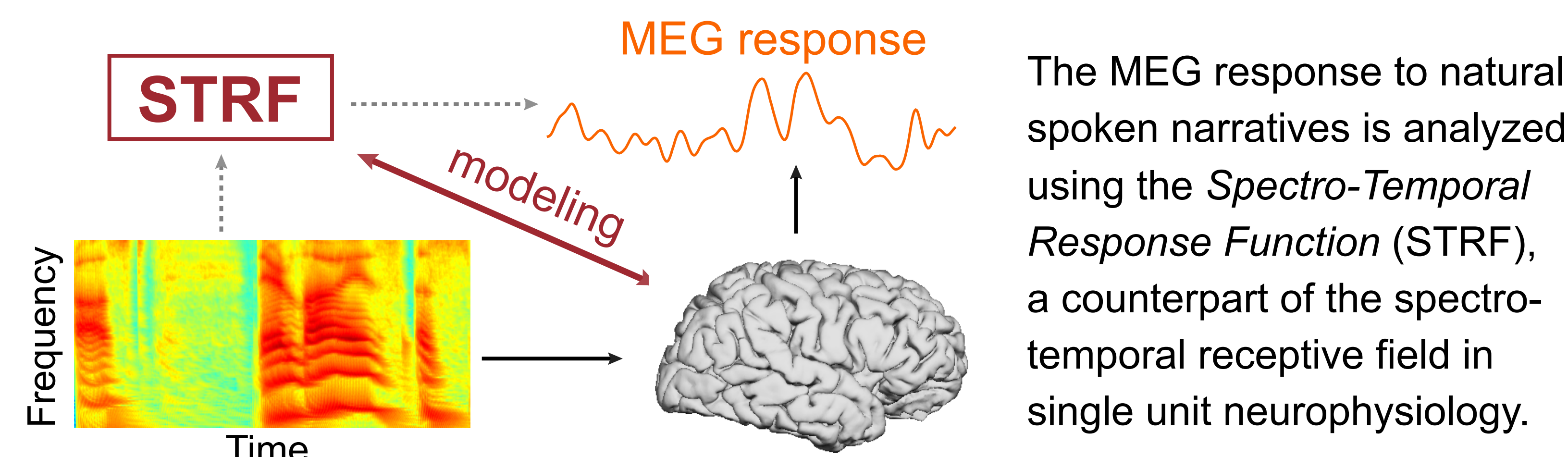


Introduction

1. What is the neural code of spectro-temporal modulations of continuous speech in human auditory cortex?
2. How does human auditory cortex segregate and maintain attention on a speech target in the presence of a concurrent speech masker?

These two questions are addressed by recording the *magnetoencephalography* (MEG) response from human subjects actively listening to spoken narratives. MEG is an non-invasive neural recording tool, with millisecond level time resolution.



Experimental Procedures

Stimulus & Procedure

Dichotic speech mixture

- Two 2 minute long spoken narratives (from *the Legend of Sleepy Hollow*, by Washington Irving) were played simultaneously to the two ears respectively.
- The stimulus was played 6 times. The subjects focused on one ear at a time and switched focus after every repetition.
- After every minute, the subject were asked a question about the comprehension of the story attended to. 90% of the questions were correctly answered.

Monaural speech

- In a separate session, each spoken narrative was played monaurally 4 times.

