

Efficient Analysis of Evoked Potentials using Linear Mixed-Effects Regression

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Introduction

One in five adults suffer from chronic pain, which is the result of disturbed processes in the central nervous system.

- Early detection of these disturbances enables better treatments and less clinical efforts per patient.
- Disturbances in nociceptive processing can be studied by estimation of the evoked potential in response to a stimulus.

The maximum number of stimuli in experiments with human subjects is limited.

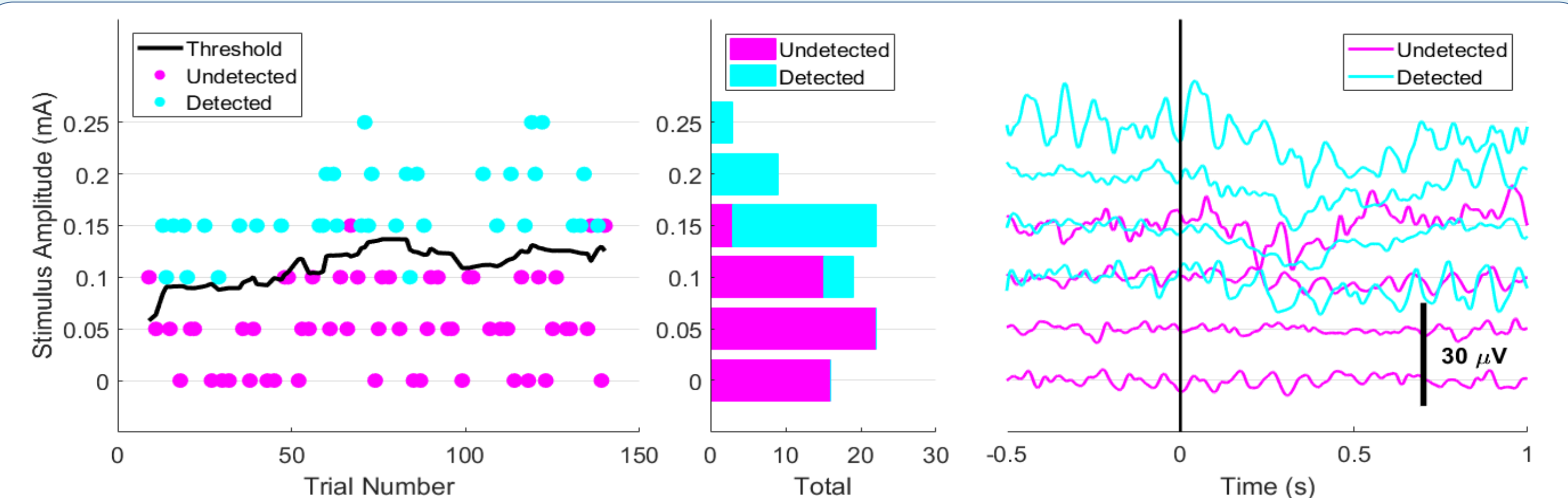
- Stimulus selection methods are used for a more efficient probing of the stimulus parameter space.
- This results in different amounts of trials per stimulus property, which is problematic for conventional averaging.

An analysis method which is robust to variations in the amount of trials is provided by a linear mixed model.

Analysis

During estimation of the detection probability stimulus parameters are constantly varied, leading to a very low number of trials per combination of parameters.

- Very poor SNR ratio using conventional averaging.
- Pooling data provokes confounding and reduces the percentage of variance explained by our estimate.

