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Noise Source Identification in an Under-Determined System by Convex Optimization

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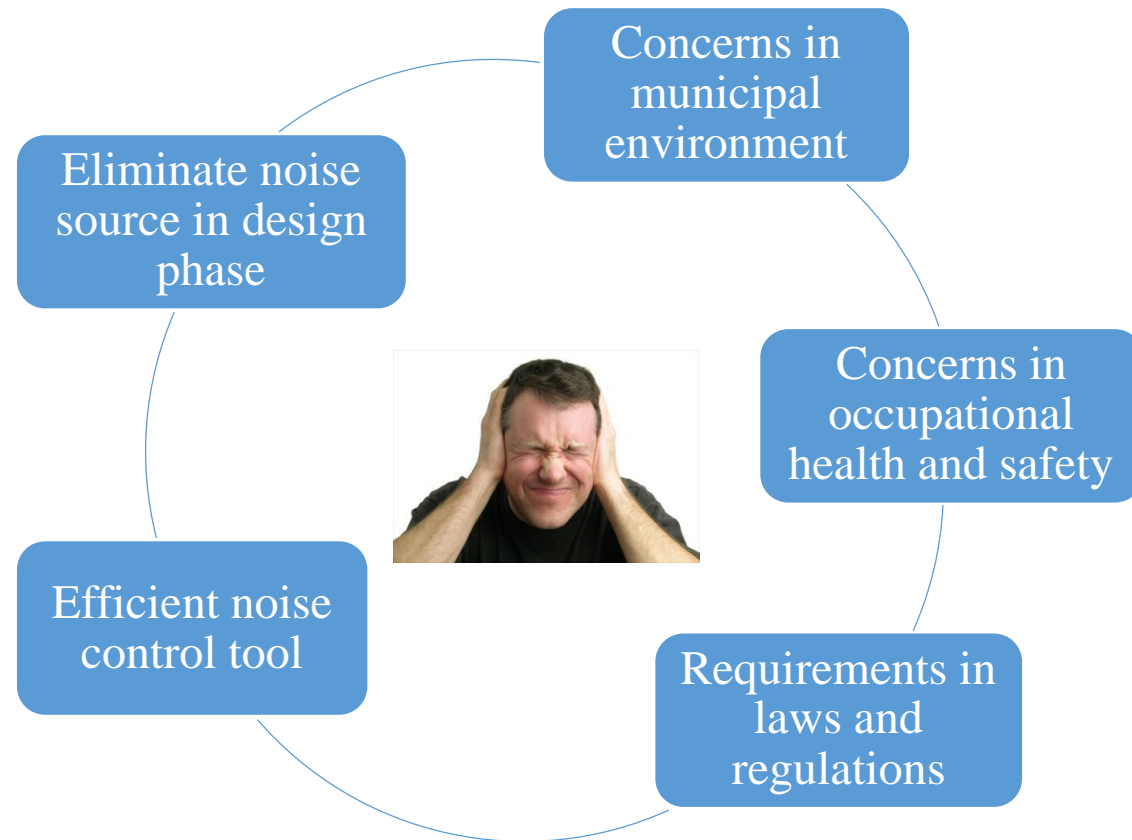
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Impact of Noise Control Engineering CHICAGO, ILLINOIS

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Sound Source Localization

- Importance of Sound Source Identification



- Near-field Acoustics Holography (NAH)

- Inverse Fourier Method
- Statistically Optimized Near Acoustical Holography (SONAH)
- Inverse Boundary Element Method (IBEM)
- Equivalent Source Method (ESM)
- Inverse Radiation Mode
 - Jiawei Liu, “Noise source Identification based on an Inverse Radiation Mode Procedure”, Noise-Con 16, Providence, Rhode Island.

Motivation of Current Work

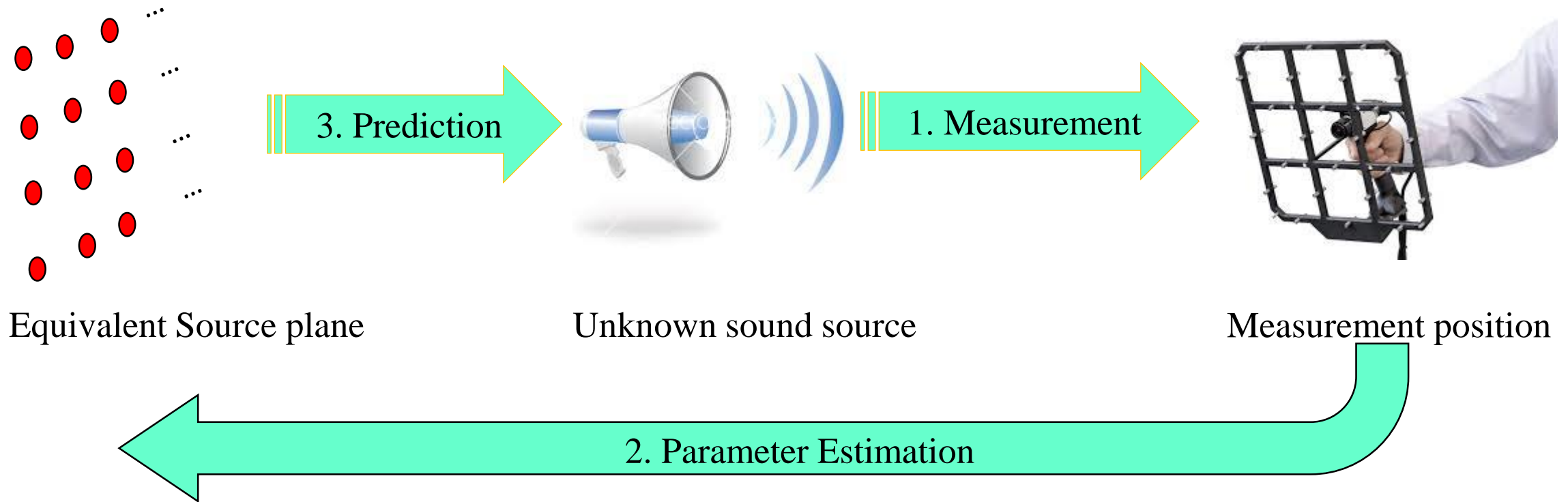
- NAH is a powerful tool to identify sound source
 - Measurements can be taken away from the source and sound field can be visualized in three-dimensional space
 - Large number of measurements is required to avoid different measurement errors: e.g., spatial aliasing, windowing errors, etc.
 - Economically costly, and hard to perform
- Motivation
 - Using a small number of microphone measurements to accurately identify major sound source locations
 - Encourage wide application of NAH in industry



Figure: LOUD 1020-node microphone array

Monopole Based Equivalent Source Method

- Idea of monopoles at fixed locations



● monopoles are used as low order equivalent sources in the present work

Mathematical Formulation

- Expression of a monopole with source strength S

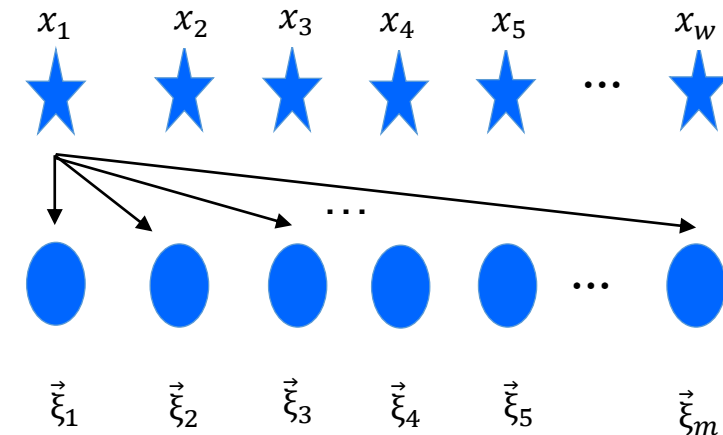
$$P_{S0}(\vec{X}|\vec{X}_0, \omega) = S \cdot P_0(\vec{X}|\vec{X}_0, \omega) = \frac{S e^{-jk\|\vec{X}-\vec{X}_0\|}}{4\pi \|\vec{X}-\vec{X}_0\|},$$

Field point position
Monopole location

- The equation of the model-generated acoustic field at all locations can be derived in a matrix form:

$$\begin{bmatrix} P_1(\vec{\xi}_1, \omega) \\ P_2(\vec{\xi}_2, \omega) \\ \vdots \\ P_M(\vec{\xi}_M, \omega) \end{bmatrix} = \begin{bmatrix} g_1(\vec{\xi}_1|\vec{X}_1, \omega) & g_2(\vec{\xi}_1|\vec{X}_2, \omega) & \dots & g_w(\vec{\xi}_1|\vec{X}_w, \omega) \\ g_1(\vec{\xi}_2|\vec{X}_1, \omega) & g_2(\vec{\xi}_2|\vec{X}_2, \omega) & \dots & \dots \\ \vdots & \vdots & \ddots & \vdots \\ g_1(\vec{\xi}_M|\vec{X}_1, \omega) & g_2(\vec{\xi}_M|\vec{X}_2, \omega) & \dots & g_w(\vec{\xi}_M|\vec{X}_w, \omega) \end{bmatrix} \begin{bmatrix} S_1(\omega) \\ S_2(\omega) \\ \vdots \\ S_w(\omega) \end{bmatrix}$$

