

# Statistical Room Acoustics in Acoustic Sensor Networks

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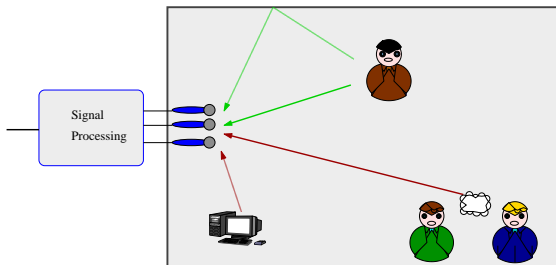
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# Outline

- 1 Acoustic Sensor Networks
- 2 Multi-channel Wiener Filter
- 3 Distributed MWF
- 4 Spatial expectation of output SNR using statistical room acoustics
- 5 Experimental results
- 6 Conclusion

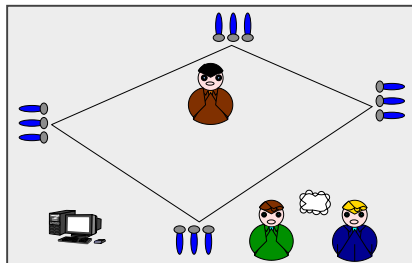
## Acoustic Sensor Networks

- Signal acquisition in **adverse acoustic environments**
- "Traditional" microphone arrays:
  - Limited number of microphones (specific configuration)
  - Microphones possibly at large distance from desired source → background noise and reverberation



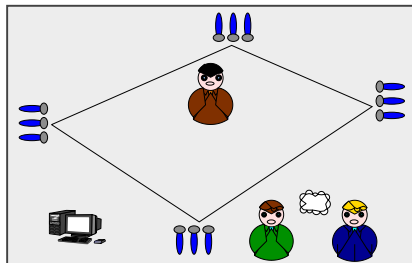
## Acoustic Sensor Networks

- Signal acquisition in **adverse acoustic environments**
- Acoustic sensor networks:
  - Network of a large number of **spatially distributed** nodes, typically at unknown positions
  - More information about spatial sound field (microphones with higher SNR, direct-to-reverberant ratio)
  - Wired or wireless data transmission



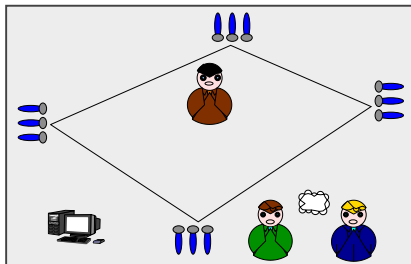
## Acoustic Sensor Networks

- Signal acquisition in **adverse acoustic environments**
- Prototype applications:
  - Hearing aids using extra microphones (room, other HA, ...)
  - Video-conferencing using all microphones on laptops / room
  - Surveillance



## Acoustic Sensor Networks

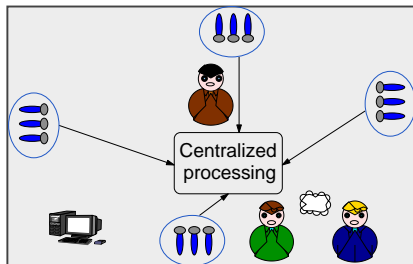
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  - *Dynamic array configuration:* large number of microphones at unknown positions, dynamic subset selection



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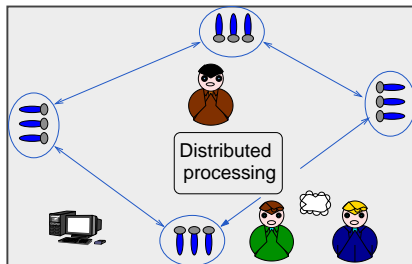
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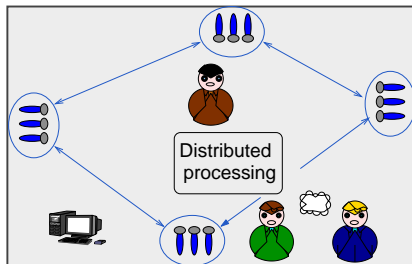
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  - *Dynamic array configuration:* large number of microphones at unknown positions, dynamic subset selection
  - *Distributed and collaborative algorithms:* power and complexity constraints, effect of limited bandwidth and coding
  - *Calibration and synchronisation issues*



