

Some vision now in IA
So what is the best classifier to know
if it will rain tomorrow at Calais ?

Implicit Times series segmentation by clustering

Emilie Poisson Caillault.

LISIC, Laboratoire Informatique Signal Image Côte d'Opale

Univ. Littoral Côte d'Opale

RAINSMORE, IA for water

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No, so it will be sunshine
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2001 : Polytech'Nantes engineer in computer sciences

2001 : DEA Automation and Computer Sciences, Ecole Centrales Nantes

2005 : PhD, Univ. Nantes at LC2N (IRCCYN Lab).

Architecture and Training of a hybrid Neuro-Markovian System for On-Line Handwriting Recognition

Keywords : TDNN, SDNN, SD-TDNN, MS-TDNN, global discriminant training, MLE-MMI, Mask/Filter in convolution layer.

2006 : Assistant Professor - Univ Littoral in data science and machine learning

2014 : IFREMER delegation

2020 : HDR degree

Contributions to the classification and segmentation of Time series by statistical unsupervised or guided learning

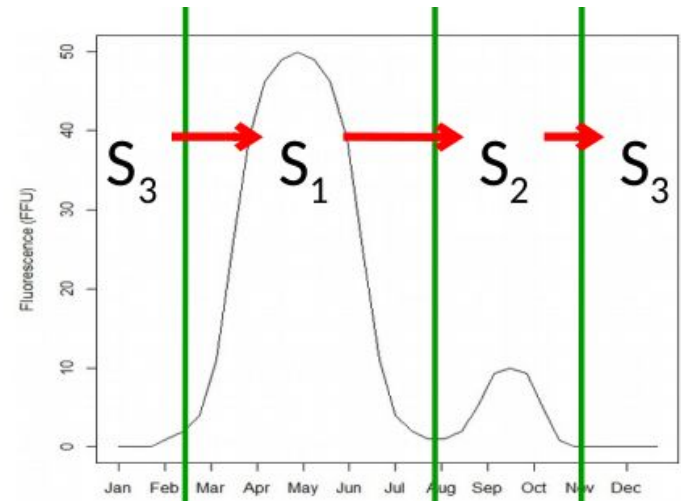
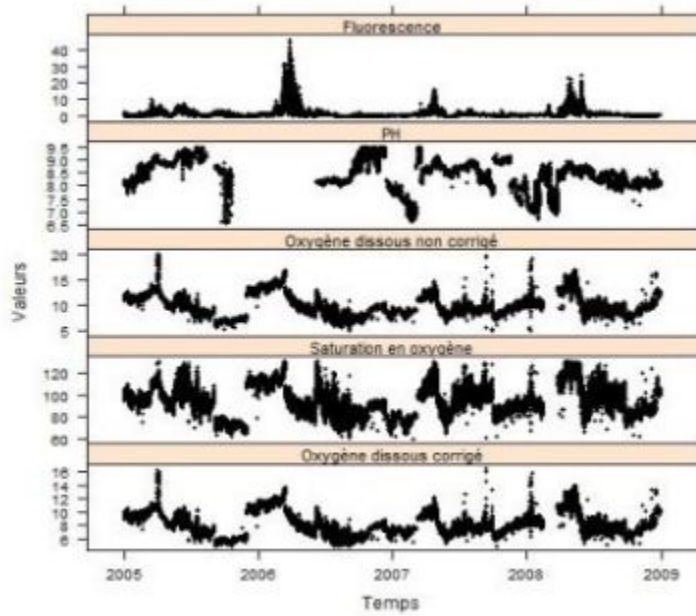
Keywords : similarity, DTW-criteria, DTW-imputation, spectral clustering and multi level approach

