

Sound Field Recording and Reproduction Using Small Number of Microphones and Loudspeakers

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Institut Langevin
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About me

➤ Shoichi Koyama, Ph.D.

- 2007 B.E. and 2009 M.S. degrees from UTokyo
- 2009 – 2014: Nippon Telegraph and Telephone Corp.
- 2014: Ph.D. (Inf Sci&Tech) from UTokyo
- 2014 – present: Assistant Prof. (Research Associate) at UTokyo
- 2016 – present: Visiting researcher at Paris Diderot University (Paris7) / Institut Langevin



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UTokyo and Institut Langevin



➤ The University of Tokyo (Tokyo, Japan):

- Our lab has just started on Apr. 2014
- 5 staffs including Prof. Hiroshi Saruwatari and 9 students
- Audio, speech, and music signal processing



➤ Institut Langevin (Paris, France):

- Research institute for “*physical waves*” including optics and acoustics
- Working with Prof. Laurent Daudet
- Staying for 2 years (until Mar. 2018)



Institut **Langevin**
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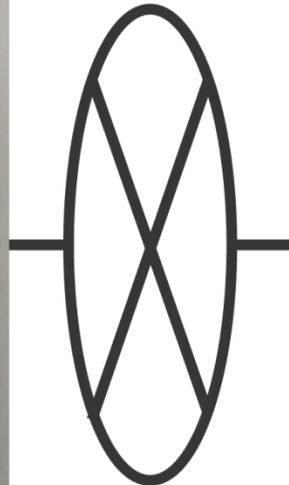
Real-time Sound Field Transmission System

System developed while I worked at NTT [Koyama+ IEICE Trans 2014]

Kanagawa



Network



Tokyo

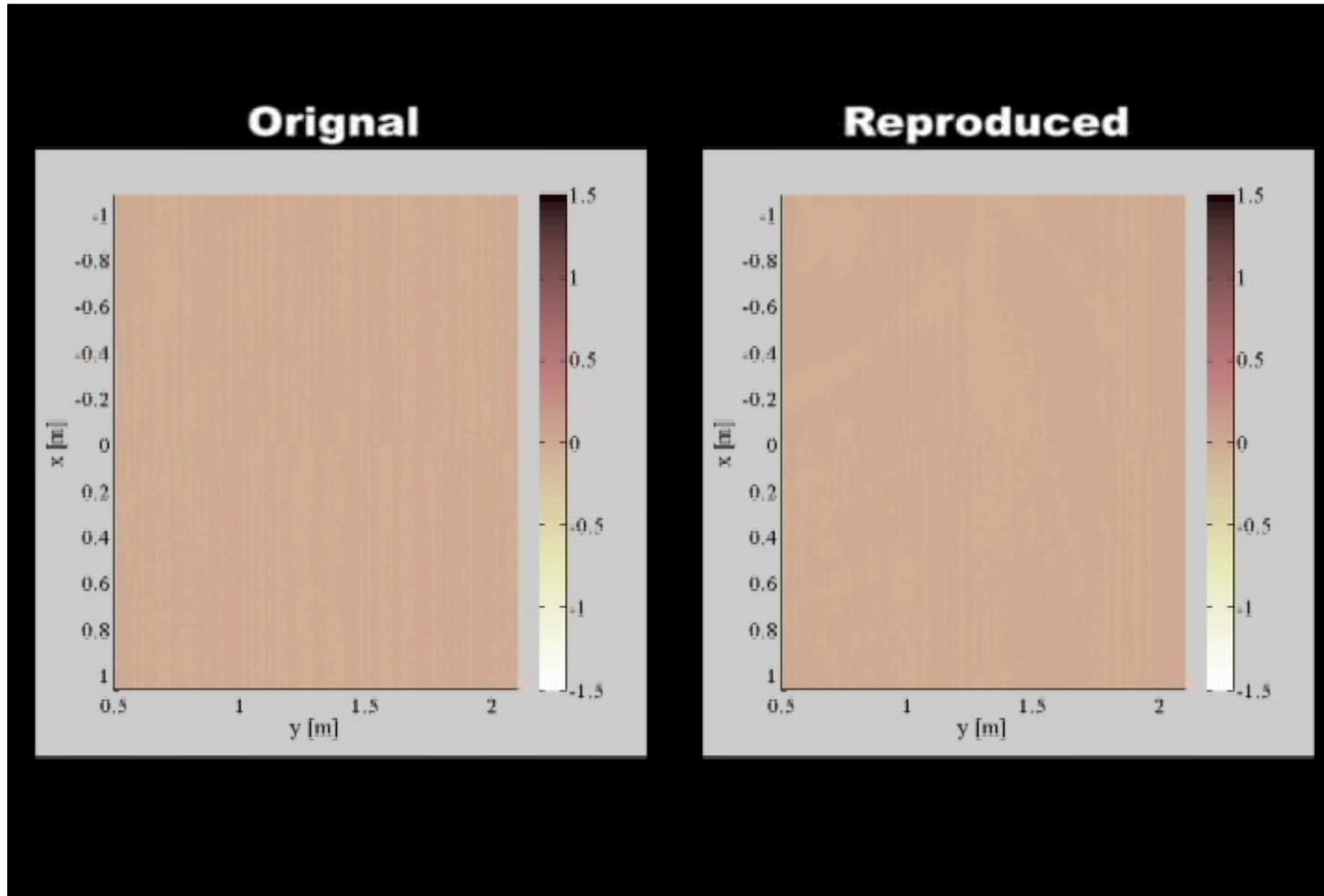


- Loudspeakers (for high freq.): 64, 6cm intervals
- Loudspeakers (for low freq.): 32, 12cm intervals
- Microphones: 64, 6cm intervals
- Array size: 3.84 m
- Sampling freq.: 48 kHz, Delay: 152 ms



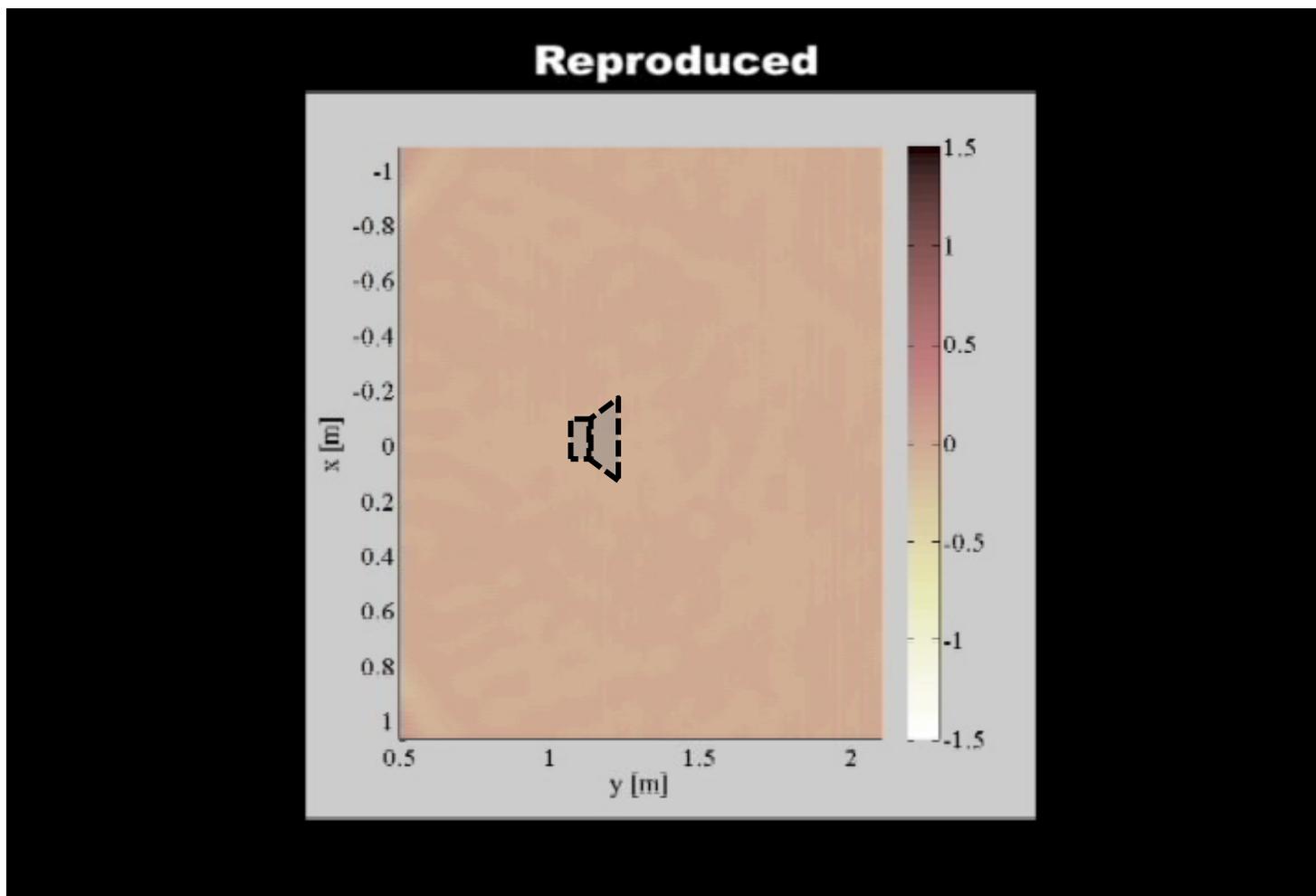
Visualization of Reproduced Sound Field

- Source signal: Low-passed pulse (0 – 2.6kHz)
- Source: Loudspeaker, Position: (-1.0, -1.0, 0.0) m



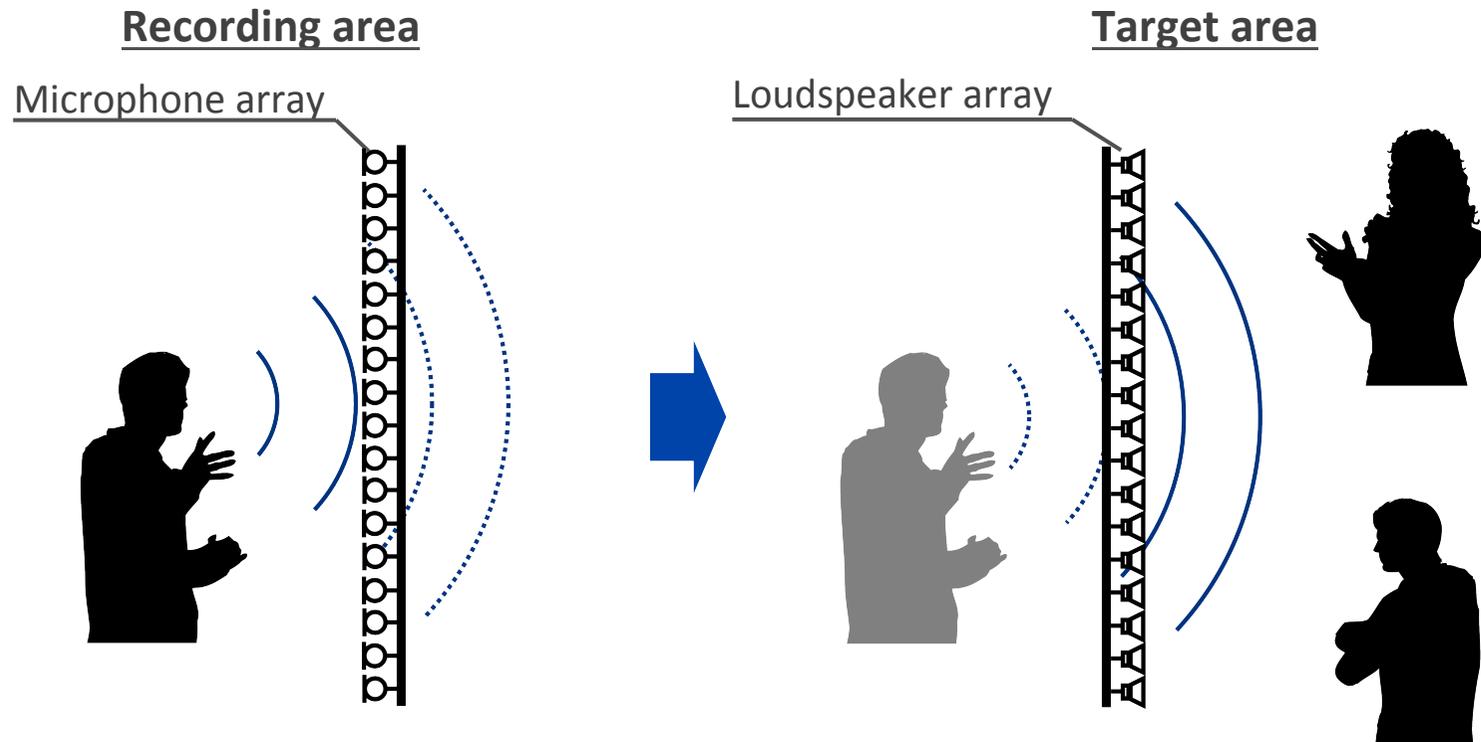
Visualization of Reproduced Sound Field

- Source signal: Low-passed pulse (0 – 2.6kHz)
- Source: Loudspeaker, Position: (0.0, -1.0, 0.0) m, 2.0 m forward shift



Today's Topic

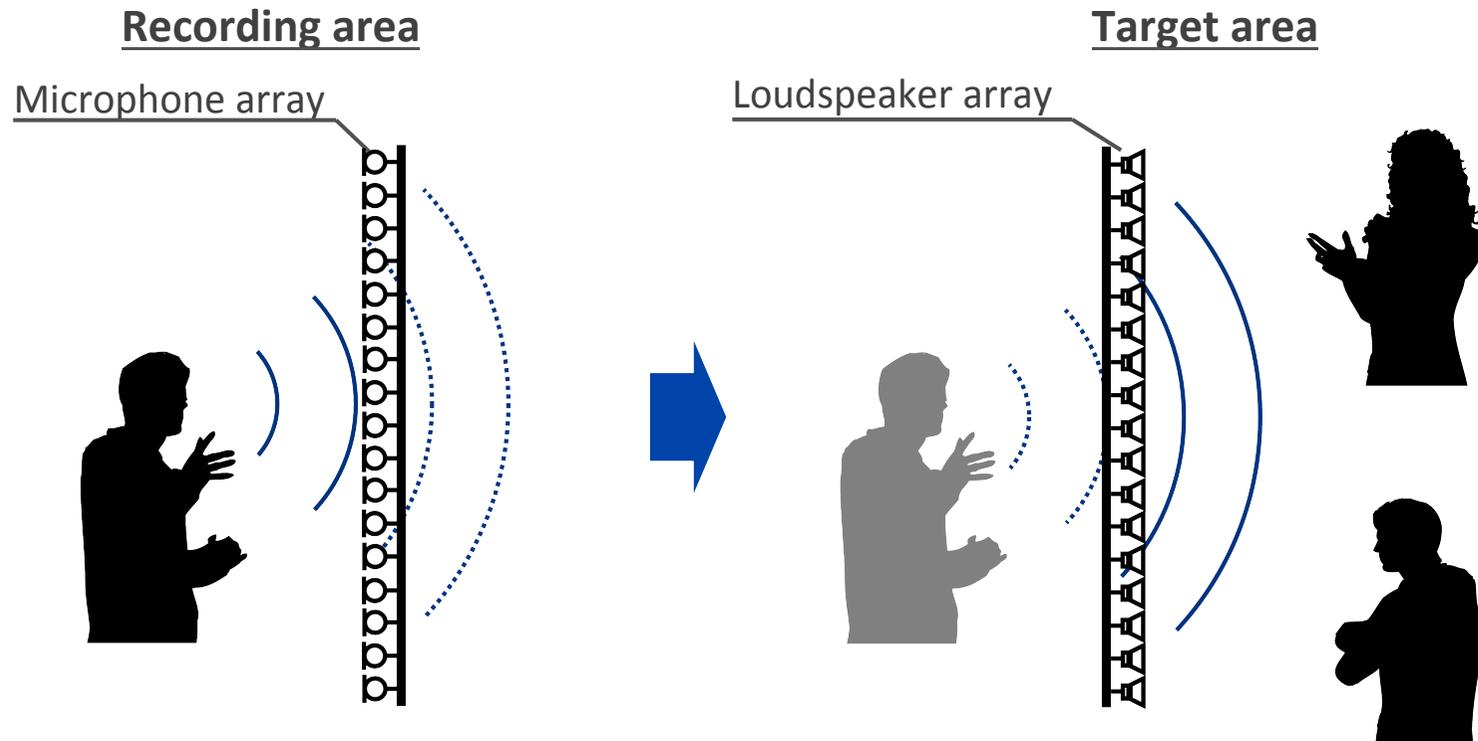
How to reduce microphones and loudspeakers in recording and reproduction?