Equivalent Sources



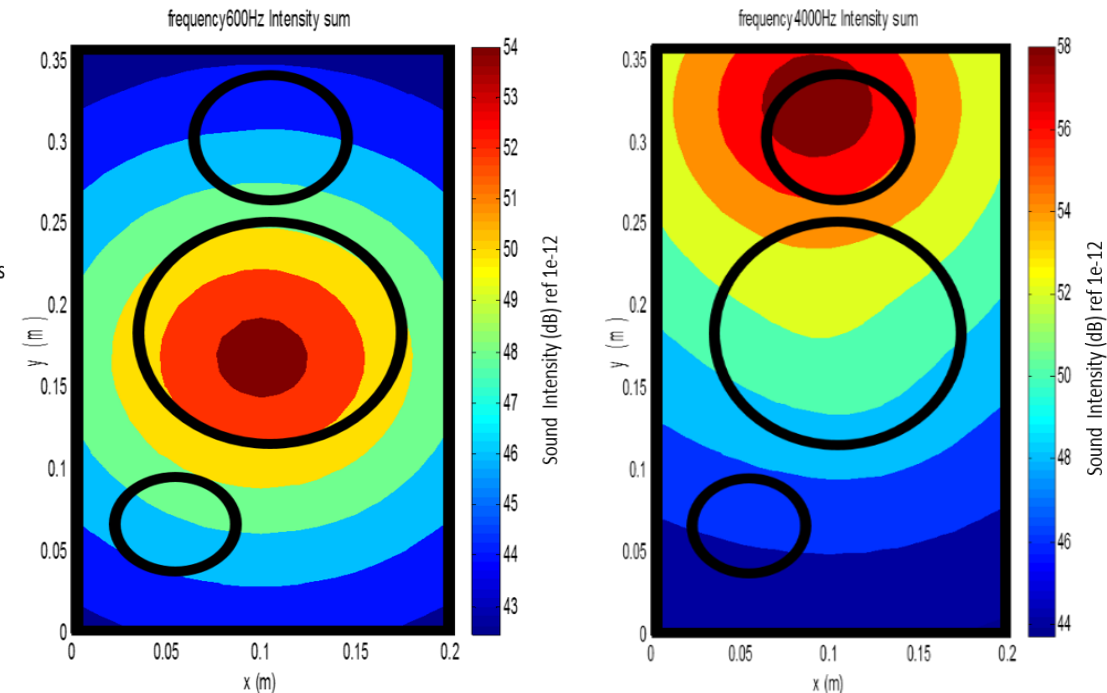
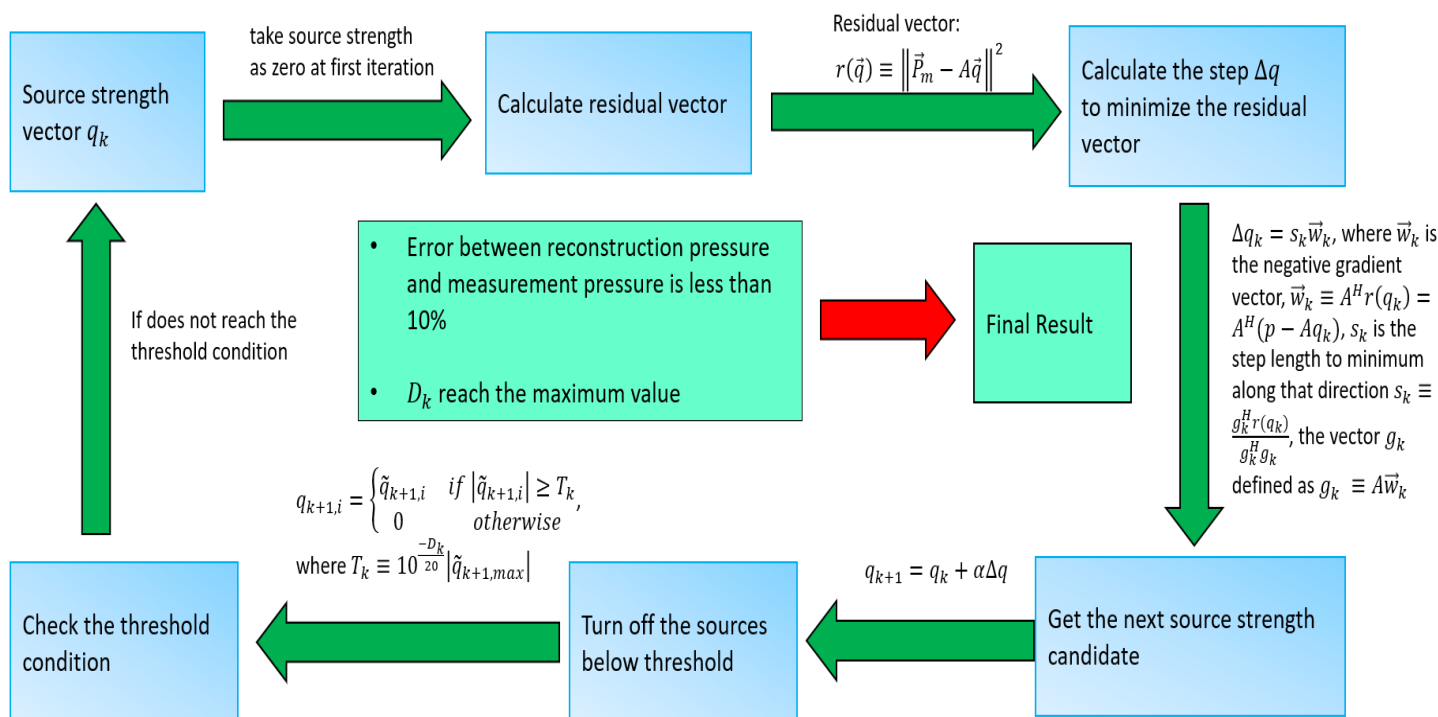
microphones

- Objective function: $\min \|\vec{P} - A(\vec{X})\vec{S}\|^2$

Previous Study

- Wideband Acoustical Holography (WBH)

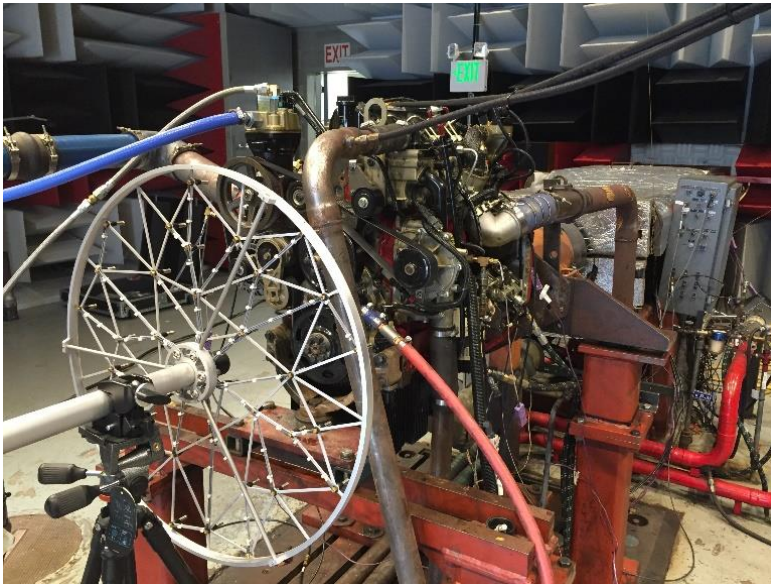
- When there is only one major sound source present, the monopole-based Wideband Acoustical Holography (WBH) can localize the sound source location and reconstruct the sound field when the system is under-determined. (T. Shi, Y. Liu, and J. Stuart Bolton. "The Use of Wideband Acoustical Holography for Noise Source Visualization." In *INTER-NOISE and NOISE-CON Congress and Conference Proceedings*, vol. 252, no. 2, pp. 479-490. Institute of Noise Control Engineering, 2016.)