JERICO project
CPER IDEAL
ORIENTOI application
LISIC/IFREMER PhD supervision

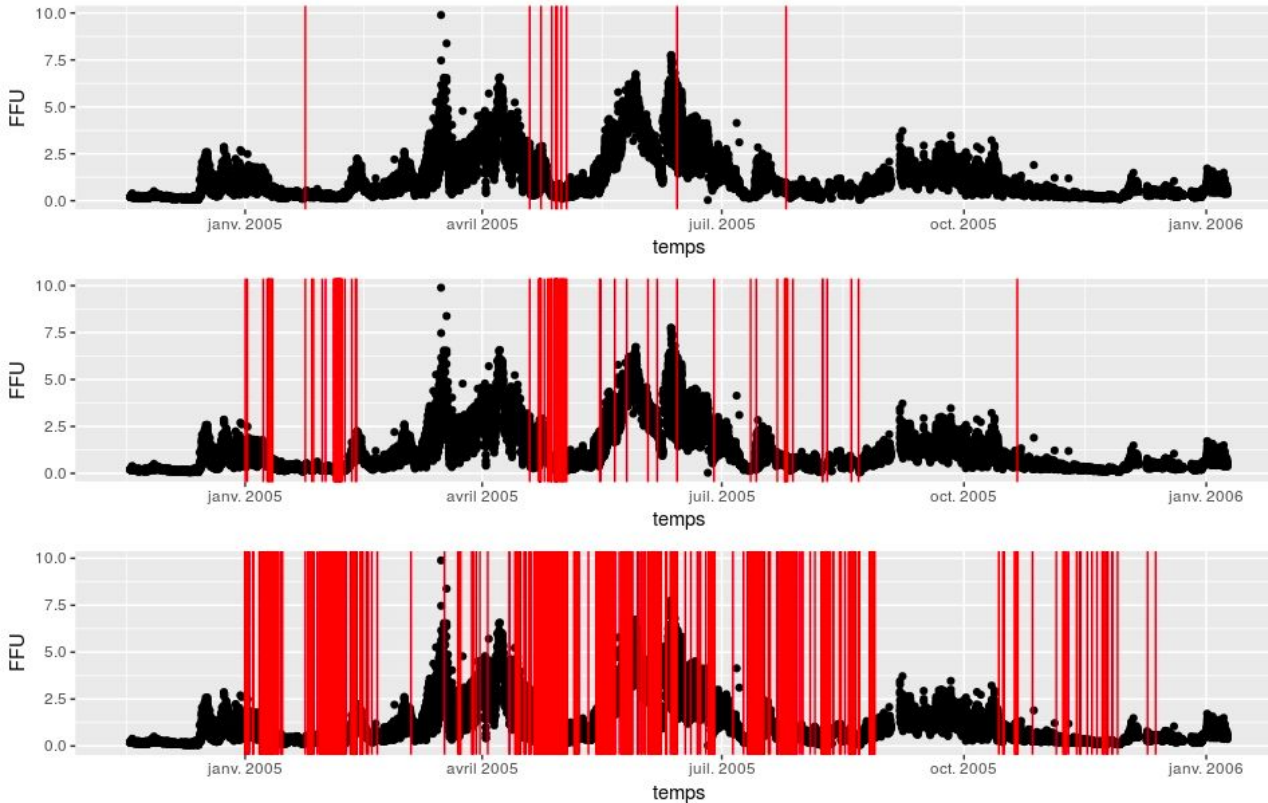
- a. Pattern clustering and classification
- b. time series
- c. convolutional neural networks
- d. hidden markov models
- e. fully unsupervised or constrained spectral clustering
- f. elastic distance metrics for signal comparison
- g. environmental science computing

Time Series or Spatial segmentation by clustering

Environmental state ?



Event or region detection



Approaches

Univariate :

- Breakpoints, PIP, trend
- Explicit segmentation
- Implicit segmentation

Multivariate :

- Explicit segmentation
- Scattering moments
- Implicit segmentation

cut process:

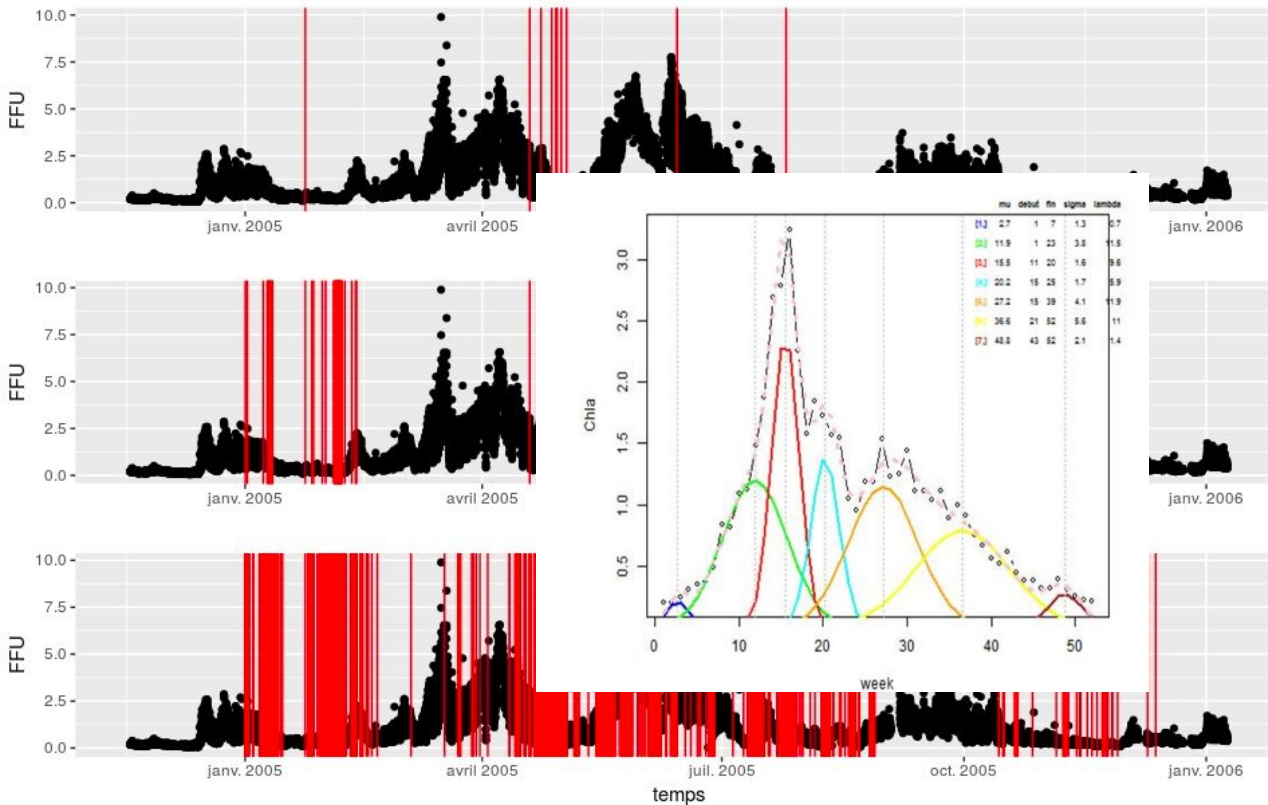
- Suitable for trend analysis
- Imposes clustering/matching before labelling



