

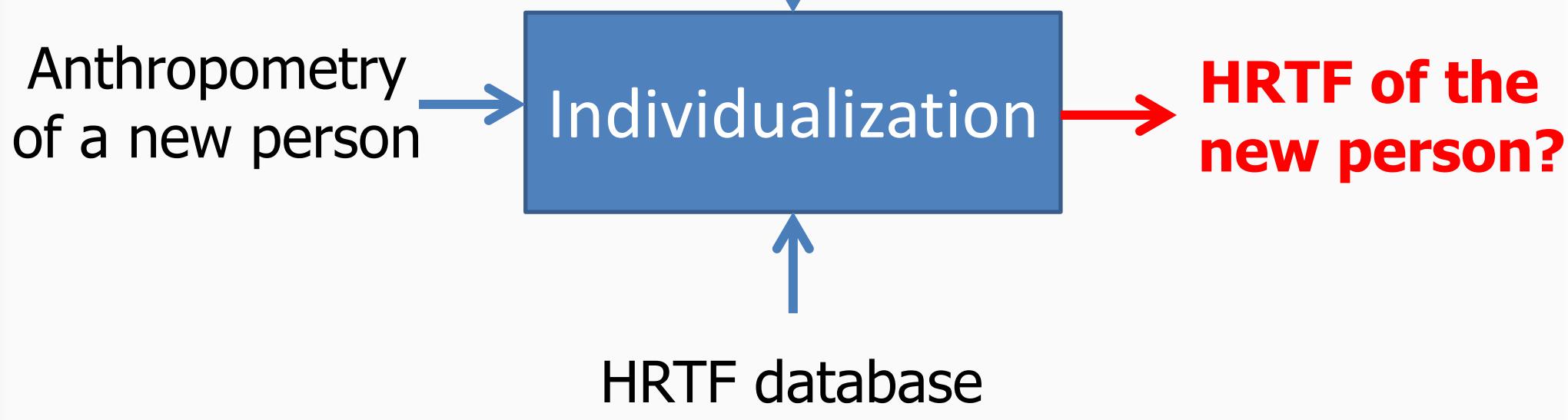
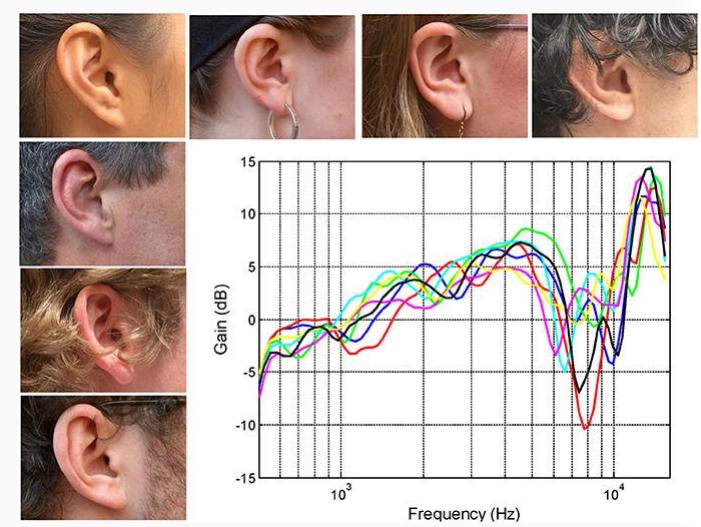
On the Preprocessing and Postprocessing of HRTF Individualization Based on Sparse Representation of Anthropometric Features

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Motivation

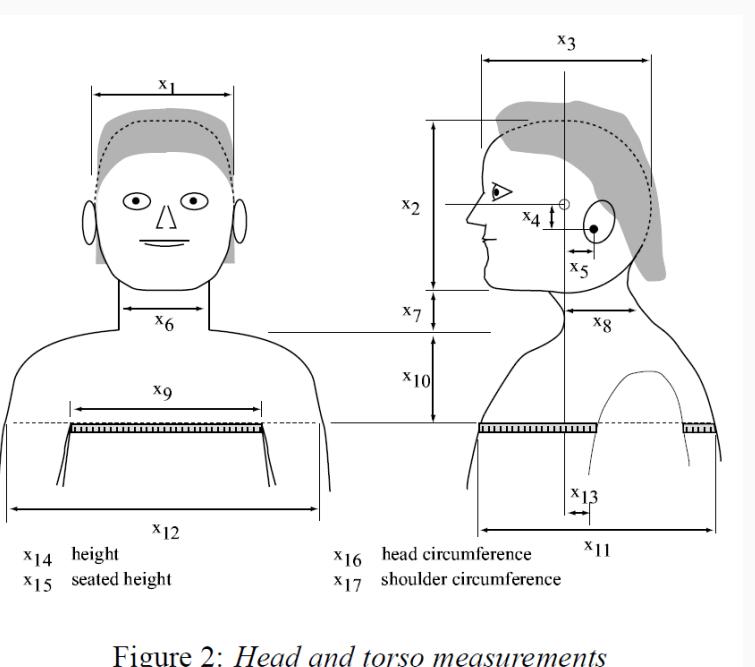
- Head-related transfer functions (HRTFs) contain sound localization cues and are commonly used in 3D audio reproduction;
- HRTFs are highly individualized [1-2];
- HRTFs are closely related to anthropometry (torso, head, pinna);
- Anthropometry can be used for HRTF individualization.