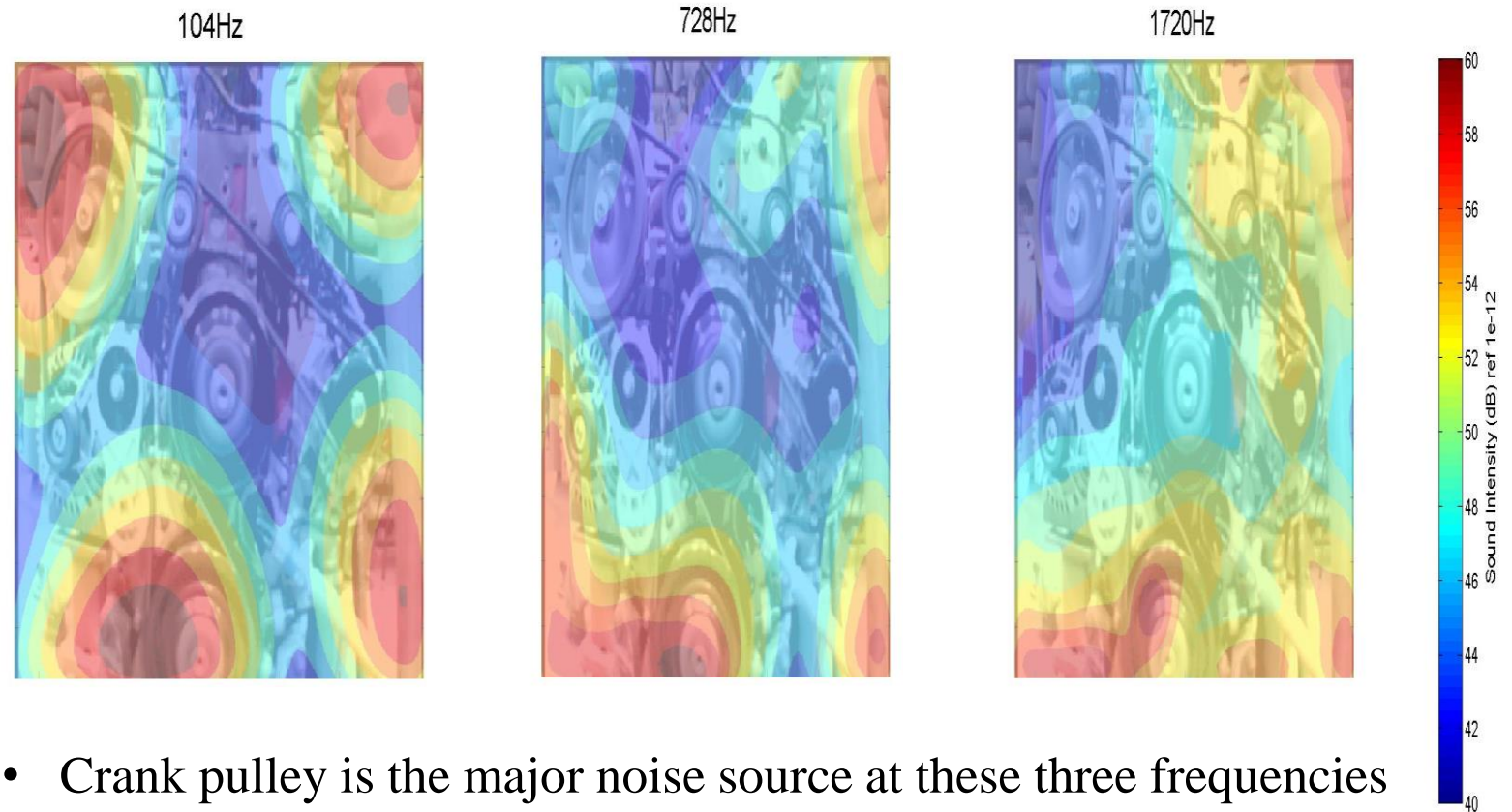
- Good results with small number of microphones

Previous Study

- Wideband Acoustical Holography (WBH)
 - Combined with Partial Field Decomposition (PFD), WBH can identify complex sound sources with a small number of measurements, e.g., diesel engine. (T. Shi, Y. Liu, J. Stuart Bolton, F. Eberhardt, and W. Frazer. "Diesel Engine Noise Source Visualization with Wideband Acoustical Holography." No. 2017-01-1874. SAE Technical Paper, 2017.)



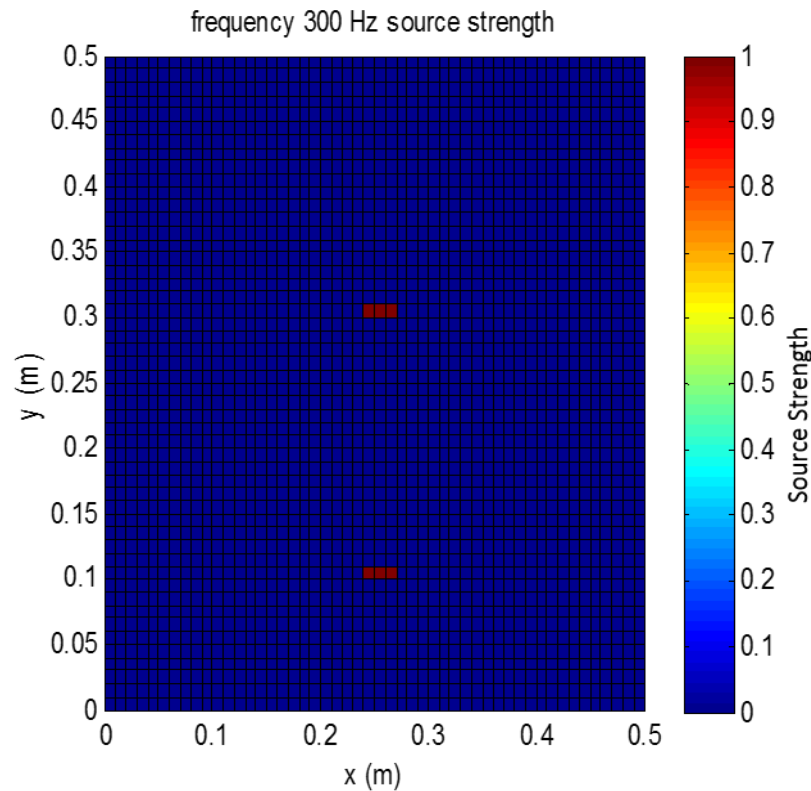
*Experimental setup in Cummins
Walesboro Noise and Vibration Lab,
Columbus, IN*



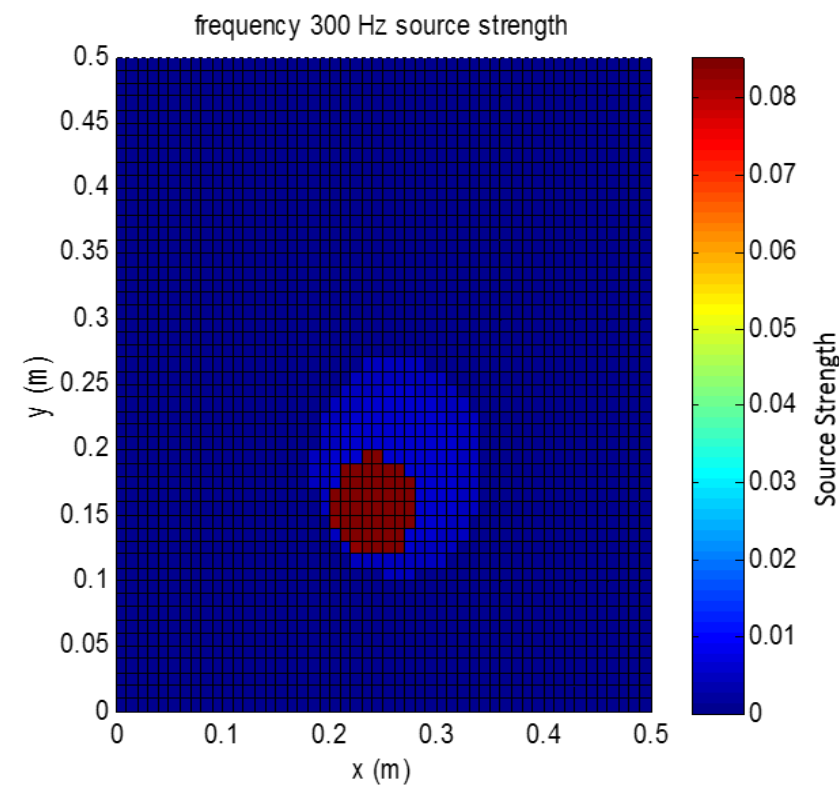
- Crank pulley is the major noise source at these three frequencies

Previous Study

- Wideband Acoustical Holography (WBH)
 - From simulation, it was found that when using WBH method, closely-positioned sources cannot be separated in space and cannot recover appropriate source strength, especially at low frequency. (T. Shi and J. S. Bolton. "Separation of closely-spaced acoustics sources in an under-determined system with convex optimization." *The Journal of the Acoustical Society of America* 143, no. 3 (2018): 1872-1872.)