.... Tedious for the expert

.... Costly in terms of calculation

Event or region detection



Approaches

Univariate :

- Breakpoints, PIP, trend
- **Explicit segmentation**
- Implicit segmentation

Multivariate :

- **Explicit segmentation**
- **Scattering moments**
- Implicit segmentation

Article : Towards Chl-a Bloom Understanding by EM-based Unsupervised Event Detection. Emilie Poisson CAILLAULT and Alain LEFEBVRE. Full accepted paper. OCEANS 2017 MTS/IEEE, Aberdeen, Scotland, 06/2017

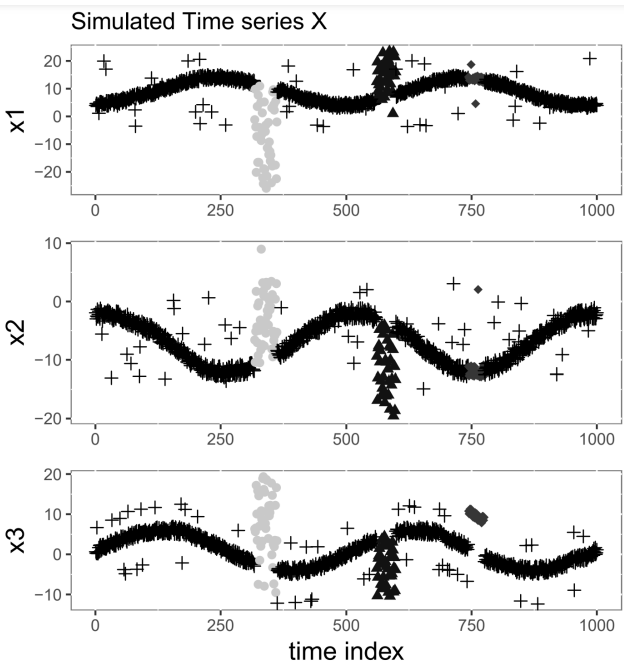
Detection of mixture of patterns Requires a priori

- Forms of event (gaussian?)
- Series statistics

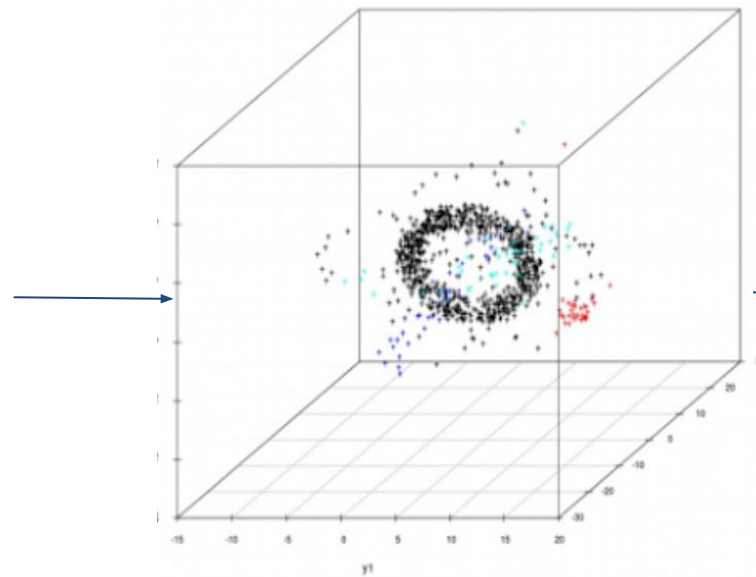


Implicit segmentation by clustering approach

- 1- Compute similarities between Observation features $\rightarrow W$
- 2 - Apply Partitioning algorithm in this Observation space
- 3 - Analyse obtained dynamics and sometimes correct it.



