

# Sound Field Recording and Reproduction and Its Extension to Super-resolution

**Shoichi Koyama**

Graduate school of information science and technology,  
The University of Tokyo



# About me

## ➤ Shoichi Koyama, Ph.D.

– 2007 B.E. and 2009 M.S. degrees from UTokyo

- *Source localization with weighted integral method*

– 2009.04 – 2014.03 NTT Media Intelligence Lab.

- *Sound field recording and reproduction*



– 2014.01 Ph.D. (Inf. Sci.&Tech.) from UTokyo

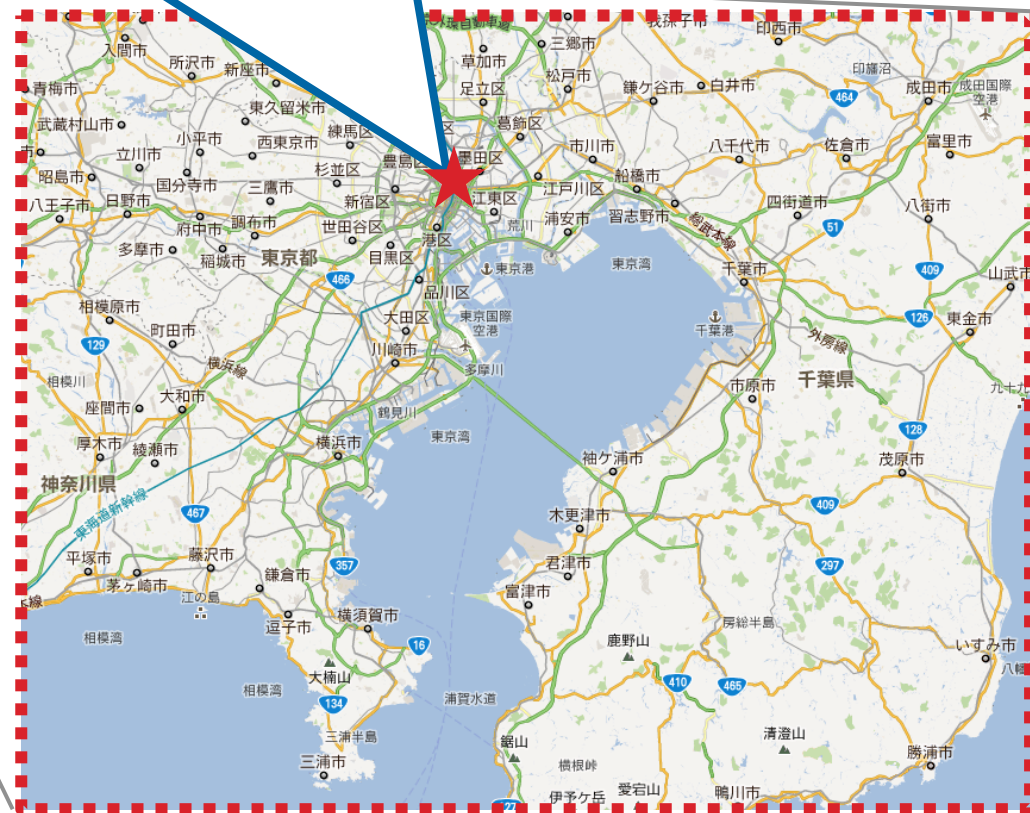
– 2014.04 – Assistant Prof. (Research Associate) at UTokyo

- *Super-resolution in sound field recording and reproduction*



# System #1 Lab., The University of Tokyo

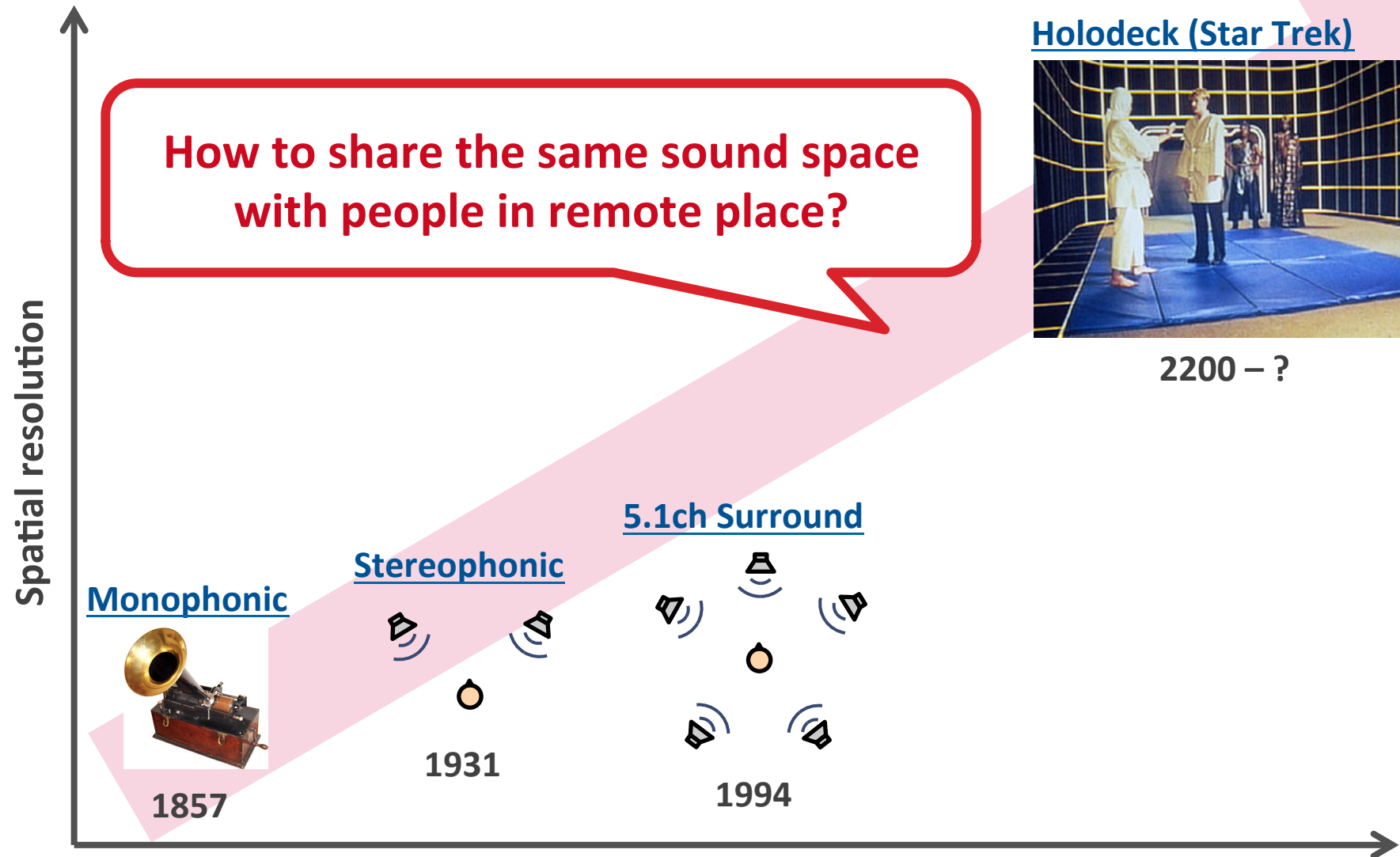
- Our lab has just started on Apr. 2014
- Staffs: Prof. Hiroshi Saruwatari & me
  - 6 students
  - Audio, speech, and music signal processing



April 17, 2015

# Recording and Reproduction

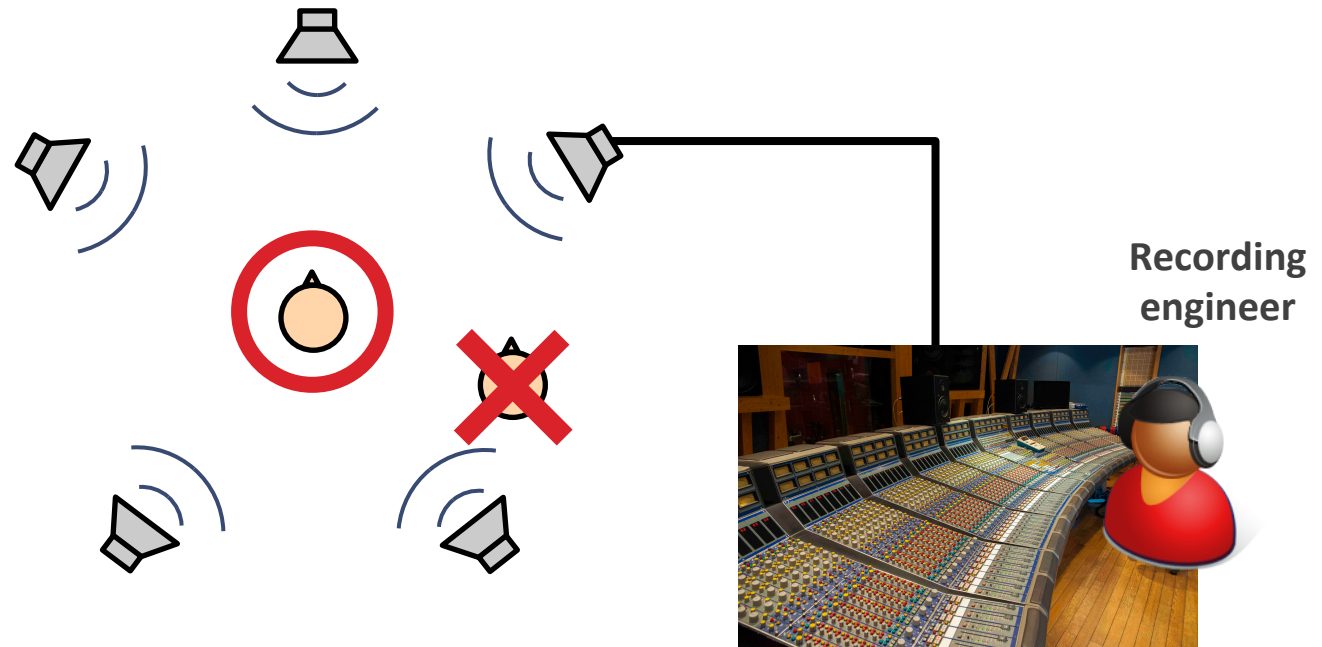
## History of audio reproduction system





# Conventional audio systems

- Stereo, 5.1ch surround, etc...

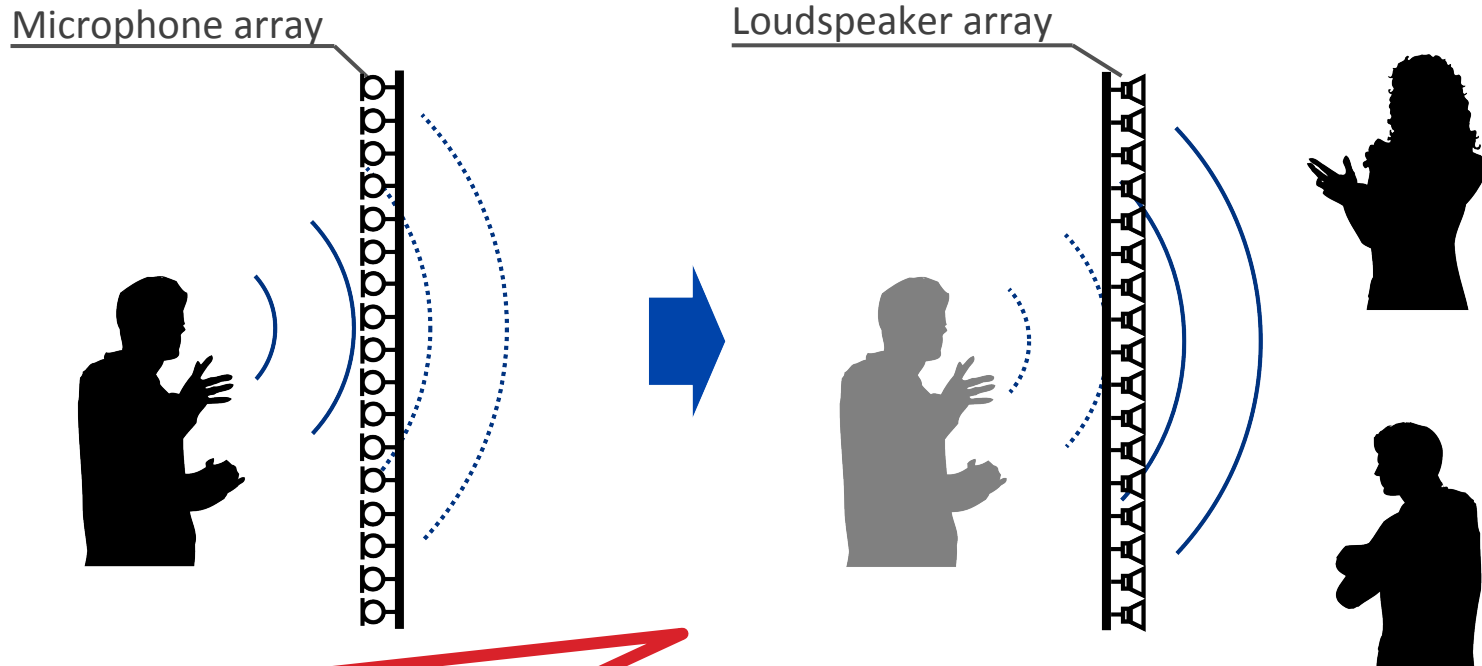


## Drawbacks

- ☹ Listening position is limited around center of loudspeakers (sweet spot)
- ☹ Reproduced sound is artificially designed by recording engineers

# Sound field reproduction for audio system

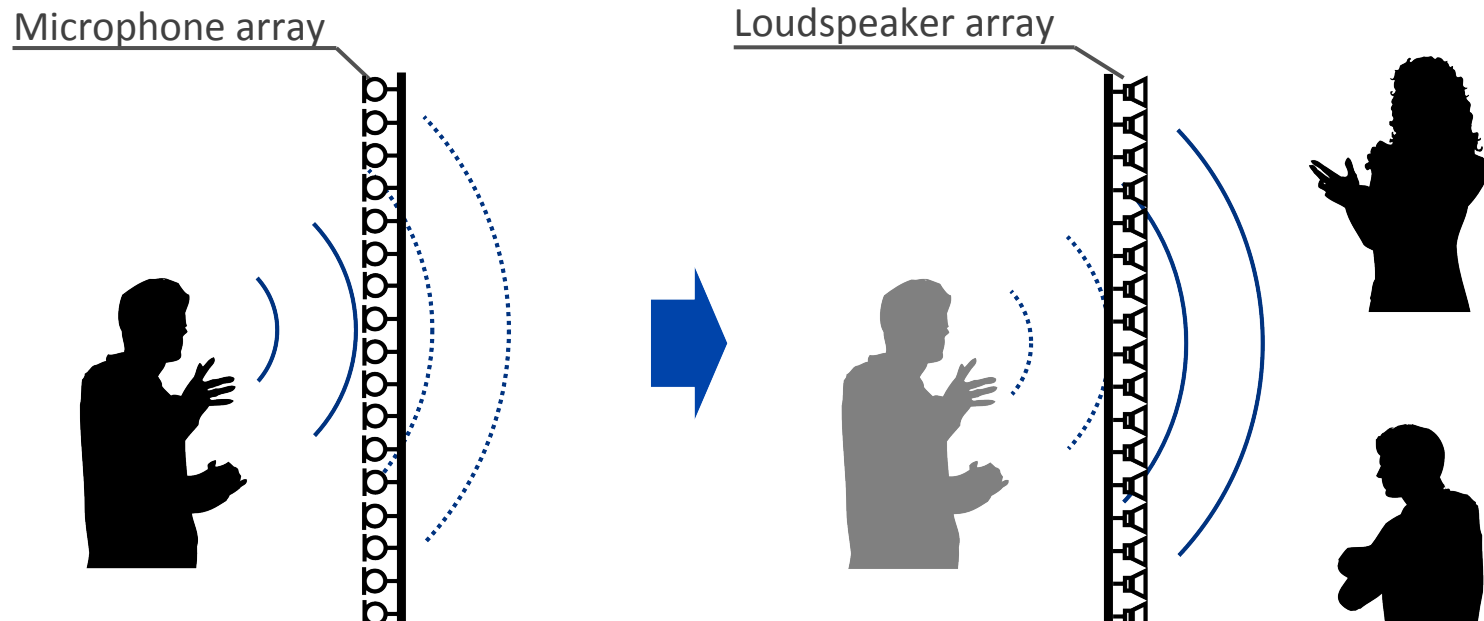
## Sound field recording and reproduction



- ☐ Large listening area can be achieved
- ☐ Listeners can perceive source distance
- ☐ Real-time recording and reproduction can be achieved without recording engineers

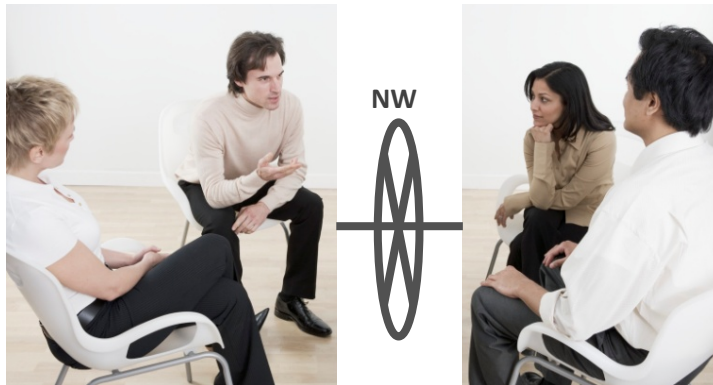
# Sound field reproduction for audio system

## Sound field recording and reproduction

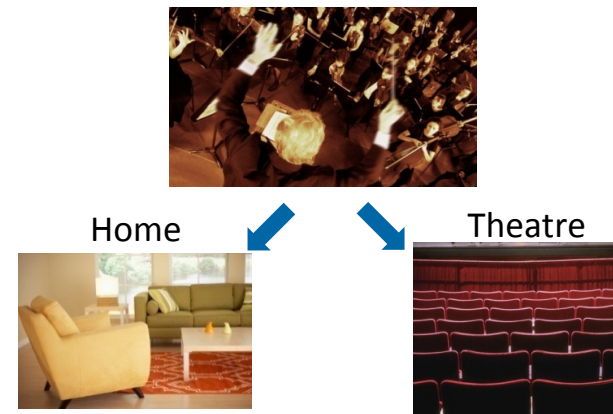


## Applications

### Telecommunication system



### Live broadcasting



# Outline

## ➤ **Signal Conversion for Sound Field Recording and Reproduction**

- Signal conversion from signals received by microphone array into driving signals of loudspeakers for reconstruction of sound field
- Fast and stable signal conversion method for sound field recording and reproduction is proposed

## ➤ **Super-resolution in Sound Field Recording and Reproduction**

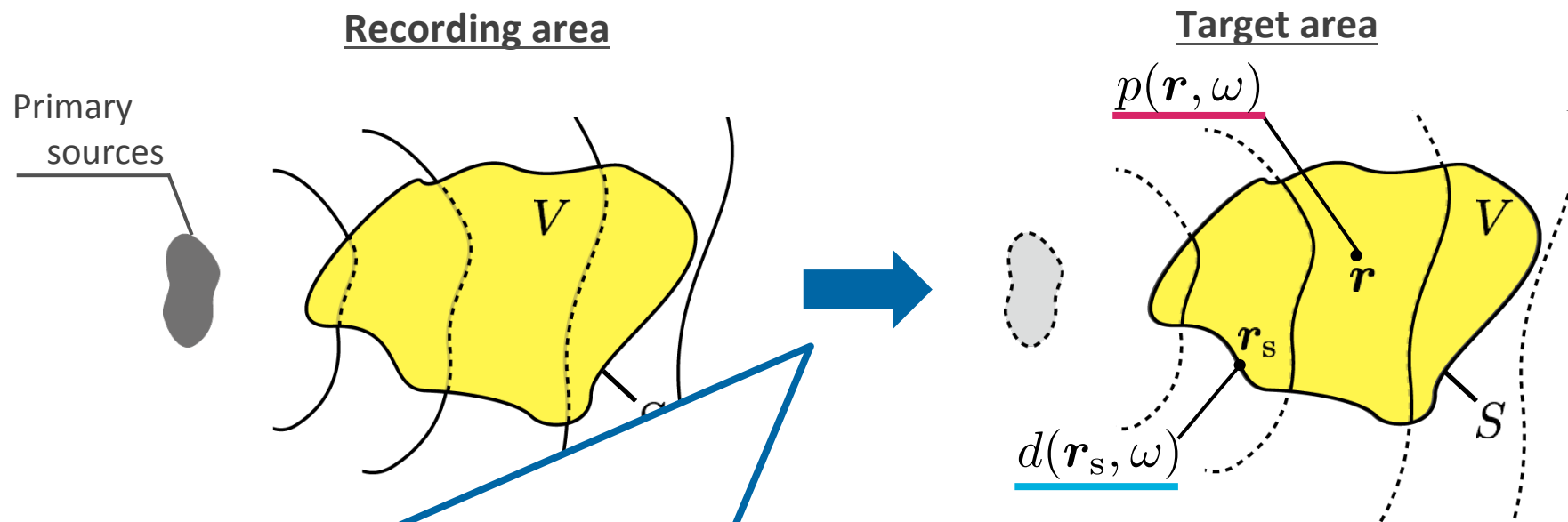
- Improve reproduction accuracy at frequencies above spatial Nyquist frequency determined by intervals of array elements
- Two approaches of super-resolution are introduced