True source distribution



WBH reconstructed sources

Convex Optimization

- Under-determined system
- Low spatial sampling rate



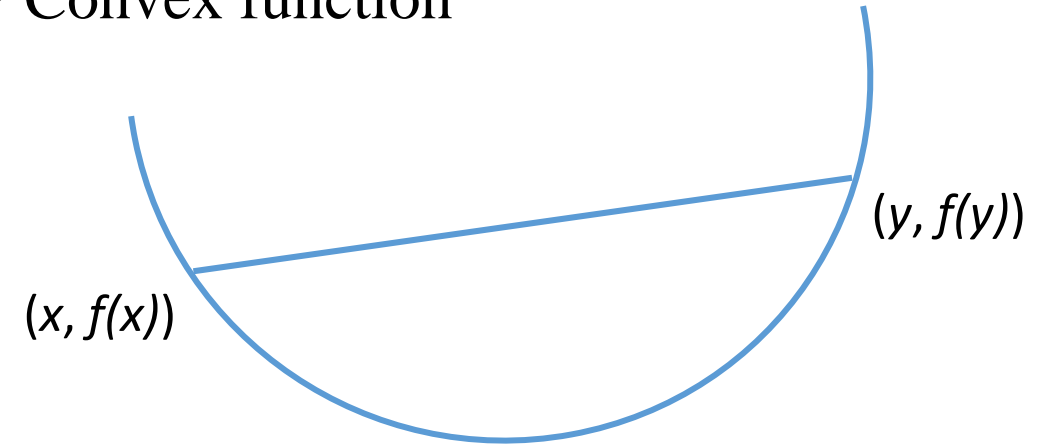
- Nyquist-Shannon sampling theorem
- Compressive Sampling (CS)



$$\min \left\| \vec{P} - A(\vec{X})\vec{S} \right\|^2$$

- Convex function
- Convex Optimization

➤ Convex function



$$\bullet \quad f(\theta x + (1 - \theta)y) \leq \theta f(x) + (1 - \theta)f(y)$$

➤ Convex optimization problem

$$\text{minimize } f_0(x)$$

$$\text{subject to } f_i(x) \leq 0, \quad i = 1, \dots, m$$

$$h_i(x) = 0, \quad i = 1, \dots, m$$

Objective Function Formulation

➤ Objective function

$$\text{minimize } \underbrace{\|\vec{S}\|_1}_{\text{Solution sparsity}} + \lambda \underbrace{\|A\vec{S} - \vec{P}_m\|_2}_{\text{Solution accuracy}}$$

Weighting parameter λ

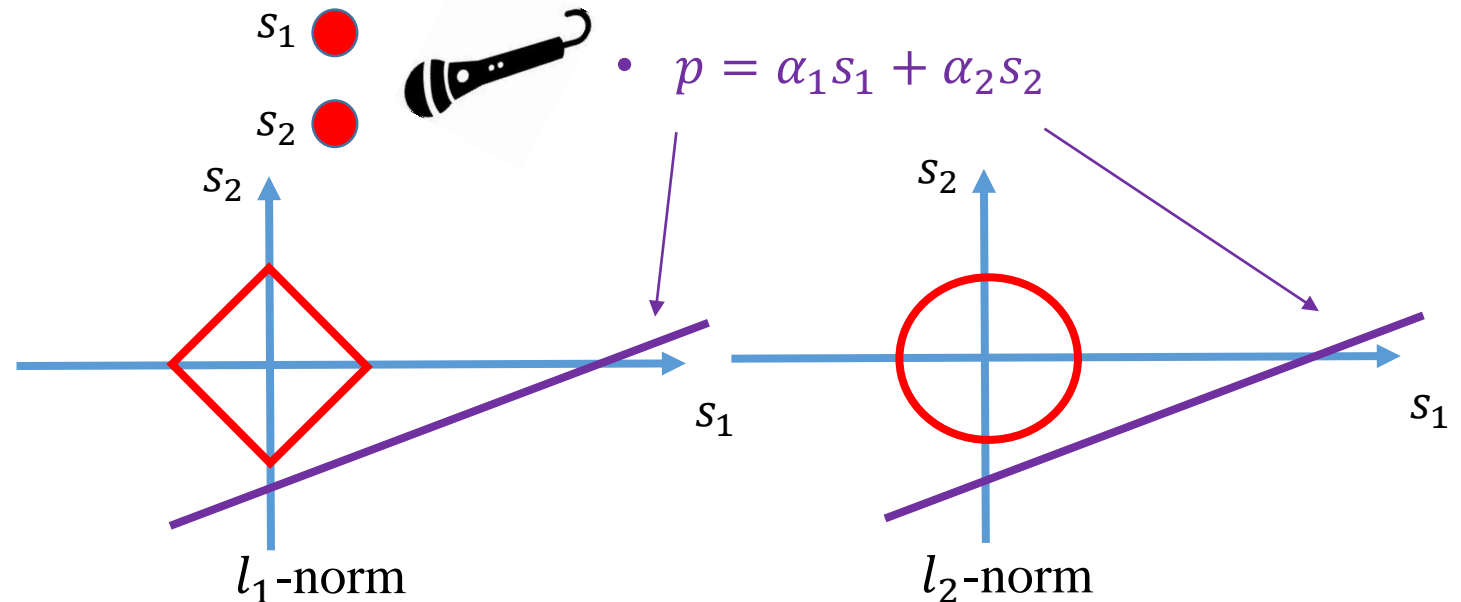


- Careful choosing of the weighting parameter
- l_1 -norm for source strength and l_2 -norm for residual

- M. Grant, S. Boyd, and Y. Ye *CVX: software for disciplined convex programming*

➤ l_1 -norm and l_2 -norm

- l_1 -norm: $\|\vec{S}\|_1 = \sum_{i=1}^m |s_i|$
- l_2 -norm: $\|\vec{S}\|_2 = \sqrt{s_1^2 + s_1^2 + \dots + s_m^2}$
- Choice of l_1 -norm ensures solution sparsity



Simulated Closely-Positioned Sources

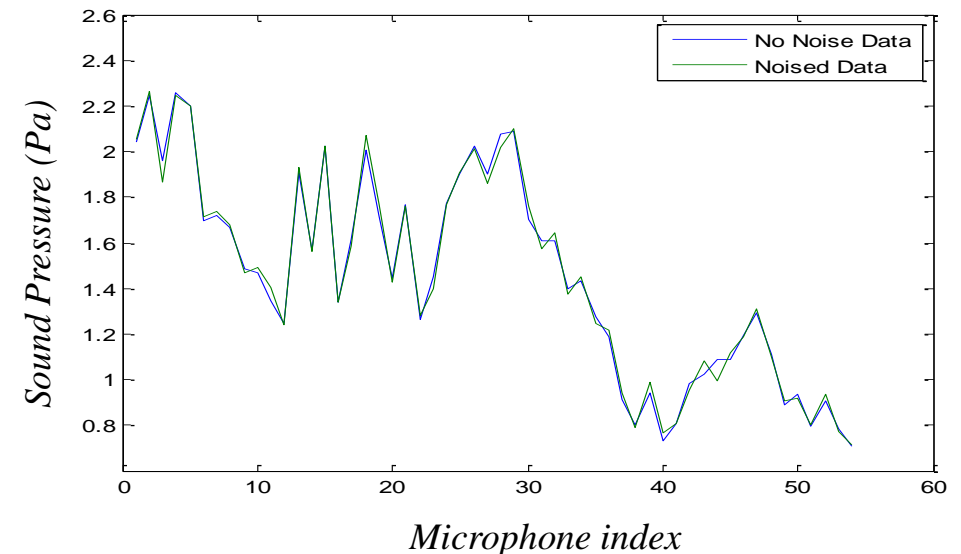
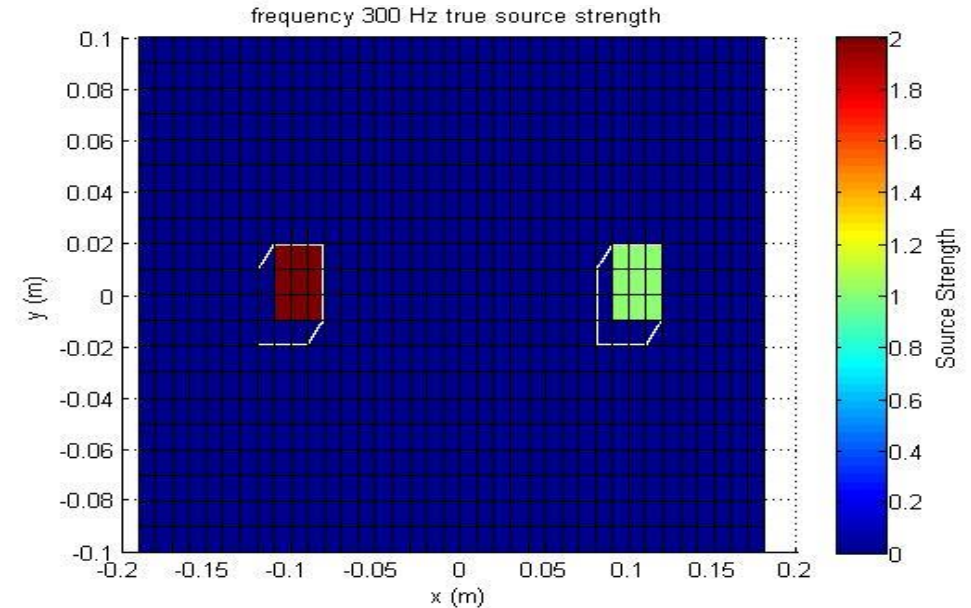
➤ Simulation set up

- Two simulated sources were placed 0.2 m from each other, source at left was composed of nine monopoles with source strength two, and source at right composed of nine unit source strength monopoles.
- 54 virtual microphones measurement 0.23 m in front of virtual sources
- 300Hz, wavelength $\lambda=1.14\text{m}$; 2000Hz, wavelength $\lambda=0.17\text{m}$

➤ Equivalent source plane

- -0.19 - 0.18 m, in x -direction
- -0.1 - 0.1 m, in y -direction
- 0.01 m spacing in both x - and y -direction, 798 monopoles.
- 0.02 m in z -direction