(x_1, x_2, x_3, t)



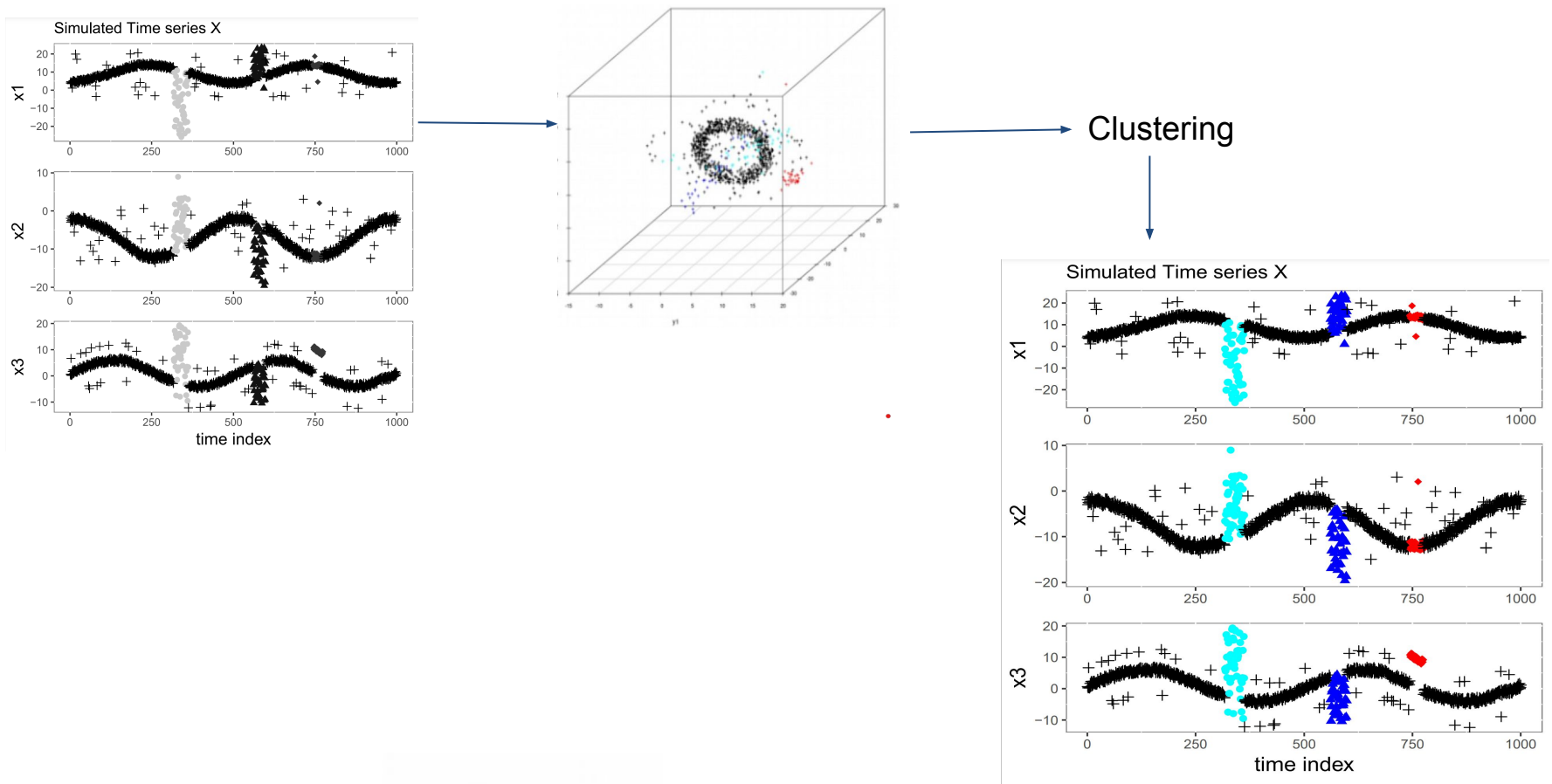
(x_1, x_2, x_3)

