

CR09 - HMM for time series classification and filtering

Introduction

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Let me introduce myself!

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PhD in image processing
in Brest



Master in telecommunication
in Lille

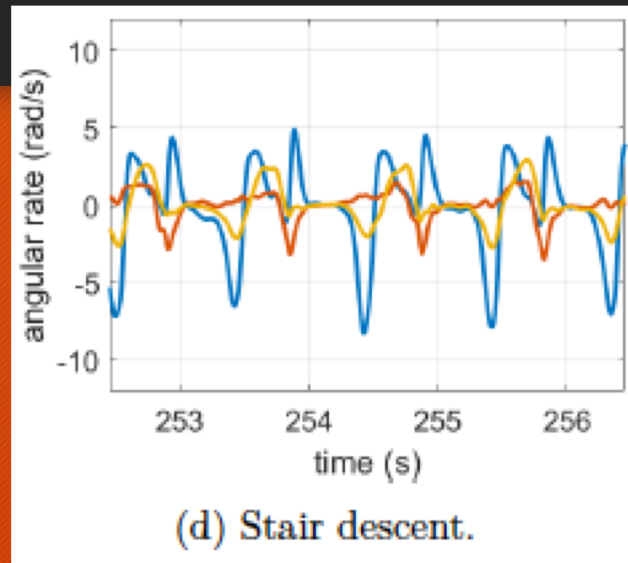
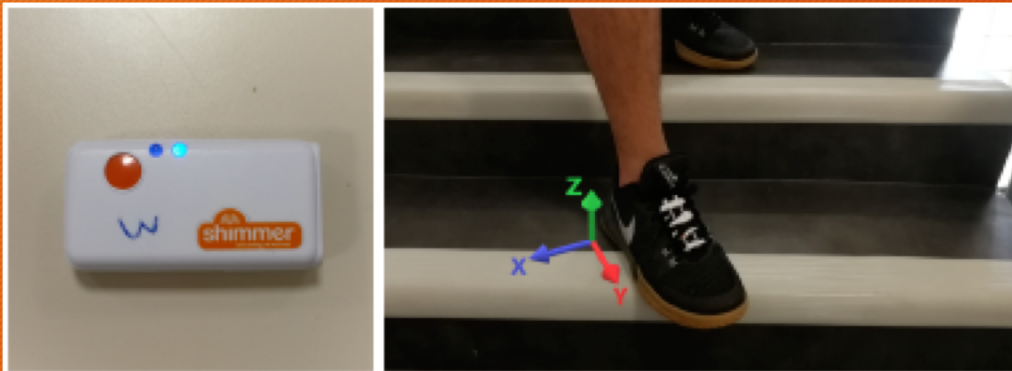
Full professor
in Computer Sciences
In Lyon