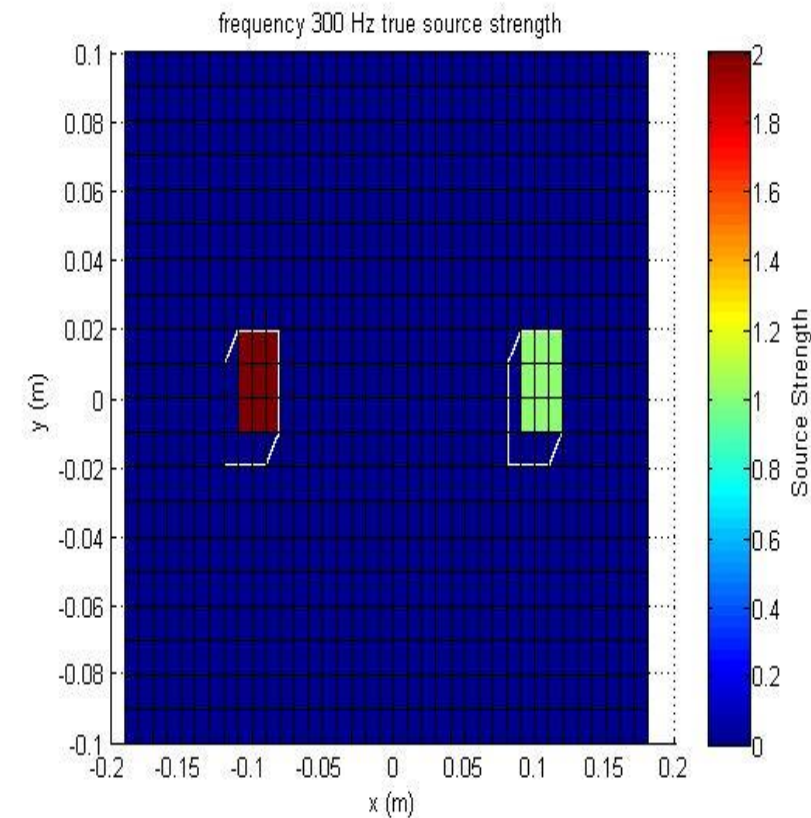
➤ White Gaussian Random Noise added into the virtual measurement, SNR = 30 dB



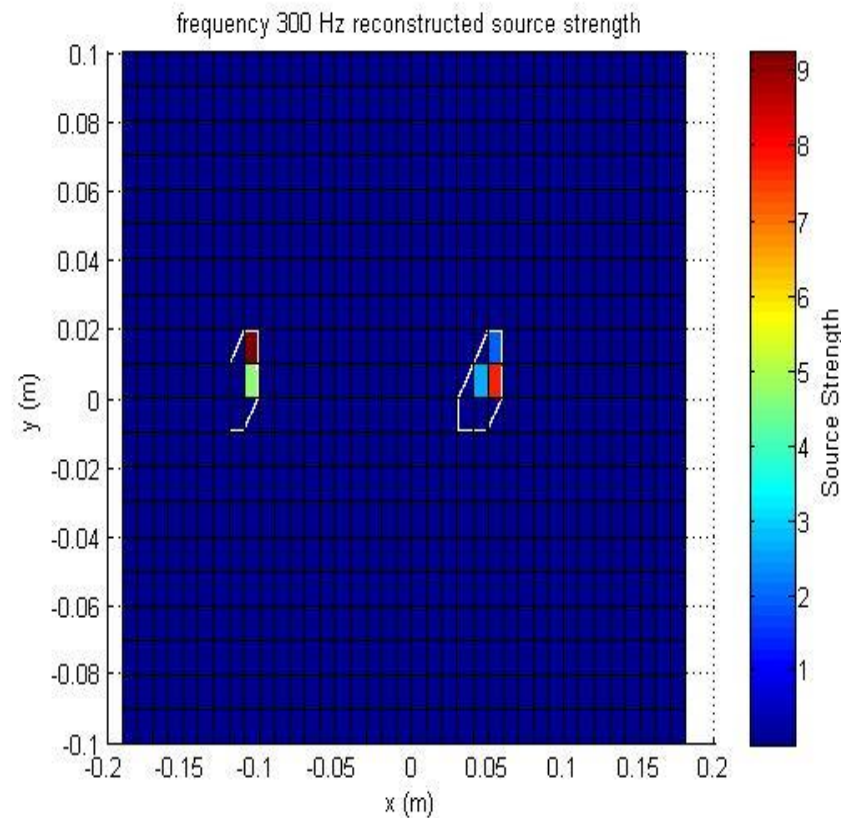
Reconstruction on Equivalent Source Plane at 300 Hz

$$\text{minimize } \|\vec{\hat{S}}\|_1 + 2 \left\| A(\vec{\hat{X}}_S) \vec{\hat{S}} - \vec{\hat{P}}_m \right\|_2$$

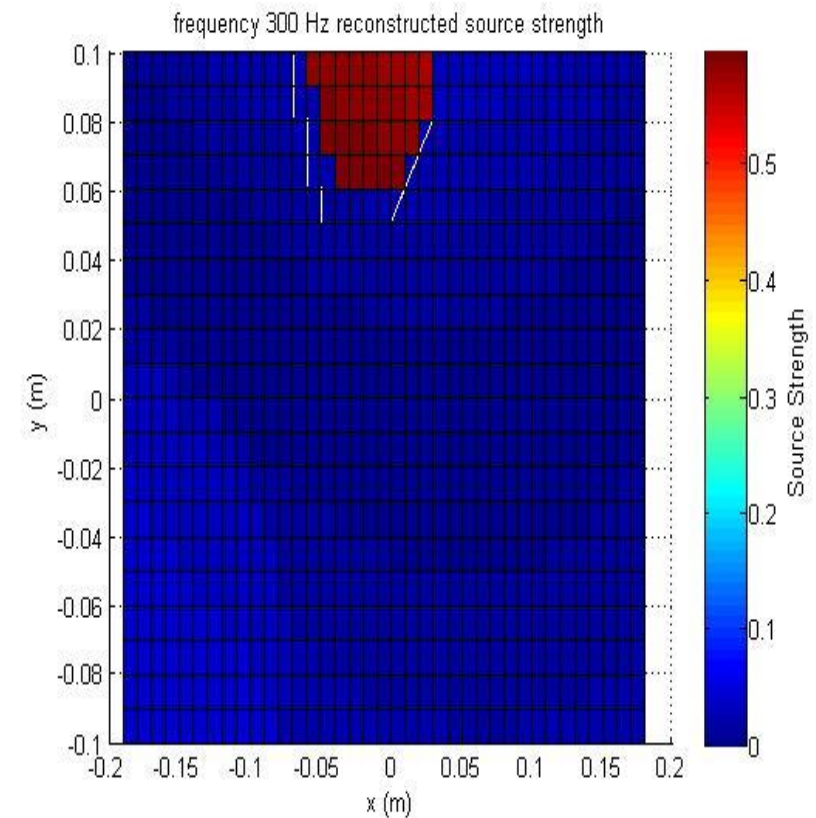
Error < 10%, D_max = 60



True sources distribution
Total source strength 27



Convex Optimization reconstructed sources
Total source strength 25.73



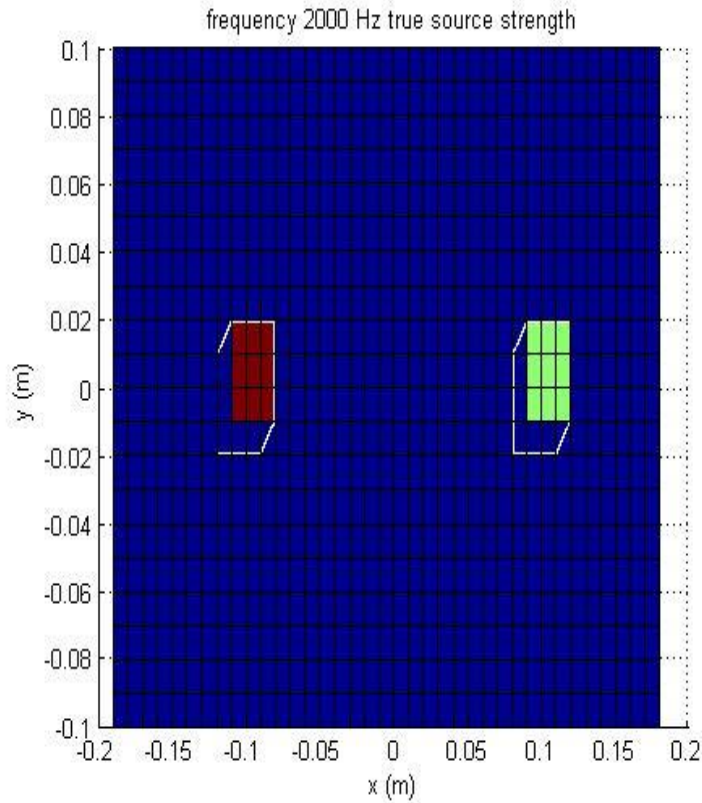
WBH reconstructed sources
Total source strength 46.53

- Two separated sources were identified by Convex Optimization near true source location with nearly-correct source strength, but WBH failed to find either correct source location or source strength.