MEG Recording and Processing

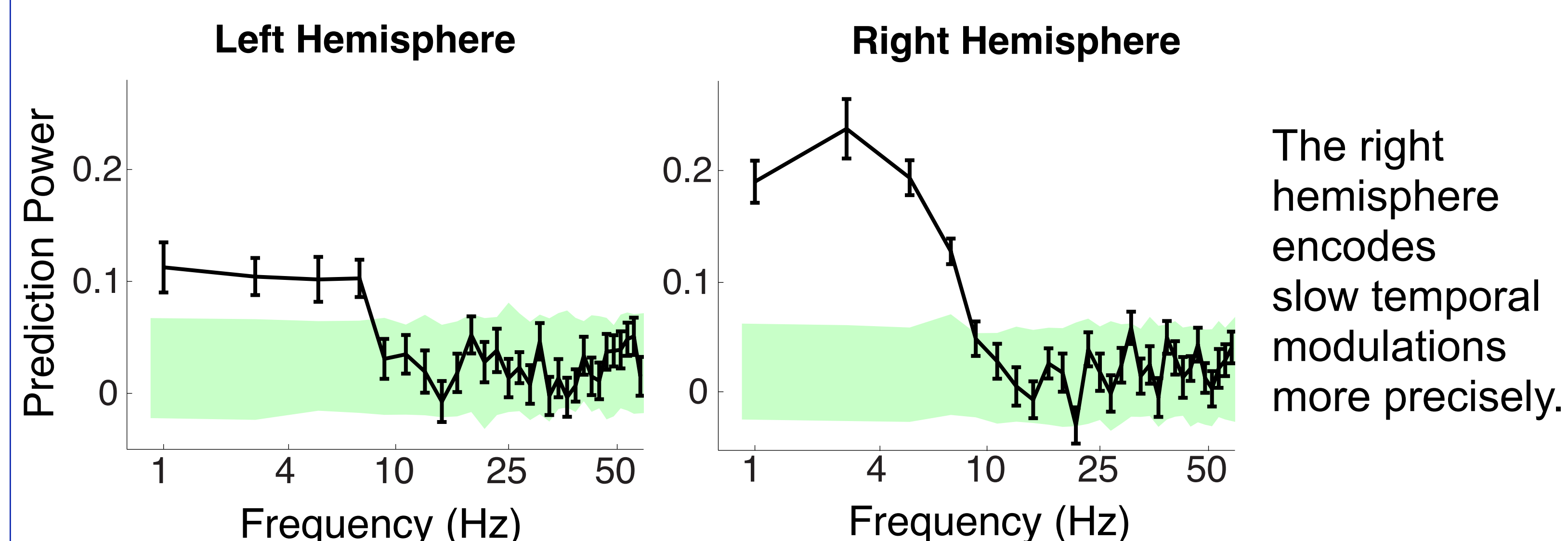
- 157 channel whole-head MEG system, sampled at 1 kHz, with a 60 Hz notch filter.
- Denoising Source Separation (DSS) assisted equivalent current dipole fitting localizes the neural source bilaterally to the *superior temporal gyrus*.
- The moment of the equivalent current dipole in each auditory cortex is reconstructed using the generalized least squares method, as a function of time.

STRF

- The STRF is estimated using boosting with 10-fold cross validation, based on a sub-cortical spectro-temporal representation of speech.

