabilities uncover temporal correlations.
- Drawback: we lose information about the actual values.
 - ⇒ Ordinal analysis gives complementary information to that gained with other analysis tools.

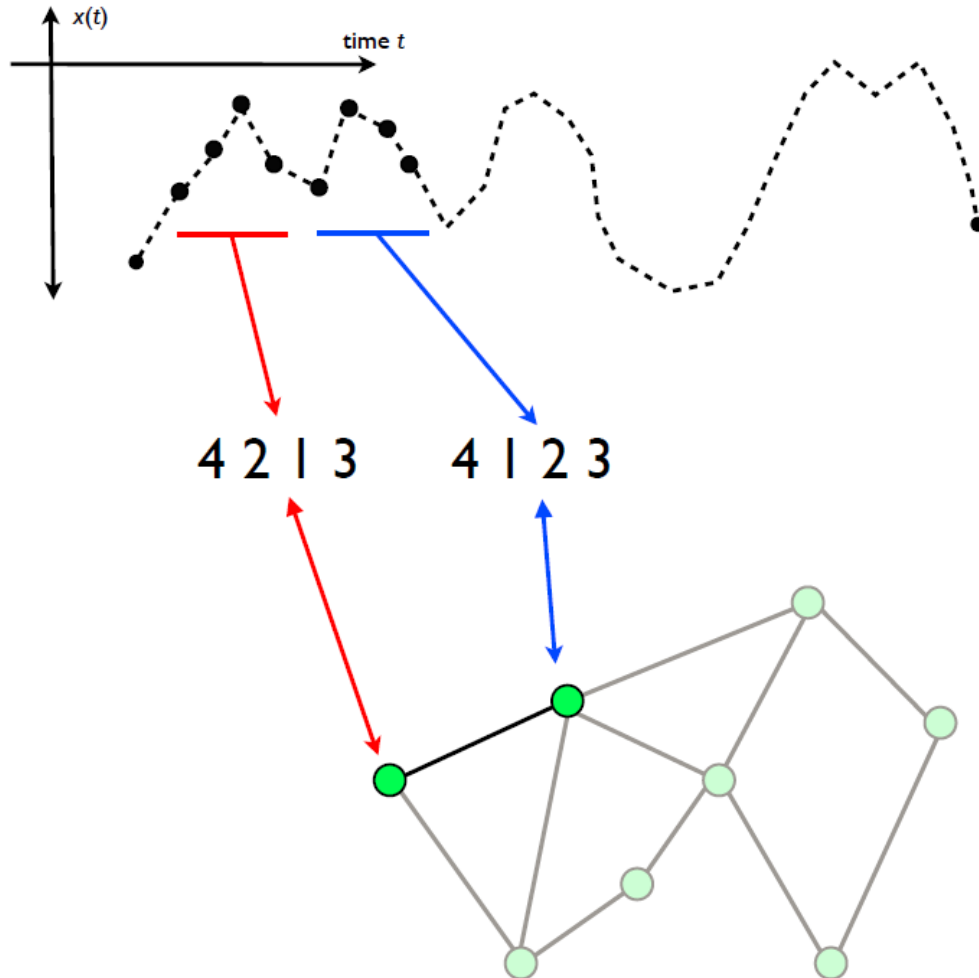
The number of patterns increases as D!



Opportunity: turn a time-series into a network by using the patterns as the “nodes” of the network.



The network nodes are the “ordinal patterns”, and the links?



- The links are defined in terms of the probability of pattern “ β ” occurring after pattern “ α ”.
- Weighs of nodes: the probabilities of the patterns ($\sum_i p_i = 1$).
- Weighs of links: the probabilities of the transitions ($\sum_j w_{ij} = 1 \forall i$).

\Rightarrow **Weighted and directed network**

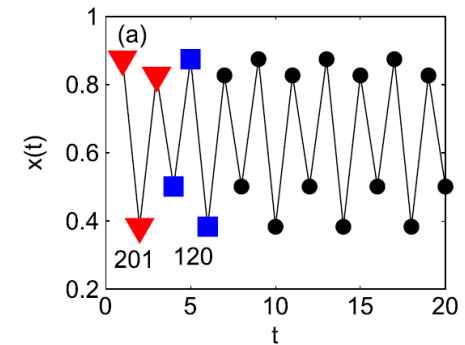
Three network-based diagnostic tools

- Entropy computed from the weights of the nodes (**permutation entropy**)

$$s_p = -\sum p_i \log p_i$$

- Entropy computed from weights of the links (**transition probabilities**, '01'→'01', '01'→'10', etc.)

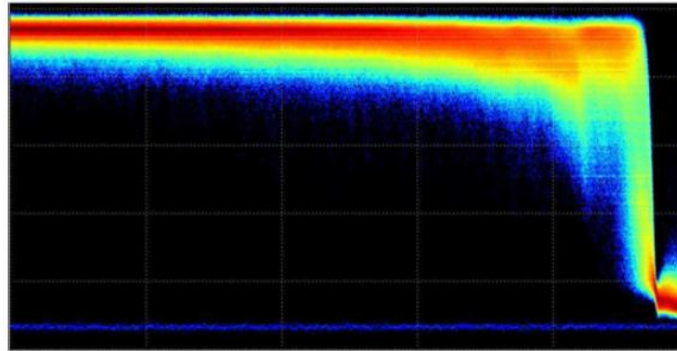
$$w_{ij} = \frac{\sum_{t=1}^{L-1} n[s(t) = i, s(t+1) = j]}{\sum_{t=1}^{L-1} n[s(t) = i]}$$



- Asymmetry coefficient: normalized difference of transition probabilities, $P('01' \rightarrow '10') - P('10' \rightarrow '01')$, etc.

$$a_c = \frac{\sum_i \sum_{j \neq i} |w_{ij} - w_{ji}|}{\sum_i \sum_{j \neq i} (w_{ij} + w_{ji})}$$

(0 in a fully symmetric network;
1 in a fully directed network)