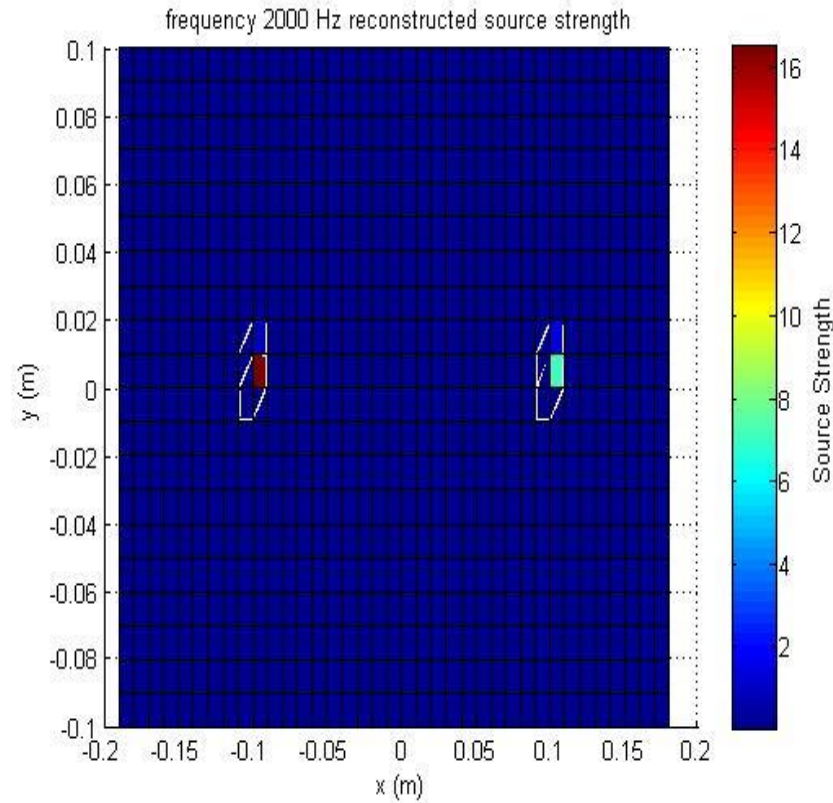
Reconstruction on Equivalent Source Plane at 2000 Hz

$$\text{minimize } \|\vec{S}\|_1 + 2 \|A(\vec{X}_S)\vec{S} - \vec{P}_m\|_2$$

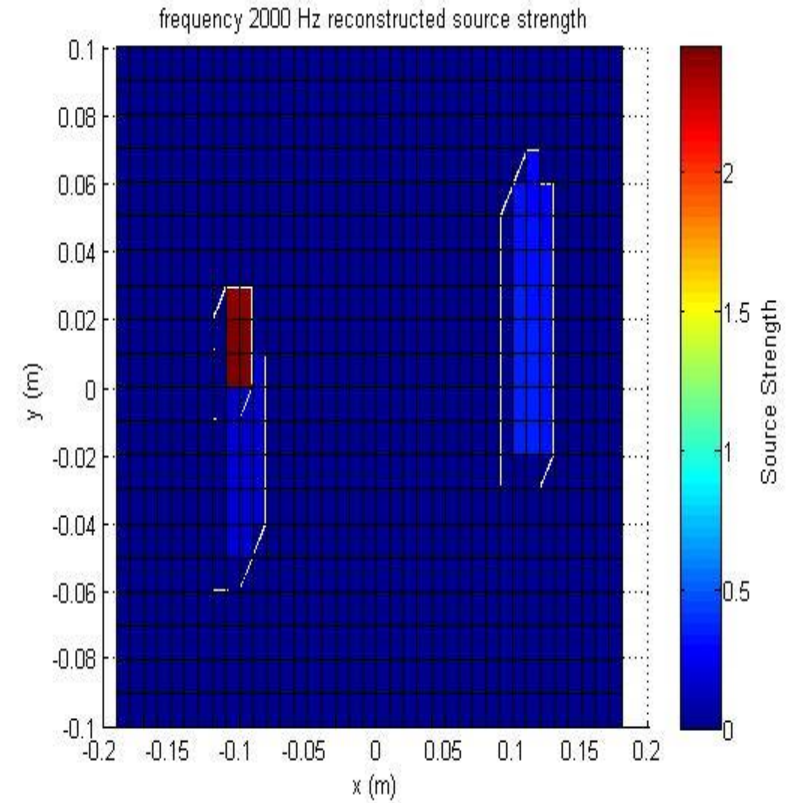
Error < 10%, D_max = 60



True sources distribution
Total source strength 27



Convex Optimization reconstructed sources
Total source strength 26.58



WBH reconstructed sources
Total source strength 27.28

- Convex Optimization reconstruction result is more concentrated and location is more accurate than WBH, and WBH underestimates the weaker source.

Loudspeaker Test

➤ Experimental set up

- Test with loudspeaker (Infinity Primus P163) as a noise source
- White noise as input
- Brule and Kjaer 18 channel irregular array
- 54 microphones measurement 0.23 m in front of loudspeaker, 10 second measurement duration

➤ Equivalent source plane

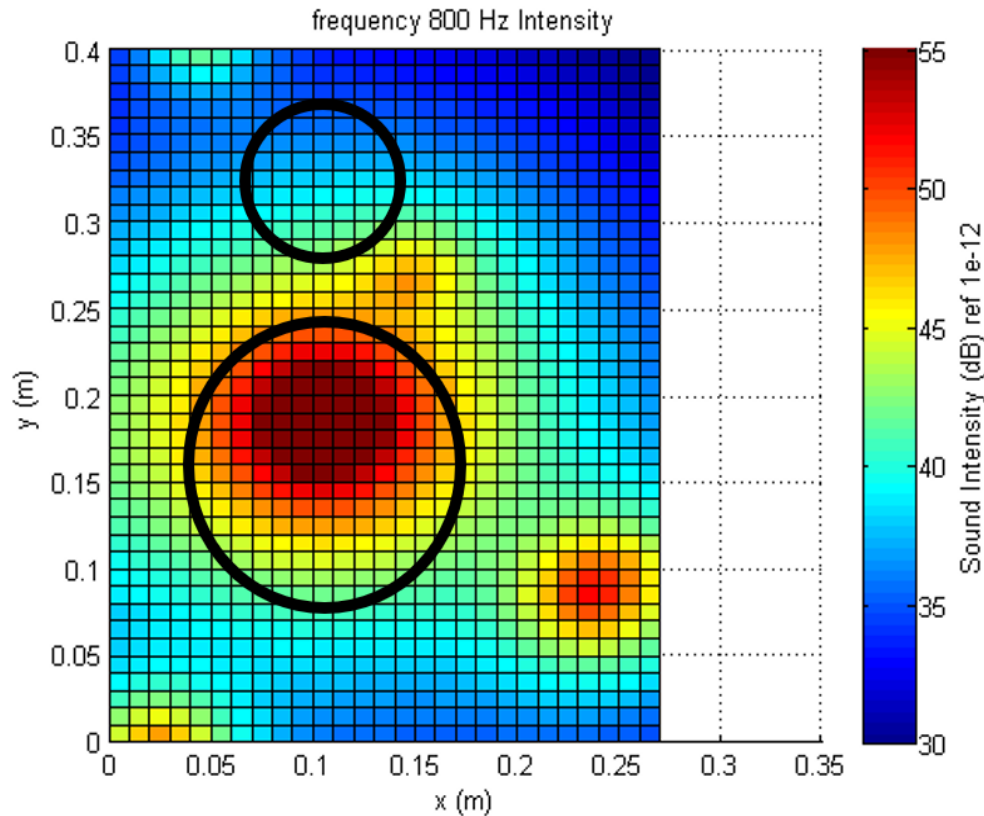
- -0.2 - 0.4 m, in x -direction
- -0.2 - 0.4 m, in y -direction
- 0.01 m spacing in both x - and y -direction, 3721 monopoles.
- 0.02 m in z -direction
- Sound intensity was reconstructed on loudspeaker front face