Assistant professor
in Computer Sciences
in Marseilles

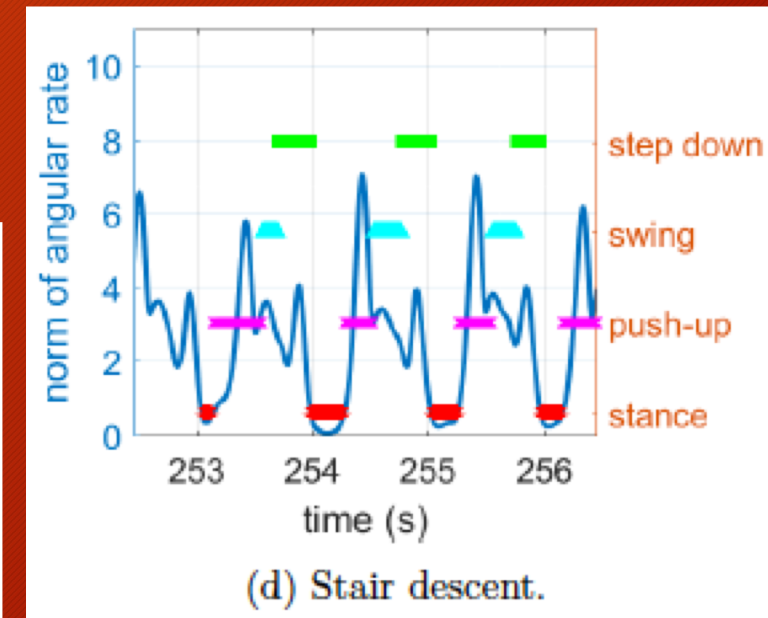
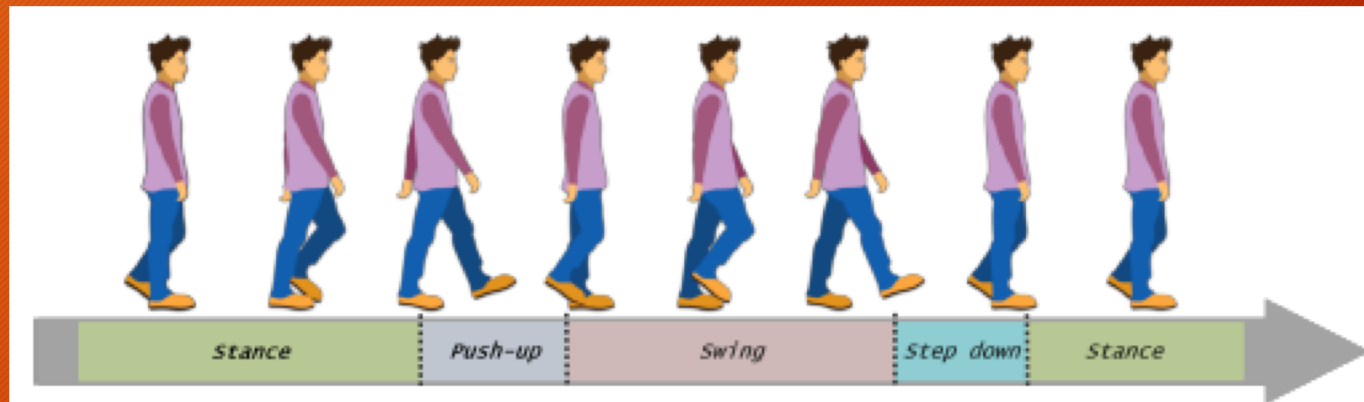
Application : Human activity monitoring

