



MRC Cognition  
and Brain  
Sciences Unit



UNIVERSITY OF  
CAMBRIDGE

# EEG/MEG 1:

Averaging

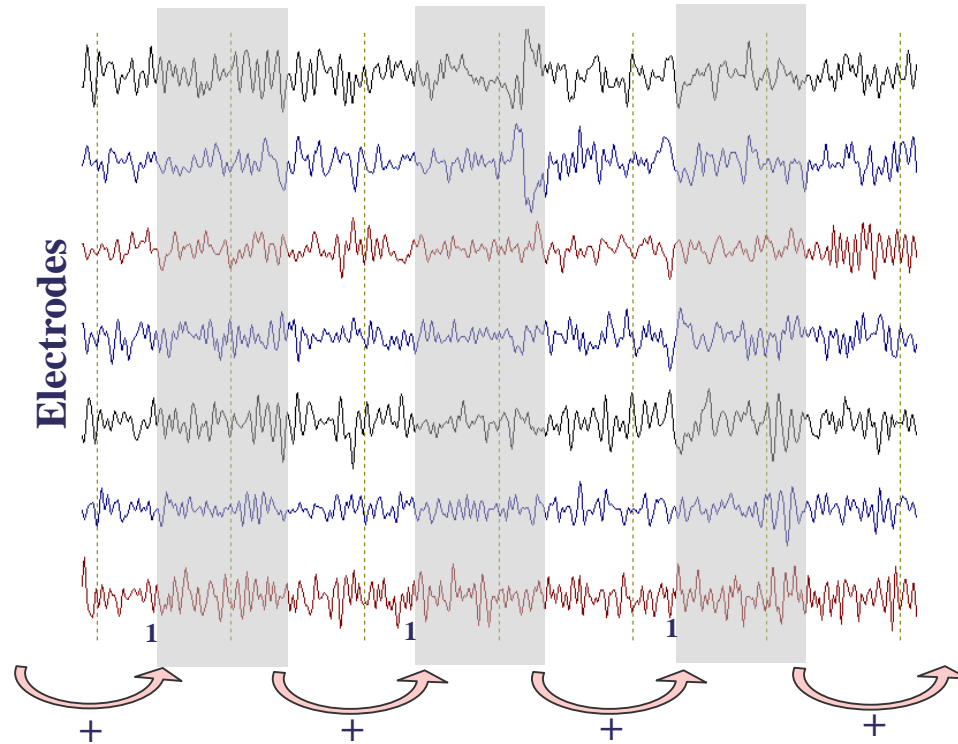
Olaf Hauk

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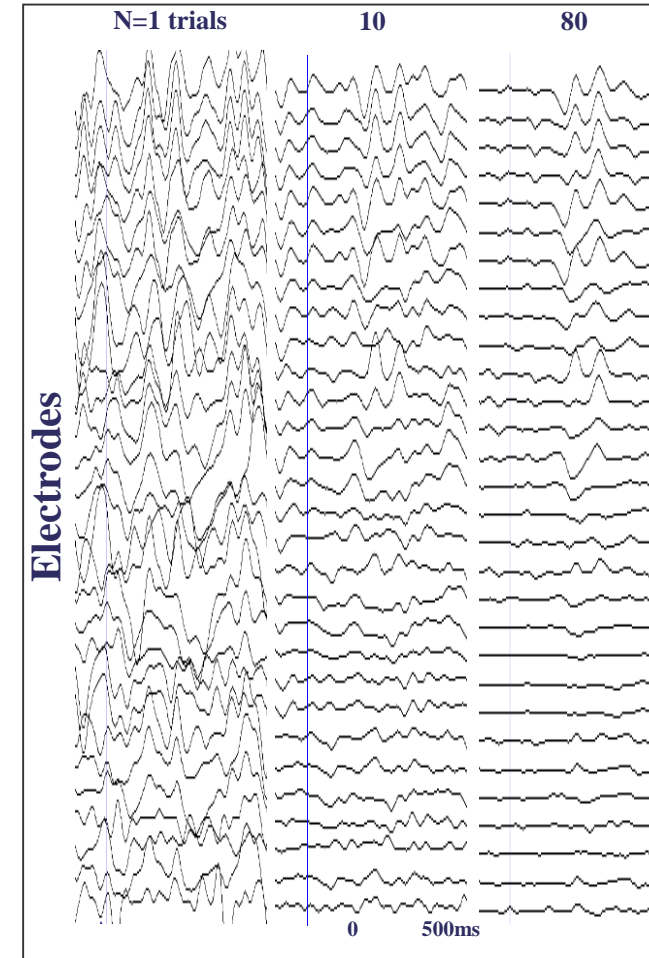
# Event-Related Potentials and Fields (ERPs and ERFs)

