

Advanced Control Systems Detection, Estimation, and Filtering

***Graduate Course on the
MEng PhD Program
Spring 2012/2013***

Chapter 1 Motivation

Instructor:

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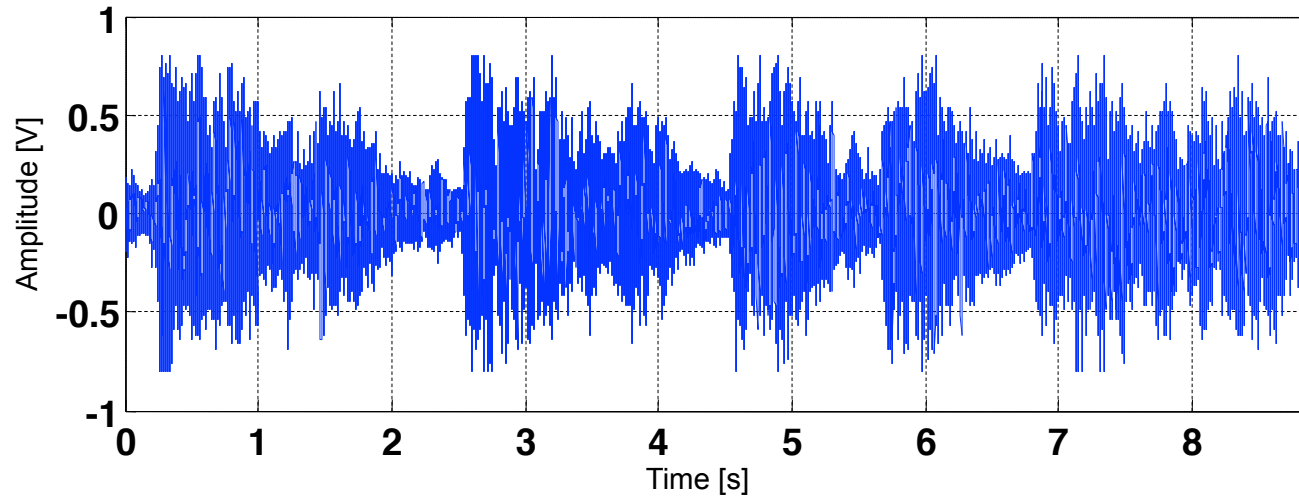
Summary:

- Motivation for estimation, detection, filtering, and identification in stochastic signal processing
- Methodologies on how to design optimal estimation algorithms
- Characterization of estimators and tools to study their performance
- To provide an overview in all principal estimation approaches and the rationale for choosing a particular technique

**Both for parameter and state estimation,
always on the presence of stochastic disturbances**

In RADAR (Radio Detection and Ranging), SONAR(sound navigation and ranging), speech, image, sensor networks, geo-physical sciences,...

Speech



Signals can be represented by functions (continuous time) or by vectors (where a sampling operation takes place)

Examples of speech/sound processing:

Automatic systems commanded by voice;

