

Kernel-Interpolation-based Filtered-x Least Mean Square for Spatial Active Noise Control in Time Domain

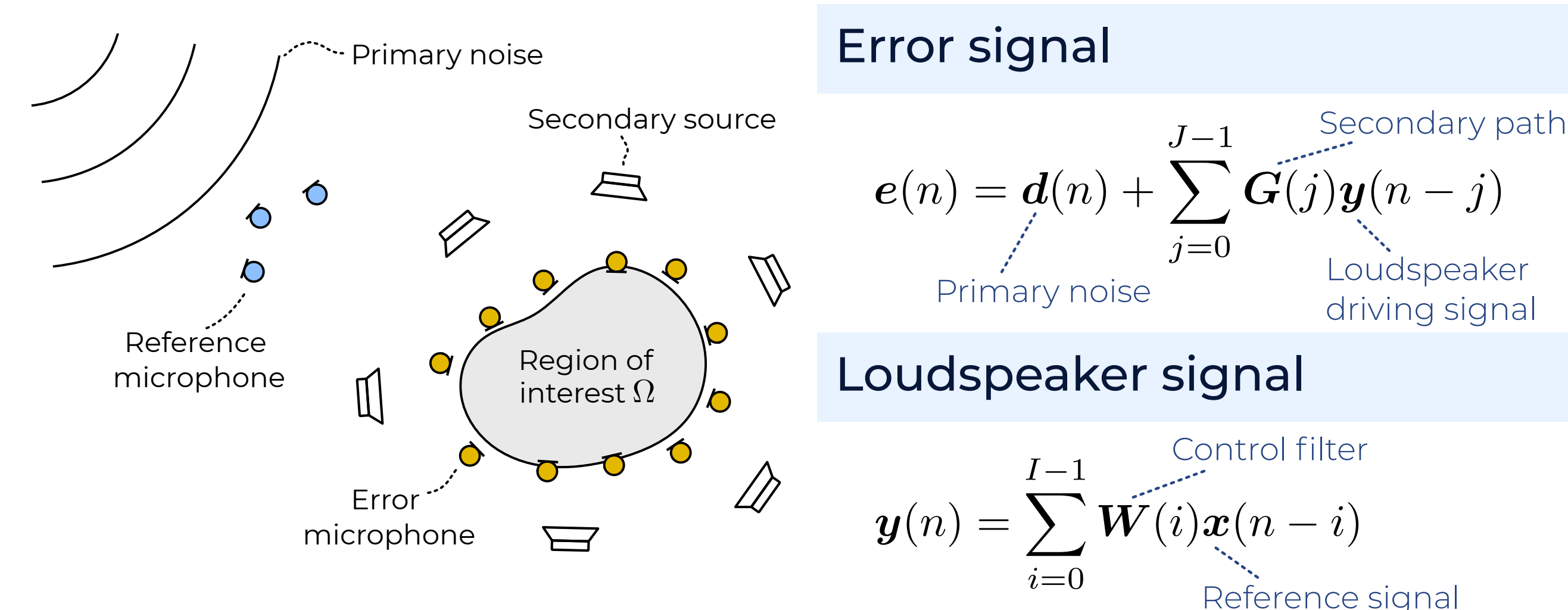
Jesper Brunnström
Shoichi Koyama
The University of Tokyo

Abstract

- A time-domain spatial active noise control algorithm for broadband noise using kernel interpolation of sound field is proposed.
- Proposed method outperform conventional multipoint pressure control with regards to regional power reduction.
- Small to no increase in computational cost compared to conventional FxLMS.
- A fast block implementation is proposed, significantly reducing computational cost at a minor performance cost.

Problem statement

Goal: Broadband noise reduction over a continuous region with flexible array placement and low computational cost.



