



# Auditory Reverse Correlation on a Phoneme-Discrimination Task: Assessing the Effect of Different Types of Background Noise

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## 1. Introduction

■ **Reverse correlation** is a behavioral method that can be used to explore which properties of a sound are used as perceptually-relevant cues. In our case, the method assesses which acoustic cues of the **phonemes** /b/ and /d/ relate to correct or incorrect **listener responses** in a phoneme-in-noise task. Based on the tested noises, a fine-grained time-frequency map of acoustic cues can be obtained: The **auditory classification image (ACI)** (Varnet et al. 2013, 2015). We used three noise types and compared the robustness of the ACIs with respect to **noise statistics**.

■ In this poster, we focus on the effect of the **statistics** of the background **noises** with respect to the efficiency and robustness of the method. We used three different noise types that have a flat long-term spectrum, but differ in the amount of temporal envelope fluctuations: **White noise (WN)**, white noise low-pass filtered in the modulation power spectrum (**MPS**), and bump noise (**BP**).

## 2. Methods: Experiments

■ **Stimuli**: two natural male speech productions of /aba/ and /ada/ were taken from the OLLO database (Meyer et al. 2010, speaker S43M). The sounds were pre-processed to have equal duration and same acoustic energy in the first syllable. The speech level was adjusted according to the trial SNR, using one of three noise types.

■ **Background noises**: 4000 realizations of each noise –WN, MPS, and BP– were used. The noises were presented at 65 dB SPL. The total level of the noisy trials was varied (roved) by  $\pm 2.5$  dB.

■ **Participants**: two normal-hearing listeners.

■ **Task**:

Each participant performed **4000 phoneme categorizations** for each noise, indicating whether the **last syllable** was /ba/ or /da/. The SNR was continuously adapted using a one-up one-down weighted rule to ensure a correct response rate of 70.7%. In the fastACI toolbox (Osses & Varnet, 2021a):

```
experiment = 'speechACI_Logatome-abda-S43M';
fastACI_experiment(experiment, Subject_ID, noise_type);
```

■ **Data analysis**: The probability of “ga” answer was linked via a **Generalized Linear Model (GLM)** to a time-frequency representation in an ERB space of the presented stimulus. As we used in a previous study (Osses & Varnet 2021b), the GLM was fitted by **penalized likelihood maximization** with **sparseness prior**, a tradeoff between fitting the data well and obtaining a smooth ACI.

■ The ACI ( $\beta$ , Fig. 1) shows how the **presence of energy at each time-frequency point influences the decision** (i.e. which parts of the stimulus serve as cues for categorization). **Positive clusters of weights** correspond to regions favoring response “ba”, whereas **negative clusters** correspond to “da” regions.