Automatic translation; Voice recognition

Synthesis of voice

Echograms

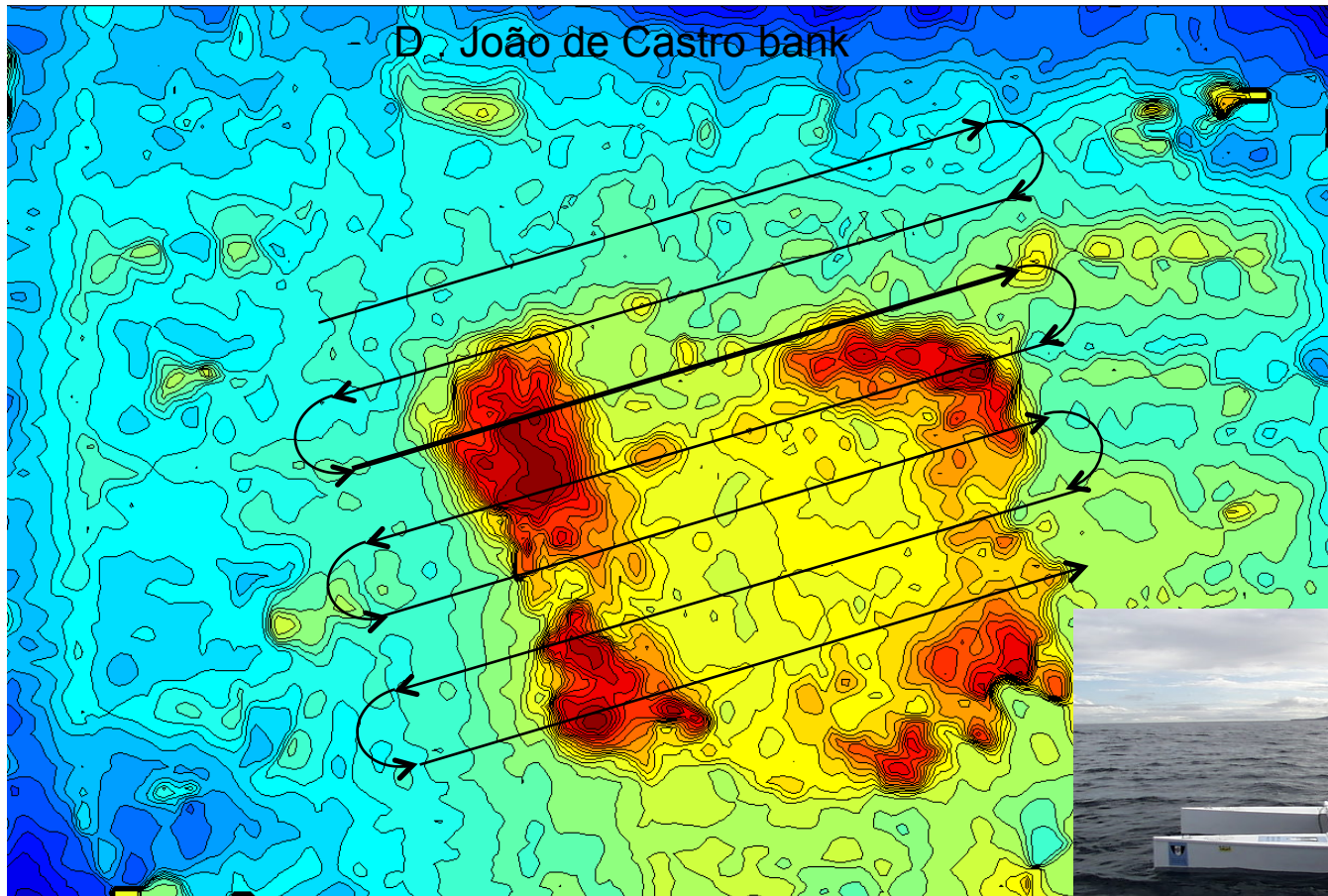
The quest for hydrothermal vents

D. João de Castro bank



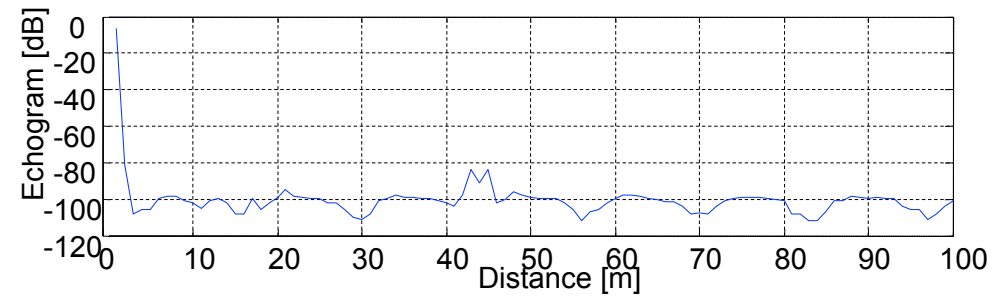
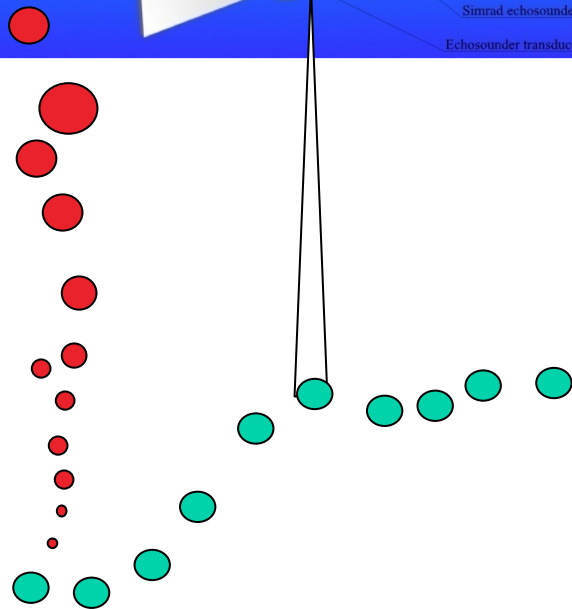
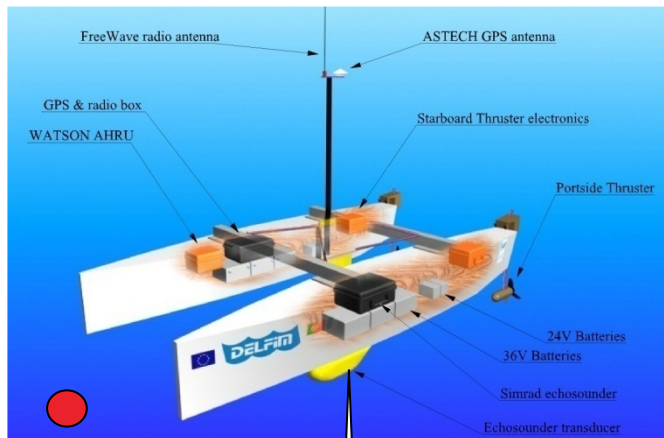
Echograms

The quest for hydrothermal vents (cont.)



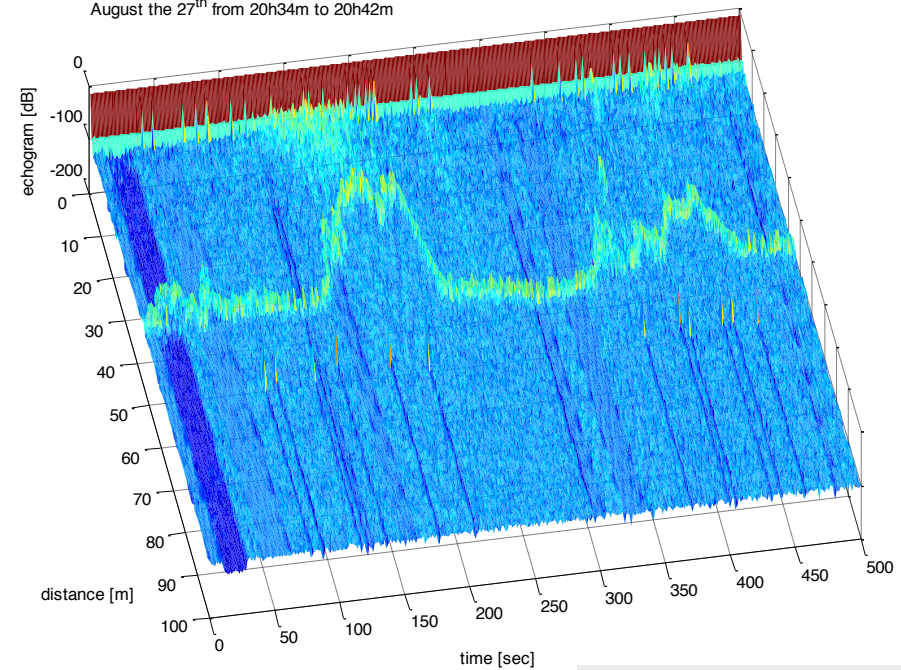
Echograms

The quest for hydrothermal vents (cont.)