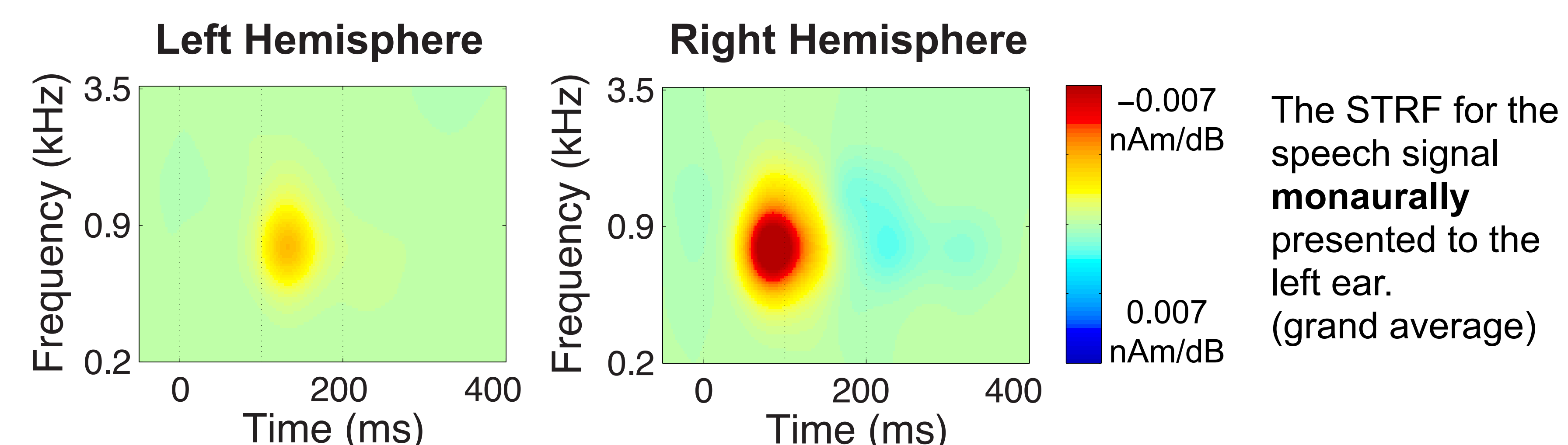
Neural tracking of slow modulations

The **predictive power** is the correlation between STRF model prediction and real MEG measurement.



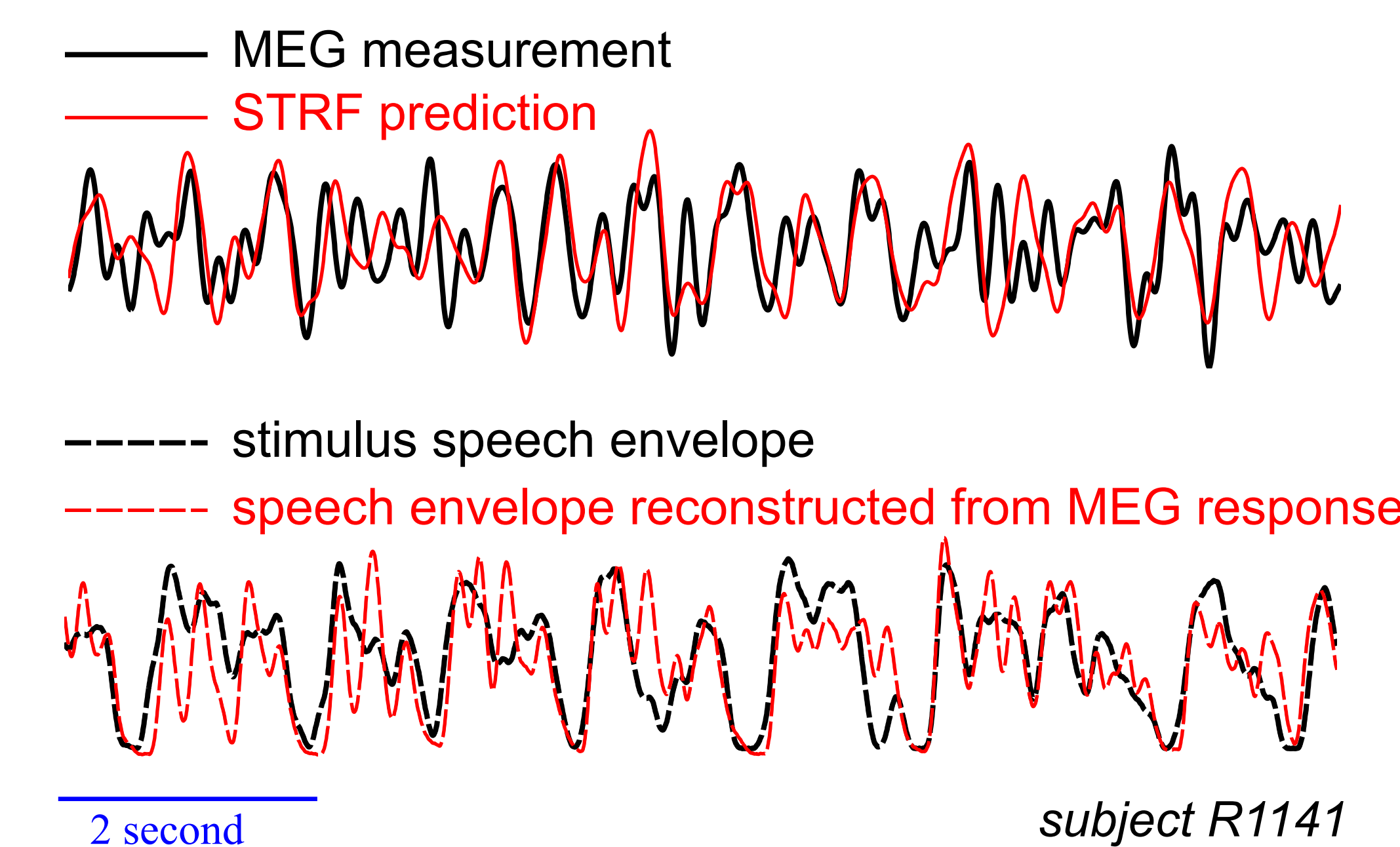
In the low frequency region (<10 Hz), the neural response can be predicted by the spectro-temporal modulations of the stimulus speech, using the STRF.