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Inertial Measurement Unit



Gait cycle

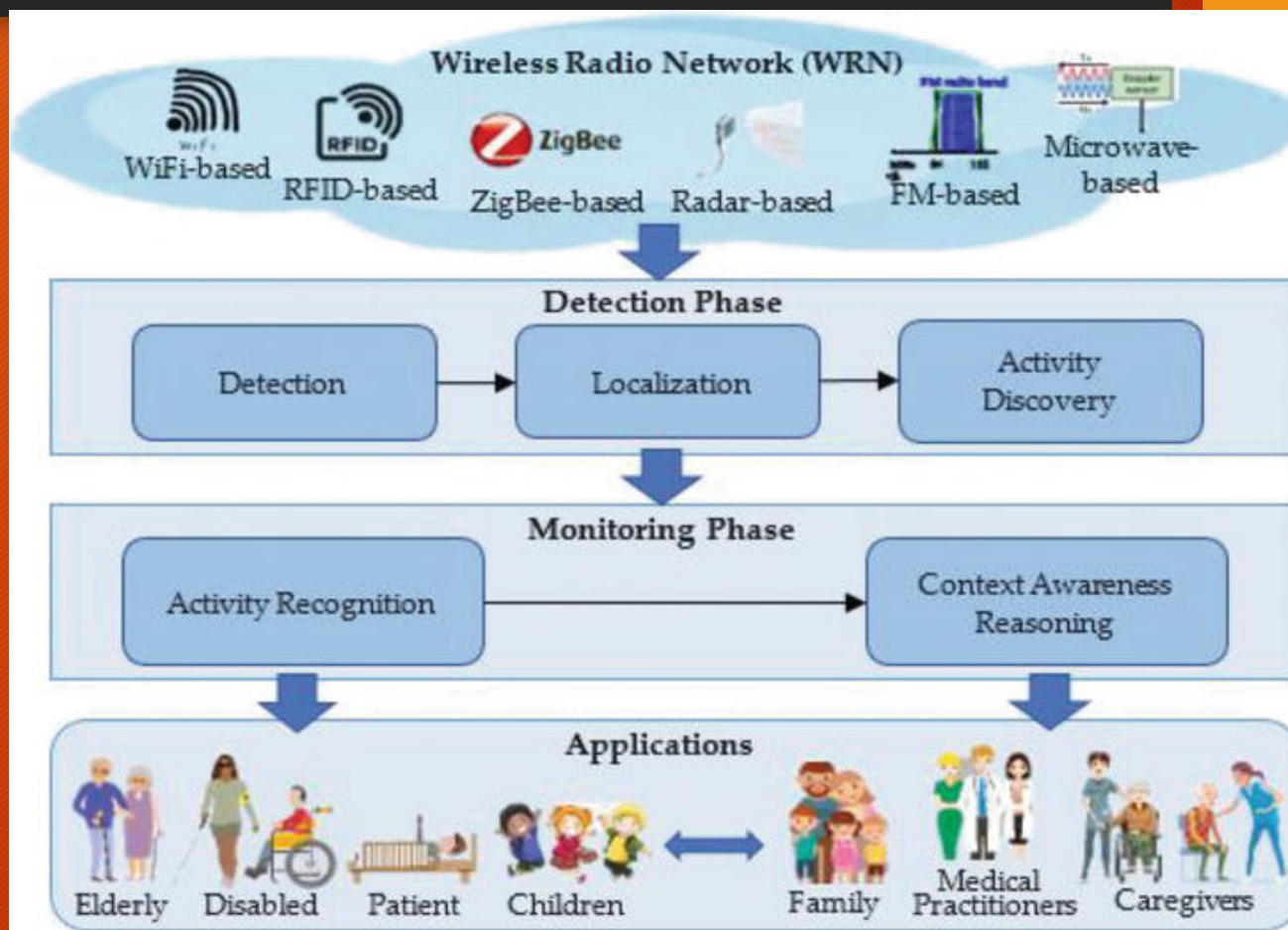


Application : Human activity monitoring using IoT network

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Home automation / domotics.

Example: Development of surveillance system for the elderly using wireless beacon signal analysis, such as WIFI, Bluetooth, ZigBee sensors...

