

# A Novel Thresholding Technique for the Denoising of Multicomponent Signals

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# Plan

- 1 Introduction
- 2 Definitions
- 3 Signal Denoising from STFT
- 4 New Method for the Denoising of the STFT of MCSs
- 5 Conclusions and Perspectives

# Context

- Denoising oscillatory signals or superposition of AM/FM modes (called *multicomponent signals* (MCSs)) from their Short-Time Fourier Transform (STFT).
- Illustration of the STFTs of oscillatory signals: real PCG (phonocardiogram) signals (SiSEC 2016 database):

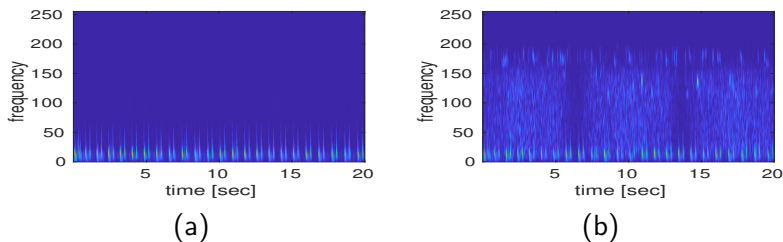


Figure: (a): STFT of noise-free PCG; (b): STFT of noisy PCG

# Objectives

- Hard-Thresholding (HT) probably the most popular thresholding technique: **keep only STFT coefficients above a certain level in terms of their magnitude** before signal reconstruction [1].
- Our goal here is to improve HT in the context of MCS: **one wants to recover the modes while denoising them.**

[1] D. Donoho and I. Johnstone, *Ideal spatial adaptation via wavelet shrinkage*, Biometrika, vol. 81, pp. 425-455, 1994.

# STFT and Reconstruction Formula

- The time-frequency representation our denoising algorithm is based on is STFT computed as:

$$V_f^g\left(m, \frac{k}{N}\right) = \sum_{n \in \mathbb{Z}} f[n]g[n - m]e^{-i2\pi \frac{k(n-m)}{N}}.$$

in which  $N$  is the number of frequency bins.

- STFT is invertible through (assuming  $\|g\|_2 = 1$ ):

$$f[n] = \sum_{m \in \mathbb{Z}} \sum_{k=0}^{N-1} V_f^g\left(m, \frac{k}{N}\right)g[n - m] \frac{e^{i2\pi \frac{k(n-m)}{N}}}{N},$$

# Multicomponent Signals (MCSs)

- *Multicomponent signals* are defined as:

$$f[n] = \sum_{l=1}^K f_l[n], \text{ with } f_l[n] = A_l[n]e^{i2\pi\phi_l[n]}, \quad n = 0, \dots, L-1$$

for some  $K$ , where  $A_l[n] > 0$ ,  $\phi'_l[n] > 0$  for any  $n$  and  $l$ ,  $(m, \phi'_l(m))$  is called *ridge* associated with the  $l$ th mode.

- $f_l$ s, called *modes or components*, are separated with resolution  $\Delta$ , if for all  $l \in \{1, \dots, K-1\}$ :

$$\phi'_{l+1}[n] - \phi'_l[n] > 2\Delta \quad \forall n.$$

# HT Denoising Technique

- Assume  $f$  complex signal contaminated by complex white Gaussian noise  $\Phi$  with variance  $\sigma^2$ :  $\tilde{f}[n] = f[n] + \Phi[n]$
- $V_{\Phi}^g(m, \frac{k}{N})$  also Gaussian with zero mean and satisfies:

$$\text{Var} \left( \Re \left\{ V_{\Phi}^g \left( m, \frac{k}{N} \right) \right\} \right) = \text{Var} \left( \Im \left\{ V_{\Phi}^g \left( m, \frac{k}{N} \right) \right\} \right) = \sigma^2 \|g\|_2^2,$$

- HT in that context can be viewed as thresholding the coefficients:

$$\overline{V}_{\tilde{f}}^g \left( m, \frac{k}{N} \right) = \begin{cases} V_{\tilde{f}}^g \left( m, \frac{k}{N} \right), & \text{if } |V_{\tilde{f}}^g \left( m, \frac{k}{N} \right)| \geq 3\sigma \|g\|_2 \\ 0 & \text{otherwise,} \end{cases}$$

$\frac{\Re \{ V_{\Phi}^g(m, \frac{k}{N}) \}^2 + \Im \{ V_{\Phi}^g(m, \frac{k}{N}) \}^2}{\sigma^2 \|g\|_2^2}$  is  $\chi_2$  distributed with two degrees of freedom, a threshold of 9 is associated with a level of significance 0.01

- Reconstruction: replace  $V_{\tilde{f}}^g(m, \frac{k}{N})$  by  $\overline{V}_{\tilde{f}}^g(m, \frac{k}{N})$  in reconstruction formula.