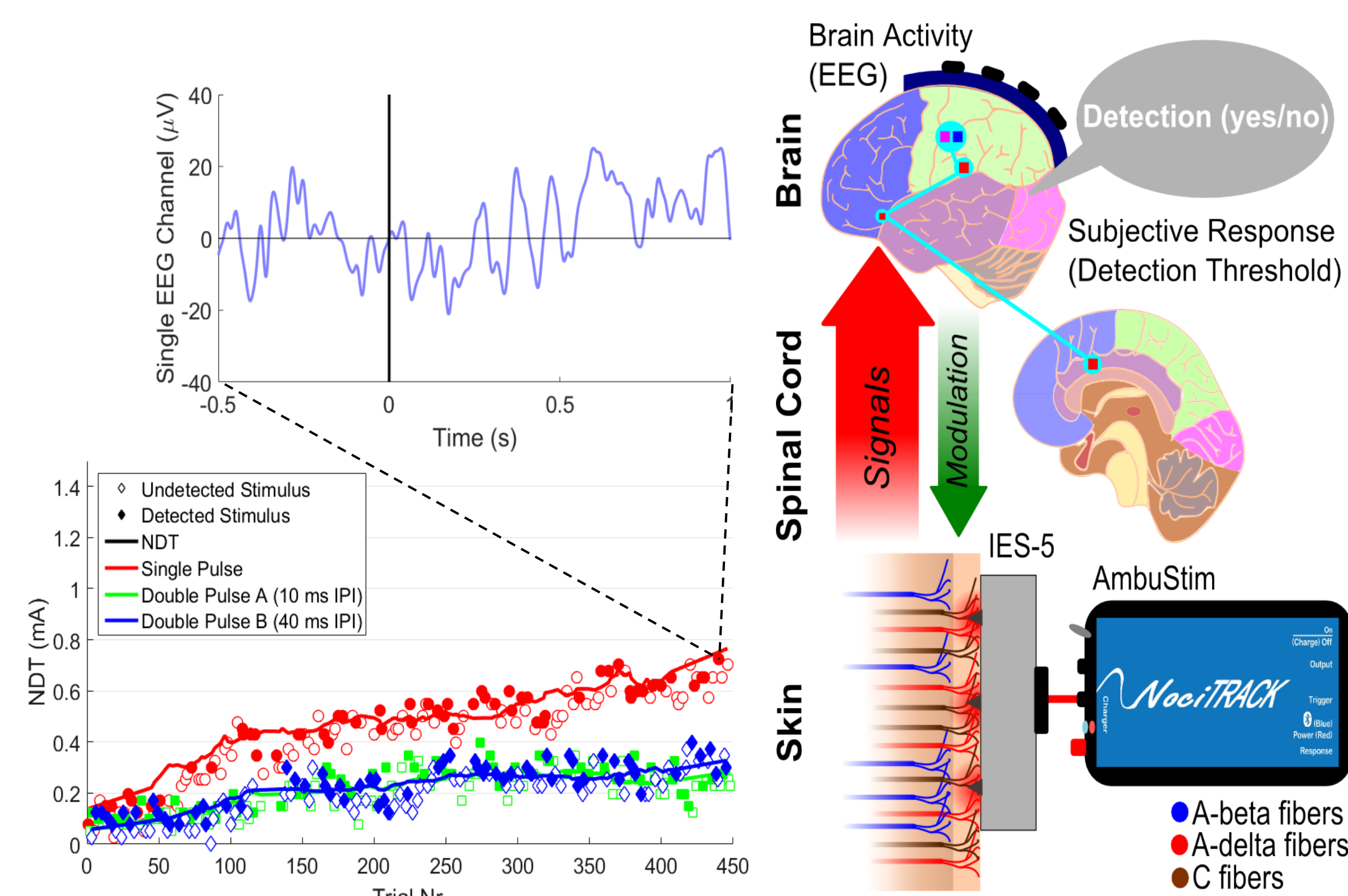


Experiment

Quantitative assessment of nociceptive sensitivity

- Detection probability of stimuli
- Neurophysiological response to stimuli

1. Specific stimulation of nociceptive nerve fibers.
2. Tracking of the nociceptive detection probability and threshold using an adaptive stimulus sequence (multiple threshold tracking, MTT).
3. Measurement of the subjective response with respect to every stimulus (detected or undetected).
4. Measurement of the EEG signal with respect to every stimulus.