## Objectives

- Performance of signal enhancement algorithms depends on **acoustical scenario**:
  - microphone configuration  $\mathbf{P}_{mic}$
  - desired source position  $\mathbf{p}_s$
  - noise position(s)  $\mathbf{p}_n$  and SNR (*not for diffuse noise*)
  - room properties (e.g.  $T_{60}$ , dimensions, reflection coefficients)

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  - ① Microphone subset selection
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  - ② Optimisation of microphone positions
    - For a *given* room, optimise positions of distributed microphones (e.g. using average performance)

## Objectives

- **Analyse performance of signal enhancement algorithms for different acoustical scenarios**
- Approaches:
  - ① Using **measurements** (RIRs, noise coherence)
    - Most accurate
    - Time-consuming if large number of source positions and microphones configurations need to be compared

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    - Numerical simulation

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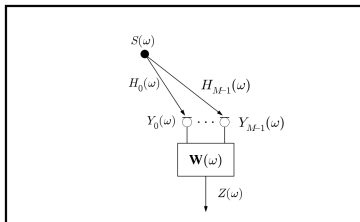
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**This presentation:** analytical expression for the spatially averaged output SNR of Multi-Channel Wiener Filter (MWF), *given* relative distance between source and microphones



## Signal model and configuration



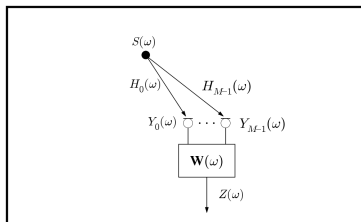
- Microphone signals in frequency-domain

$$\mathbf{Y}(\omega) = \mathbf{X}(\omega) + \mathbf{V}(\omega) = \mathbf{H}(\omega)S(\omega) + \mathbf{V}(\omega)$$

$$\mathbf{Y}(\omega) = [Y_0(\omega) \cdots Y_{M-1}(\omega)]^T, \quad \mathbf{H}(\omega) = [H_0(\omega) \cdots H_{M-1}(\omega)]^T$$

- Output signal:  $Z(\omega) = \mathbf{W}^H(\omega)\mathbf{Y}(\omega) = Z_x(\omega) + Z_v(\omega)$

## Signal model and configuration



- Desired source at position  $\mathbf{p}_s = [x_s \ y_s \ z_s]^T$
- Microphones at positions  $\mathbf{p}_m = [x_m \ y_m \ z_m]^T$ ,  $m = 0 \cdots M-1$
- Relative distance between source and microphones

$$\mathbf{D} = \begin{bmatrix} D_0 \\ \vdots \\ D_{M-1} \end{bmatrix} = \begin{bmatrix} \|\mathbf{p}_s - \mathbf{p}_0\| \\ \vdots \\ \|\mathbf{p}_s - \mathbf{p}_{M-1}\| \end{bmatrix}$$

## Multi-channel Wiener Filter

- **Goal:** MMSE estimate of speech component  $X_{m_0}$

$$\xi(\mathbf{W}) = \mathcal{E} \left\{ \left| X_{m_0} - \mathbf{W}^H \mathbf{Y} \right|^2 \right\}$$

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- **Solution:**

$$\mathbf{W}_{m_0} = \Phi_y^{-1} \Phi_x \mathbf{e}_{m_0} \quad \mathbf{e}_{m_0} = [0 \dots 1 \dots 0]^T$$

- Speech and noise correlation matrices:

$$\Phi_y = \Phi_x + \Phi_v$$

$$\Phi_x = \mathcal{E}\{\mathbf{X}\mathbf{X}^H\}, \quad \Phi_v = \mathcal{E}\{\mathbf{V}\mathbf{V}^H\}$$

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- **Output SNR:**

$$\text{SNR}_{\text{out}} = \frac{\mathcal{E}\{|Z_x|^2\}}{\mathcal{E}\{|Z_v|^2\}} = \frac{\mathbf{W}_{m_0}^H \Phi_x \mathbf{W}_{m_0}}{\mathbf{W}_{m_0}^H \Phi_v \mathbf{W}_{m_0}}$$