- Insufficient number of array elements: ***spatial aliasing artifacts***
 - Low reproduction accuracy at high frequencies
 - Inaccurate frequency characteristics (*coloration effect*)

Today's Topic

How to reduce microphones and loudspeakers in recording and reproduction?



Improve reproduction accuracy using prior information

- **Reduction of the number of microphones** [Koyama+ ICASSP2014, 2015]
 - Improve reproduction accuracy within predefined near-field source area
- **Reduction of the number of loudspeakers** [Ueno+ ICASSP2017, HSCMA2017]
 - Improve reproduction accuracy within predefined listening area

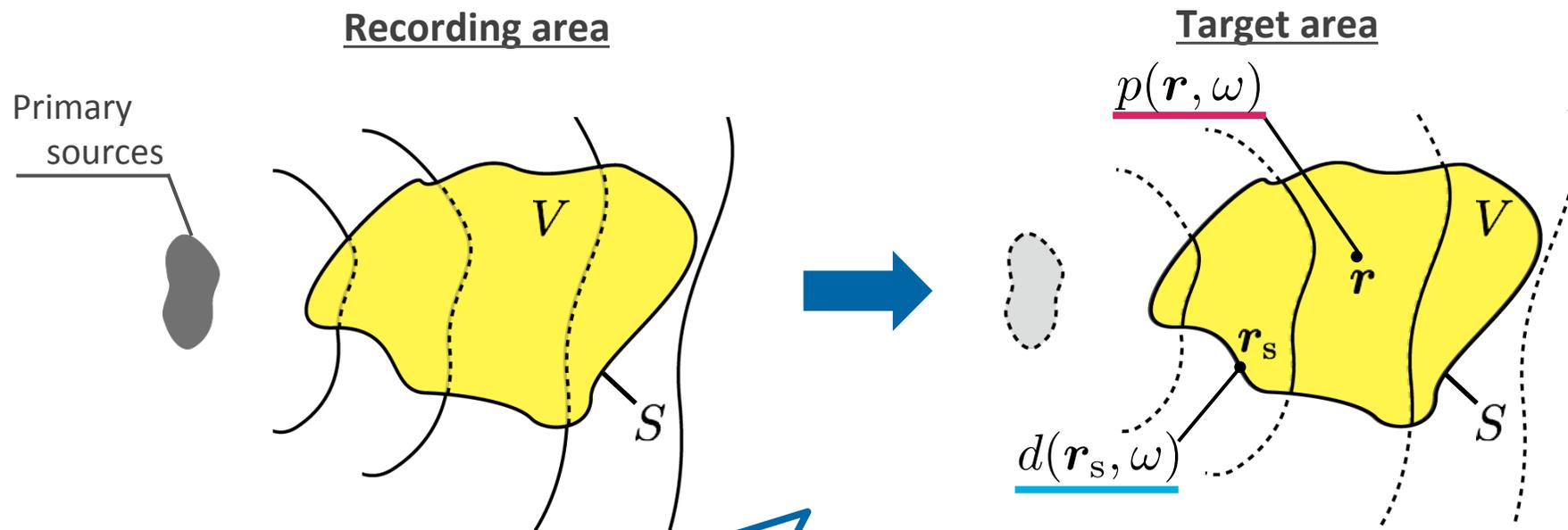


REDUCTION OF THE NUMBER OF MICROPHONES

October 24, 2017

Sound Field Recording and Reproduction

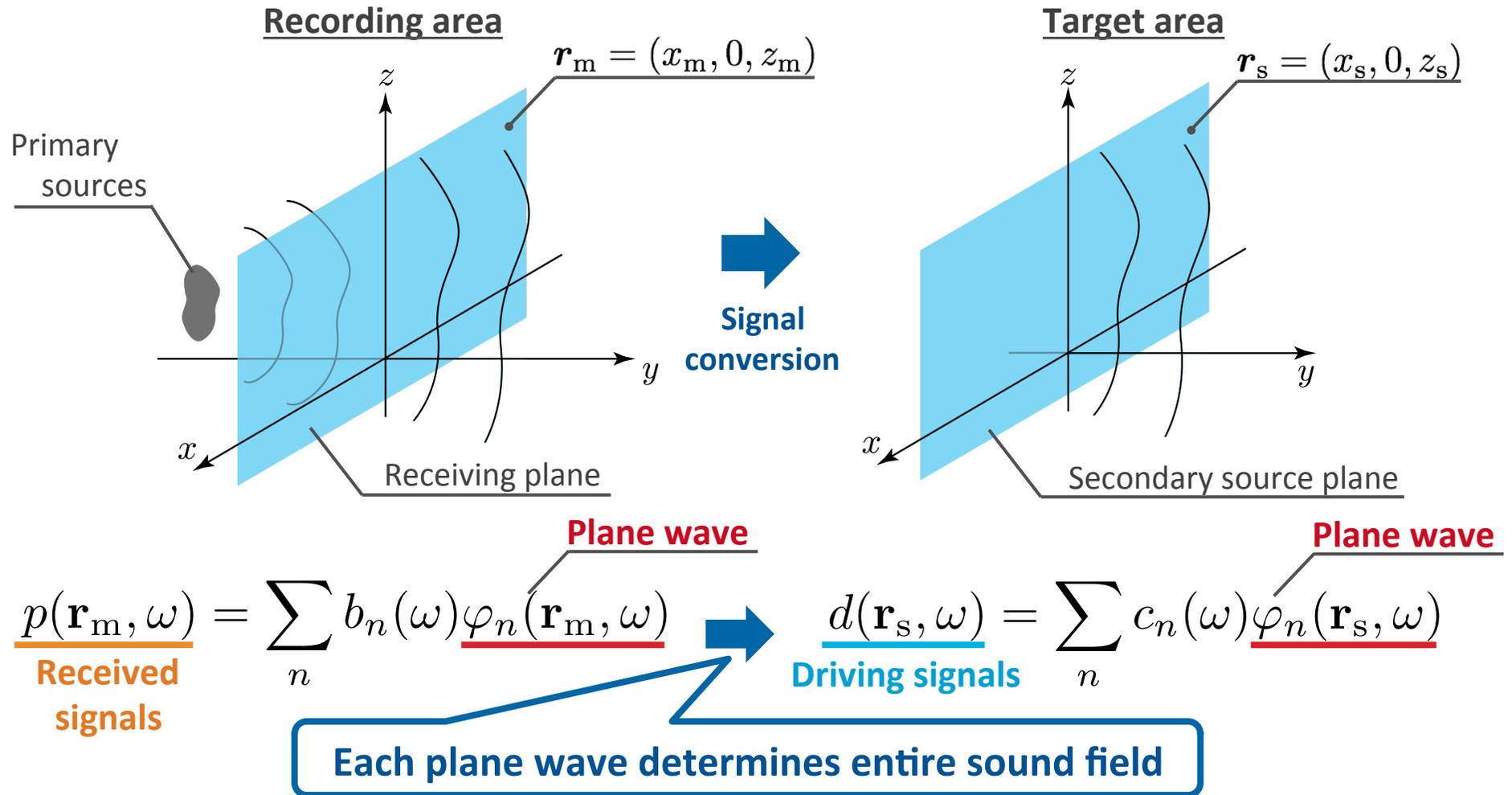
Obtain driving signals $d(\mathbf{r}_s, \omega)$ of secondary sources (= loudspeakers) arranged on S to reconstruct desired sound field inside V



When sound pressures at multiple positions are only known in the recording area, typical strategy to obtain the driving signals is plane wave or harmonic decomposition of the captured sound field

Conventional: WFR filtering method

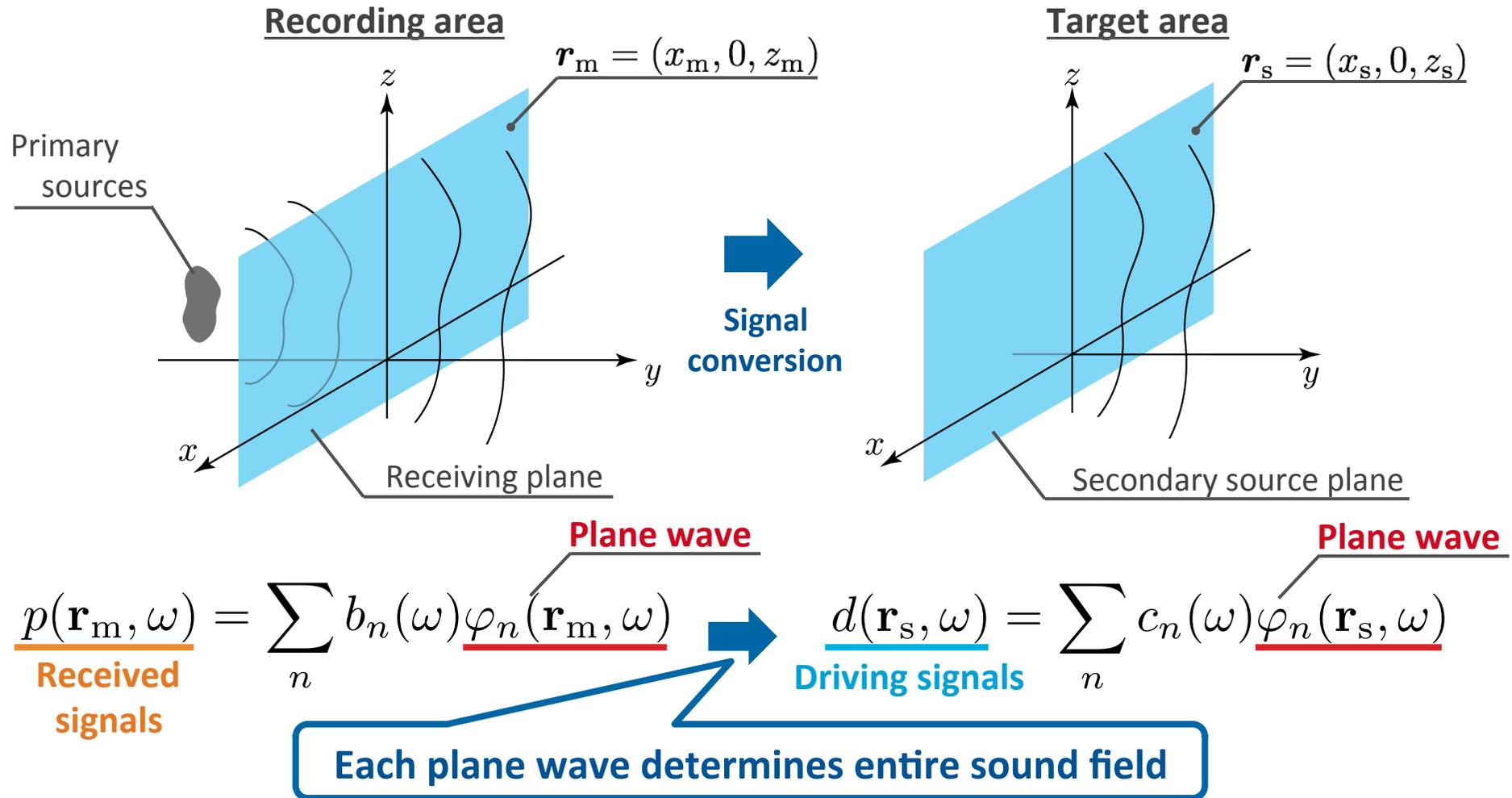
[Koyama+ IEEE TASLP 2013]



Signal conversion based on plane-wave decomposition

Conventional: WFR filtering method

[Koyama+ IEEE TASLP 2013]



Spatial aliasing artifacts due to plane wave decomposition →
Significant error above spatial Nyquist freq of microphone array

Sound Field Decomposition in Recording

$$\underbrace{p(\mathbf{r}, \omega)}_{\text{Received signals}} = \sum_n b_n(\omega) \underbrace{\varphi_n(\mathbf{r}, \omega)}_{\text{Basis function}}$$

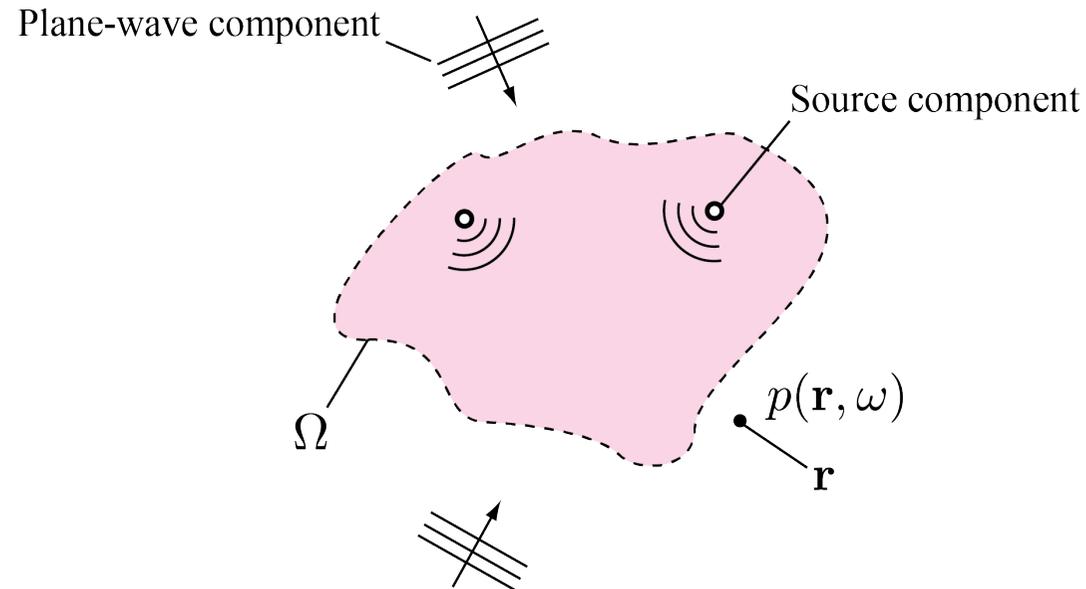
- ✧ Plane wave / harmonic decomposition suffers from spatial aliasing artifacts because many basis functions are used
- ✧ Observed signals should be represented by a few basis functions for accurate interpolation of sound field
- ✧ Appropriate basis function may be close to pressure distribution originating from near-field sound sources
- ✧ To obtain driving signals of loudspeakers, basis functions must be elementary solutions of Helmholtz equation (e.g. Green functions)

