Noise-reduction process and useful signal interpretation on recorded passive acoustic signals using time-frequency representations

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Introduction (1/4)

A major challenge in underwater communication:
 Collect and distribute subsurface data from multiple distributed instruments in real time

Problem: To achieve useful spatial coverage, the subsurface measurement involves multiple instruments deployed with separation of several kilometers

Seafloor wires and cables?

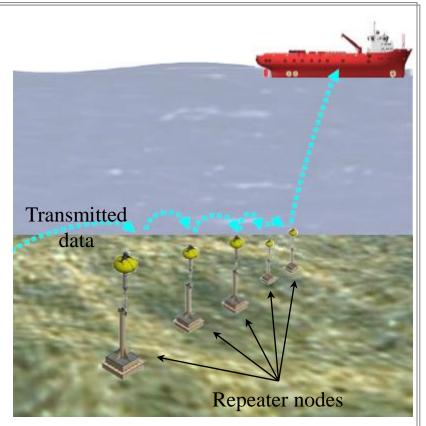
⇒Deployment cost, Incompatible with bottom-fishing activities

Show Array of networked acoustic modems ⇒Example: FRONT* Project

*Front-Resolving Observation Network with Telemetry, funded by the NOPP

Introduction (2/4)

Networked acoustic modems: Based on the use of repeater nodes \Rightarrow Individual acoustic modems \Rightarrow Use to relay the message > Repeater node principle: Decoding received data from the previous node Sencoding and sending data to the next one



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Image from Sonardyne International

Problem: strong dependency with the modulation techniques
What happens in case of different systems on the same area?

Introduction (3/4)

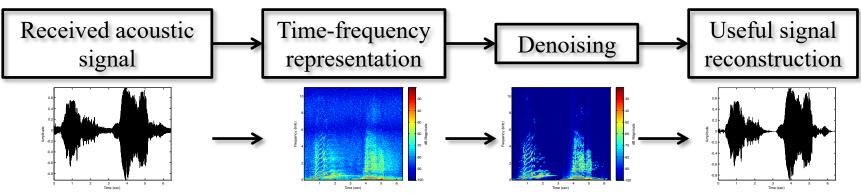
Idea: Develop an acoustic repeater with no dependency on the type of modulation (blind system)

Substitution System System

- Principle: Applied a denoising process on the received signal and amplified the resulting data before resending them
- Several denoising process based on:
 Signal and/or noise statistics knowledge lack of knowledge ?
 Gaussian noise assumptions disturbing signal = realization of a non-Gaussian process ?
 - > Noise reduction in time domain \Leftrightarrow smoothing effect

Introduction (4/4)

Proposition: take advantage of a time-frequency plane



♦ Use of time-frequency plane ⇔ Gaussian properties
 ♥ Suited to non-stationary signals ⇔ time resolution preservation

Which kind of time-frequency transformation ?
 Time and frequency resolutions <> Heisenberg principle
 Narrow frequency resolution <> high amount of computations
 Must make possible the time domain reconstruction

Contents

➤ The Hearingogram

- The denoised *Hearingogram*
- ➤ Useful signal reconstruction

> Experiments

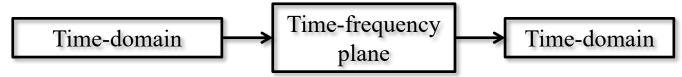
- Simulated signal
- Real underwater records



The *Hearingogram* (1/3)

≻ Main idea

- Solve three main problems:
 - ⇒Time-frequency approaches ⇔ resolution problems or interferences
 - ⇒Narrow frequency resolution ⇔ memory problems and high computing time
 - ⇒Invertible process:



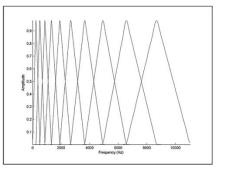
STake into account the human physiology

The *Hearingogram* (2/3)

> The Mel's filters

Human ear = filters concentrated only on certain frequency components

SMel's filters non-uniformly spaced on the frequency axis



⇒Triangular shaped filters bank in accordance with the Mel's scale:

$$mel(\nu) = 2595 \log_{10} \left(1 + \frac{\nu}{700} \right)$$

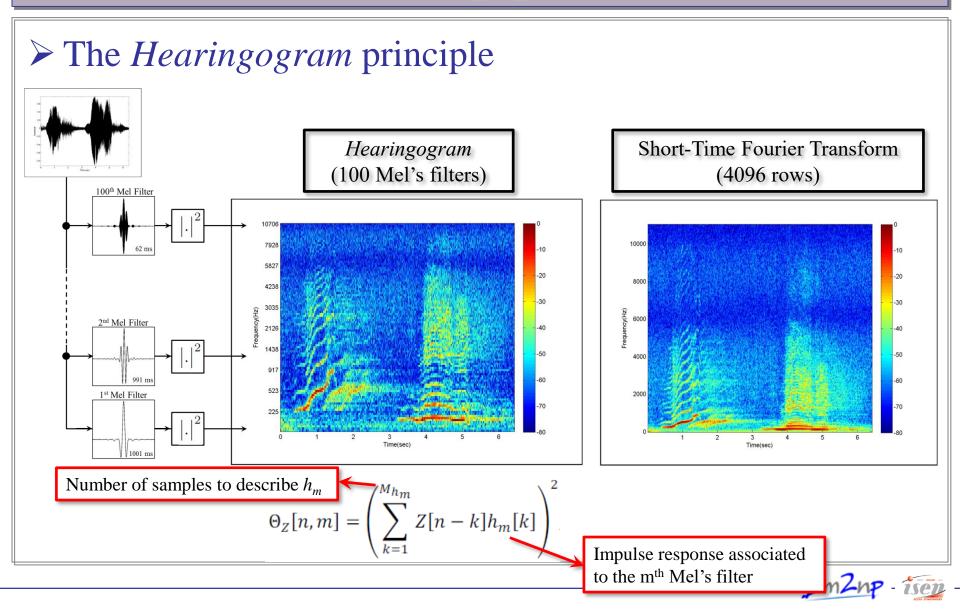
Center frequencies

$$v_m = 700 \left(10^{\frac{mel(v_m)}{2595}} - 1 \right)$$

with: $mel(v_m) = \frac{m}{M+1} (mel(v_{max}) - mel(v_{min}))$

The *Hearingogram* (3/3)

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The denoised *Hearingogram* (1/2)

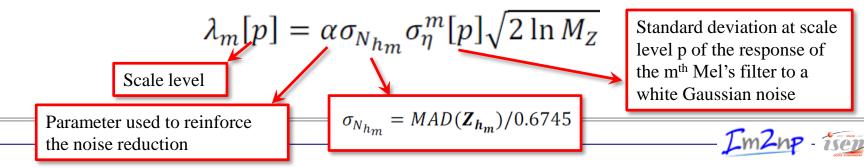
> Principle

Statistical properties: the disturbing terms in the received signal become Gaussian, before the square magnitude, in $\Theta_Z[n,m]$

$$\boldsymbol{N}_{\boldsymbol{h}_{m}} \hookrightarrow \mathcal{N}\left(0, \sigma_{N_{\boldsymbol{h}_{m}}}^{2}\right)$$

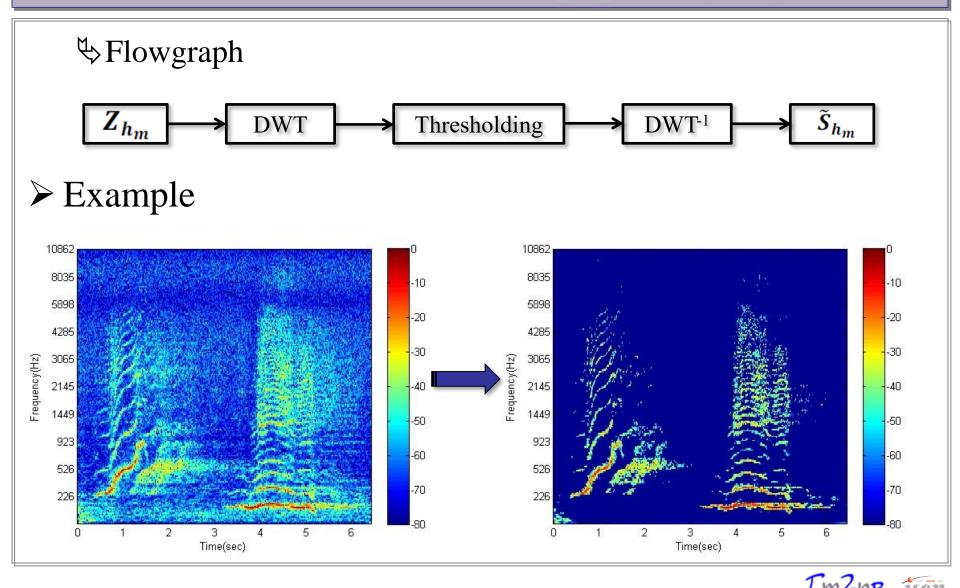
 \clubsuit Gaussian noise \Leftrightarrow wavelet based denoising method

- ⇒Discrete Wavelet Transform (DWT) following the Mallat multiresolution algorithm
- ⇒Wavelet coefficients thresholding setting to zero the coefficients lower than the *universal* threshold:



The denoised *Hearingogram* (2/2)

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Denoised useful signal reconstruction (1/2)

➢ Principle

SWhole Mel's filters bank = band-pass filter

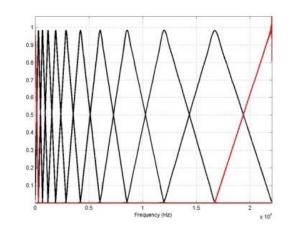
$$H^{Mel}(\nu) = \sum_{m=1}^{M} H_m(\nu) = 1, \forall \nu \in [\nu_1; \nu_M]$$

with $[mel(v_{min}); v_1[$ and $]v_M; mel(v_{max})]$ as transition widths

⇒ Energy conservation ⇔ add two filters

$$H(\nu) = H_0(\nu) + H_{M+1}(\nu) + H^{Mel}(\nu)$$

 $= 1,$
 $\forall \nu \in [0; F_s/2]$



M+1

m=0

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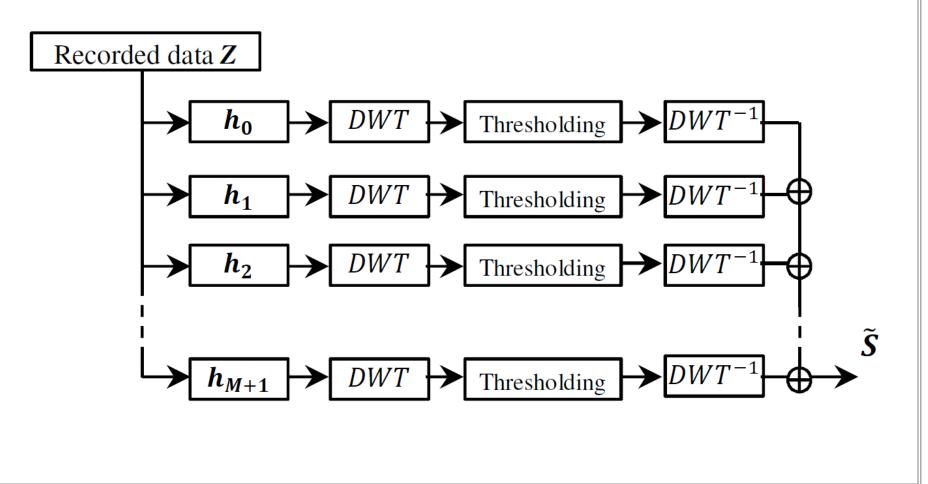
Associated impulse response

 $\boldsymbol{h} = TF^{-1}[H(\nu)] \cong \delta$

 \Rightarrow Approximation of the useful signal: $\tilde{s} = \sum \tilde{s}_{h_m}$

Denoised useful signal reconstruction (2/2)



