

Real-time Multichannel Speech Separation and Enhancement using a Beamspace-domain-based Lightweight CNN

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Noisy and reverberant environment





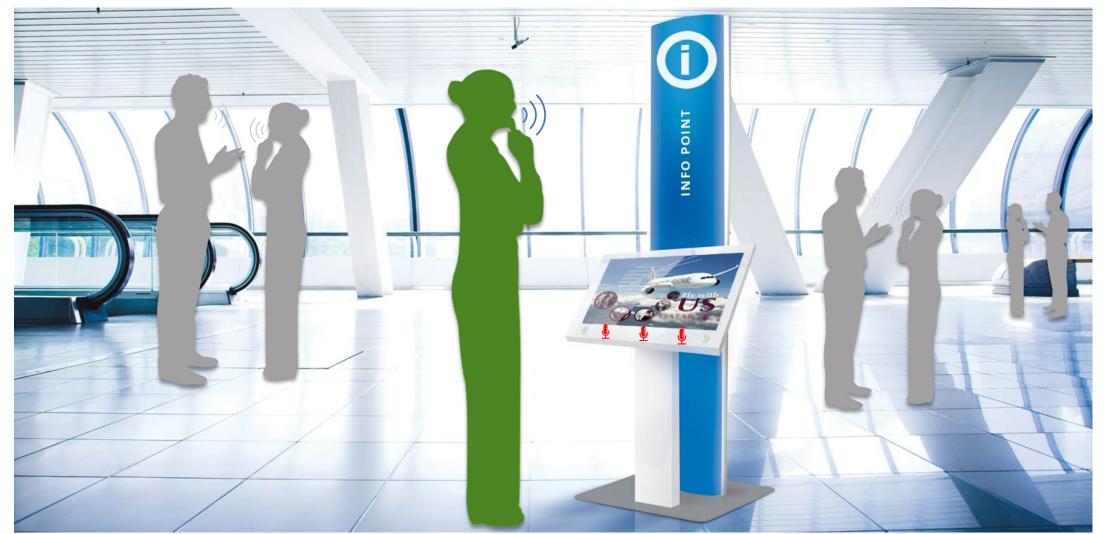






Extract the talker in front of the array

case 1











Discard all the interferers

case 2

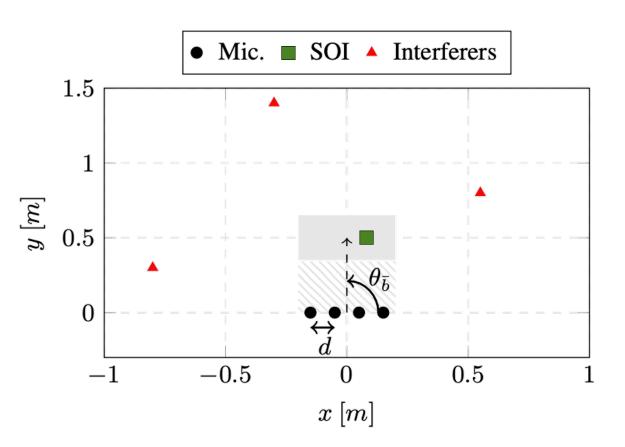






SETUP and GOALS

- Uniform Linear Array (**ULA**)
- ✤ 1 Signal Of Interest (SOI)
 - within a region in front of the array
 - DOA $\approx \theta_{\bar{b}} = 90^{\circ}$
- \checkmark $R \in \{0, \dots, 4\}$ interferers
 - in the noisy and reverberant room



OBJECTIVES

- □ **Real-time model** for the SOI separation and enhancement
- □ Evaluation on **real recordings** whereas training on simulated data
- □ Robust system with respect to multiple array geometries and acoustic conditions



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SIGNAL MODEL and BACKGROUND

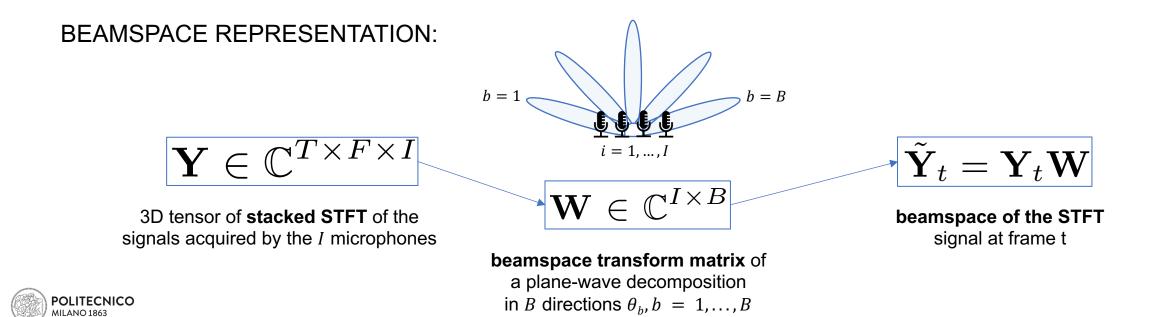
I microphones with *d* inter-sensor spacing

