

# VERY LOW BITRATE SPATIAL AUDIO CODING WITH DIMENSIONALITY REDUCTION



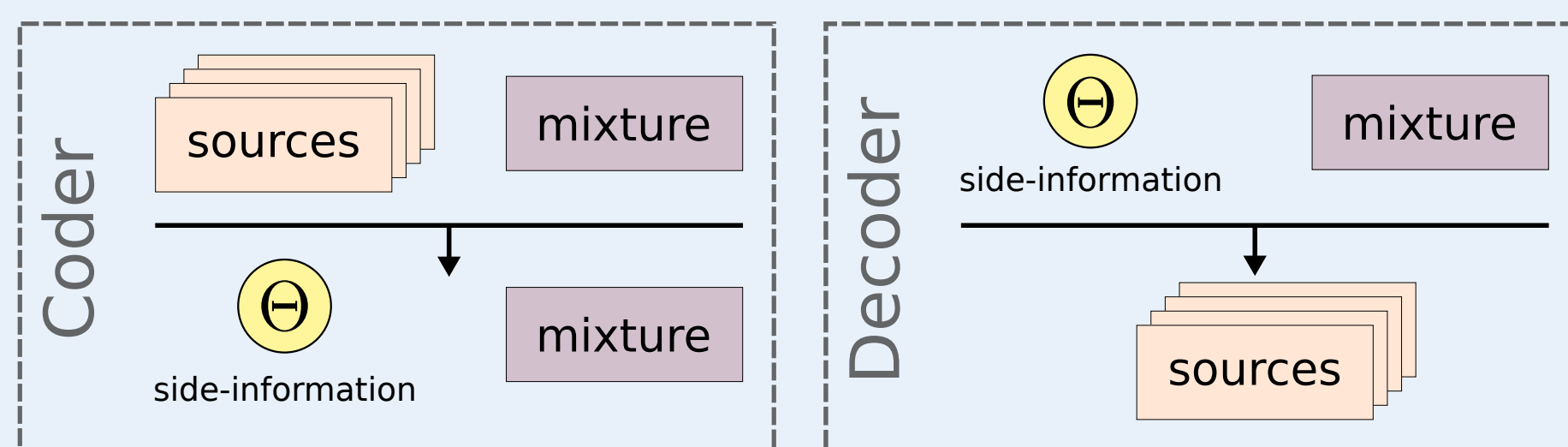
Christian Rohlfing<sup>1</sup>, Jeremy E. Cohen<sup>2</sup>, Antoine Liutkus<sup>3</sup>

<sup>1</sup>Institut für Nachrichtentechnik, RWTH Aachen University

<sup>2</sup>Department of Image and Signal-processing, Gipsa-lab, CNRS, Grenoble, France