Spectro-temporal response function for speech



The STRF indicates the neural response is following the spectro-temporal modulations of speech at latency of about 100 ms, where the contralateral response is enhanced.

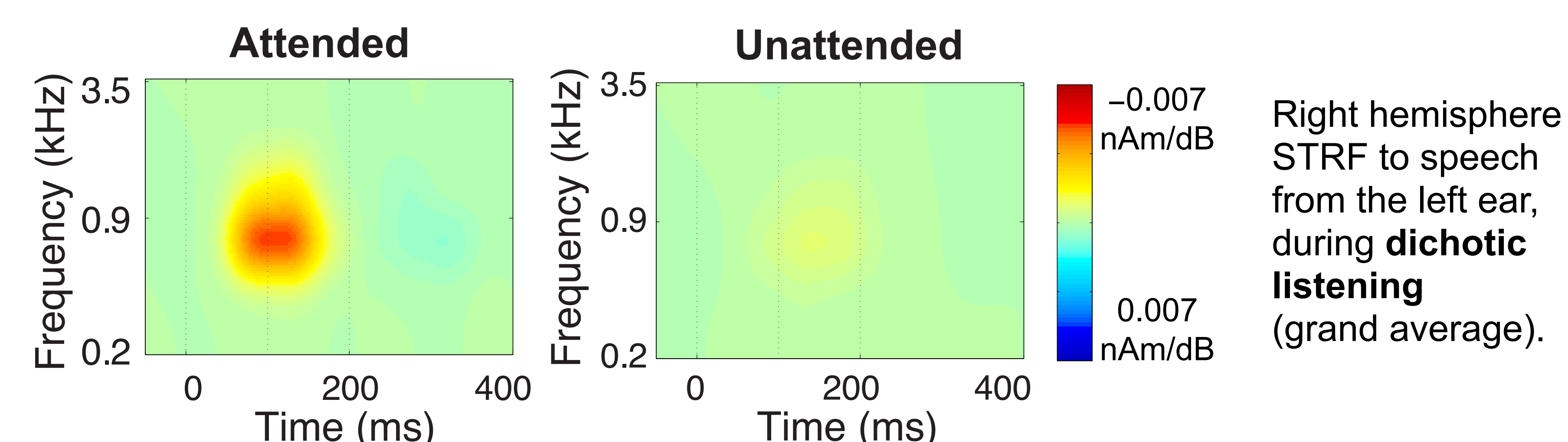
The peak at about 100 ms is called the **M100-like response**. The polarity of this peak is the same as the polarity of the M100 (N1m) response to tone pip.



