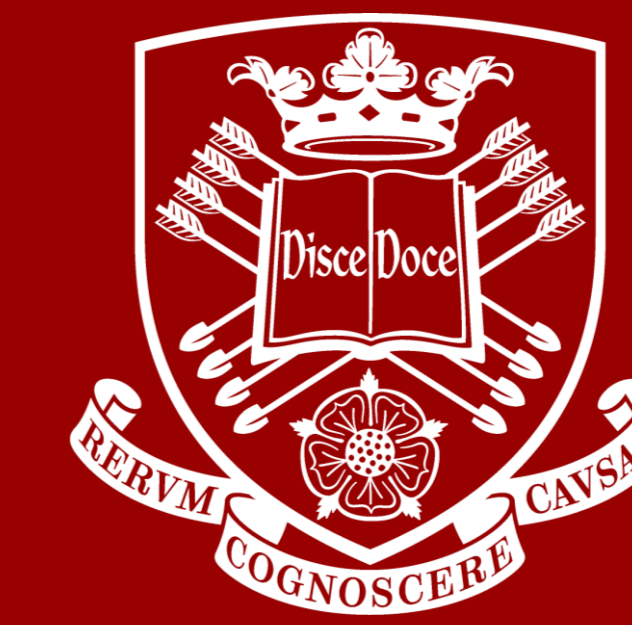


# Exploiting non-negative matrix factorization for binaural sound source localization in the presence of directional interference



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An existing binaural localization system based on a deep neural network (DNN) performs well in reverberant conditions, but performance deteriorates in conditions with directional interference. To overcome this limitation, the system is extended with a separation stage based on nonnegative matrix factorization (NMF). Different approaches to validating and training the system are explored. Extending the system in this way greatly improves performance in conditions with directional interference.

**Task to solve:**

- ◆ Predict azimuth angle  $\phi$  of a speech source
- ◆ One non-speech masking source is present

**Problem with existing system:**

- ◆ Masker provides competing ITD and ILD cues

**Solution:**

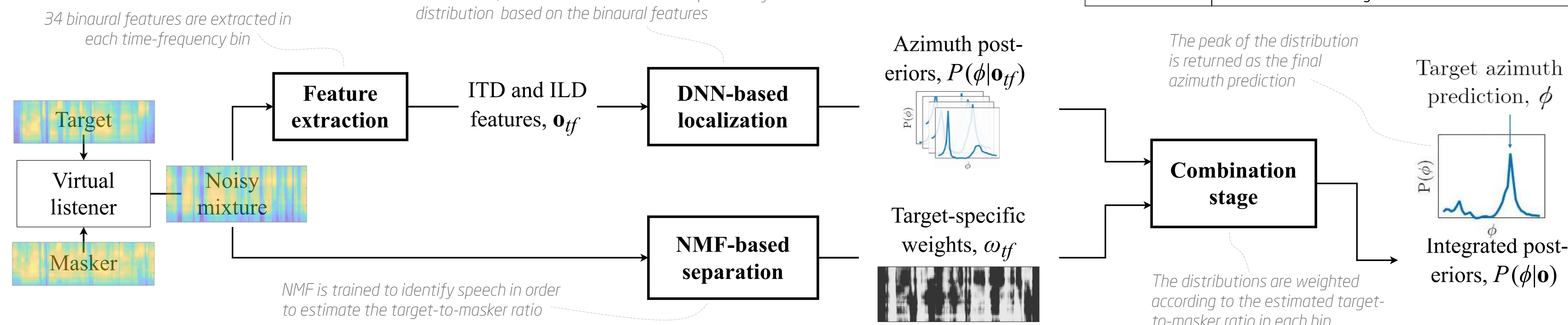
- ◆ Add separation stage to isolate speech cues

**The study of sound localization** has been an active research field for over a hundred years. Today, some relevant applications include<sup>1</sup>:

- Hearing aids, where beamforming is used for spatial filtering.
- Social robots that need to be able to communicate with humans.
- Auditory scene analysis, where spatial positioning of sources is important.

Our study extends an existing machine listening system with a separation stage. This extension allows it to cope with directional interference.

For each bin, the DNN returns an azimuth probability distribution based on the binaural features