# HT in the Context of MCS

- **Ridges extraction:**  $\phi'_l$  approximated by some  $\psi_l$  following an existing approach [2].
- **STFT thresholding:** Select the coefficients above the threshold provided by HT in the vicinity of the detected ridge.

$$\eta_{l,m}^{[1]} = \operatorname{argmax}_k \left\{ \frac{k}{N} < \psi_l(m), |V_{\tilde{f}}^g(m, \frac{k}{N})| < 3\sqrt{2}\sigma \|g\|_{2,m} \right\}$$

$$\eta_{l,m}^{[2]} = \operatorname{argmin}_k \left\{ \frac{k}{N} > \psi_l(m), |V_{\tilde{f}}^g(m, \frac{k}{N})| < 3\sqrt{2}\sigma \|g\|_{2,m} \right\},$$

and then define:  $J_{l,m} := [N\psi_l(m) - \eta_{l,m}^{[1]}, N\psi_l(m) + \eta_{l,m}^{[2]}]$ .

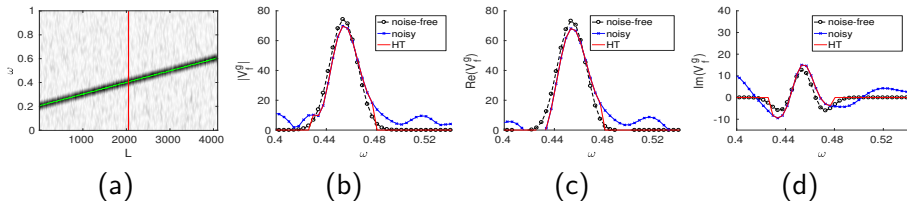
- **Mode reconstruction:**

$$f_l[n] = \sum_{m \in \mathbb{Z}} \sum_{k \in J_{l,m}} V_{\tilde{f}}^g(m, \frac{k}{N}) g[n-m] \frac{e^{i2\pi \frac{k(n-m)}{N}}}{N}.$$

[2] R. Carmona, W. Hwang, and B. Torr esani, *Characterization of signals by the ridges of their wavelet transforms*, IEEE Transactions on Signal Processing, vol. 45, no. 10, pp. 2586-2590, Oct 1997.



# Illustration of HT on a Mono Component Signal



**Figure:** (a): modulus of the STFT of a noisy linear chirp (SNR = 0 dB) along with its ridge detection (green line); (b): modulus of STFTs (noise-free, noisy and hard-thresholded) at  $m = L/2$  (red line); (c): the real part of STFT displayed in (b); (d): same as (c) but for the imaginary part.

Several drawbacks of HT:

- Potential shift of the local maxima compared with noise-free case
- Lack of symmetry and regularity (the noise-free transform has the regularity of  $g$ )

## Local Linear Chirp Mode Approximation

- Second-order phase approximation close to the  $l$ th ridge:

$$V_{f_l}^g(m, \frac{k}{N}) \approx \left| V_{f_l}^g(m, \frac{k}{N}) \right| e^{i2\pi\Psi_l(m, \frac{k}{N})},$$

when  $g$  is the Gaussian window  $\sigma_s^{-\frac{1}{2}} e^{-\pi\frac{x^2}{\sigma_s^2}}$ , with

$$\Psi_l(m, \frac{k}{N}) = -\frac{\phi_l''(m)}{2r(m)^2} \left( \frac{k}{N} - \phi_l'(m) \right)^2 - \frac{\theta(m)}{4\pi} + \phi_l(m)$$

$$\left| V_{f_l}^g(m, \frac{k}{N}) \right| = A_l(m) \sigma_s^{-\frac{1}{2}} r(m)^{-\frac{1}{2}} e^{-\frac{\pi \left( \frac{k}{N} - \phi_l'(m) \right)^2}{\sigma_s^2 r(m)^2}}$$

with  $\theta(m) = \arctan(-\phi_l''(m)\sigma_s^2)$  and  $r(m) = \sigma_s^{-2} (1 + \phi_l''(m)^2 \sigma_s^4)^{\frac{1}{2}}$

- $\left| V_{f_l}^g(m, \frac{k}{N}) \right|_k$ ,  $\Re \left( V_{f_l}^g(m, \frac{k}{N}) \right)_k$  and  $\Im \left( V_{f_l}^g(m, \frac{k}{N}) \right)_k$  even functions in the vicinity of the  $l$ th ridge, and all reach an extremum at  $\frac{k}{N} = \phi_l'(m)$ .