Identifying early signs of upcoming transition

- “optical big data”: useful for testing novel diagnostic tools



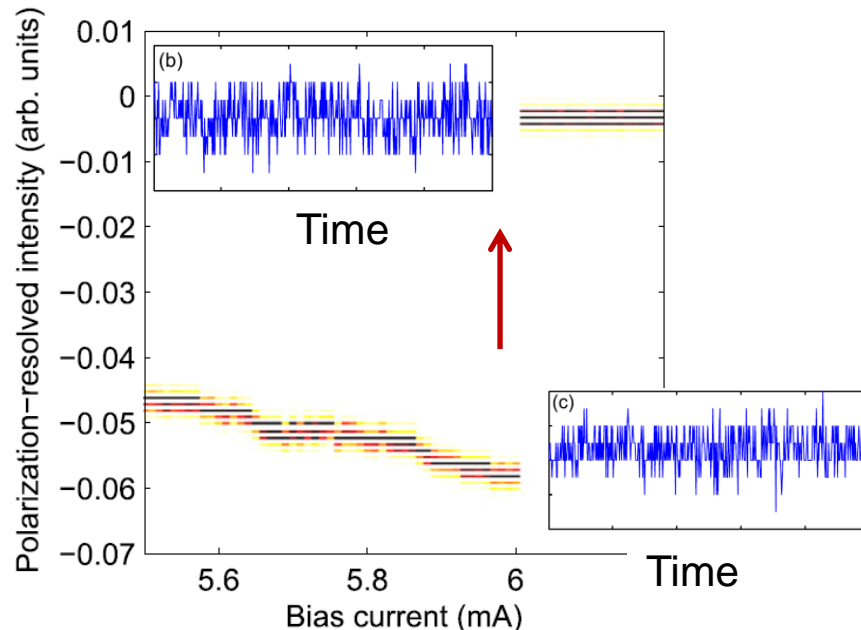
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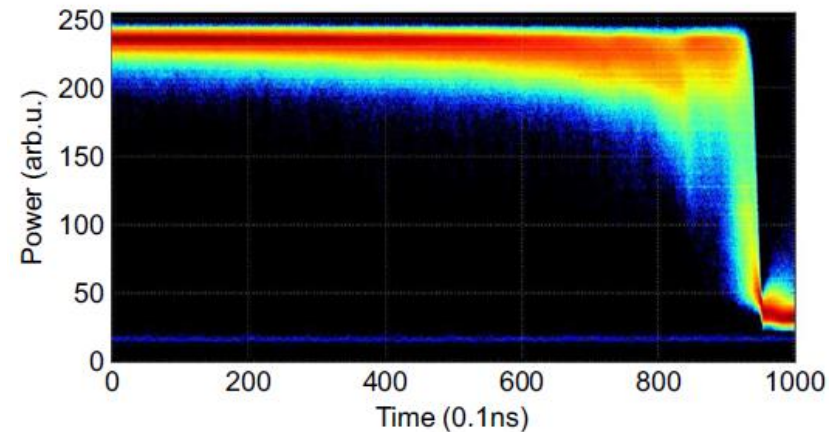
Experimental data from INLN & Bangor University
(S. Barland & Y. Hong)

VCSEL polarization-resolved intensity: two sets of experiments

- Time series recorded with pump current constant in time.
- Record the turn-on of the orthogonal mode.



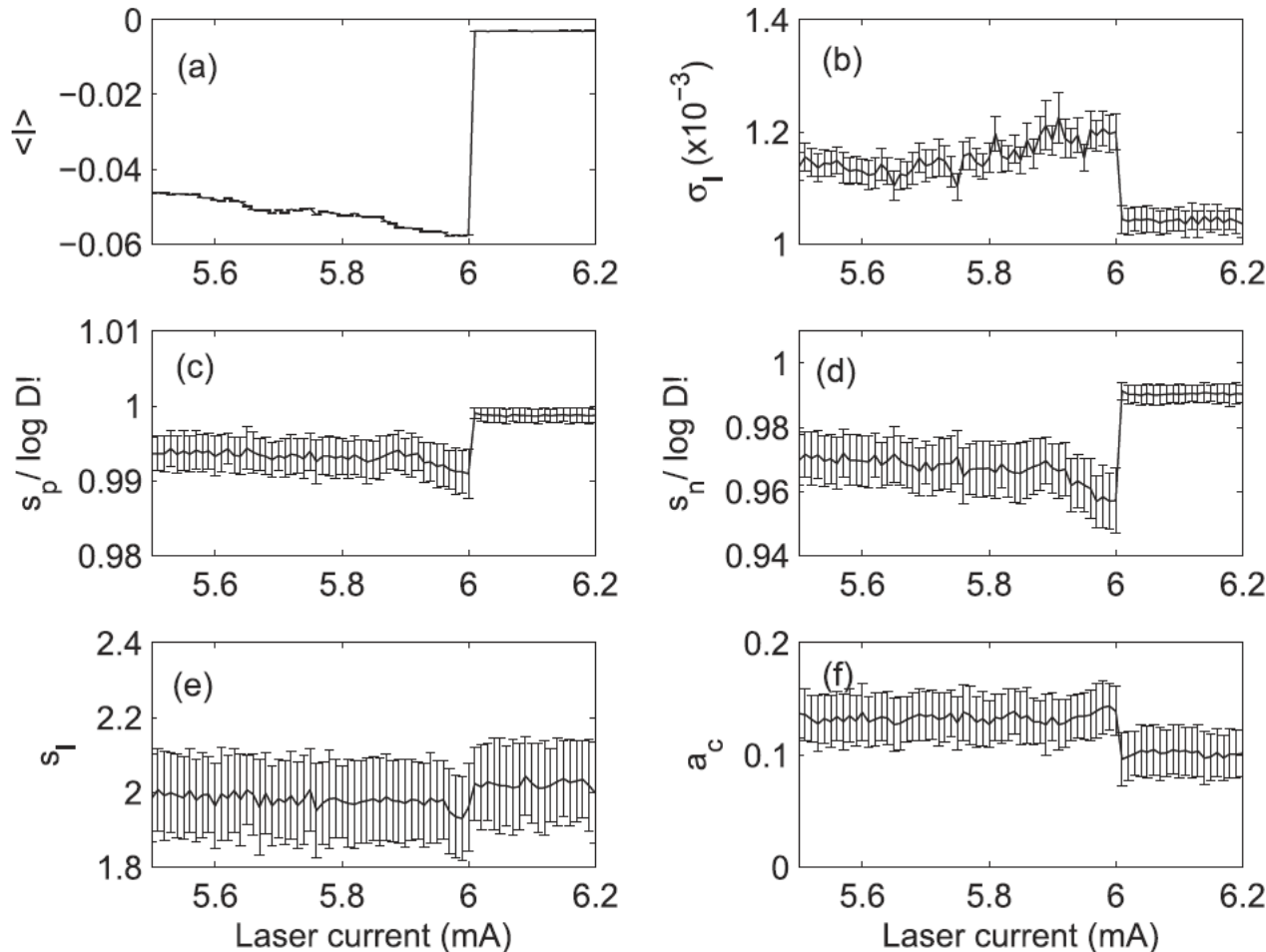
- Time series recorded with pump current varying in time.
- Record the turn-off of the fundamental mode.