In this paper, we aim to answer:

- Whether the preprocessing and postprocessing methods affect the performance of HRTF individualization?
- If so, what is the best preprocessing and postprocessing techniques?
- And, how good is it?

CIPIC Anthropometric data



Var	Measurement	μ	σ	%
x_1	head width	14.49	0.95	13
x_2	head height	21.46	1.24	12
x_3	head depth	19.96	1.29	13
x_4	pinna offset down	3.03	0.66	43
x_5	pinna offset back	0.46	0.59	254
x_6	neck width	11.68	1.11	19
x_7	neck height	6.26	1.69	54
x_8	neck depth	10.52	1.22	23
x_9	torso top width	31.50	3.19	20
x_{10}	torso top height	13.42	1.85	28
x_{11}	torso top depth	23.84	2.95	25
x_{12}	shoulder width	45.90	3.78	16
x_{13}	head offset forward	3.03	2.29	151
x_{14}	height	172.43	11.61	13
x_{15}	seated height	88.83	5.53	12
x_{16}	head circumference	57.33	2.47	9
x_{17}	shoulder circumference	109.43	10.30	19
d_1	cavum concha height	1.91	0.18	
d_2	cavum concha height	0.68	0.12	35
d_3	cavum concha width	1.58	0.28	35
d_4	fossa height	1.51	0.33	44
d_5	pinna height	6.41	0.51	16
d_6	pinna width	2.92	0.27	18
d_7	intertragal incisure width	0.53	0.14	51
d_8	cavum concha depth	1.02	0.16	32
θ_1	pinna rotation angle	24.01	6.59	55
θ_2	pinna flare angle	28.53	6.70	47

35 subjects * [37 features & 1250 HRTFs]

Figures from [3]

HRTF Individualization based on Sparse Representations of Anthropometric features

Anthropometry database $A : S$ subjects * 1 set of Anthropometry
Anthropometry of a new person A_1 : 1 subjects * 1 set of Anthropometry