<sup>3</sup>Inria, speech processing team, Villers-lès-Nancy, France

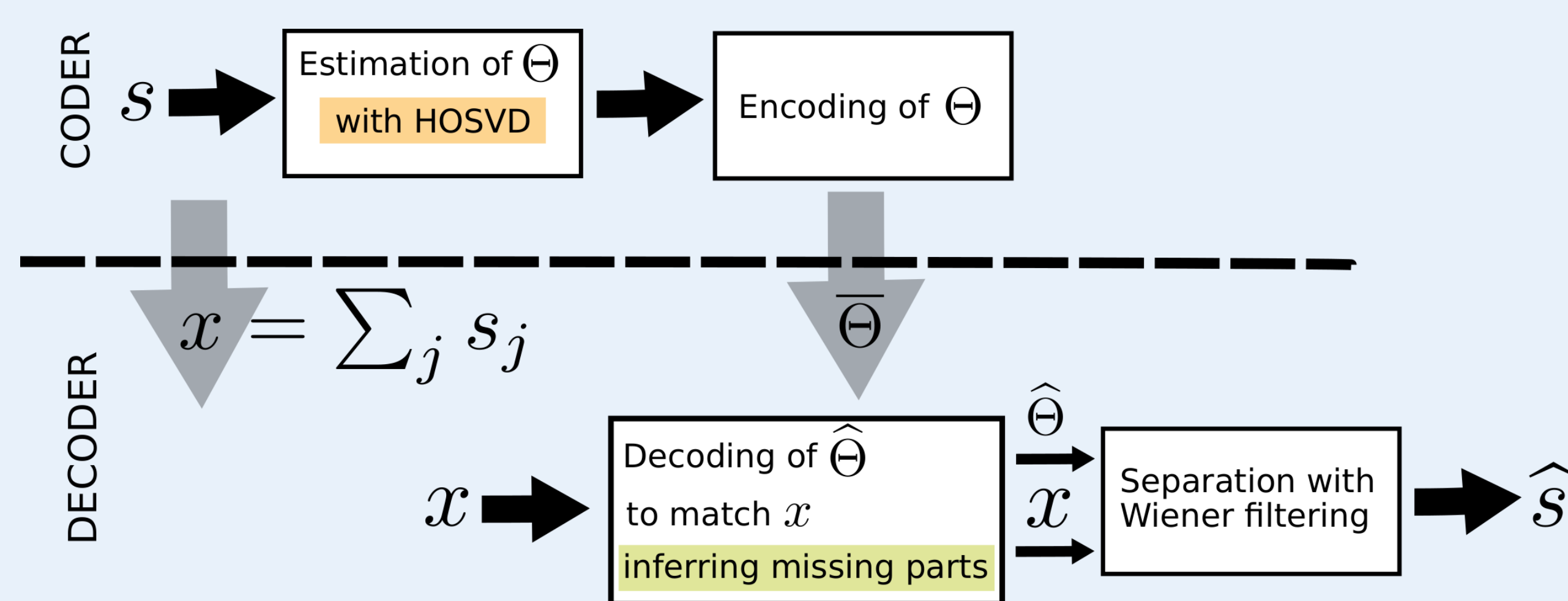
## Introduction



Reference ISS method [1] consists of two steps:

- **Coder:** Sources and mix perfectly known. Nonnegative Tensor Factorization (NTF) of source spectrograms with NTF. Encoding of resulting parameters  $\Theta$ .
- **Decoder:** Only mix available. Wiener filtering used to recover sources from mix assisted by transmitted NTF parameters  $\Theta$ .

## Dimension reduction for ISS



### Restriction of classical NTF-based ISS methods:

Temporal dimension of source spectrogram can be very large (e.g. for full-length tracks).

⇒ Growth of temporal dimension of NTF parameters (bitrate) and computational complexity of NTF.

### Proposed solution:

- Replace NTF with Higher-order SVD in coder.
  - Do not send temporal parameter but estimate it at decoder.
- ⇒ Size of side-information independent of track length!

## Higher-order SVD at coder

Replace classical NTF with higher-order SVD model:

$$\mathcal{V}_s = (\mathbf{U}_F \otimes \mathbf{U}_T \otimes \mathbf{U}_J) \mathcal{G}$$

**Randomized SVD** reduces memory costs and numerical complexity:  $\mathcal{O}(FT(R_F + R_T))$  (vs. classical NTF with  $R$  components:  $\mathcal{O}(FTJR)$ )

**Transmitted parameters** do *not* include  $\mathbf{U}_T$  but only  $\mathbf{U}_F$ ,  $\mathbf{U}_J$  and  $\mathcal{G}$ .

## Parameter estimation at decoder

Model of mixture  $\mathcal{V}_x \approx (\mathbf{U}_F \otimes \mathbf{U}_T \otimes \mathbf{u}_J) \mathcal{G}$  with  $\mathbf{u}_J = \sum_{j=1}^J [\mathbf{U}_J]_j$