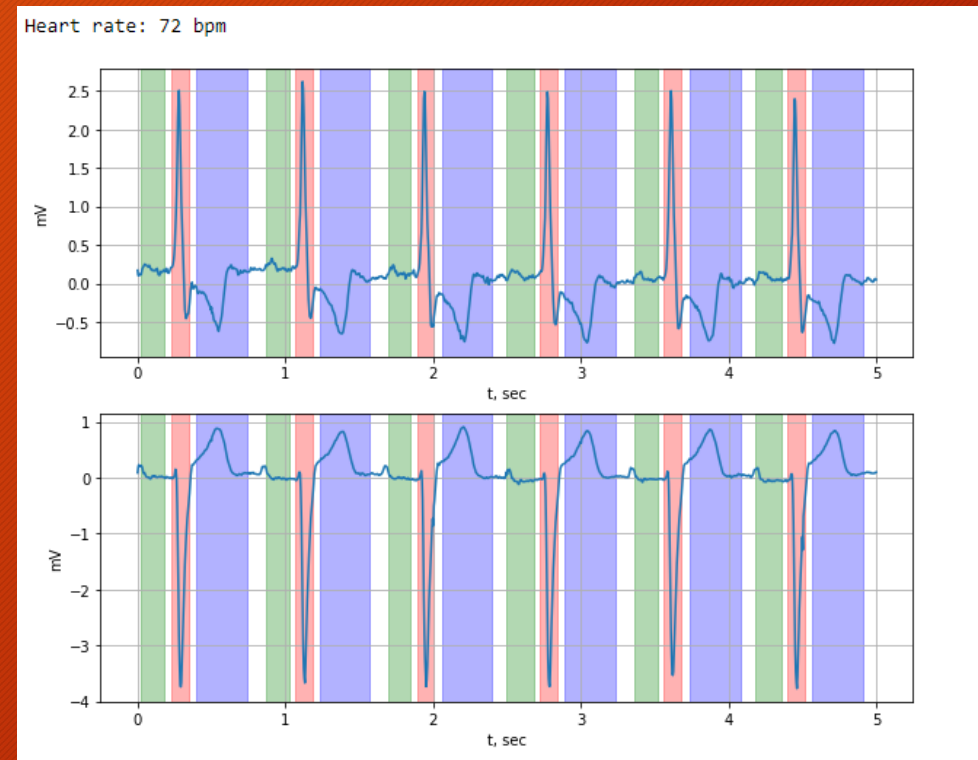
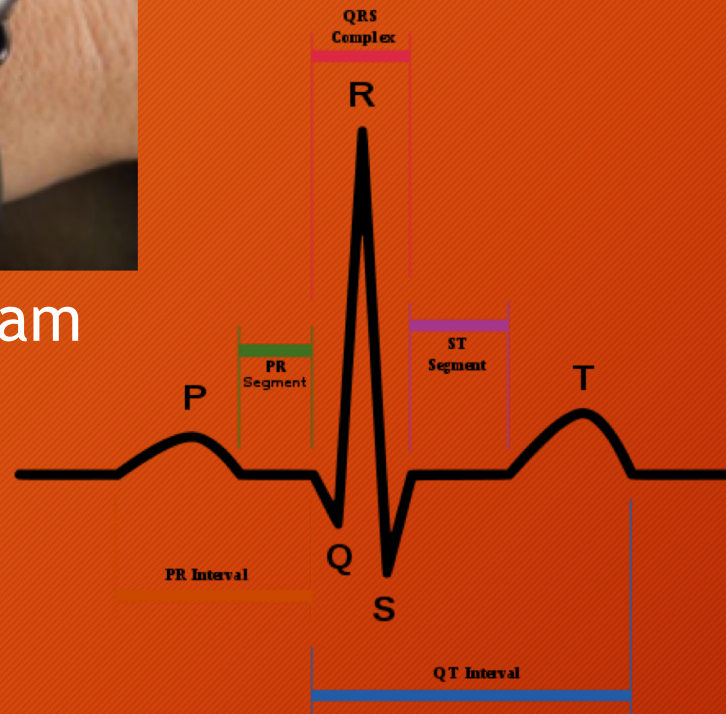


Markov models everywhere!