HRTF database $H : S$ subjects * 1 set of HRTF

HRTF of a new person H_1 : 1 subjects * 1 set of HRTF

1 set of Anthropometry : F features

1 set of HRTF: D directions * K points

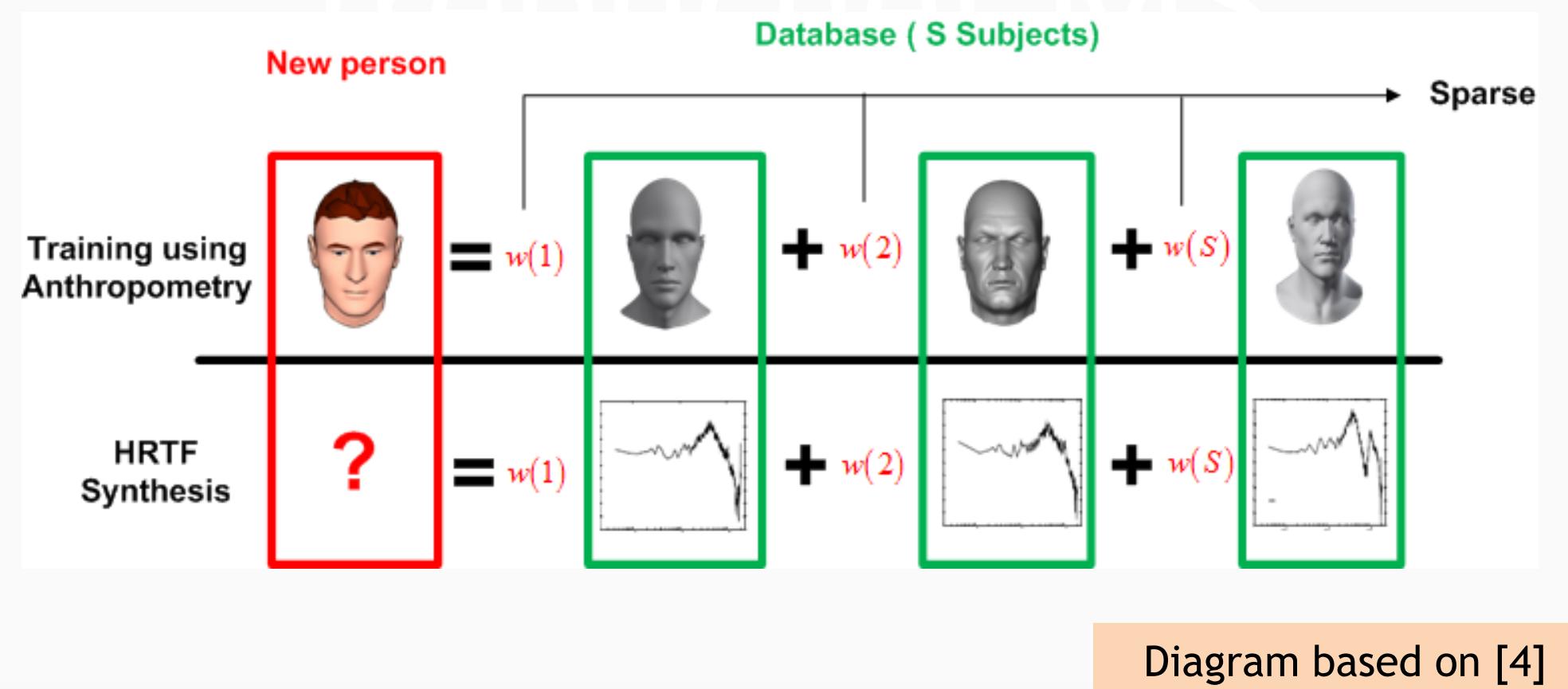


Diagram based on [4]

Preprocessing and Postprocessing in HRTF Individualization

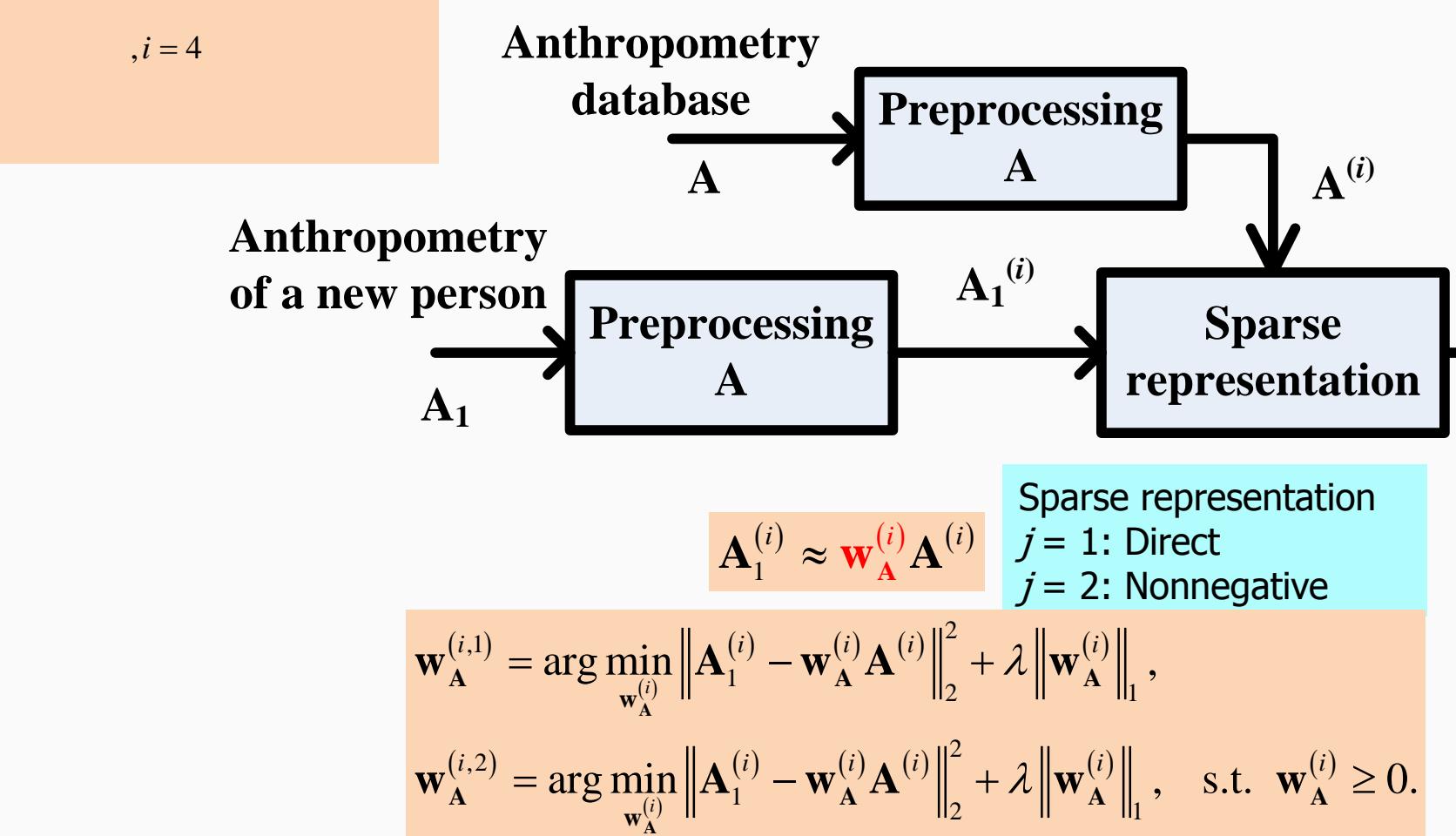
$$\mathbf{A}^{(i)}(f) = \begin{cases} \mathbf{A}(f) - \min[\mathbf{A}(f)] & , i=1 \\ \frac{\max[\mathbf{A}_0(f)] - \min[\mathbf{A}_0(f)]}{\text{std}[\mathbf{A}_0(f)]} & , i=2 \\ \frac{\mathbf{A}(f) - \text{mean}[\mathbf{A}(f)]}{\text{std}[\mathbf{A}_0(f)]} & , i=3 \\ \frac{\mathbf{A}(f)}{\text{std}[\mathbf{A}_0(f)]} & , i=4 \end{cases} \quad \forall f = 1, 2, \dots, F,$$