Averaging assumes that the signal in the j -th trial consists of the real signal and random trial-specific background activity.

$$y_j(\tau) = \underbrace{\beta_0(\tau)}_{\text{Real signal}} + \underbrace{\eta_j(\tau)}_{\text{Noise/Background activity}} \quad (1)$$

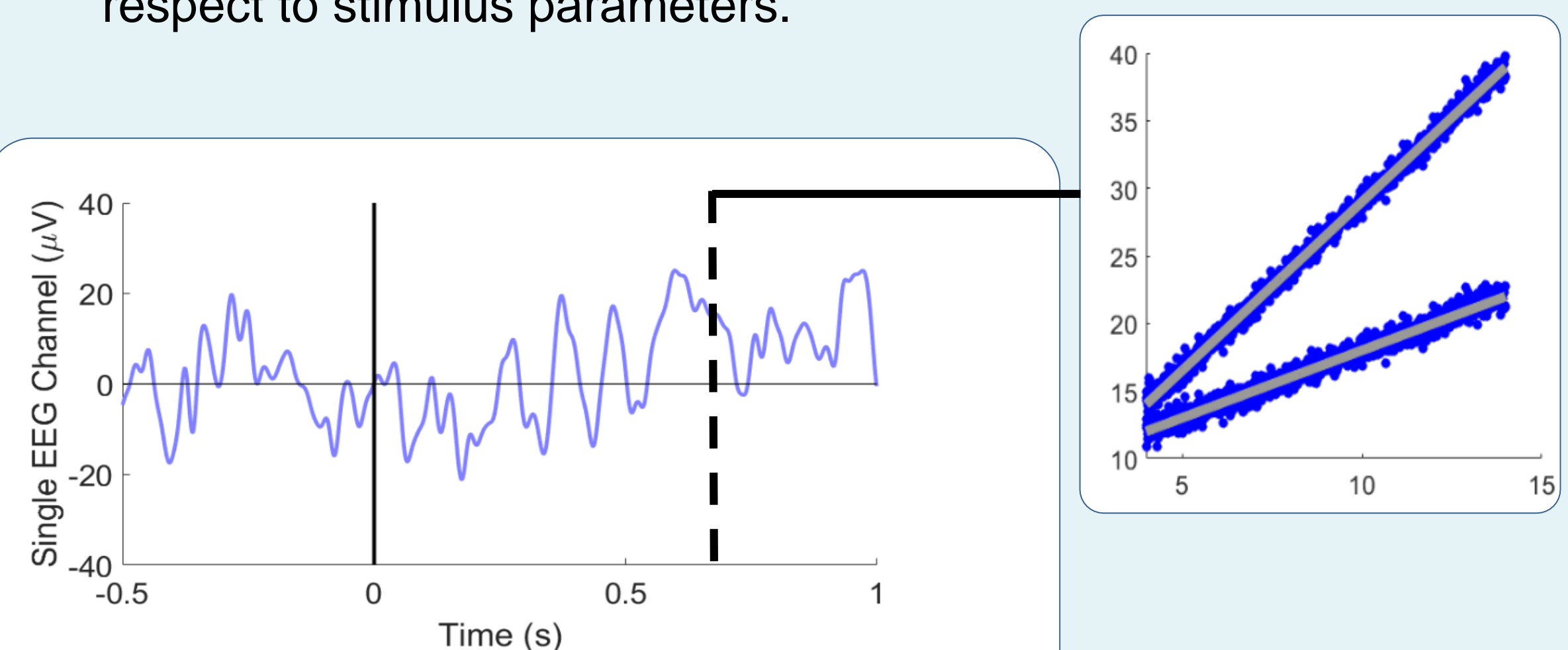
We can improve averaging by accounting for variation of the signal.

- With respect to stimulus parameters ($\beta_0(\tau) + \sum_{k=1}^p \beta_k(\tau)x_{ijk}$).
- With respect to the subject ($v_{i0}(\tau) + \sum_{k=1}^p v_{ik}(\tau)z_{ijk}$).