## Multi-channel Wiener Filter

- For a **single speech source**:  $\Phi_x = \phi_s \mathbf{H} \mathbf{H}^H$ ,  $\phi_s = \mathcal{E}\{|S|^2\}$

$$\mathbf{W}_{m_0} = \frac{\Phi_v^{-1} \mathbf{H}}{\phi_s^{-1} + \Lambda} \mathbf{H}_{m_0}^*, \quad \text{SNR}_{\text{out}} = \phi_s \Lambda$$

$$\Lambda = \mathbf{H}^H \Phi_v^{-1} \mathbf{H}$$

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- Homogeneous noise field**:  $\Phi_v = \phi_v \Gamma_v$

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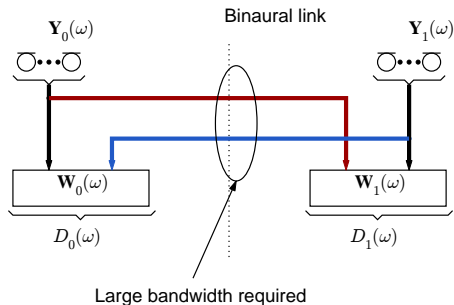
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$$\text{SNR}_{\text{out}} = \frac{\phi_s}{\phi_v} \rho \quad \rho = \mathbf{H}^H \Gamma_v^{-1} \mathbf{H}$$

SNR improvement only depends on Acoustical Transfer Function vector  $\mathbf{H}$  and noise coherence matrix  $\Gamma_v$

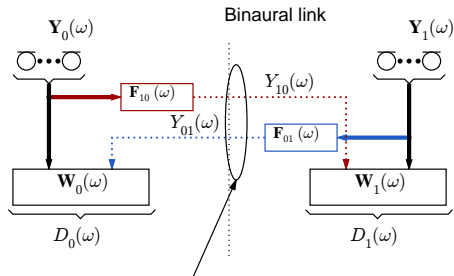


## Intermezzo: Distributed MWF



- All microphone signals are transmitted over a wireless link

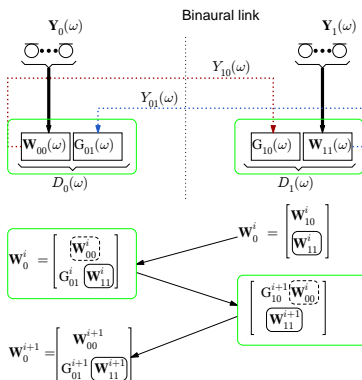
## Intermezzo: Distributed MWF



Required bandwidth can be reduced

- All microphone signals are transmitted over a wireless link
- Reduce bandwidth requirement of wireless link by transmitting one signal
  - Iterative distributed binaural MWF scheme (DB-MWF)

## Distributed MWF



- In each iteration  $F_{10}$  is equal to  $W_{00}$  from previous iteration, and  $F_{01}$  is equal to  $W_{11}$  from previous iteration
- **Converges to centralized MWF !**

## Rate constraints

- **Objective:** investigate influence of link bandwidth on performance of **binaural MWF algorithm**
- The signal  $\mathbf{Y}_{01}$  is encoded at finite bitrate  $R$  before transmission

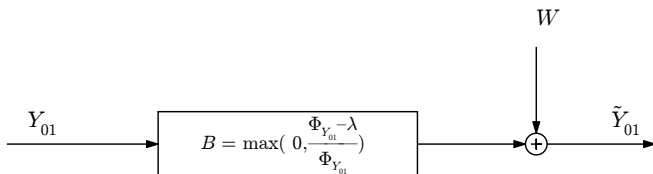
$$R(\lambda) = \frac{1}{4\pi} \int_{-\infty}^{\infty} \max(0, \log_2 \frac{\Phi_{\mathbf{Y}_{01}}}{\lambda}) d\omega$$

$$D(\lambda) = \frac{1}{2\pi} \int_{-\infty}^{\infty} \min(\lambda, \Phi_{\mathbf{Y}_{01}}) d\omega$$

- $\Phi_{\mathbf{Y}_{01}}$  is the PSD of  $\mathbf{Y}_{01}$
- Parameter  $\lambda$  links the transmission rate to the distortion

## Rate constraints

Upper bound on achievable performance can be calculated using forward channel representation



$$\Phi_W = \max\left(0, \lambda \frac{\Phi_{\mathbf{Y}_{01}} - \lambda}{\Phi_{\mathbf{Y}_{01}}}\right).$$