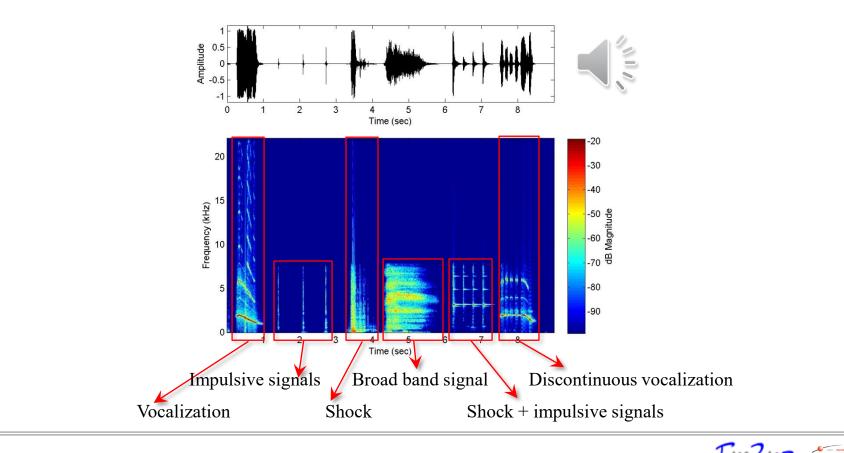


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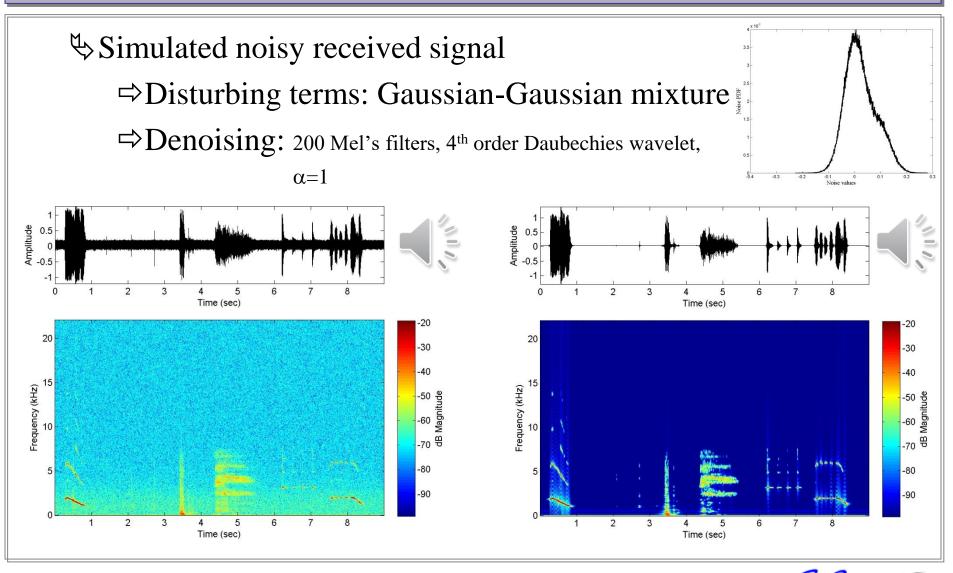
Experiments (1/5)

Simulated data

Stest signal (sampling frequency: 44 100 Hz)



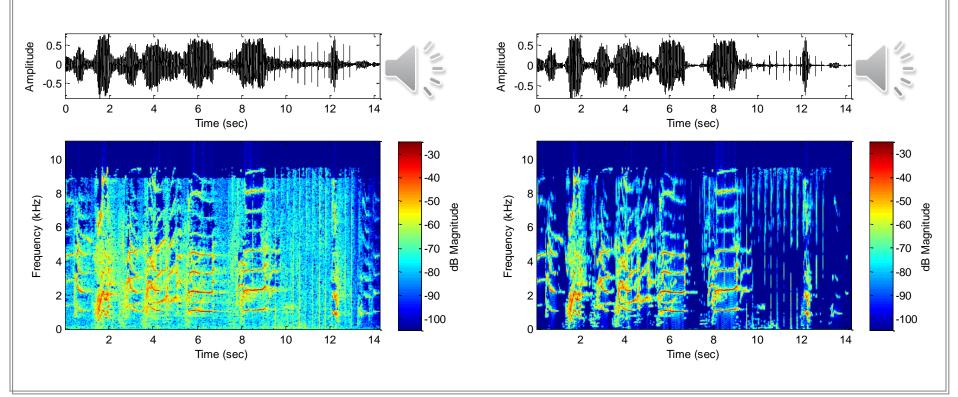
Experiments (2/5)



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Experiments (4/5)

Real underwater records
Killer whale vocalization
(sampling frequency: 22 050 Hz)

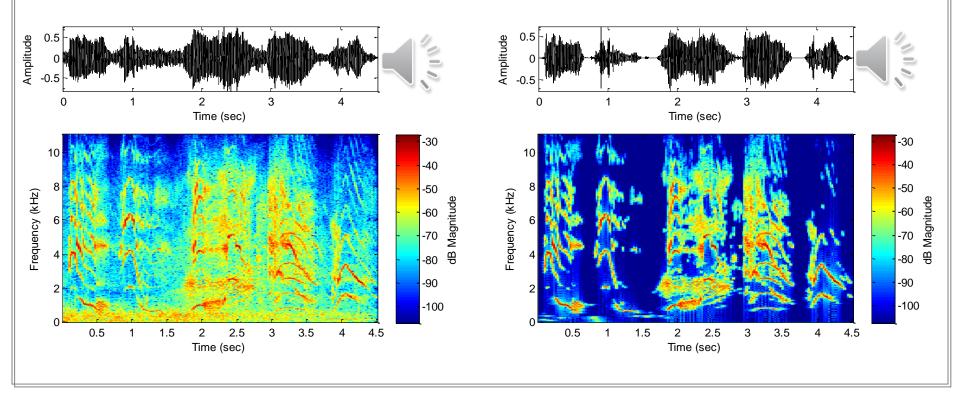


m2np ise

Experiments (5/5)

Solphin sounds (sampling frequency: 22 050 Hz)





Concluding remarks

- Results obtained on real and simulated data reveal the efficiency of the denoising process
- Method can be easily parallelized compatibility
 Real-time

Noise reduction

No assumption or information about the useful signal and noise
 Blind process fully automatic (key point in fully automatic and operational systems)

Next step: sea trials in a context of wireless underwater communications





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