Clustering

$(x_1, x_2, x_3) + \text{label}$

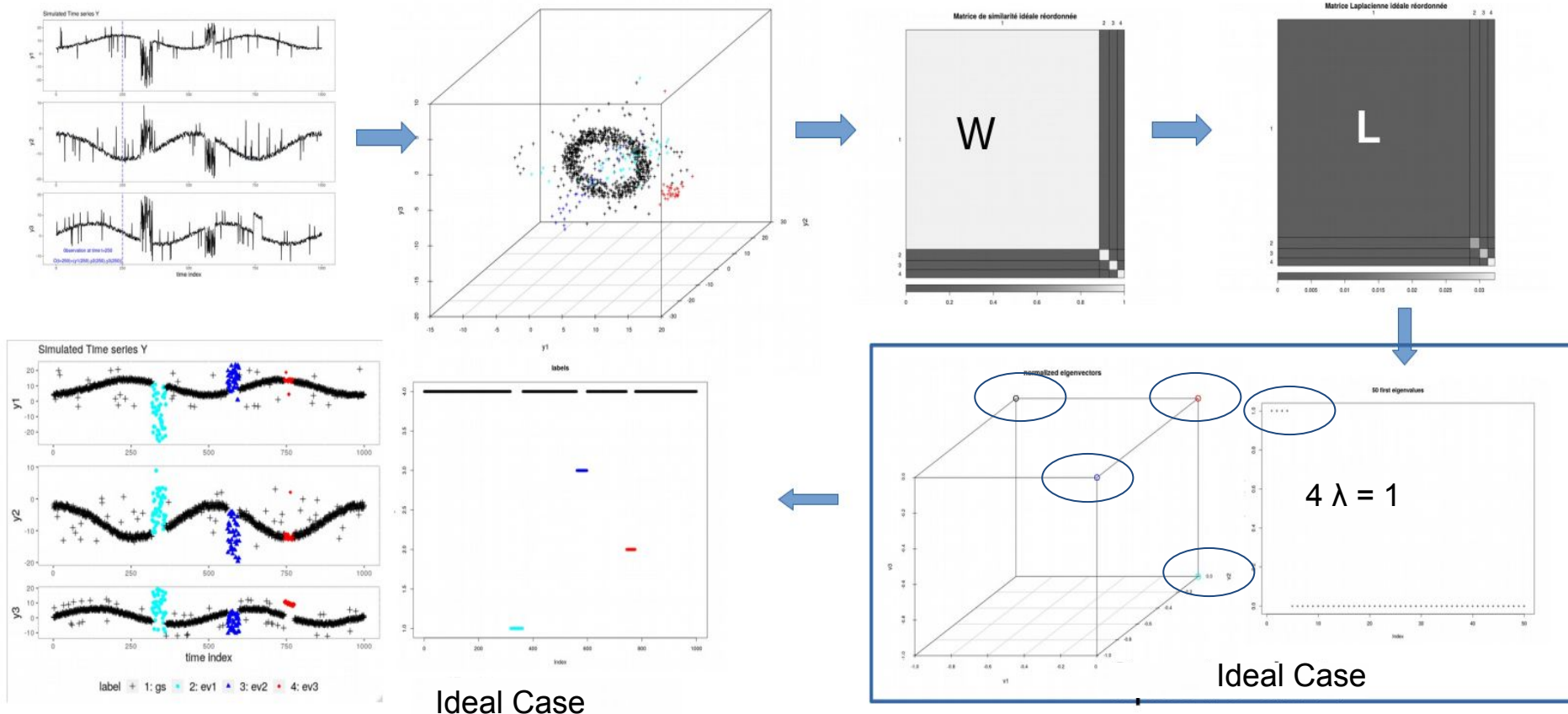
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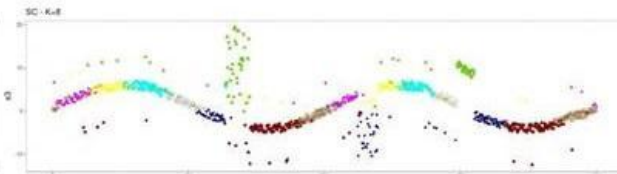
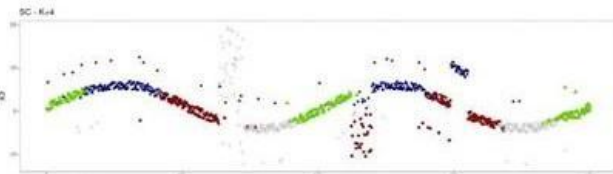
Implicit segmentation by **spectral clustering** approach

- 1- Compute similarities between Observation features -> W
- 2- Compute Laplacian matrix from W
- 3- Extract eigenvectors V and eigenvalues -> detect K principal values
- 4 - Partitioning data in the normed K -first vector eigenspace U (PAM)

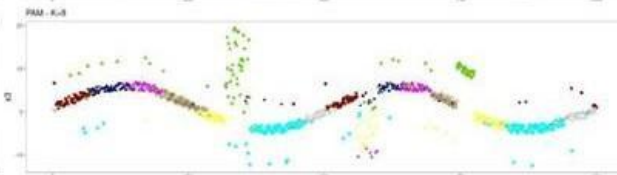
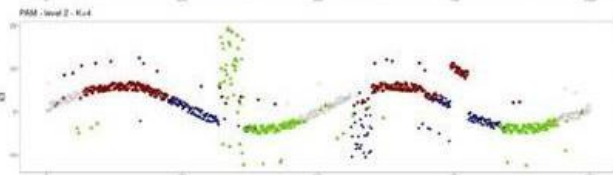