➔ ***Sound field decomposition***

Sound field decomposition into elementary solutions of Helmholtz equation is necessary

Generative model of sound field

[Koyama+ ICASSP 2014]



➤ Inhomogeneous Helmholtz eq.

$$\left[(\nabla^2 + k^2) \underline{p(\mathbf{r}, \omega)} = -\underline{Q(\mathbf{r}, \omega)} \right]$$

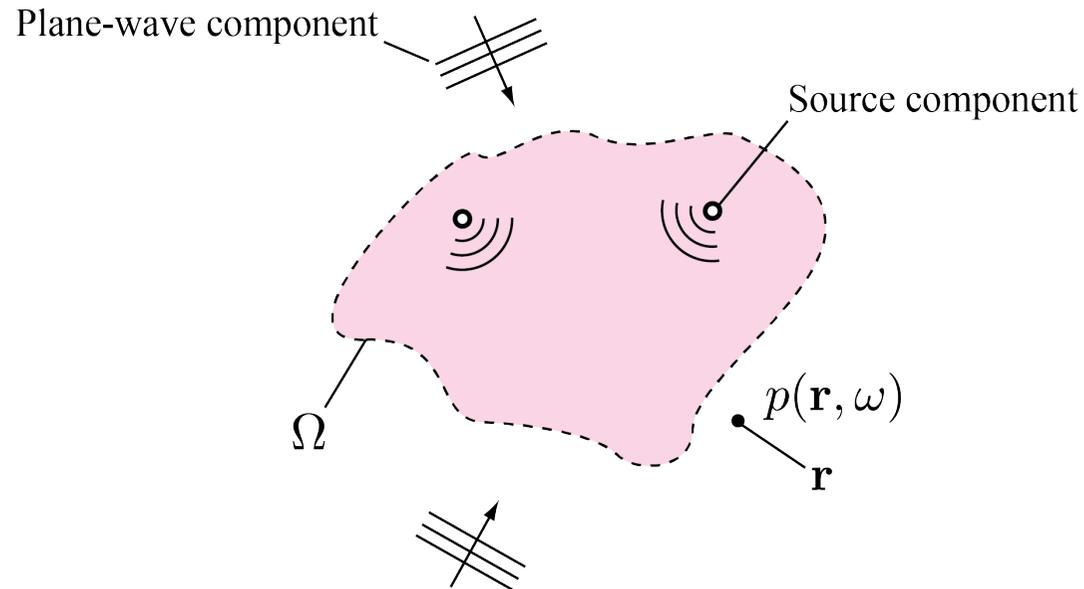
Unknown boundary condition on room boundary

Distribution of
source components

Sound field consisting of near-field source and far-field
plane-wave components

Generative model of sound field

[Koyama+ ICASSP 2014]



➤ Inhomogeneous Helmholtz eq.

$$\underline{p(\mathbf{r})} = p_P(\mathbf{r}) + p_H(\mathbf{r})$$

Sum of particular + homogeneous solutions

$$= \int_{\mathbf{r}' \in \Omega} \underline{Q(\mathbf{r}')} \underline{G(\mathbf{r}|\mathbf{r}')} d\mathbf{r}'$$

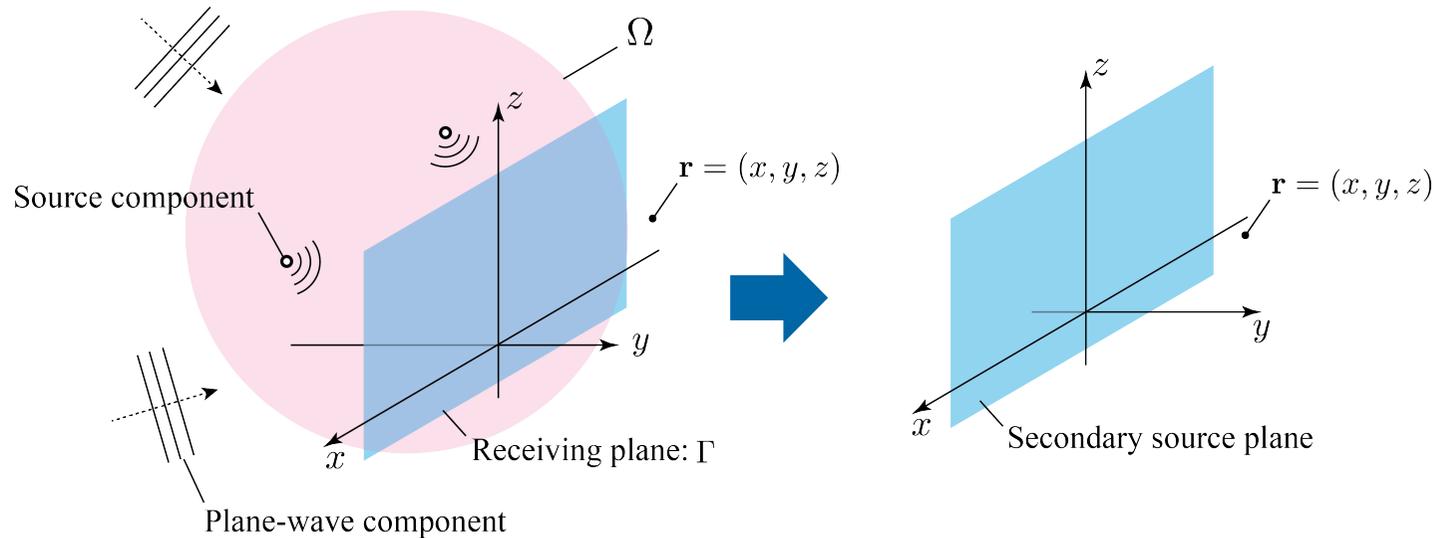
Green's function

$$+ \frac{1}{4\pi^2} \int_{-\infty}^{\infty} dk_x \int_{-\infty}^{\infty} dk_z P_H(k_x, k_z) e^{j(k_x x + k_y y + k_z z)}$$

Plane wave

Generative model of sound field

- Observe sound pressure distribution on plane Γ



- Conversion into driving signals

$$d(\mathbf{r}) = \frac{\partial p(\mathbf{r})}{\partial y} \Big|_{y=0}$$

$$= \int_{\mathbf{r}' \in \Omega} \underbrace{Q(\mathbf{r}')}_{\text{Direct source components}} \frac{\partial G(\mathbf{r}|\mathbf{r}')}{\partial y} \Big|_{y=0} d\mathbf{r}' + \frac{1}{4\pi^2} \int_{-\infty}^{\infty} dk_x \int_{-\infty}^{\infty} dk_z jk_y P_H(k_x, k_z) e^{j(k_x x + k_z z)}$$

Synthesize monopole sources [Spors+ AES Conv. 2008]

Direct source components

Reverberant components

Applying WFR filtering method [Koyama+ IEEE TASLP 2013]

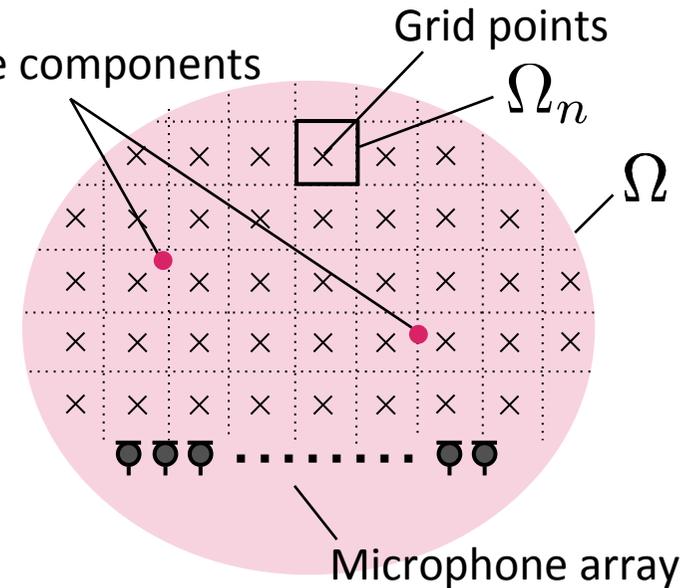
Decomposition into two components can lead to higher reproduction accuracy above spatial Nyquist freq

Sparse sound field representation

➤ Sparsity-based signal decomposition

$$\underline{p}(\mathbf{r}) = \int_{\mathbf{r}' \in \Omega} \underline{Q}(\mathbf{r}') \underline{G}(\mathbf{r}|\mathbf{r}') d\mathbf{r}' + \underline{p}_H(\mathbf{r})$$

$(\mathbf{r} \in \Gamma)$



Discretization

$$\underline{y} = \underline{D} \underline{x} + \underline{z}$$

Observed signal

Dictionary matrix of Green's functions

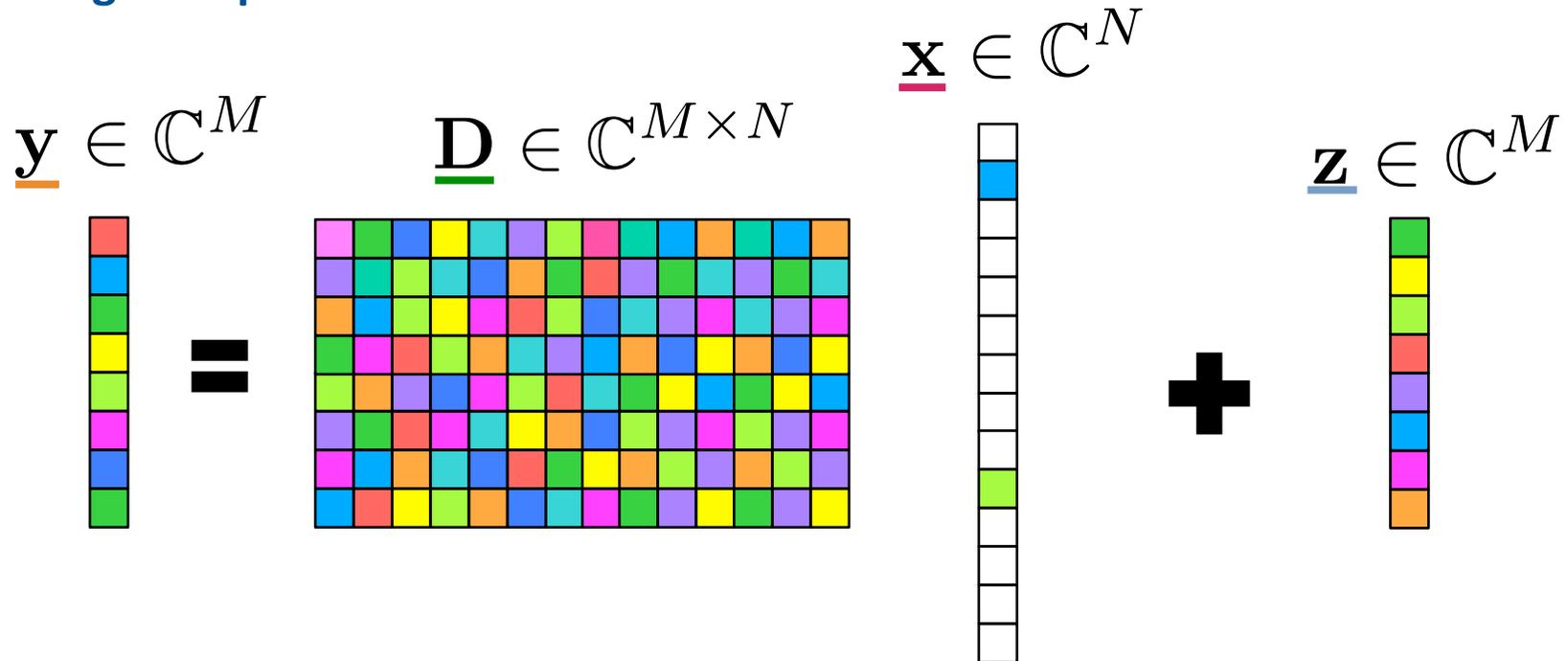
Distribution of source components

Reverberant components

A few elements of \underline{x} has non-zero values
under the assumption of spatially sparse source distribution

Sparse signal decomposition

➤ Sparse signal representation in vector form



➤ Signal decomposition based on sparsity of $\underline{\mathbf{x}}$

minimize $\|\underline{\mathbf{x}}\|_p^p$ ($p \leq 1$)

subject to $\|\underline{\mathbf{y}} - \underline{\mathbf{D}}\underline{\mathbf{x}}\|_2^2 < \epsilon$

