

Filtering out time-frequency areas using Gabor multipliers

A. Marina Krémé^{1,2}, Valentin Emiya², Caroline Chaux¹ and Bruno Torrèsani¹

¹ *Institut de Mathématiques de Marseille, Aix Marseille Univ, CNRS, France.*

² *Laboratoire d'Informatique et des Systèmes, Aix Marseille Univ, CNRS, France.*



Institut de Mathématiques de Marseille



IMM - UMR CNRS 7373

Aix-Marseille Université • CNRS • Ecole Centrale de Marseille



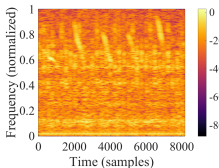
Outline

- 1 Filtering out time-frequency areas
 - Problem overview and baseline solutions
 - Problem statement
- 2 Gabor transform -Gabor multiplier
 - Gabor transform
 - Gabor multiplier
 - Some properties
- 3 Analytical solution
 - TF areas filtering: normal equations
 - TF areas filtering: analytical solution
 - TF areas filtering: λ setting
- 4 Numerical simulations
 - Problem data
 - Comparative reconstruction results and SNRs
 - More experiments
- 5 Conclusion

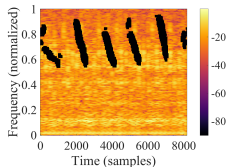
Outline

- 1 Filtering out time-frequency areas
 - Problem overview and baseline solutions
 - Problem statement
- 2 Gabor transform -Gabor multiplier
- 3 Analytical solution
- 4 Numerical simulations
- 5 Conclusion

Problem overview and baseline solutions

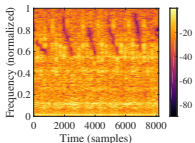


Obs: target + perturbation

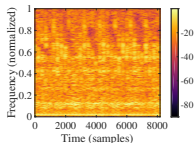


Ω : masked area

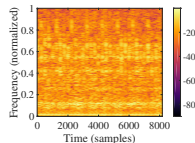
Goal: filter out signal perturbation and minimize the artefacts



ZerVal method:
fill Ω with zeros
Audible artefact



RandVal method:
interp in Ω + random
phases [ANSYS]



RedEnergy
Our method

Problem of filtering out time-frequency areas formulation

Time-frequency transform: $\mathcal{T} : \mathbb{R}^L \longrightarrow \mathbb{R}^{M \times N}$

The perturbation signal \mathbf{x}_{per} is well localized in a region Ω and the target signal \mathbf{x}_{ref} is everywhere in TF plane

Observations: $\mathbf{x}_0 = \mathbf{x}_{ref} + \mathbf{x}_{per}$

Problem of filtering out time-frequency areas formulation

$$\mathbf{x}_\lambda^* = \underset{\mathbf{x} \in \mathbb{C}^L}{\operatorname{argmin}} \underbrace{\|\mathcal{T}\mathbf{x} - \mathcal{T}\mathbf{x}_0\|_\Omega^2}_{\text{data fitting}} + \lambda \underbrace{\|\mathcal{T}\mathbf{x}\|_\Omega^2}_{\text{energy within } \Omega}, \quad \lambda > 0,$$

$$\text{where } \|\mathbf{y}\|_\Omega^2 := \sum_{k \in \Omega} |\mathbf{y}[k]|^2.$$

Outline

- 1 Filtering out time-frequency areas
- 2 Gabor transform -Gabor multiplier**
 - Gabor transform
 - Gabor multiplier
- 3 Analytical solution
- 4 Numerical simulations
- 5 Conclusion

Gabor transform

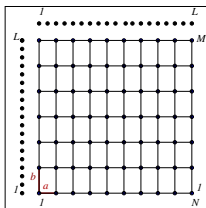


Figure: Lattice: $\Lambda = \mathbb{Z}_M \times \mathbb{Z}_N$ with $N = L/a$ and $M = L/b$

For a window $\mathbf{g} \in \mathbb{R}^L$, the Gabor atom \mathbf{g}_{mn} at TF point $(m, n) \in \Lambda$:

$$\mathbf{g}_{mn}[l] = \mathbf{g}[l - na]e^{2i\pi mbl/M}, \forall l \in \mathbb{Z}_L.$$

Gabor transform:

$$\mathcal{V}_{\mathbf{g}}\mathbf{x}[m, n] = \langle \mathbf{x}, \mathbf{g}_{mn} \rangle = \sum_{l=0}^{L-1} \mathbf{x}[l]\mathbf{g}[l - na]e^{-2i\pi mbl/M}.$$

