

EFFICIENT MULTICHANNEL NONLINEAR ACOUSTIC ECHO CANCELLATION BASED ON A COOPERATIVE STRATEGY

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1. Outline

Goal

An efficient approach for Multichannel Nonlinear Acoustic Echo Cancellation (NAEC)

Challenges

- Simultaneously excited nonlinearities
- High dimensionality of the problem
- High computational cost

Proposed Method:

Cooperative Multichannel AEC (CM-AEC)

- Decomposes the problem into multiple smaller problems
- Allows different particle filters to share their state
- Efficient model by coupling microphones
- Two variants are studied and realized

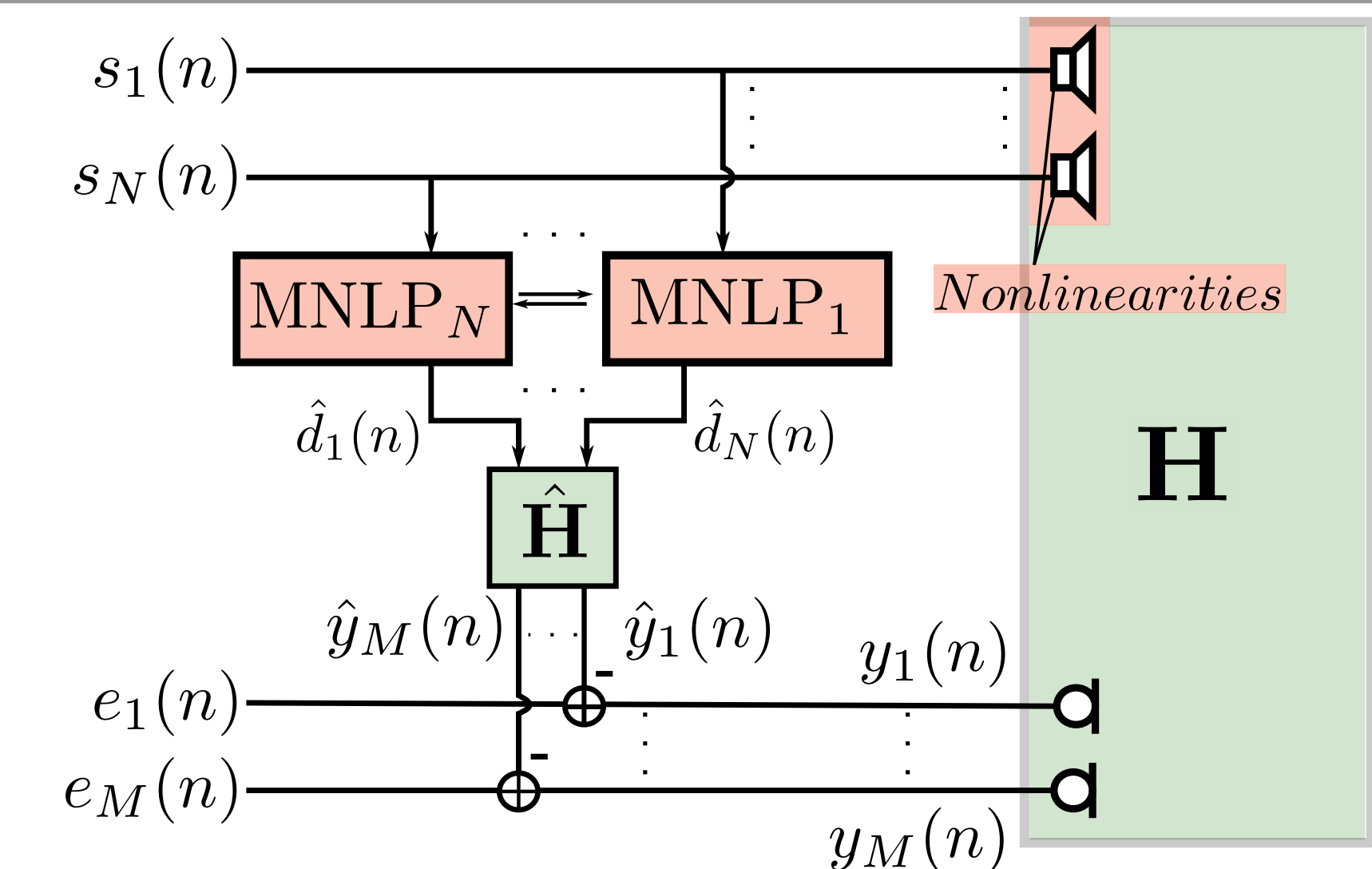
2. Problem Formulation

Echo paths

- Each loudspeaker introduces nonlinear distortions to the far-end signal $\{s_i\}_{i=1}^N$
- Linear transmission characteristics from the N loudspeakers to the M microphones are represented by \mathbf{H}