Cost function: Total sound power within region of interest

$$\mathcal{L} = \mathbb{E} \left[\int_{\Omega} u(n, \mathbf{r})^2 d\mathbf{r} \right]$$

Sound pressure

Kernel Interpolation

- Estimate sound pressure function from error microphone measurements.
- Optimal interpolation filter is given in closed form in frequency domain.
- Obtain time domain filter through inverse Fourier transform.

$$u(n, \mathbf{r}) = \sum_{i=-\infty}^{\infty} \mathbf{z}^T(i, \mathbf{r}) e(n-i)$$

$$\mathbf{z}(i, \mathbf{r}) = \mathcal{F}^{-1} \left[\mathbf{z}(\omega, \mathbf{r}) \right] (i) = \mathcal{F}^{-1} \left[[(\mathbf{K}(\omega) + \lambda \mathbf{I})^{-1}]^T \boldsymbol{\kappa}(\omega, \mathbf{r}) \right] (i)$$

- Interpolation filter is determined by the kernel function.
- This kernel function imposes requirement that pressure function must satisfy Helmholtz equation.
- Only positions of microphones and region of interest is required to compute the filter.

$$\kappa(\omega, \mathbf{r}, \mathbf{r}') = j_0 \left(\frac{\omega}{c} \|\mathbf{r} - \mathbf{r}'\| \right)$$

Spherical Bessel function

Speed of sound

$$\boldsymbol{\kappa}(\omega, \mathbf{r}) = [\kappa(\omega, \mathbf{r}, \mathbf{r}_1), \dots, \kappa(\omega, \mathbf{r}, \mathbf{r}_M)]$$

$$\mathbf{K}(\omega) = \begin{bmatrix} \kappa(\omega, \mathbf{r}_1, \mathbf{r}_1) & \dots & \kappa(\omega, \mathbf{r}_M, \mathbf{r}_1) \\ \vdots & \ddots & \vdots \\ \kappa(\omega, \mathbf{r}_1, \mathbf{r}_M) & \dots & \kappa(\omega, \mathbf{r}_M, \mathbf{r}_M) \end{bmatrix}$$

Kernel-interpolation-based FxLMS

- Express cost function using kernel interpolation filter and error signal.

$$\mathcal{L} = \mathbb{E} \left[\sum_{i,j=-\infty}^{\infty} e^T(n-i) \boldsymbol{\Gamma}(i,j) e(n-j) \right]$$

$$\boldsymbol{\Gamma}(i,j) = \int_{\Omega} \mathbf{z}(i, \mathbf{r}) \mathbf{z}^T(j, \mathbf{r}) d\mathbf{r}$$

- Gradient is obtained with a simplified interpolation weighting filter.

$$\nabla \mathcal{L}(i) = \sum_{k=-\infty}^{\infty} \sum_{j=0}^{J-1} \mathbf{G}^T(j) \mathbf{A}(k) \mathbb{E} \left[e(n) \mathbf{x}^T(n-i-j-k) \right]$$

$$\mathbf{A}(k) = \sum_{j=-\infty}^{\infty} \boldsymbol{\Gamma}(j, j+k) = \mathcal{F}^{-1} \left[\int_{\Omega} \mathbf{z}^*(\omega, \mathbf{r}) \mathbf{z}^T(\omega, \mathbf{r}) d\mathbf{r} \right] (k)$$

Simplify to obtain a practical algorithm

- Truncate and delay weighting filter to obtain casual linear phase filter.

$$\bar{\mathbf{A}}(i) = \begin{cases} \mathbf{A}(i-K) & i \in \{0, \dots, 2K\} \\ \mathbf{0} & \text{otherwise} \end{cases}$$

- Delay error signal by K samples to obtain correct cross-correlation.
- If secondary path is estimated offline, it can be combined ahead of time with the weighting filter to save computations.

$$\mathbf{H}(i) = \sum_{j=0}^{2K} \bar{\mathbf{A}}^T(j) \mathbf{G}(i-j)$$

- Update filter with instantaneous value instead of expected value.

$$\mathbf{W}_{n+1}(i) = \mathbf{W}_n(i) - \mu \sum_{j=0}^{J+2K-1} \mathbf{H}^T(j) e(n-K) \mathbf{x}^T(n-i-j)$$