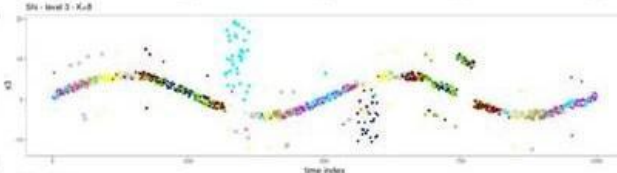
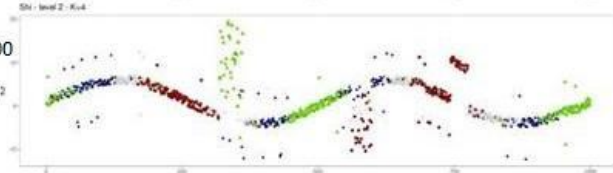
NJW-SC
Ng et al, 2001



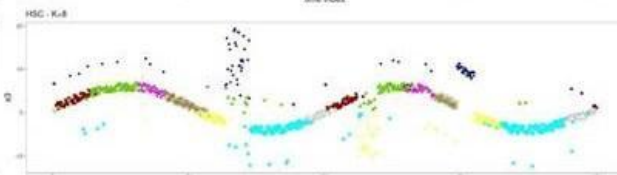
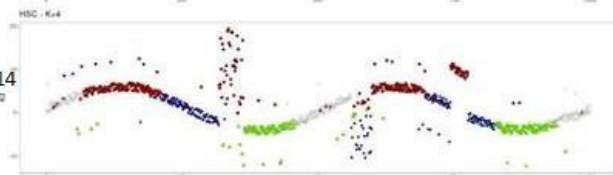
PAM-SC



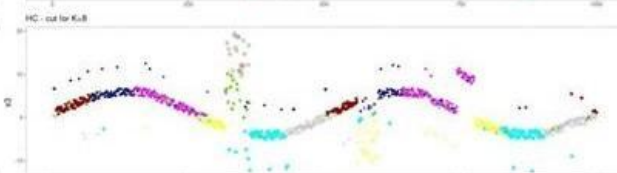
Bi-SC
Shi et Malik, 2000



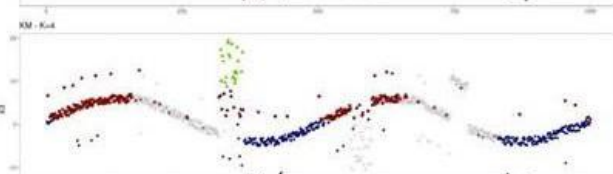
H-SC
S-Garcia et al. 2014



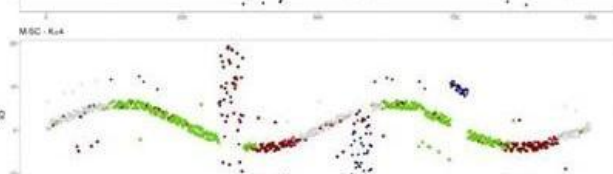
HC



K-means



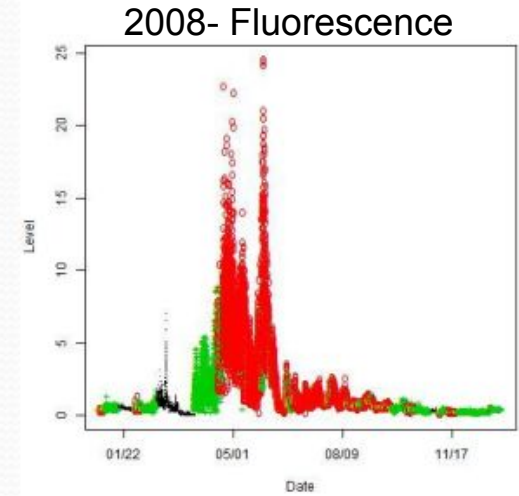
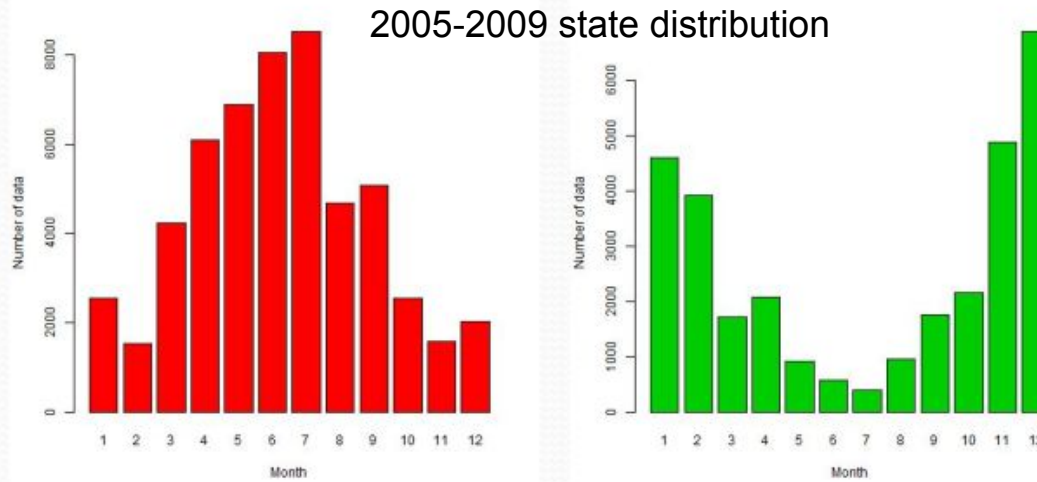
M-SC