Minimize ℓ_p -norm of $\underline{\mathbf{x}}$

Group sparsity based on physical properties

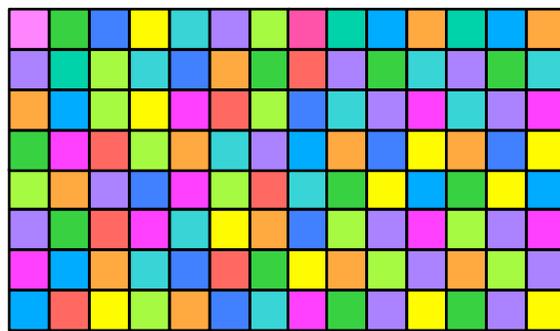
➤ Sparse signal representation in vector form

$$\underline{\mathbf{y}} \in \mathbb{C}^M$$



=

$$\underline{\mathbf{D}} \in \mathbb{C}^{M \times N}$$



\mathbf{x}

Structure of sparsity induced by physical properties



+

$$\underline{\mathbf{z}} \in \mathbb{C}^M$$

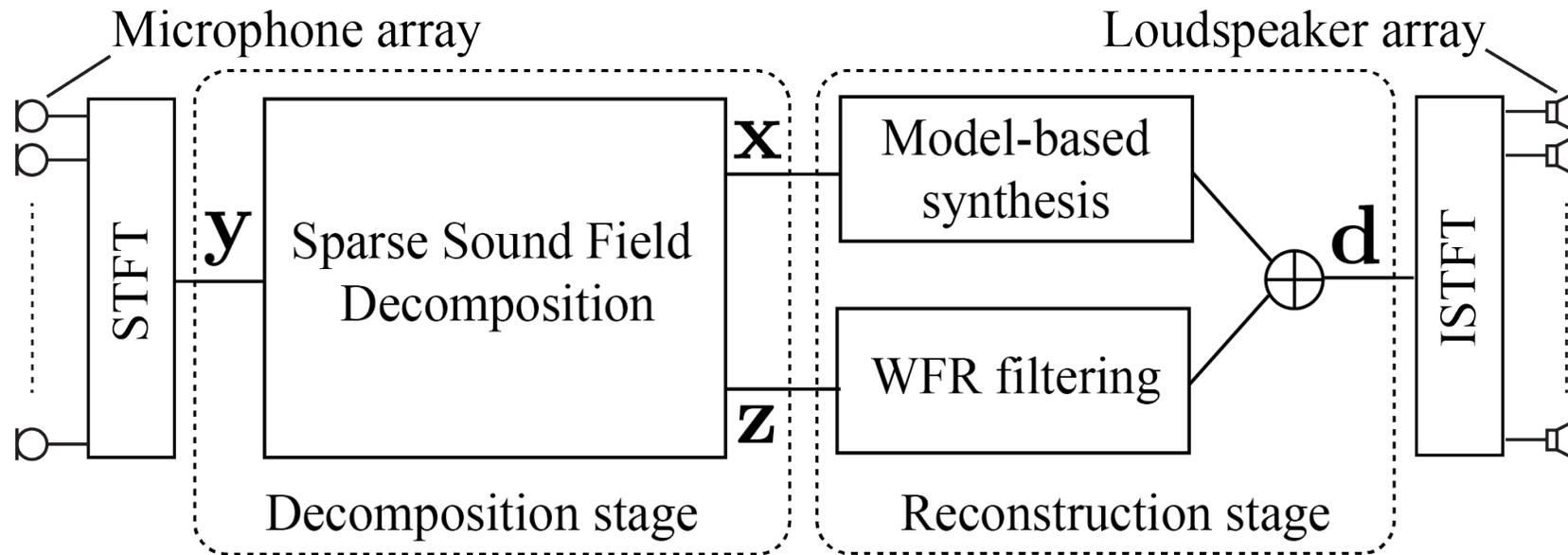


➤ Group sparse signal models for accurate and robust decomposition

- Multiple time frames
- Temporal frequencies
- Multipole components

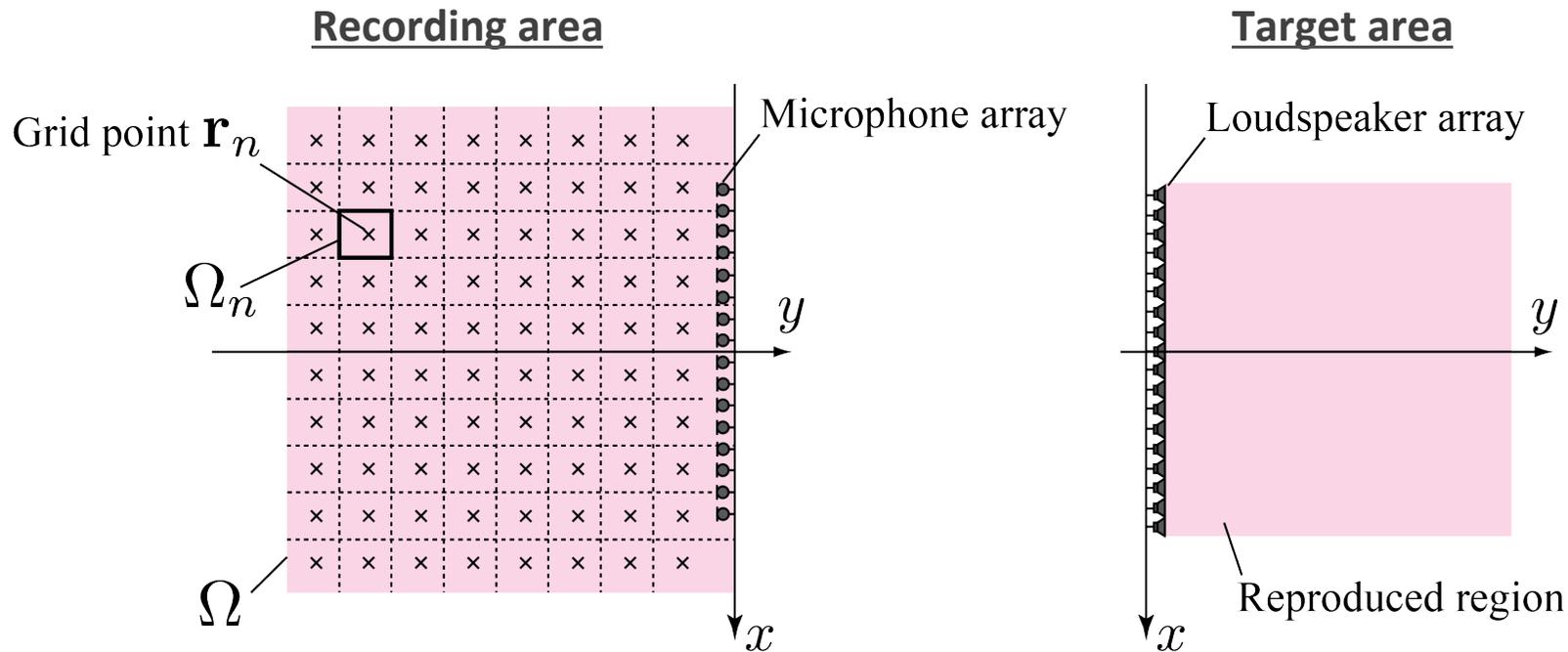
➤ Decomposition algorithm extending FOCUSS [Koyama+ ICASSP 2015]

Block diagram of signal conversion



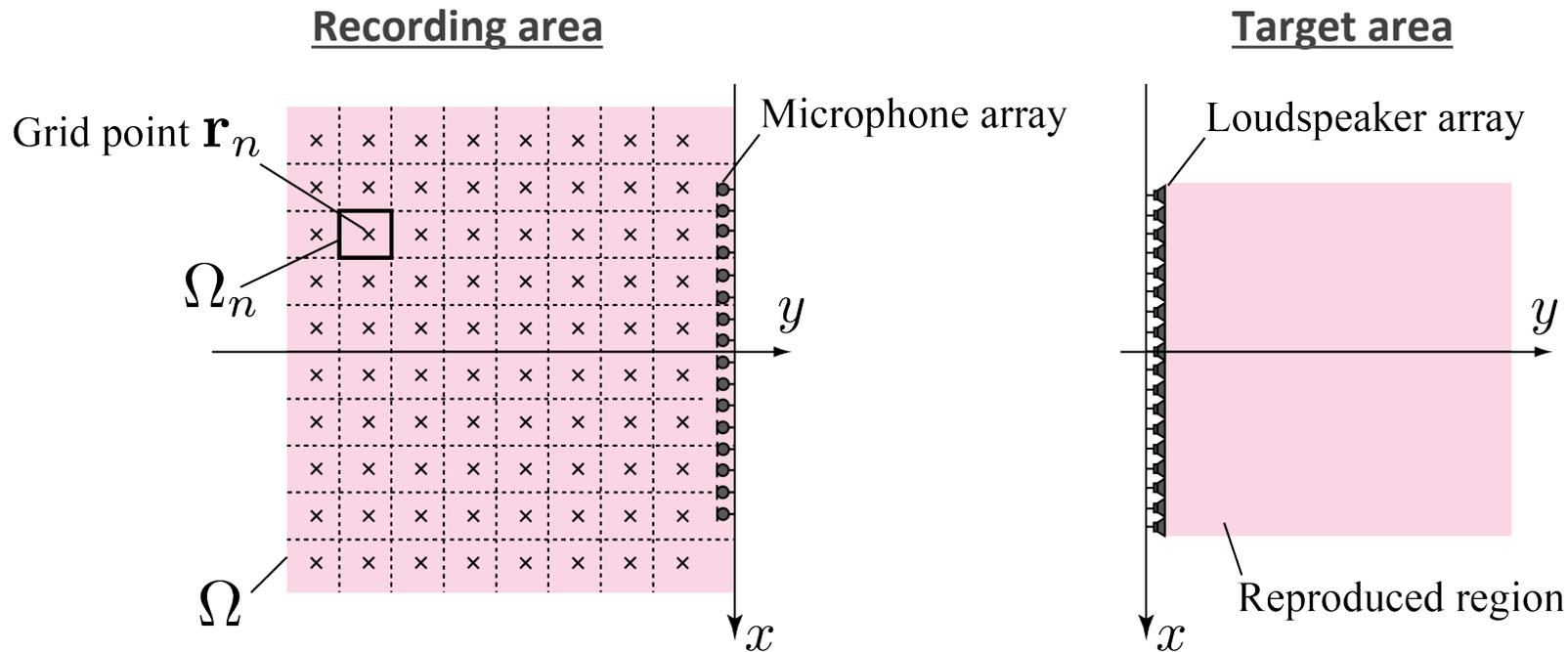
- **Decomposition stage**
 - Group sparse decomposition of y
- **Reconstruction stage**
 - x and z are respectively converted into driving signals
 - d is obtained as sum of two components

Simulation Experiment



- Proposed method (**Proposed**), method based on sparse circular harmonics decomposition (**CH**) WFR filtering method (**WFR**), and Sound Pressure Control method (**SPC**) were compared
- 32 microphones (0.06 m intervals) and 48 loudspeakers (0.04 m intervals)
- Ω : Rectangular region of 2.4x2.4 m, Grid points: (0.01 m, 0.02 m) intervals
- Source directivity: unidirectional
- Source signal: single frequency sinewave

Simulation Experiment



➤ **Signal-to-distortion ratio of reproduction (SDRR)**

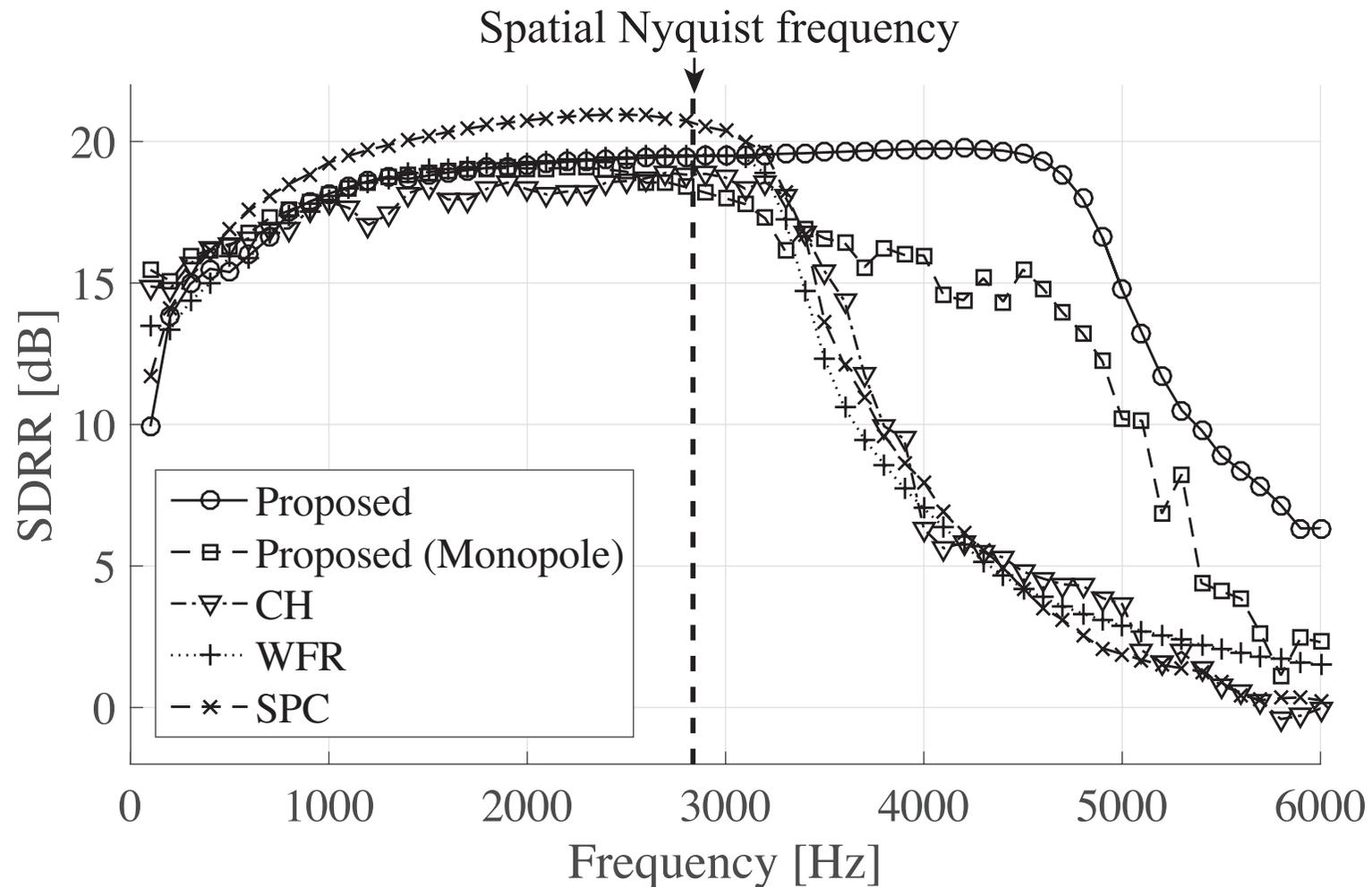
$$\text{SDRR} = 10 \log_{10} \frac{\sum_i \sum_j \sum_k |\bar{p}_{\text{org}}(x_i, y_j, t_k)|^2}{\sum_i \sum_j \sum_k |\bar{p}_{\text{org}}(x_i, y_j, t_k) - \bar{p}_{\text{rep}}(x_i, y_j, t_k)|^2}$$