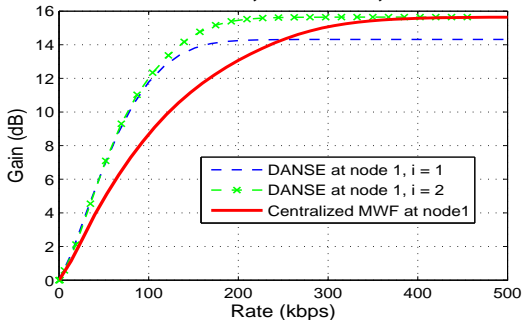
- The resulting MSE for SDW-MWF

$$\xi(R) = \frac{1}{2\pi} \int_{-\infty}^{\infty} \left( \Phi_{X_{0,1}} - \Phi_{\tilde{\mathbf{Y}}_0 X_{0,1}}^H \Phi_{\tilde{\mathbf{Y}}_0}^{-1} \Phi_{\tilde{\mathbf{Y}}_0 X_{0,1}} \right) d\omega$$

## Rate constraints

- Effect on performance of **DB-MWF-algorithm**
  - Single signal is compressed/transmitted in each iteration
  - Spread iterations over subsequent frames

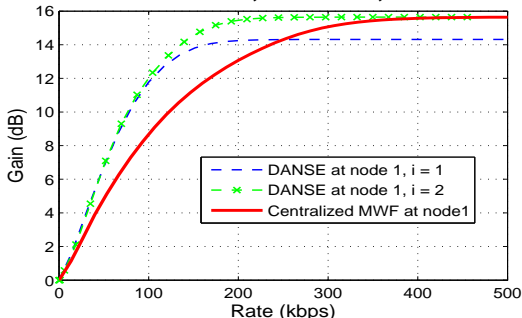
**SNR = 20 dB, SIR= 0 dB, rate R**



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- ⇒ DB-MWF-algorithm converges after  $K = 2$  iterations, moreover achieving **highest performance gain**

## Multi-channel Wiener Filter

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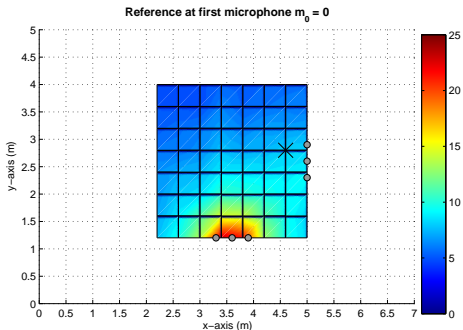
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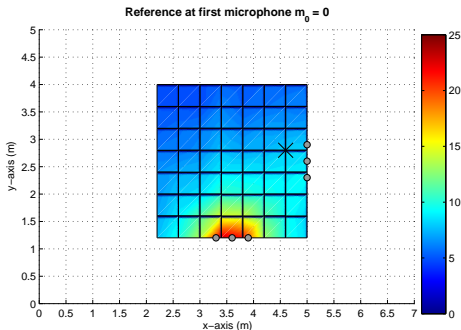
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**Objective:** (approximate) analytical expression for the output SNR of the MWF using **statistical room acoustics**

## Statistical properties of ATFs

- Decomposition of ATFs  $\mathbf{H}$  in direct and reverberant component

$$\mathbf{H}(\theta) = \mathbf{H}_d(\theta) + \mathbf{H}_r(\theta)$$

- Stochastic variable  $\theta = [\mathbf{p}_s, \mathbf{p}_0 \cdots \mathbf{p}_{M-1}]$

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<sup>1</sup>M. R. Schroeder, "Frequency correlation function of frequency responses in room", *Journal of the Acoustical Society of America*, vol. 34, no.12, pp. 1819-1823, Dec. 1962

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- Statistical properties of ATFs under specific conditions<sup>1</sup>
  - room dimensions large relative to wavelength of signals
  - frequencies above Schröder frequency  $f_g = 2000 \sqrt{T_{60}/V}$   
( $V$  = volume of room,  $T_{60}$  = reverberation time)
  - microphones and source at least half a wavelength away from walls
- spatial expectation  $\mathcal{E}_\theta\{\cdot\} =$  ensemble average over all realizations of  $\theta$

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## Statistical properties of ATFs

- Statistical properties of ATFs (fixed relative distance  $\mathbf{D}$ )