# Data Averaging

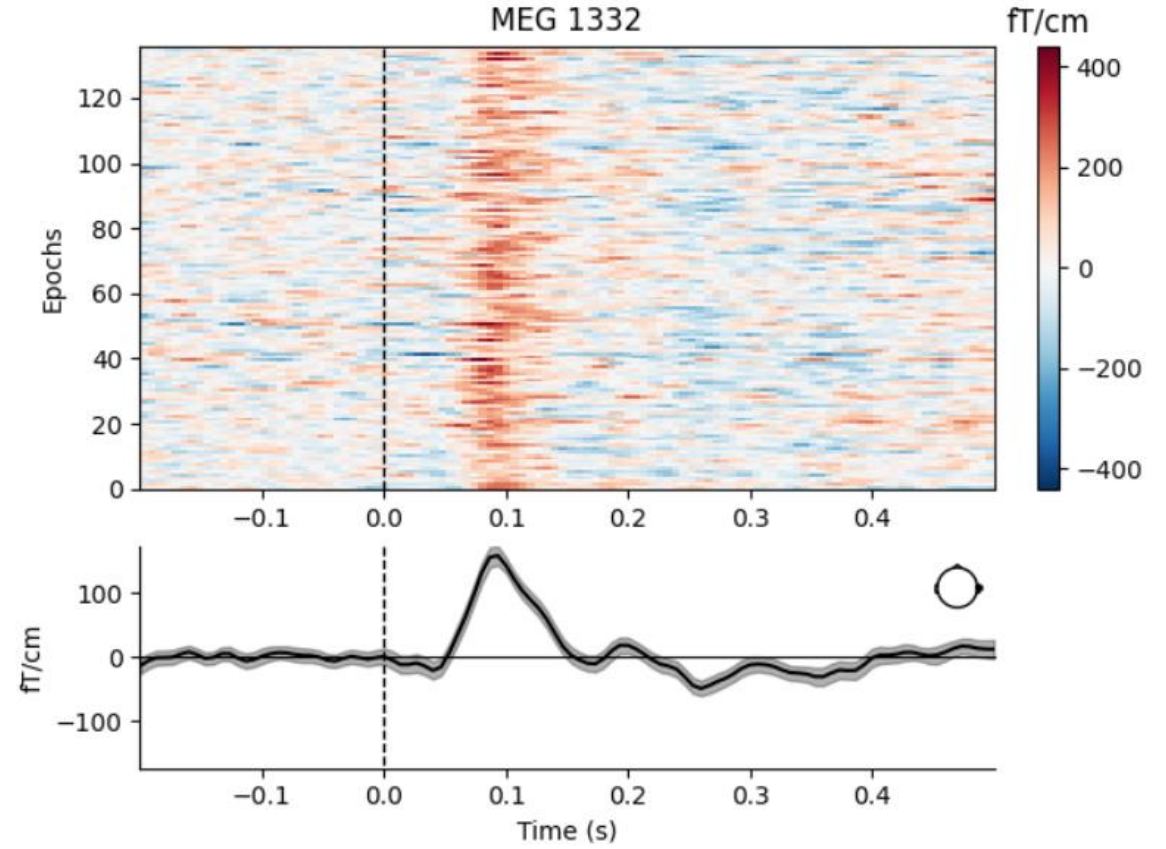
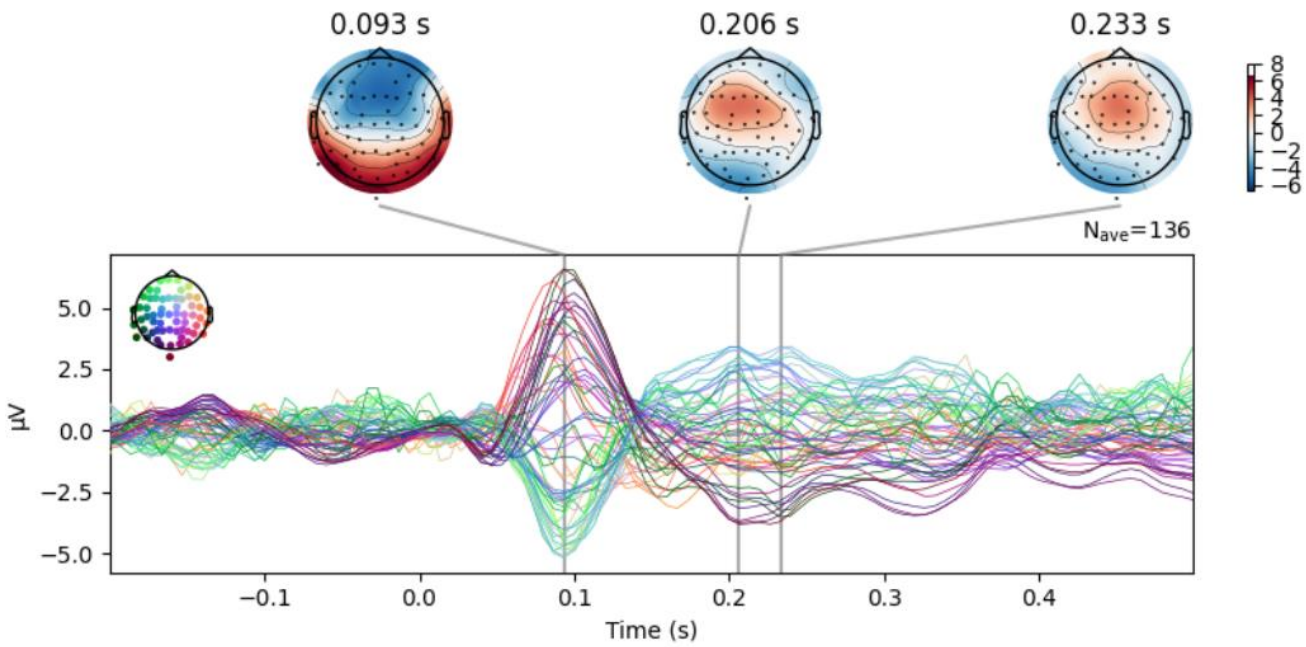
Continuous “raw” data:



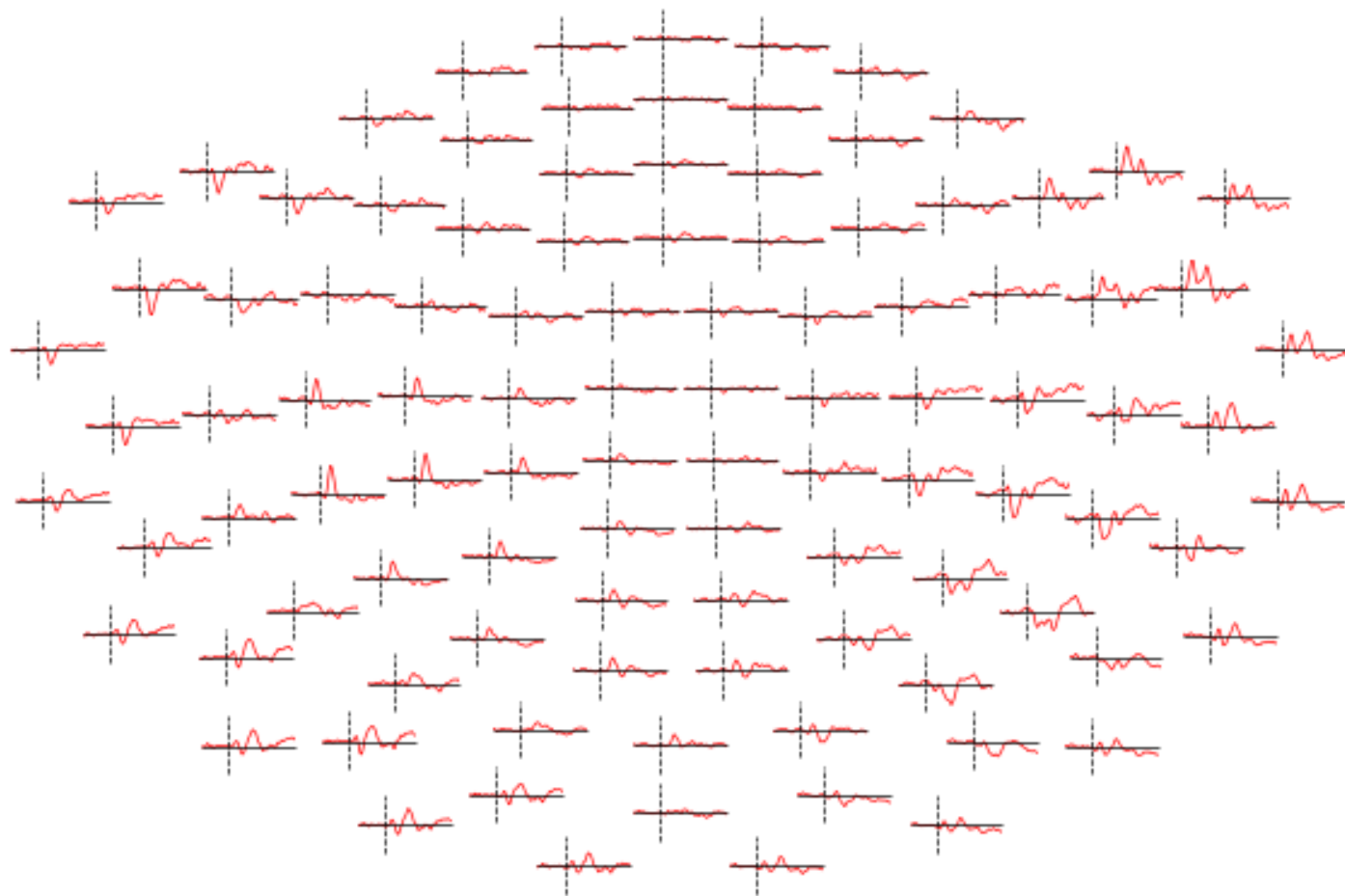
Averaged data:



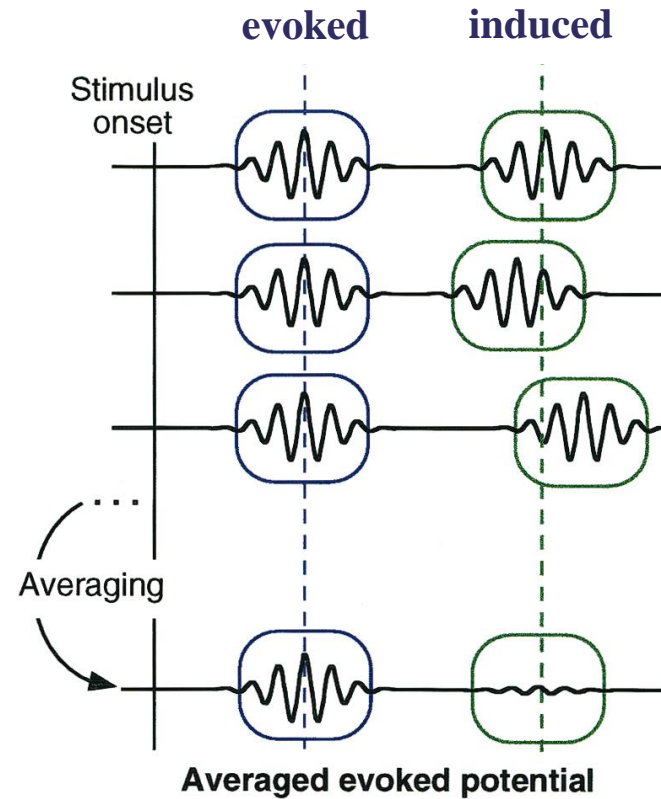
# Event-Related Potentials



# Event-Related Fields



# Evoked and Induced Activity



Tallon-Baudry & Bertrand, TICS 1999

Temporal jitter across trials has a larger effect on higher frequencies, and they are more likely to be attenuated by averaging.

# Data Averaging

The necessary number of trials depends on effect size, noise, variability across participants, your stats etc. – the more the better if feasible.

For random noise, variance goes down with  $n$ , and standard deviation with  $\sqrt{n}$ .

For “one-off” artefacts, amplitude in the average goes down with  $n$ .

“Robust Averaging” procedures exist (e.g. in SPM) that weigh epochs with an estimate of their reliability (e.g. distance to mean).

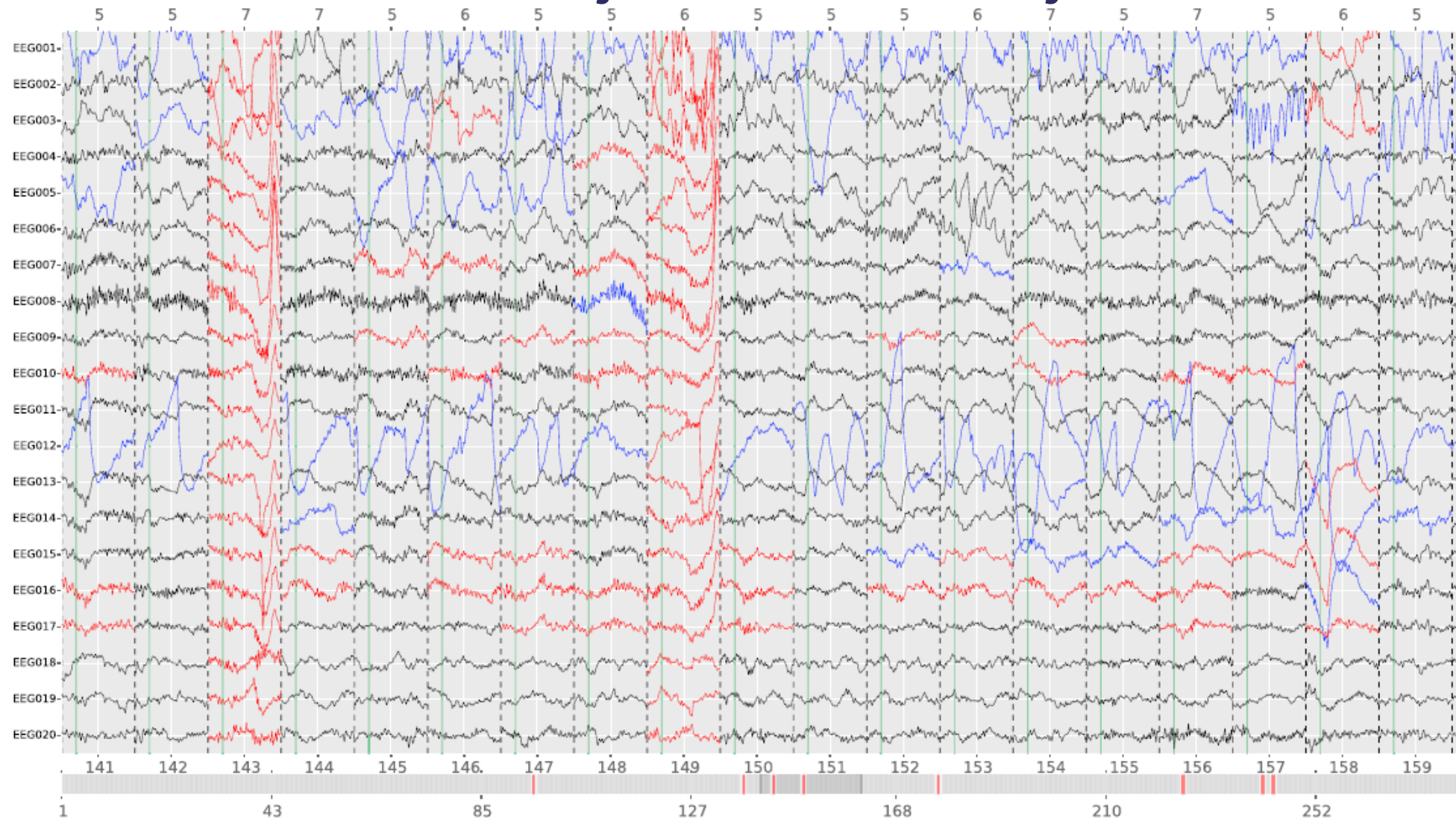
The average will be affected by the amplitude as well as jitter of evoked activity. Amplitude differences between conditions may therefore reflect true activation differences or different variability across trials.