Is it possible to anticipate the PS?

No if the mechanisms that trigger the PS are fully stochastic.

Results for constant pump current & turn-on of the orthogonal mode

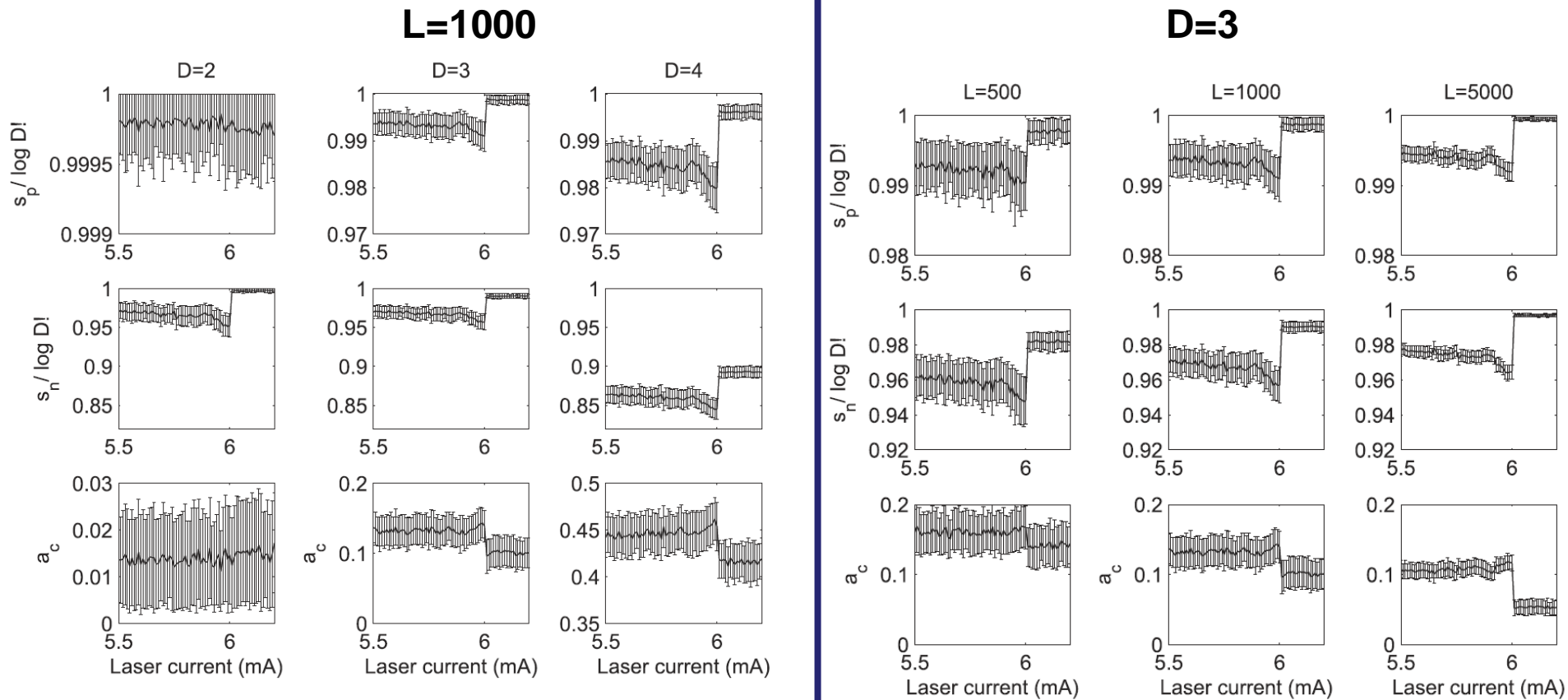


⇒ Despite of the stochasticity of the time-series, the measures “anticipate” the PS.

⇒ Deterministic mechanisms involved.

Error bars computed from 100 non-overlapping windows with $L=1000$ data points each. Length of the pattern $D=3$.