# **SIGNAL CONVERSION FOR SOUND FIELD RECORDING AND REPRODUCTION**

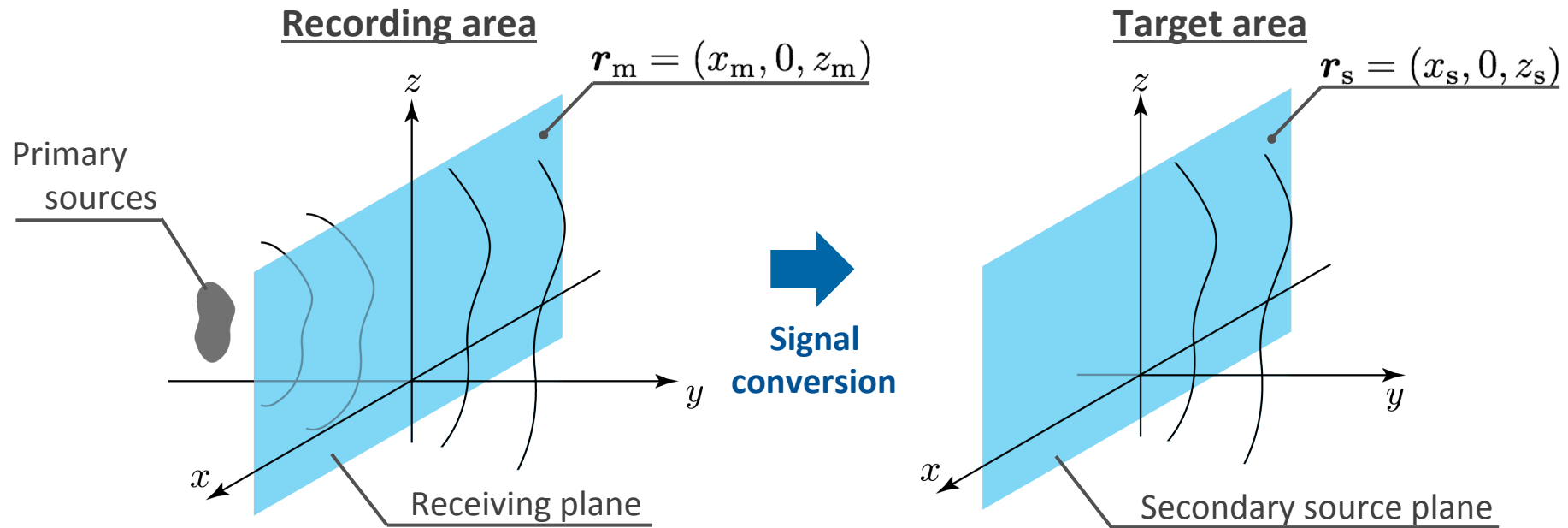
# Sound Field Recording and Reproduction

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



- ✧ Inherently, sound pressure and its gradient on  $S$  is required to obtain  $\underline{d(\mathbf{r}_s, \omega)}$ , but sound pressure is usually only known
- ✧ Our objective: Fast and stable signal conversion for sound field recording and reproduction with ordinary acoustic sensors and transducers

# Recording and Reproduction with Planar Arrays



Desired sound field based on Rayleigh integral

$$p_{\text{des}}(\mathbf{r}, \omega) = -2 \iint_{-\infty}^{\infty} \frac{\partial p(\mathbf{r}_m, \omega)}{\partial y_m} \underbrace{G(\mathbf{r} | \mathbf{r}_m, \omega)}_{\text{Green's function}} dx_m dz_m$$

Synthesized sound field by secondary sources

$$p_{\text{syn}}(\mathbf{r}, \omega) = \iint_{-\infty}^{\infty} \underbrace{d(\mathbf{r}_s, \omega)}_{\text{Transfer function}} \underbrace{H(\mathbf{r} - \mathbf{r}_s, \omega)}_{\text{Transfer function}} dx_s dz_s$$

Approximate secondary sources as point source ( $\underline{H} \simeq \underline{G}$ )

$$\underline{d}(\mathbf{r}_s, \omega) = -2 \frac{\partial p(\mathbf{r}_m, \omega)}{\partial y_m}$$

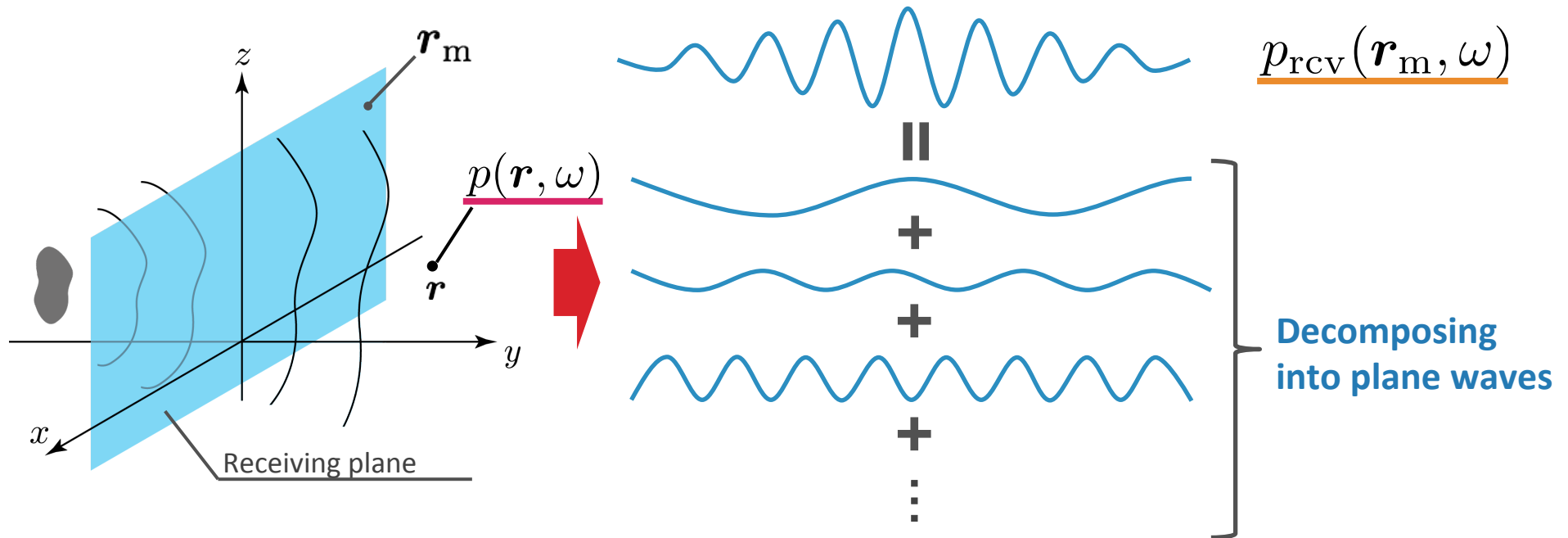
⊗  $\underline{p}(\mathbf{r}_m, \omega)$  is only known

Distribution of sound pressure gradient needs to be estimated



# Formulation in Spatial Frequency Domain

Sound pressure distribution is represented as a sum of plane waves



Since each plane wave uniquely determines propagation in  $y$  direction, sound pressure gradient can be estimated

➡ Plane wave decomposition corresponds to spatial Fourier transform

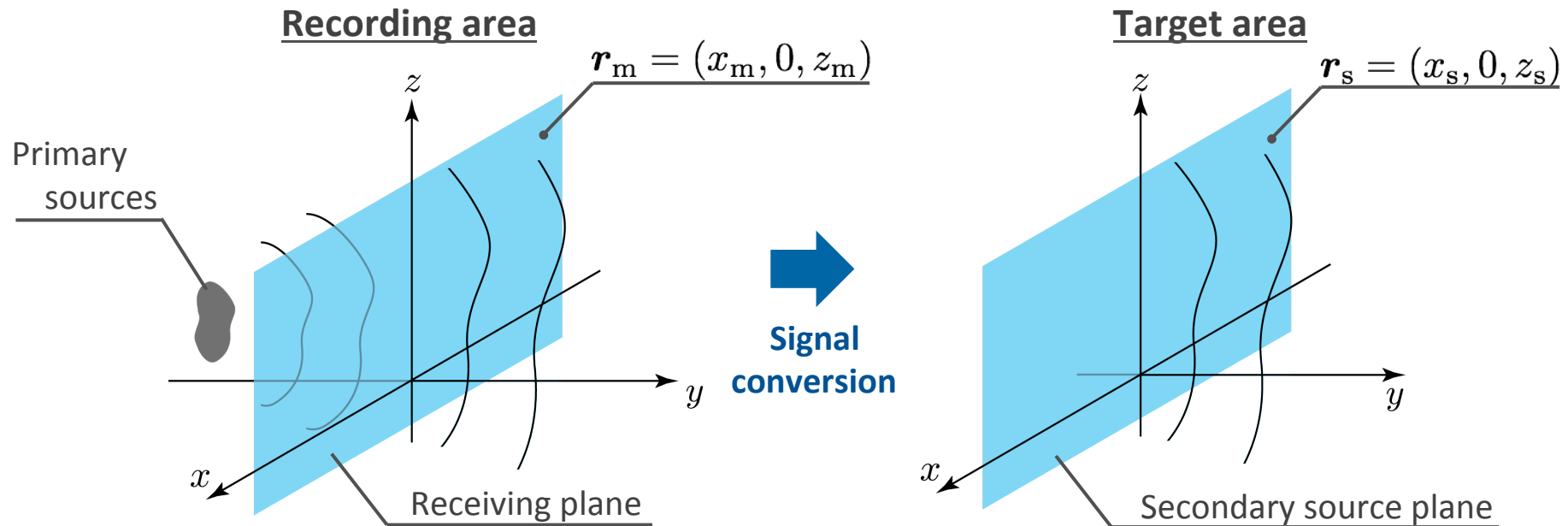
$$\underline{p(\mathbf{r}, \omega)} = \frac{1}{4\pi^2} \int_{-\infty}^{\infty} dk_x \int_{-\infty}^{\infty} dk_z \underline{P_{\text{rcv}}(k_x, 0, k_z, \omega)} e^{j(k_x x + k_y y + k_z z)}$$

Spatial frequency spectrum  
on receiving plane

$$(k_y = \pm \sqrt{k^2 - k_x^2 - k_z^2})$$

# Signal Conversion in Spatial Freq Domain

[Koyama+ IEEE TASLP 2013]



**Desired sound field based on Rayleigh integral**

$$\underline{p_{\text{des}}(\mathbf{r}, \omega)} = -2 \iint_{-\infty}^{\infty} \frac{\underline{\partial p(\mathbf{r}_m, \omega)}}{\partial y_m} \underline{G(\mathbf{r}|\mathbf{r}_m, \omega)} dx_m dz_m$$

Green's function

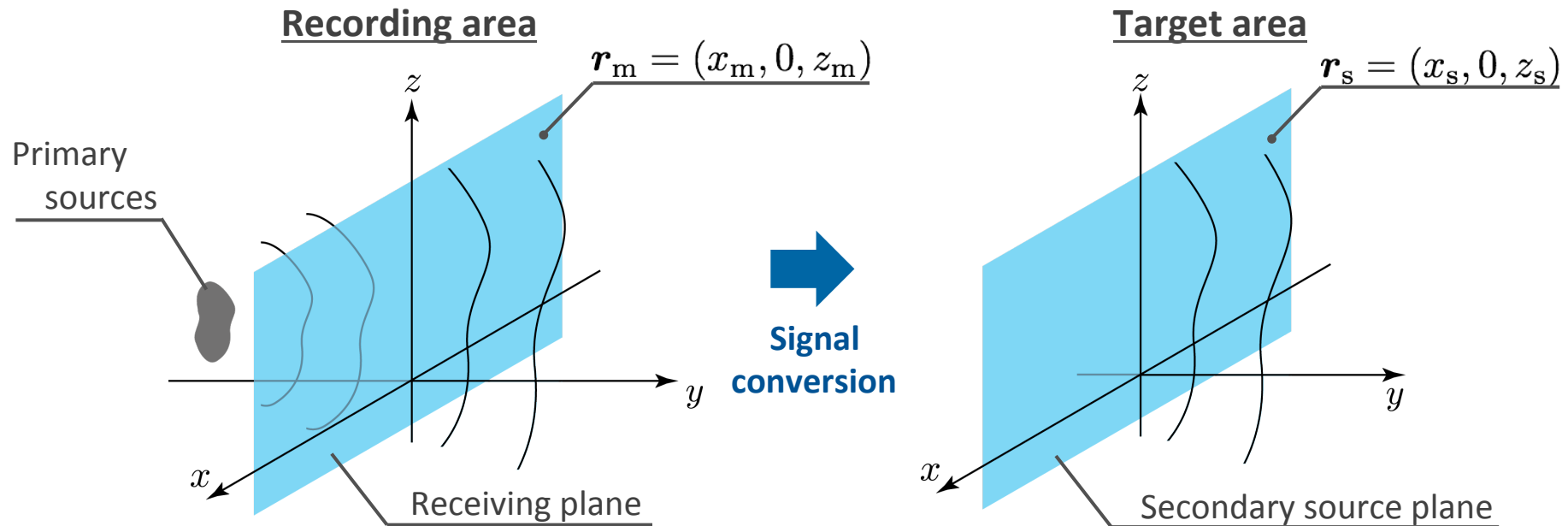
**Synthesized sound field by secondary sources**

$$\underline{p_{\text{syn}}(\mathbf{r}, \omega)} = \iint_{-\infty}^{\infty} \underline{d(\mathbf{r}_s, \omega)} \underline{H(\mathbf{r} - \mathbf{r}_s, \omega)} dx_s dz_s$$

Transfer function

# Signal Conversion in Spatial Freq Domain

[Koyama+ IEEE TASLP 2013]



Desired sound field based on Rayleigh integral in spatial freq domain

$$\underline{P_{\text{des}}(k_x, y, k_z, \omega)} = -2jk_y \underline{P_{\text{rcv}}(k_x, 0, k_z, \omega)} \underline{G(k_x, y, k_z, \omega)}$$

Green's function

Synthesized sound field by secondary sources in spatial freq domain

$$\underline{P_{\text{syn}}(k_x, y, k_z, \omega)} = \underline{D(k_x, k_z, \omega)} \underline{H(k_x, y, k_z, \omega)}$$

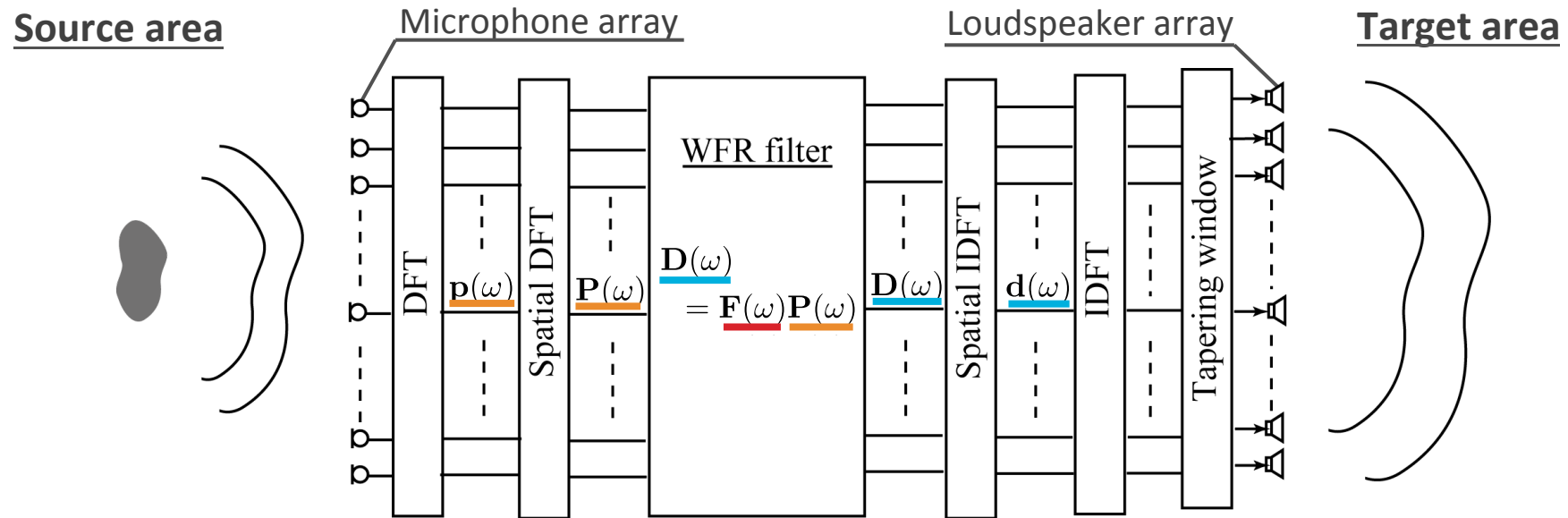
Transfer function

$$\underline{D(k_x, k_z, \omega)} = -2jk_y \underline{P_{\text{rcv}}(k_x, 0, k_z, \omega)}$$

$d(\mathbf{r}_s, \omega)$  can be  
obtained only  
from  $p_{\text{rcv}}(\mathbf{r}_m, \omega)$ !

# Wave Field Reconstruction (WFR) Filtering

**Wave Field Reconstruction (WFR) Filter:** Driving signal can be obtained by applying spatio-temporal filter [Koyama+ IEEE TASLP2013]



➤ Linear arrays of microphones and loudspeakers: 2D Convolution of **WFR filter**  $\mathbf{F}(\omega)$

$$\mathbf{D}(\omega) = \mathbf{F}(\omega)\mathbf{P}(\omega)$$

Driving signals in spatial freq domain

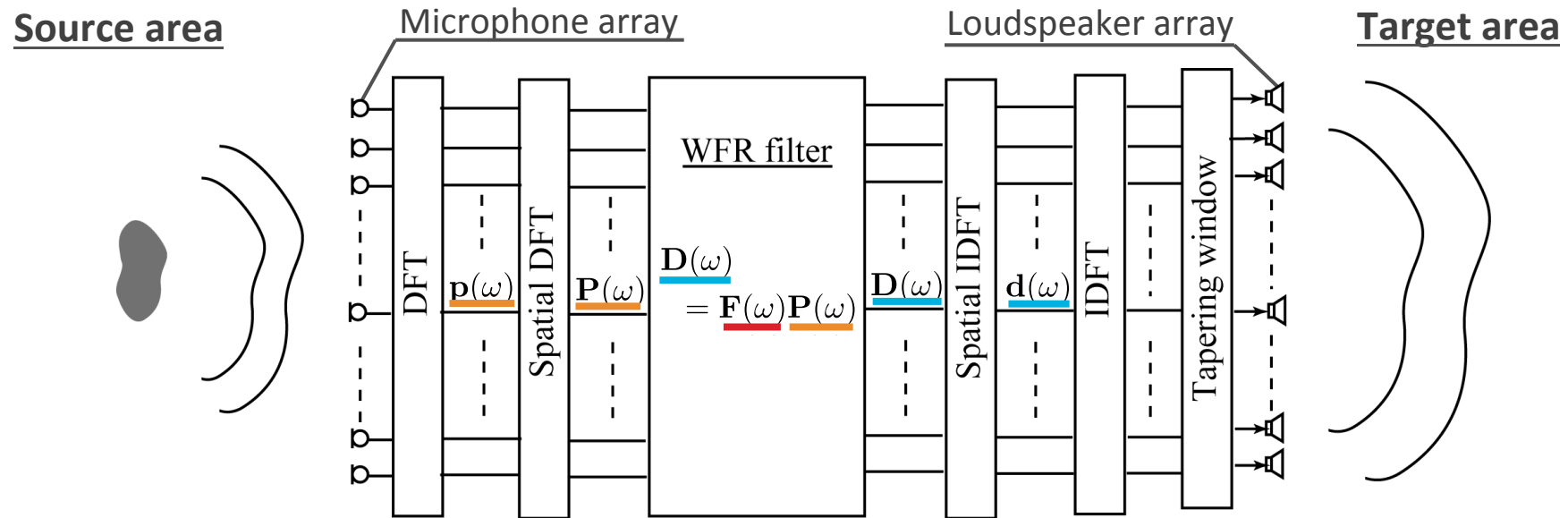
Diagonal matrix consisting of

$$F_i(\omega) = -4j \frac{\exp\left(\sqrt{k^2 - k_{x,i}^2} y_{\text{ref}}\right)}{H_0^{(1)}\left(\sqrt{k^2 - k_{x,i}^2} y_{\text{ref}}\right)}$$

