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# Audio declipping

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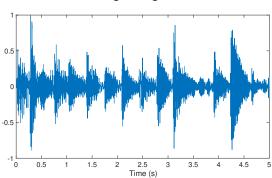
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## Audio Declipping

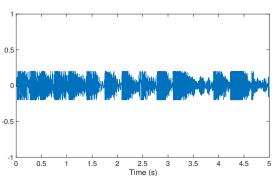




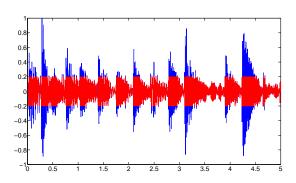
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# Audio Declipping





## Audio Declipping

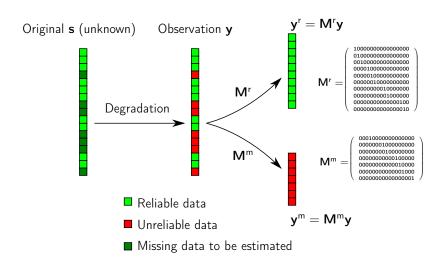


Goal:

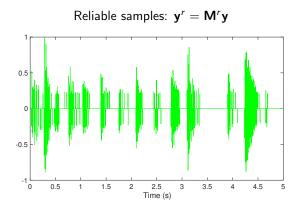
Can we get a good estimation of the original signal (blue) from the clipped one (red) ?

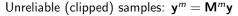
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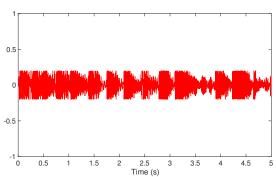
#### Reliable vs Unreliable coeff.



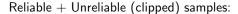
### Reliable vs Unreliable coeff.

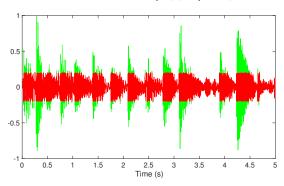






#### Reliable vs Unreliable coeff.





## Audio inpainting: forward problem [A. Adler, V. Emiya et A/]

We have then:

$$\mathbf{y}^r = \mathbf{M}^r \mathbf{y} = \mathbf{M}^r \mathbf{s}$$

where

- $\mathbf{s} \in \mathbb{R}^N$  is the unknown "clean" signal;
- $\mathbf{y}^r \in \mathbb{R}^M$  are the "reliable" sample of the observed signal
- $\mathbf{M}^r \in \mathbb{R}^{M \times N}$  is the matrix of the reliable support of  $\mathbf{x}$

we can also define the "missing" samples as

$$y^m = M^m y = M^m s$$

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## Inverse problem: data term

Using the reliable coefficients, we must have

$$\mathbf{y}^r = \mathbf{M}^r \mathbf{s}$$

where  $\mathbf{M}^r$  select the reliable samples. We can use a simple  $\ell_2$  loss

$$\hat{\mathbf{s}} = \operatorname*{argmin}_{\mathbf{s}} \frac{1}{2} \|\mathbf{y}^r - \mathbf{M}^r \mathbf{s}\|_2^2$$

We must take the clipped samples into account

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## Inverse problem: clipping constraints

For audio declipping, we can add the following constraint

$$\hat{\mathbf{s}} = \underset{\mathbf{s}}{\operatorname{argmin}} \frac{1}{2} \| \mathbf{y}' - \mathbf{M}' \mathbf{s} \|_{2}^{2}$$
  
s.t.  $\mathbf{M}^{m^{+}} \mathbf{\Phi} \alpha > \theta^{clip}$   
 $\mathbf{M}^{m^{-}} \mathbf{\Phi} \alpha < -\theta^{clip}$ 

#### where

- $\mathbf{M}^{m^+}$  (resp.  $\mathbf{M}^{m^-}$ ) select the positive (resp. negative) clipped samples.
- $\theta^{clip}$  is the clip threshold (here  $\theta^{clip} = 0.2$ )

Problem: infinite solutions! We must add some constraints on s

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## Audio declipping: use a dictionnary

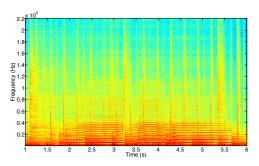
Let  $\Phi$  a dictionnary such that:

$$\mathsf{s} = \mathsf{\Phi} \alpha$$

where  $\alpha$  are **sparse** synthesis coefficients

Audio signal: use the short time Fourier transform

$$s(t) = \mathbf{\Phi} \boldsymbol{\alpha} = \sum_{n,f} \alpha_{n,f} \varphi_{n,f}(t)$$