Original pressure distribution

Reproduced pressure distribution

Frequency vs. SDR

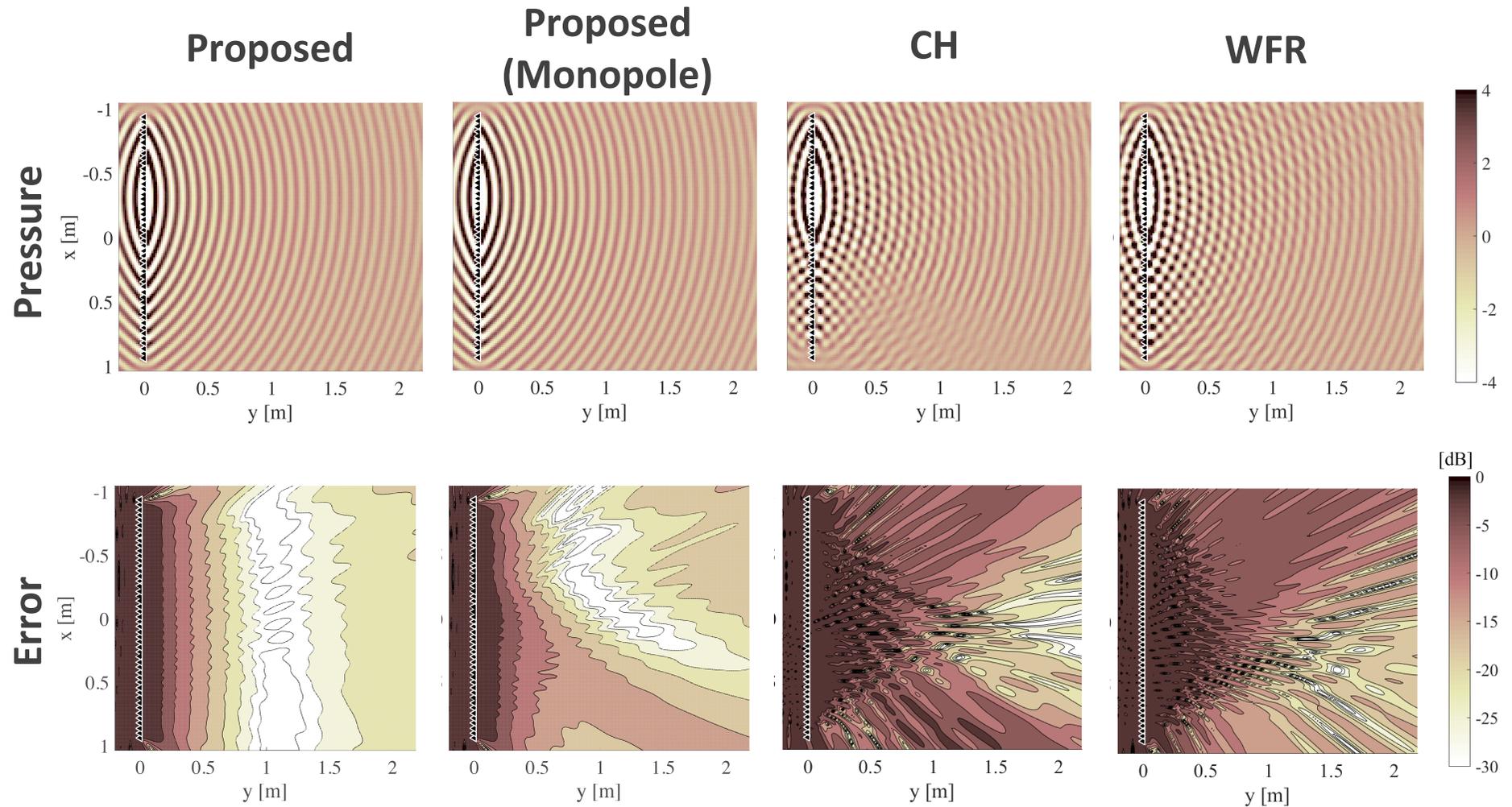
➤ Source location: (-0.32, -0.84, 0.0) m



SDRRs above spatial Nyquist frequency were improved

Reproduced sound pressure distribution (4.0 kHz)

➤ Source location: (-0.32, -0.84, 0.0) m



SDRR: 19.7 dB

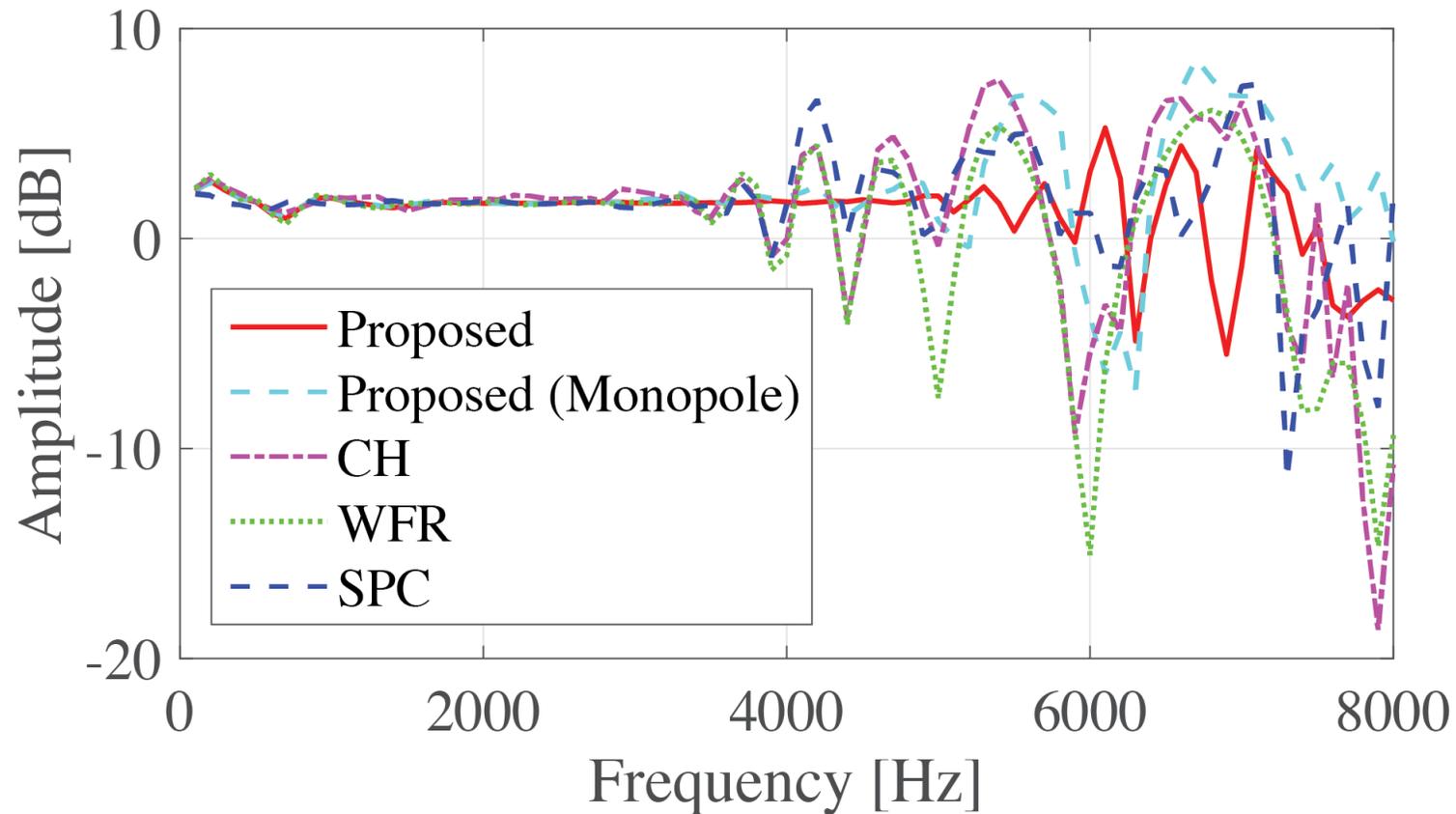
16.0 dB

6.3 dB

7.0 dB

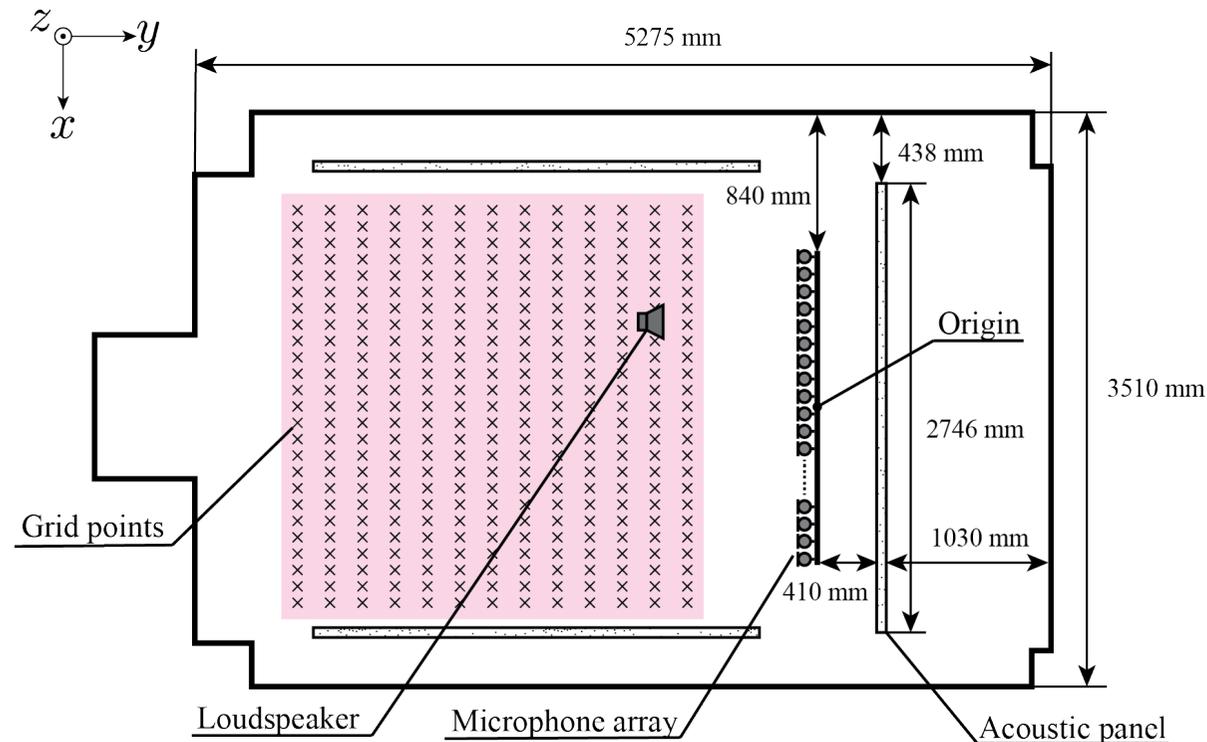
Frequency response of reproduced sound field

➤ Frequency response at (0.0, 1.0, 0.0) m



Reproduced frequency response was improved

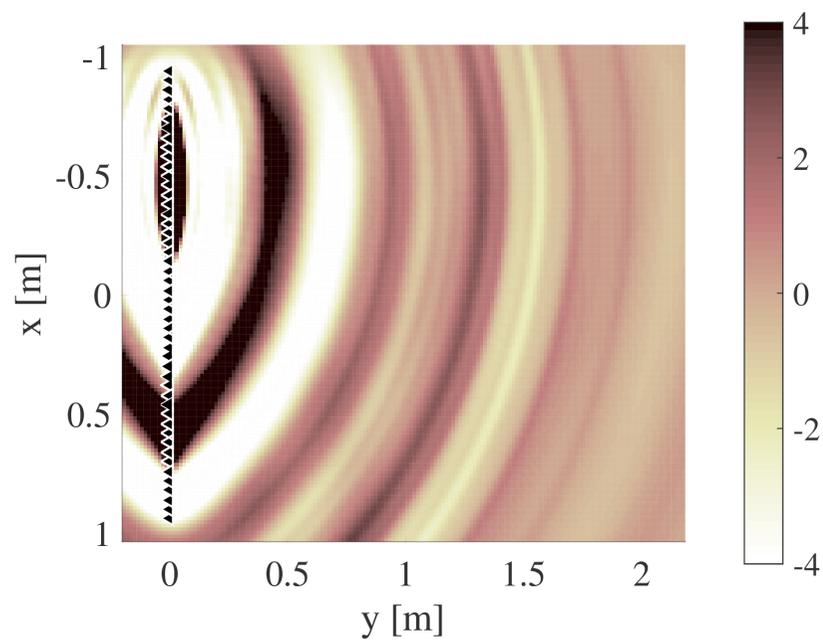
Experiments using real data



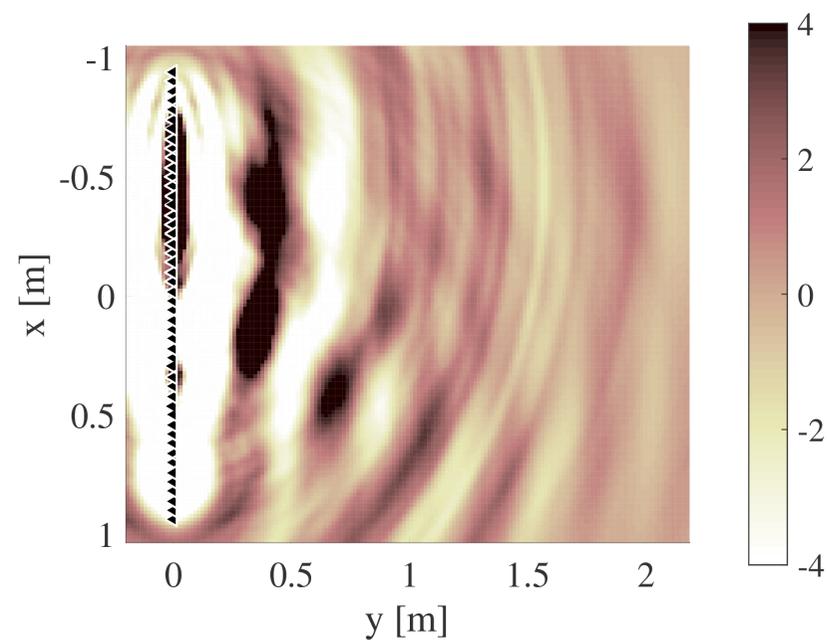
- Proposed method (**Proposed**) and WFR filtering method (**WFR**), were compared
- Same experimental setting as the previous one
- Reproduced region was simulated as free field
- Source signal: speech

Reproduced sound pressure distribution

Proposed



WFR



Reduction of The Number of Microphones

- Conventional plane wave decomposition is suffered from spatial aliasing artifacts
- Sound field representation using near-field source and plane wave components
- Sound field decomposition based on spatial sparsity of near-field source components
- Group sparsity based on physical properties of sound field
- Experimental results indicated that reproduction accuracy above spatial Nyquist frequency can be improved



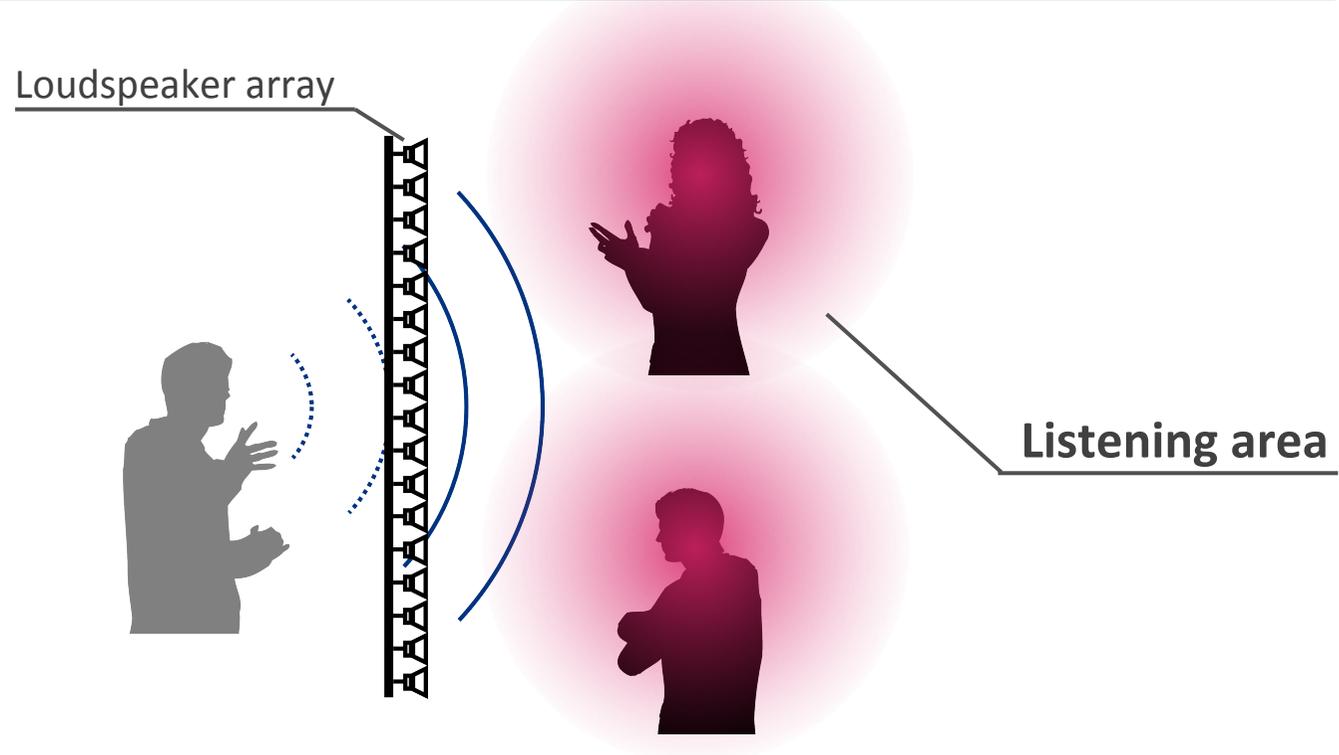
REDUCTION OF THE NUMBER OF LOUDSPEAKERS

October 24, 2017

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Listening-area-informed sound field reproduction