Received signals in spatial freq domain

# Wave Field Reconstruction (WFR) Filtering

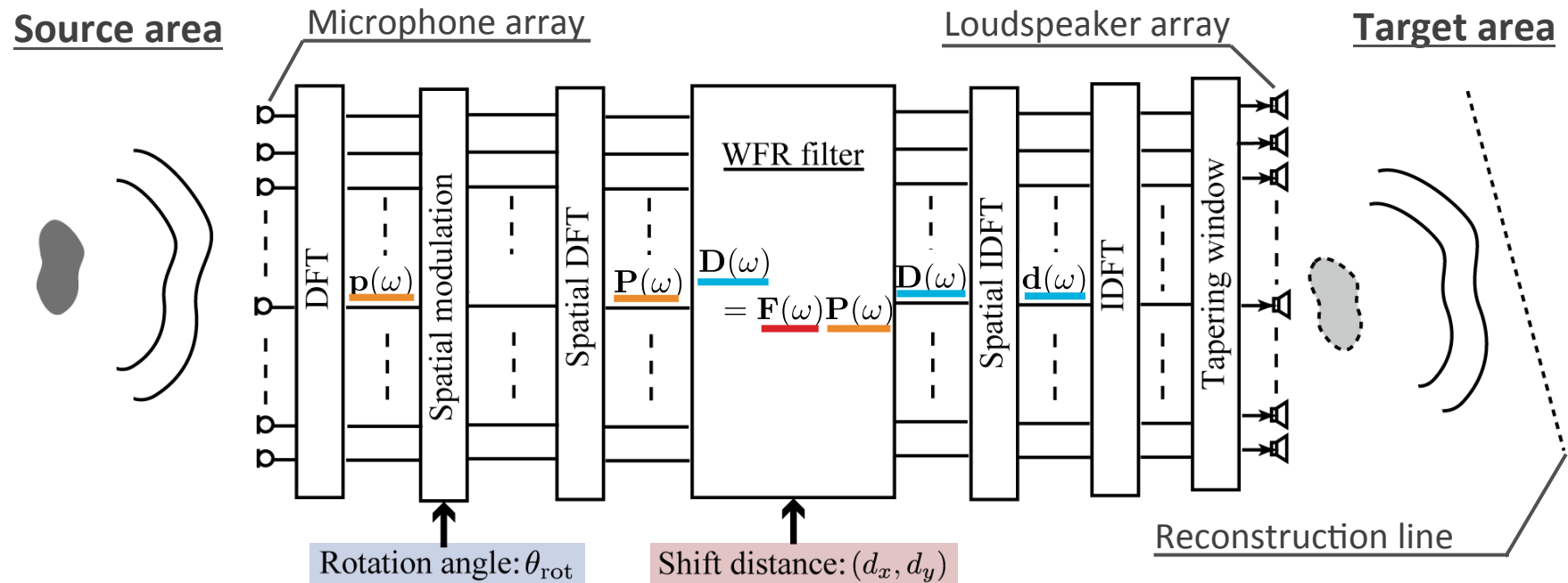
**Wave Field Reconstruction (WFR) Filter:** Driving signal can be obtained by convolution of spatio-temporal filter [Koyama+ IEEE TASLP2013]



- **Linear arrays of microphones and loudspeakers: 2D Convolution of WFR filter  $F(\omega)$** 
  - ✧ **Filter for signal conversion is stable compared to that designed by numerical approach (e.g., Least squares)**
  - ✧ **Fast computation of signal conversion can be achieved by using FFT algorithm**

# Wave Field Reconstruction (WFR) Filtering

Position of reproduced sound field can be controlled by spatial phase shift and modulation



- ✧ Reconstruction line can be shifted by spatial phase shift and rotated by spatial modulation
- ✧ It is possible to recreate virtual sound sources in front of loudspeaker array

# Practical Implementation

Real-time sound field transmission system by NTT [Koyama+ IEICE Trans 2014]

Kanagawa



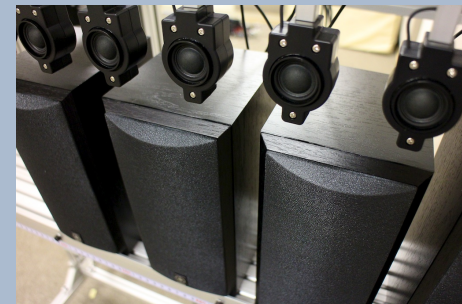
Network



Tokyo

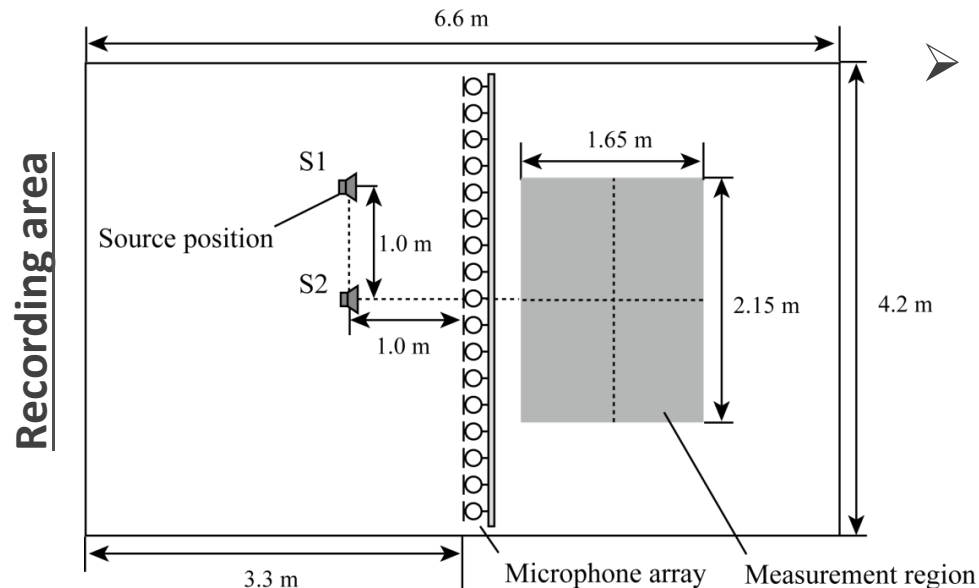


- Loudspeakers (for high freq.): 64, 6cm intervals
- Loudspeakers (for low freq.): 32, 12cm intervals
- Microphones: 64, 6cm intervals
- Array size: 3.84 m
- Sampling freq.: 48 kHz, Delay: 152 ms



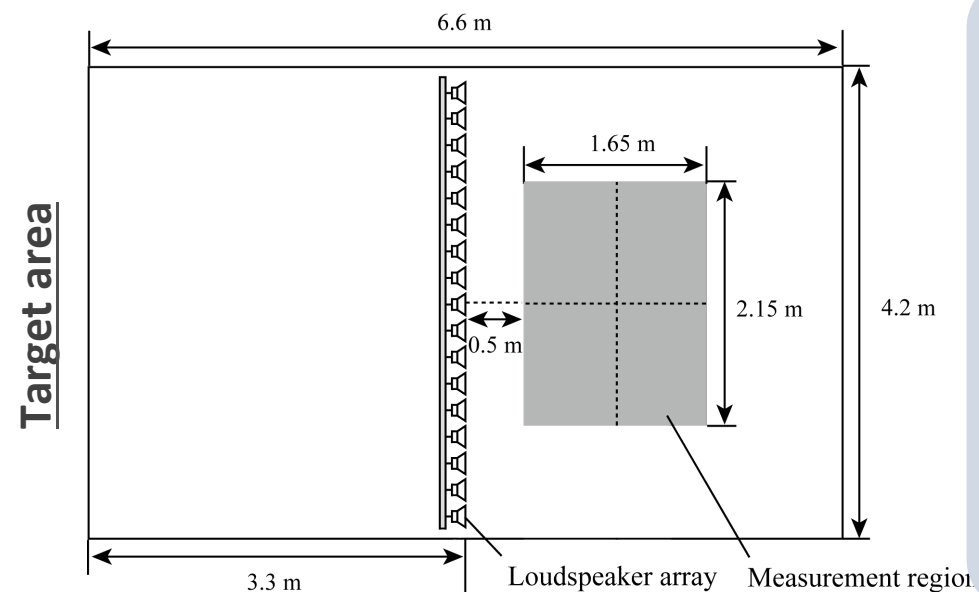


# Measurement experiments

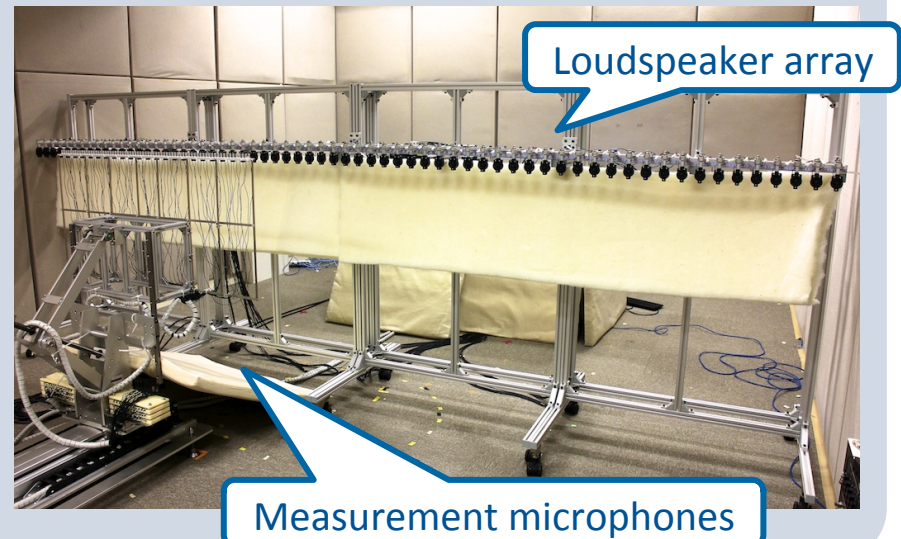


## ➤ Visualize reproduced sound field

- Impulse responses were measured on planar region of 2.15 x 1.65 m at 1.5 cm intervals (144x108 points)
- Reverberation time ( $T_{60}$ ): 167ms
- Source: Loudspeaker
- Array of smaller loudspeakers was only used

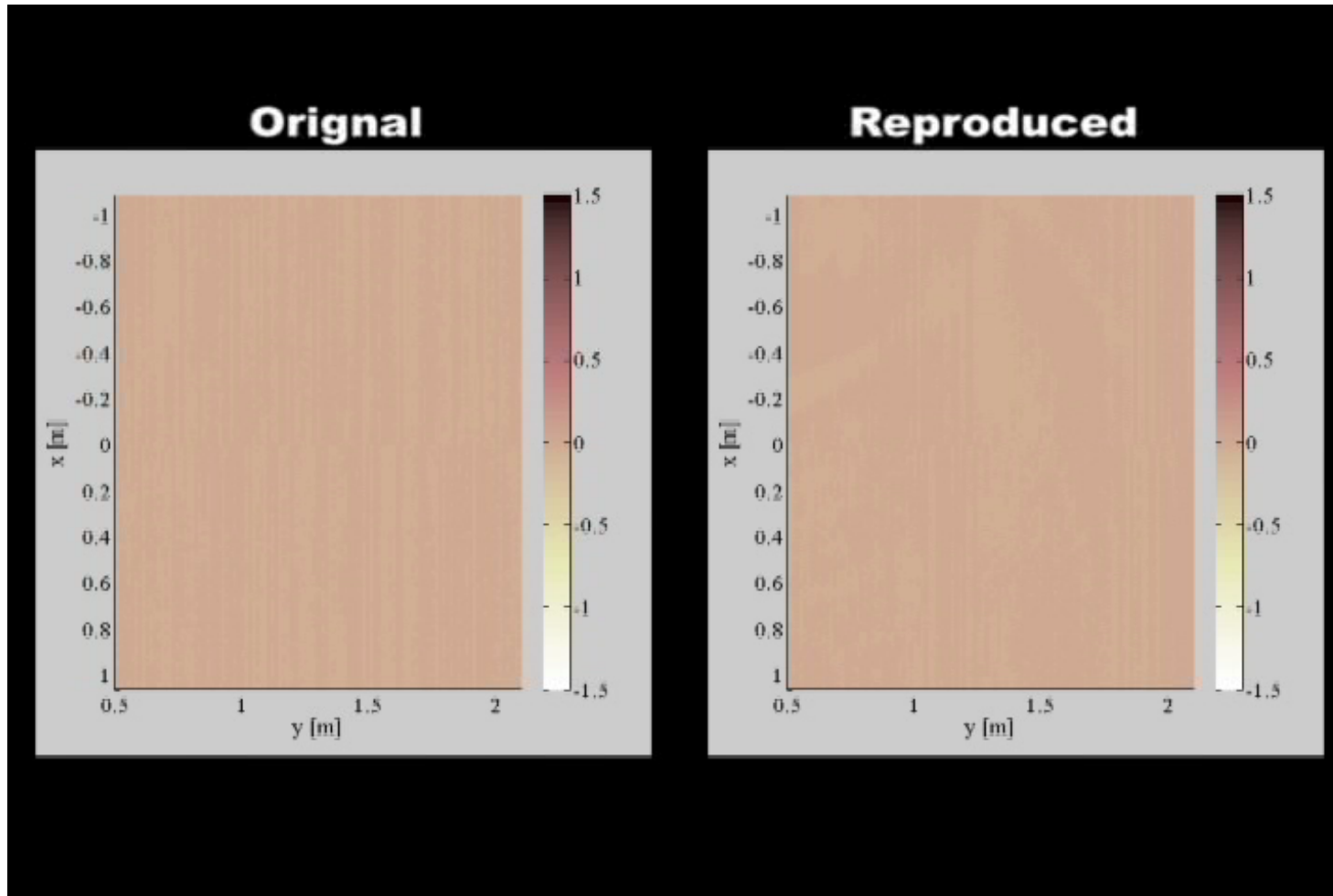


## Impulse response measurement



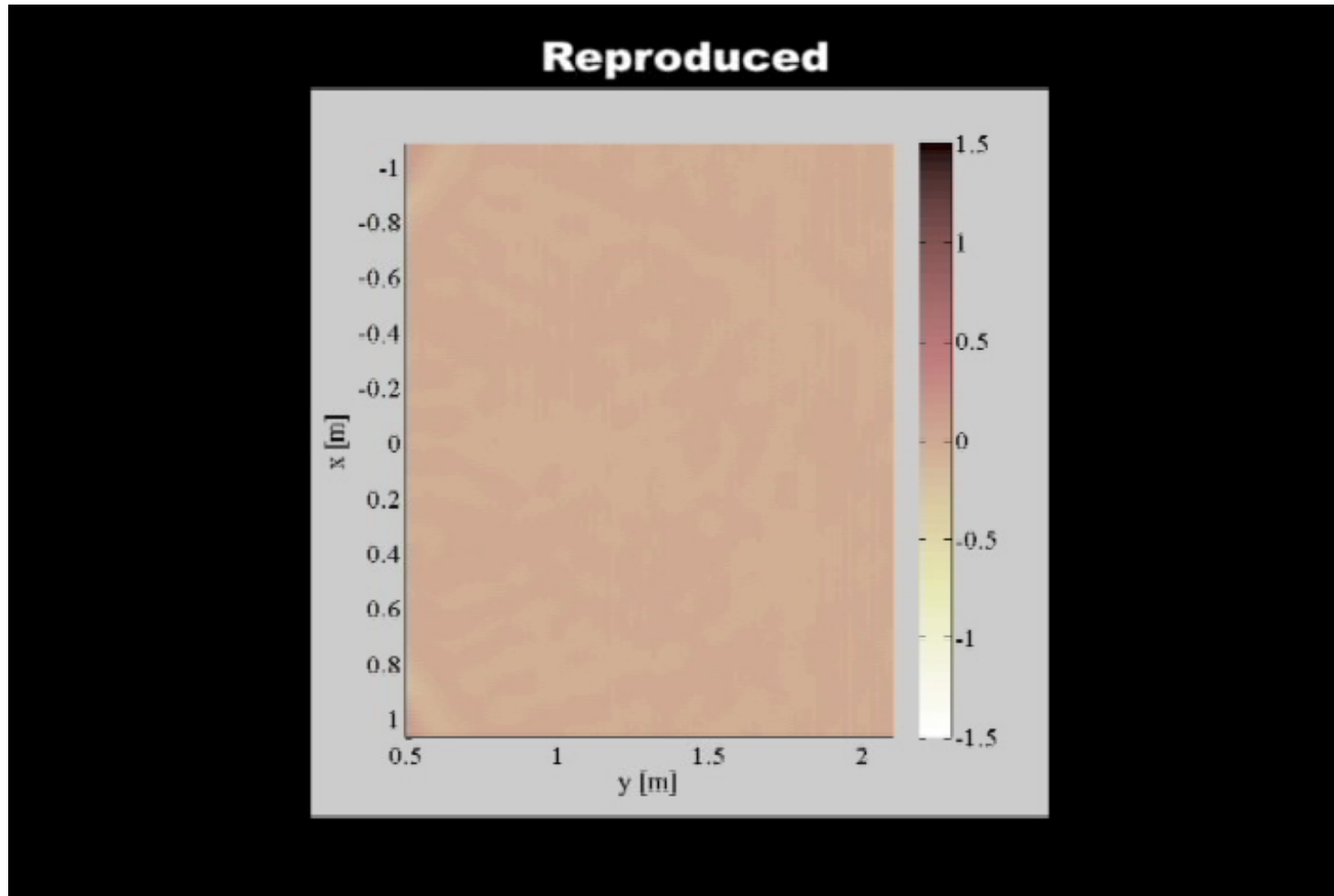
# Measurement of Reproduced Sound Field

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

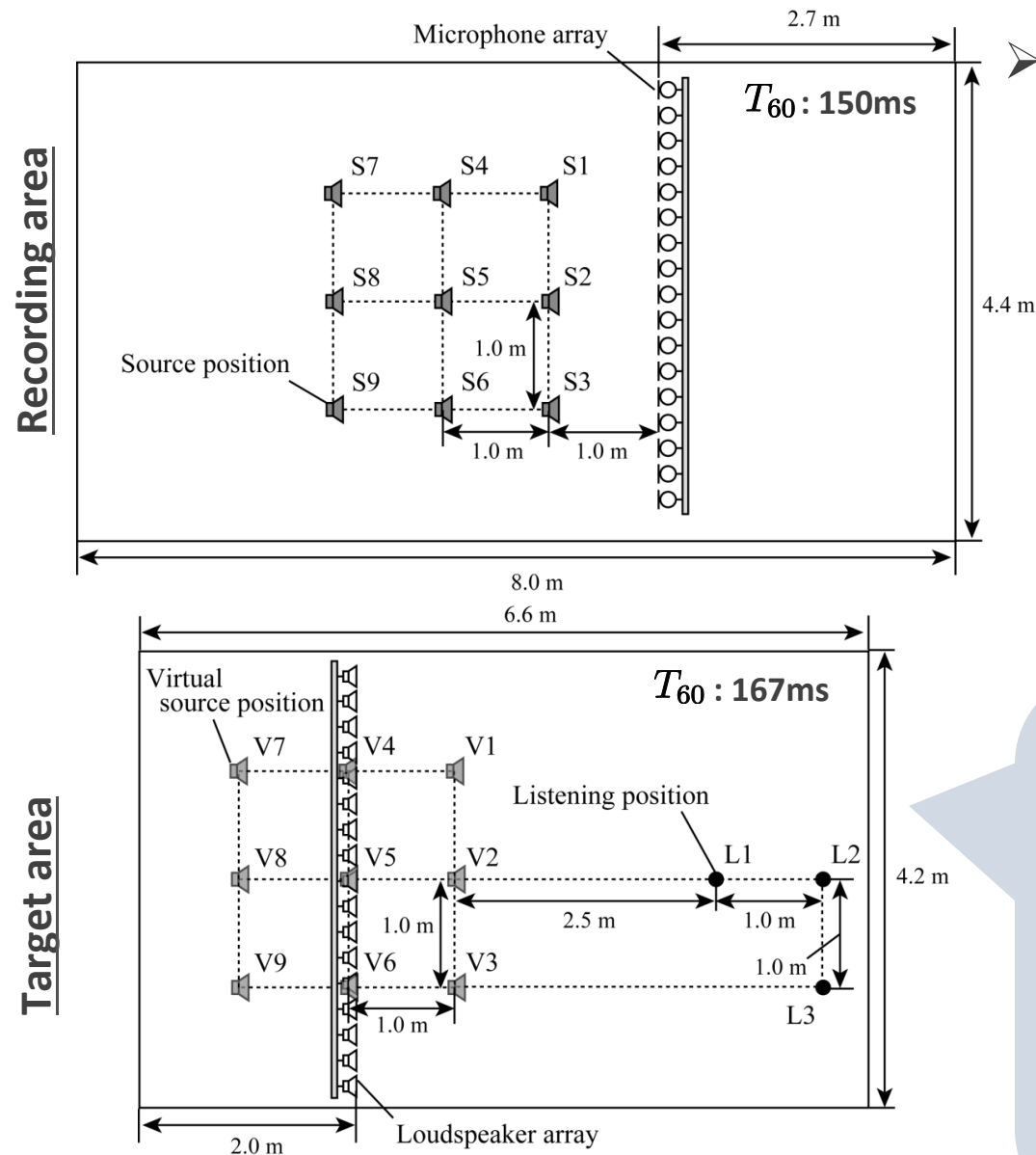


# Measurement of Reproduced Sound Field

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



# Sound Localization Listening Test



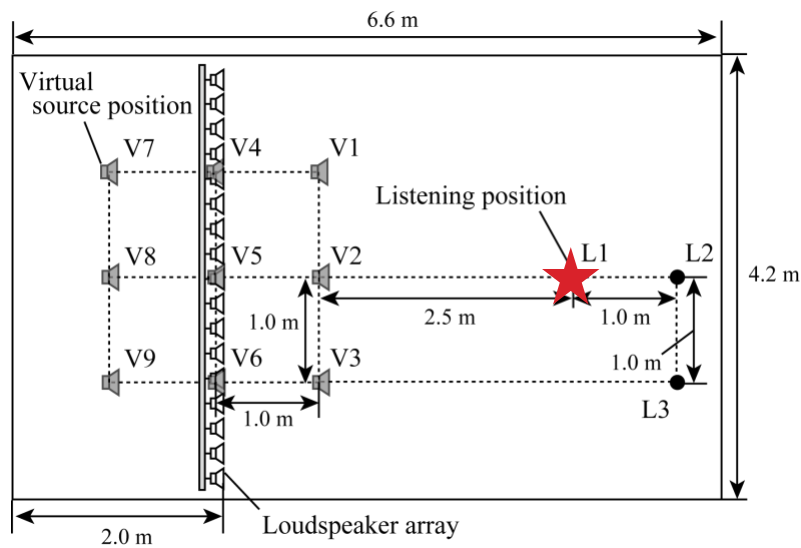
## Sound localization test

- 6 listeners answered sound location from 9 positions
- Compare localization accuracy between real and virtual sound sources
- Source signal: Speech (4 sec)
- Each sound source position was randomly selected at 144 times in total