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Electrocardiogram



ECG signal annotation

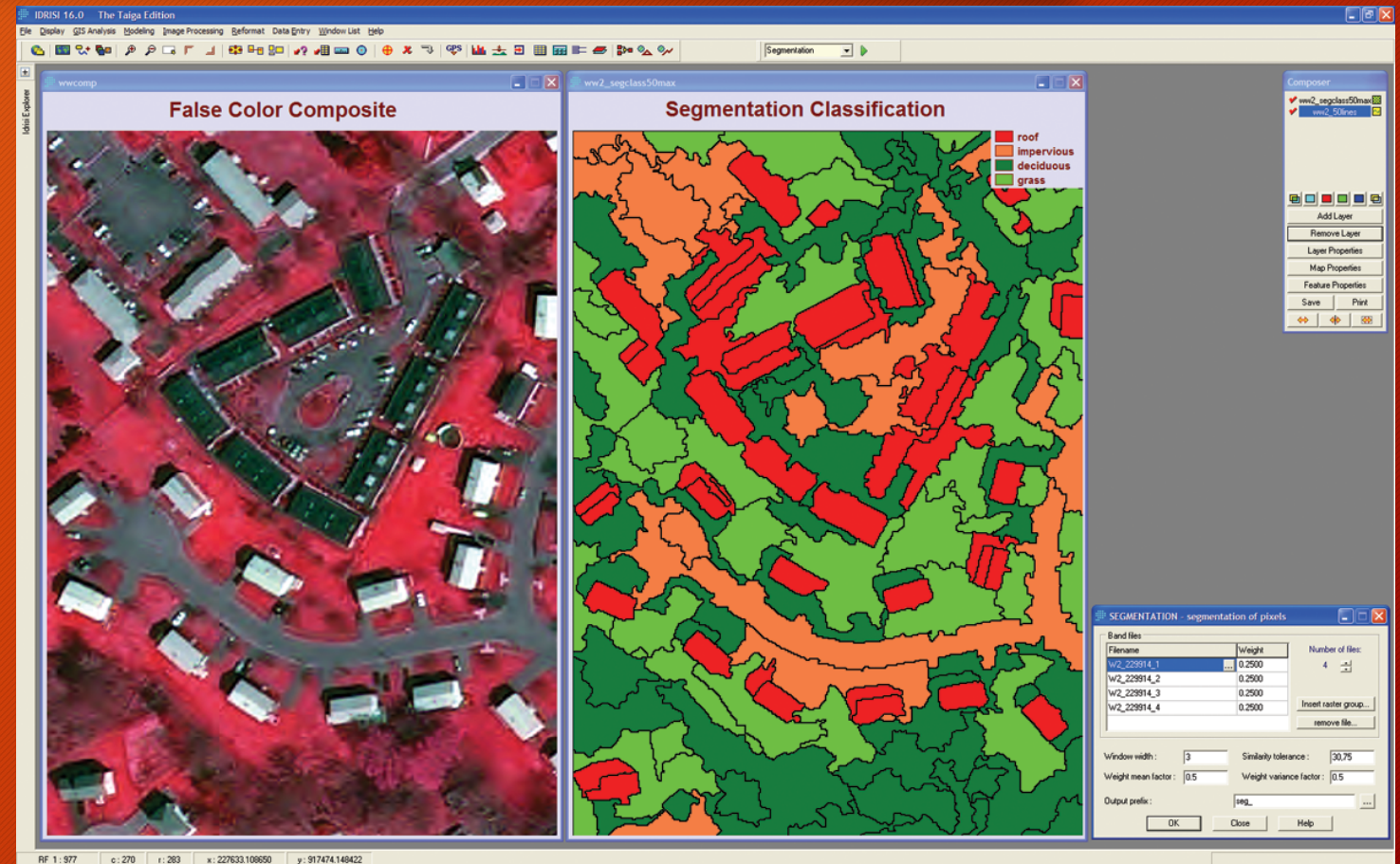
Markov models everywhere!

Hidden Markov Random Field Model

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Satellite image
segmentation

Aerial photography for
Woburn, Massachusetts in
2005.