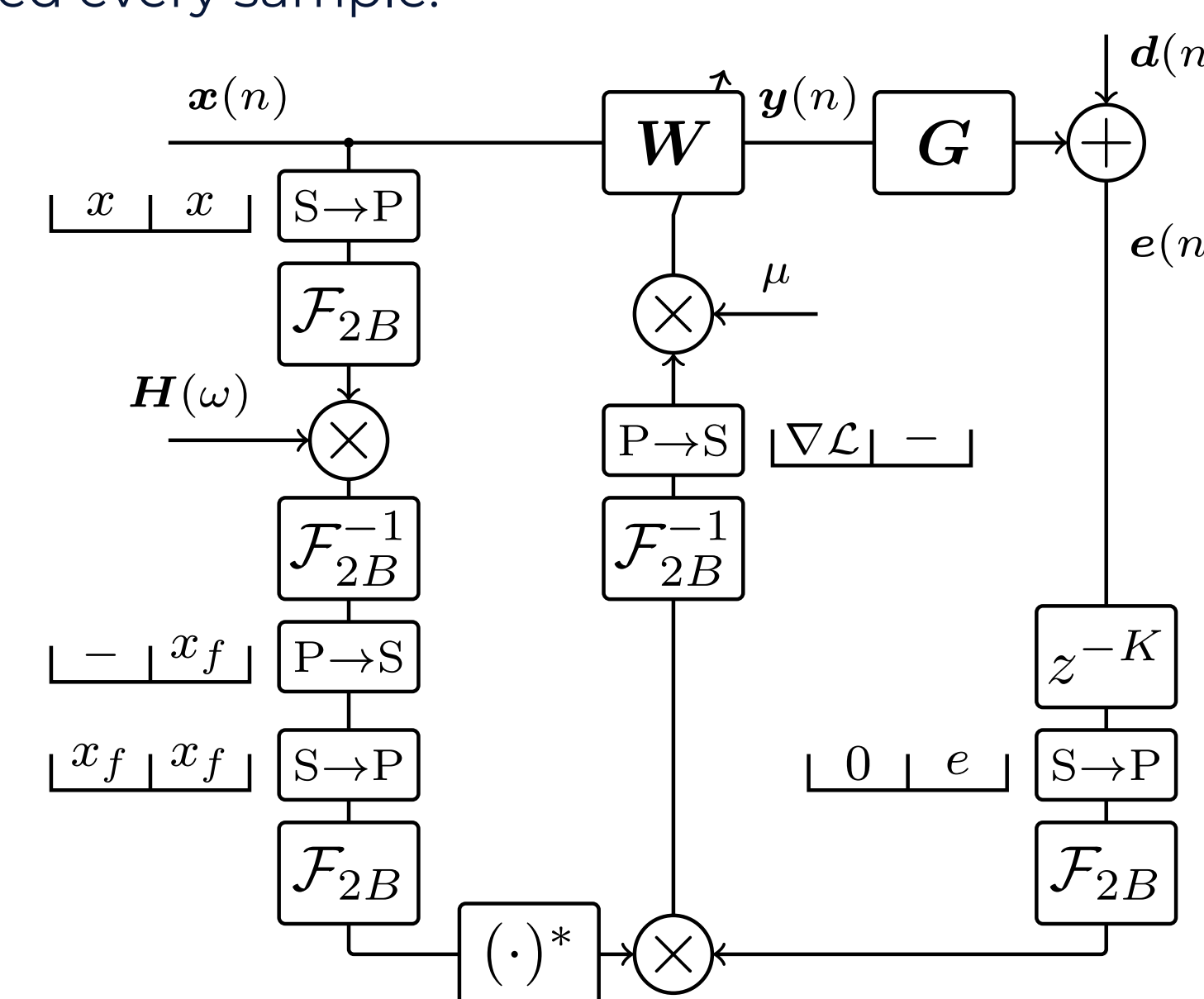
Fast block KI-FxLMS

Block-based implementation to reduce computational cost

- Update filter once per block.
- Compute convolutions and correlations with overlap-save and fast Fourier transform.
- Control filter is obtained in time domain to easily maintain causality.
- Loudspeaker signals are generated every sample.