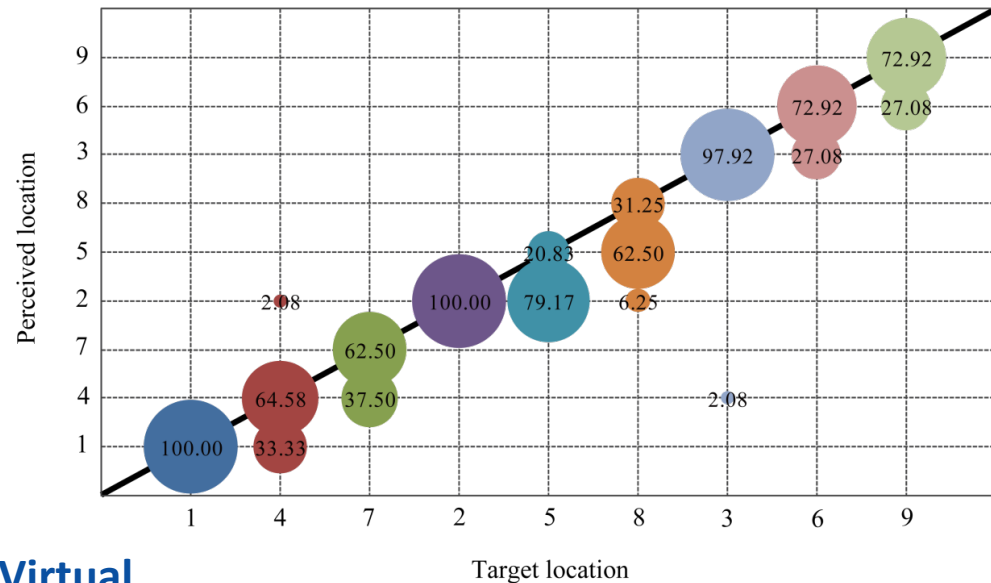
## Sound localization listening test



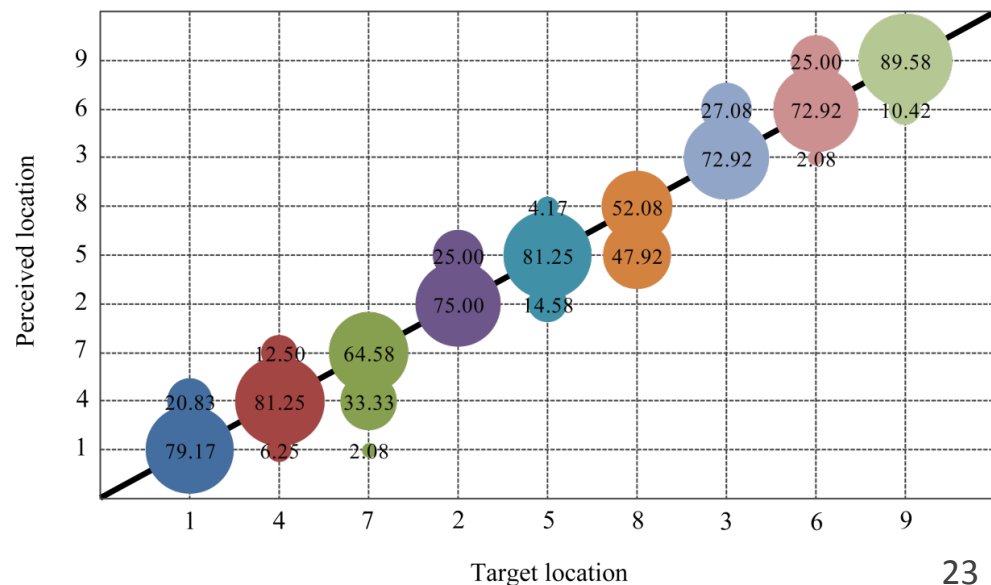
# Listening Test



## Real



## Virtual

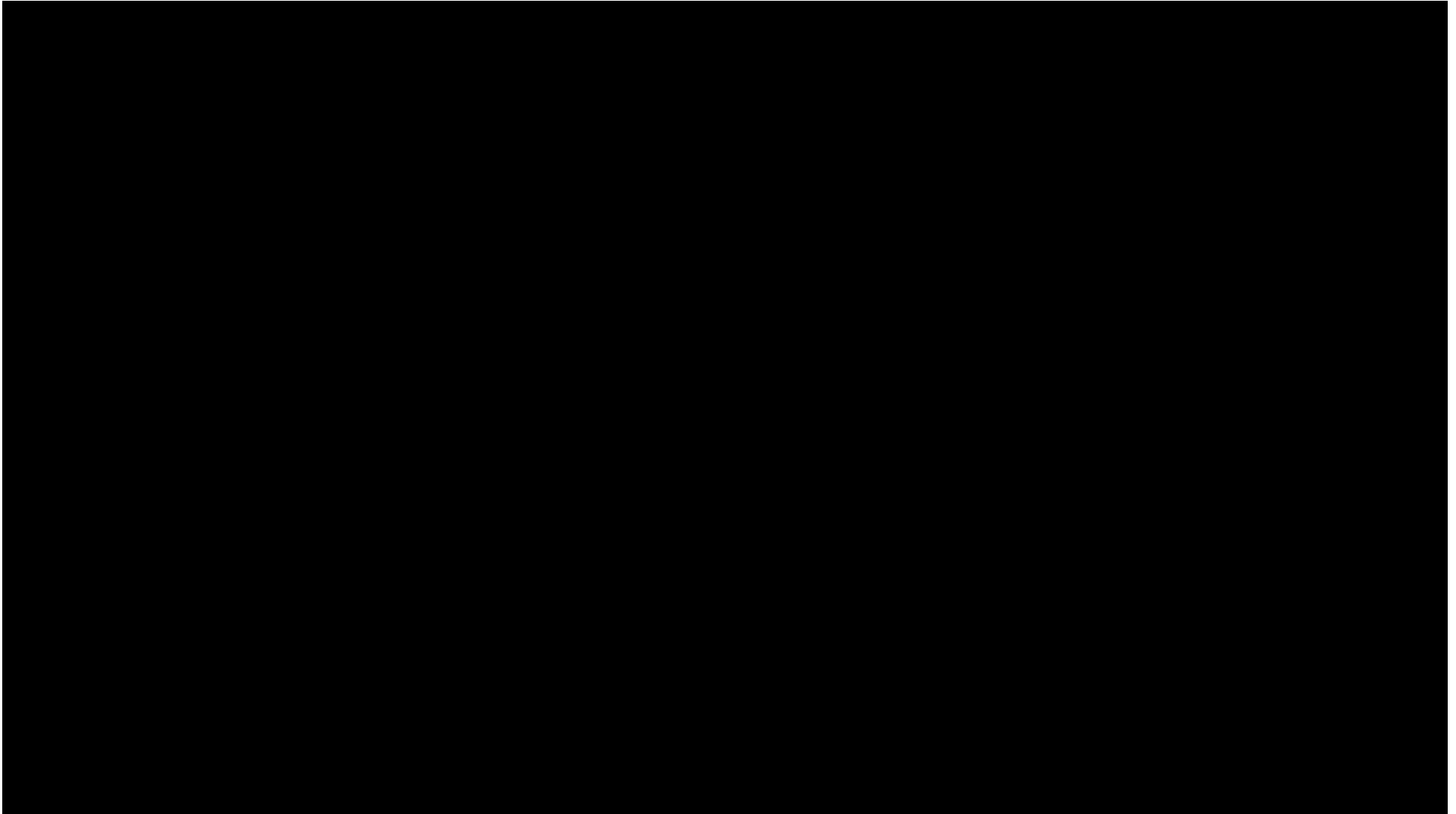


## ➤ Result of perceived location

- Almost the same localization accuracy
- Accuracy rate of virtual sources at V1 - V3 was slightly low
- Low accuracy rate of real sound sources in line with front-facing position (V5 & V8)

# Demonstration System with 3D image

DigInfo TV: <http://www.diginfo.tv/v/13-0016-r-en.php>



# Signal Conversion for Recording and Reproduction

- ✧ Fast and stable signal conversion for sound field recording and reproduction
- ✧ WFR filter was analytically derived in spatio-temporal freq domain
- ✧ Signal conversion by spatio-temporal convolution of WFR filter
- ✧ Real-time sound field transmission system and its evaluation

## Related publications

- S. Koyama, *et al.* "Wave field reconstruction filtering in cylindrical harmonic domain for with-height recording and reproduction," *IEEE/ACM Trans. Audio, Speech, Lang. Process.*, vol. 22, no. 10, pp. 1546-1557, 2014.
- S. Koyama, *et al.* "Real-time sound field transmission system by using wave field reconstruction filter and its evaluation," *IEICE Trans. Fund. Electron. Comm. Comput. Sci.*, vol. E97-A, no. 9, pp. 1840-1848, 2014.
- S. Koyama, *et al.* "Analytical Approach to Wave Field Reconstruction Filtering in Spatio-temporal Frequency Domain," *IEEE Trans. Audio, Speech, Lang. Process.*, vol. 21, no. 4, 2013.
- S. Koyama, *et al.* "Reproducing Virtual Sound Sources in Front of a Loudspeaker Array Using Inverse Wave Propagator," *IEEE Trans. Audio, Speech, Lang. Process.*, vol. 20, no. 6, 2012.

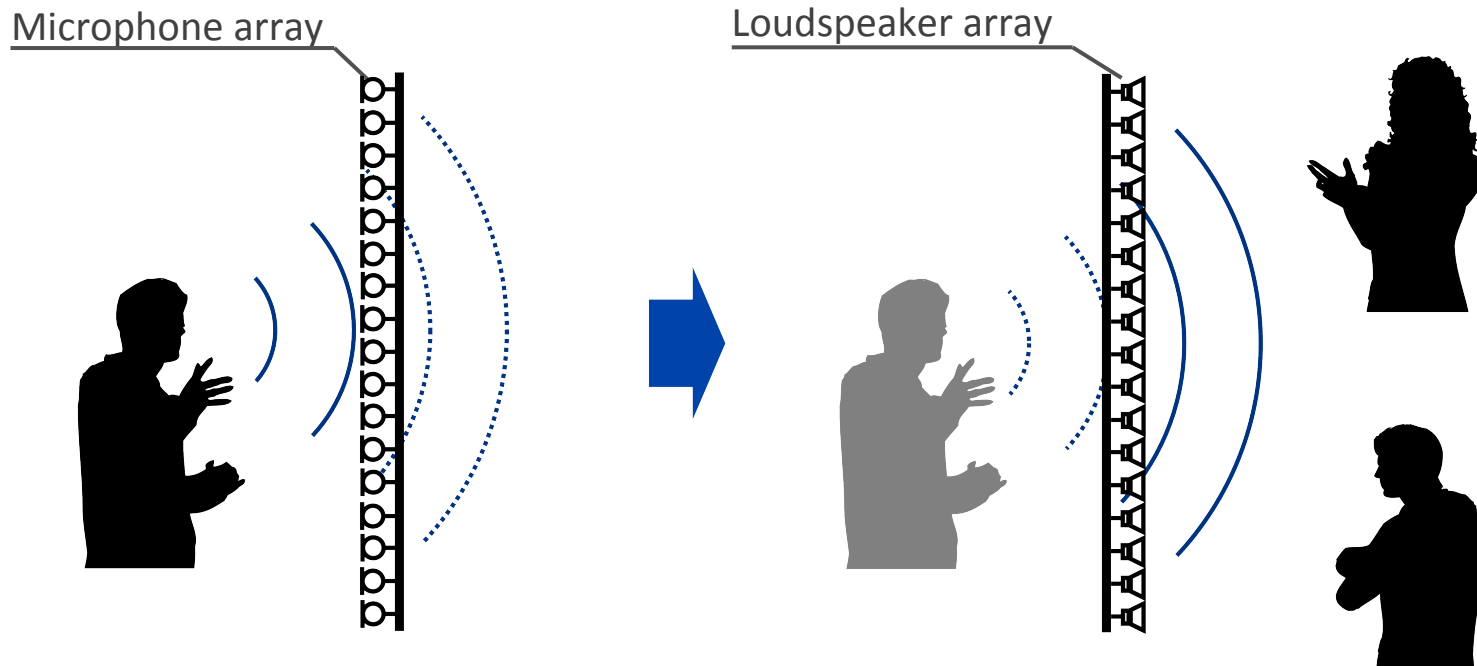




# **SUPER-RESOLUTION IN SOUND FIELD RECORDING AND REPRODUCTION**

# Sound field reproduction for audio system

## Super-resolution in sound field recording and reproduction

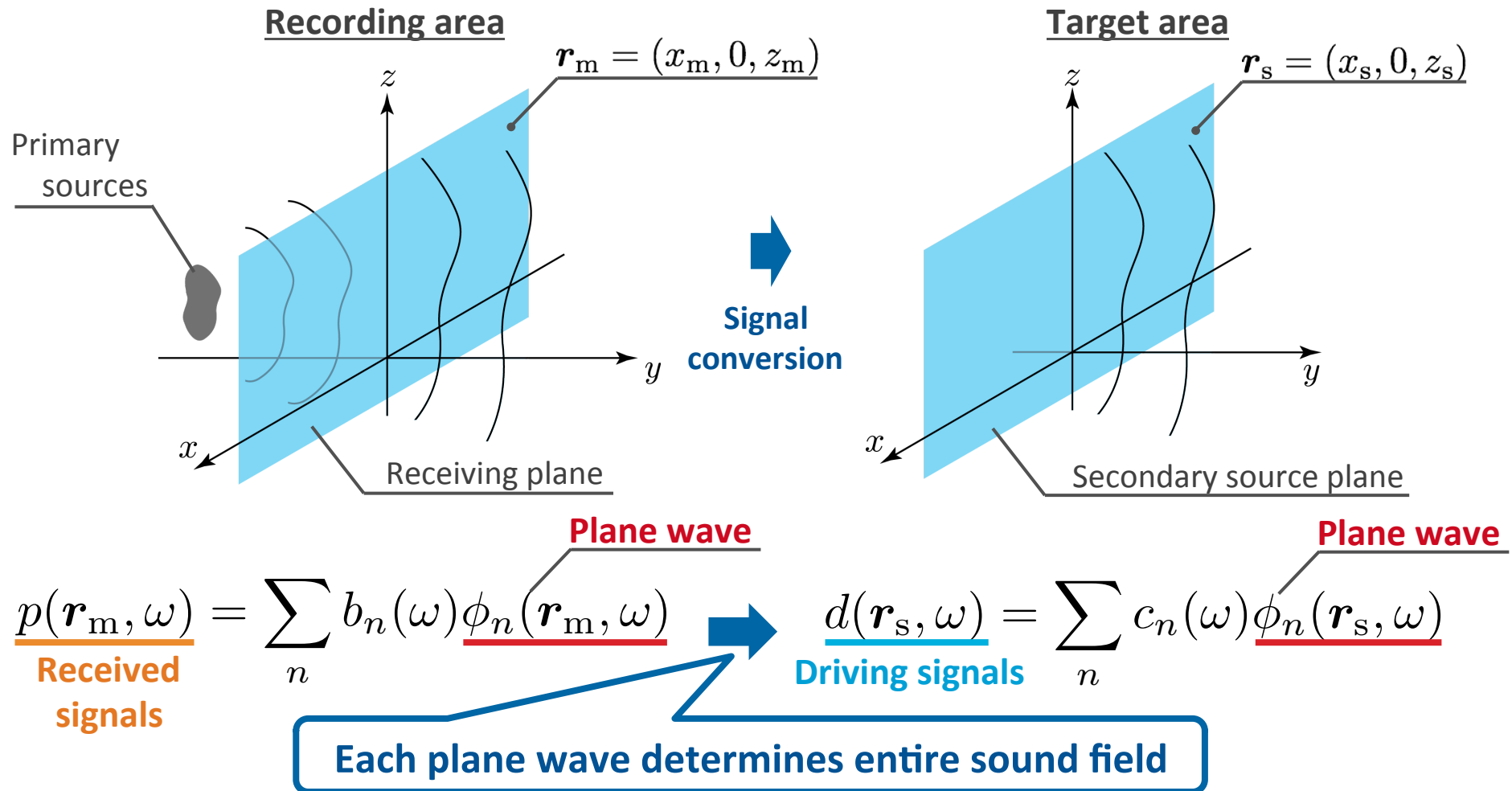


### Improve reproduction accuracy when less microphones than loudspeakers

- # of microphones > # of loudspeakers [Ahrens+ AES Conv. 2010]
  - Higher reproduction accuracy within local region of target area
- # of microphones < # of loudspeakers [Koyama+ WASPAA 2013], [Koyama+ ICASSP 2014]
  - Higher reproduction accuracy of sources in local region of recording area

# Wave field reconstruction (WFR) filtering method

[Koyama+ IEEE TASLP 2013]



Spatial aliasing artifacts due to plane wave decomposition →  
Significant error at high freq. even when microphone < loudspeaker

# Sound field decomposition for super-resolution



## Plane wave decomposition

$$\underline{p(\mathbf{r}, \omega)} = \sum_n b_n(\omega) \underline{\phi_n(\mathbf{r}, \omega)}$$

**Plane wave**

- Computationally efficient
- Larger effect of spatial aliasing

## Decompose into source signal, location, transfer function, etc...

$$\underline{p(\mathbf{r}, \omega)} = \sum_k s_k(\omega) \underline{h_k(\mathbf{r}, \omega)}$$

**Transfer function**

- Smaller effect of spatial aliasing
- Accurate estimation in real environment is very difficult