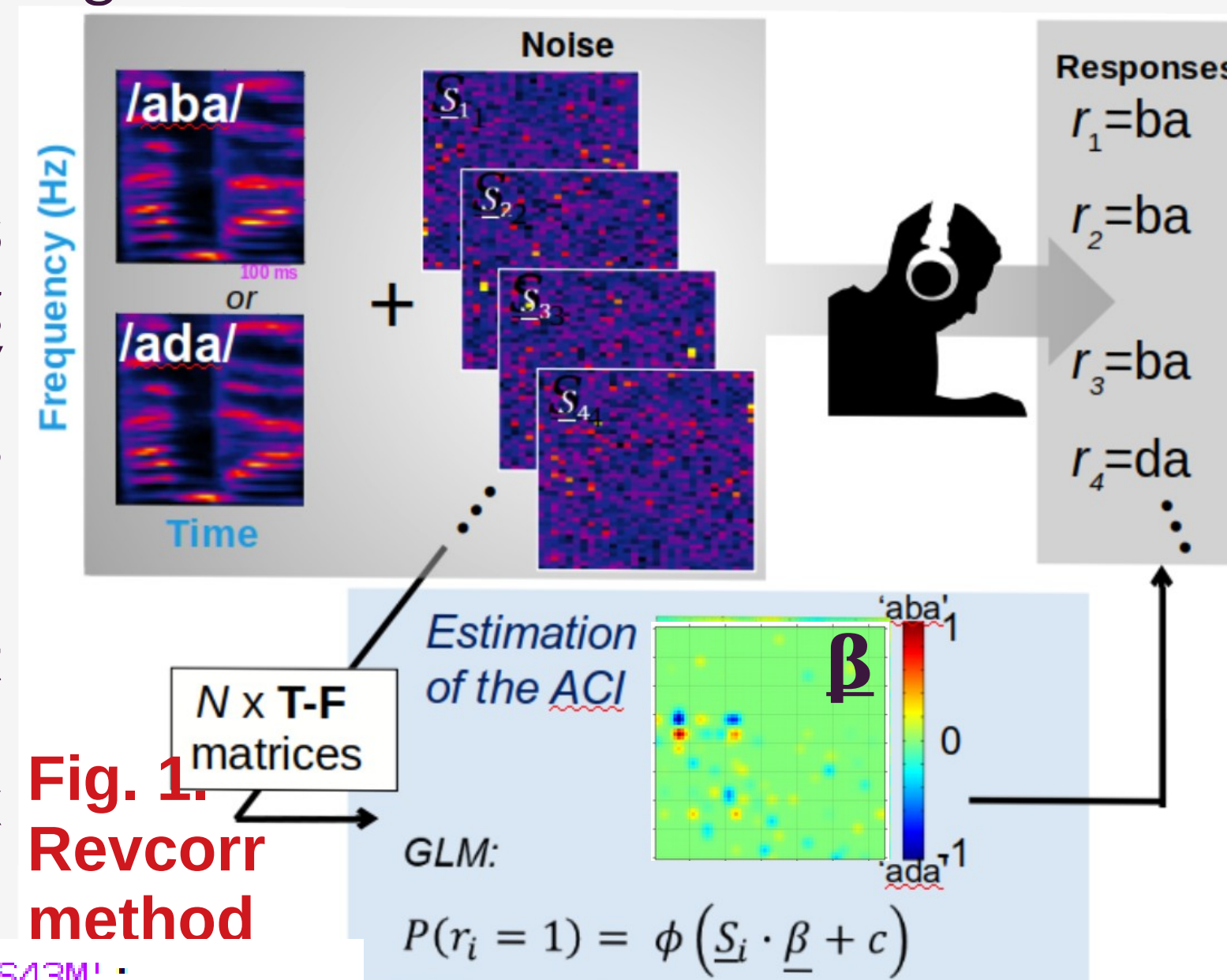


Fig. 1  
Revcorr method

## 3. Methods: Noise characterization

Fig. 2A All sounds are sampled at 16 kHz and have a duration of 0.85 ms.