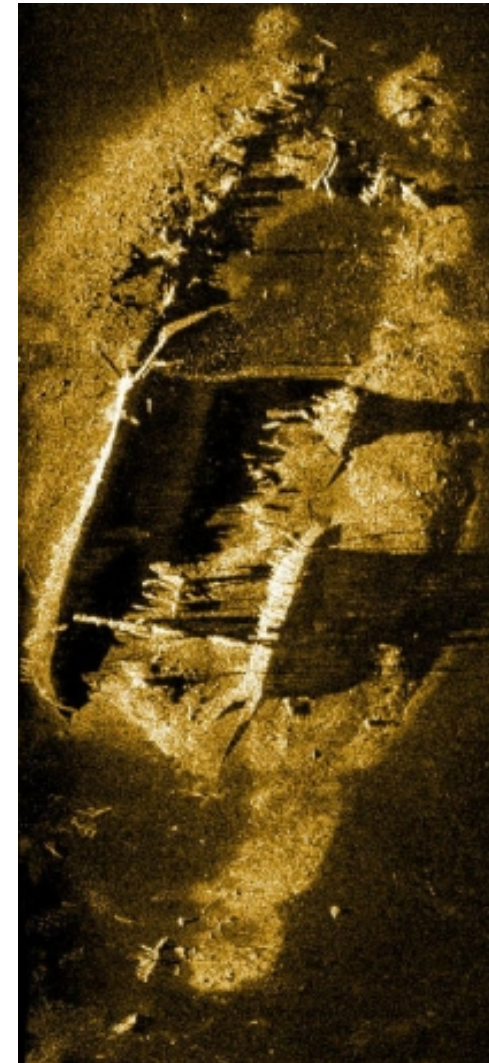
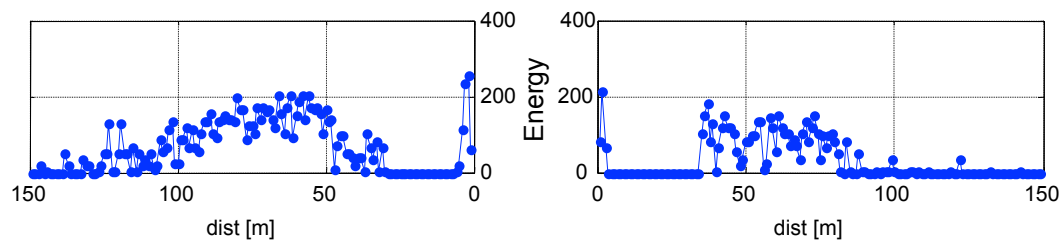
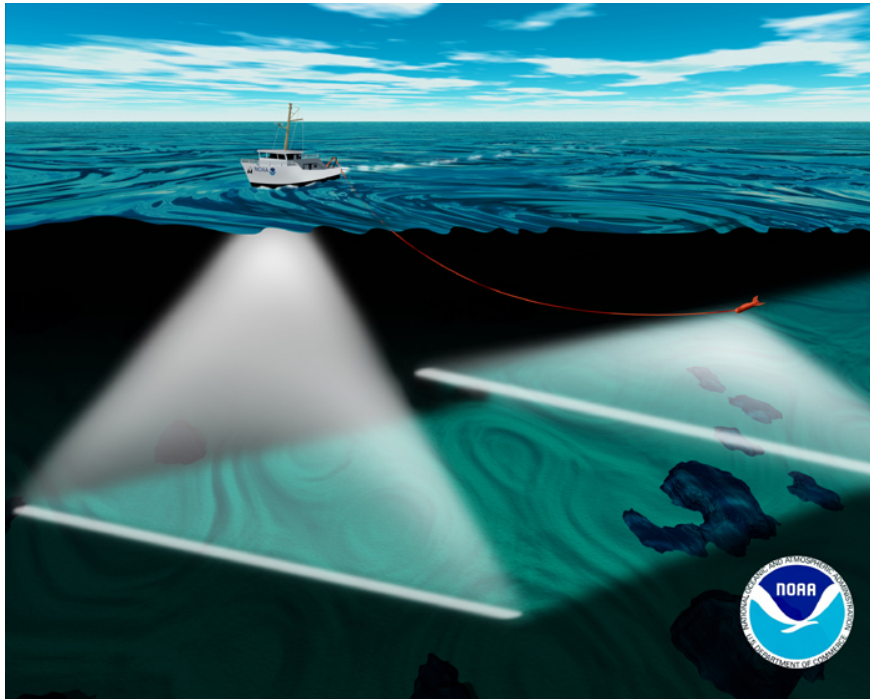