**Intermediate representation of sound field is required for super-resolution in recording and reproduction**

# Two Approaches to Super-resolution

## ➤ **Source-Location-Informed Sound Field Recording and Reproduction** [Koyama+ WASPAA 2013]

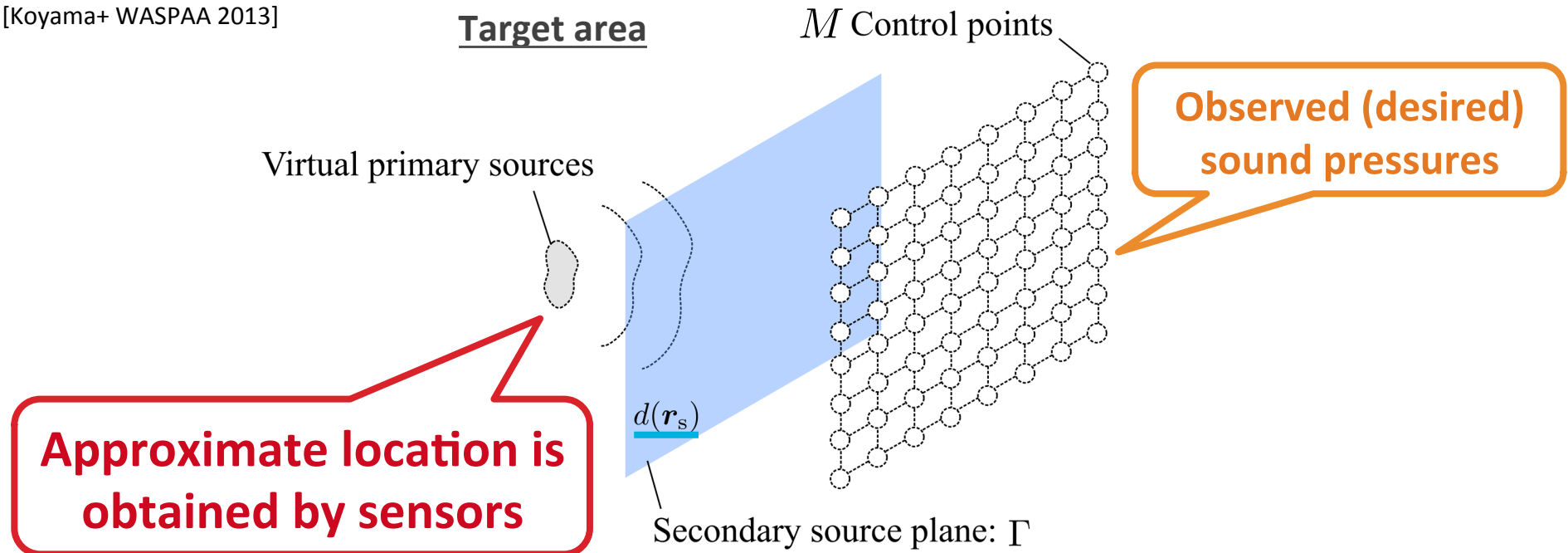
- Assume that approximate locations of sound sources in recording area are known
- Driving signals of loudspeakers are obtained by using prior information on source locations

## ➤ **Sparse Sound Field Representation for Super-resolution in Recording and Reproduction** [Koyama+ ICASSP 2014]

- Signals received by microphone array are decomposed into monopole source and plane wave components
- Sparse signal decomposition algorithm is applied under the assumption of sparse distribution of monopole source components

# Source-Location-Informed Recording and Reproduction

[Koyama+ WASPAA 2013]



## Constraint on driving signals

$$\underbrace{\mathbf{p}_{\text{des}}}_{\text{Desired pressures}} = \int_{\Gamma} \underbrace{d(\mathbf{r}_s)}_{\text{Transfer function}} \mathbf{G}(\mathbf{r}_s) d\Gamma + \epsilon$$

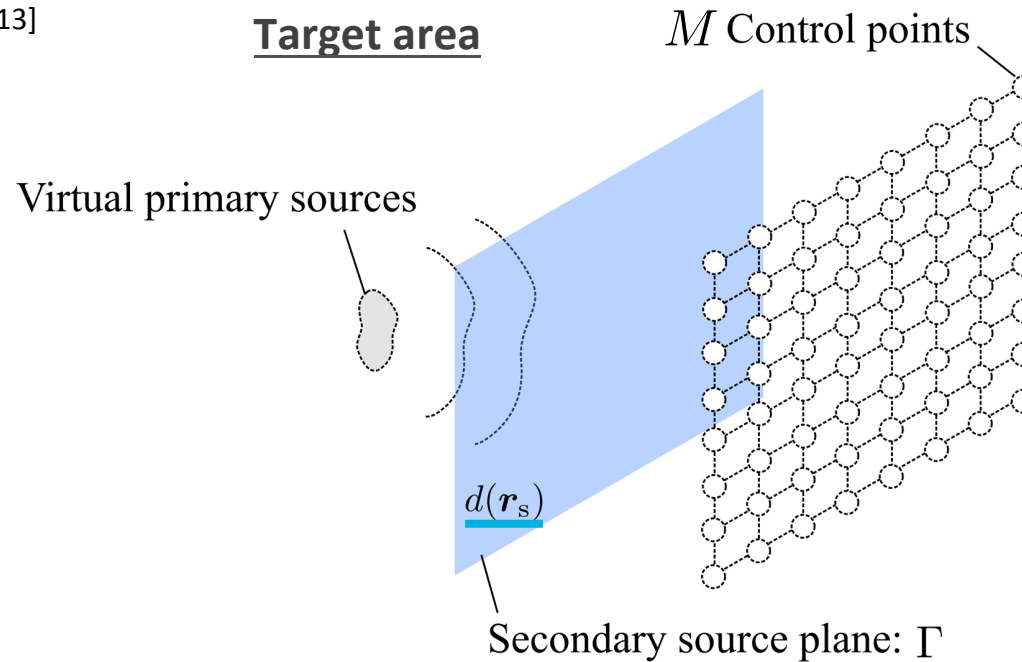
## Linear combination of spatial basis functions

$$\underbrace{\hat{d}(\mathbf{r}_s)}_{\text{}} = \sum_{n=1}^N c_n \underbrace{\phi_n(\mathbf{r}_s)}_{\text{}} = \underbrace{\boldsymbol{\phi}^T(\mathbf{r}_s)}_{\text{}} \mathbf{c}$$

Optimize  $\boldsymbol{\phi}(\mathbf{r}_s)$  and  $\mathbf{c}$  by using prior information on source locations

# MAP Estimation of Driving Signals

[Koyama+ WASPAA 2013]



## Maximum *a posteriori* (MAP) estimation

$$(\hat{\mathbf{c}}, \hat{\phi}(\mathbf{r}_s)) = \arg \max_{\mathbf{c}, \phi(\mathbf{r}_s)} \pi(\underline{d(\mathbf{r}_s)} = \phi^T(\mathbf{r}_s)\mathbf{c} | \underline{\mathbf{p}_{des}})$$

Bayes' rule



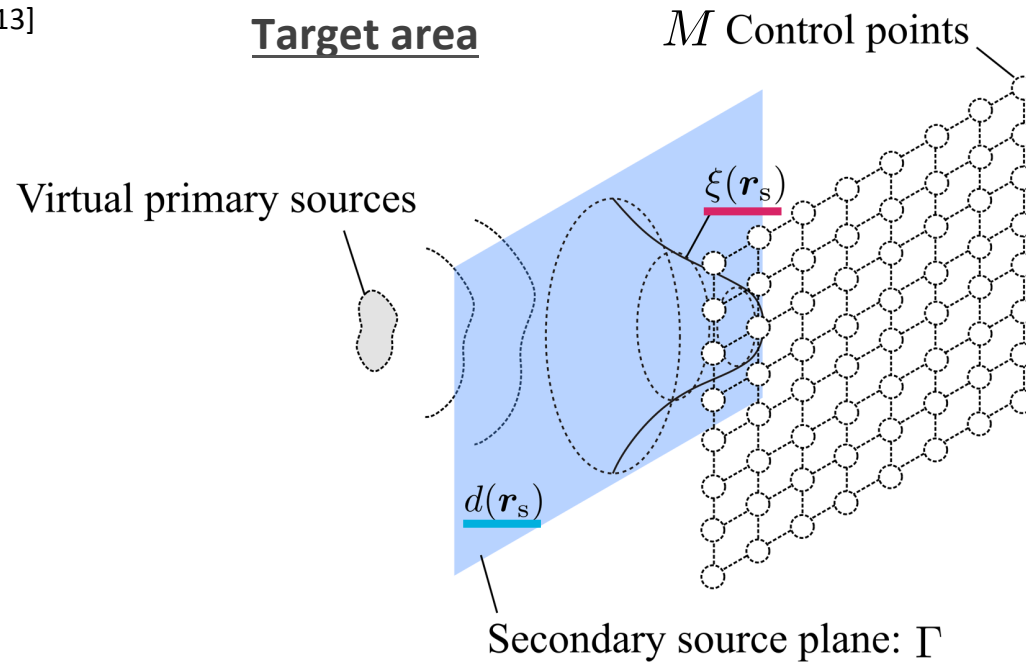
$$\pi(d(\mathbf{r}_s) | \mathbf{p}_{des}) \propto \pi(\underline{\mathbf{p}_{des}} | \underline{d(\mathbf{r}_s)}) \pi(\underline{d(\mathbf{r}_s)})$$

Likelihood function    Prior distribution



# MAP Estimation of Driving Signals

[Koyama+ WASPAA 2013]



## Maximum *a posteriori* (MAP) estimation

$$(\hat{\mathbf{c}}, \hat{\phi}(\mathbf{r}_s)) = \arg \max_{\mathbf{c}, \phi(\mathbf{r}_s)} \pi(\underline{d(\mathbf{r}_s)} = \phi^T(\mathbf{r}_s)\mathbf{c} | \underline{\mathbf{p}_{des}})$$

Bayes' rule



$$\pi(d(\mathbf{r}_s) | \mathbf{p}_{des}) \propto \pi(\underline{\mathbf{p}_{des}} | \underline{d(\mathbf{r}_s)}) \pi(\underline{d(\mathbf{r}_s)})$$

Likelihood function    Prior distribution

Use amplitude distribution of  $\underline{d(\mathbf{r}_s)}$  obtained from prior source locations

# MAP Estimation of Driving Signals

- Assume spatial basis functions are  $M$  orthogonal functions, which satisfies the following relation of singular value decomposition

$$\xi(\mathbf{r}_s) \underline{\mathbf{G}^H(\mathbf{r}_s)} = \underline{\phi^T(\mathbf{r}_s)} \mathbf{\Lambda} \mathbf{U}^H$$

- Optimal spatial basis functions and their coefficients

$$\underline{\hat{\phi}(\mathbf{r}_s)} = \underline{\xi(\mathbf{r}_s)} \mathbf{\Lambda}^{-1} \mathbf{U}^T \underline{\mathbf{G}^*(\mathbf{r}_s)}$$

$$\hat{\mathbf{c}} = (\alpha \mathbf{I} + \mathbf{\Lambda}^2)^{-1} \mathbf{\Lambda} \mathbf{U}^H \underline{\mathbf{p}_{\text{des}}}$$

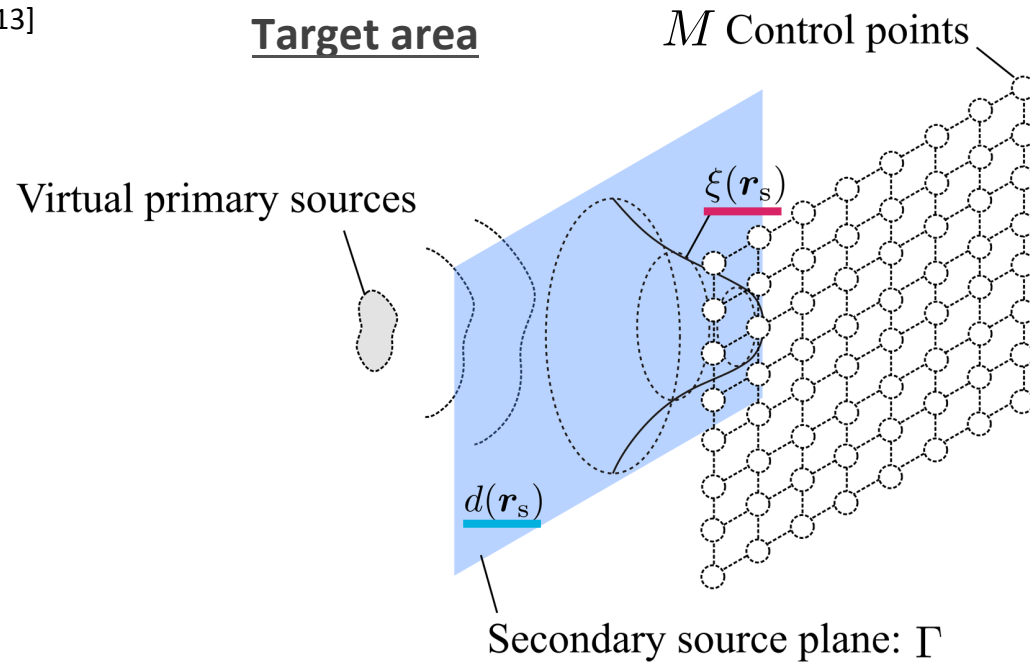
( $\alpha$  : regularization parameter)

- Driving signals obtained by MAP estimation

$$\hat{d}(\mathbf{r}_s) = \underline{\xi(\mathbf{r}_s)} \underline{\mathbf{G}^H(\mathbf{r}_s)} \mathbf{U} (\alpha \mathbf{I} + \mathbf{\Lambda}^2)^{-1} \mathbf{U}^H \underline{\mathbf{p}_{\text{des}}}$$

# Predicted Amplitude Distribution of Driving Signals

[Koyama+ WASPAA 2013]



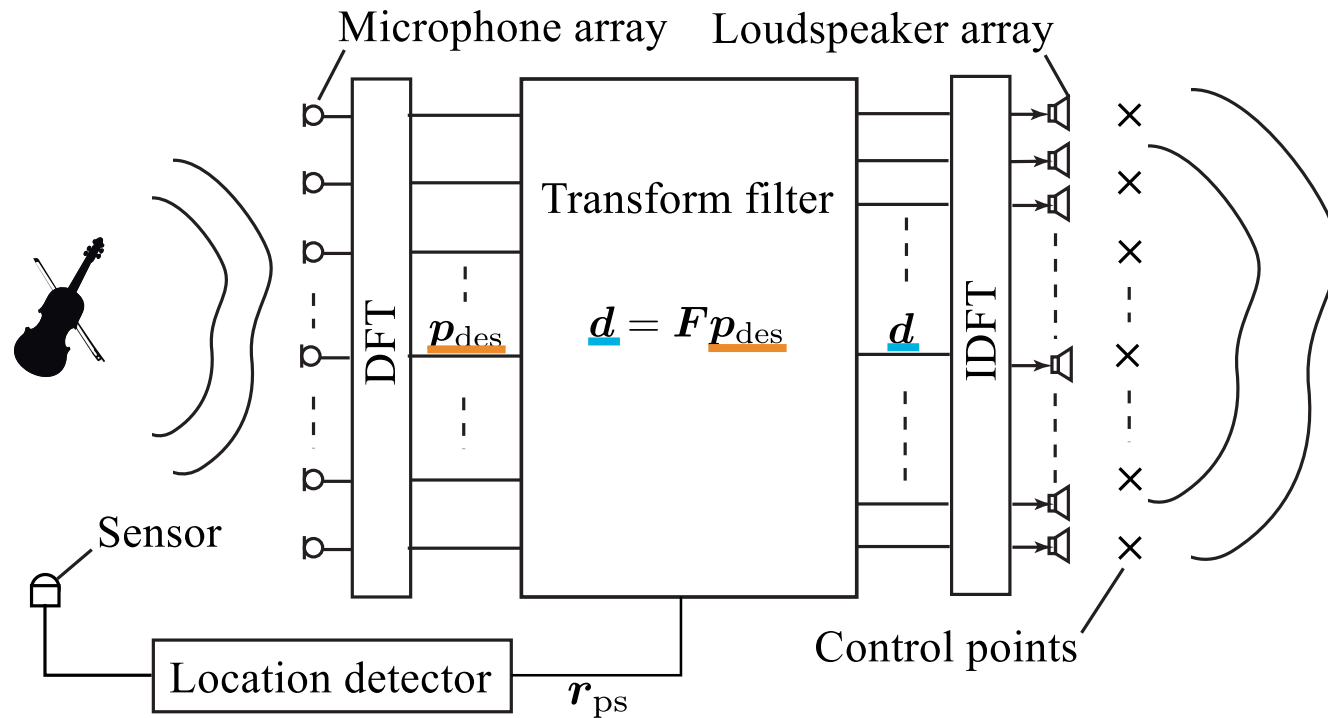
Assume spherical wave propagation from prior source locations

$$\bar{d}(\mathbf{r}_s) = 2 \frac{\partial \underline{p_{\text{des}}(\mathbf{r}_s)}}{\partial y_s}$$

Normalization

$$\simeq (y - y_{\text{ps}}) \frac{jk|\mathbf{r}_s - \mathbf{r}_{\text{ps}}| - 1}{2\pi|\mathbf{r}_s - \mathbf{r}_{\text{ps}}|} e^{jk|\mathbf{r}_s - \mathbf{r}_{\text{ps}}|} \rightarrow \underline{\xi(\mathbf{r}_s)} = |\underline{\bar{d}(\mathbf{r}_s)}| / \int_{\mathbf{r}_s \in \Gamma} |\underline{\bar{d}(\mathbf{r}_s)}|^2 d\Gamma$$

# Algorithm



## Discretize secondary source distribution

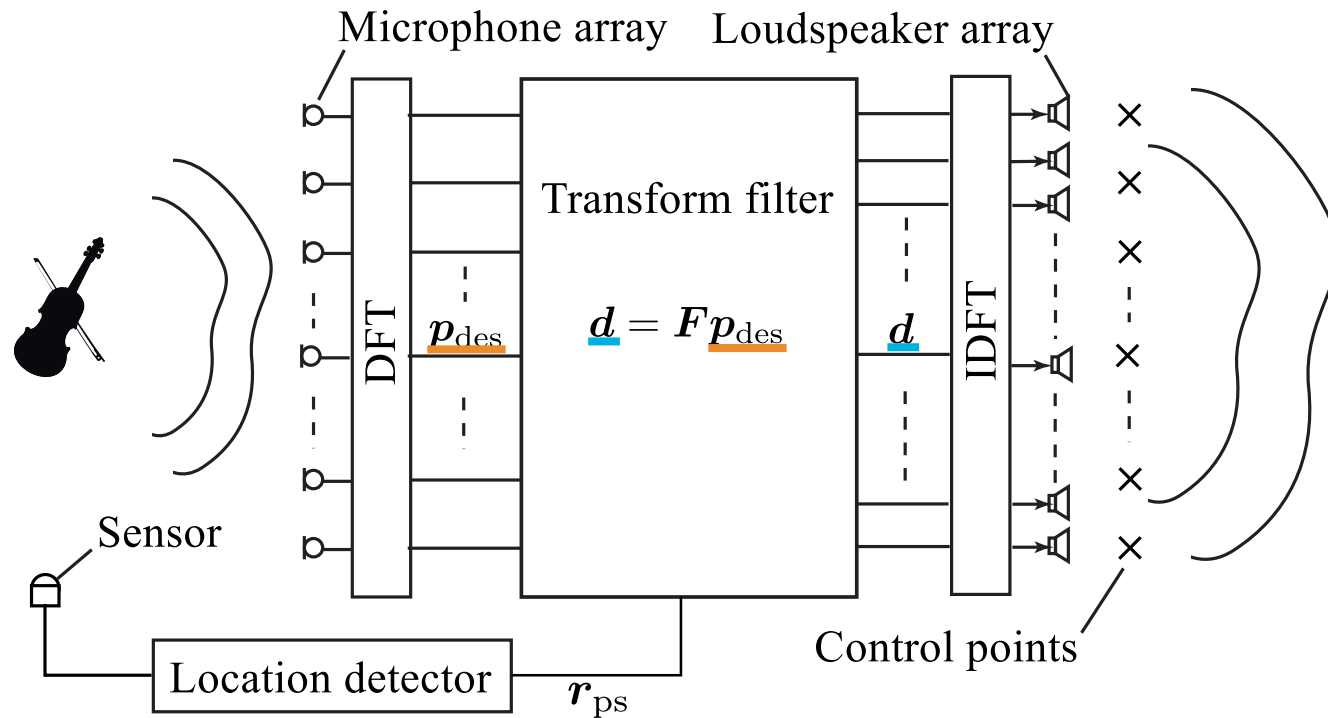
$$\underline{d} = [d(\mathbf{r}_{s,1}), d(\mathbf{r}_{s,2}) \cdots, d(\mathbf{r}_{s,L})]^T$$

$$\mathbf{H} = [\mathbf{G}(\mathbf{r}_{s,1}), \mathbf{G}(\mathbf{r}_{s,2}) \cdots, \mathbf{G}(\mathbf{r}_{s,L})]$$

## Amplitude distribution for prior

$$\underline{\xi} = \text{diag}([\xi(\mathbf{r}_{s,1}), \xi(\mathbf{r}_{s,2}) \cdots, \xi(\mathbf{r}_{s,L})])$$

