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# Audio Inpainting, Source Separation, Audio Compression All with a Unified Framework Based on NTF Model

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**Keywords:** Audio Inpainting, Source Separation, Informed Source Separation (ISS), Compressive Sampling, Nonnegative Tensor Factorization (NTF)

We propose a general signal recovery algorithm, that can recover  $J$  audio source signals in time,  $s_{jt}^\tau, j \in \llbracket 1, J \rrbracket, t \in \llbracket 1, T \rrbracket$ , given the quantized source samples,  $y_{jt}^\tau$ , each observed on the support  $\Omega_j^\tau$  and the quantized mixture samples,  $x_t^\tau = \sum_{j=1}^J s_{jt}^\tau + a_t^\tau$ , observed on  $\Xi^\tau$  with quantization noise  $a_t^\tau$ . The sources are modelled in the short time Fourier transform (STFT) domain with a normal distribution ( $s_{jfn} \sim \mathcal{N}_c(0, v_{jfn})$ ) where the variance tensor  $\mathbf{V} = [v_{jfn}]_{j,f,n}$  has the following low-rank non-negative tensor factorization (NTF) structure [7],  $v_{jfn} = \sum_{k=1}^K q_{jk} w_{fk} h_{nk}$ . This model is parametrized by  $\theta = \{\mathbf{Q}, \mathbf{W}, \mathbf{H}\}$ , with  $\mathbf{Q} = [q_{jk}]_{j,k} \in \mathbb{R}_+^{J \times K}$ ,  $\mathbf{W} = [w_{fk}]_{f,k} \in \mathbb{R}_+^{F \times K}$  and  $\mathbf{H} = [h_{nk}]_{n,k} \in \mathbb{R}_+^{N \times K}$ .

We propose to recover the source signals with a generalized expectation-maximization (GEM) algorithm [4] based on multiplicative update (MU) rules [5]. The Algorithm is briefly described in Algorithm 1. Using the proposed approach, it is possible to solve a number of existing and new problems in audio signal processing:

- **Audio inpainting:** It is possible to recover arbitrary time domain losses in audio signals for applications such as signal declipping. NTF

model is used for the *first time* for the recovery of arbitrary time domain losses [3].

- **Joint audio inpainting and source separation:** It is possible to jointly perform audio inpainting and source separation to improve the performance of both tasks. Audio inpainting and source separation are performed jointly for the *first time* [1].
- **Compressed sampling-based informed source separation:** It is possible to recover the sources from their random samples and the mixture via compressive sampling-based informed source separation. This new ISS scheme uses a simple encoder that has properties of distributed coding [8, 6] and it competes with traditional ISS. The concept of distributed coding and of compressive sampling based scheme is introduced for the *first time* in the informed source separation problem [2].

The presentation and the poster will include various new results for the proposed algorithm with comparisons to state of the art methods in the different problems discussed above.

\* The first and second authors have contributed equally for this work.

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**Algorithm 1** GEM algorithm for NTF model estimation. Details can be found in [1, 2, 3]

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- 1: **procedure** ESTIMATESOURCES-NTF( $\Xi^\tau, x^\tau, \{y_{jt}^\tau\}_{j=1}^J, \{\Omega_j^\tau\}_{j=1}^J$ )
  - 2:     Initialize non-negative  $\mathbf{Q}, \mathbf{W}, \mathbf{H}$  randomly and apply any known prior on  $\mathbf{Q}, \mathbf{W}, \mathbf{H}$
  - 3:     **repeat**
  - 4:         Estimate sources ( $s_{jfn}$ ), given  $\mathbf{Q}, \mathbf{W}, \mathbf{H}, x^\tau, \Xi^\tau, \{y_{jt}^\tau\}_{j=1}^J, \{\Omega_j^\tau\}_{j=1}^J$  with Wiener filtering
  - 5:         Apply constraints in the time domain, and estimate posterior power spectra of the sources ( $\tilde{\mathbf{P}}$ )
  - 6:         Update  $\mathbf{Q}, \mathbf{W}, \mathbf{H}$  given  $\tilde{\mathbf{P}}$  using MU rules
  - 7:     **until** convergence criteria met
  - 8: **end procedure**
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## References

- [1] Bilen, Ç., A. Ozerov, and P. Pérez (2015a, October). Audio declipping via nonnegative matrix factorization. In *IEEE Workshop on Applications of Signal Processing to Audio and Acoustics*.
- [2] Bilen, Ç., A. Ozerov, and P. Pérez (2015b, October). Compressive sampling-based informed source separation. In *IEEE Workshop on Applications of Signal Processing to Audio and Acoustics*.
- [3] Bilen, Ç., A. Ozerov, and P. Pérez (2015c, August). Joint audio inpainting and source separation. In *International Conference on Latent Variable Analysis and Signal Separation*.
- [4] Dempster, A., N. Laird, and D. Rubin. (1977). Maximum likelihood from incomplete data via the EM algorithm. *Journal of the Royal Statistical Society. Series B (Methodological)* 39, 1–38.
- [5] Févotte, C., N. Bertin, and J. Durrieu (2009, March). Nonnegative matrix factorization with the Itakura-Saito divergence. With application to music analysis. *Neural Computation* 21(3), 793–830.
- [6] Girod, B., A. Aaron, S. Rane, and D. Rebollo-Monedero (2005, January). Distributed video coding. *Proceedings of the IEEE* 93(1), 71 – 83.
- [7] Ozerov, A., C. Févotte, R. Blouet, and J.-L. Durrieu (2011, May). Multichannel nonnegative tensor factorization with structured constraints for user-guided audio source separation. In *IEEE International Conference on Acoustics, Speech, and Signal Processing (ICASSP'11)*, Prague, pp. 257–260.
- [8] Xiong, Z., A. Liveris, and S. Cheng (2004, September). Distributed source coding for sensor networks. *IEEE Signal Processing Magazine* 21(5), 80–94.