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## Inverse problem: a constrained sparse problem

Using the dictionnary  $\Phi$  + sparsity

$$\begin{split} \hat{\boldsymbol{\alpha}} &= \operatorname*{argmin}_{\mathbf{s}} \frac{1}{2} \| \mathbf{y'} - \mathbf{M'} \mathbf{\Phi} \boldsymbol{\alpha} \|_2^2 + \lambda \| \boldsymbol{\alpha} \|_1 \\ \text{s.t.} \quad \mathbf{M}^{m^+} \mathbf{\Phi} \boldsymbol{\alpha} &> \theta^{clip} \\ \mathbf{M}^{m^-} \mathbf{\Phi} \boldsymbol{\alpha} &< -\theta^{clip} \end{split}$$

#### where

- $\mathbf{M}^{m^+}$  (resp.  $\mathbf{M}^{m^-}$ ) select the positive (resp. negative) clipped samples.
- $\theta^{clip}$  is the clip threshold (here  $\theta^{clip} = 0.2$ )
- $\hat{\mathbf{s}} = \hat{\alpha}$

#### Problems:

- the proximity operator has no closed form
- Cannot use simple algorithms such as (F)ISTA

### Rewrite the constraints

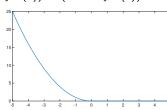
Idea: use a  $\ell_2$  loss on the clipped samples if the constraint is not respected

If 
$$y^m(t) > \theta^{clip}$$
  
then  $\mathcal{L}(\theta^{clip} - y^m(t)) = 0$ 

Inverse problem

If 
$$\hat{y}^m(t) < \theta^{clip}$$

else 
$$\mathcal{L}(\theta^{clip} - y^m(t)) = (\theta^{clip} - y^m(t))^2$$



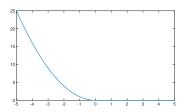
#### Rewrite the constraints

The squared hinge loss:

$$\mathcal{L}(\theta^{clip} - \mathbf{y}^m) = [\theta^{clip} - \mathbf{y}^m]_+^2$$

$$= \sum_{t:y^m(t)>0} (\theta^{clip} - y^m(t))_+^2 + \sum_{t:y^m(t)<0} (-\theta^{clip} + y^m(t))_+^2$$

$$= [\theta^{clip} - \mathbf{M}^m \mathbf{\Phi} \alpha]_+^2$$



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# Audio declipping: (convex unconstrained) inverse problem

We consider the following **unconstrained** convex problem:

$$\boldsymbol{\alpha} = \operatorname*{argmin}_{\boldsymbol{\alpha}} \frac{1}{2} \| \mathbf{y}^r - \mathbf{M}^r \mathbf{\Phi} \boldsymbol{\alpha} \|_2^2 + \frac{1}{2} [\boldsymbol{\theta}^{\textit{clip}} - \mathbf{M}^m \mathbf{\Phi} \boldsymbol{\alpha}]_+^2 + \lambda \| \boldsymbol{\alpha} \|_1$$

which is under the form

$$f_1(\alpha) + f_2(\alpha)$$

with  $f_1$  Lipschitz-differentiable and  $f_2$  semi-convex.

We can apply (relaxed)-ISTA directly!

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Inverse problem

# FISTA for declipping

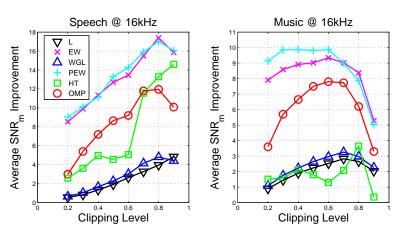
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Inverse problem

## Thresolding operators



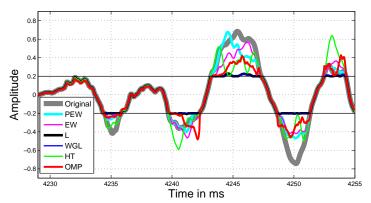
#### Numerical results



Average  $SNR_{miss}$  for 10 speech (left) and music (right) signals over different clipping levels and operators. Neighborhoods extend 3 and 7 coefficients in time for speech and music signals, respectively.

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#### Numerical results: zoom on reconstructions



Declipped music signal using different operators for clip level  $\theta^{clip}=0.2$  using the Lasso, WGL, EW, PEW, HT, and OMP operators. Neighborhood size for WGL and PEW was 7.

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# Original Vs clipped Vs declipped Signal