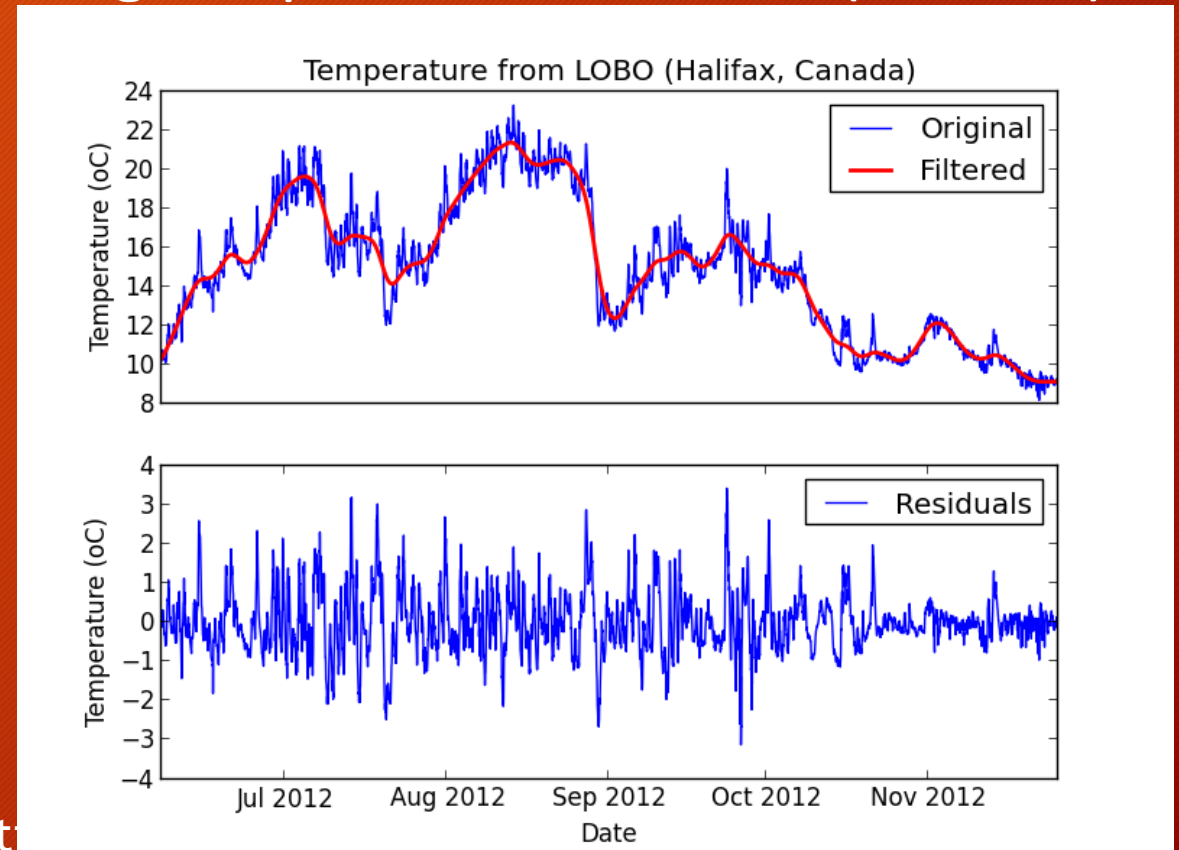
Markov models everywhere!

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Kalman-like filtering

LOBO instrument (Land/Ocean Biogeochemical Observator)

Temperature data measured in the southwest edge of peninsular Halifax (Canada)



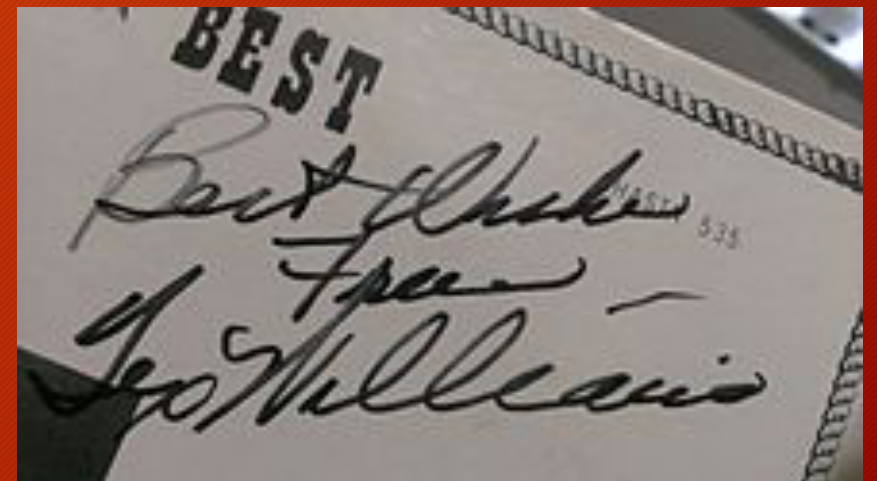
Markov chain models everywhere!