Influence of the length of the pattern (D) and of length of the time-series (L)

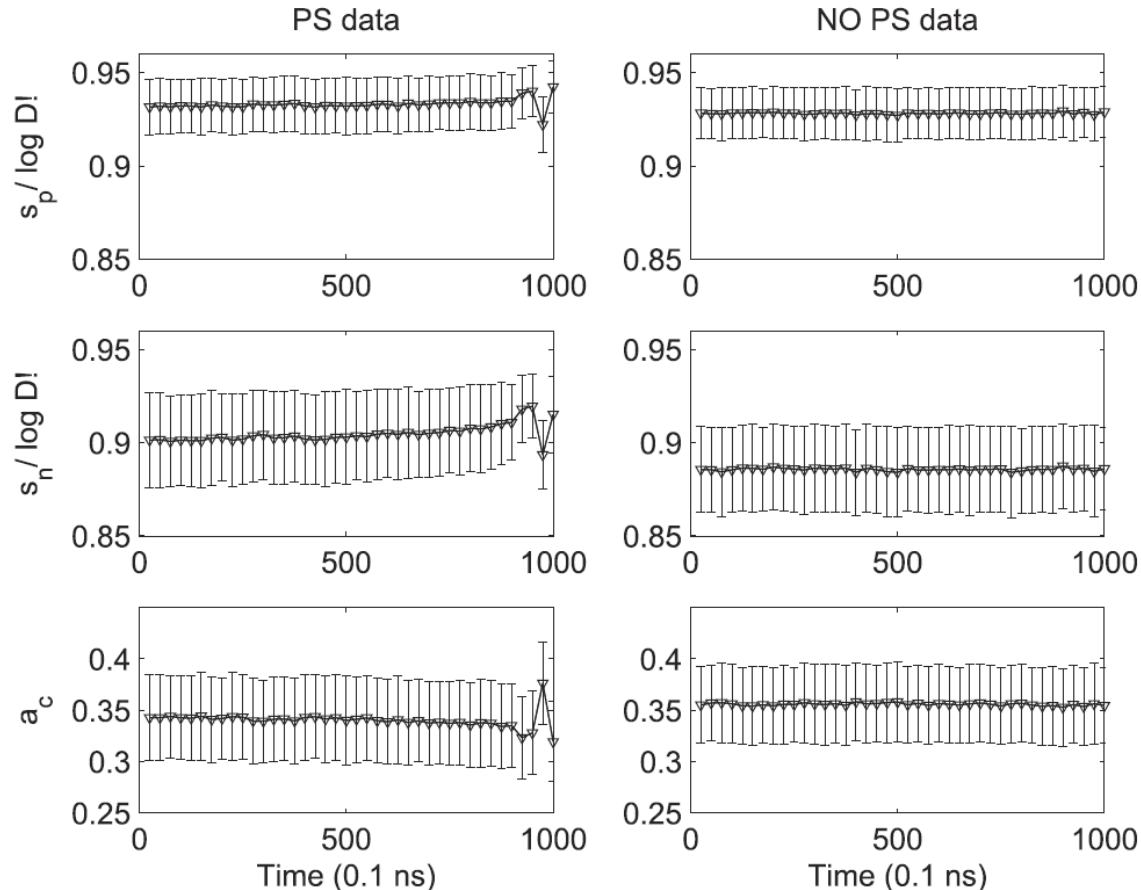


⇒ Transition detected even for short dataset (L=500 with D=3).
Open issues: How to quantify performance? Optimal D depends on L?

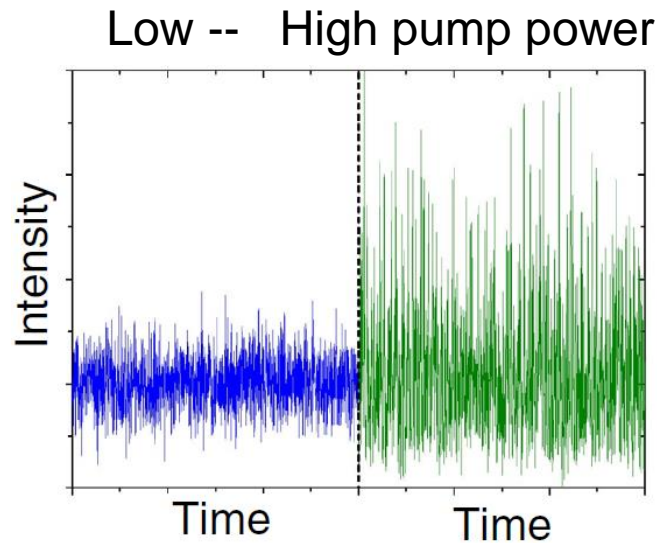
Results for time-varying pump current & turn-off of the fundamental mode

Slightly different optical feedback conditions result in PS or no PS.

Analysis done with $D=3$, error bars computed with 1000 time series $L=500$.



Another open issue: comparison with other diagnostic tools



Characterizing the laminar-turbulence transition in a fiber laser



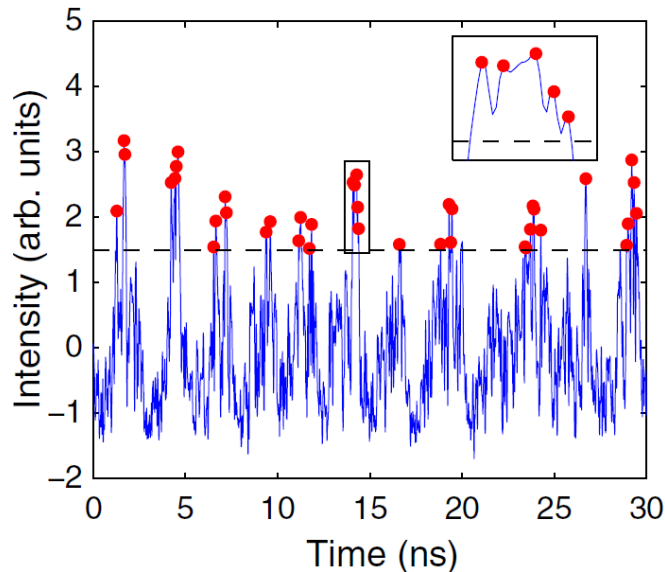
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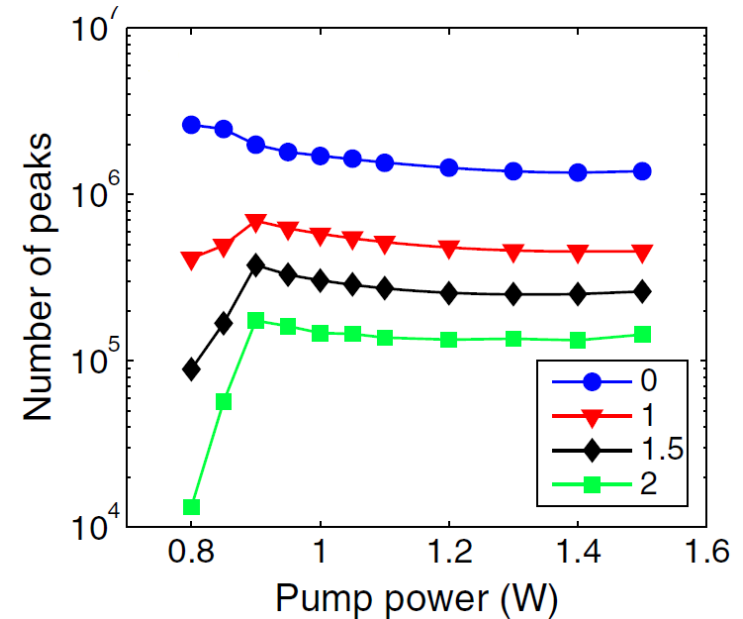
Experimental data from Aston University, UK
(Prof. Turitsyn' group)