# Artefact Rejection

1. Usually, epochs are excluded from averaging when their data exceed some maximum-minimum criterion.
2. Make sure “chronically bad channels” are excluded from this procedure (or there won’t be any data left to average).
3. Prior to any procedure that combines signals across channels, such as average reference, SSP or ICA, bad channels should be removed or interpolated (or signals from bad channels may be projected into the good ones).
5. Appropriate filtering and artefact correction (e.g. ICA) should be applied beforehand (but don’t feel too safe: artefacts may slip through).
6. It may be wasteful to reject whole epochs because of artefacts in few channels, especially when data are “precious”, e.g. for patient data and small datasets. There are algorithms that attempt to deal with this, e.g. “Autoreject” (<https://autoreject.github.io/stable/index.html>).
7. The proof of the pudding is in the eating: Check data quality by visual inspection, compute SNRs, etc.



# Artefact Rejection – “Autoreject”

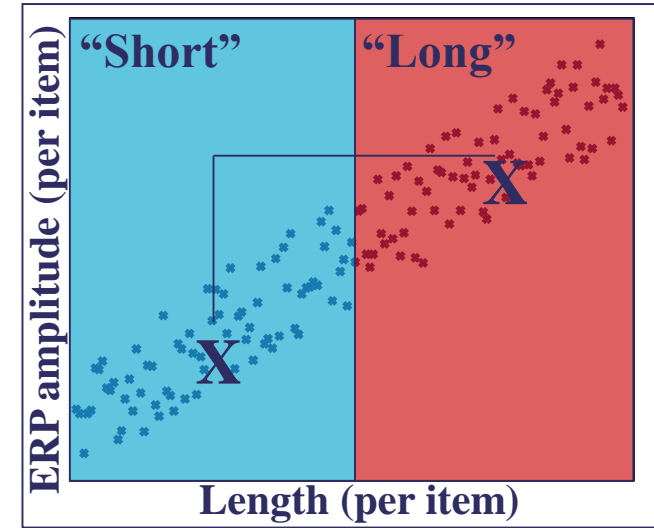
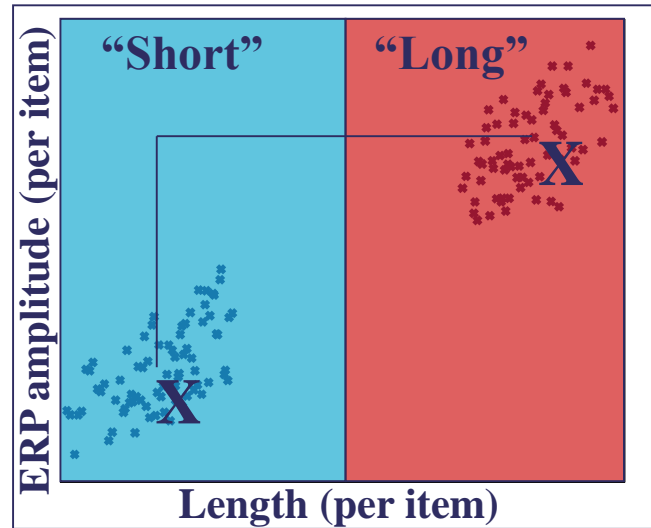


Good Bad and interpolated Bad not interpolated

# Parametric vs Factorial Designs

Consider parametric analysis/GLM if stimulus variables are continuous.