Delfim at D. João de Castro Bank

August the 27th from 20h34m to 20h42m



Sidescan Sonar Imaging



GPS – Global Positioning System

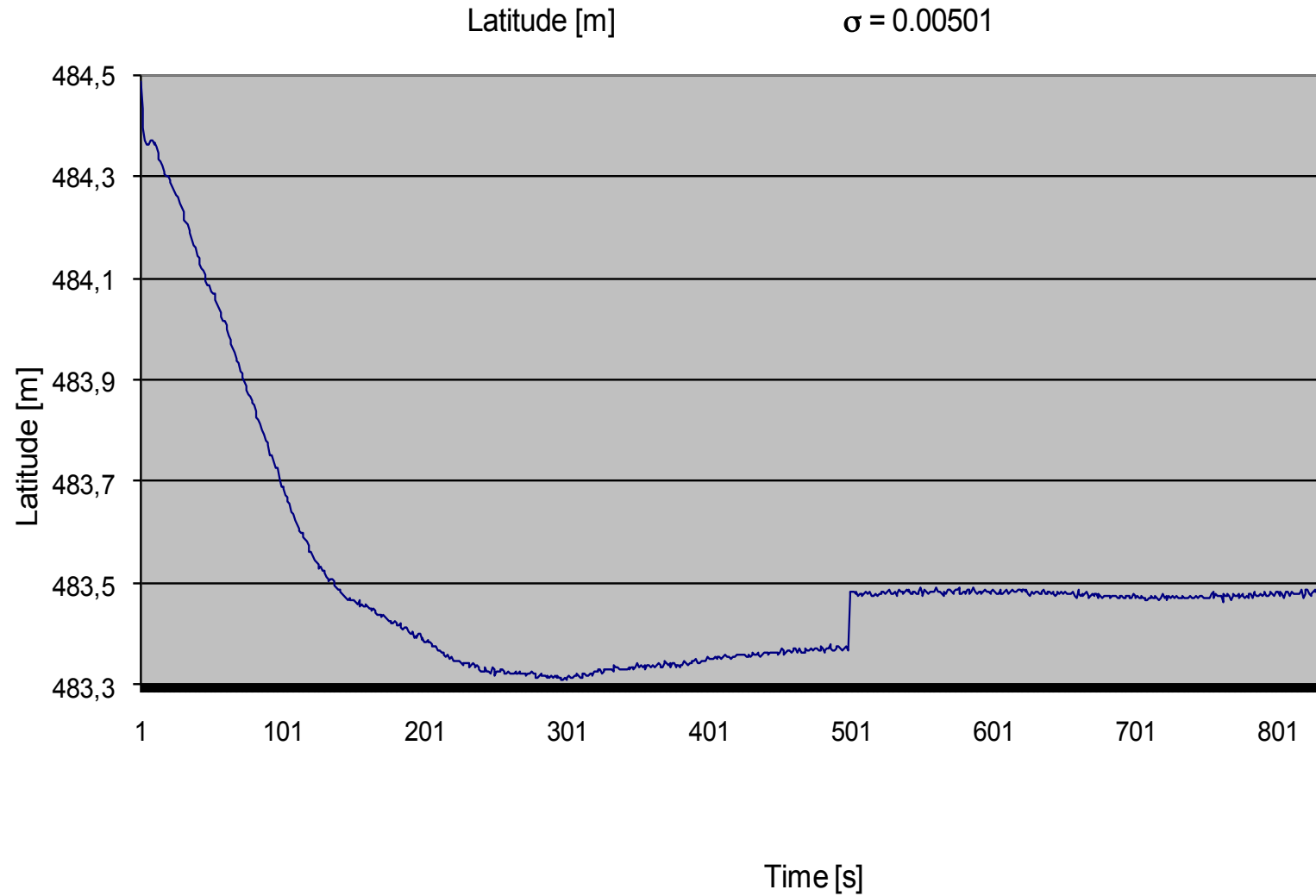
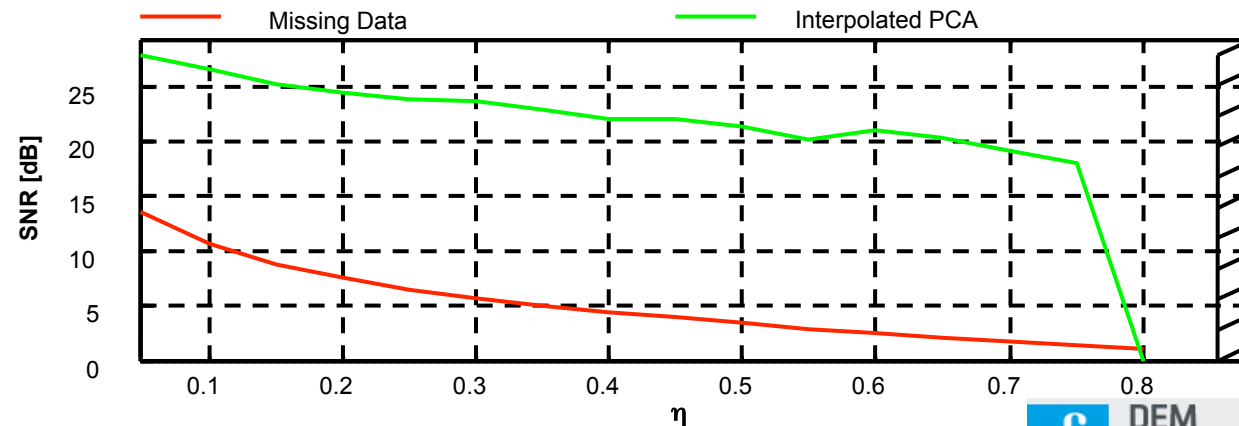
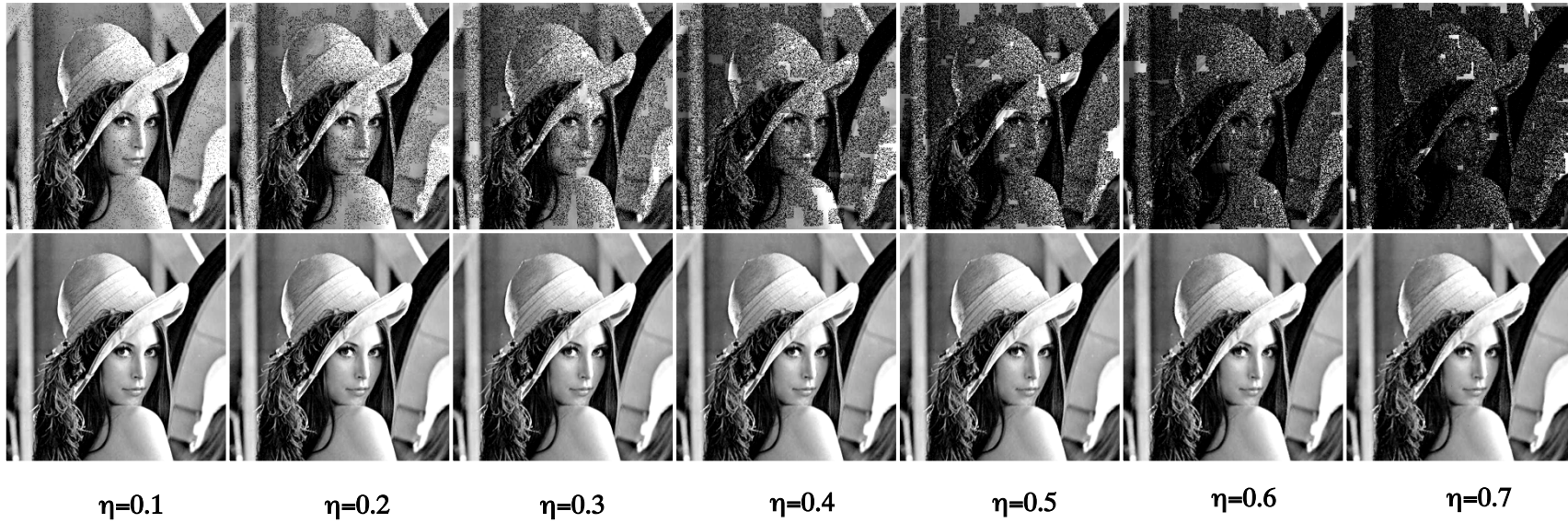
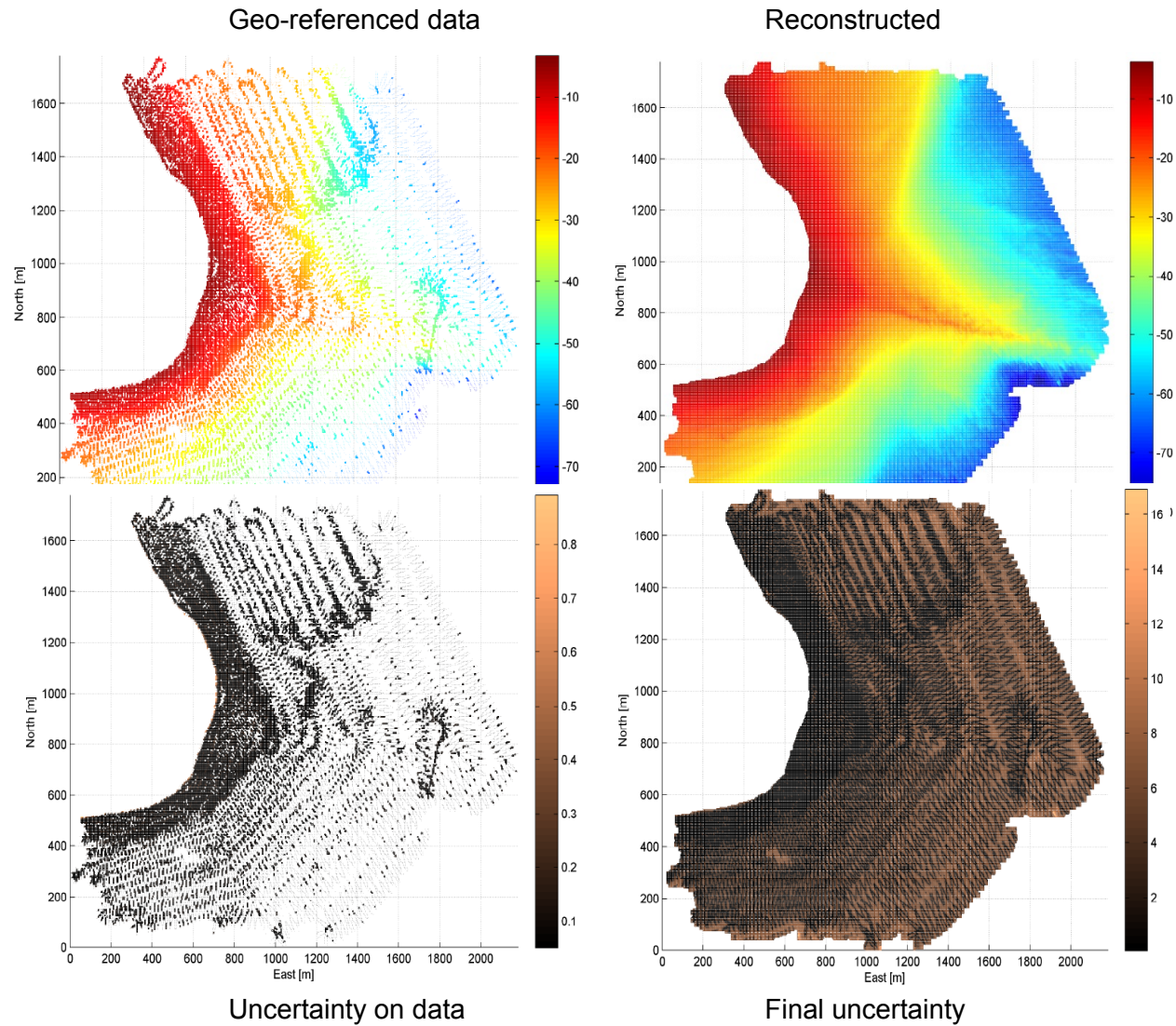