Analysis of the intensity peaks higher than a threshold

Each time series is first normalized to $\langle I \rangle = 0$ and $\sigma = 1$



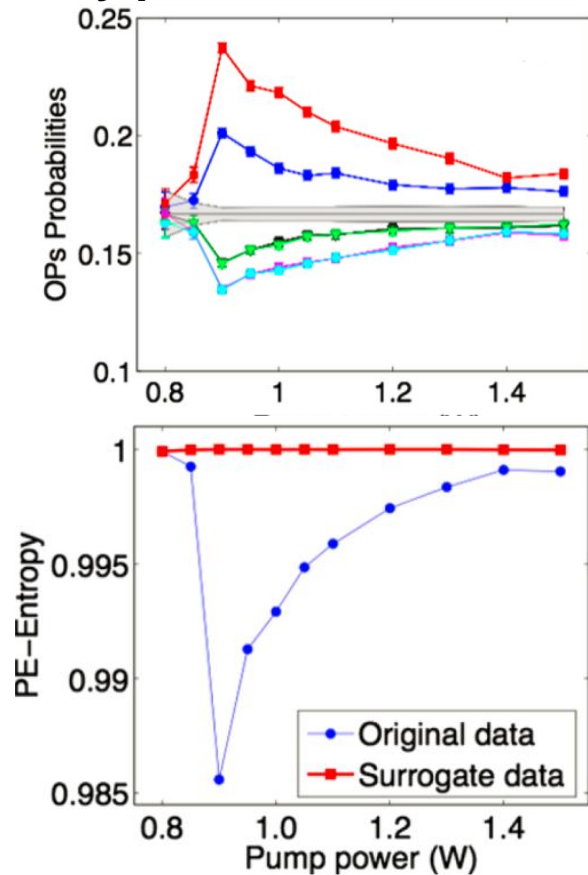
$$\{I_{\max,i}\}$$



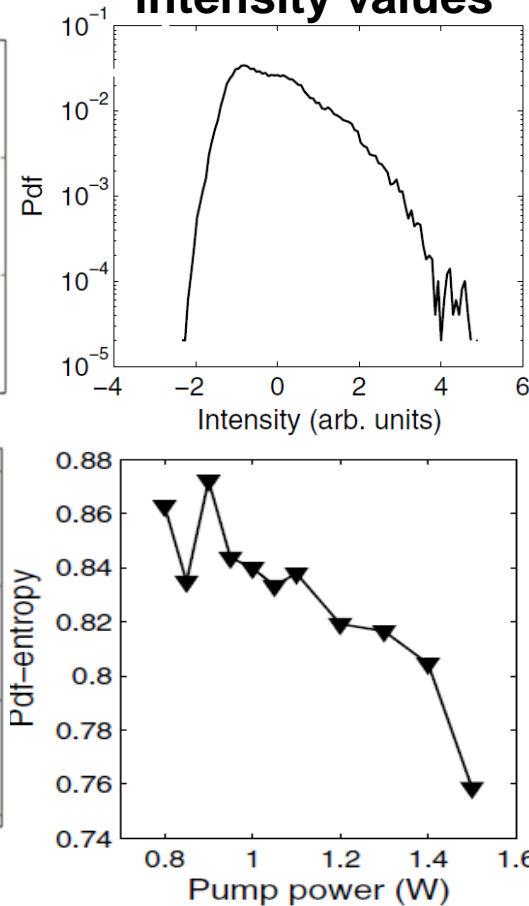
In each time series 5×10^7
data points.
Sampling time $dt = 12.5$ ps