A1 Direct path component independent of realization of  $\theta$

$$\mathcal{E}_{\theta}\{H_{m,d}(\boldsymbol{\theta})|D_m\} = H_{m,d} = \frac{e^{-j\frac{\omega}{c}D_m}}{4\pi D_m} \quad \forall m$$

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A3 Spatial expectation of spectrum of reverberant component

$$\mathcal{E}_{\theta}\{|H_{m,r}(\boldsymbol{\theta})|^2\} = \frac{1-\bar{\alpha}}{\pi\bar{\alpha}A} \quad (\bar{\alpha} = \frac{0.161V}{AT_{60}}, A = \text{total surface area})$$

A4 Spatially expected correlation between reverberant components

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A5 Direct and reverberant components are uncorrelated

$$\mathcal{E}_{\theta}\{H_{m,d}(\boldsymbol{\theta})H_{n,r}^*(\boldsymbol{\theta})|D_m, D_n\} = 0, \quad \forall m, n$$

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A6 
$$\mathcal{E}_{\theta}\{|H_m(\boldsymbol{\theta})|^2 | D_m\} = \frac{1}{(4\pi D_m)^2} + \frac{1-\bar{\alpha}}{\pi\bar{\alpha}A}$$



## Output SNR using statistical properties ATFs

- For each realization of  $\theta$

$$\rho(\theta) = \mathbf{H}_d^H(\theta)\mathbf{\Gamma}_v^{-1}\mathbf{H}_d(\theta) + \mathbf{H}_d^H(\theta)\mathbf{\Gamma}_v^{-1}\mathbf{H}_r(\theta) + \mathbf{H}_r^H(\theta)\mathbf{\Gamma}_v^{-1}\mathbf{H}_d(\theta) + \mathbf{H}_r^H(\theta)\mathbf{\Gamma}_v^{-1}\mathbf{H}_r(\theta)$$

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$$\begin{aligned} \rho(\theta) = \sum_{m=1}^M \sum_{n=1}^M \check{\gamma}_{mn} & \left( H_{d,m}^*(\theta)H_{d,n}(\theta) + H_{d,m}^*(\theta)H_{r,n}(\theta) \right. \\ & \left. + H_{r,m}^*(\theta)H_{d,n}(\theta) + H_{r,m}^*(\theta)H_{r,n}(\theta) \right) \end{aligned}$$

with

$$\mathbf{H}_i^H \mathbf{\Gamma}_v^{-1} \mathbf{H}_j = \sum_{m=1}^M \sum_{n=1}^M \check{\gamma}_{mn} H_{i,m}^* H_{j,n}$$

- $\check{\gamma}_{mn}$  coefficients of the matrix  $\mathbf{\Gamma}_v^{-1}$

## Output SNR using statistical properties ATFs

- Spatial expectation of  $\text{SNR}_{\text{out}}$  given  $\mathbf{D}$

$$\begin{aligned} \mathcal{E}_{\theta}\{\rho(\theta)|\mathbf{D}\} &= \sum_{m=1}^M \sum_{n=1}^M \check{\gamma}_{mn} \left( \mathcal{E}_{\theta}\{H_{d,m}^*(\theta)H_{d,n}(\theta)|\mathbf{D}\} + \mathcal{E}_{\theta}\{H_{d,m}^*(\theta)H_{r,n}(\theta)|\mathbf{D}\} \right. \\ &\quad \left. + \mathcal{E}_{\theta}\{H_{r,m}^*(\theta)H_{d,n}(\theta)|\mathbf{D}\} + \mathcal{E}_{\theta}\{H_{r,m}^*(\theta)H_{r,n}(\theta)\} \right) \end{aligned}$$

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$$\mathcal{E}_\theta\{\text{SNR}_{\text{out}}(\theta)|\mathbf{D}\} = \frac{\phi_s}{\phi_v} \sum_{m=1}^M \sum_{n=1}^M \check{\gamma}_{mn} \left( \frac{e^{j\frac{\omega}{c}(D_n - D_m)}}{(4\pi)^2 D_m D_n} + \frac{1 - \bar{\alpha}}{\pi \bar{\alpha} A} \frac{\sin \frac{\omega}{c} \|\mathbf{p}_m - \mathbf{p}_n\|}{\frac{\omega}{c} \|\mathbf{p}_m - \mathbf{p}_n\|} \right)$$

