

Introduction

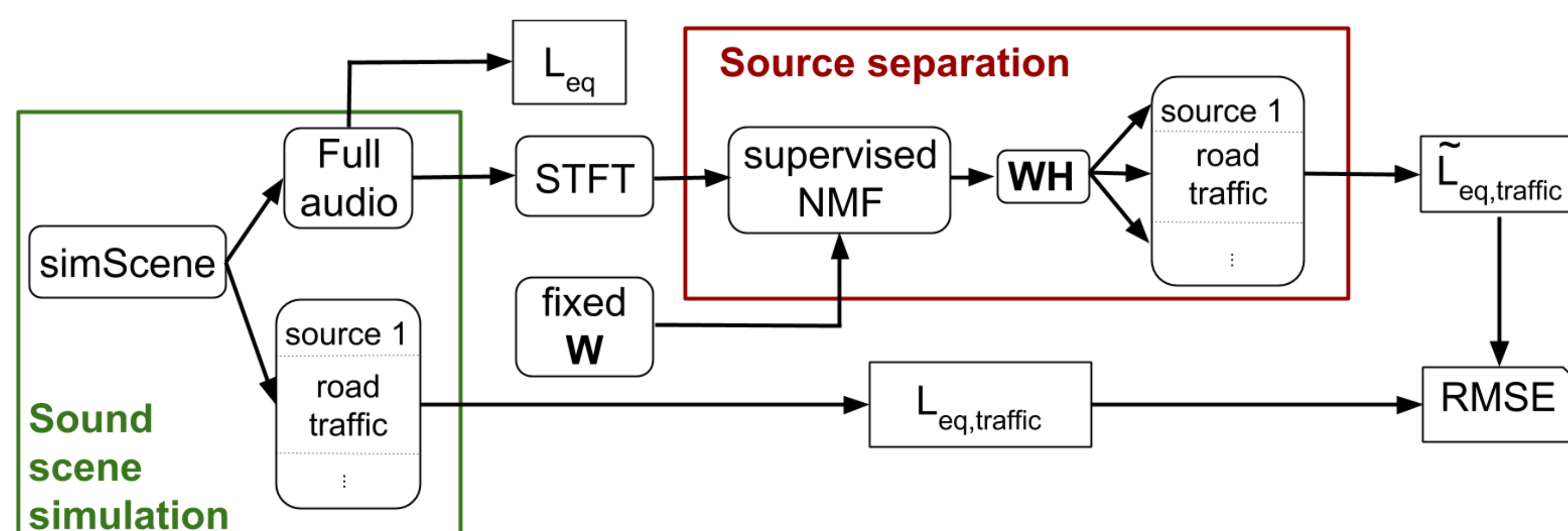
- **Context:** European Directive 2002/49/EC imposes to cities over 100 000 inhabitants to realize road traffic noise maps
- Aim to improve road traffic simulated noise maps by means of with acoustic measurements

=> **Need for a method that separates road traffic from the other existing sound sources within measurements**

- **Approach:** Usual methods for urban sound environments separation do not deal with overlapping sounds
- The NMF seems to suit urban sound mixture requirements but it is used yet only for audio or musical applications

=> **Adapt and test the NMF on a set of simulated urban sound mixtures**

Method



Sound scene simulation

Simulation of a set of sound scenes with the *simScene*^a software:

- Realization of sound mixtures from a database of isolated sounds
- Control of the event/background ratio
- Control of the time of presence of each sound source

It gives as output:

- A complete audio file with all the mixed sound sources and an audio file for each sound class,
- => Creation of specific urban sound environments (animated or quiet streets, parks...)

Source separation

Supervised Non-negative Matrix Factorization:

$$V \approx WH$$

- $V_{F \times N} \in \mathbb{R}^+$, the audio spectrogram of the sound mixture,
- $W_{F \times K} \in \mathbb{R}^+$, fixed dictionary composed of a set of K audio spectra of urban sound sources
- $H_{K \times N} \in \mathbb{R}^+$, the feature matrix standing for the temporal variation of the spectrum found iteratively by the minimisation of the cost function

$$\min D(V||WH) \quad \text{w.r.t. } W > 0, H > 0$$

3 tested cost functions :

- The Euclidian distance (Euc. Dist)
- The Kullback-Leibler divergence (K-L div)
- The Itakura –Saito divergence (I-S div)

- W is composed of 3 classes (*car*, *bird*, *horn*)
- H is updated with the maximisation-minimisation algorithm^b