Image with missing data

Reconstruction



Bathymetric survey

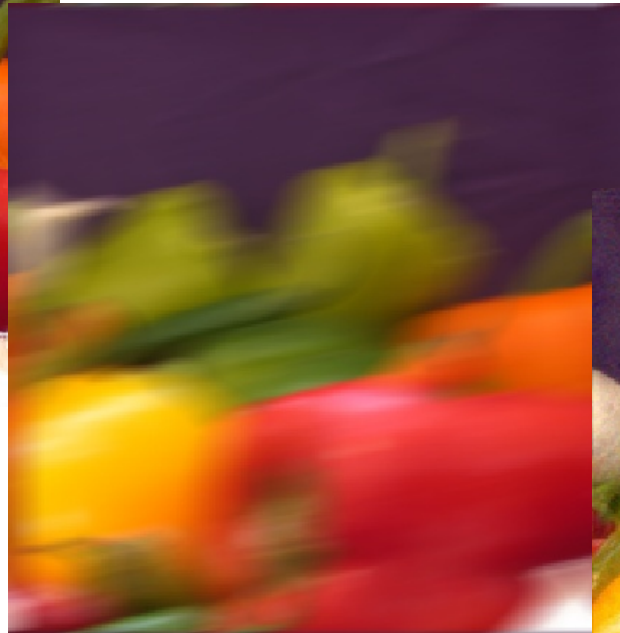


Deblurring an image

Original Image



Blurred Image



Restored image



Causes:

- Out of focus acquisition
- Camera-object movement
- Shaking
- Shallow field of view ...

Stock Exchange

Models that explain evolution of phenomena

- Causality
- Number of parameters
- Type of model
- Uncertainty

Is it possible to predict the market price tomorrow, next week, next month, next year,?...?

Courtesy Jornal de Negócios



GPS Intelligent Buoys (GIB)-ACSA/ORCA

Tracking with a Sensor Network

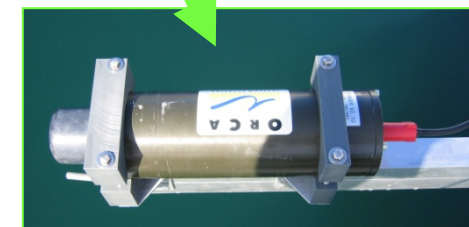
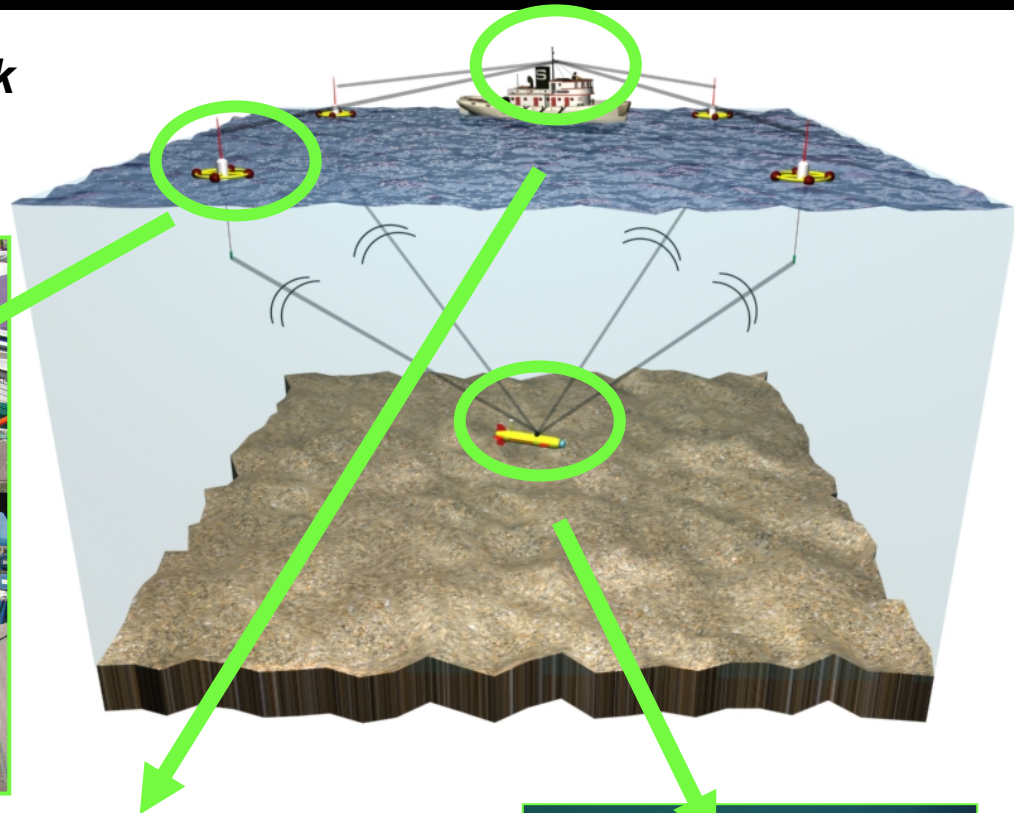