J speakers

- **γ diffuse noise** component
- *v* additive noise component

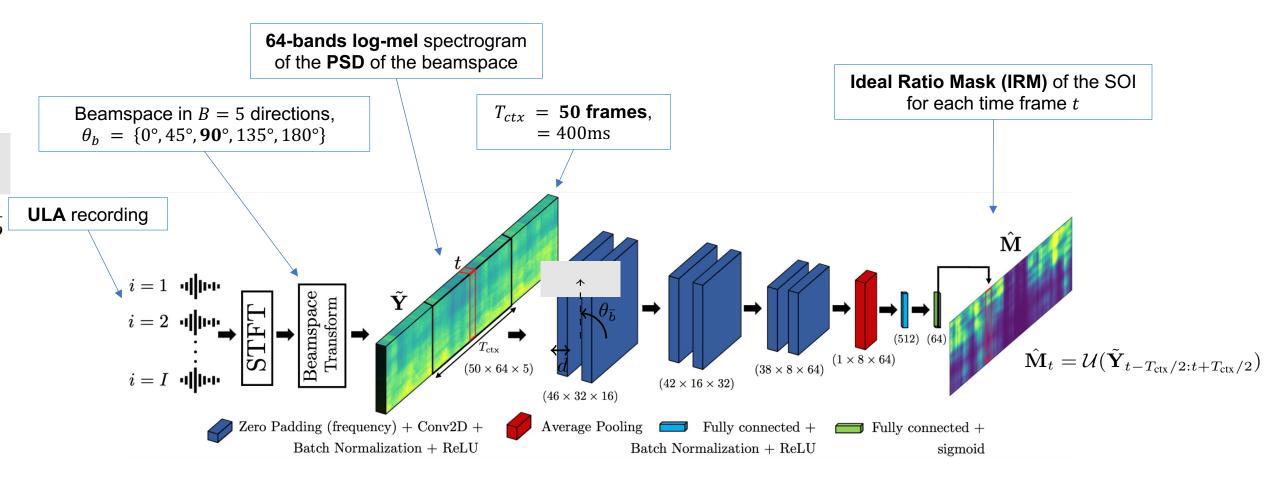
$$y_{i}[t,f] = \sum_{j=1}^{J} h_{j,i}[t,f]s_{j}[t,f] + \gamma_{i}[t,f] + v_{i}[t,f]$$
$$= \sum_{j=1}^{J} x_{j,i}[t,f] + \gamma_{i}[t,f] + v_{i}[t,f],$$

STFT representation of the signal acquired by the *i*th microphone





PROPOSED METHOD



$$\mathcal{L}(t) = \frac{1}{F} \sum_{f=1}^{F} (\mathbf{M}_{t,f} - \hat{\mathbf{M}}_{t,f})^2$$

$$\hat{\mathbf{X}}_{ar{j},t} = \hat{\mathbf{M}}_t \odot ilde{\mathbf{Y}}_{ar{b},t}$$

Final estimate of the **desired signal** \bar{j} at frame t



Loss function

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DATASET GENERATION

Extensive **simulation campaign** by sampling with a uniform distribution the operational ranges

RIRs computed with **gpuRIR** [1]

ULA setup	I = 3/4, d = 20/30mm
Room dimensions	$L_x \in [3,8]m$; $L_y \in [3,8]m$; $L_z \in [2.6,4]m$
Т60	[0.2, 1.4] s
SOI presence	80/20 % of rooms with / without SOI
R number of interferers	from 0 to 4
SIR (loudness simulation)	[-3,3] dB
SDR (babble noise)	[-3, 60] dB
SNR (microphone noise)	[30, 70] dB
Array Gain (signal dynamic)	[-40, -1] dB
LibriSpeech dataset	5 sec signals
Total training rooms	250,000