■ **Power spectrum**: 512-point FFT, 90% overlap, 32-ms analysis windows

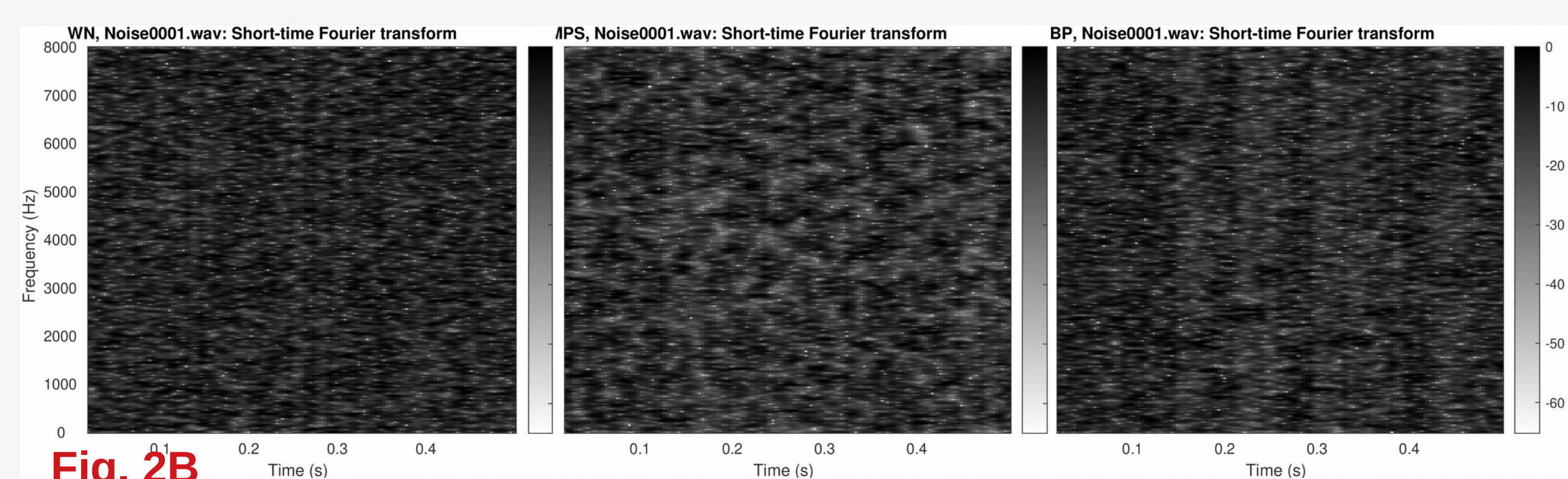


Fig. 2B

■ **Band level** within 1-ERB. bands (auditoryfilterbank.m from AMT):

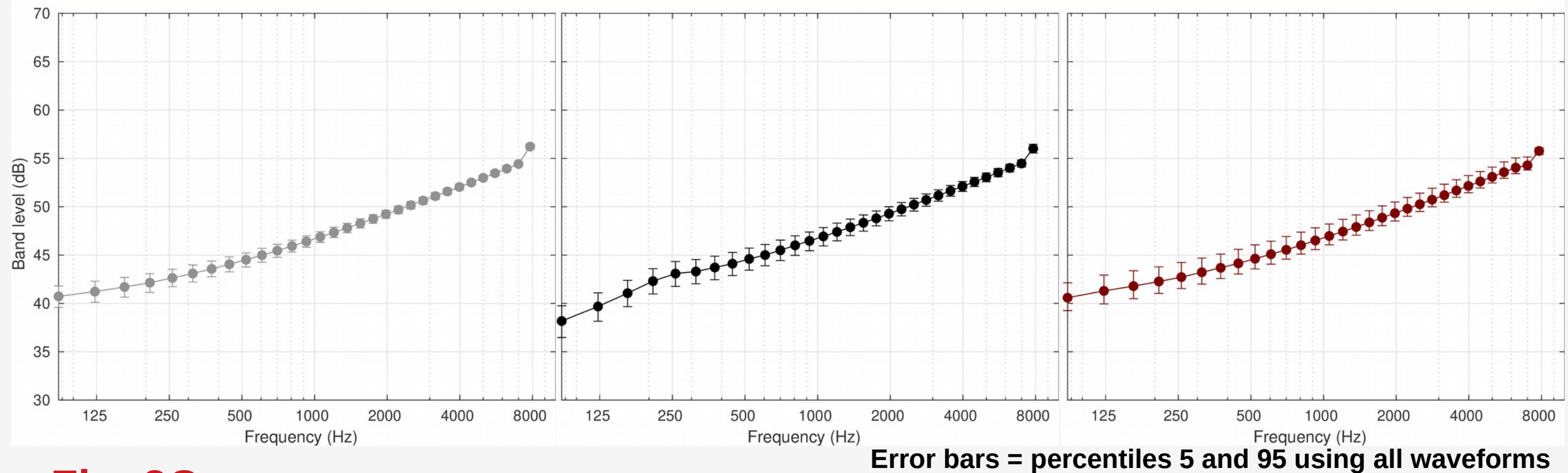


Fig. 2C

■ **Modulation power spectrum**:

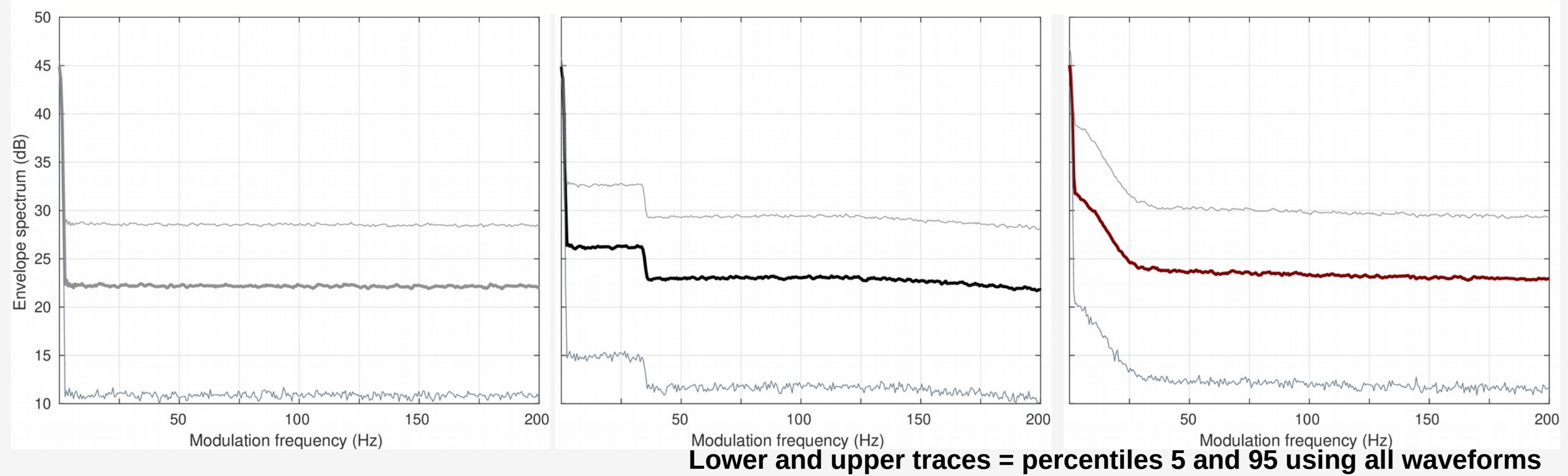
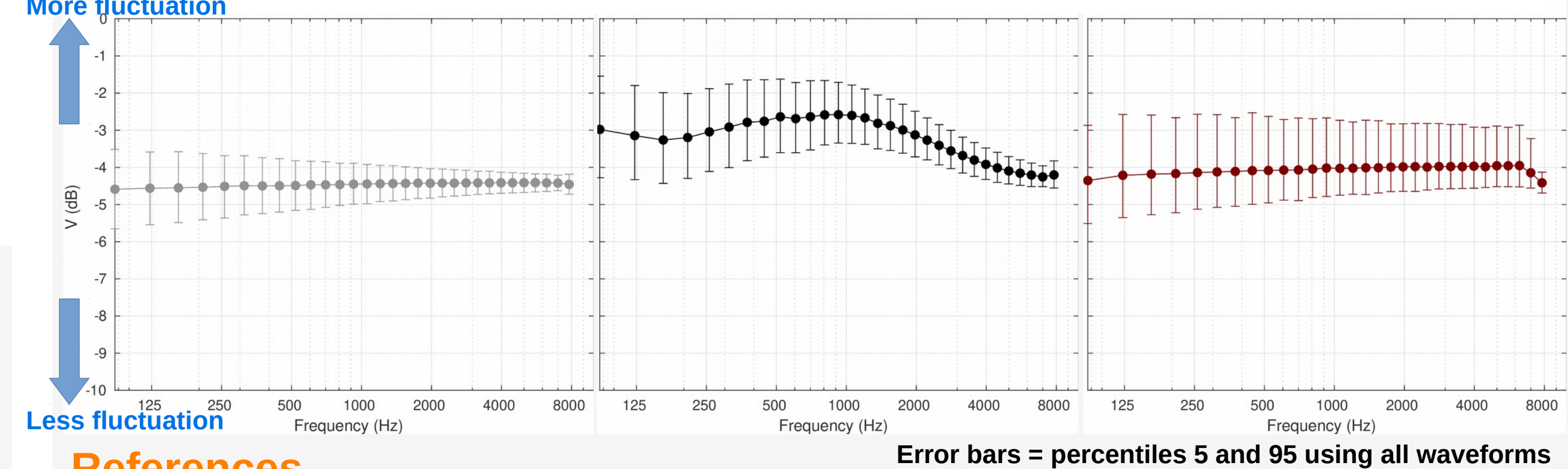