Surface buoys with

- DGPS receivers
- Hydrophones
- Radio link

Control Station

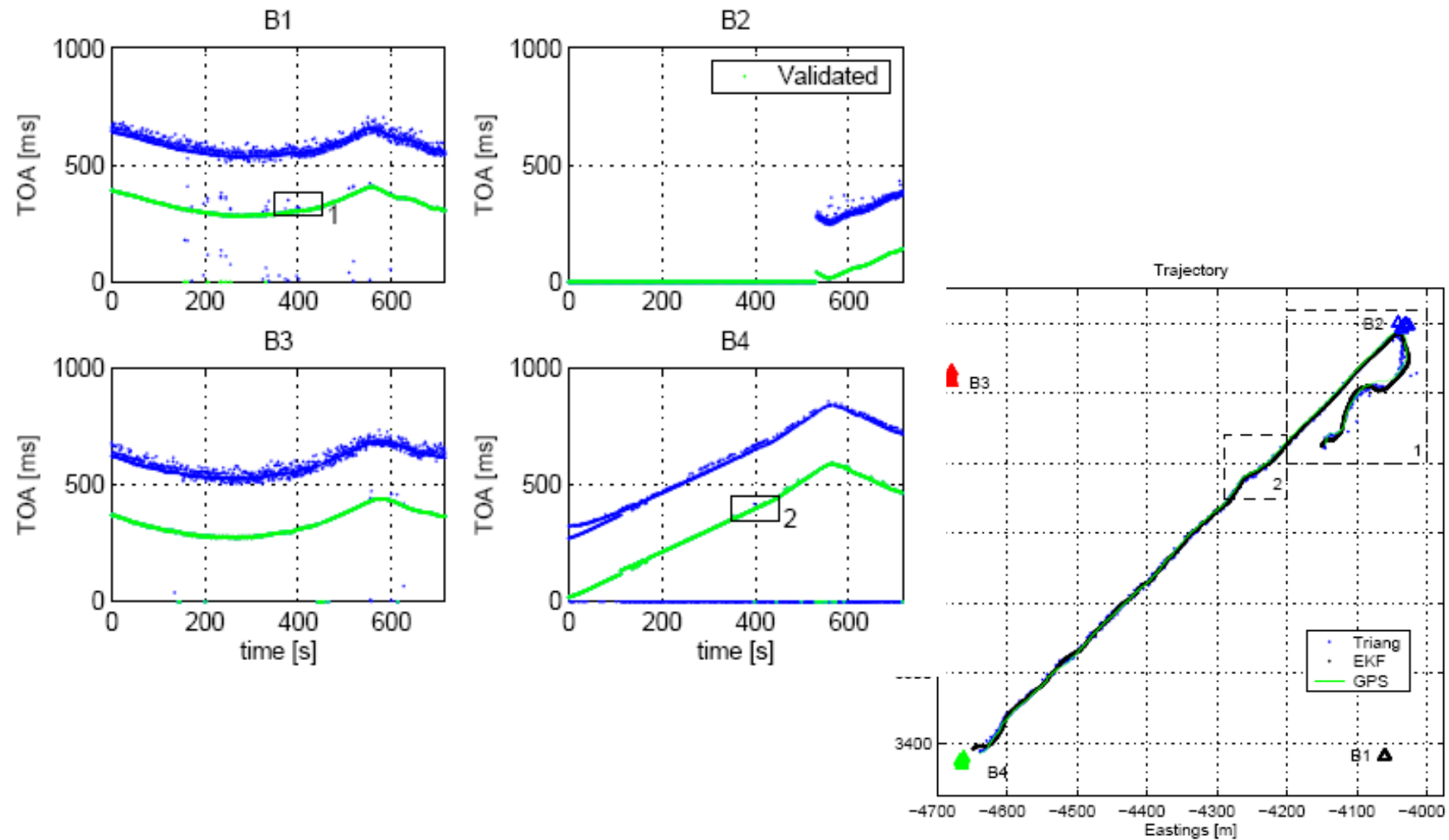
- DGPS receiver
- Radio link
- PC with tracking software



Acoustic pinger

GPS Intelligent Buoys (GIB)-ACSA/ORCA

Tracking with a Sensor Network (cont.)



The mathematical estimation problem

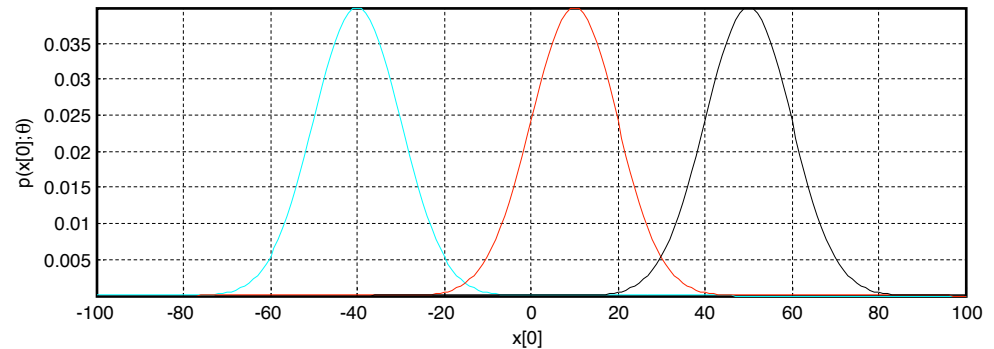
To be possible to design estimators, first the data must be modeled.