Echo paths approximation

- Each nonlinearity is approximated by a Memoryless Nonlinear Preprocessor (MNLP) $\hat{d}_i(n) = \sum_{l=1}^{L_a} \hat{a}_{i,l} \cdot f_{i,l}(s_i(n)) = \hat{\mathbf{a}}_i^T \mathbf{f}_i(s_i(n))$
- \mathbf{H} is approximated by block matrix $\hat{\mathbf{H}}$



3. The Cooperative Multichannel AEC

Identification of The Nonlinear Part

- Estimating the matrix $\hat{\mathbf{A}} = [\mathbf{a}_1, \dots, \mathbf{a}_N] =$

$$\hat{\mathbf{A}} = [\mathbf{a}_1, \dots, \mathbf{a}_N] = \begin{bmatrix} \hat{a}_{1,1} & \dots & \hat{a}_{N,1} \\ \vdots & \ddots & \vdots \\ \hat{a}_{1,L_a} & \dots & \hat{a}_{N,L_a} \end{bmatrix}$$

via submatrices

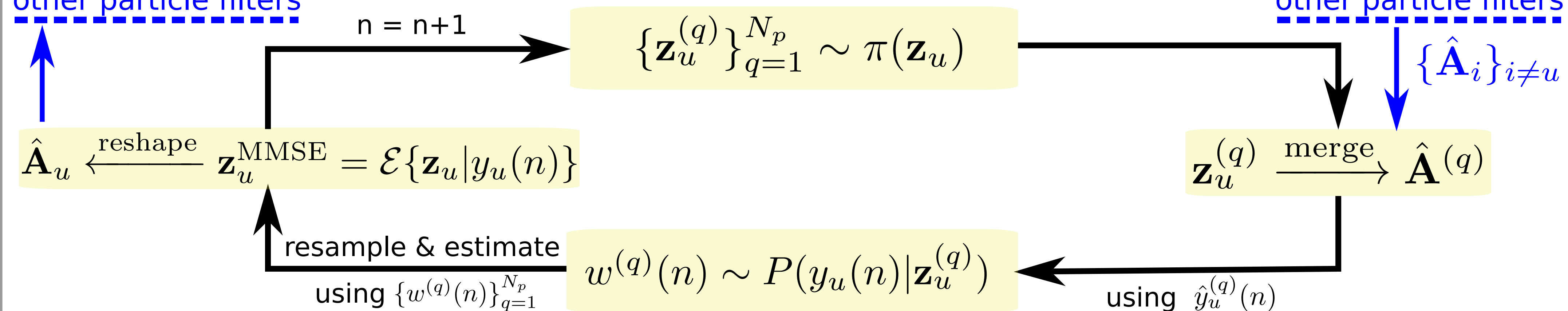
$$\hat{\mathbf{A}}_1 = \begin{bmatrix} \hat{a}_{1,1} & \dots & \hat{a}_{1,L_a} \\ \vdots & \ddots & \vdots \\ \hat{a}_{1,L_a} & \dots & \hat{a}_{1,L_a} \end{bmatrix}$$

$$\hat{\mathbf{A}}_U = \begin{bmatrix} \hat{a}_{U,1} & \dots & \hat{a}_{U,L_a} \\ \vdots & \ddots & \vdots \\ \hat{a}_{U,L_a} & \dots & \hat{a}_{U,L_a} \end{bmatrix}$$

Estimating $\hat{\mathbf{A}}_u, u \in \{1, \dots, U\}$

- $\hat{\mathbf{A}}_u$ is arranged as $\mathbf{z}_u = [\hat{\mathbf{a}}_{u,1}^T, \dots, \hat{\mathbf{a}}_{u,L_a}^T]^T$ and adapted by a particle filter using $y_u(n)$ as follows:

other particle filters



Two variants • CM Gaussian Particle Filter (CM-GPF) [1] • CM Elitist Resampling Particle Filter (CM-ERPF) [2]

Identification of The Linear Part

- The block matrix $\hat{\mathbf{H}}$ is adapted by the Generalized Frequency Domain Adaptive Filtering (GFDAF) algorithm [3]