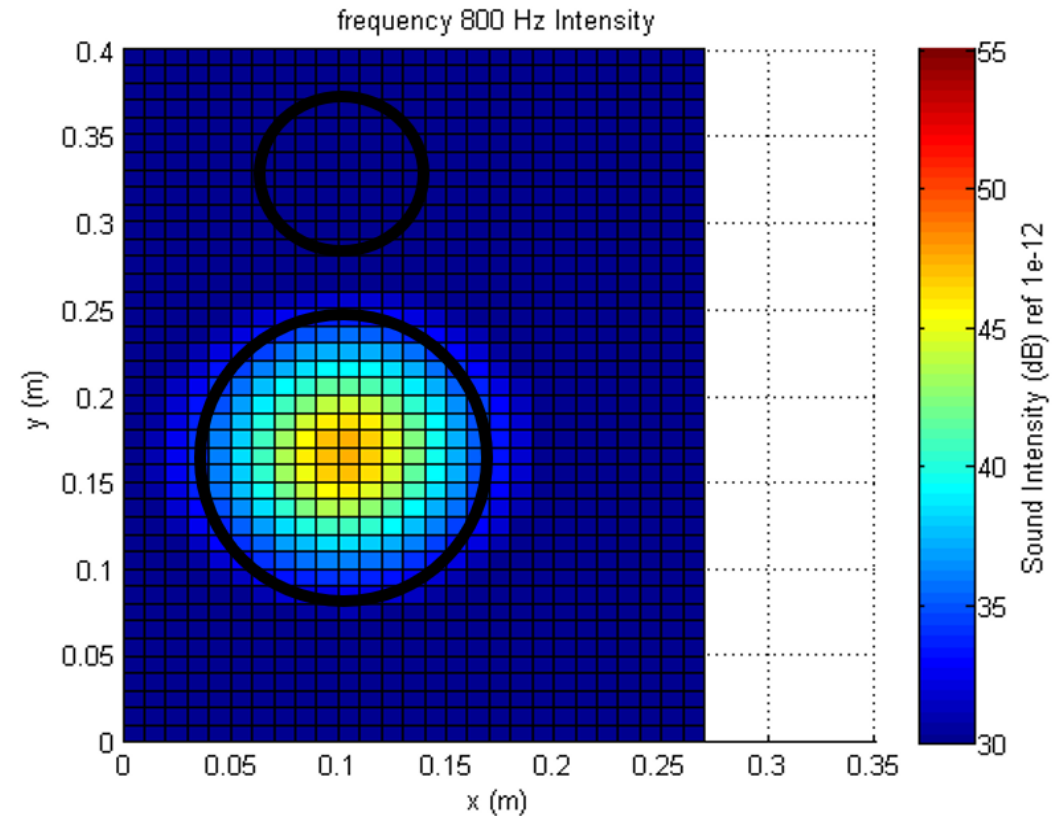
Sound Intensity Reconstruction on Loudspeaker Front Face at 800 Hz

$$\text{minimize } \|\vec{S}\|_1 + 10 \left\| A(\vec{X}_S) \vec{S} - \vec{\hat{P}}_m \right\|_2$$

Error < 10%, D_max = 100



Convex Optimization



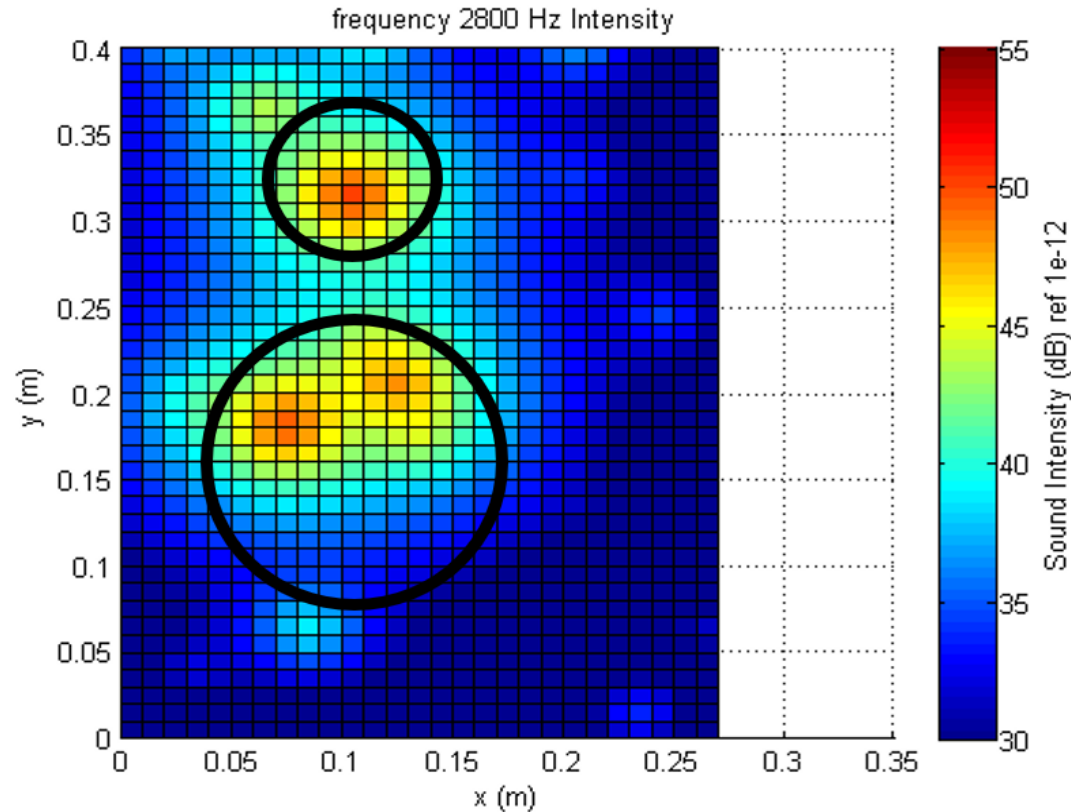
WBH

- WBH works well when only a single significant source
- Ghost sources appear in convex optimization case

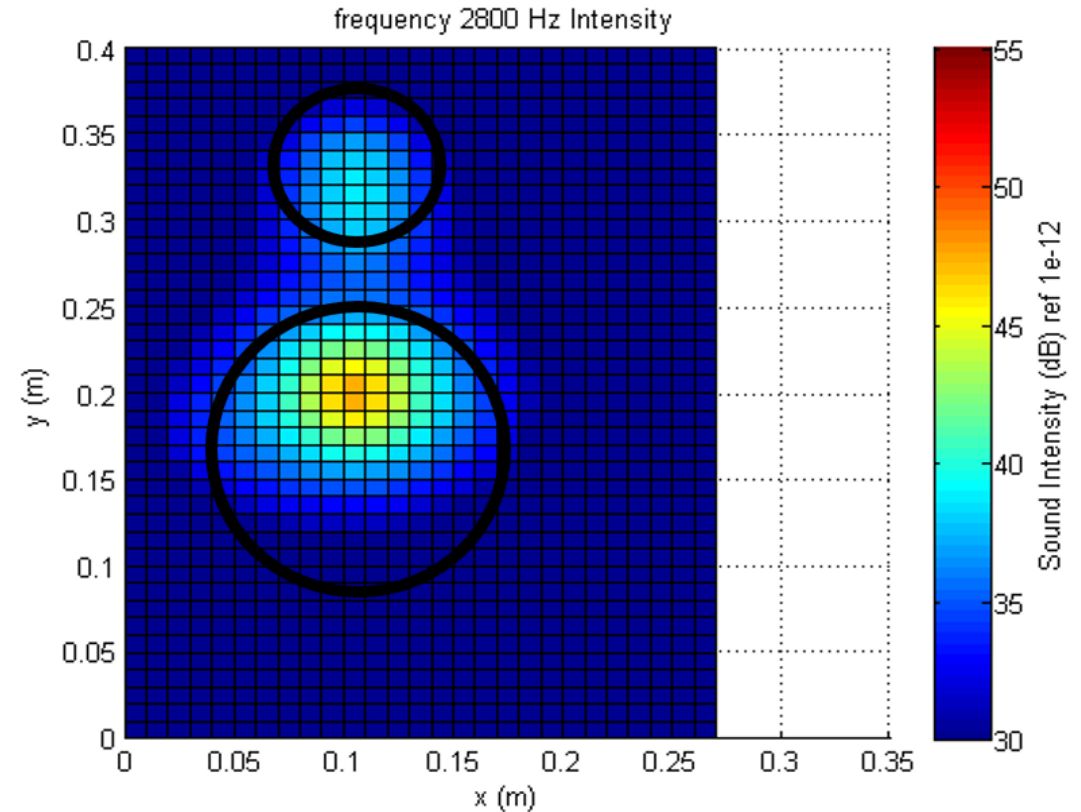
Sound Intensity Reconstruction on Loudspeaker Front Face at 2800 Hz

$$\text{minimize } \|\vec{S}\|_1 + 10 \|A(\vec{X}_S)\vec{S} - \vec{P}_m\|_2$$

Error < 10%, D_max = 100



Convex Optimization



WBH

- WBH underestimates contribution from tweeter
- Convex optimization gives more balanced result