Gabor multiplier

For suitably chosen \mathbf{g} and Λ (i.e. generating a tight Parseval frame) then, $\forall \mathbf{x} \in \mathbb{C}^L$, we have:

$$\mathbf{x} = \sum_{m,n} \mathcal{V}_{\mathbf{g}} \mathbf{x}[m, n] \mathbf{g}_{mn} = \sum_{m,n} \langle \mathbf{x}, \mathbf{g}_{mn} \rangle \mathbf{g}_{mn} .$$

Gabor multiplier is similarly defined via a transfer function also called mask \mathbf{m} in the time-frequency domain, i.e.,

Definition

The Gabor multiplier associated to (\mathbf{g}, Λ) with mask $\mathbf{m} \in \mathbb{R}^{M \times N}$ is defined by:

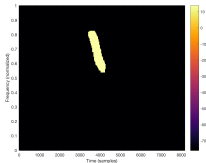
$$\mathfrak{M}_{\mathbf{m}} \mathbf{x} = \sum_{m,n} \mathbf{m}[m, n] \langle \mathbf{x}, \mathbf{g}_{mn} \rangle \mathbf{g}_{mn} ,$$

where \mathbf{m} denotes also the operator of pointwise multiplication by \mathbf{m} .

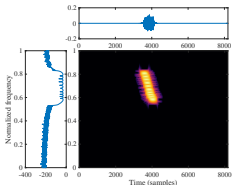
Main property [H. G. Feichtinger and K. Nowak, 2002.]

If \mathbf{m} is real-valued then $\mathfrak{M}_{\mathbf{m}}$ is Hermitian. Then there is an orthonormal basis of \mathbb{C}^L formed by $\mathfrak{M}_{\mathbf{m}}$ eigenvectors.

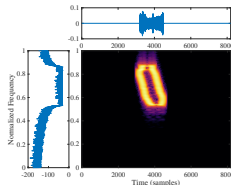
Illustration: eigenvectors associated with large eigenvalues tend to follow TF localization properties of the mask.



(a) mask



(b) Large eigenvalue



(c) Small eigenvalue

Outline

- 1 Filtering out time-frequency areas
- 2 Gabor transform -Gabor multiplier
- 3 Analytical solution**
 - TF areas filtering: normal equations
 - TF areas filtering: analytical solution
 - TF areas filtering: λ setting
- 4 Numerical simulations
- 5 Conclusion

TF areas filtering: normal equations

Objective function:

$$f(\mathbf{x}) = \|\mathcal{V}_{\mathbf{g}}\mathbf{x} - \mathcal{V}_{\mathbf{g}}\mathbf{x}_0\|_{\Omega}^2 + \lambda \|\mathcal{V}_{\mathbf{g}}\mathbf{x}\|_{\Omega}^2$$

Normal equations:

$$\begin{aligned} (\mathfrak{M}_{\overline{\Omega}} + \lambda \mathfrak{M}_{\Omega})\mathbf{x} &= \mathfrak{M}_{\overline{\Omega}}\mathbf{x}_0 \\ [I + (\lambda - 1)\mathfrak{M}_{\Omega}]\mathbf{x} &= (I - \mathfrak{M}_{\Omega})\mathbf{x}_0 \end{aligned}$$

$[I + (\lambda - 1)\mathfrak{M}_{\Omega}]$ is invertible for $\lambda > 0$ due to:

If $\mathbf{m} \in \mathbb{C}^{M \times N}$, then $\mathfrak{M}_{\mathbf{m}}$ defines a bounded operator with operator norm $\|\mathfrak{M}_{\mathbf{m}}\|_{op} \leq C\|\mathbf{m}\|_{\infty}$, where C is a constant. In particular, if \mathbf{g} and Λ generate a Parseval frame, then $\|\mathfrak{M}_{\mathbf{m}}\|_{op} \leq \|\mathbf{m}\|_{\infty}$.

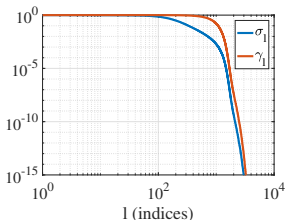
TF areas filtering: analytical solution

\mathfrak{M}_Ω is diagonalizable, so there is a unitary matrix U and a diagonal matrix $D = \text{diag}(\sigma_1, \dots, \sigma_L)$, $\sigma_1 \geq \dots \geq \sigma_L$ such that $\mathfrak{M}_\Omega = UDU^{-1}$.