where $\mathbf{A}_0 = [\mathbf{A} - \mathbf{A}_1]$

Preprocessing of Anthropometry
 $i = 1$: Direct
 $i = 2$: Min-max normalization
 $i = 3$: Standard score
 $i = 4$: Standard deviation normalization

Preprocessing of HRTF
 $m = 1$: Magnitude
 $m = 2$: Log magnitude
 $m = 3$: Power

$$\mathbf{H}^{(m)}(d, k) = \begin{cases} \mathbf{H}(d, k) & , m=1 \\ 20 \log_{10} [\mathbf{H}(d, k)] & , m=2 \\ [\mathbf{H}(d, k)]^2 & , m=3 \end{cases} \quad \forall d = 1, 2, \dots, D; k = 1, 2, \dots, K$$



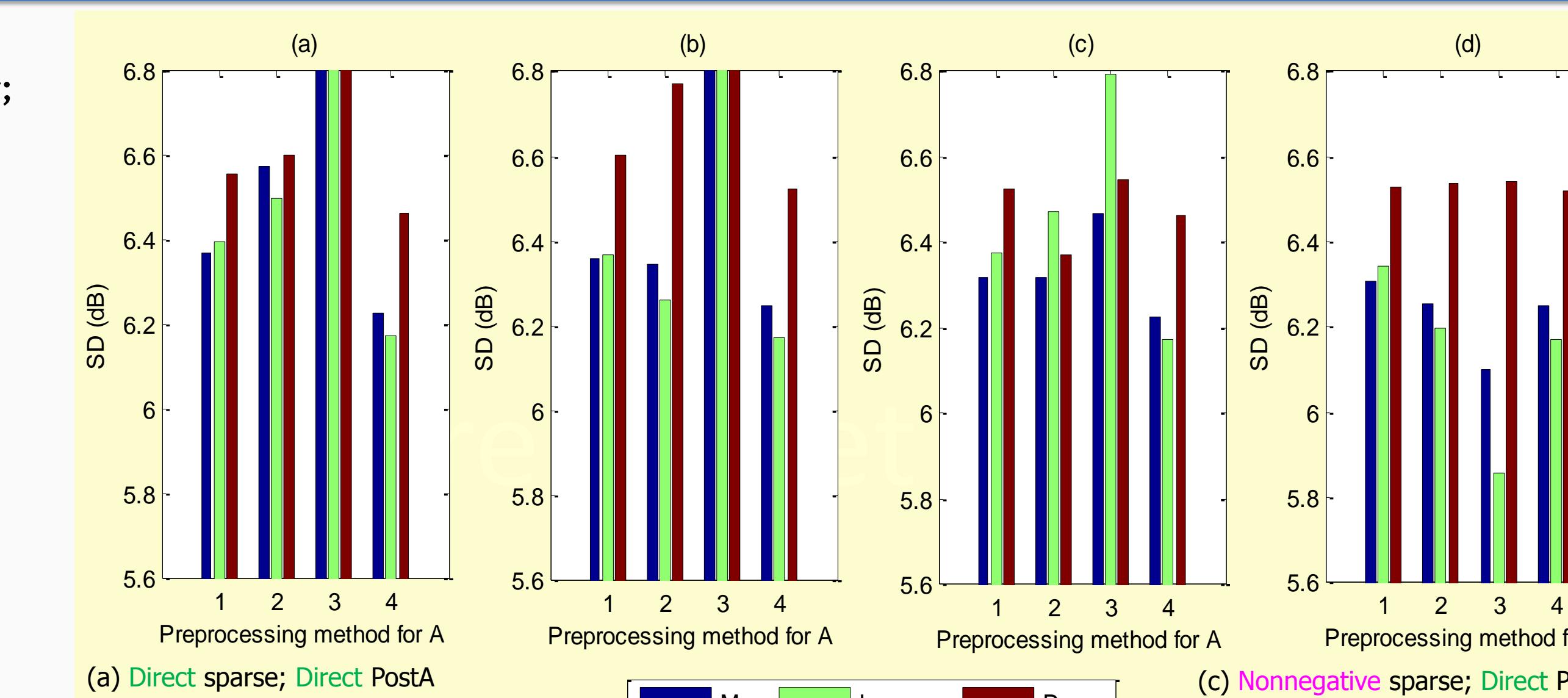
In total, we have $4 \times 3 \times 2 \times 2 = 48$ variants of methods!