Highly-accurate sound field reproduction
exploiting prior information on listening area



- Probability distribution on the listeners' position is given
- High reproduction accuracy within the listening area is achieved

Problem statement

- Sound field synthesized by L discrete secondary sources in 2D

$$p_{\text{syn}}(\mathbf{r}) = \underline{\mathbf{h}}(\mathbf{r})^{\top} \underline{\mathbf{d}}$$

Transfer function

$$\underline{\mathbf{h}}(\mathbf{r}) = [h(\mathbf{r}, \mathbf{r}_1), \dots, h(\mathbf{r}, \mathbf{r}_L)]$$

Driving signals

$$\underline{\mathbf{d}}(\mathbf{r}) = [d(\mathbf{r}_1), \dots, d(\mathbf{r}_L)]$$

- Expectation minimization of pressure error

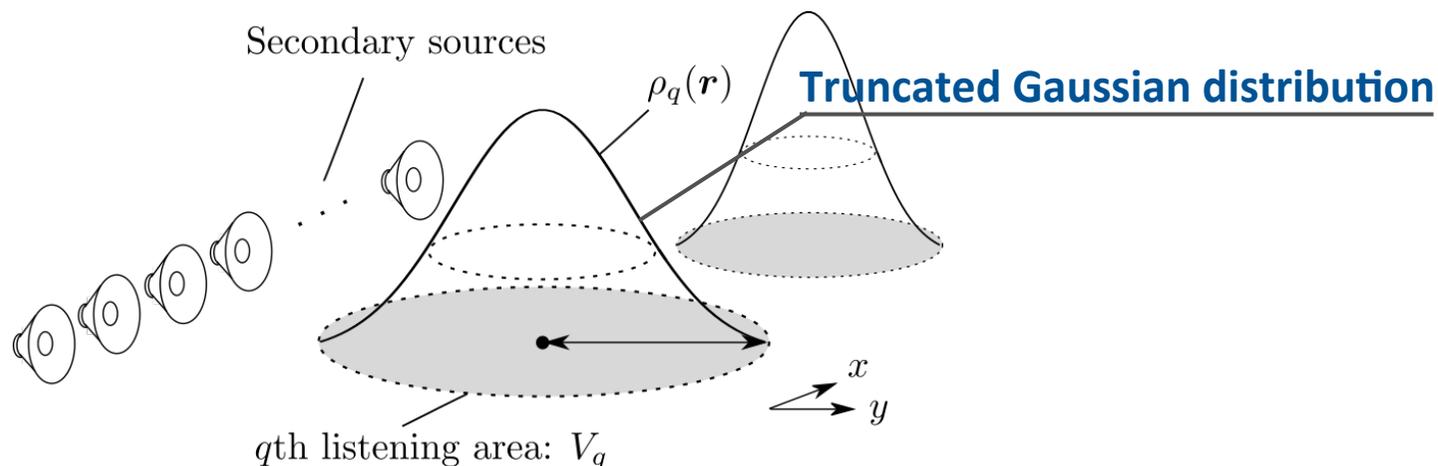
$$\underset{\underline{\mathbf{d}}}{\text{minimize}} \int_{\mathbf{r} \in V} \underline{\rho}(\mathbf{r}) \left| \underline{\mathbf{h}}(\mathbf{r})^{\top} \underline{\mathbf{d}} - p_{\text{des}}(\mathbf{r}) \right|^2 d\mathbf{r}$$

Given desired pressure

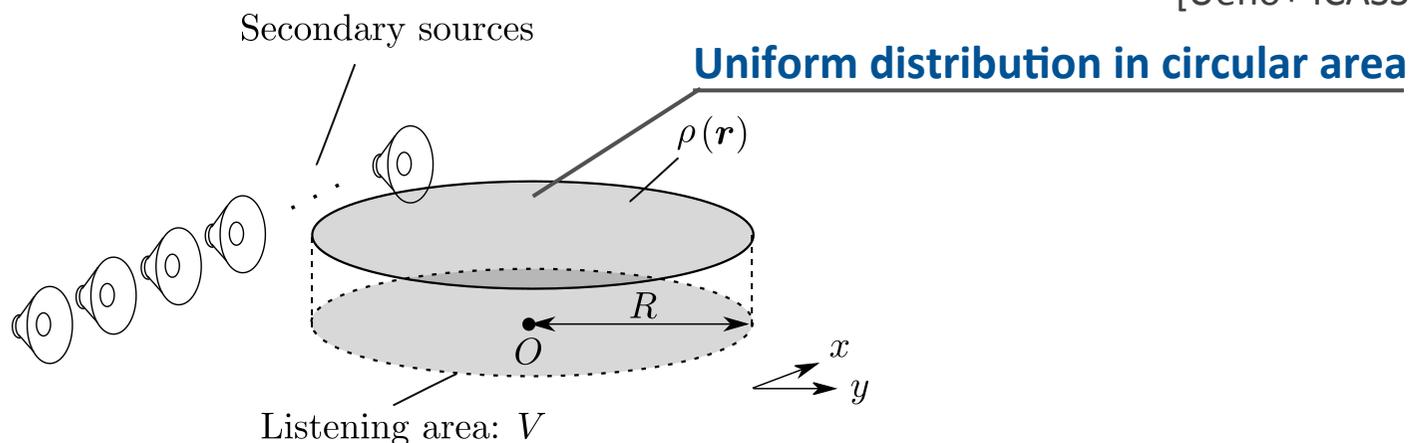
Prior distribution on the listeners' position inside listening area V

Problem statement

- $\rho(\mathbf{r})$ is a mixture of truncated Gaussian distributions [Ueno+ HSCMA 2017]



- $\rho(\mathbf{r})$ is a uniform distribution in single or multiple circular areas [Ueno+ ICASSP 2017]



Harmonic Expansion of Objective Function

- Objective function

$$\mathcal{J} = \int_{\mathbf{r} \in V} \rho(\mathbf{r}) \left| \mathbf{h}(\mathbf{r})^\top \mathbf{d} - p_{\text{des}}(\mathbf{r}) \right|^2 d\mathbf{r}$$

Difficult to minimize analytically due to multiple integrals

- Circular harmonic expansion

Basis functions: $\varphi_m(\mathbf{r}) = J_m(kr)e^{jm\phi}$ ($\mathbf{r} = (r, \phi)$: Polar coordinates)



$$\left\{ \begin{array}{l} \mathbf{h}(\mathbf{r}) \simeq \begin{bmatrix} c_{-M,1} & \cdots & c_{M,1} \\ \vdots & \ddots & \vdots \\ c_{-M,L} & \cdots & c_{M,L} \end{bmatrix} \begin{bmatrix} \varphi_{-M}(\mathbf{r}) \\ \vdots \\ \varphi_M(\mathbf{r}) \end{bmatrix} = \mathbf{C}^\top \underline{\varphi}(\mathbf{r}) \\ p_{\text{des}}(\mathbf{r}) \simeq \begin{bmatrix} b_{-M} & \cdots & b_M \end{bmatrix} \begin{bmatrix} \varphi_{-M}(\mathbf{r}) \\ \vdots \\ \varphi_M(\mathbf{r}) \end{bmatrix} = \mathbf{b}^\top \underline{\varphi}(\mathbf{r}) \end{array} \right.$$

Harmonic Expansion of Objective Function

➤ Objective function

$$\mathcal{J} = \int_{\mathbf{r} \in V} \underline{\rho(\mathbf{r})} \left| \underline{\mathbf{h}(\mathbf{r})}^\top \underline{\mathbf{d}} - p_{\text{des}}(\mathbf{r}) \right|^2 d\mathbf{r}$$



Approximation using circular harmonic expansion

$$\begin{cases} \underline{\mathbf{h}(\mathbf{r})} \simeq \mathbf{C}^\top \underline{\varphi(\mathbf{r})} \\ p_{\text{des}}(\mathbf{r}) \simeq \mathbf{b}^\top \underline{\varphi(\mathbf{r})} \end{cases}$$

$$\mathcal{J} \simeq (\mathbf{C}\underline{\mathbf{d}} - \mathbf{b})^\text{H} \underbrace{\left\{ \int_{\mathbf{r} \in V} \underline{\rho(\mathbf{r})} \underline{\varphi(\mathbf{r})}^* \underline{\varphi(\mathbf{r})}^\top d\mathbf{r} \right\}}_{\underline{\mathbf{W}}} (\mathbf{C}\underline{\mathbf{d}} - \mathbf{b})$$

Integrals including $\rho(\mathbf{r})$ and $\varphi(\mathbf{r})$ are only required to be calculated

Harmonic Expansion of Objective Function

- Each element of **W** can be analytically calculated as

$$w_{m,n} = \int_{\mathbf{r} \in V} \underline{\rho(\mathbf{r})} \underline{\varphi_m(\mathbf{r})}^* \underline{\varphi_n(\mathbf{r})} d\mathbf{r}$$

(m,n) element
of **W**

$$= \int_{\mathbf{r} \in V} J_m(kr) e^{-jm\phi} J_n(kr) e^{jn\phi} d\mathbf{r}$$

$$= \int_0^R J_m(kr) J_n(kr) r dr \int_0^{2\pi} e^{j(n-m)\phi} d\phi$$

$$= \delta_{m,n} \pi R^2 \{ J_m(kR)^2 - J_{m-1}(kR) J_{m+1}(kR) \}$$

W is a diagonal matrix with positive value

- **W** can be analytically calculated for uniform distribution
- Similar results can be obtained for truncated Gaussian distribution

Optimal Driving Signals

- Objective function using circular harmonic expansion

$$\mathcal{J} \simeq (\mathbf{C}\underline{\mathbf{d}} - \mathbf{b})^H \underbrace{\left\{ \int_{\mathbf{r} \in V} \underline{\rho(\mathbf{r})} \underline{\varphi(\mathbf{r})}^* \underline{\varphi(\mathbf{r})}^T d\mathbf{r} \right\}}_{\underline{\mathbf{W}}} (\mathbf{C}\underline{\mathbf{d}} - \mathbf{b})$$

- **Optimal driving signals**

$$\underline{\hat{\mathbf{d}}} = (\mathbf{C}^H \underline{\mathbf{W}} \mathbf{C} + \lambda \mathbf{I})^{-1} \mathbf{C}^H \underline{\mathbf{W}} \mathbf{b}$$

- Objective function is simply minimized and optimal driving signals can be obtained

Relationship with mode-matching method

➤ Proposed method

$$\hat{\underline{\mathbf{d}}} = (\mathbf{C}^H \underline{\mathbf{W}} \mathbf{C} + \lambda \mathbf{I})^{-1} \mathbf{C}^H \underline{\mathbf{W}} \mathbf{b}$$

- Circular harmonics are optimally weighted based on prior information

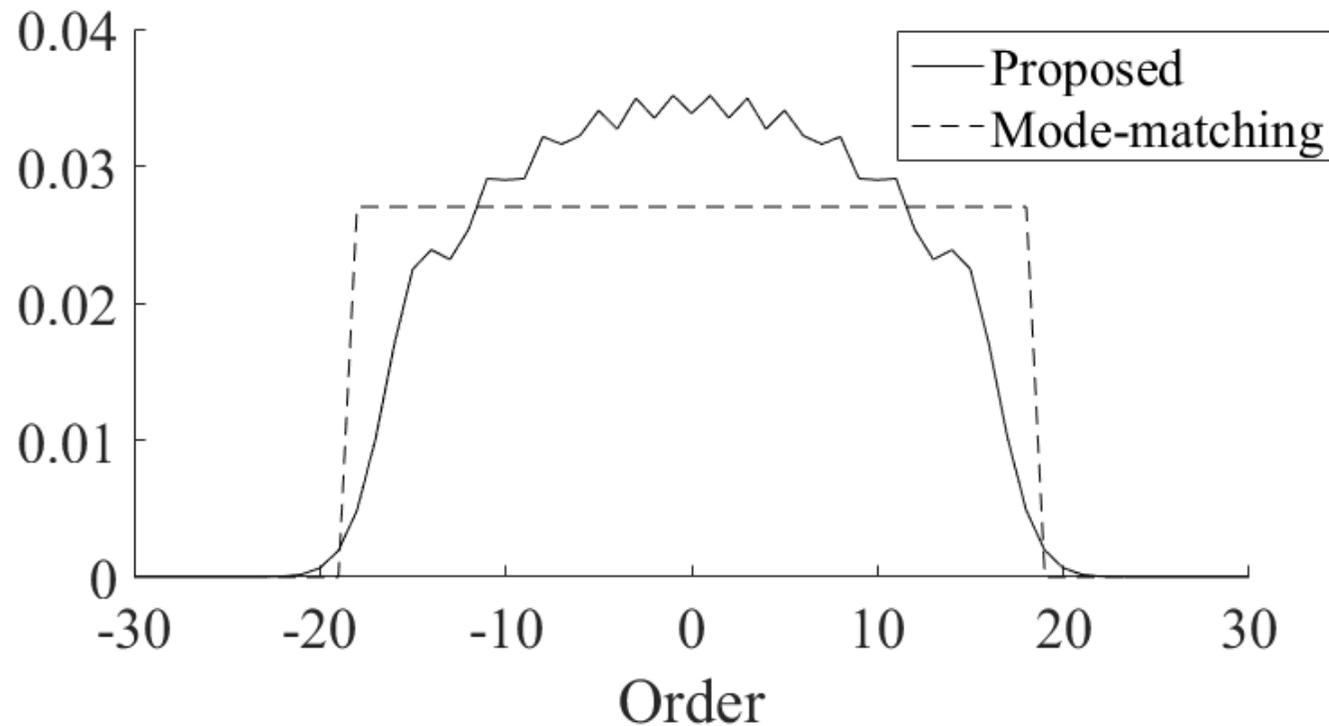
➤ Mode-matching method: least squares solution of $\mathbf{C} \underline{\mathbf{d}} = \mathbf{b}$

$$\hat{\underline{\mathbf{d}}} = (\mathbf{C}^H \mathbf{C} + \lambda \mathbf{I})^{-1} \mathbf{C}^H \mathbf{b}$$