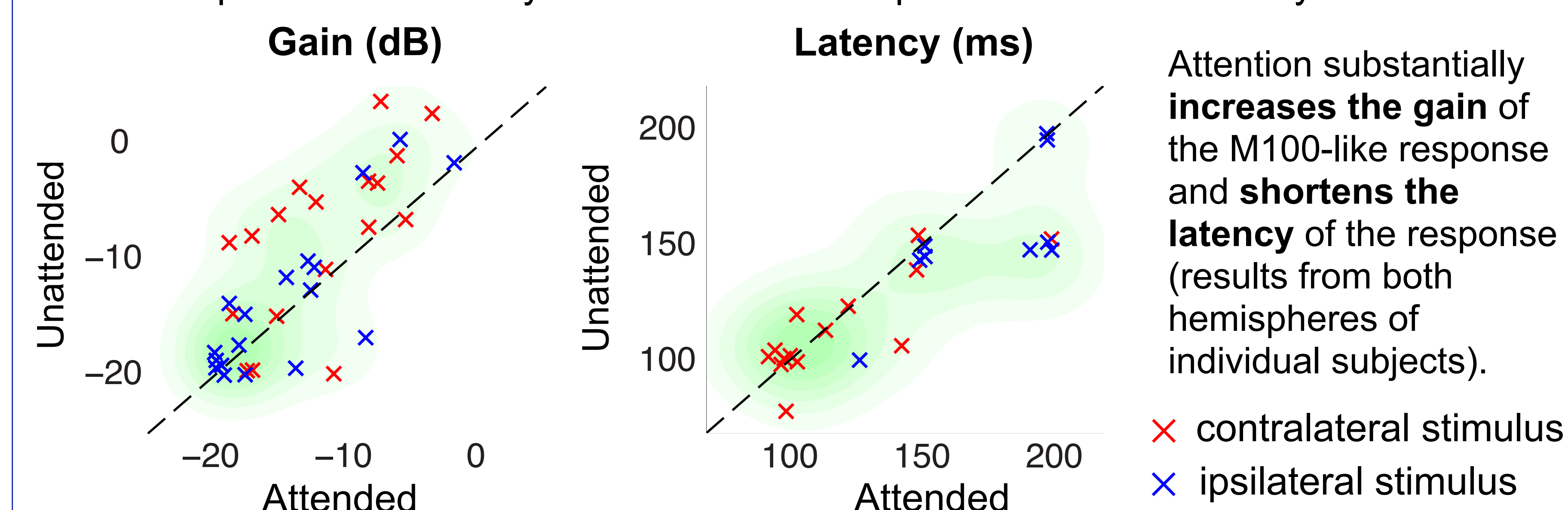
The inverse of the STRF can be used to *decode* the speech envelope information from the MEG response.

The neural activity in human auditory cortex precisely encodes slow (<10 Hz) temporal modulations of speech.