Example I:

Assume that one sample x is available (scalar example, i.e. $N=1$) with constant **unknown** mean θ .

The probability density function (PDF) is

$$p(x; \theta) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-\theta)^2}{2\sigma^2}}$$



For instance if $x[0] < 0$ it is doubtful that the unknown parameter is $\gg 0$.

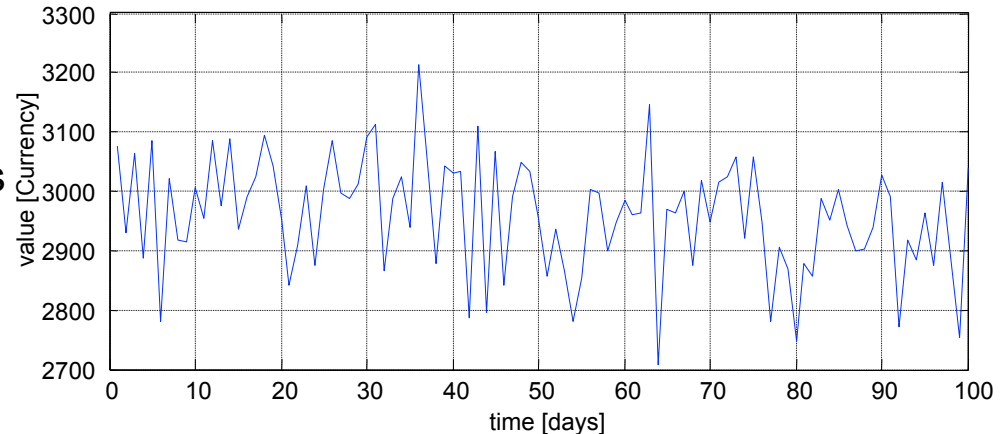
In a actual problem, we are not given a PDF, but must be chosen to be consistent with the data and with the prior knowledge.

The mathematical estimation problem

Example II:

Now the following sequence \mathbf{x} is given.

Note that the value along time appears to be decreasing. Lets consider that the phenomena is described by



$$x[n] = A + Bn + w[n] \quad n = 0, 1, \dots, N-1$$

where A and B are constant unknown parameters and $w[n]$ is assumed to be white Gaussian noise, with PDF $N(0, \sigma^2)$. For $\theta = [A \ B]$ and $\mathbf{x} = [x[0] \ x[1] \ \dots \ x[n]]$ the data PDF is

$$p(\mathbf{x} | \theta) = \frac{1}{(2\pi\sigma^2)^{\frac{N}{2}}} e^{-\frac{1}{2\sigma^2} \sum_{n=0}^{N-1} (x[n] - A - Bn)^2}$$

Where the uncertainty in the samples is assumed to be uncorrelated.

The performance of the estimators is dependent on the models used, so they must be mathematically treatable..

The mathematical estimation problem

Classical estimation techniques

Parameters are assumed deterministic but unknown

Bayesian techniques

Parameters are used to be unknown but are stochastic also described by a PDF.

The joint PDF would then be

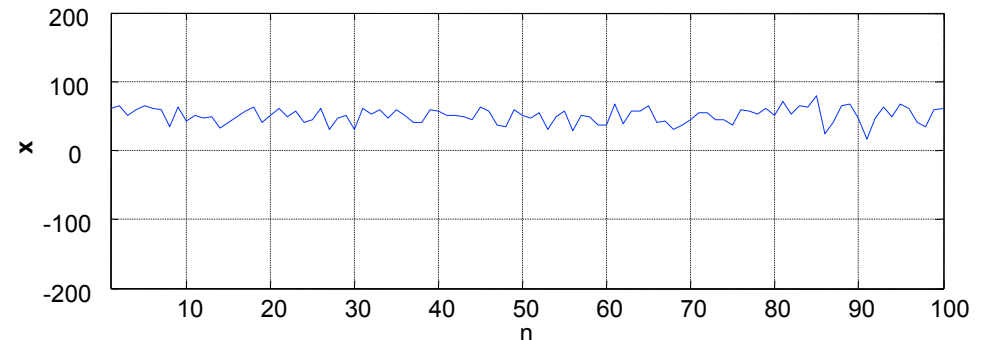
$$p(\mathbf{x}, \theta) = p(\mathbf{x} | \theta) p(\theta)$$

The diagram shows the equation $p(\mathbf{x}, \theta) = p(\mathbf{x} | \theta) p(\theta)$ with two green circles around the terms $p(\mathbf{x} | \theta)$ and $p(\theta)$. An arrow points from the circle around $p(\mathbf{x} | \theta)$ to the text "Dependence of data on the parameters". Another arrow points from the circle around $p(\theta)$ to the text "Prior knowledge".

Exploiting simple estimators

Example III (Quiz):

Given a data sequence from a signal with PDF as described by one of three models
Which one is the correct model?



First scenario:

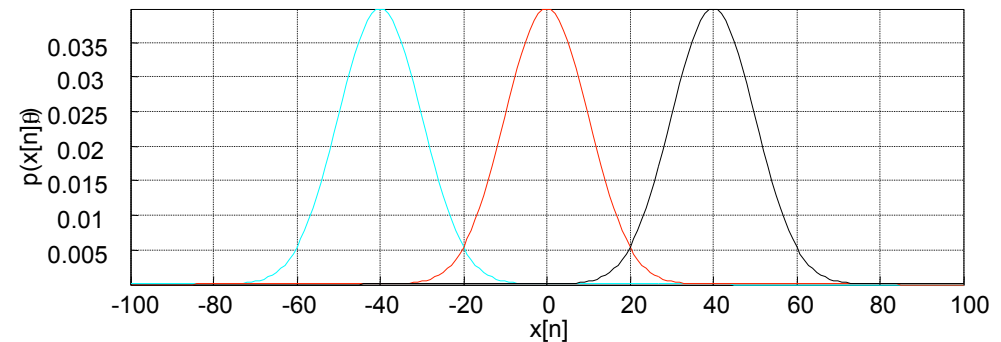
$$N = 100$$

$$\theta \in \{-40, 0, 40\}$$

$$\sigma^2 = 10^2$$

For the signal

$$x[n] = \theta + w[n] \quad n = 0, 1, \dots, N-1$$



The answer is obvious:

$$\theta = 40!$$

Exploiting simple estimators

Second scenario

(lousy sensor quality or lousy data):