Th = 2: number of
peaks $> 10^4$ for all
values of the pump
power

Ordinal analysis of the intensity peaks above $Th = 2$



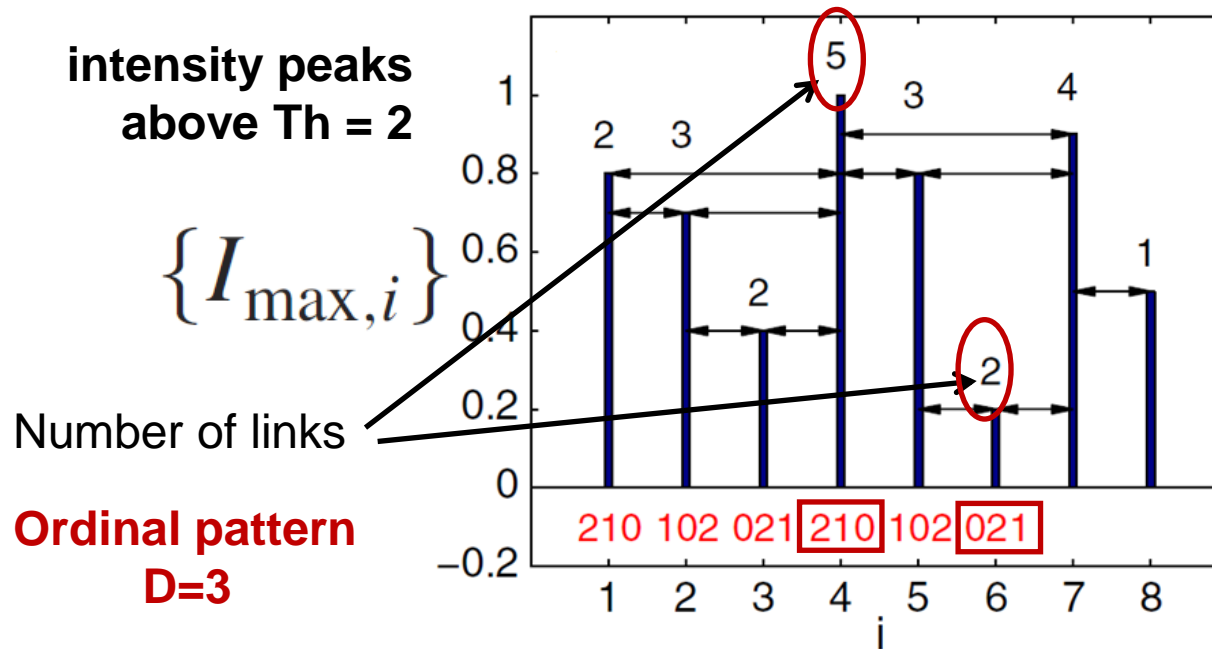
Histogram of raw intensity values



- Sharp transition not captured by standard histogram analysis.
- Different entropy behavior.

Second diagnostic tool: horizontal visibility graph (HVG)

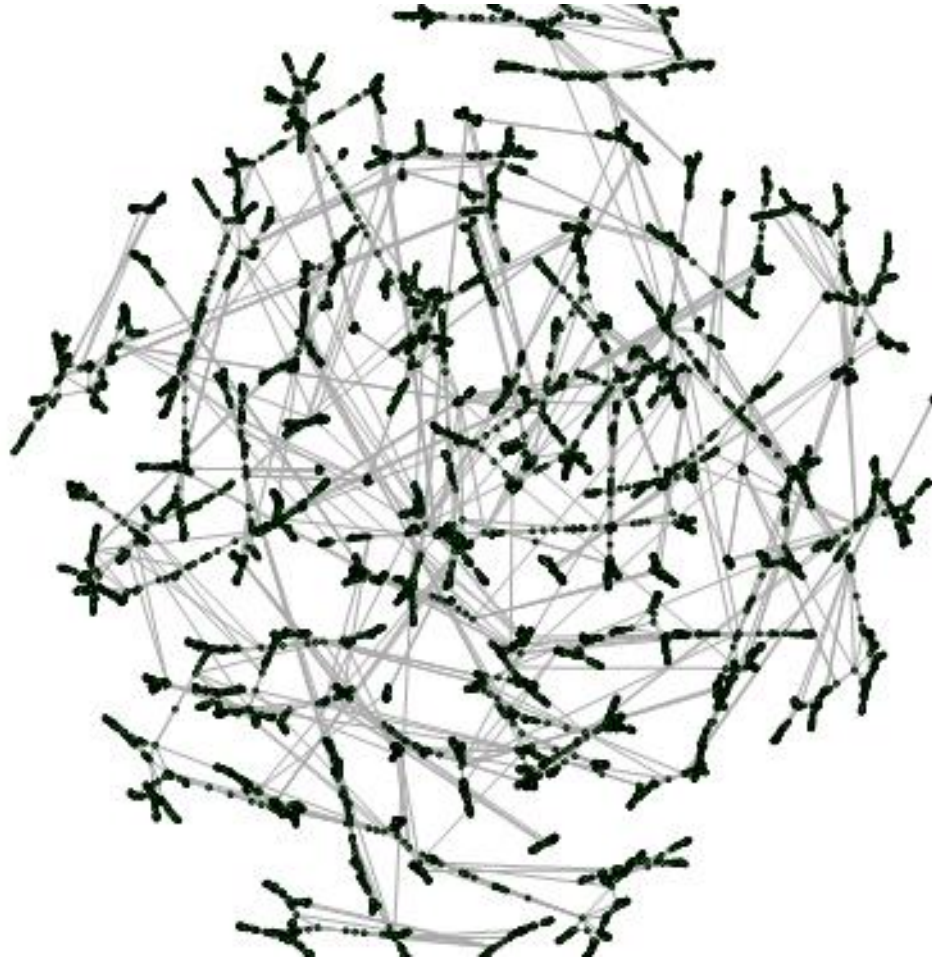
A time-series is represented as a graph, where each data point is a node



- Rule: data points i and j are connected if there is “visibility” between them: $I_{\max,i}$ and $I_{\max,j} > I_{\max,n}$ for all $n, i < n < j$