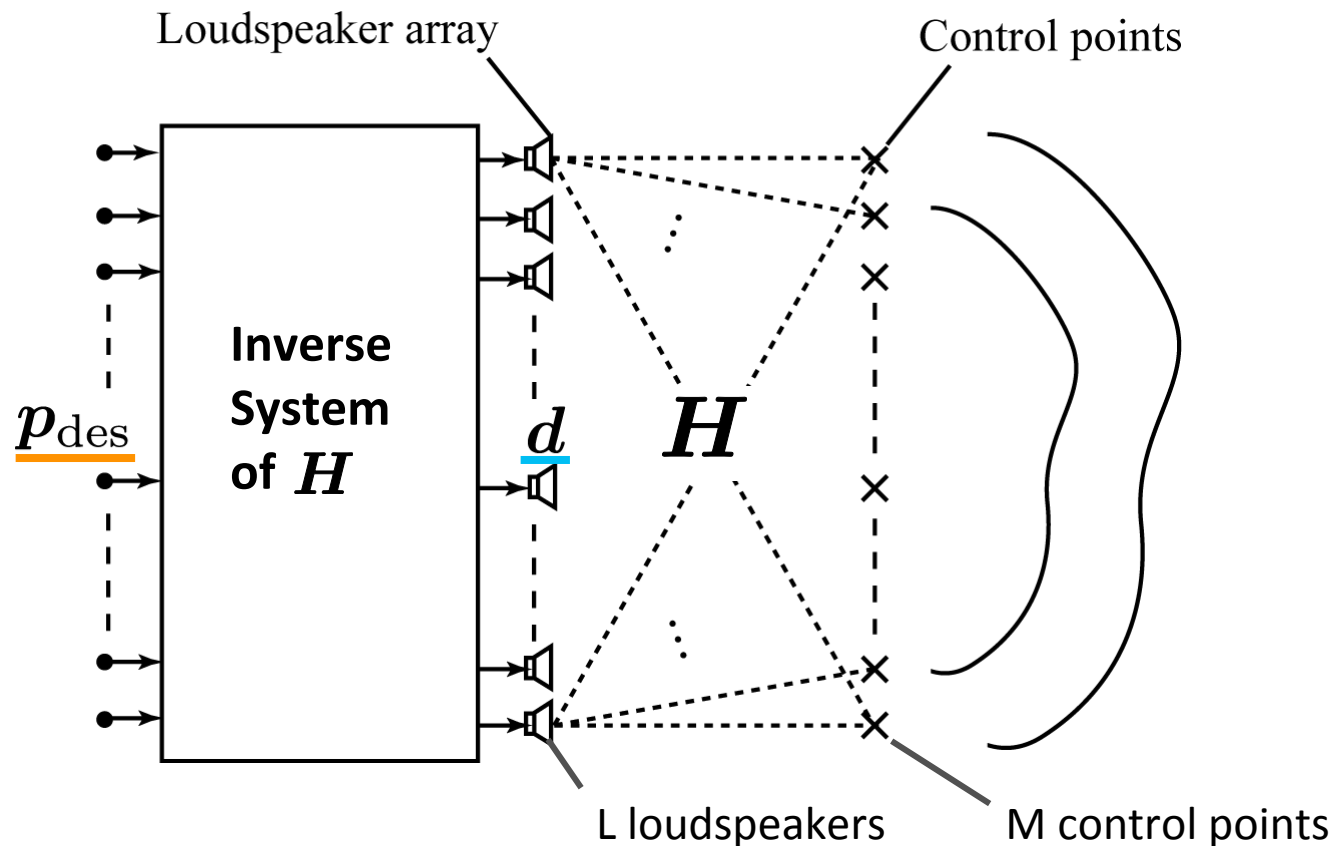
# Algorithm



1. Detect source location ( $r_{\text{ps}}$ )
2. Calculate amplitude distribution  $\xi$
3. Calculate  $R \in \mathbb{C}^{M \times M}$  as  $R = \underline{H} \xi H^H$
4. Eigenvalue decomposition of  $R = U \Lambda^2 U^H$
5. Obtain transform filter as  $F = \underline{\xi} H^H U (\alpha I + \Lambda^2)^{-1} U^H$

# Relation to Sound Pressure Control method

[Gautheir+ JASA 2005]

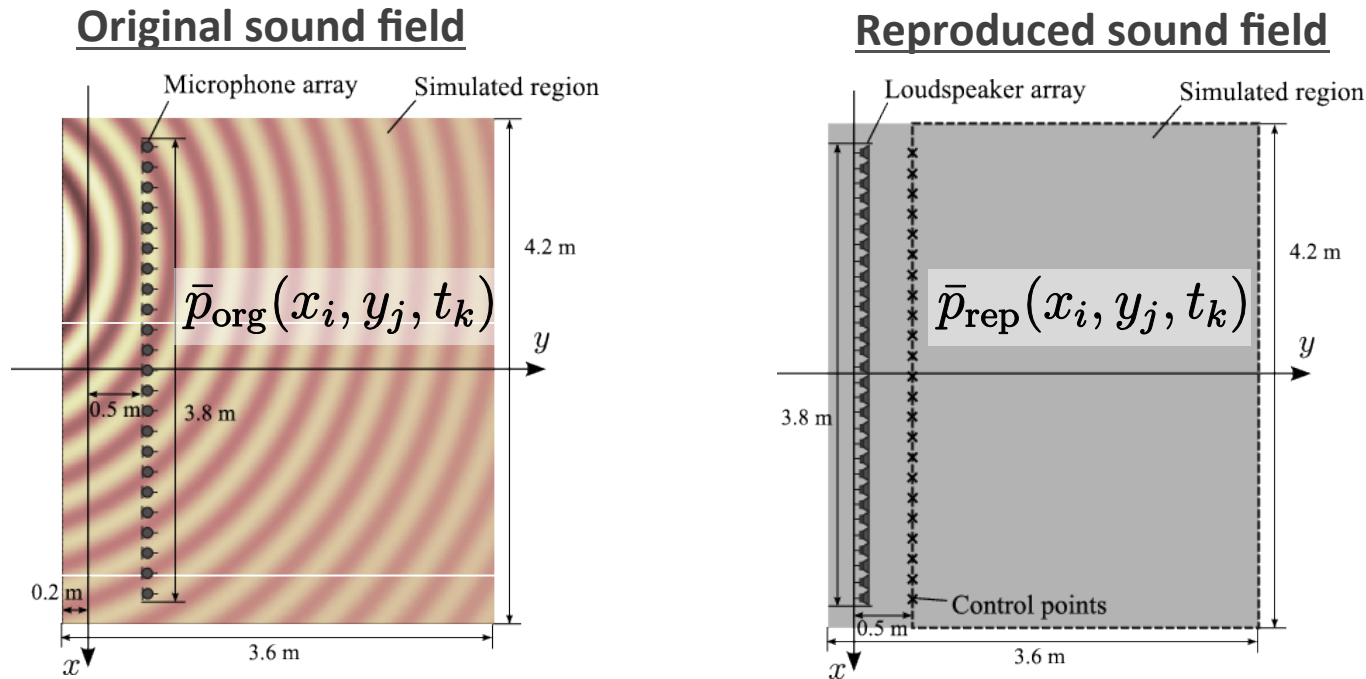


Inverse filter based on minimum-norm solution

$$\hat{d} = H^H (H H^H + \beta I)^{-1} p_{\text{des}}$$

Correspond to the case that  $\xi(r_s)$  is uniform distribution!

# Simulation Experiments

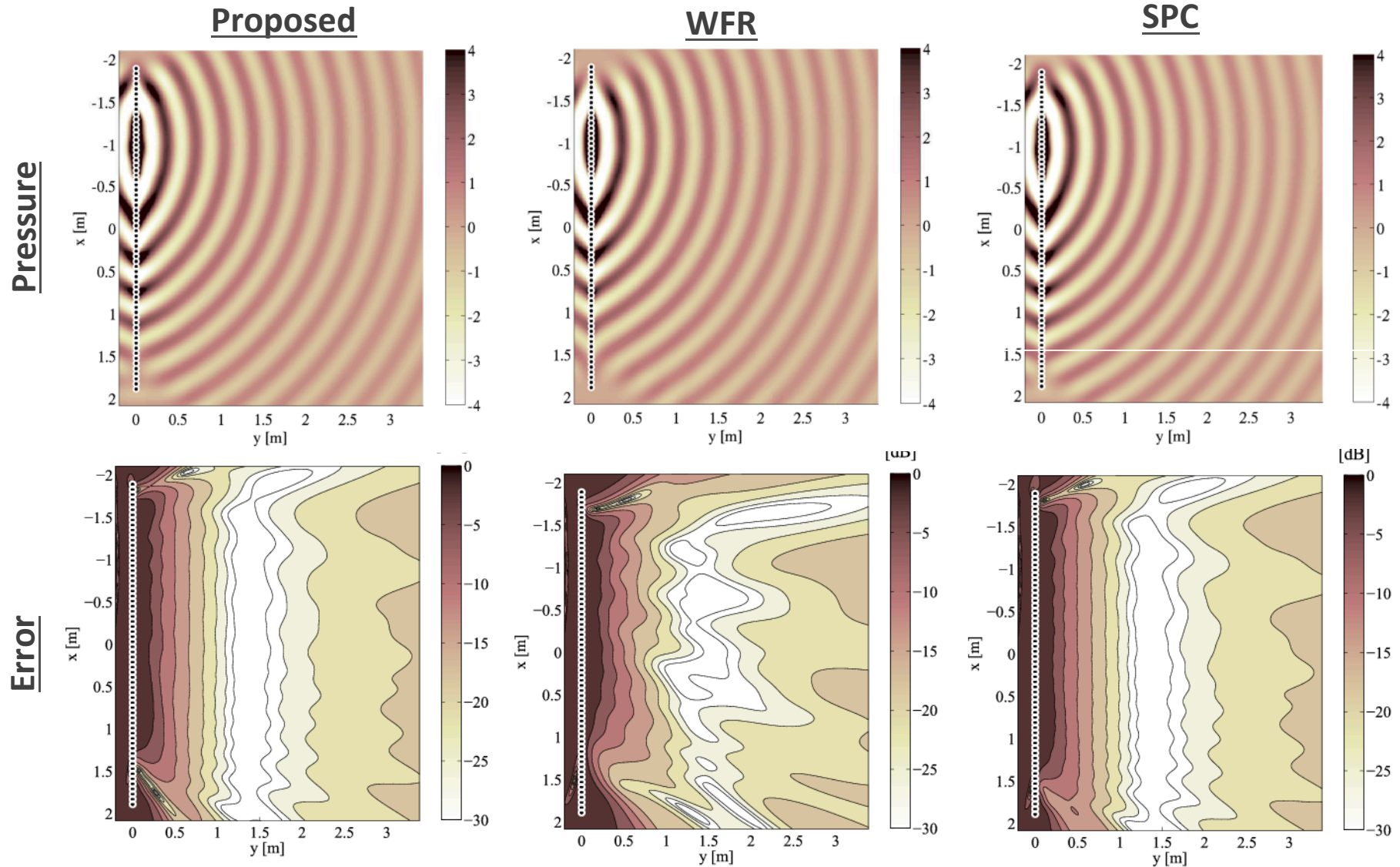


- Proposed method (**Proposed**), WFR filtering method (**WFR**), and sound pressure control method (**SPC**) methods were compared
- 32 microphones (12 cm intervals) and 64 loudspeakers (6cm intervals)
- Signal-to-distortion ratio (SDR) for evaluation

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

# Reproduced Sound Pressure Distribution (1.0 kHz)

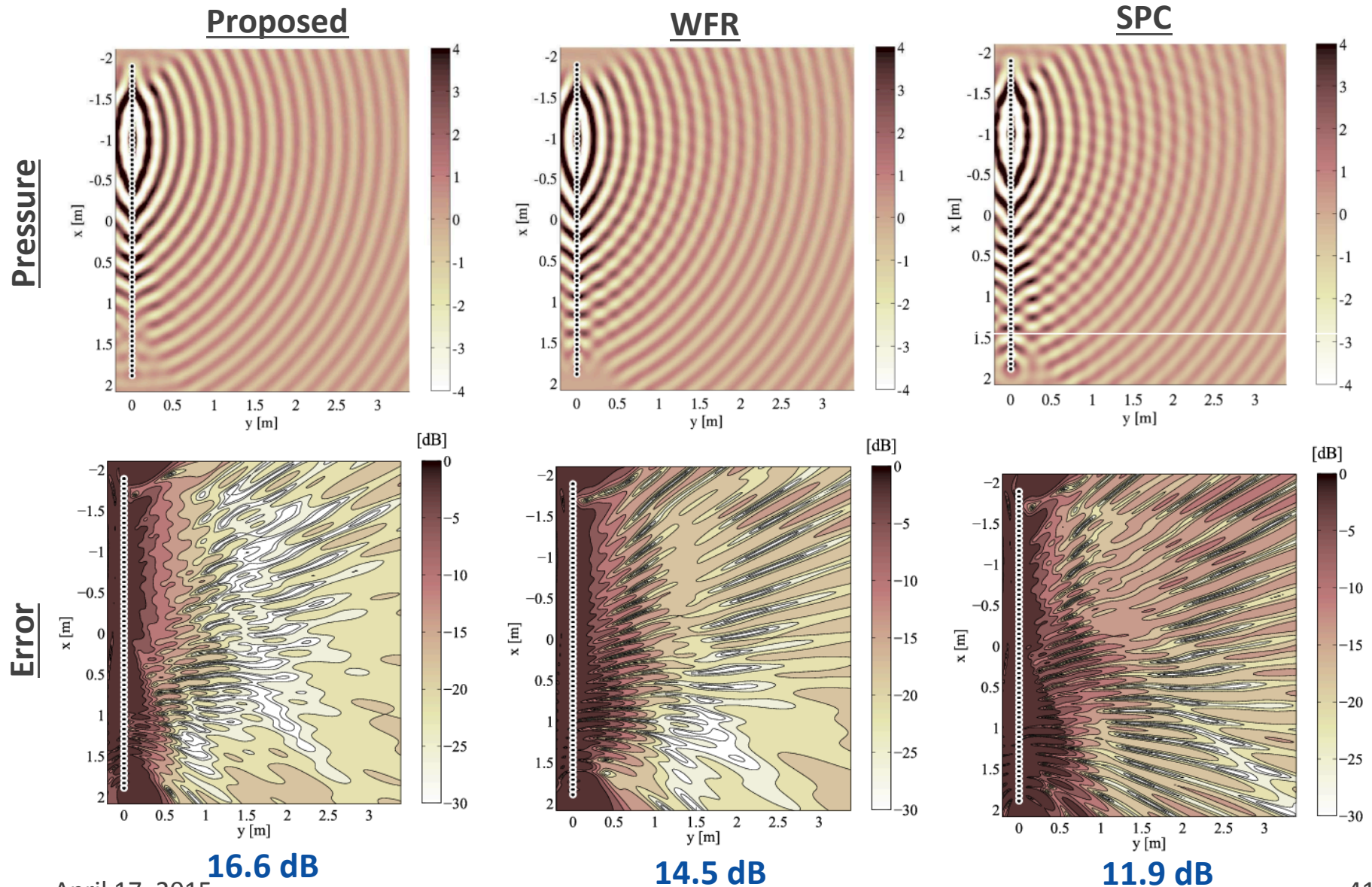
- Source location: (-1.0, -1.0, 0.0) m, Exact prior source location was given





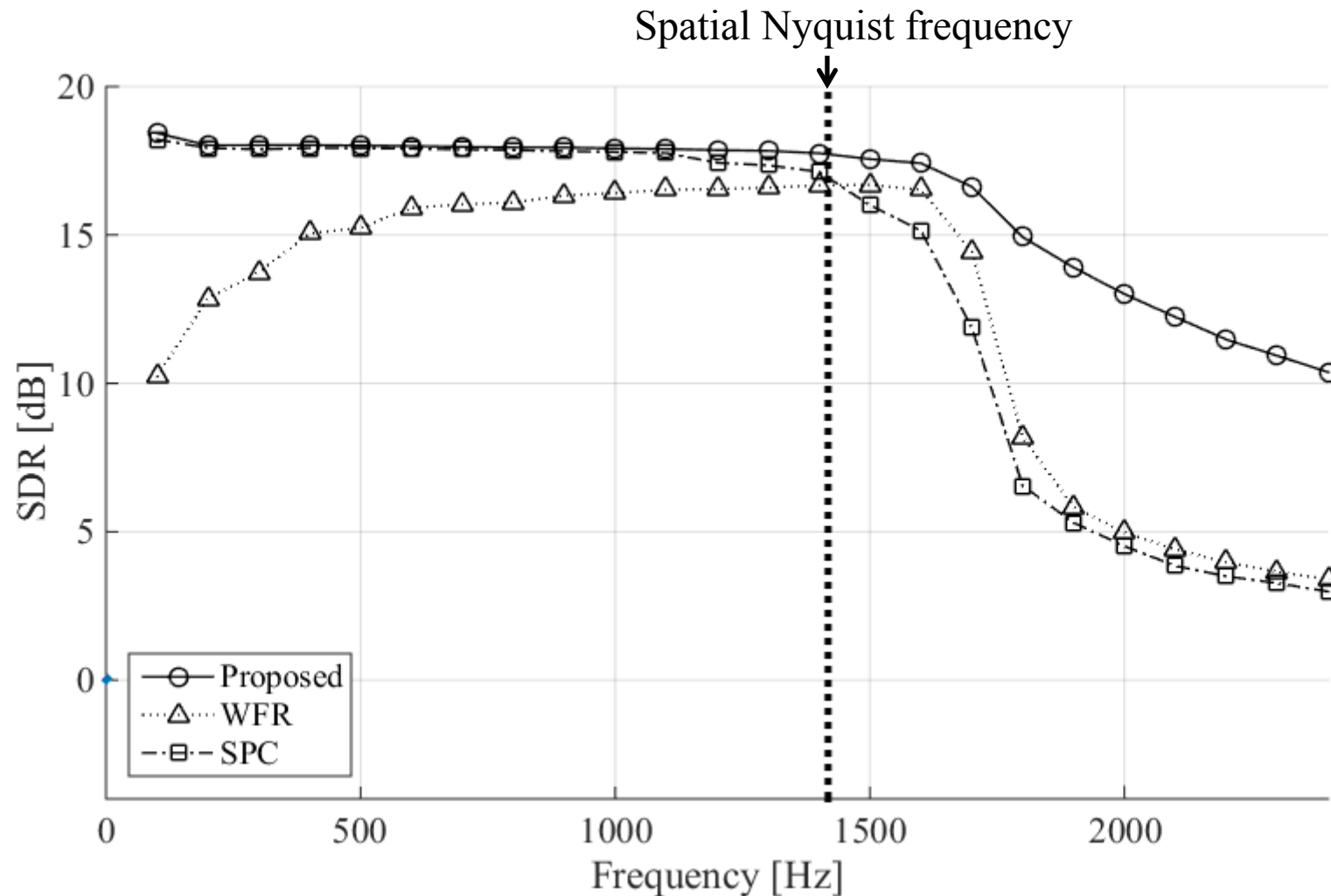
# Reproduced Sound Pressure Distribution (1.7 kHz)

- Source location: (-1.0, -1.0, 0.0) m, Exact prior source location was given



# Frequency vs. SDR

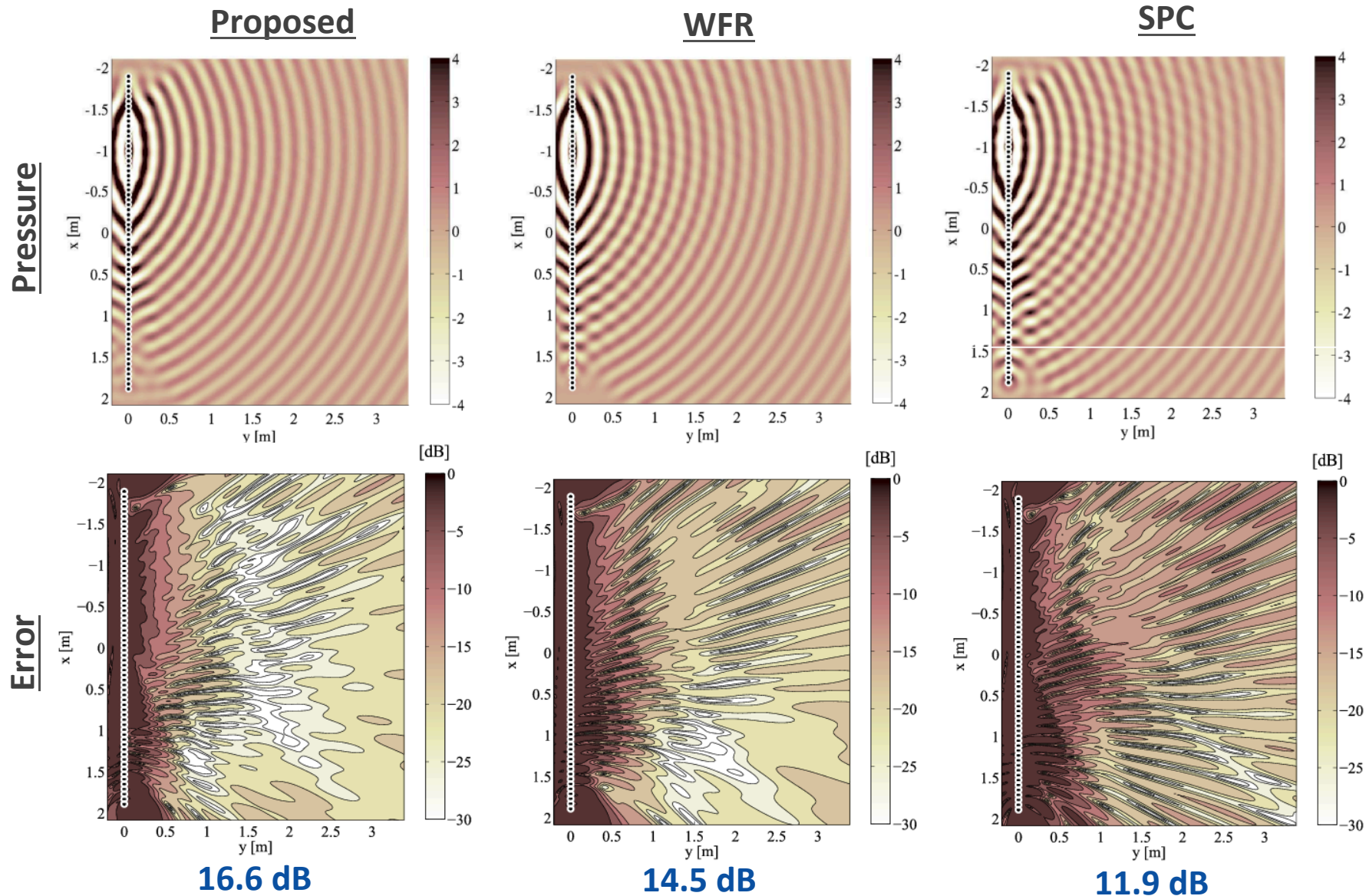
- Source location: (-1.0, -1.0, 0.0) m, Exact prior source location was given