Spatial cue based speech segregation



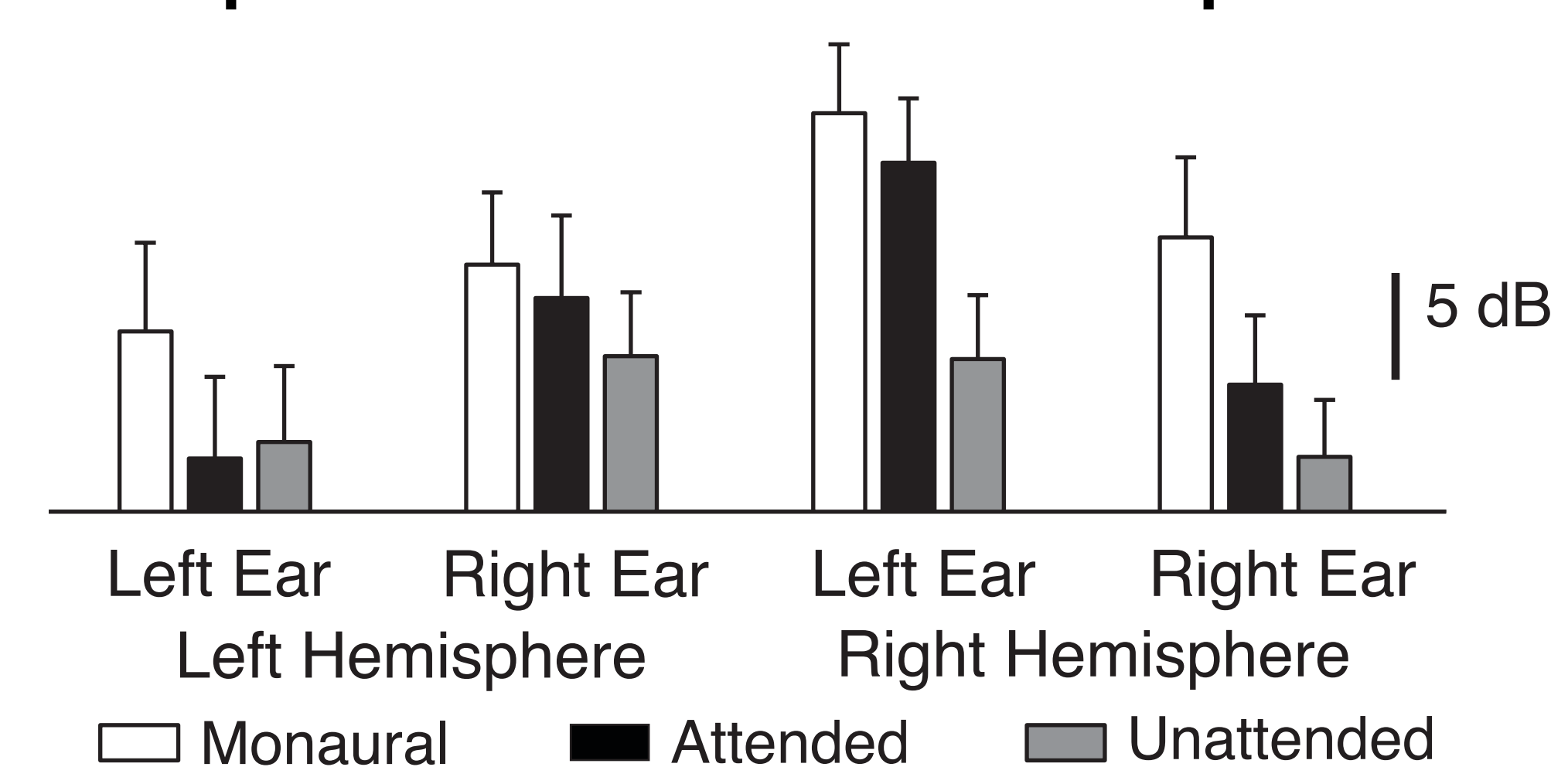
The amplitude and latency of the M100-like response are modulated by attention.



Purely spatial cue based speech segregation occurs in human auditory cortex, within 100 ms, reflected by the different attentional gains of the target and masker speech.

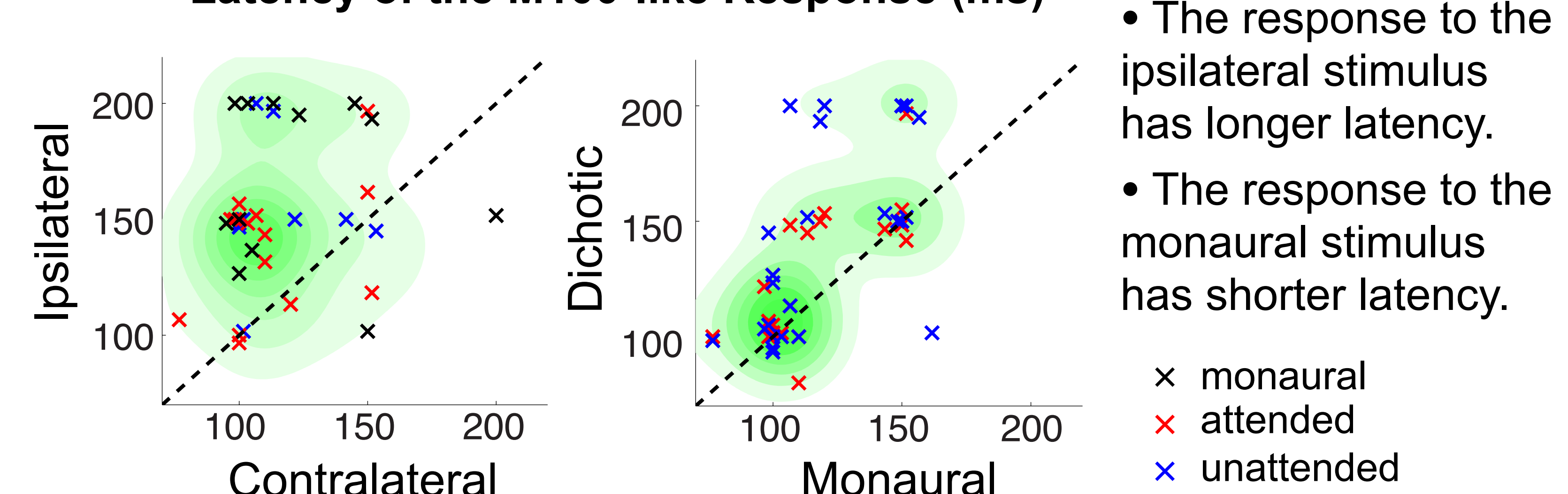
Binaural interaction & Hemispheric lateralization