Experimental Results

- CIPIC HRTF database;
- Cross validation technique to selection the regularization parameter;
- $S_{\text{test}} = 35$ test cases, all 1250 directions, and full frequency range.

$$\text{Spectral distortion SD}^{(i,j,l,m,n)} = \sqrt{\frac{1}{S_{\text{test}}} \sum_{s=1}^S \sum_{d=1}^D \sum_{k=1}^K \left[20 \log_{10} \frac{\hat{\mathbf{H}}_s^{(i,j,l,m,n)}(d, k)}{\mathbf{H}_s(d, k)} \right]^2} \quad [\text{dB}]$$

Sparse representation	PostA	PreH	PreA			
			Direct	Min-max	Standard score	Standard deviation
Direct	Mag	Mag	6.37	6.57	81.00	6.23
		Log mag	6.40	6.50	21.61	6.17
		Power	6.56	6.60	78.94	6.46
	Power	Mag	6.36	6.35	15.97	6.25
Nonnegative	Mag	Log mag	6.37	6.26	8.89	6.17
		Power	6.60	6.77	25.21	6.52
		Mag	6.32	6.32	6.47	6.23
	Log mag	6.38	6.47	6.79	6.17	
Nonnegative	Power	Mag	6.52	6.37	6.55	6.46
		Log mag	6.31	6.26	6.10	6.25
		Power	6.53	6.54	6.54	6.52



Direct sparse representation

- PreA: standard deviation best, standard score worst;
- PreH: log mag best, power worst;
- PostA: minimal effect for good PreA, PreH.

Nonnegative sparse representation

- Better than corresponding direct sparse representation (esp. standard score);
- Trend in PreA/PreH not obvious;
- Normalized PostA can improve the performance (esp. standard score).

CONCLUSIONS

- Introduced preprocessing and postprocessing in HRTF individualization based on sparse representation of anthropometric features.
- Investigated 48 variants of preprocessing and postprocessing methods, and found
 - Preprocessing and postprocessing methods do affect the performance of HRTF individualization, though the effects differ in different combinations;
 - Adding nonnegative constraints in sparse representation improves the performance;
 - The best combination for HRTF individualization is < standard score + log magnitude + nonnegative + normalized >.
- Established the lower bound for this type of HRTF individualization and verified that “our best” combination outperforms existing approaches and is quite close to the lower bound.
- Future work: subjective evaluation of HRTF individualization.

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[2] S. Carlile (2014) The plastic ear and perceptual relearning in auditory spatial perception. Front. Neurosci. 8:237. doi: 10.3389/fnins.2014.00237

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[5] S. J. Kim, K. Koh, M. Lusig, S. Boyd, and D. Gorinevsky, “An interior-point method for large-scale l1-regularized least squares,” J. Selected topics in signal processing, vol. 1, no. 4, pp. 606-617, Dec. 2007.