Analytical solution

$$\mathbf{x}_\lambda^* = \mathbf{x}_0 - U \underbrace{\text{diag} \left(\frac{\lambda \sigma_l}{1 - (1 - \lambda) \sigma_l} \right)}_{\gamma_l, l=1, \dots, L} U^{-1} \mathbf{x}_0, \lambda > 0$$

Typical behavior of eigenvalues ($\lambda = 0.1$): $0 \leq \sigma_l, \gamma_l \leq 1$



Efficient tuning of the hyperparameter

Computational issues

- 1 Most costly part: diagonalization of \mathfrak{M}_Ω . It does not depend on λ and is done only once 😊.
- 2 Low rank approximations can be used to reduce computing time 😊.

Tuning of the regularization parameter λ

- λ can be tuned to match a prescribed target energy $E = \|\mathcal{V}_g \mathbf{x}_\lambda^*\|_\Omega^2$.
- We set $E = \|\mathcal{V}_g \mathbf{x}_\lambda^*\|_{\Omega'}^2$, from a region Ω' similar to Ω and not affected by the perturbation.

Outline

- 1 Filtering out time-frequency areas
- 2 Gabor transform -Gabor multiplier
- 3 Analytical solution
- 4 Numerical simulations**
 - Problem data
 - Comparative reconstruction results and SNRs
 - More experiments
- 5 Conclusion

Problem data

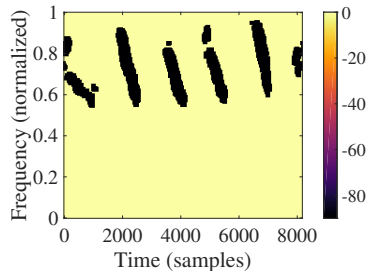
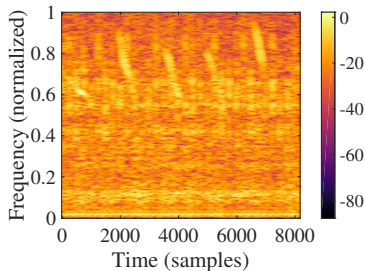
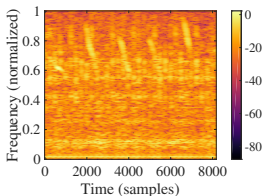


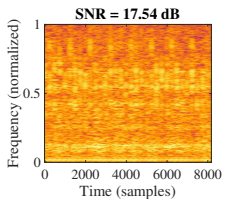
Figure: Observations (left), Mask (right)

- signal length: $L = 8192$
- sampling frequency: $f_s = 8000$ Hz
- \mathbf{x}_{ref} : wide-band engine sound and \mathbf{x}_{per} : chirp-like bird song
- \mathbf{g} : hanning window of length 128 (16ms) , hop size $a = 32$ (4ms) , and frequency shift $b = 512$ (16Hz).

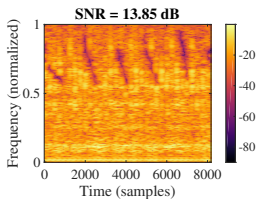
Comparative reconstruction results



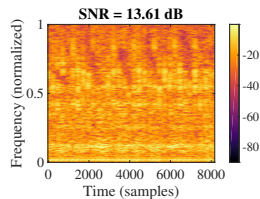
$$SNR(\mathbf{x}_{ref}, \mathbf{x}) = 20 \log_{10} \left(\frac{\|\mathbf{x}_{ref}\|}{\|\mathbf{x}_{ref} - \mathbf{x}\|} \right)$$



(a) RedEnerg



(b) ZerVal



(c) RandVal

SNRs for several targets and perturbations

	Target			perturbation
	car	train	aircraft	
RedEnerg	25.05	23.88	26.80	Beeps
ZerVal	24.23	16.52	24.73	
RandVal	24.90	16.15	24.73	
RedEnerg	18.88	22.49	20.72	Finger snap
ZerVal	17.18	20.70	16.59	
RandVal	16.92	16.17	15.93	
RedEnerg	21.67	20.97	17.96	Clicks
ZerVal	17.61	13.98	12.33	
RandVal	10.57	5.60	11.07	
RedEnerg	18.01	21.52	20.04	Birdsong
ZerVal	17.4	21.52	19.71	
RandVal	17.13	21.28	19.56	

Outline

- 1 Filtering out time-frequency areas
- 2 Gabor transform -Gabor multiplier
- 3 Analytical solution
- 4 Numerical simulations
- 5 Conclusion**

Conclusion

- We have addressed the problem of estimating a target signal, with no assumption on its contents, when perturbed by an additive signal that is well located in a region Ω of the TF plane.
- We have proposed an optimization problem in which the energy in Ω is controlled.
- We have used the nice properties of the Gabor multipliers for its analytical and numerical resolution.
- The proposed method outperforms some industrial baseline systems in terms of reconstruction SNR.
- Ongoing work: efficient algorithms for speeding up computations.

ama-marina.kreme@univ-amu.fr

Thank you for your attention!