Amplitude of the M100-like Response



- The response is stronger in the right hemisphere.
- The response is stronger for monaural stimuli than dichotic stimuli.
- Each hemisphere is more responsive to the contralateral stimulus.

Latency of the M100-like Response (ms)



- The response to the ipsilateral stimulus has longer latency.
- The response to the monaural stimulus has shorter latency.

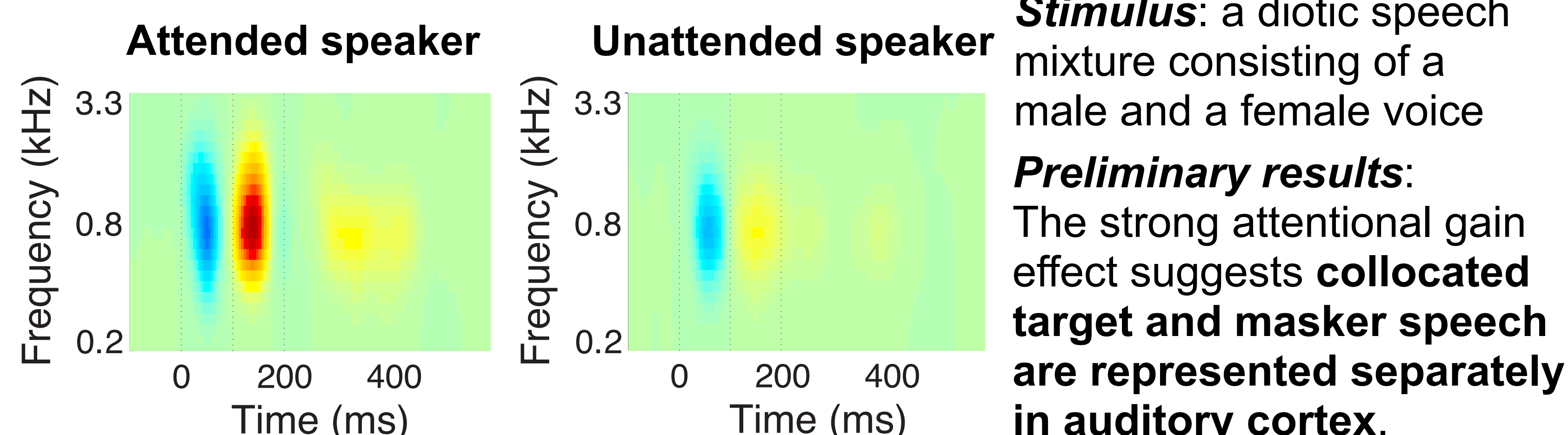
Each hemisphere responds stronger and faster to contralateral stimuli. The response is weakened and delayed by stimuli in the other ear.

Summary & Future work

• A faithful and robust neural coding of the temporal modulations of speech is observed using noninvasive neural recording methods. This neural code is strongly modulated by attention.

• A paradigm is provided to study the neural processing of continuous speech in various natural auditory scenes.

To examine whether speech segregation as a general phenomenon occurs in human auditory cortex, we investigate the neural processing underlying **speaker feature cue based speech segregation**.



Reference

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S.V. David, N. Mesgarani & S.A. Shamma, *Network: Comput. Neural Syst.* **18** (2007)
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