Fig. 2D

■ **V metric**: Ratio between the standard deviation and the mean envelope (Kohlrausch et al. 1997) applied to each 1-ERB-wide band



## References

- Meyer et al. (2010). Human phoneme recognition depending on speech-intrinsic variability. J. Acoust. Soc. Am., 128, 3126–3141.
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## 3. Experimental results

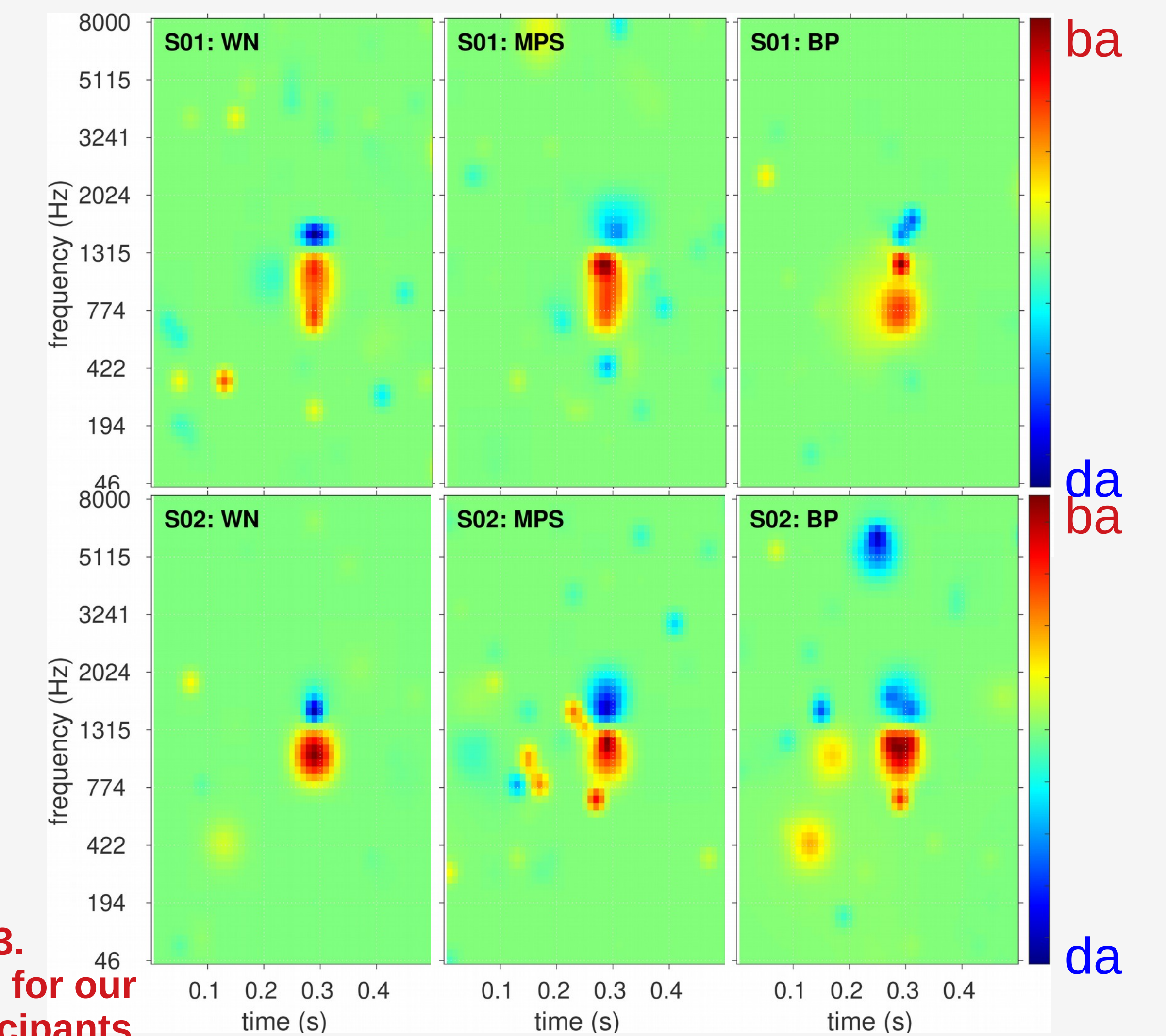


Fig. 3.  
ACIs for our participants

■ In line with previous studies using WN, all ACIs revealed a primary cue around the time onsets of the second syllable and the frequency region of the F1 and F2 formants.