(still less common in EEG/MEG than in fMRI analysis)

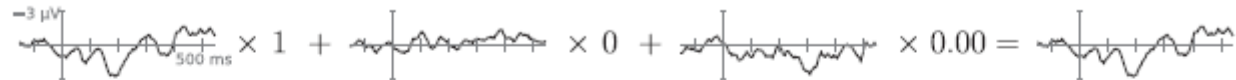


# “Single-trial” analysis

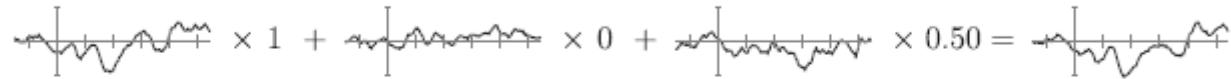
$\beta_1$      $x_{1i}$      $\beta_2$      $x_{2i}$      $\beta_3$      $x_{3i}$     Predicted ERP

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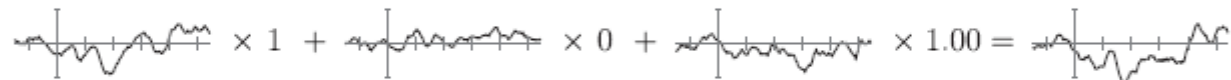
*a* with 0% expectancy:



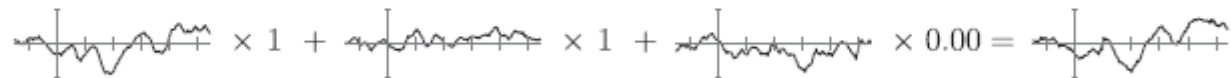
*a* with 50% expectancy:



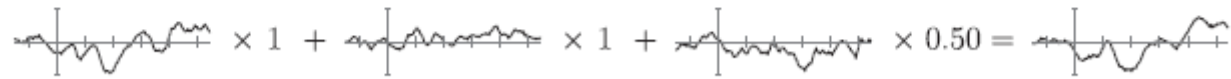
*a* with 100% expectancy:



*an* with 0% expectancy:



*an* with 50% expectancy:

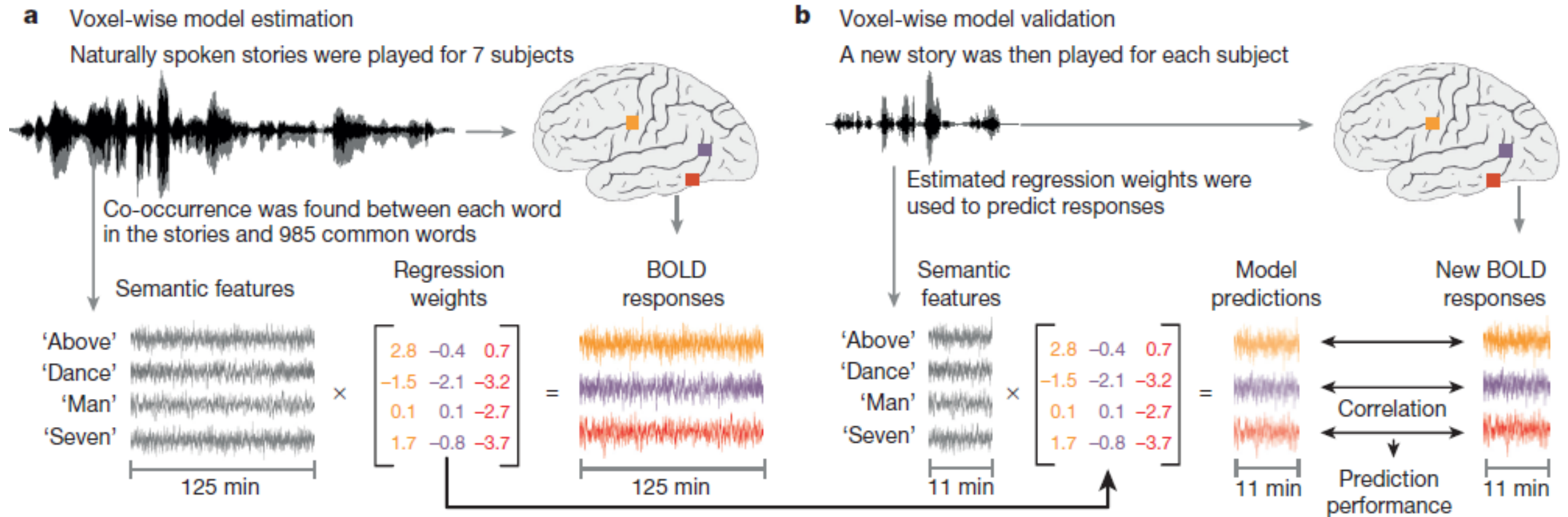


*an* with 100% expectancy:

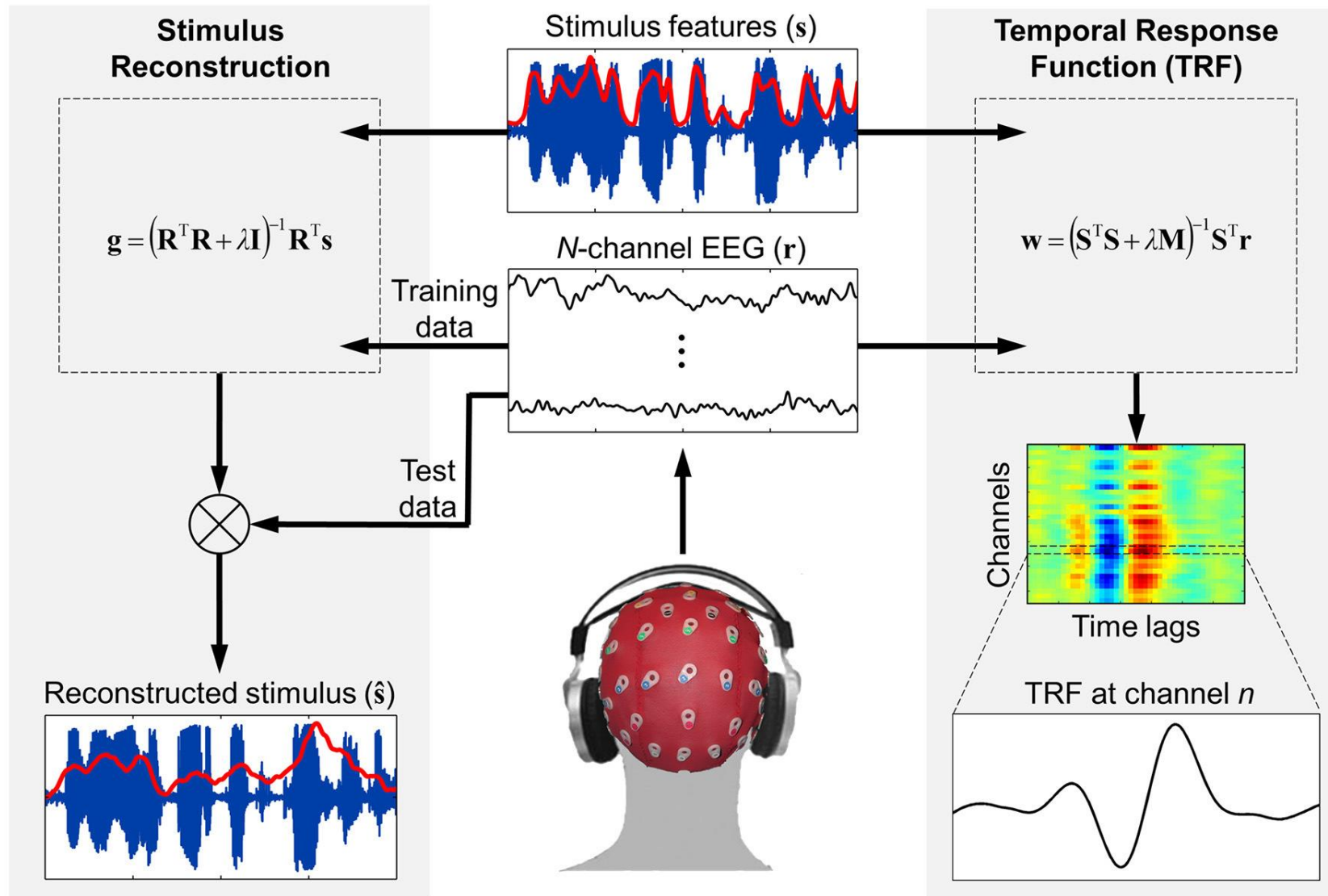


# Parametric vs Factorial Designs

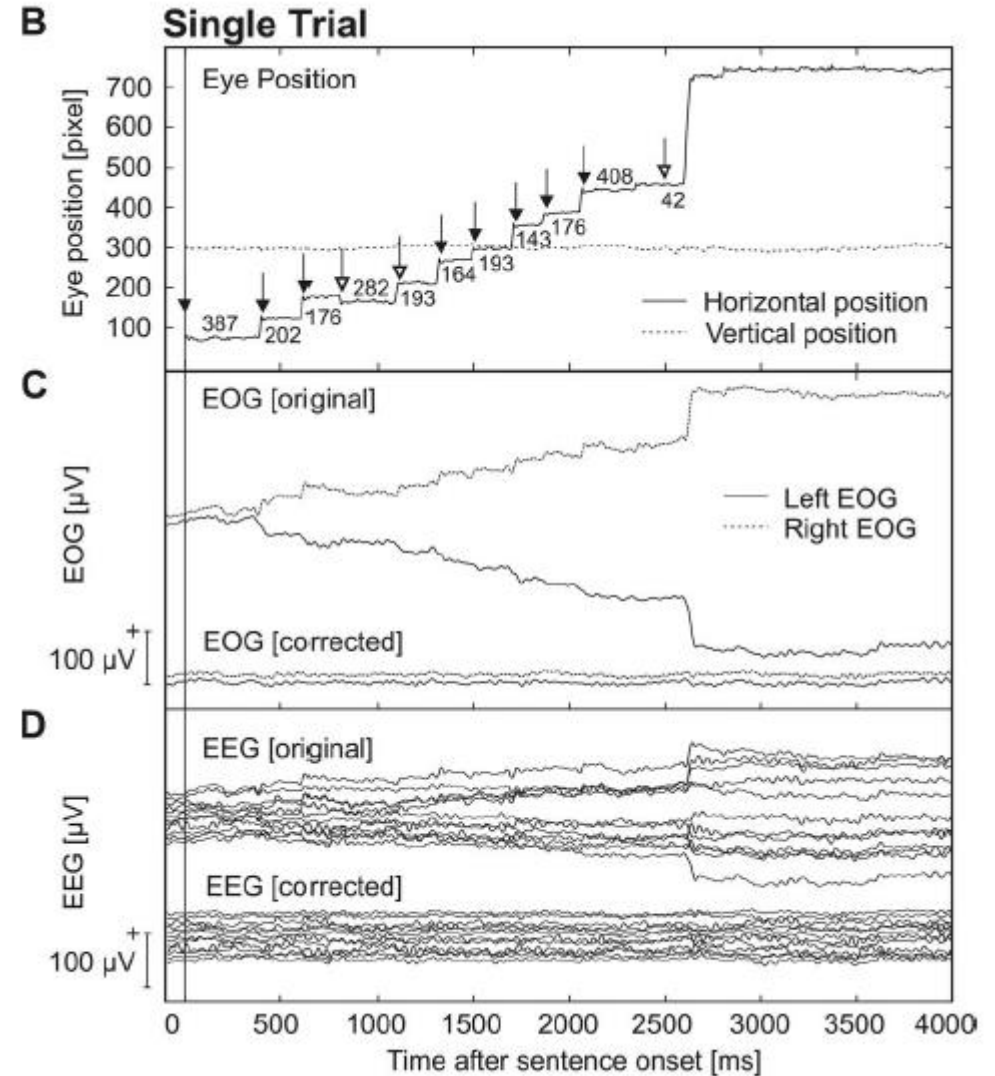
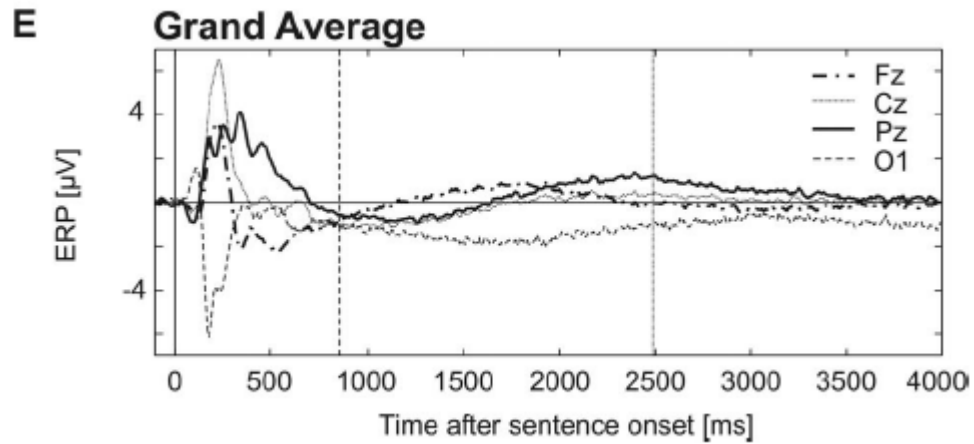