4. Evaluation

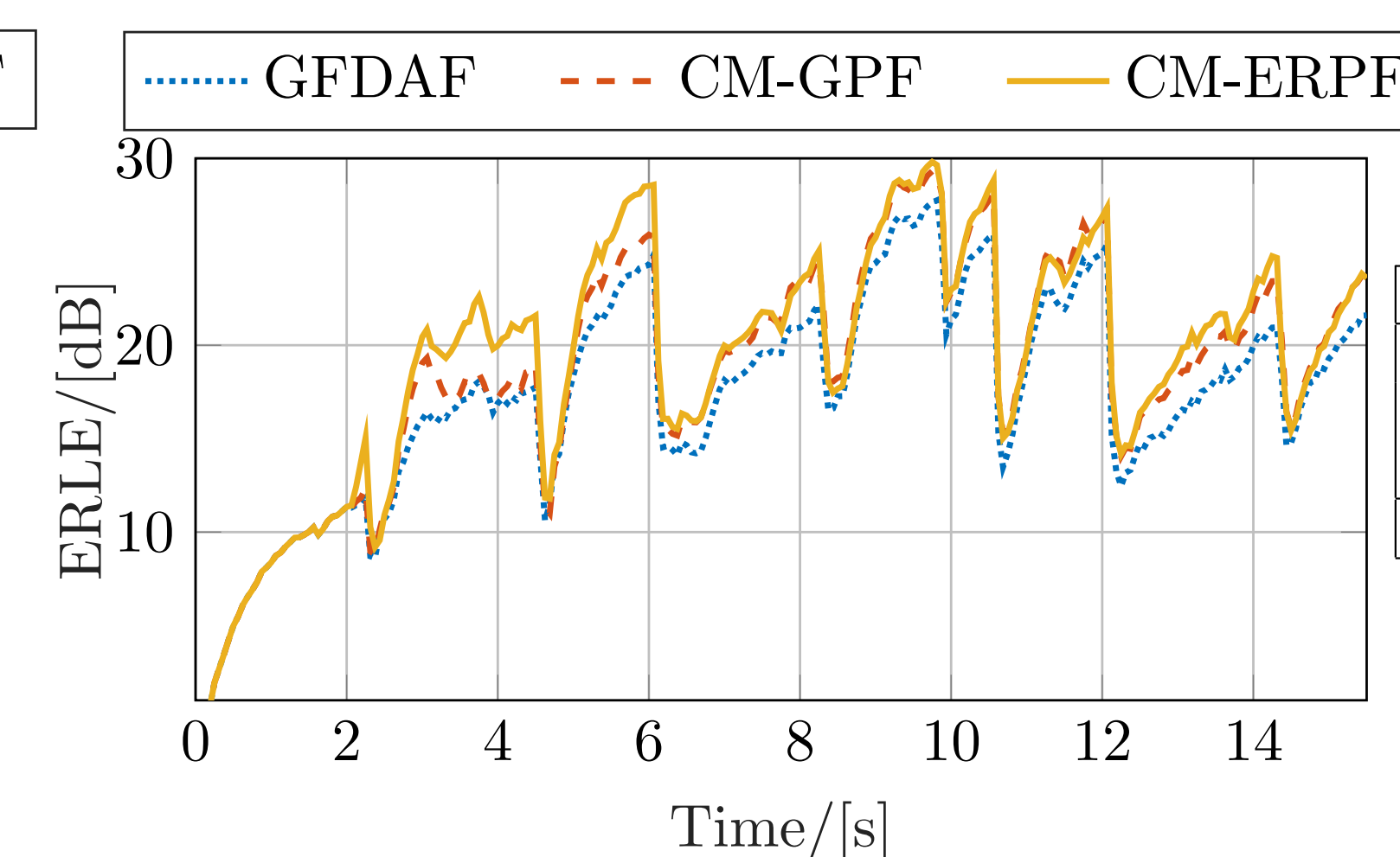
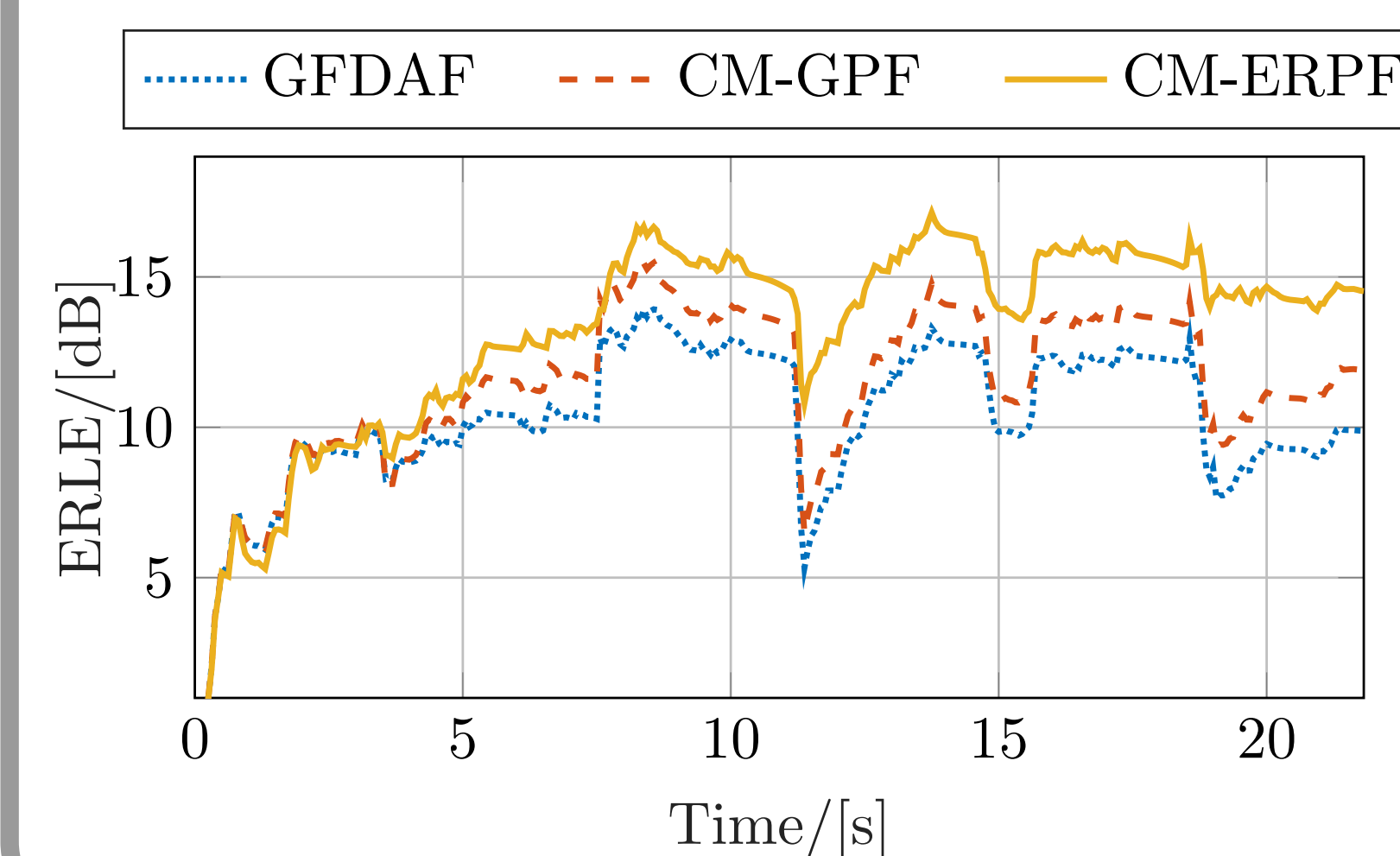
Synthesized nonlinear distortions