^aM. Rossignol & al., *simScene: a web-based acoustic scenes simulator*, 1st Web Audio Conference, 2015

^bC. Févotte & J. Idier, *Algorithms for nonnegative matrix factorization with the β -divergence*, Neural Computation, vol 23, no 9, 2011

Experiment and first results

- 20 simple scenes composed of 3 sound classes (car, horn, bird)
- W is composed of theses 3 classes
- NMF is performed with 100 iterations

=> $L_{eq,traffic}$ and $\tilde{L}_{eq,traffic}$ estimated for each scene

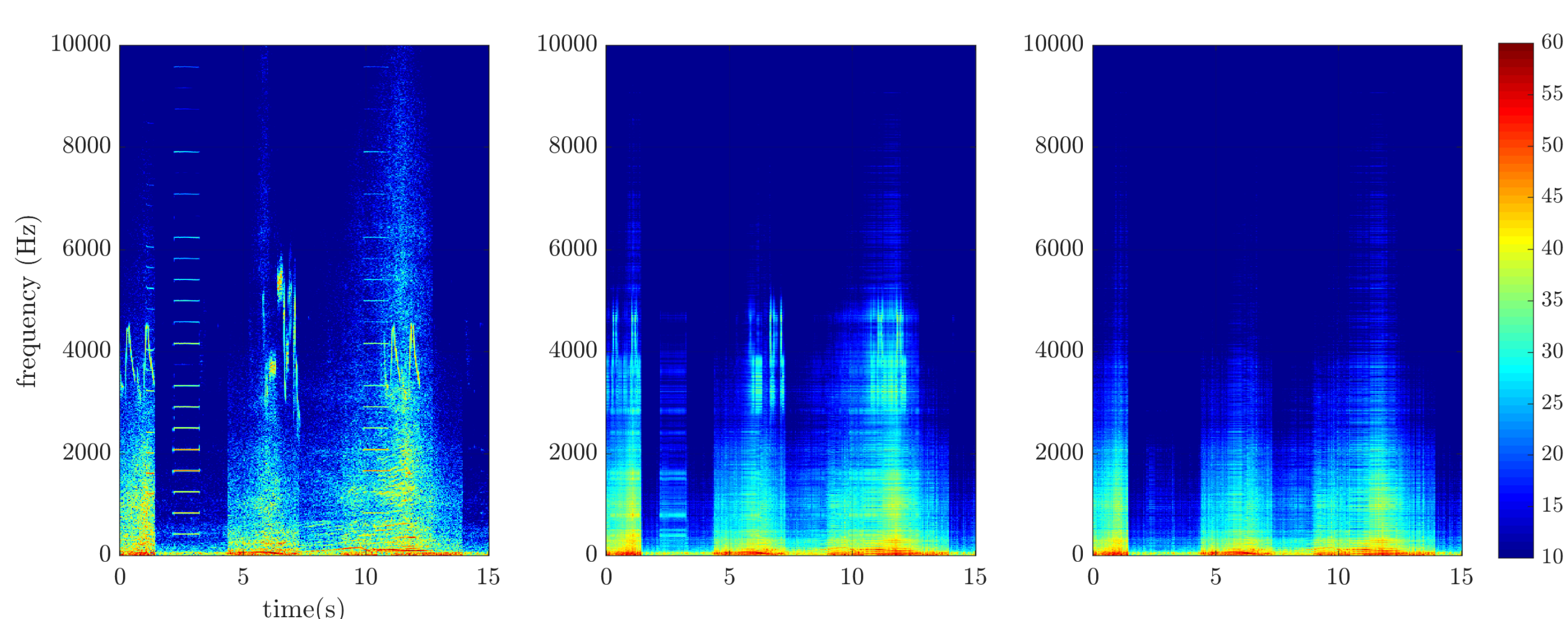


Fig 1: on the left the spectrogram of a scene created with *simScene*, in the middle the estimated spectrogram after 100 iterations, on the right the road traffic spectrogram estimated after source separation

- Estimation of the performance of the implemented NMF by the RMSE depending on number of iterations and the cost function:

$$RMSE = \sqrt{\frac{1}{N} \sum_{i=1}^{20} (L_{eq,traffic}^i - \tilde{L}_{eq,traffic}^i)^2}$$