More computationally efficient than KI-FxLMS when the control filter length is over 32

- Block length and control filter length is B .
- Fast Fourier transform (FFT) are performed for sequences of length $2B$.
- Frequency domain filters are obtained as the FFT of B filter coefficients followed by B zeroes.
- Serial to parallel (S→P) operation concatenates 2 blocks worth of samples.
- Parallel to serial (P→S) operation discards one blocks worth of samples.

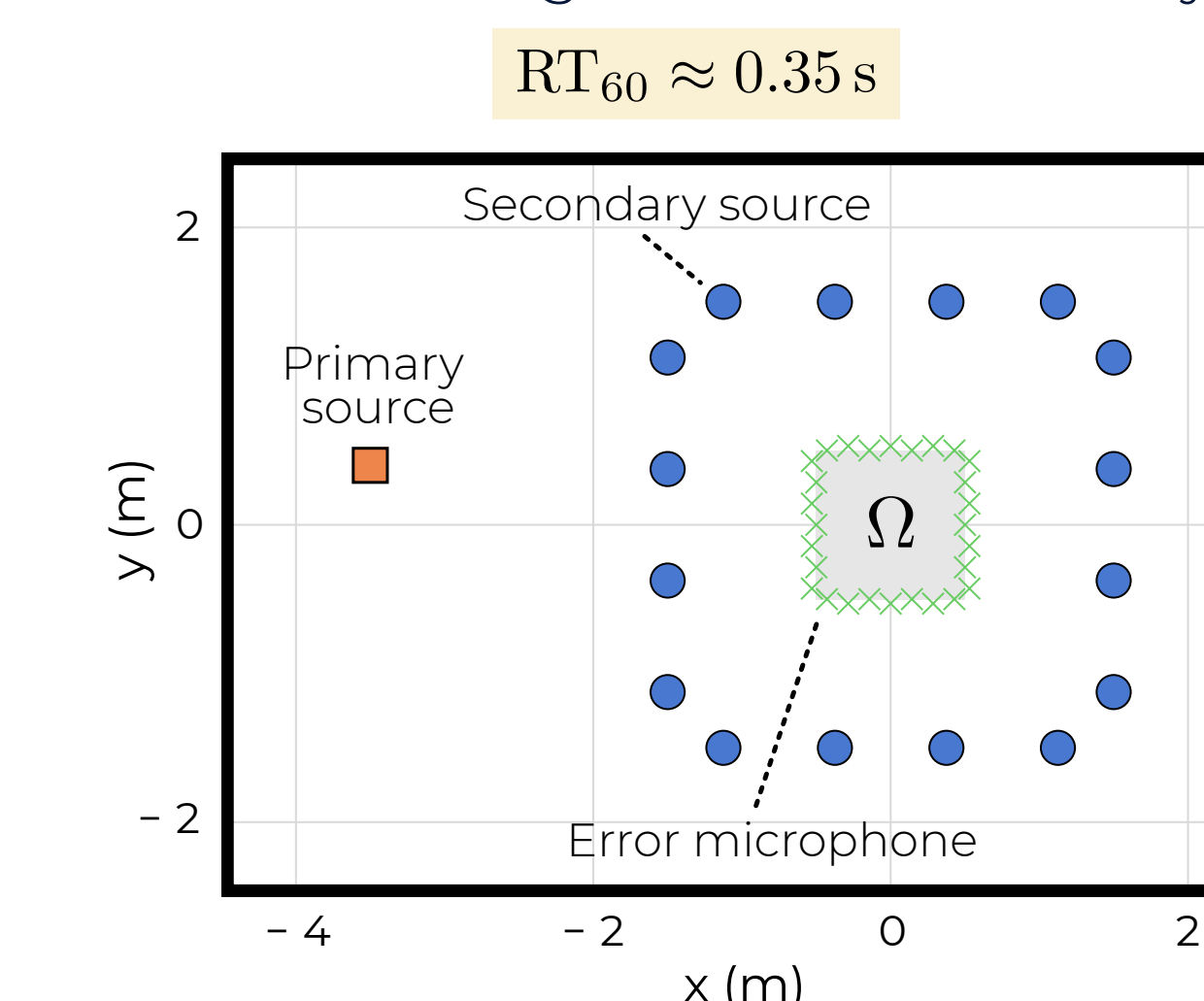


Evaluation

- Comparison between FxLMS and KI-FxLMS