- Circular harmonics have to be truncated at appropriate order
- Truncation at $M = \lceil kR \rceil$ is empirically known to give high performance within circular region of radius R

Relationship with mode-matching method

- Weight on circular harmonics when $k = 36.9$ rad/m and $R = 0.4$ m



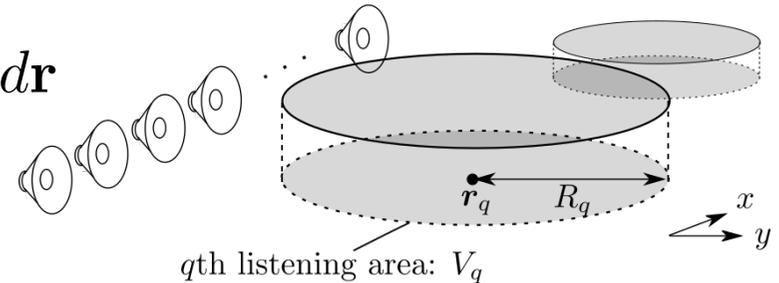
Optimal weight on circular harmonics based on prior information

Extension to Multiple Listening Areas

➤ Objective function

$$\mathcal{J} = \sum_{q=1}^Q \int_{\mathbf{r} \in V_q} \rho(\mathbf{r}) \left| \mathbf{h}(\mathbf{r})^T \mathbf{d} - p_{\text{des}}(\mathbf{r}) \right|^2 d\mathbf{r}$$

qth listening area ($q = 1, \dots, Q$)



Approximation using circular harmonic expansion

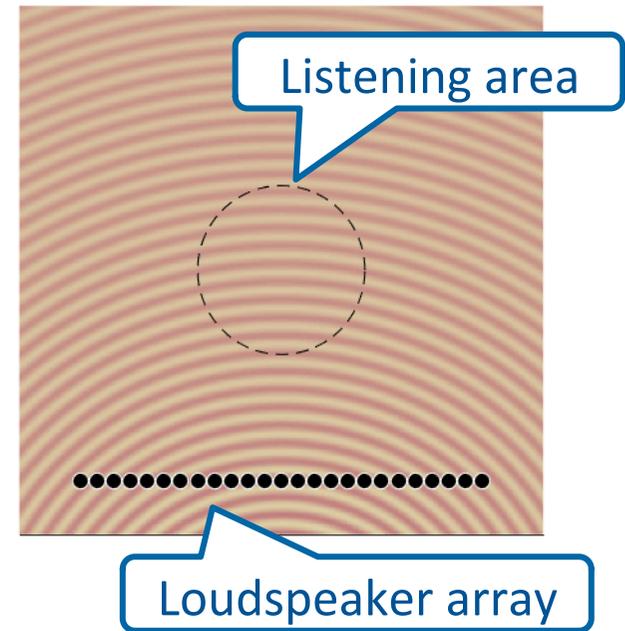
$$\mathcal{J} \simeq \sum_{q=1}^Q \left(\mathbf{C}^{(q)} \mathbf{d} - \mathbf{b}^{(q)} \right)^H \mathbf{W}^{(q)} \left(\mathbf{C}^{(q)} \mathbf{d} - \mathbf{b}^{(q)} \right)$$

➤ Optimal driving signals

$$\hat{\mathbf{d}} = \left(\sum_{q=1}^Q \mathbf{C}^{(q)H} \mathbf{W}^{(q)} \mathbf{C}^{(q)} + \lambda \mathbf{I} \right)^{-1} \sum_{q=1}^Q \mathbf{C}^{(q)H} \mathbf{W}^{(q)} \mathbf{b}^{(q)}$$

Simulation Experiment

- Array geometry
 - Linear: 25 loudspeakers, 0.16 m intervals
 - Circular: 64 loudspeakers, 2.0 m radius
- Desired sound field: cylindrical wave
- Listening area: two circular areas
- Compared method:
 - **Proposed**
 - **MM**: Mode-matching method
 - **CD**: Continuous distribution method (WFS/HOA) [Spors+ 2008, Poletti 2005]
 - **CD w/ BL**: CD with band limitation [Ahrens+ 2009, 2011]



- Evaluation: Signal-to-Distortion Ratio of Reproduction (SDRR)
 - Original pressure distribution

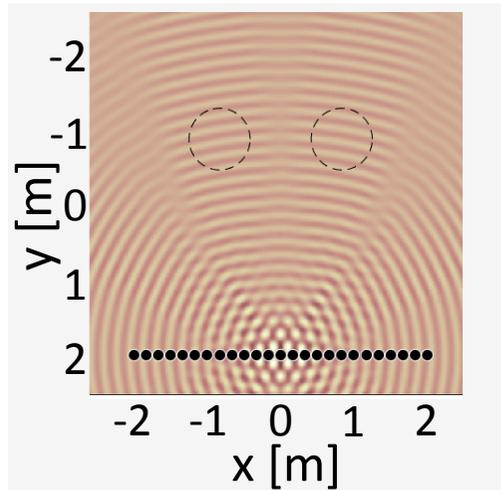
$$\text{SDRR} = 10 \log_{10} \frac{\sum_i \sum_j \sum_k |\bar{p}_{\text{org}}(x_i, y_j, t_k)|^2}{\sum_i \sum_j \sum_k |\bar{p}_{\text{org}}(x_i, y_j, t_k) - \bar{p}_{\text{rep}}(x_i, y_j, t_k)|^2}$$

Reproduced pressure distribution

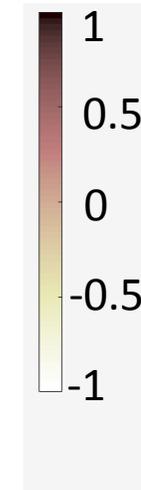
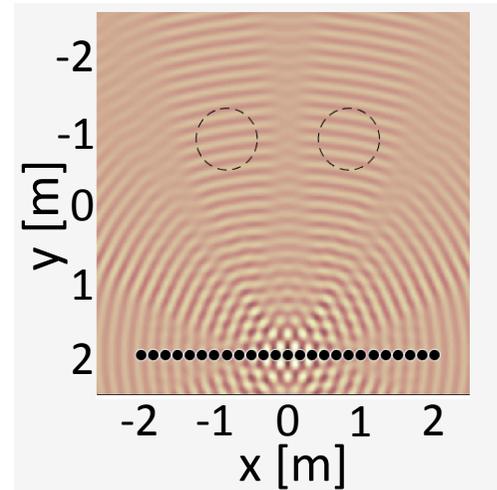
Pressure distribution (2 kHz, linear array)

Sound pressure distribution

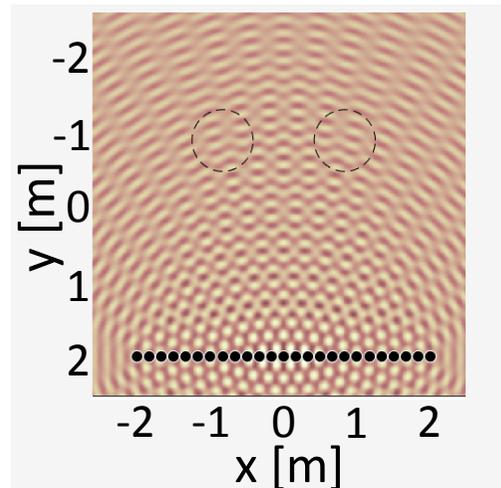
Proposed (SDRR = 68.51 dB)



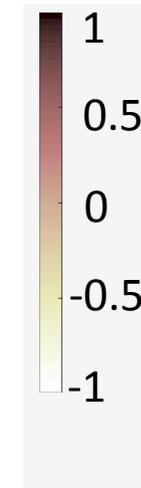
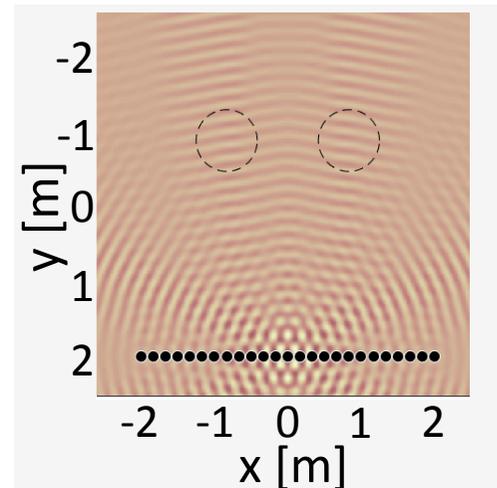
MM (SDRR = 60.42 dB)



CD (SDRR = -0.48 dB)

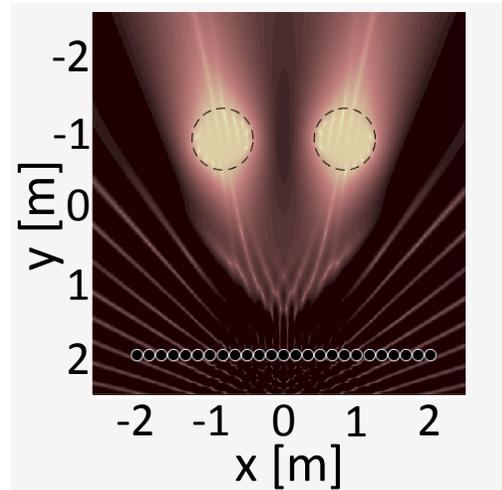


CD w/ BL (SDRR = 5.81 dB)

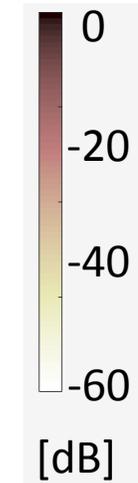
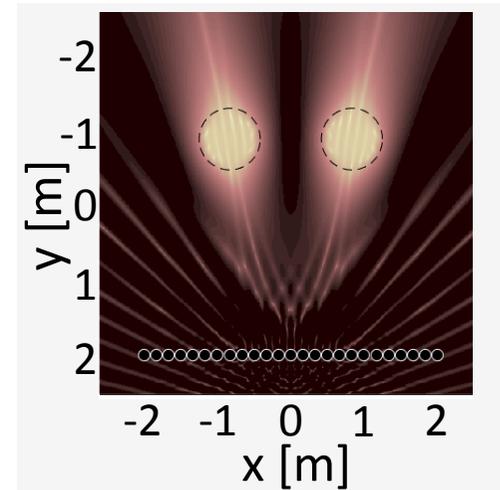


Error distribution (2 kHz, linear array)

Proposed (SDRR = 68.51 dB)

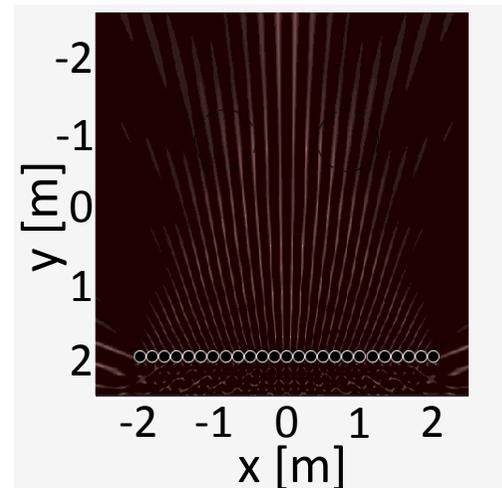


MM (SDRR = 60.42 dB)

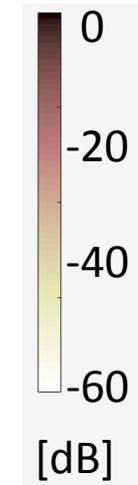
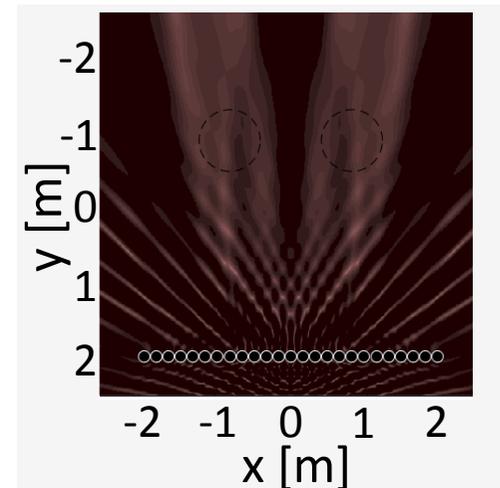


Normalized error distribution

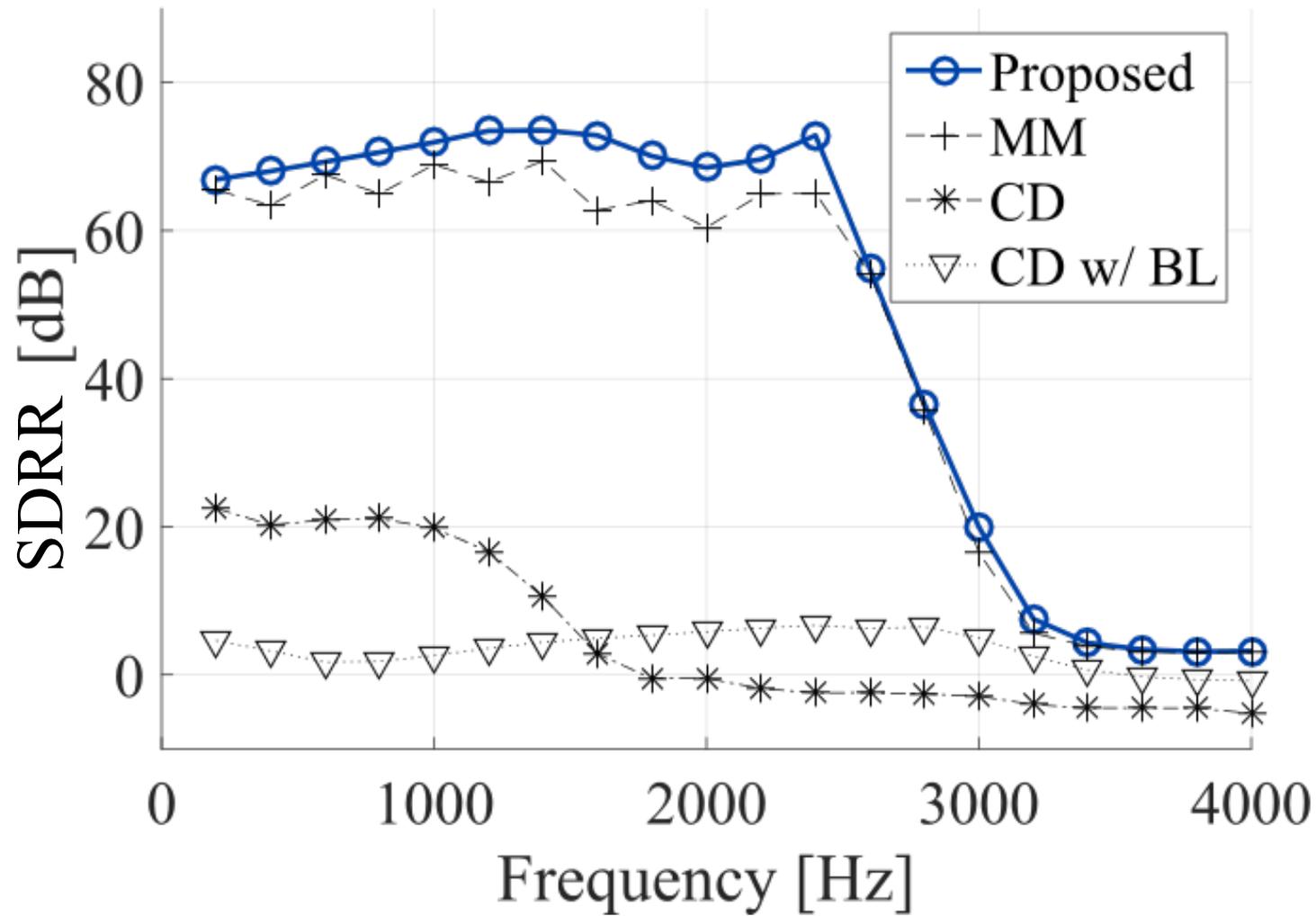
CD (SDRR = -0.48 dB)