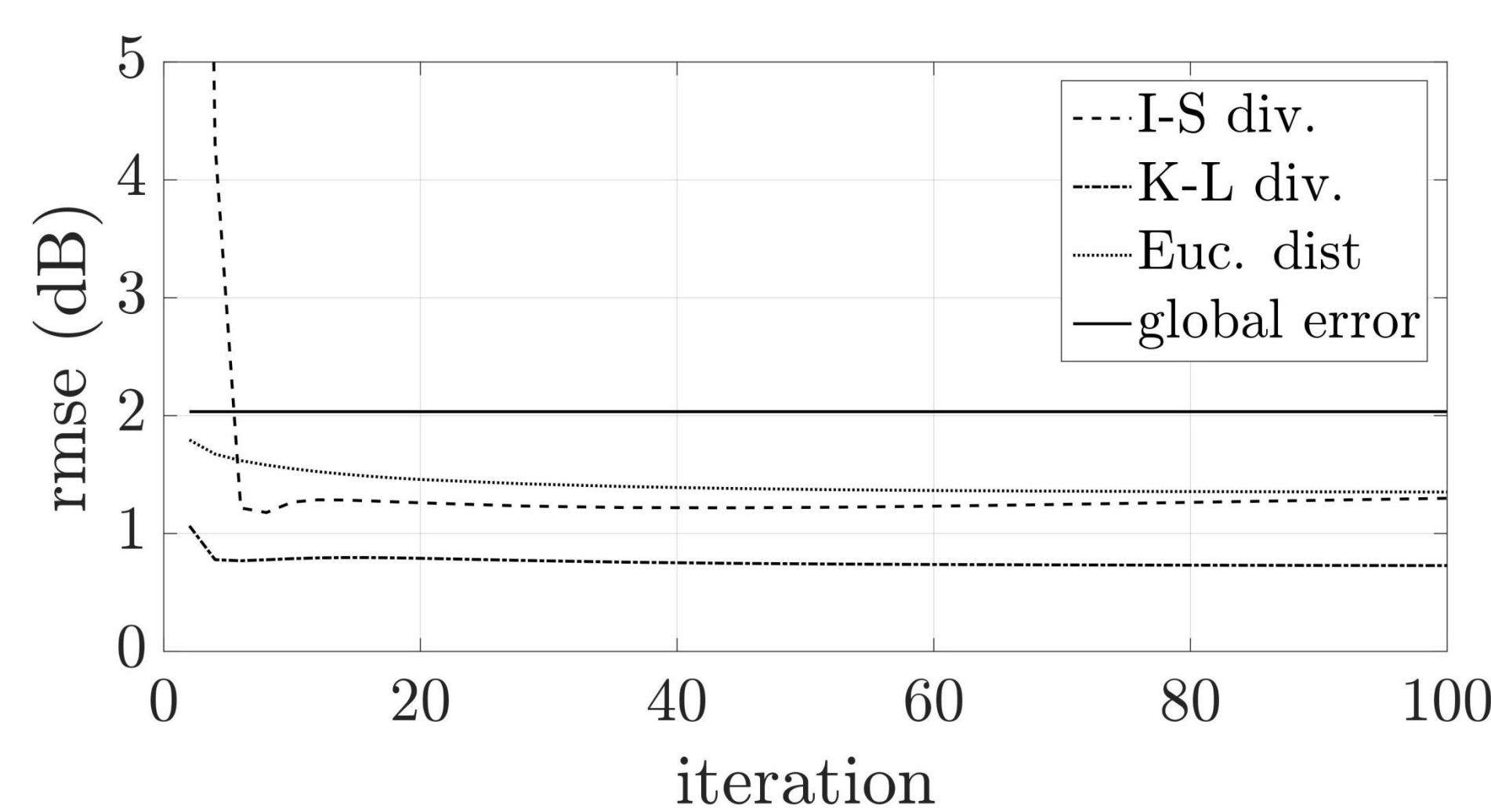


Fig 2: comparison of the RMSE for the 3 cost functions. We add the RMSE between L_{eq} and $L_{eq,traffic}$ without source separation

Conclusion and investigations

- First results for the road traffic source separation in an sound urban context within simple simulated sound scene
- Low RMSE values and improvement of the estimated road traffic sound levels show the interest of the NMF in that context
- The K-L divergence seems the most promising approach

Future works:

- Building a set of more complete and realistic sound scenes
- Design an experimental plan in order to optimize the modeling parameters
- Add temporal constraints on H to modelize more realistic comportements (smooth NMF)