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Also (wikipedia):

- Finance and econometry (stock exchange),
- Cryptanalysis (used to breach cryptographic security systems),
- Speech coding and synthesis, handwriting recognition,
- Biology : Gene prediction, bio-sequence alignment, DNA motif recovery...

Handwritten Text Recognition (HTR), is the ability of a computer to interpret intelligible handwritten input from sources such as paper documents, photographs, touch-screens and other devices.



Our stochastic point of view on TS analysis

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- What we observe through an instrument or a sensor is degraded by many noises: Sensor noise (electronic), Transmission noise (coding error)...
- The noise is stochastic: it cannot be modelled by a deterministic equation. But its behaviour can be characterized statistically, through a probability law.
- Consequently, if the same experiment is repeated twice, the resulting time series will be different, ie samples will not be exactly the same.

Notations : the observations

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- One observation emitted by a sensor (or whatever) will be denoted by the lower-case letter y .

- A series of observations of length N will be denoted by

$$\mathbf{y} = \mathbf{y}_1^N = \{y_1, y_2, \dots, y_n, \dots, y_N\}$$

- The series is modeled by a stochastic process with as many random variables as there are samples:

$$\mathbf{Y} = \mathbf{Y}_1^N = \{Y_1, Y_2, \dots, Y_n, \dots, Y_N\}$$

Each random variable Y_n is assumed to be real-valued (for the requirement of our course only) and characterized by a probability density function (pdf, e.g. Gaussian).

Notations : the states (or classes)

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- The series of states / labels will be modeled by a stochastic process with as many random variables as there are samples:

$$\mathbf{X} = \mathbf{X}_1^N = \{X_1, X_2, \dots, X_n, \dots, X_N\}$$

Each random variable X_n is assumed to be

- Discrete-valued for data classification $X_n \in \Omega = \{1, \dots, K\}$
 - Real-valued for data filtering
- A realization of the stochastic process will be denoted by

$$\mathbf{x} = \mathbf{x}_1^N = \{x_1, x_2, \dots, x_n, \dots, x_N\}$$

Preparation to upcoming lab and homework

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- Install Python 3.x through the Anaconda distribution:
<https://www.anaconda.com/distribution/>
- The distribution includes a text editor to write programs called « Spyder ».
- Beginners can follow Spyder tutorial from:
https://www.youtube.com/watch?v=a1P_9fGrfnU



CR09 Lesson agenda

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- **Session 1 (8h): Bayesian Decision (BD) theory and Mixture Model (MM)**
 - One 2h lab session with Python
 - Grading: a 1h sitting exam, no document (weight: 25%)
- **Session 2 (14h): Hidden Markov chain (and beyond)**
 - Two 2h lab session with Python
 - A 1h conference about a self-quantified application of HMM
 - Grading: a 2h sitting-exam, no document (weight : 50%)
- **Session 3 (8h): from Kalman filter to particle filter**
 - Grading : A 2h lab session with Python (weight : 25%)

Slides (completed and updated periodically), exercise and lab statements collected at: <http://perso.ec-lyon.fr/derrode.stephane/Teaching.php>

- Pattern recognition and Machine Learning, Christopher M. Bishop, Springer
- Bayesian smoothing and filtering, Simo Sarkka, Cambridge University Press, https://users.aalto.fi/~ssarkka/pub/cup_book_online_20131111.pdf

Keywords : Bayesian decision theory, mixture model, HMM, Kalman filtering, particle filtering