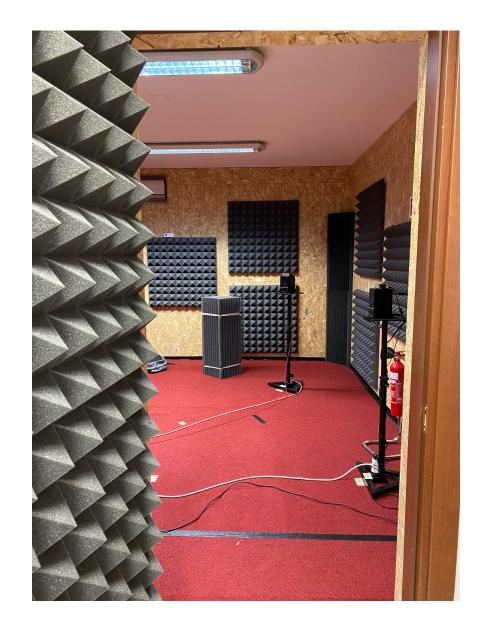
[1] D. Diaz-Guerra, et al. "gpuRIR: A python library for room impulse response simulation with gpu acceleration," Multimedia Tools and Applications, 2021.



operational ranges

EVALUATION

- Evaluation on real recordings in ETSI room with different ULA unseen during training:
 - Varying number of sensors I
 - Varying inter-sensor distance *d*
 - \rightarrow I = 3/4, d = 20/26mm
 - $\rightarrow I = 5, d = 52mm$
 - Comparison wrt
 - NBDF method [2]
 - $\widetilde{Y}_{90^{\circ}}$ input beamformer steering to the SOI region



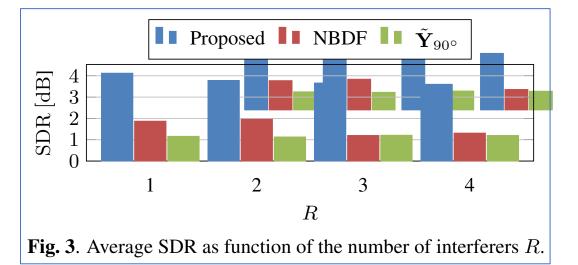




RESULTS	ULA setups Metrics	I = 4, d = 26 mm Proposed NBDF $\tilde{\mathbf{Y}}_{90^{\circ}}$			$I = 3, d = 52 \mathrm{mm}$ Proposed NBDF $\tilde{\mathbf{Y}}_{90^{\circ}}$			$I = 4, d = 52 \mathrm{mm}$ Proposed NBDF $\tilde{\mathbf{Y}}_{90^{\circ}}$			Average over test sets Proposed NBDF $\tilde{\mathbf{Y}}_{90^{\circ}}$		
	SIR	9.46	8.5	1.62	8.5	10.48	0.93	6.47	10.84	0.97	8.31	10.05	1.18
	SAR	7.73	2.99	-	9.34	6.05	-	7.58	3.08	-	8.29	4.29	-
国政的部分委员 (11)	SDR	4.15	0.04	1.6	4.63	3.29	0.92	2.28	0.75	0.96	3.79	1.59	1.17
	PESQ	1.66	1.19	1.71	1.86	1.39	1.79	1.66	1.24	1.79	1.73	1.27	1.76
	ESTOI	0.57	0.44	0.58	0.61	0.52	0.61	0.6	0.44	0.62	0.59	0.46	0.6
	$\mathrm{R}_{\mathrm{soi}}$	-5.75	-4.7	-	-2.61	-1.77	-	-6.81	-3.81	-	-4.67	-3.25	-
	$\mathrm{R}_{\mathrm{interf}}$	-17.54	-13.47	-	-14.31	-14.48	-	-15.7	-15.2	-	-15.65	-14.32	-

Table 1. Comparison of the average metrics between the proposed method, the NBDF approach and the beamformer $\tilde{\mathbf{Y}}_{90^\circ}$ for the different test sets and for the average results.

COMPARISON	NBDF	Proposed solution				
# parameters	4,006,236	120,000				
MACS/frame	198.5 millions	1.06 millions				



Perceptual intelligibility of the devised solution outperforms both NBDF and $\tilde{Y}_{90^{\circ}}$

https://polimi-ispl.github.io/beamspace_cnn_speech_separation.github.io/



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CONCLUSION

SOI speech extraction and enhancement in noisy and reverberant environments

Lightweight CNN architecture for **real-time computation**

Robust system wrt **setup generalization**:

- number of speakers and microphones
- inter-sensor spacing
- **reverberation** time of the room
- **noise** components of the array and the environment

Generalization wrt different array geometries





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