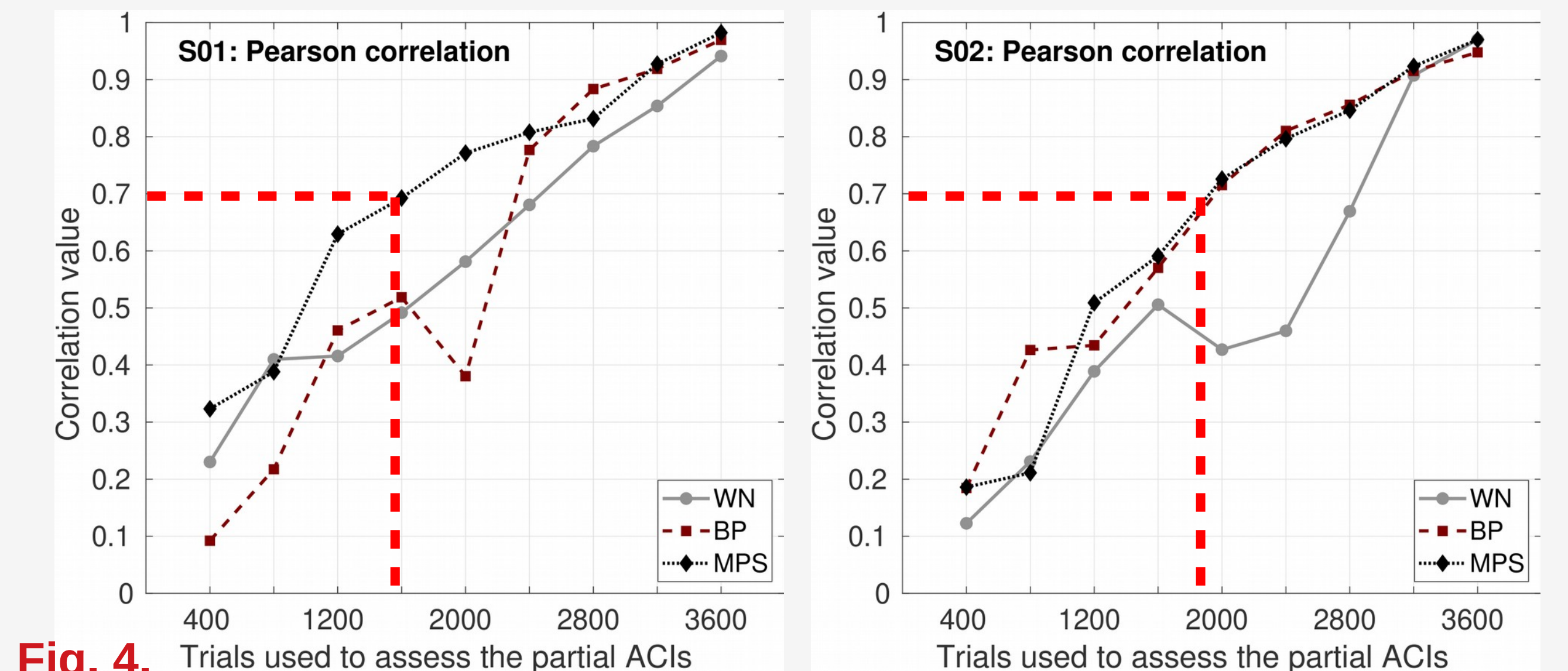


Fig. 4. Trials used to assess the partial ACIs  
Correlation between “partial ACIs” and the ACI obtained with 4000 trials

## 5. Summary

Our analyses suggest that the reverse correlation method applied to a consonant-in-noise discrimination task, the **ACI**, converges more quickly to a stable result when the background noises contain dominant components in the modulation frequency range between 0 and 40 Hz, which is the case for **MPS** and **BP** noises. The prominent envelope fluctuations in this range lead to more systematic confusion errors compared to white noise and, therefore, to higher prediction accuracy and more robust reverse correlation results.

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