Simulation in reverberant space

- 28 error microphones
- 16 loudspeakers
- Reference signal obtained directly



Source types

- **Noise** - White noise bandlimited to 100-500 Hz
- **Song** - High Horse by artist Secret Mountains
- **Instrumental** - Instrumental version of Song

Regional power reduction

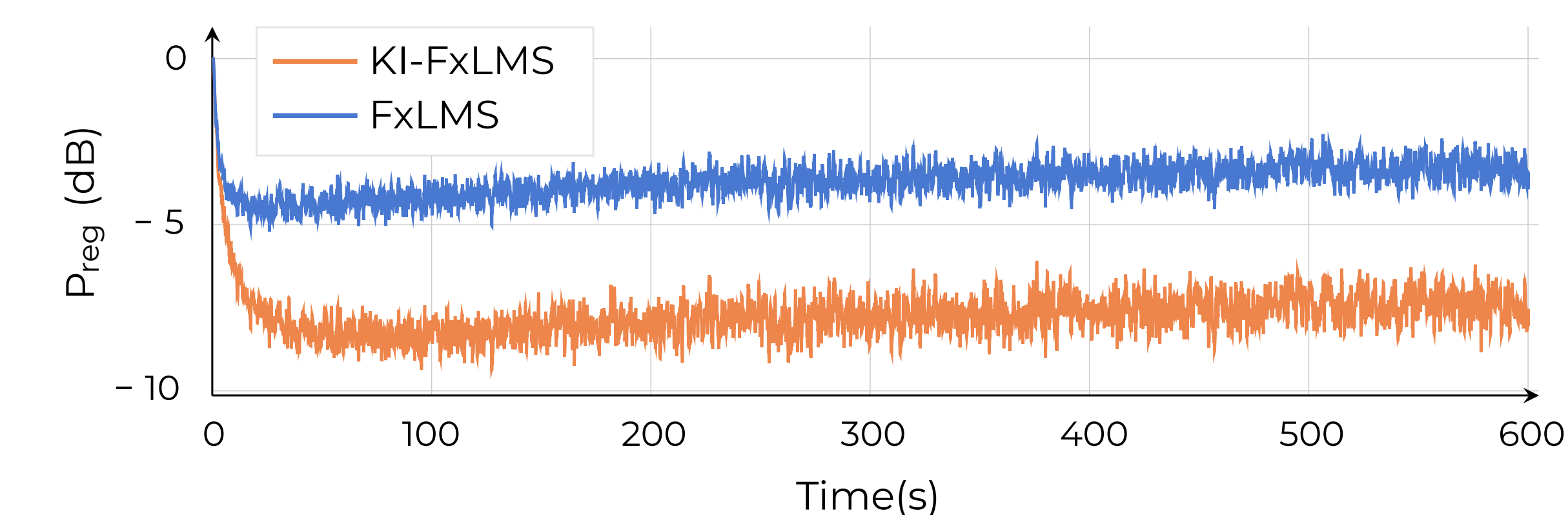
- Measured by discretizing the region of interest with 400 equally spaced points.

$$P_{\text{reg}}(n) = \frac{\sum_{\nu} \sum_{\tau} u(n-\tau, \mathbf{r}_{\nu})^2}{\sum_{\nu} \sum_{\tau} u_p(n-\tau, \mathbf{r}_{\nu})^2}$$

