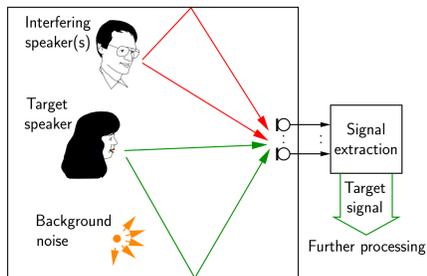


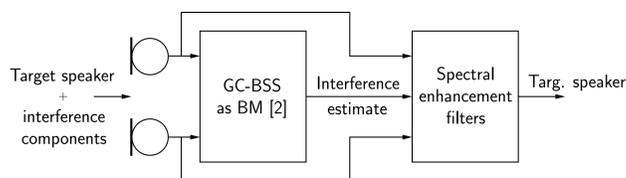
1. Introduction

Signal extraction scenario:



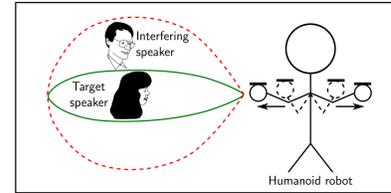
Goal: Extraction of desired source

Blind signal extraction (BSE) scheme [1]:



- Geometrically-constrained BSS (GC-BSS) acts as blocking matrix to estimate interference components
- Wiener-type spectral enhancement filters derived from interference estimate
- **Extraction performance based on quality of interference estimate**

Idea:



- Place microphones on movable limbs of a humanoid robot
- **Use movable microphones to adapt microphone array topology to improve target signal suppression of GC-BSS**
- Improved target signal suppression of GC-BSS leads to higher signal extraction performance of BSE [1]

2. Geometrically-constrained BSS

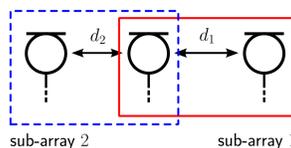
Cost function of GC-BSS [2]:

$$J_{\text{total}}(\mathbf{W}) = J_{\text{BSS}}(\mathbf{W}) + \gamma_C J_C(\mathbf{W}),$$

- \mathbf{W} : Matrix containing demixing filter weights
- $J_{\text{BSS}}(\mathbf{W})$: Generic cost function of TRINICON BSS [3]
- Minimization of $J_{\text{BSS}}(\mathbf{W})$ yields statistically independent output channels
- $J_C(\mathbf{W})$: Penalty term to force spatial null towards direction of target source in one output channel
- γ_C : Weight to control importance of penalty term, typically $0.4 < \gamma_C < 0.6$
- Direction of arrival ϕ_t of target source required for $J_C(\mathbf{W})$

3. Proposed Array Topology Adaptation Algorithm

Main concept:



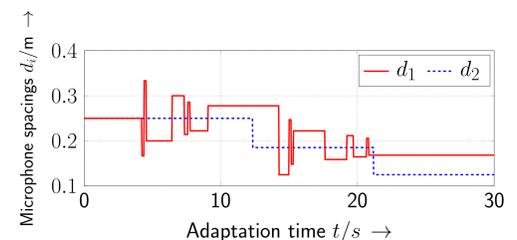
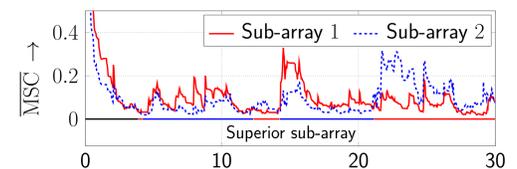
- Configuration of microphone array into two sub-arrays
- **Use inferior sub-array to create new competitor for superior sub-array**
- Adaptation of microphone spacing d_{inf} of inferior sub-array in dependence of microphone spacing d_{sup} of superior sub-array

$$d_{\text{inf}} = a_n d_{\text{sup}}, \quad n \in \{1, 2, 3, \dots\},$$

with

$$a_n = \left(\frac{1+n}{2+n} \right)^{(-1)^{n+1}} = \left\{ \frac{2}{3}, \frac{4}{3}, \frac{4}{5}, \frac{6}{5}, \frac{8}{7}, \dots \right\}$$

Example of array topology adaptation:



- $\phi_t = 0^\circ$, $\phi_{\text{int}} = 40^\circ$, $T_{60} = 100$ ms, Source distance: 1.0 m

4. Experiments

Evaluation measure:

- Weighted Magnitude Squared Coherence (MSC)

$$\overline{\text{MSC}} = \frac{1}{\sum_{\nu=0}^{\nu_{\text{max}}} W(\nu)} \sum_{\nu=0}^{\nu_{\text{max}}} W(\nu) \frac{|S_{y_1 y_2}(\nu)|^2}{S_{y_1 y_1}(\nu) \cdot S_{y_2 y_2}(\nu)},$$

with

$W(\nu)$: Window function

$S_{y_1 y_1}$, $S_{y_1 y_2}$, $S_{y_2 y_2}$: Auto- and cross-power spectrum densities of output channels

- Segmental target suppression gain

$$\Delta \overline{\text{TS}}_{\text{seg}} = \overline{\text{TS}}_{\text{seg}, y_2} - \overline{\text{TS}}_{\text{seg}, x} \text{ dB},$$

with

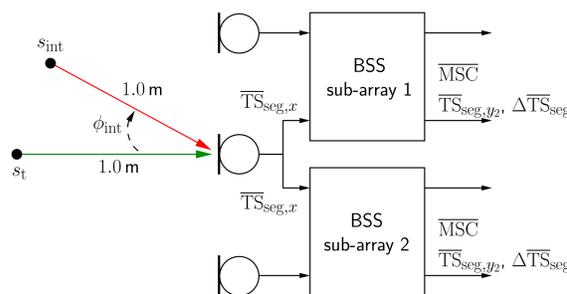
$$\overline{\text{TS}}_{\text{seg}, y_2} = \frac{1}{K_S} \sum_{m=1}^{K_S} \left(10 \log_{10} \left(\frac{\sum_{\kappa=1}^{N_S} y_{2, \text{int}}^2(\kappa + m N_S)}{\sum_{\kappa=1}^{N_S} y_{2, t}^2(\kappa + m N_S)} \right) \right) \text{ dB}$$

$\overline{\text{TS}}_{\text{seg}, x_2}$: Analogously to $\overline{\text{TS}}_{\text{seg}, y_2}$

$y_{2, t}$, $y_{2, \text{int}}$: Target and interference components in second output channel

$N_S = 256$, K_S : Block length and number of blocks used to compute $\overline{\text{TS}}_{\text{seg}, y}$ and $\overline{\text{TS}}_{\text{seg}, y_2}$

Experimental Setup:



- $T_{60} = 100$ ms, Source distance: 1.0 m
- $\phi_t = 0^\circ$, $\phi_{\text{int}} \in \{20^\circ, 40^\circ, 60^\circ, 80^\circ\}$
- $f_s = 16$ kHz, signal length: 60 s, Topology adaptation phase: 30 s
- Initial microphone spacings $d_i = 0.25$ m
- Room impulse responses simulated using image method by Allen and Berkley [4]
- BSS: filter length $L = 512$, block length $N = 2L$

Results:

Adaptation	ϕ_{int}	20°	40°	20°	80°	Avg.
No	$\Delta \overline{\text{TS}}_{\text{seg}}$ [dB]	8.01	8.76	9.96	10.07	9.20
Yes	$\Delta \overline{\text{TS}}_{\text{seg}}$ [dB]	8.49	11.56	10.94	12.02	10.75

5. Conclusions

- Proposed array topology adaptation algorithm leads to improved separation, and thus, higher target source suppression performance
- By using array topology adaptation for humanoid robots, BSE algorithm [1] is expected to yield better signal extraction performance

References:

- [1] K. Reindl, Y. Zheng, A. Schwarz, S. Meier, R. Maas, A. Sehr, and W. Kellermann, "A Stereophonic acoustic signal extraction scheme for noisy and reverberant environments," Computer Speech and Language (CSL), vol. 27, no. 3, pp. 726–745, May 2012.
- [2] Y. Zheng, K. Reindl, and W. Kellermann, "BSS for improved interference estimation for blind speech signal extraction with two microphones," Int. Workshop on Comp. Advances in Multi-Sensor Adapt. (CAMSAP), Aruba, Dutch Antilles, pp. 253–256, Dec. 2009.
- [3] H. Buchner, R. Aichner, and W. Kellermann, "TRINICON: A versatile framework for multichannel blind signal processing," IEEE Int. Conference on Acoustics, Speech and Signal Processing (ICASSP), vol. 3, pp. iii – 889–92, May 2004.
- [4] J.B. Allen and D.A. Berkley, "Image method for efficiently simulating small-room acoustics," Journal Acoustic Society of America (JASA), vol. 65, no. 4, pp. 943–1086, April 1979.