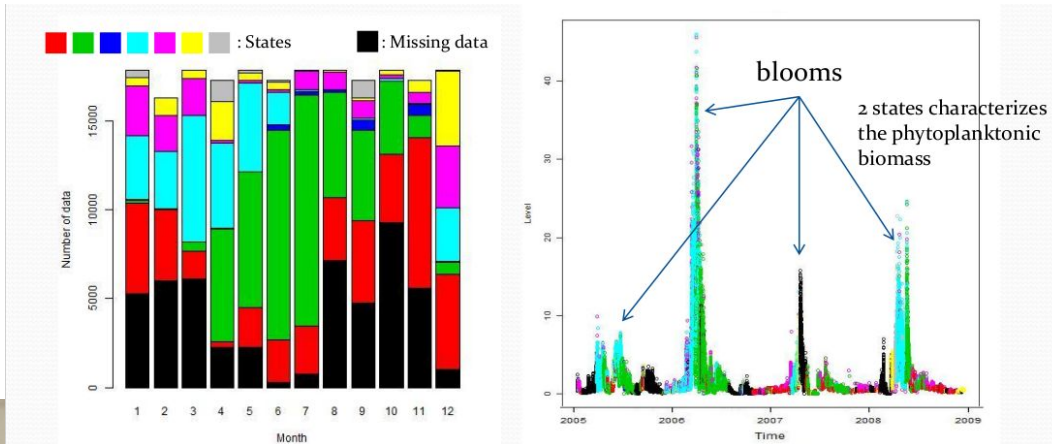
MSC
Kelly's talk

Long-term series : Marel Carnot application. (K. Rousseuw PhD)

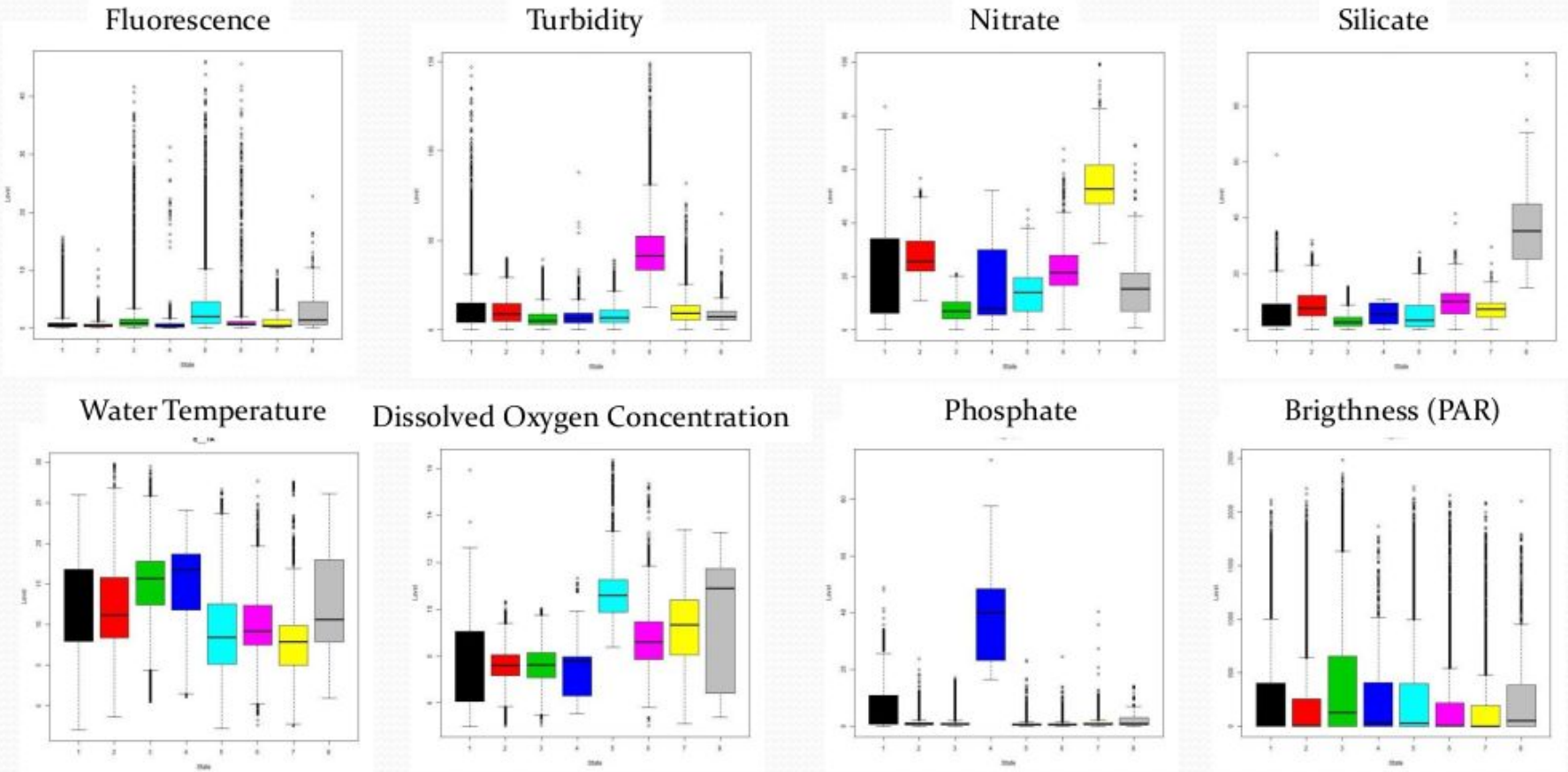
SC with K=2: Identification of non productive period vs productive period.