CD w/ BL (SDRR = 5.81 dB)



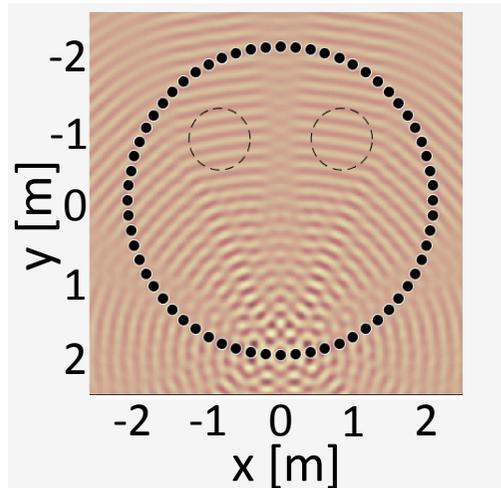
Frequency vs. SDRR (linear array)



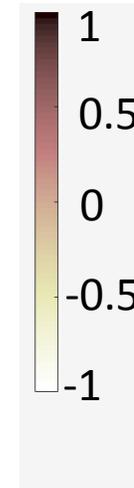
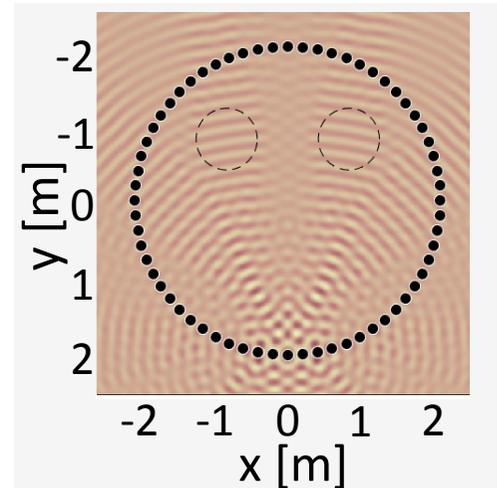
Pressure distribution (2 kHz, circular array)

Sound pressure distribution

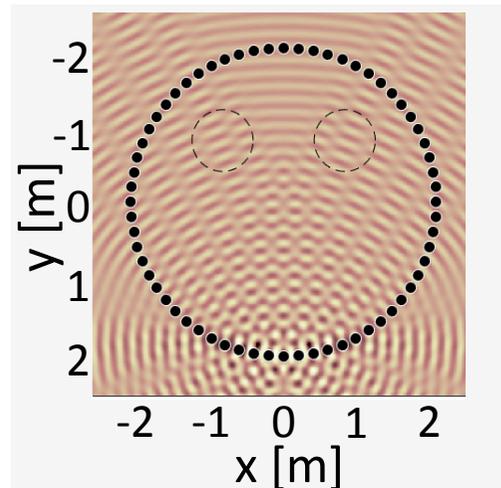
Proposed (SDRR = 66.58 dB)



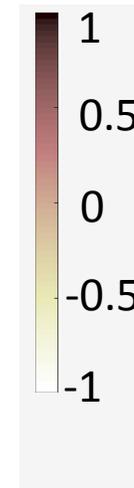
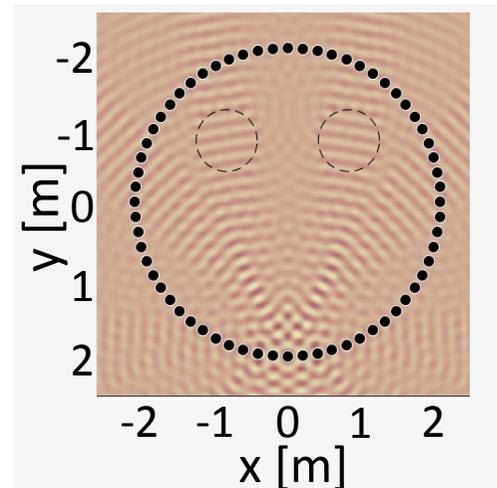
MM (SDRR = 39.63 dB)



CD (SDRR = 0.65 dB)

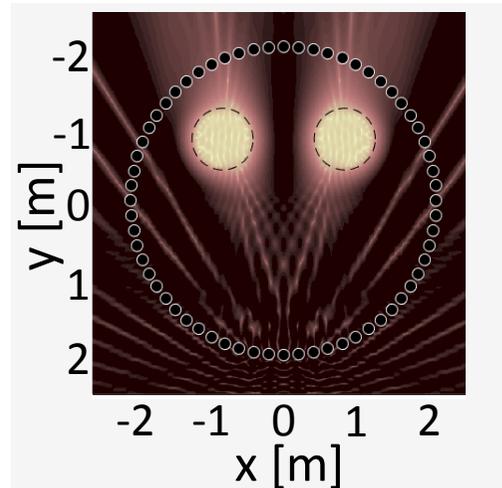


CD w/ BL (SDRR = 17.79 dB)

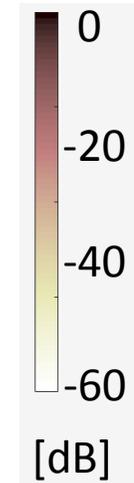
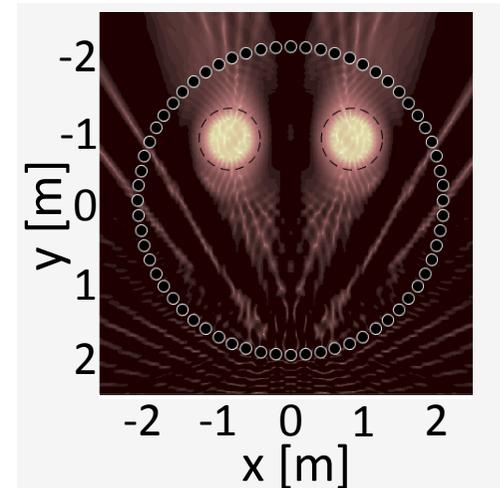


Error distribution (2 kHz, circular array)

Proposed (SDRR = 66.58 dB)

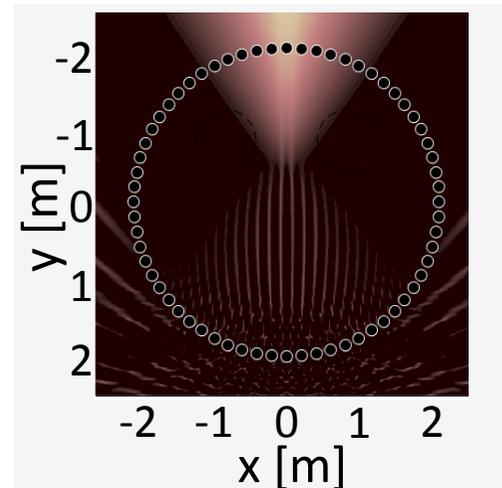


MM (SDRR = 39.63 dB)

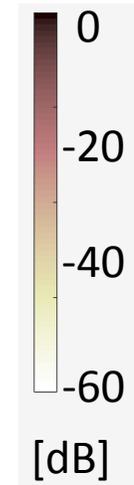
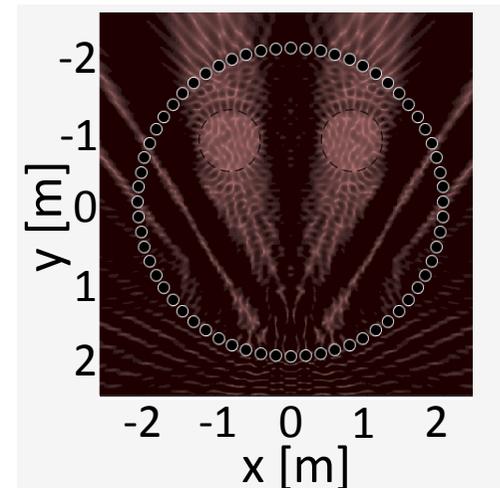


Normalized error distribution

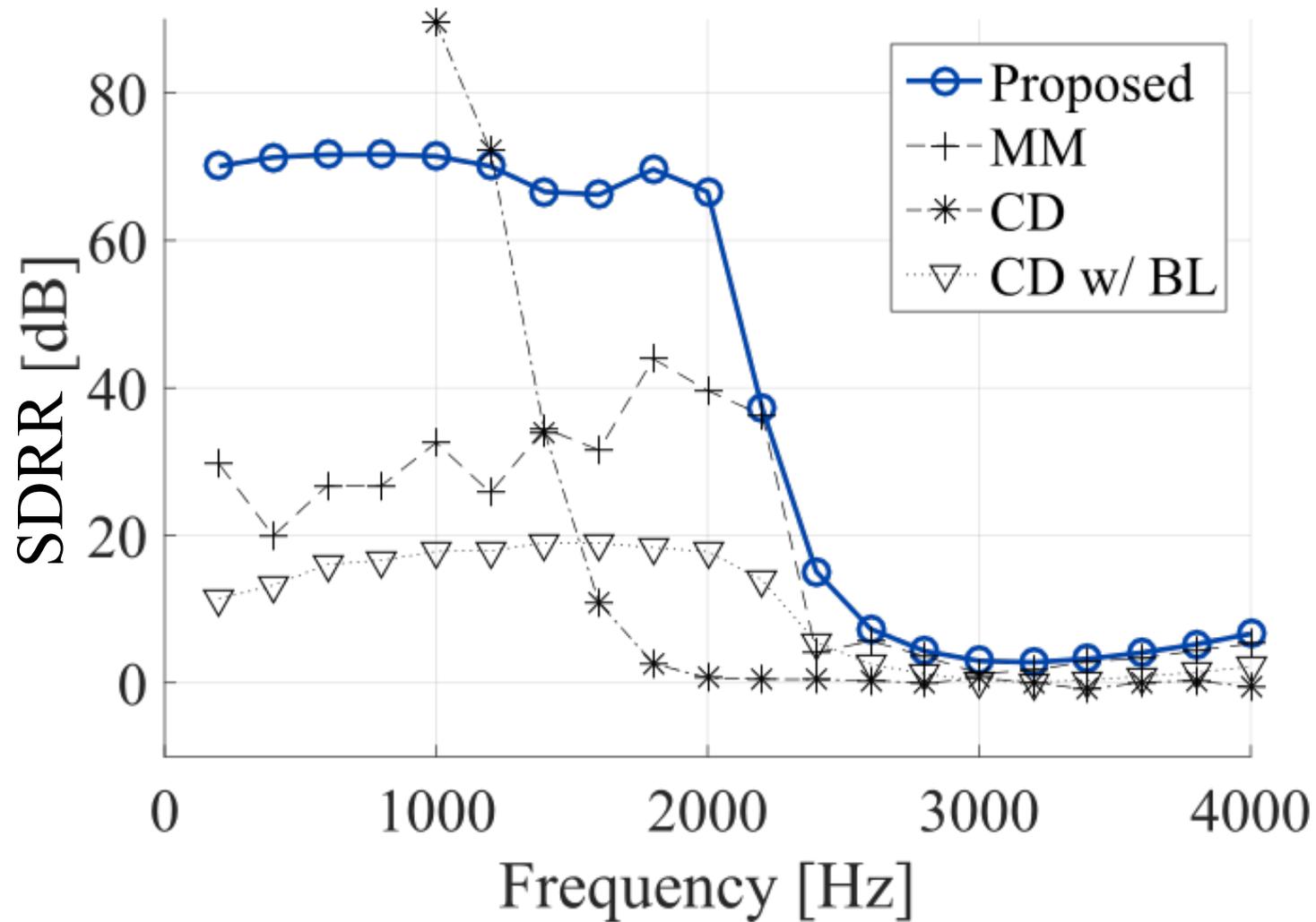
CD (SDRR = 0.65 dB)



CD w/ BL (SDRR = 17.79 dB)



Frequency vs. SDRR (circular array)



Reduction of The Number of Loudspeakers

- Sound field reproduction exploiting prior information on listening area
- Objective function is formulated as expectation minimization of spatial squared error inside listening areas
- Optimal driving signals are obtained by circular harmonic expansion
- Optimal weighting of circular harmonics can be analytically calculated based on prior information
- Experimental results indicated that high reproduction accuracy can be achieved by using the proposed method

Conclusion

- **Reduction of microphones and loudspeakers in sound field recording and reproduction**
 - Improve reproduction accuracy exploiting prior information
 - Near-field source area and source sparsity for reducing microphones
 - Probability distribution on listeners' position for reducing loudspeakers
 - These two methods can be combined

Thank you for your attention!

Related publications

- S. Koyama, *et al.* “Effect of multipole dictionary in sparse sound field decomposition for super-resolution in recording and reproduction,” *Proc. ICSV*, 2017 (to appear).
- N. Murata, S. Koyama, *et al.* “Spatio-temporal sparse sound field decomposition considering acoustic source signal characteristics,” *Proc. IEEE ICASSP*, 2017.
- N. Ueno, S. Koyama, and H. Saruwatari, “Listening-area-informed sound field reproduction based on circular harmonic expansion,” *Proc. IEEE ICASSP*, 2017.
- N. Ueno, S. Koyama, and H. Saruwatari, “Listening-area-informed sound field reproduction with Gaussian prior based on circular harmonic expansion,” *Proc. HSCMA*, 2017.
- S. Koyama and H. Saruwatari, “Sound field decomposition in reverberant environment using sparse and low-rank signal models,” *Proc. IEEE ICASSP*, 2016.
- N. Murata, S. Koyama, *et al.* “Sparse sound field decomposition with multichannel extension of complex NMF,” *Proc. IEEE ICASSP*, 2016.
- S. Koyama, *et al.* “Sparse sound field decomposition using group sparse Bayesian learning,” *Proc. APSIPA ASC*, 2015.
- N. Murata, S. Koyama, *et al.* “Sparse sound field decomposition with parametric dictionary learning for super-resolution recording and reproduction,” *Proc. IEEE CAMSAP*, 2015.
- S. Koyama, *et al.* “Structured sparse signal models and decomposition algorithm for super-resolution in sound field recording and reproduction,” *Proc. IEEE ICASSP*, 2015.
- S. Koyama, *et al.* “Sparse sound field representation in recording and reproduction for reducing spatial aliasing artifacts,” *Proc. IEEE ICASSP*, 2014.