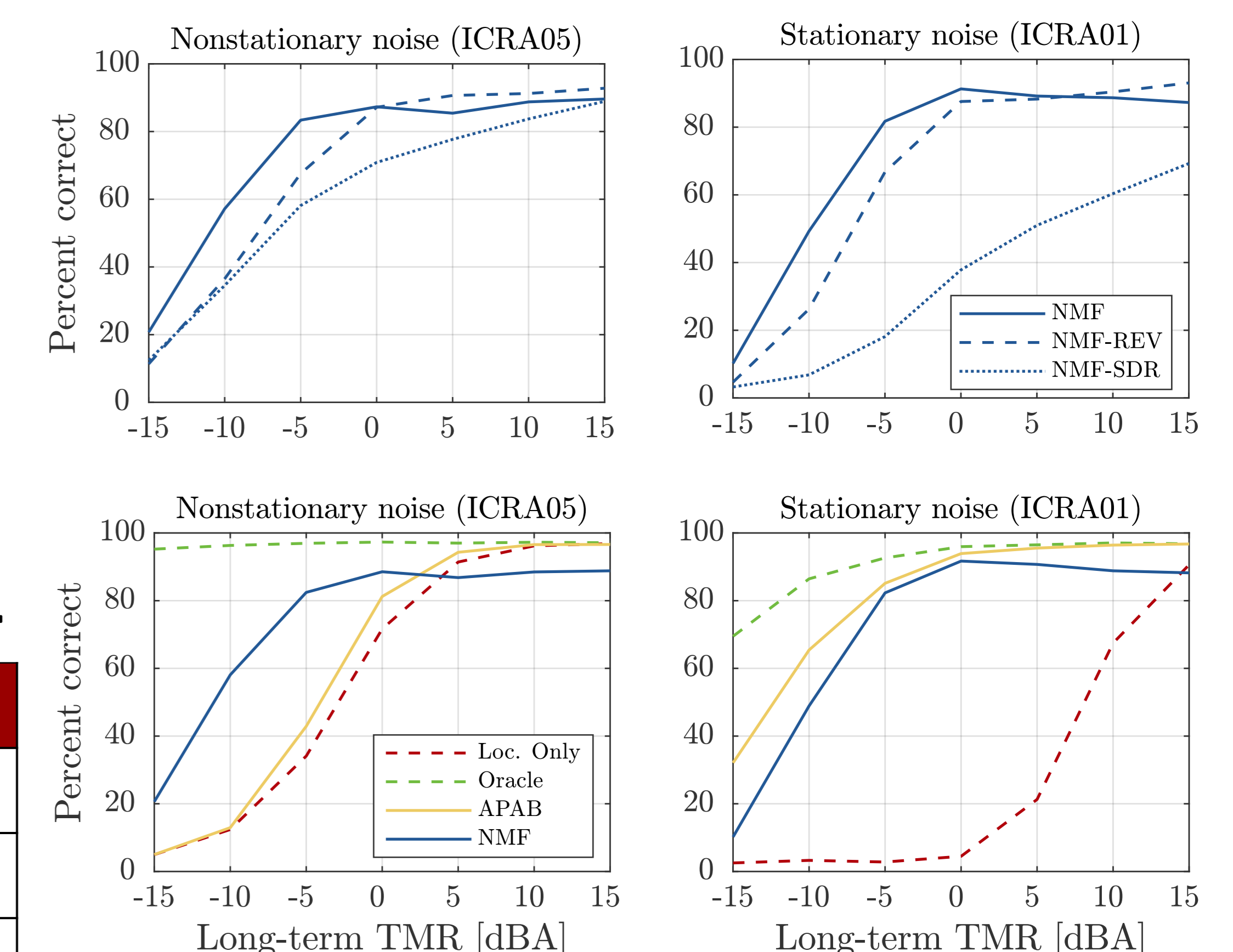
**Three main research questions** are addressed in our study:

1. Should the NMF be validated on a separation or a localization metric?
2. Should reverberant speech be used as training material for the NMF?
3. Is the proposed separation stage better than a learning-free approach?

To address these questions, three NMF variants are tested. The best of these is then compared to three baseline algorithms.

Algorithm	Validation metric	Training
NMF	Localization (PC)	Anechoic
NMF-SDR	Separation (SDR) <sup>2</sup>	Anechoic
NMF-REV	Localization (PC)	Reverberant
Baselines		Description
Loc. Only	No separation <sup>3</sup>	
Oracle	Ideal separation	
APAB	Adaptive post-filter driven by two single-channel noise reduction systems <sup>4,5</sup>	

**The results of our experiments** are based on simulations in five different rooms using both stationary and nonstationary noise types.



**To conclude**, our study showed that using an NMF-based separation stage is a useful way to deal with directional interference in a state-of-the-art localization system. It was also shown that:

1. It is of great importance that the NMF-based separation stage is validated using a localization metric.
2. Training the NMF on reverberant speech only aids localization in high target-to-masker ratio conditions.
3. NMF is better than the learning-free approach in nonstationary noise. NMF also performs well in stationary noise, but not as well as the baseline.

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<sup>2</sup>E. Vincent, R. Gribonval, and C. Févotte: "Performance measurements in blind audio source separation," *IEEE Transactions on Audio, Speech and Language Processing*, vol. 14, pp. 1462-1469, 2006

<sup>3</sup>N. Ma, J. A. Gonzales, and G. J. Brown: "Robust binaural localization of a target sound source by combining spectral source models and deep neural networks," *IEEE Workshop on Applications of Signal Processing to Audio and Acoustics*, 2016

<sup>4</sup>Y. Ephraim, and D. Malah: "Speech enhancement using a minimum-mean square error short-time spectral amplitude estimator," *IEEE Transactions on Acoustics, Speech, and Signal Processing*, 1984, vol. 32, pp. 1109-1121

<sup>5</sup>C. Ris, and S. Dupont: "Assessing local noise level estimation methods: Application to noise robust ASR," *Speech Communication*, 2001, vol. 34(1-2), pp. 141-158