\Rightarrow ***Unweighted and undirected graph***

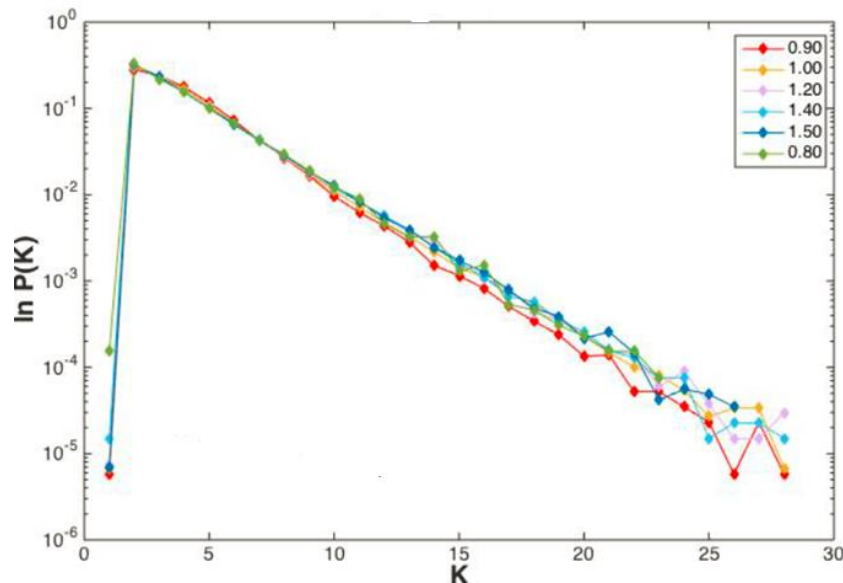
The obtained graph



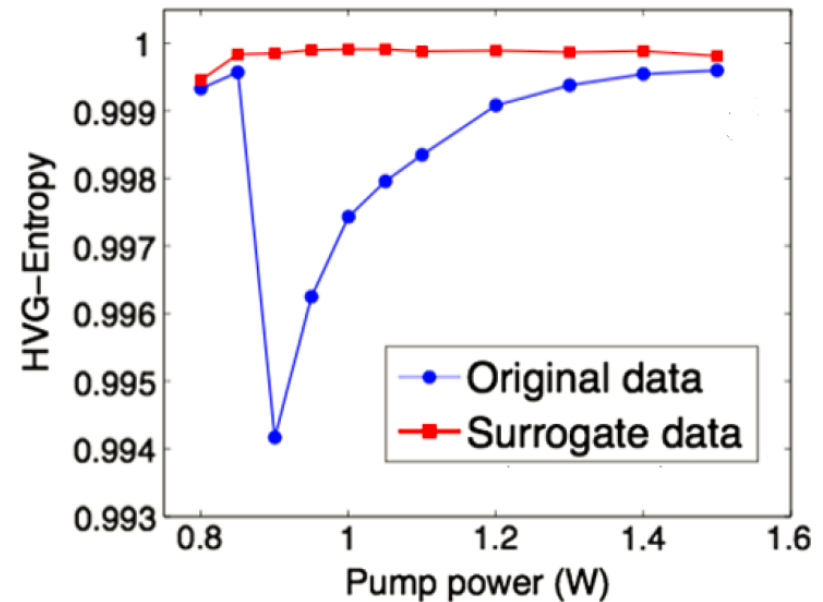
How to characterize this graph?

⇒ Degree Distribution (distribution of the number of links)

- Degree distribution for various pump powers using $Th=2$.



- Entropy of the degree distribution (normalized to the entropy of Gaussian white noise)

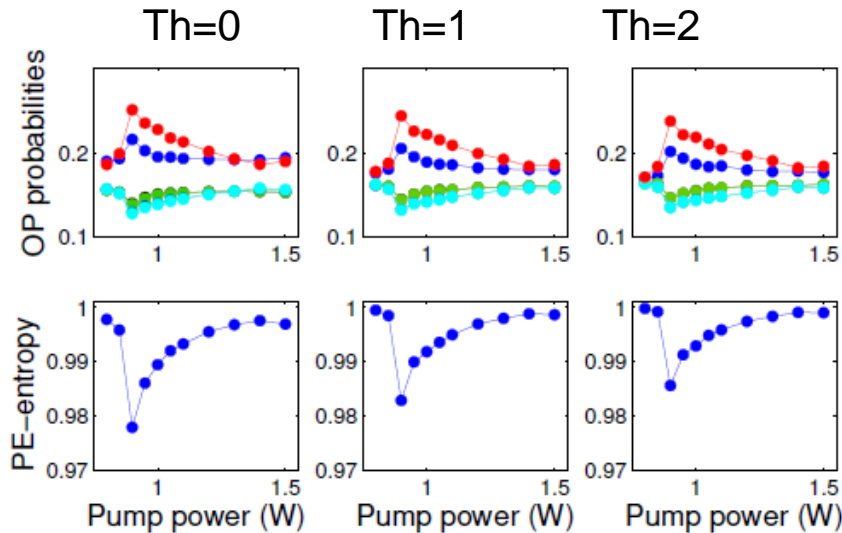


⇒ sharp transition also seen with HVG technique.

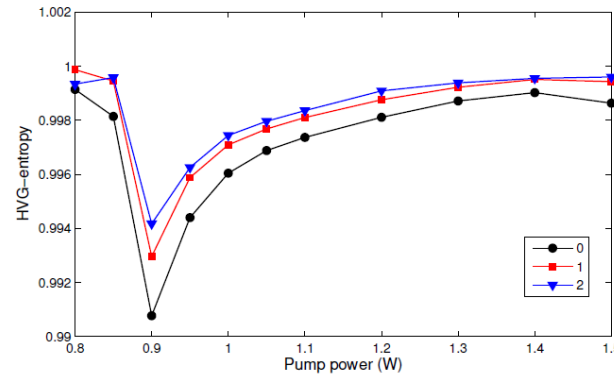
Influence of the threshold

Raw data $\{.../i...\} \Rightarrow \text{Th} \Rightarrow \{.../_{\max,i}...\}$

Ordinal analysis

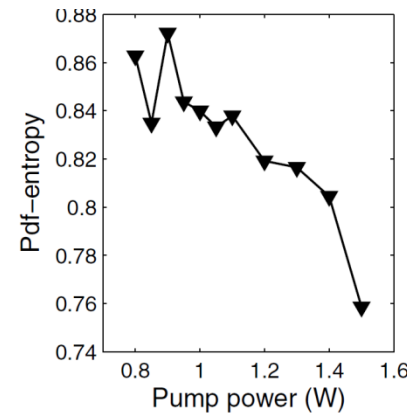
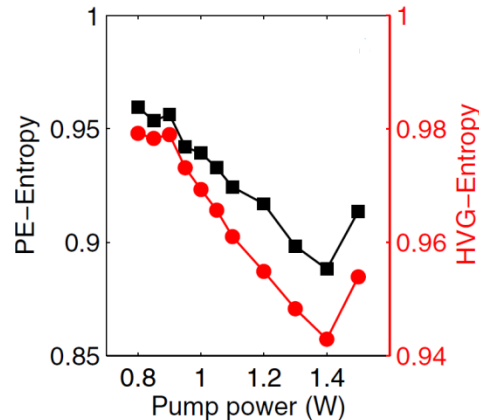


HVG analysis



⇒ sharp transition seen with different thresholds.