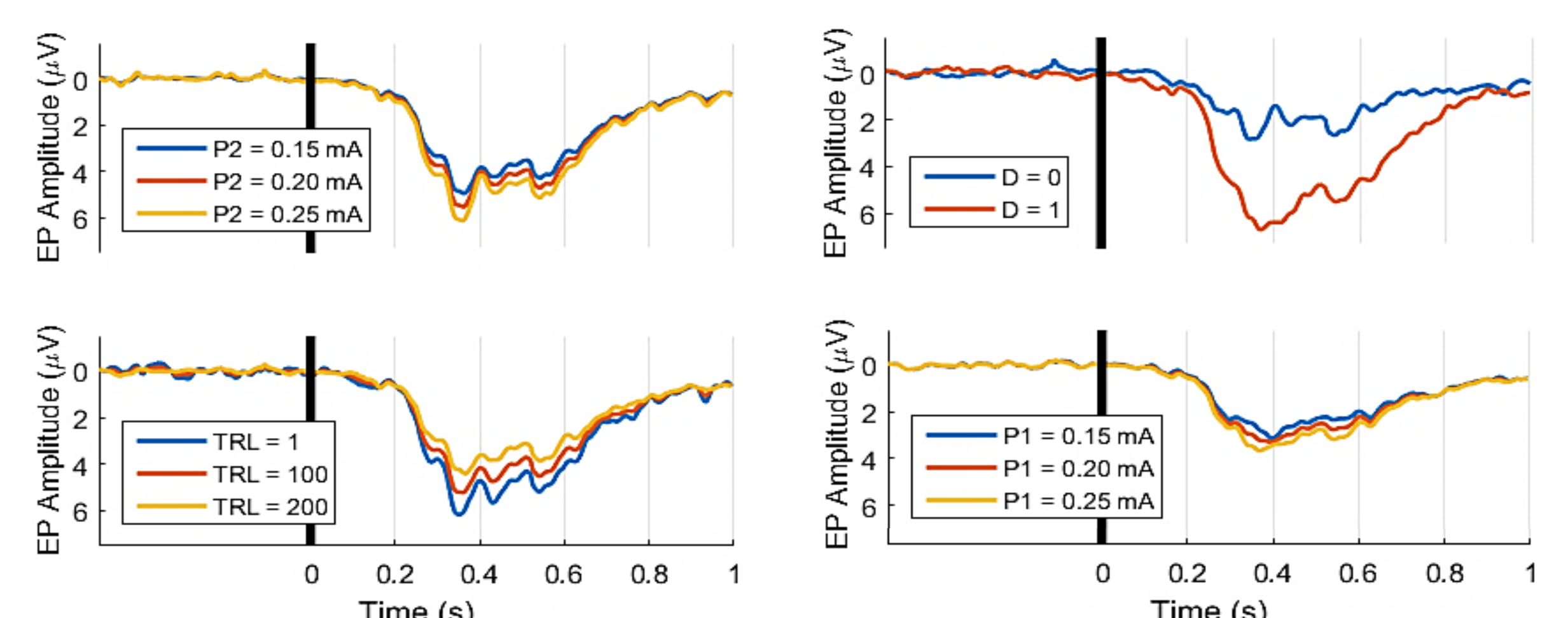
$$y_{ij}(\tau) = \underbrace{\beta_0(\tau) + \sum_{k=1}^p \beta_k(\tau)x_{ijk}}_{\text{Real signal}} + \underbrace{v_{i0}(\tau) + \sum_{k=1}^p v_{ik}(\tau)z_{ijk}}_{\text{Noise/Background Activity}} + \underbrace{\eta_{ij}(\tau)}_{\text{Noise/Background activity}} \quad (2)$$

We can analyze evoked potentials using a linear mixed model (Eq. 2).

1. Compute model coefficients for every point in time.
2. Use model coefficients to analyze variation of the evoked potential with respect to stimulus parameters.



Individual effect of stimulus parameters can be shown by prediction using the model coefficients (β_k).



Conclusion

- A linear mixed model can improve analysis of evoked potentials during multi-stimulus experiments (e.g. during multiple threshold tracking).
- The effect of experimental parameters can be analyzed using model coefficients.