### Temporal parameter estimation:

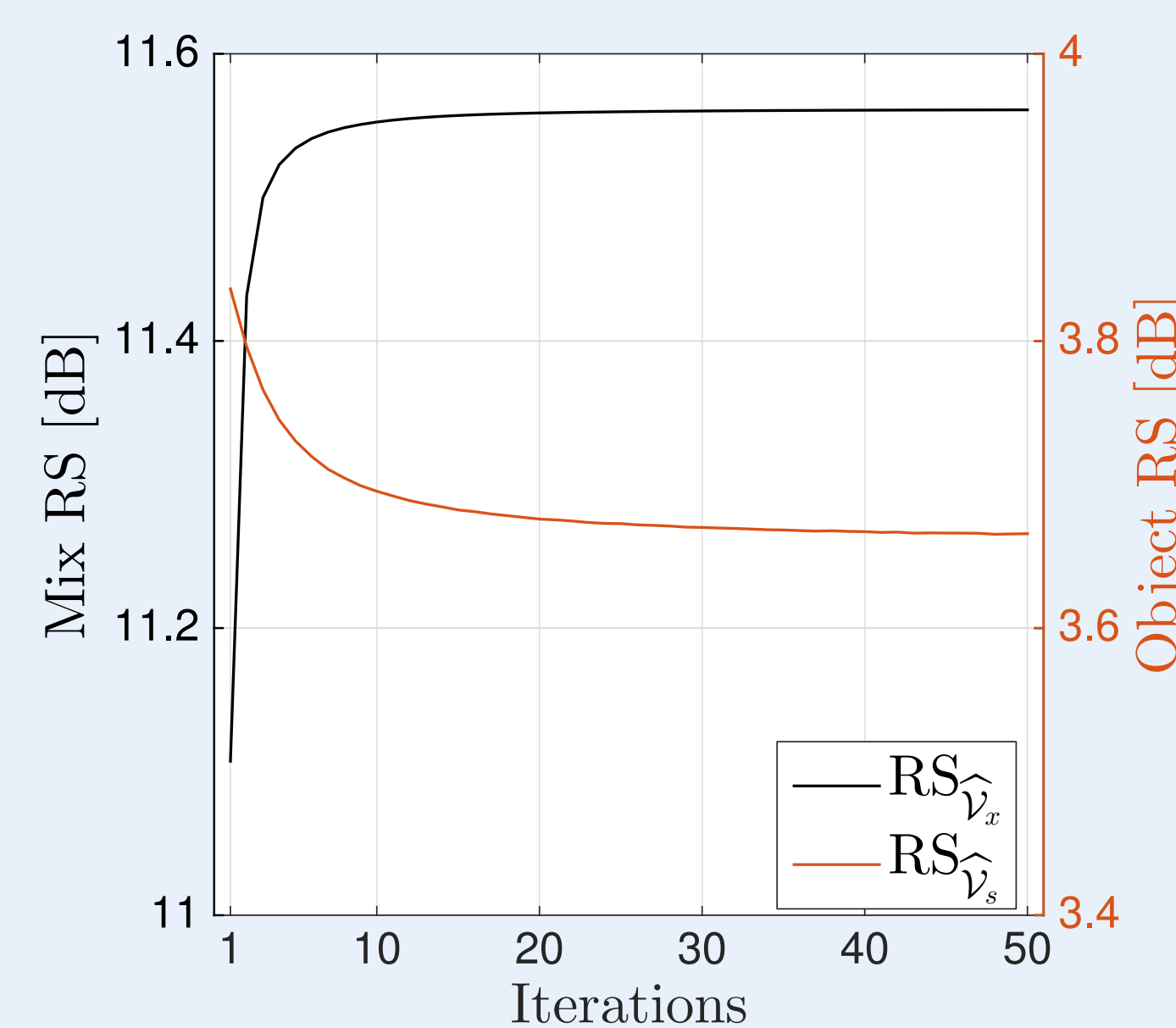
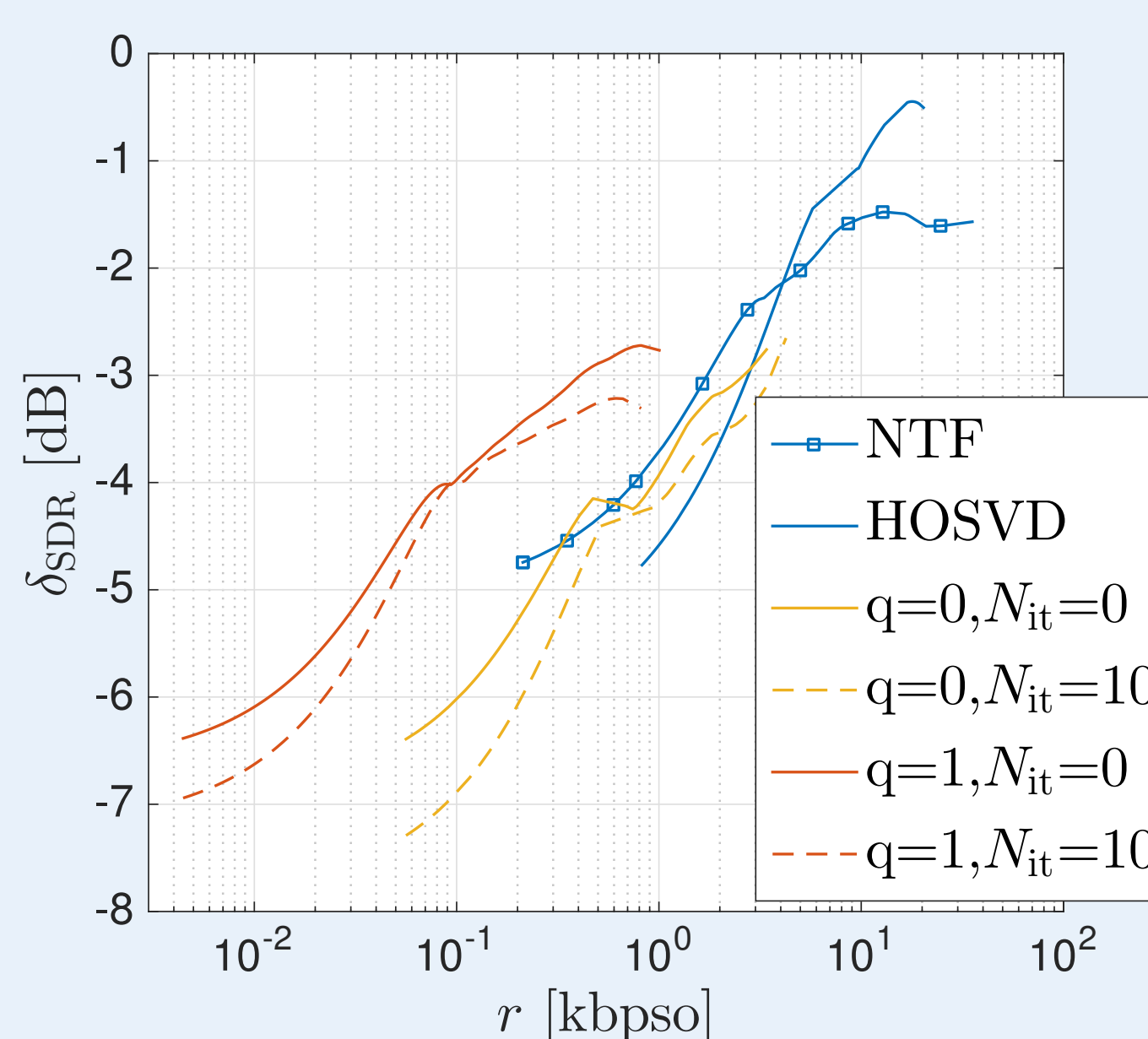
Temporal matrix  $\mathbf{U}_T$  is not transmitted. Estimate  $\mathbf{U}_T$  given the other transmitted parameters  $\bar{\Theta}$ .