**SDRs above spatial Nyquist frequency were improved**

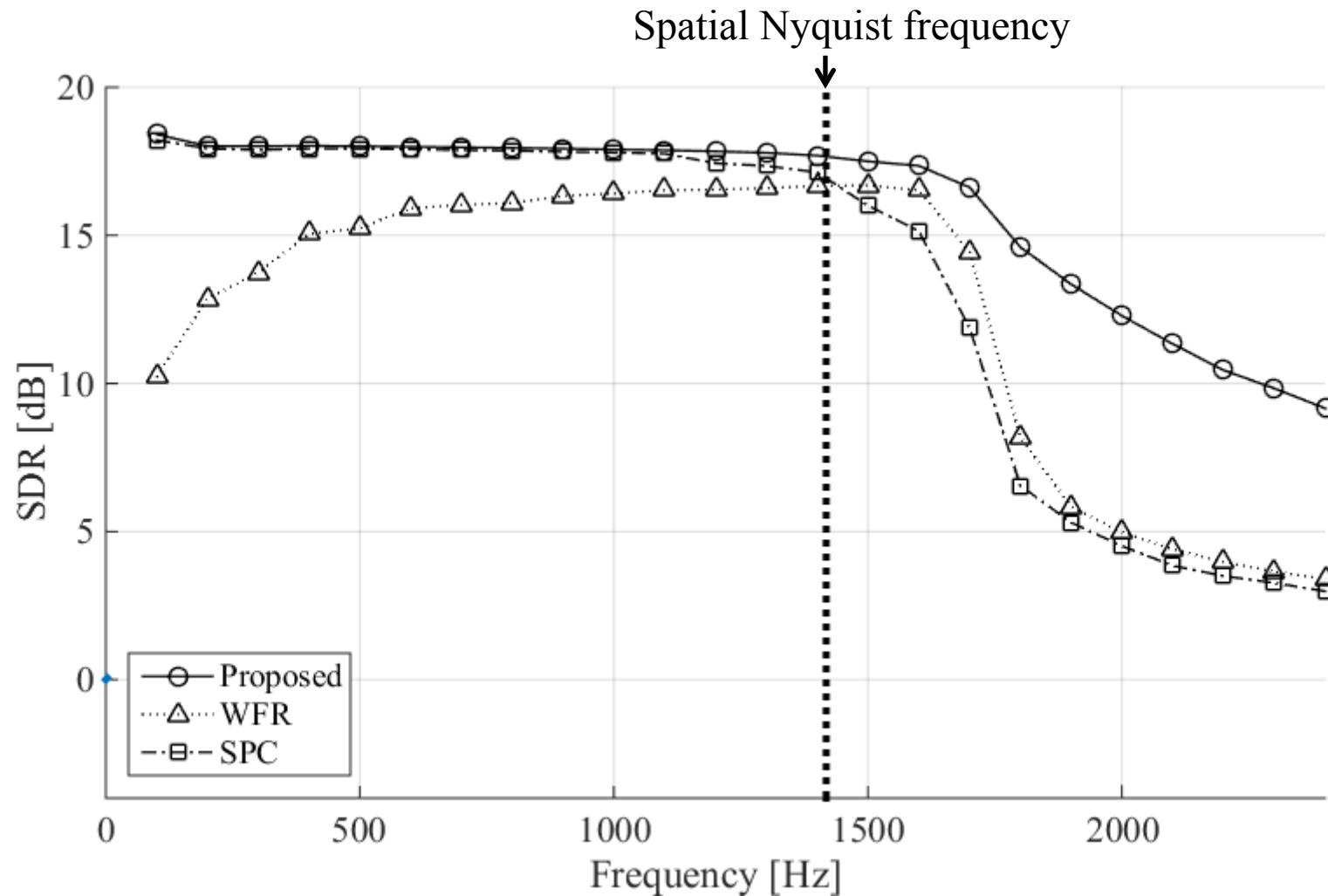
# Reproduced Sound Pressure Distribution (1.7 kHz)

- Source location: (-1.0, -1.0, 0.0) m, Prior source location: (-0.9, -1.2, 0.0) m



# Frequency vs. SDR

- Source location:  $(-1.0, -1.0, 0.0)$  m, **Prior source location:  $(-0.9, -1.2, 0.0)$  m**



**Proposed method is robust against mismatch of prior source locations**



# Two Approaches to Super-resolution

## ➤ **Source-Location-Informed Sound Field Recording and Reproduction** [Koyama+ WASPAA 2013]

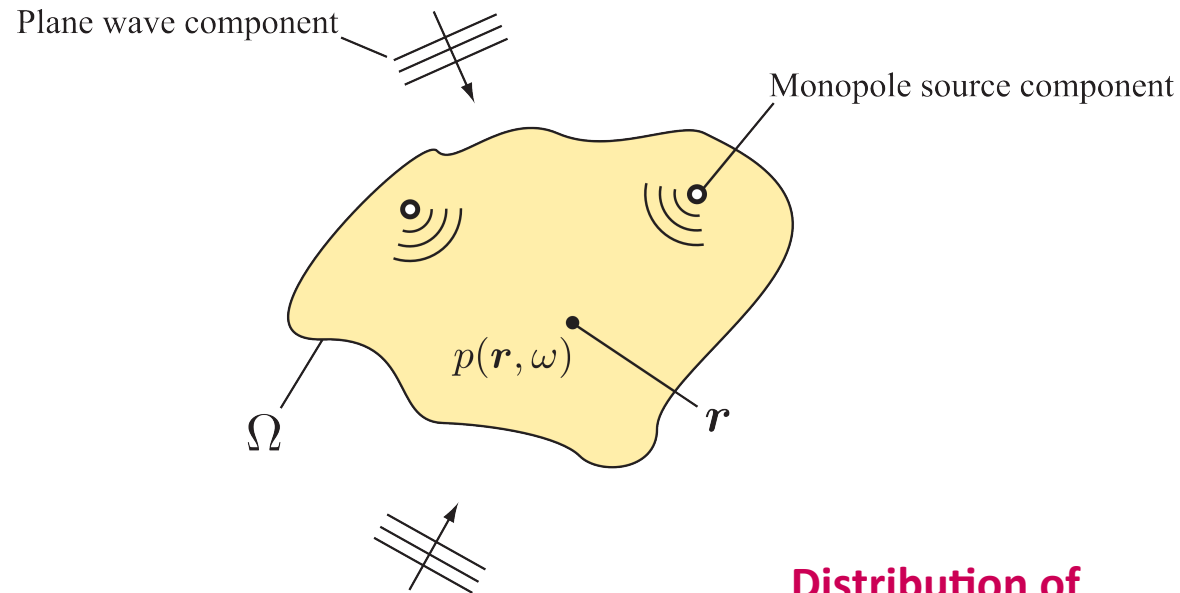
- Assume that approximate locations of sound sources in recording area are known
- Driving signals of loudspeakers are obtained by using prior information of source locations

## ➤ **Sparse Sound Field Representation for Super-resolution in Recording and Reproduction** [Koyama+ ICASSP 2014]

- Signals received by microphone array are decomposed into monopole source and plane wave components
- Sparse signal decomposition algorithm is applied under the assumption of sparse distribution of monopole source components

# Generative model of sound field

[Koyama+ ICASSP 2014]



## ➤ Inhomogeneous and homogeneous Helmholtz eq.

$$(\nabla^2 + k^2) \underline{p(\mathbf{r}, \omega)} = \begin{cases} -\underline{Q(\mathbf{r}, \omega)}, & \mathbf{r} \in \Omega \\ 0, & \mathbf{r} \notin \Omega \end{cases}$$

Distribution of  
monopole components

Inhomogeneous + homogeneous terms

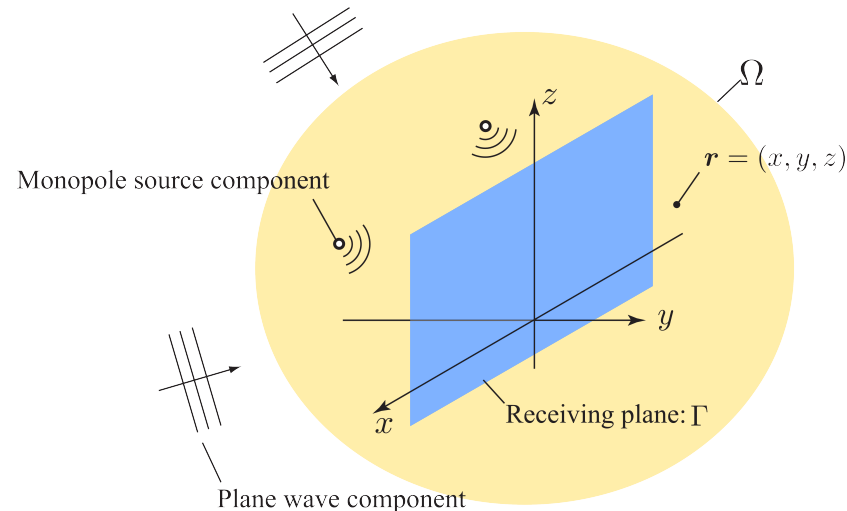


$$\begin{aligned} \underline{p(\mathbf{r}, \omega)} &= p_i(\mathbf{r}) + p_h(\mathbf{r}) \\ &= \int_{\mathbf{r}' \in \Omega} \underline{Q(\mathbf{r}')} \underline{G(\mathbf{r}|\mathbf{r}')} d\mathbf{r}' + \underline{p_h(\mathbf{r})} \end{aligned}$$

Green's function

# Generative model of sound field

## ➤ Observe sound pressure distribution on $\Gamma$



## ➤ Conversion into driving signals

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

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

**Synthesize monopole sources** [Spors+ AES Conv. 2008]

**Applying WFR filtering method** [Koyama+ IEEE TASLP 2013]

**Decomposition into two components may lead to higher reproduction accuracy above spatial Nyquist freq.**

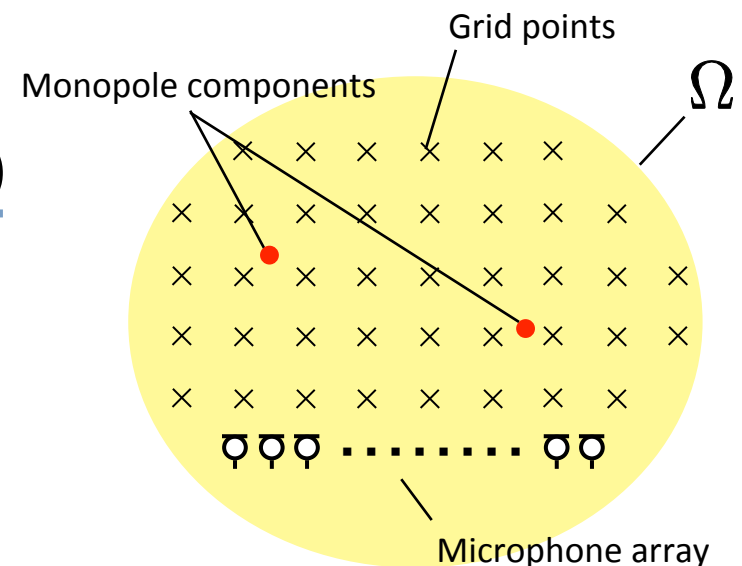
# Sparse signal decomposition

## ➤ Sparse representation of sound field

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

↓ Discretization

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



$N$  : # of grid points within  $\Omega$ ,  $M$  : # of microphones,  $N \gg M$

$\underline{y} \in \mathbb{C}^M$  : Observed signal,

$\underline{x} \in \mathbb{C}^N$  : Distribution of monopole components,

$\underline{D} \in \mathbb{C}^{M \times N}$  : Dictionary matrix of Green's functions

$\underline{h} \in \mathbb{C}^M$  : Ambient components

A few elements of  $\underline{x}$  may have non-zero components



# Sparse signal decomposition

## ➤ Sparse signal representation in vector form

$$\underline{\mathbf{y}} \in \mathbb{C}^M \quad \underline{\mathbf{D}} \in \mathbb{C}^{M \times N} \quad \underline{\mathbf{x}} \in \mathbb{C}^N \quad \underline{\mathbf{h}} \in \mathbb{C}^M$$

The diagram shows the equation  $\underline{\mathbf{y}} = \underline{\mathbf{D}} \underline{\mathbf{x}} + \underline{\mathbf{h}}$ .   
 -  $\underline{\mathbf{y}} \in \mathbb{C}^M$  is represented by a vertical column of 8 colored squares (red, blue, green, yellow, light green, magenta, blue, green).   
 -  $\underline{\mathbf{D}} \in \mathbb{C}^{M \times N}$  is represented by a 10x10 grid of various colored squares.   
 -  $\underline{\mathbf{x}} \in \mathbb{C}^N$  is represented by a vertical column of 10 squares, with the 9th square from the top being green and the others white.   
 -  $\underline{\mathbf{h}} \in \mathbb{C}^M$  is represented by a vertical column of 8 colored squares (green, yellow, light green, red, purple, blue, magenta, orange).   
 - A large black plus sign is between the  $\underline{\mathbf{D}} \underline{\mathbf{x}}$  and  $\underline{\mathbf{h}}$  terms.

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

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

$$\text{subject to } \underline{\mathbf{y}} = \underline{\mathbf{D}} \underline{\mathbf{x}}$$

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

# Structured sparsity based on physical properties

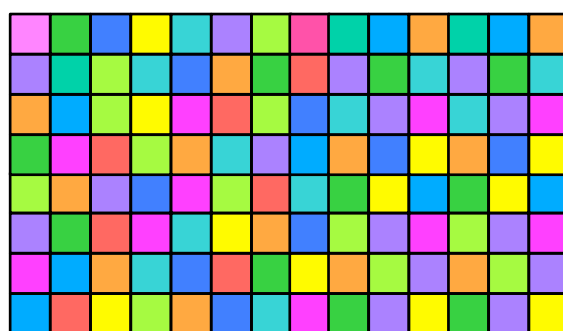
## ➤ Sparse signal representation in vector form

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



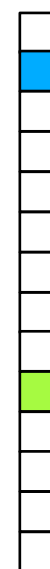
=

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



$\mathbf{x}$

Structure of sparsity induced by physical properties



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



+

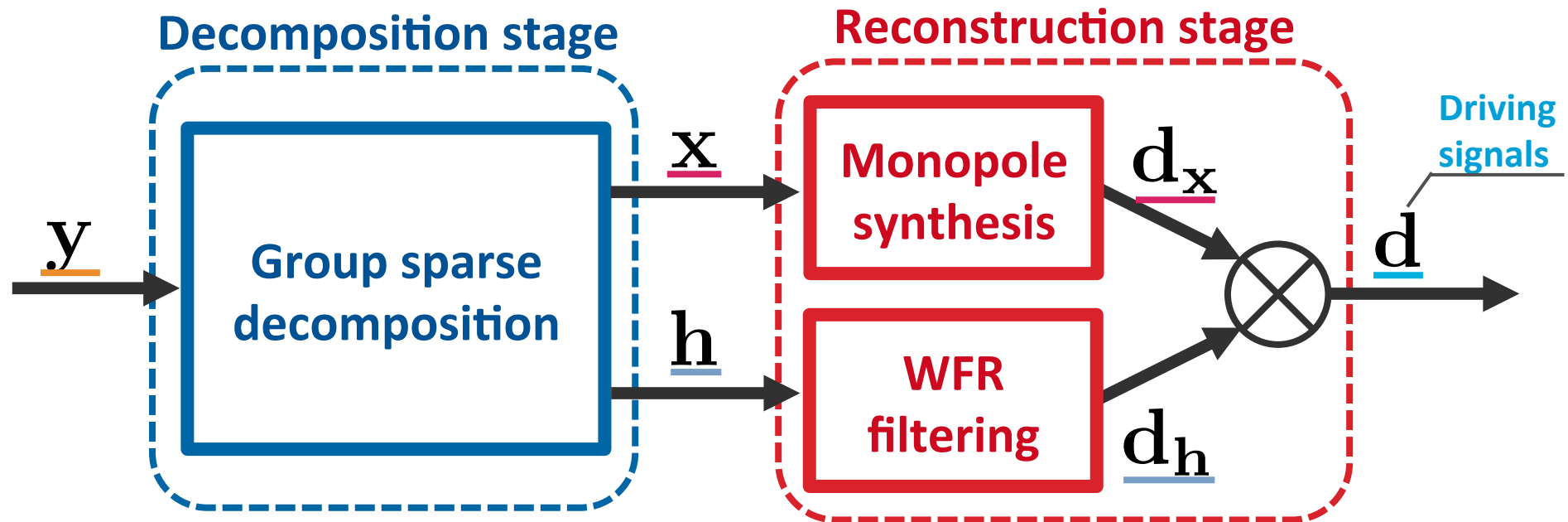
## ➤ Group sparse signal models for robust decomposition

- Multiple time frames
- Temporal frequencies
- Image sources and multipole components

## ➤ Decomposition algorithm extending M-FOCUSS

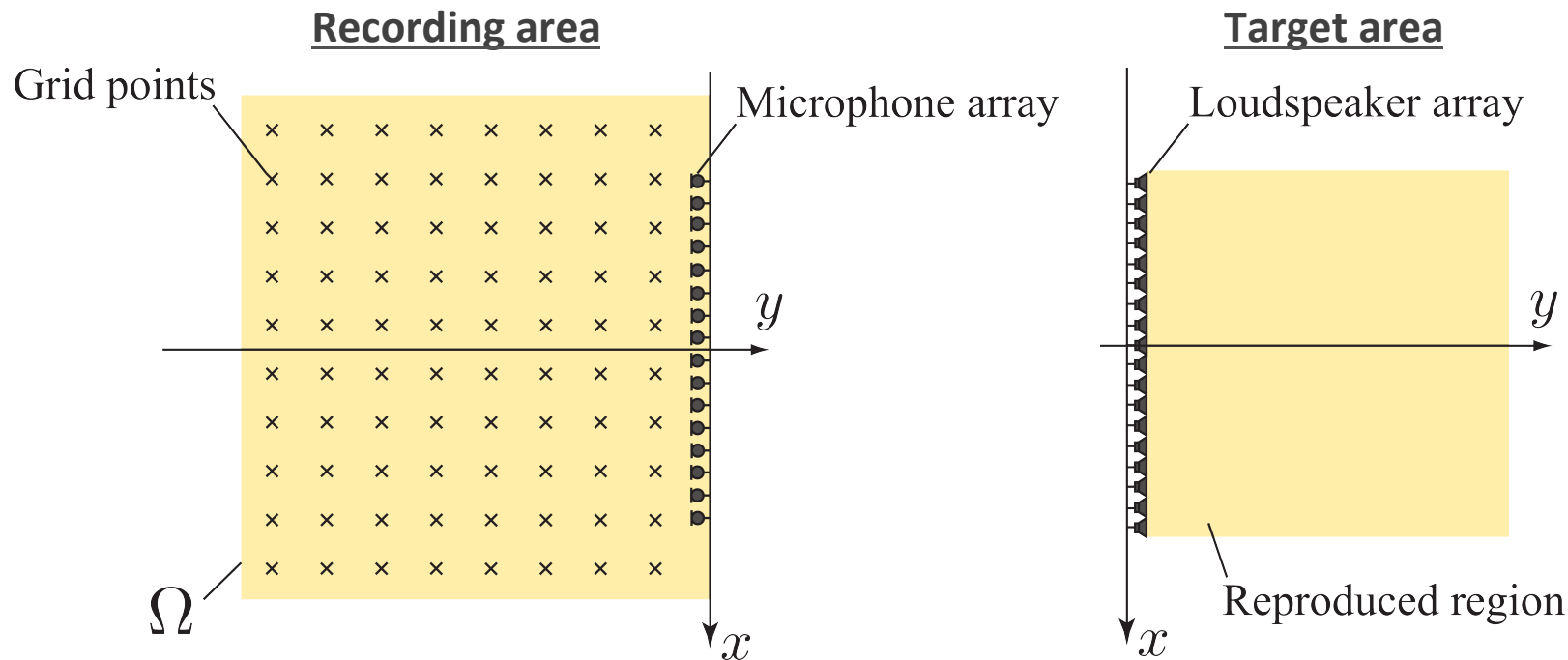
[Koyama+ ICASSP 2015 (to appear)]