Lets *repeat the problem with*

$$\sigma^2 = 100^2$$

$$N = 100$$

$$\theta \in \{-40, 0, 40\}$$

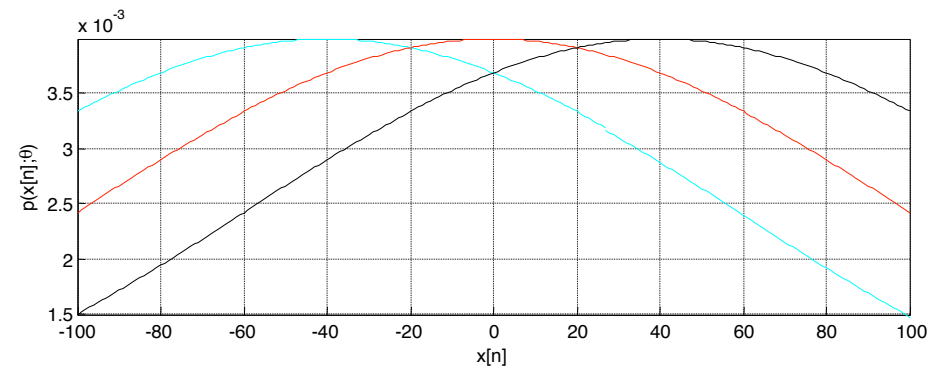
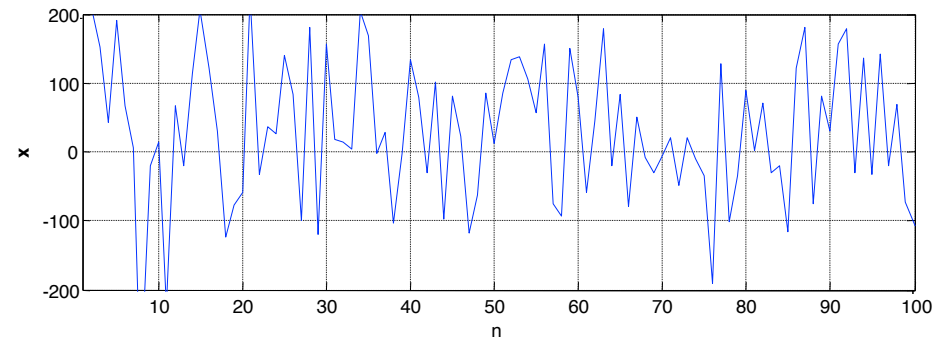
The answer is not obvious anymore!

Lets propose a couple of estimators

and to study them...

$$\hat{\theta}_1 = x[0]$$

$$\hat{\theta}_2 = \frac{1}{N} \sum_{n=0}^{N-1} x[n]$$



Assessing estimator performance

Estimators depend only on observed data thus can be viewed as a function

$$\theta = g(x[0], \dots, x[N]) = g(\mathbf{x})$$

The study of estimator properties must be done resorting to statistic tools.

Is it exact?, i.e. Does it return the true value of the unknown parameters?

Is this a good estimator? If many experiments can be performed, is it expected that the unknown parameter is achievable? Or are the results expected to be biased?

$$\begin{aligned} E[\hat{\theta}_1] &= E[x[0]] = \theta \\ E[\hat{\theta}_2] &= E\left[\frac{1}{N} \sum_{n=0}^{N-1} x[n]\right] = \\ &= \frac{1}{N} \sum_{n=0}^{N-1} E[x[n]] = \\ &= \frac{1}{N} N\theta = \theta \end{aligned}$$

Assessing estimator performance

How good is an estimator? How much uncertainty corresponds to the computed value?

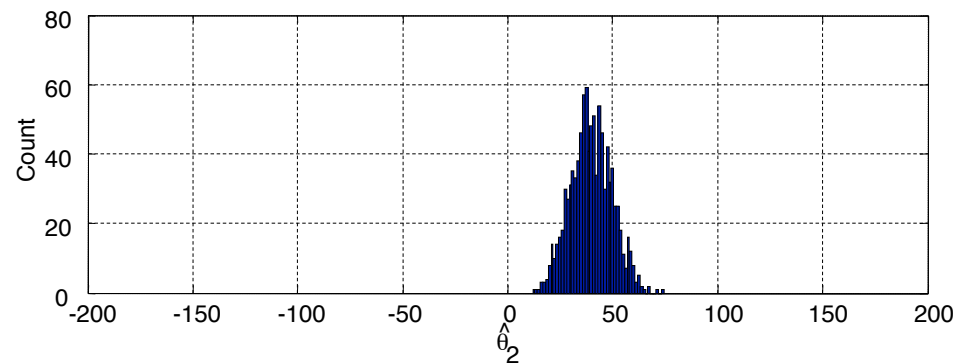
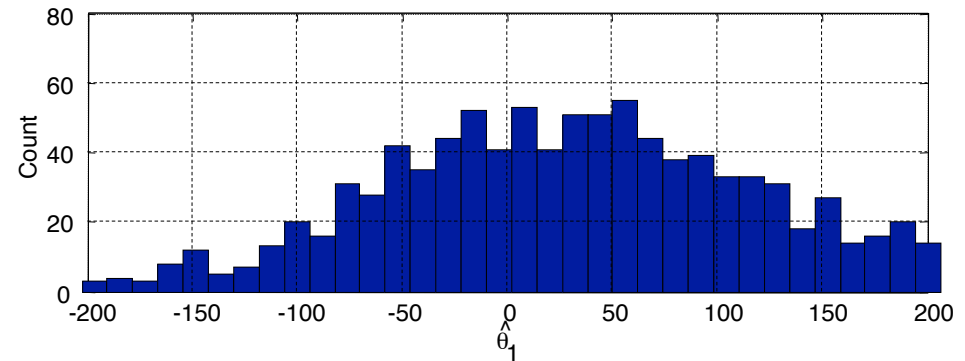
The use of computational tools is a good idea? No!

Formal methods are required

For our quiz:

$$\text{var}(\hat{\theta}_1) = \text{var}(x[0]) = \sigma^2$$

$$\text{var}(\hat{\theta}_2) = \text{var}\left(\frac{1}{N} \sum_{n=0}^{N-1} x[n]\right) = \frac{1}{N^2} \sum_{n=0}^{N-1} \text{var}(x[n]) = \frac{1}{N^2} N\sigma^2 = \frac{\sigma^2}{N}!$$



Assessing estimator performance

Questions triggered from this simple example but valid to all our problems:

- The second estimator is much better than the first estimator.
- The quality of the estimate increases with the number of points. Is it reasonable? Is it plausible?
- Do we have always data available? How to get data?
- Is this the best one can do with N samples?
- Are there better estimators that we can exploit?



Answers to this questions will be provided along the course...

Bibliography:

Further reading

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See for instance <http://www.ieee.org/portal/site>