$$\hat{\mathbf{U}}_T \leftarrow \underset{\mathbf{U}_T^T \mathbf{U}_T = \mathbf{I}}{\operatorname{argmin}} \|\mathcal{V}_x - \mathbf{N}_T \mathbf{U}_T^T\|_F^2$$

with  $\mathbf{N}_T = \mathbf{U}_F (\mathbf{u}_J \bullet_3 \mathcal{G})$

**Iterative re-estimation** of parameters  $\mathbf{U}_F$  and  $\mathcal{G}$  initialized with their quantized value obtained from coder.

## Experiments



### Reference methods:

Performance of NTF method [1] as well as plain HOSVD without quantization and with full transmission of all HOSVD parameters.

### Setup:

- Ten full-length tracks (4 sources, taken from DSD100 database)
- HOSVD operates with  $R_F, R_T \in [5, 100]$  and  $R_J = 4$
- Quantization disabled ( $q = 0$ ) or enabled ( $q = 1$ ) with  $N_{\mathcal{G}}, N_F \in [5, 1000]$  centroids for quantization of  $\mathcal{G}$  and  $\mathbf{U}_F$
- Number of iterations for re-estimation  $N_{it} \in \{0, 10\}$
- Quality measure:  $\delta_{\text{SDR}}$  over parameter bitrate  $r$  and reconstruction scores  $\text{RS}_{\hat{\mathcal{V}}_x} = 10 \log(\|\mathcal{V}_x\|_F^2 / \|\mathcal{V}_x - \hat{\mathcal{V}}_x\|_F^2)$  and  $\text{RS}_{\hat{\mathcal{V}}_s}$  over iterations

### Results:

- NTF (—) yields performance better than HOSVD (—) with full resolution: it has less parameters
- Not sending  $\mathbf{U}_T$  but estimating it at decoder, with other parameters full resolution (—), permits to reach NTF performance (—)
- Coarse quantization of  $\mathbf{U}_F$  and  $\mathcal{G}$  and not sending  $\mathbf{U}_T$  (—) leads to remarkable decrease of bitrate by a factor of almost 10.
- Iterative re-estimation of quantized parameters (—, —) is not increasing performance. Score  $\text{RS}_{\hat{\mathcal{V}}_x}$  (—) is improved by 0.5 dB. Generalization to  $\hat{\mathcal{V}}_s$  does not hold.

[1] A. Liutkus, J. Pinel, R. Badeau, L. Girin, and G. Richard, "Informed source separation through spectrogram coding and data embedding," *Signal Processing*, vol. 92, no. 8, pp. 1937 – 1949, 2012.