# Block diagram of signal conversion



- **Decomposition stage**
  - Group sparse decomposition of  $\underline{y}$
- **Reconstruction stage**
  - $\underline{x}$  and  $\underline{h}$  are respectively converted into driving signals
  - $\underline{d}$  is obtained as sum of  $\underline{d_x}$  and  $\underline{d_h}$

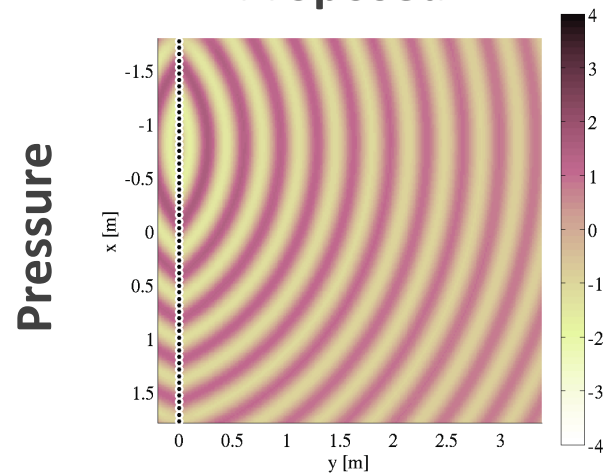
# Simulation Experiment



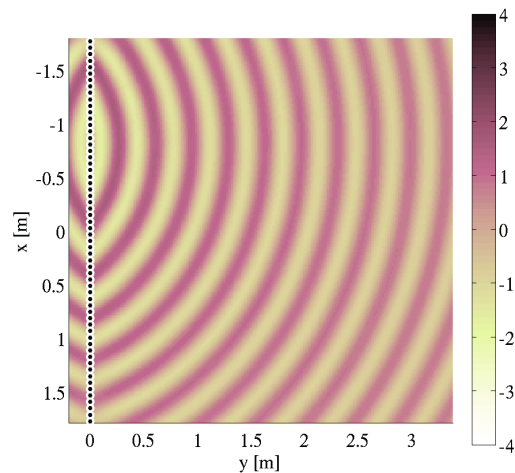
- Proposed method (**Proposed**), WFR filtering method (**WFR**), Sound pressure control method (**SPC**) were compared
- 32 microphones (12 cm intervals) and 64 loudspeakers (6 cm intervals)
- $\Omega$ : Rectangular region of 4.0x4.0 m, Grid points: (10cm, 20cm) intervals
- Source location: (-0.82, -0.86, 0.0) m

# Reproduced sound pressure distribution (1.0 kHz)

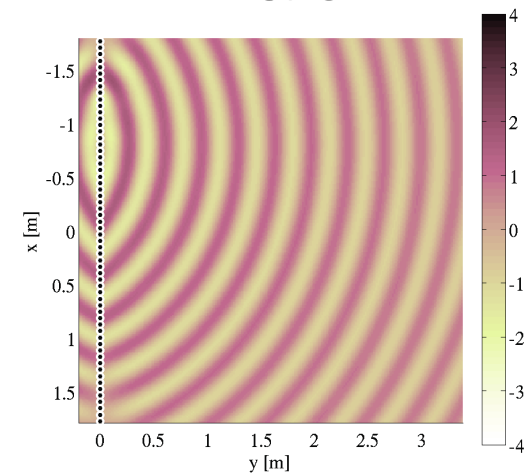
**Proposed**



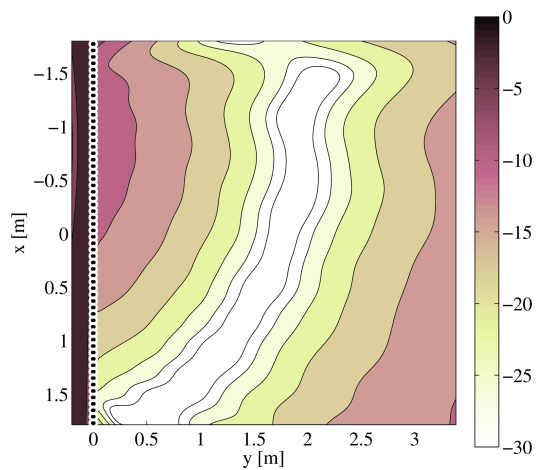
**WFR**



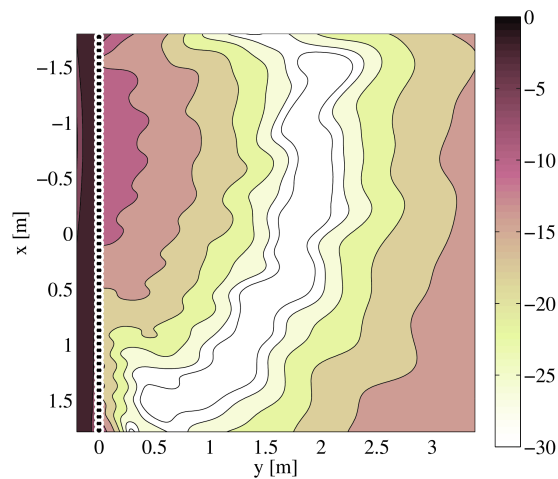
**SPC**



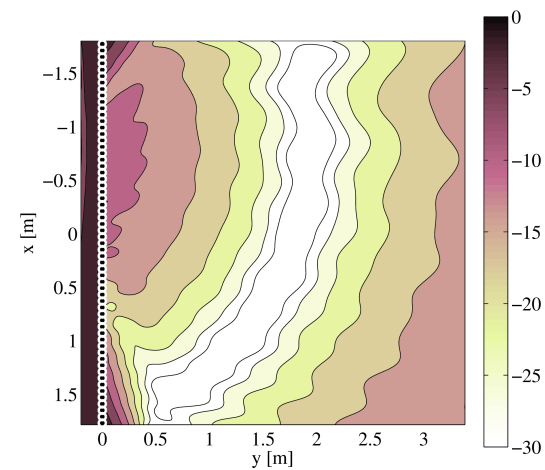
**Error**



16.4 dB



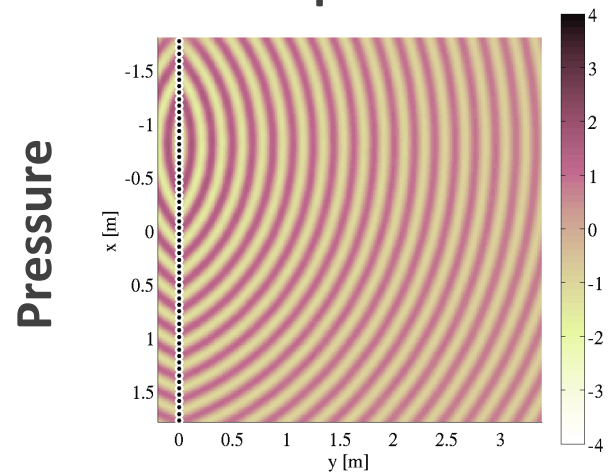
16.5 dB



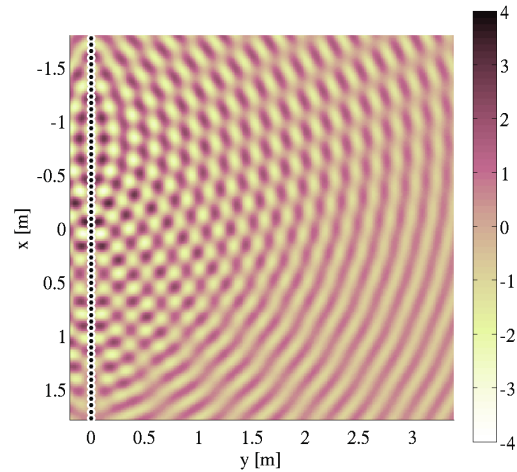
16.5 dB

# Reproduced sound pressure distribution (1.8 kHz)

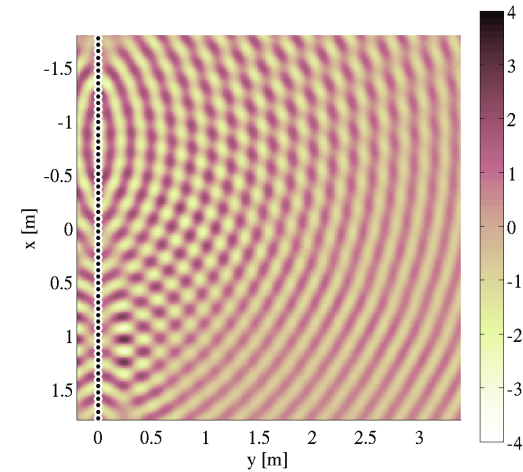
**Proposed**



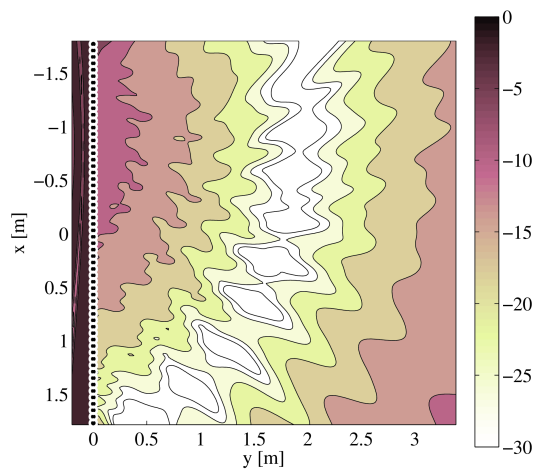
**WFR**



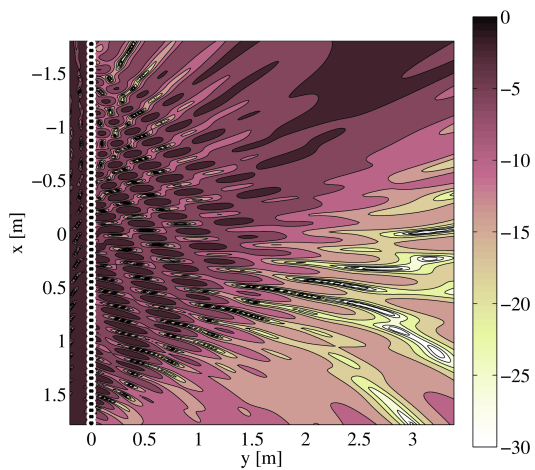
**SPC**



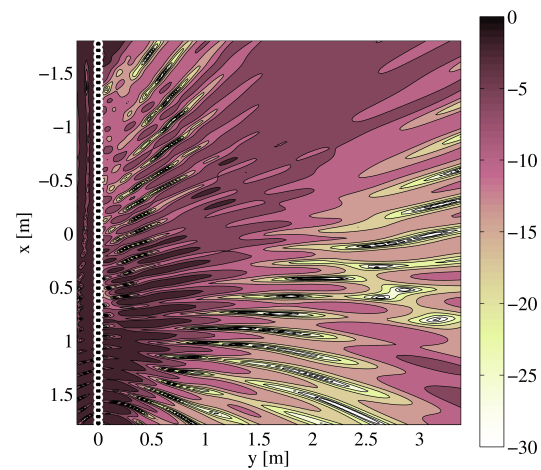
**Error**



16.2 dB

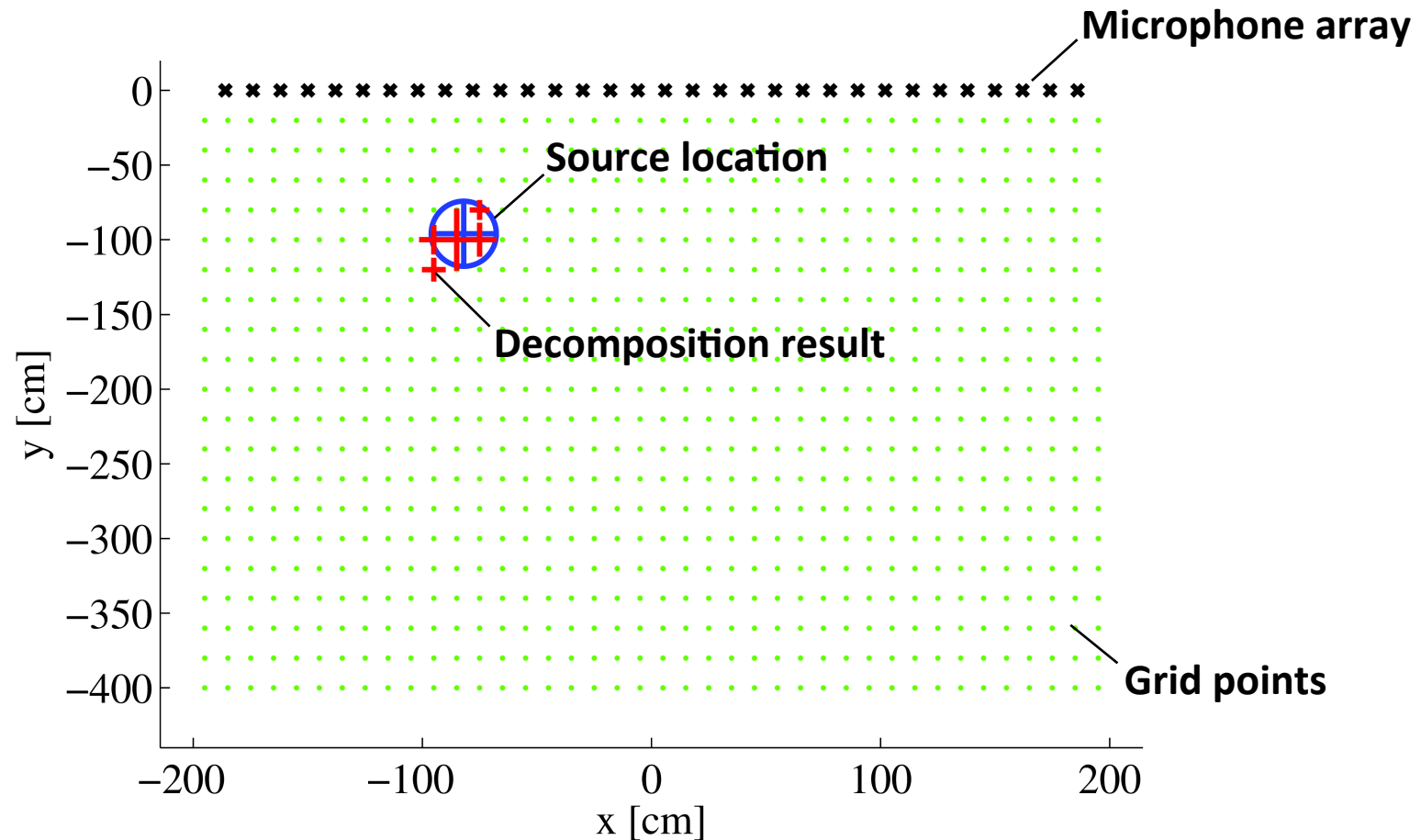


4.8 dB



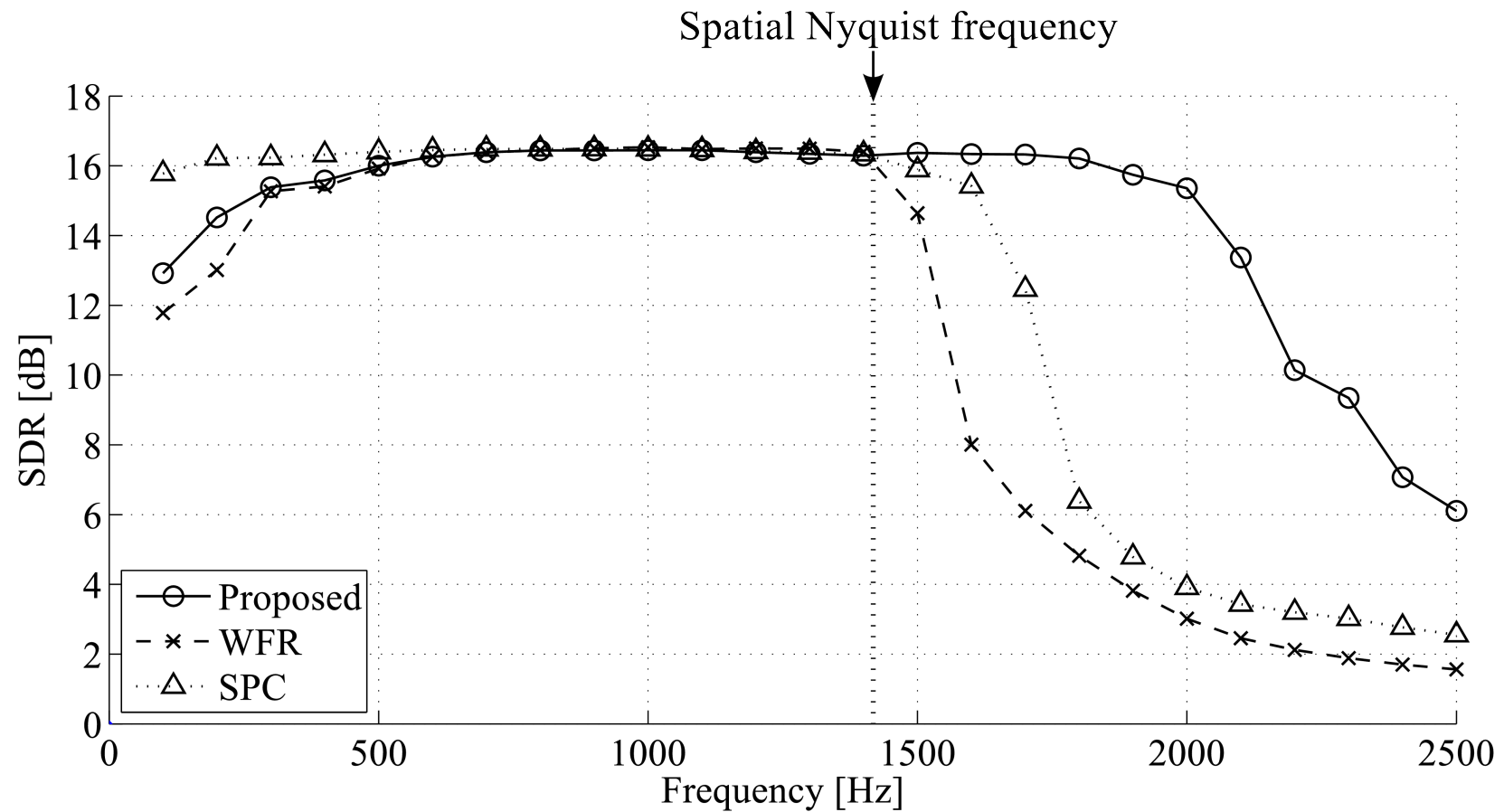
6.4 dB

# Decomposition result (1.8 kHz)



A few dictionaries around source location is used for representing observed array signals

# Frequency vs. SDR



**SDRs above spatial Nyquist frequency were improved**



# Super-resolution in Sound Field Recording and Reproduction

- ✧ Two approaches to super-resolution in sound field recording and reproduction
- ✧ Source-location-informed recording and reproduction use prior information on source locations
- ✧ Sparse sound field representation based on generative model as a sum of monopole source and plane wave components
- ✧ Reproduction accuracy above spatial Nyquist frequency was improved in both methods

## Related publications

- S. Koyama, *et al.* “Structured sparse signal models and decomposition algorithm for super-resolution in sound field recording and reproduction,” *Proc. IEEE ICASSP*, 2015 (to appear).
- S. Koyama, *et al.* “Sparse sound field representation in recording and reproduction for reducing spatial aliasing artifacts,” *Proc. IEEE ICASSP*, pp. 4476-4480, 2014.
- S. Koyama, *et al.* “MAP estimation of driving signals of loudspeakers for sound field reproduction from pressure measurements,” *Proc. IEEE WASPAA*, 2013.

# Conclusion

## ➤ Signal Conversion for Sound Field Recording and Reproduction

- Wave field reconstruction filtering method for fast and stable signal conversion for recording and reproduction
- Real-time sound field transmission system

## ➤ Super-resolution in Sound Field Recording and Reproduction

- Source-location-informed sound field recording and reproduction
- Super-resolution based on sparse sound field representation

***Thank you for your attention!***