SC with K=7: Identification of blooms, pre/post-blooms, rare events

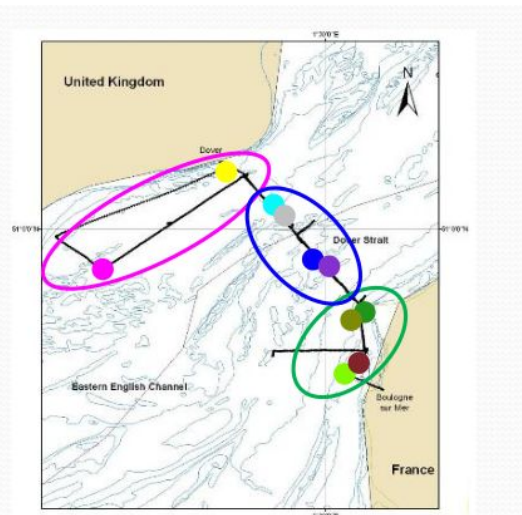
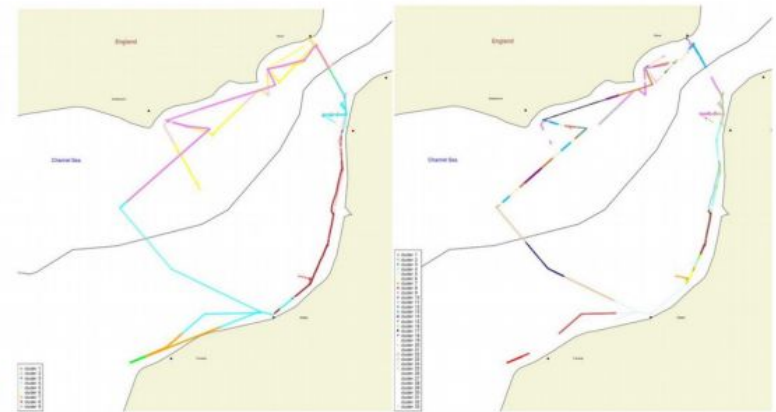
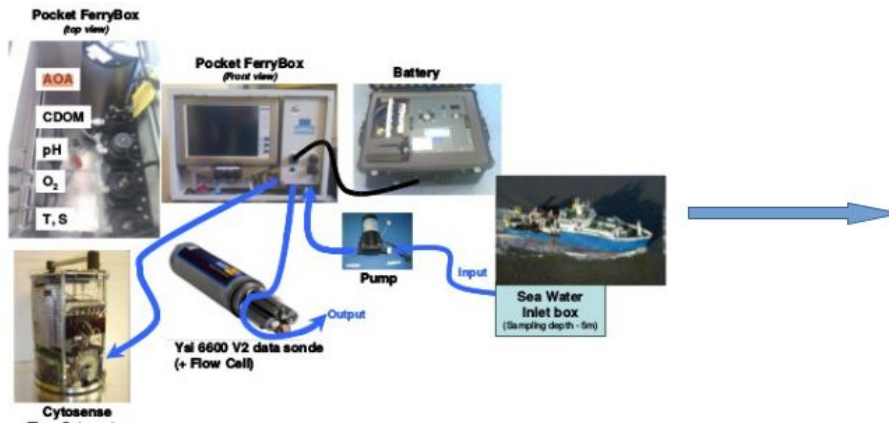


Long-term series : Marel Carnot application. (K. Rousseeuw PhD)



interface in R-package : uHMM and sClust

Short-term **spatio time** series : DYMAPHY Leg (MEPS'2019)



LEG 1 of the DYMAPHY campaign
(20 - 21th of April 2012)

Eco-regions
Coastal zone vs large zone