- 5 loudspeakers, 2 microphones
- Simulated soft-clipping nonlinearity
- Recorded impulse response with $T_{60} = 450\text{ms}$

Real nonlinear distortions

- Smart speaker prototype: 8 loudspeakers, 7 microphones
- Recording in a room with $T_{60} = 550\text{ms}$

Evaluation measure: $\text{ERLE}(n) = 10 \log_{10} \left(\frac{1}{M} \sum_{j=1}^M \frac{E\{y_j^2(n)\}}{E\{e_j^2(n)\}} \right)$



| | | Average ERLE [dB] | | |
|------|----------|-------------------|------|------|
| | SNR [dB] | GFDAF | GPF | ERPF |
| Syn | 30 | 10.4 | 11.5 | 13.4 |
| | 20 | 9.8 | 10.9 | 11.8 |
| Real | 10 | 7.1 | 7.4 | 8.1 |
| | 30 | 18.1 | 19.0 | 20.2 |

5. Conclusions

The CM-AEC approach

- is a new particle filter-based MIMO NLAEC method
- models the simultaneously excited nonlinearities using an efficient model
- was verified for both synthesized and real distortions using two different variants

References

- [1] J. Kotecha and P. Djuric, "Gaussian particle filtering," IEEE Trans. Signal Process., vol. 51, no. 10, pp. 2592–2601, Oct. 2003.
- [2] M. M. Halimeh et al., "Nonlinear acoustic echo cancellation using elitist resampling particle filter," in ICASSP, April 2018, pp. 236–240.
- [3] M. Schneider and W. Kellermann, "The generalized frequency-domain adaptive filtering algorithm as an approximation of the block recursive least-squares algorithm", EURASIP Journal on Advances in Signal Processing, 2016.