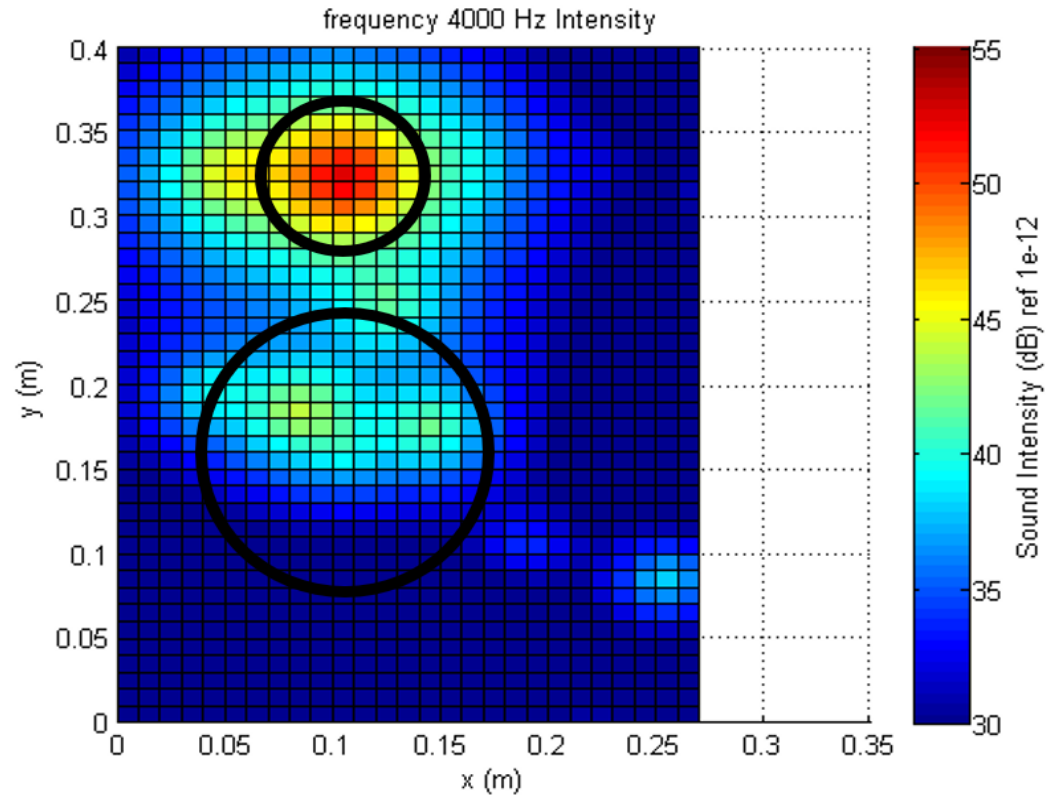
Sound Intensity Reconstruction on Loudspeaker Front Face at 4000 Hz

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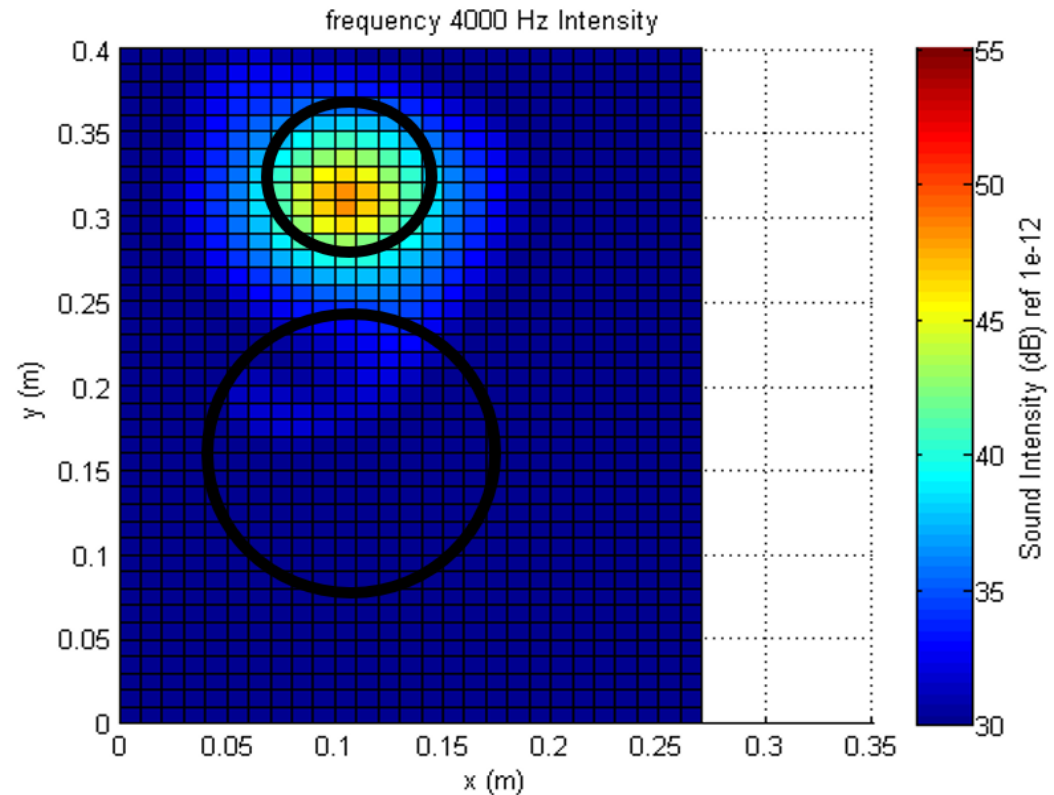


$$\text{minimize } \|\vec{S}\|_1 + 10 \|A(\vec{X}_S)\vec{S} - \vec{P}_m\|_2$$

Error < 10%, D_max = 100



Convex Optimization



WBH

- WBH underestimates contribution from diaphragm

Difference between WBH and Convex Optimization

➤ Wideband holography method **roughly** equivalent to

$$\text{minimize } \left\| A(\vec{X}_S) \vec{S} - \overrightarrow{\hat{P}_m} \right\|_2 \quad \circ \text{ Steepest gradient method}$$

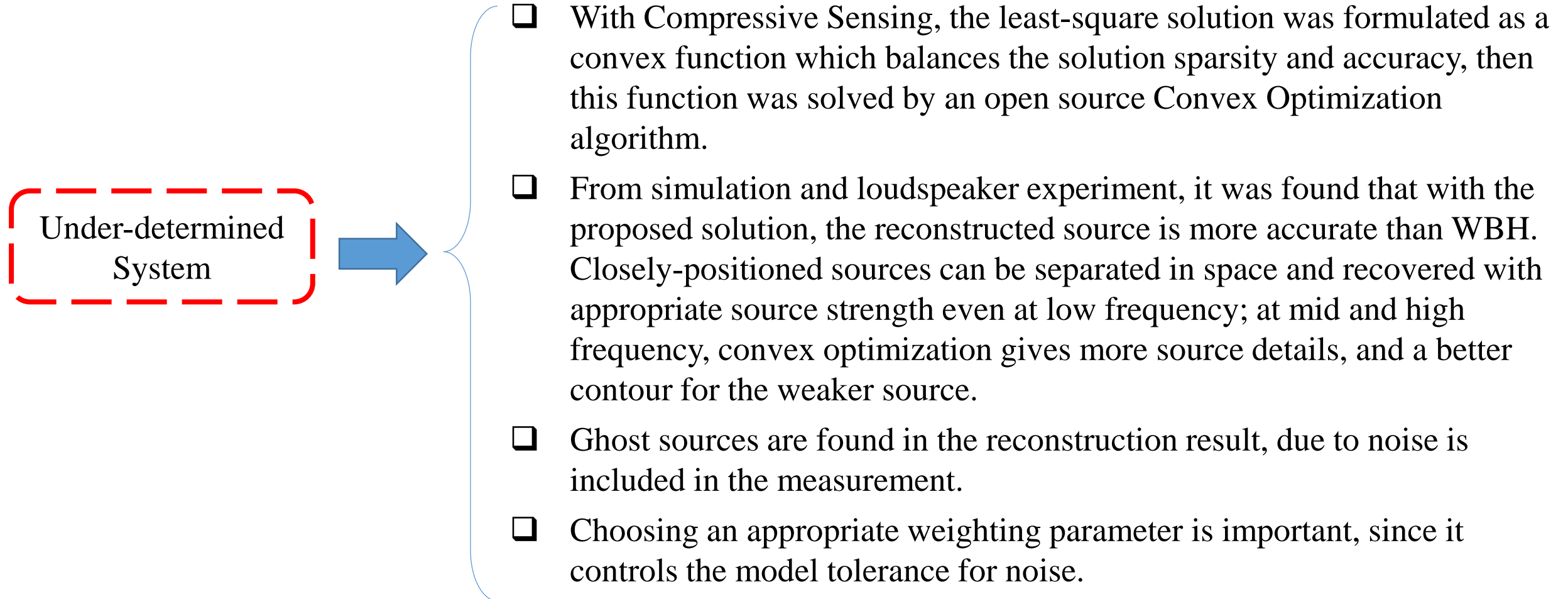
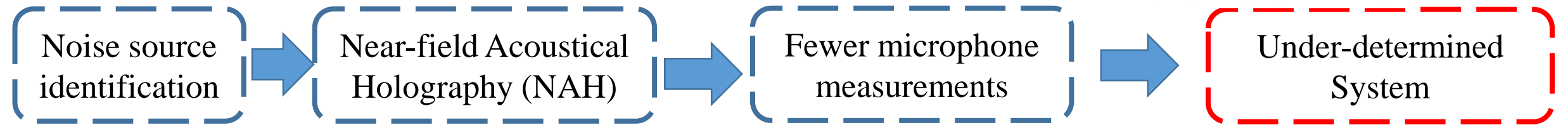
$$\text{subject to } \mathbf{card}(\vec{S}) \leq \tilde{t}_k \quad \circ \text{ Adjust } \tilde{t}_k \text{ at each iteration through } T_k$$

- Different formulation to create the solution sparsity

➤ Convex formulation

$$\text{minimize } \left\| \vec{S} \right\|_1 + \lambda \left\| A(\vec{X}_S) \vec{S} - \overrightarrow{\hat{P}_m} \right\|_2$$

Conclusions



Acknowledgement



- The authors are grateful for the funding provided by Cummins Inc., and for helpful discussions with the contract monitor, Frank Eberhardt

*Thank
you*

A detailed illustration of a fountain pen with a gold-colored nib and a black barrel, positioned as if it has just finished writing the word "you".