## Novel Algorithm for MCS Denoising

- **Compensate for potential shift:** shift the imaginary and real parts of STFT so that their extrema correspond to  $\psi_I(m)$ .
- **Compensate for lack of symmetry:** enforce the symmetry of the real and imaginary part of STFT with respect to  $\psi_I(m)$ , by defining:

$$\eta = \max(\eta_{l,m}^{[1]}, \eta_{l,m}^{[2]}) \text{ and then}$$

$$\forall \nu \in [0, \eta] : \frac{\left( V_{\tilde{f}}^g(m, \psi_I(m) + \frac{\nu}{N}) + V_{\tilde{f}}^g(m, \psi_I(m) - \frac{\nu}{N}) \right)}{2} \rightarrow X$$

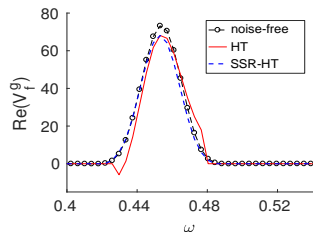
$$X \rightarrow V_{\tilde{f}}^g(m, \psi_I(m) \pm \frac{\nu}{N})$$

- **Compensate for lack of regularity:** smooth the just obtained real and imaginary parts using piecewise cubic Hermite interpolation.

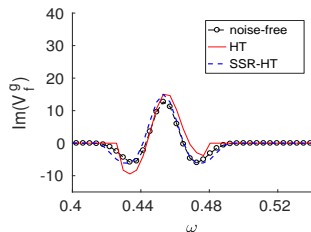
# Illustration on a Simple Example

Process called *SSR-HT*:

*Shifted-Symmetrized-Regularized Hard-Thresholding.*



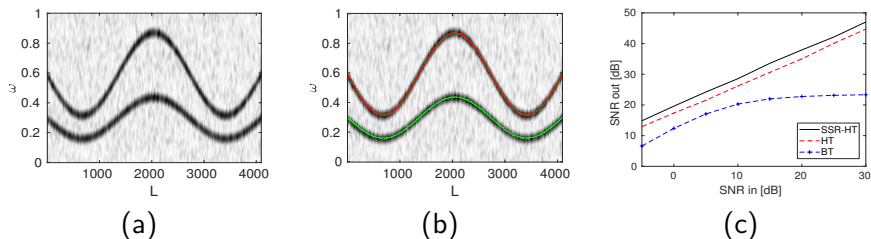
(a)



(b)

**Figure:** Illustration of SSR-HT on the real and imaginary parts of the STFT displayed in the previous Figure

# Numerical Experiments: a Simulated MCS

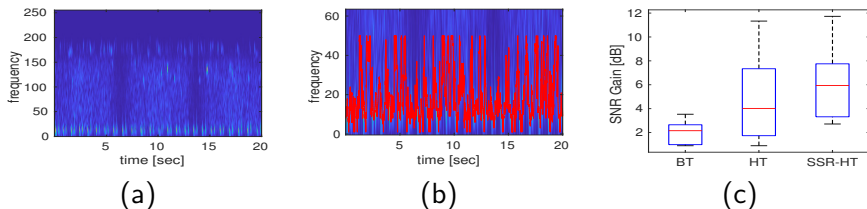


**Figure:** (a): STFT of the simulated MCS; (b): detected ridges of the signal displayed in (a); (c): denoising performance comparison of the studied methods (BT (Block-Thresholding [3]), HT (Hard-Thresholding), SSR-HT (our new method))

[3] G. Yu, S. Mallat, and E. Bacry, *Audio denoising by time-frequency block thresholding*, IEEE Transactions on Signal Processing, vol. 56, no. 5, pp. 1830-1839, May 2008.

# Numerical Experiments: Reals Noisy PCGs

- 16 PCGs were recorded with a sampling rate 1KHz by a cardiac microphone MLT210 on three healthy volunteers (SiSEC 2016 database [4]).
- Artificially contaminated by different real interference (radio, cough, pseudo-periodic breathing noise, etc.)



**Figure:** (a): STFT of a noisy PCG; (b): ridge detection on a low-pass filtered version with a cutoff frequency 64Hz of signal in (a) (assuming number of modes  $K = 1$ ); (c): denoising performance comparison (using boxplot) .

[4] A. Liutkus, et al., *The 2016 signal separation evaluation campaign*, in *Int. Conf. on Latent Variable Anal. and Sig. Separation*, 2017, pp. 323-332.

# Conclusions and Perspectives

- We have detailed a new method for the denoising of multicomponent signals based on an improvement of hard-thresholding.
- This approach can be extended to the denoising of STFT computed using any symmetric and regular window.
- Further works should be carried out to adapt the proposed new technique to noise other than Gaussian.

# Thank you for your attention!