With raw data



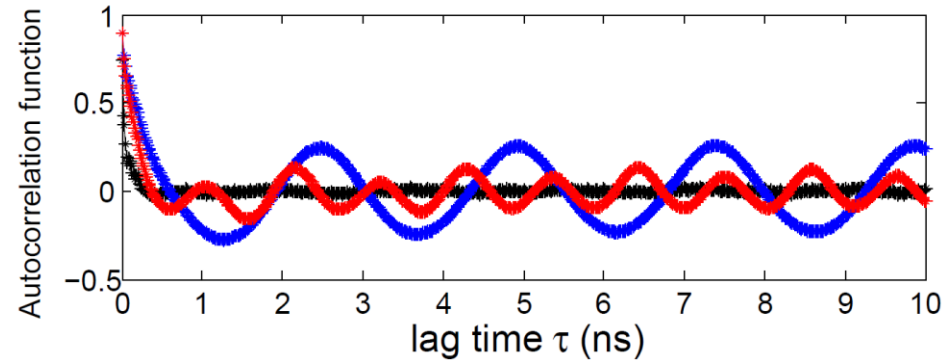
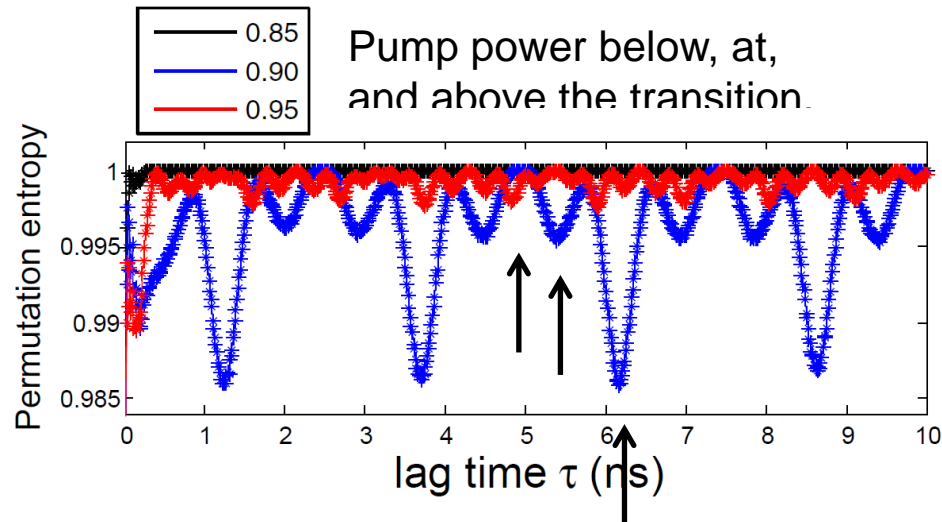
⇒ sharp transition not seen.

Can we obtain more info. from the raw data?

Ordinal analysis of **lagged** raw data

$$\{I_i, I_{i+\tau}, I_{i+2\tau}, \dots\};$$

all data points, no threshold used

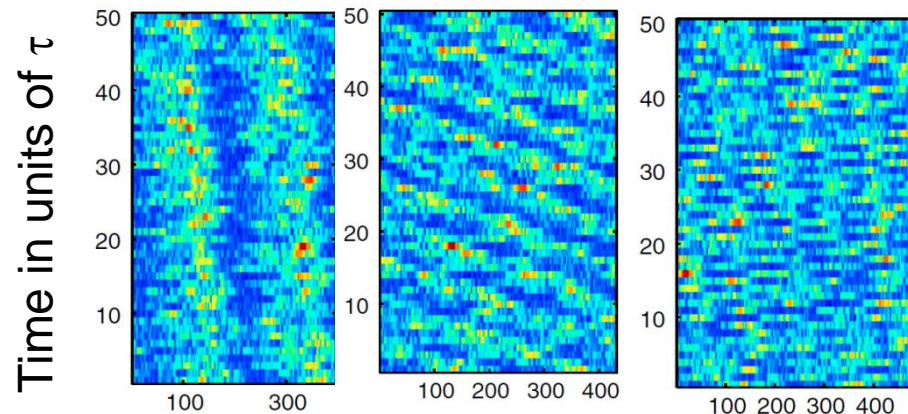


⇒ Sharp variations at the transition not captured.

Space-time representation

$$\begin{bmatrix} \dots & \dots & \dots & \dots \\ I_{2\tau+1} & I_{2\tau+2} & \dots & I_{3\tau} \\ I_{\tau+1} & I_{\tau+2} & \dots & I_{2\tau} \\ I_1 & I_2 & \dots & I_{\tau} \end{bmatrix}$$

Color: I_i



⇒ Different structures uncovered with different lags (sampling times).

Conclusions



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- Take home message:
 - The symbolic network and the horizontal visibility graph are useful tools for characterizing transitions.
- Main conclusions
 - Early-warning signs of upcoming PS validated.
 - Laminar-turbulent transition: sharp transition seen in thresholded data but not in raw data; particular time-scales identified at the transition.
- Future work:
 - Quantify the performance of the diagnostic tools.

At UPC

- Andres Aragoneses
- Laura Carpi
- Toni Pons
- Mari Carme Torrent

At URV (Tarragona)

- Sergio Gomez
- Alex Arenas

Experimental data:

- PS data from INLN (S. Barland) and Bangor University (Y. Hong)
- Fiber laser data from Aston University (S. Turitsyn)



THANK YOU FOR YOUR ATTENTION !

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Papers at <http://www.fisica.edu.uy/~cris/>

- C. Masoller et al, “*Quantifying sudden changes in dynamical systems using symbolic networks*”, New J. Phys. 17, 023068 (2015).
- A. Aragoneses et al, “*Unveiling temporal correlations characteristic of a phase transition in the output intensity of a fiber laser*”, PRL 116, 033902 (2016).