- Depends on room properties  $A$  and  $\bar{\alpha}$ , relative distance  $\mathbf{D}$  and noise coherence  $\Gamma_v$

## Experimental results

**Objective:** Compare the *analytically* computed spatial expectation  $\mathcal{E}_{\theta}\{\text{SNR}_{\text{out}}(\theta)|\mathbf{D}\}$  with the *numerically* computed spatially averaged output SNR

$$\overline{\text{SNR}}_{\text{out}} = \frac{1}{N} \sum_{i=1}^N \text{SNR}_{\text{out}}(\tilde{\theta}_i)$$

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### Simulation and experimental setup

- Room dimensions  $7 \text{ m} \times 5 \text{ m} \times 3.5 \text{ m}$  and  $T_{60} = 250 \text{ ms}$
- $M = 3$  equally spaced microphones ( $d = 4 \text{ cm}$ )
- Source located at endfire of the microphone array such that  $\mathbf{D} = [1.36 \ 1.40 \ 1.44]^T$

## Experimental results

### Simulation and experimental setup

- Sampling frequency  $f_s = 16\text{kHz}$
- Room impulse responses simulated using image model  
( $L = 4096$ )<sup>2</sup>

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## Experimental results

### Simulation and experimental setup

- Sampling frequency  $f_s = 16\text{kHz}$
- Room impulse responses simulated using image model ( $L = 4096$ )<sup>2</sup>
- Diffuse noise coherence matrix  $\Gamma_v$  theoretically computed

$$\gamma_{mn}(\omega) = \frac{\sin \frac{\omega}{c} \|\mathbf{p}_m - \mathbf{p}_n\|}{\frac{\omega}{c} \|\mathbf{p}_m - \mathbf{p}_n\|}$$

- Frequency-flat a-priori input SNR  $\frac{\phi_s}{\phi_v}$

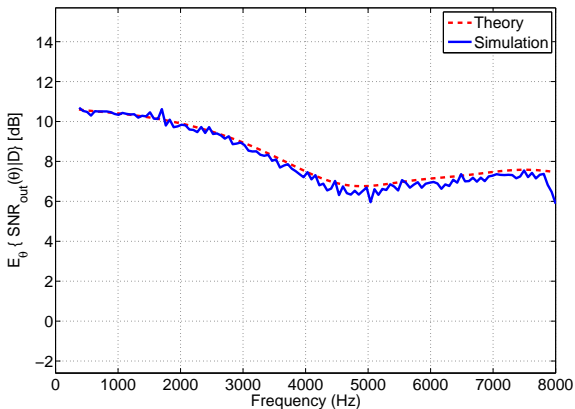
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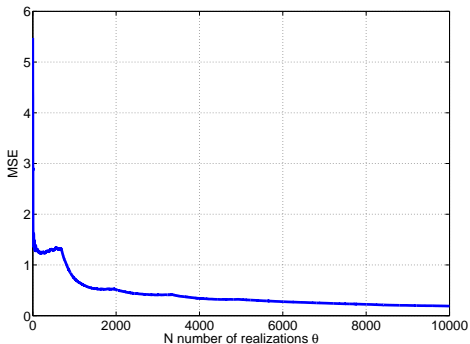
## Experimental results

- Monte Carlo simulation using 10000 realizations



## Experimental results

- $\text{MSE} = \sum_{\omega} |\mathcal{E}_{\theta}\{\text{SNR}_{\text{out}}(\theta)|\mathbf{D}\} - \overline{\text{SNR}_{\text{out}}}|^2$  as a function of number of realizations  $\theta$



The larger the number of realizations, the smaller the MSE between the analytical expression and simulations

## Conclusion and future work

- **Analytical expression** for spatial expectation of output SNR of MWF, depending on room properties, relative distance and noise coherence
- Simulation results show that analytical expression is **close to numerical simulations** (for large number of Monte Carlo realizations)

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- **Analytical expression** for spatial expectation of output SNR of MWF, depending on room properties, relative distance and noise coherence
- Simulation results show that analytical expression is **close to numerical simulations** (for large number of Monte Carlo realizations)
- **Future work**
  - For fixed microphone configuration, compute output SNR averaged over different source positions  $\Rightarrow$  **optimal subset selection**
  - Using average performance  $\Rightarrow$  **optimize positions** of distributed microphones