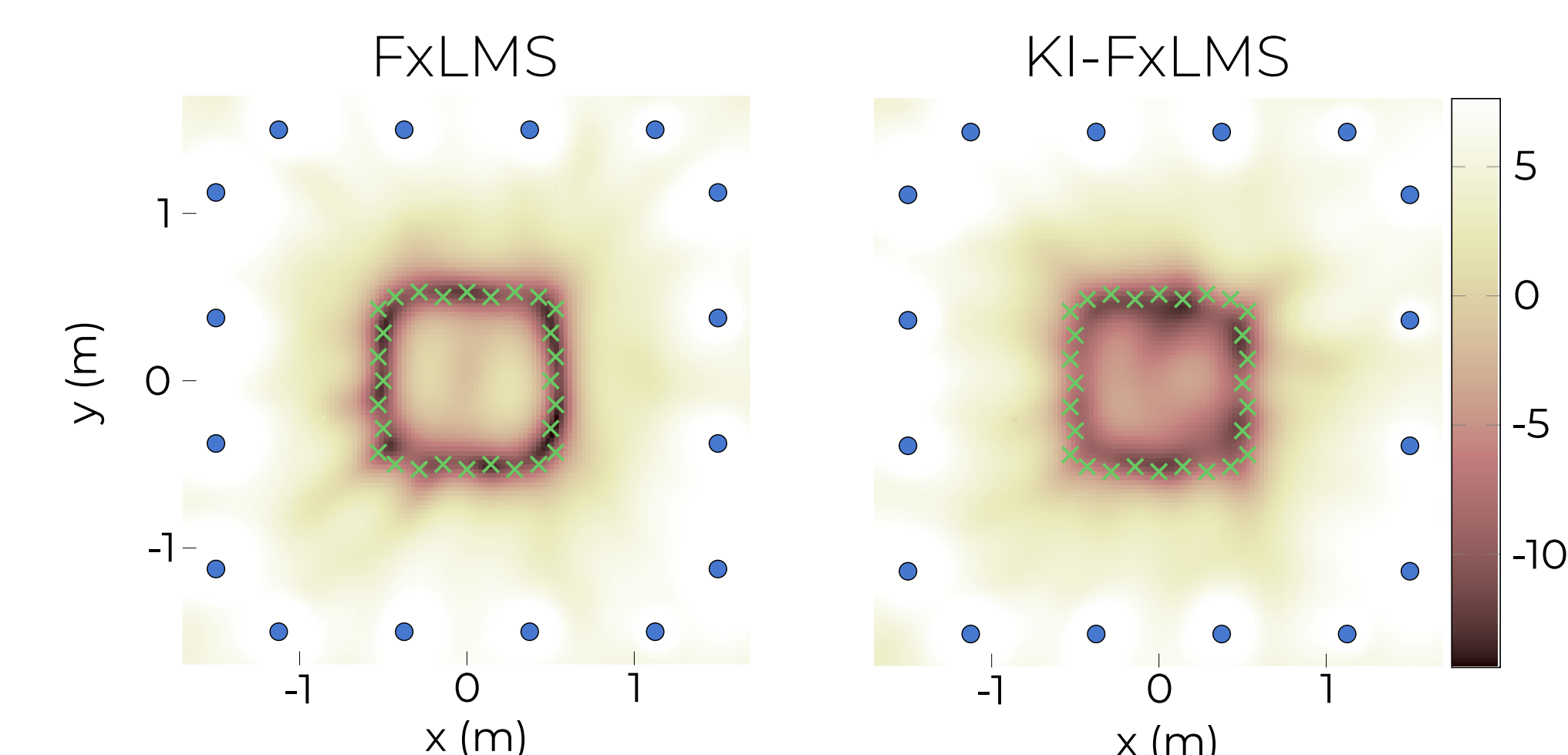
Primary noise sound pressure

Results

Source type Noise



- Power distribution in dB after 600 seconds



All source types

- Average regional power reduction in dB between 120 and 240 seconds

	Noise	Song	Instrumental
FxLMS	-3.81	-3.28	-3.13
Fast Block FxLMS	-3.88	-3.17	-3.12
KI-FxLMS	-7.88	-3.99	-4.39
Fast Block KI-FxLMS	-8.00	-3.60	-4.08