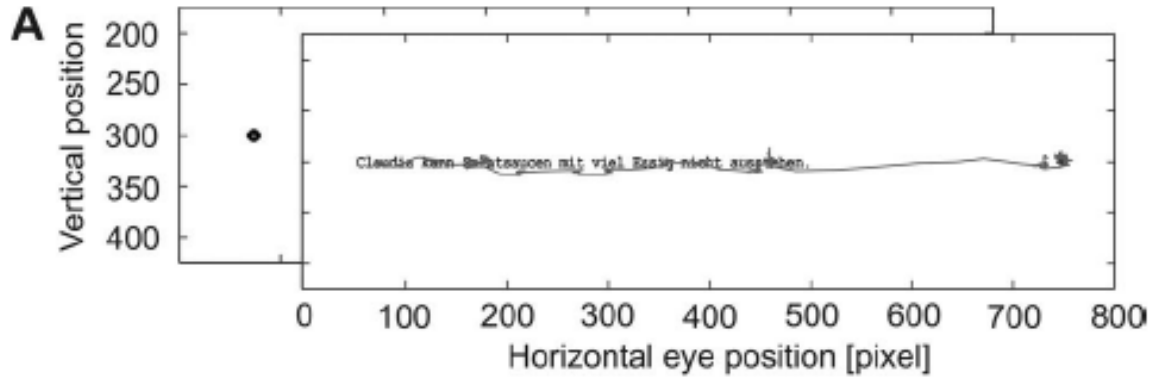
Factorial designs may not always be feasible, e.g. in naturalistic paradigms.



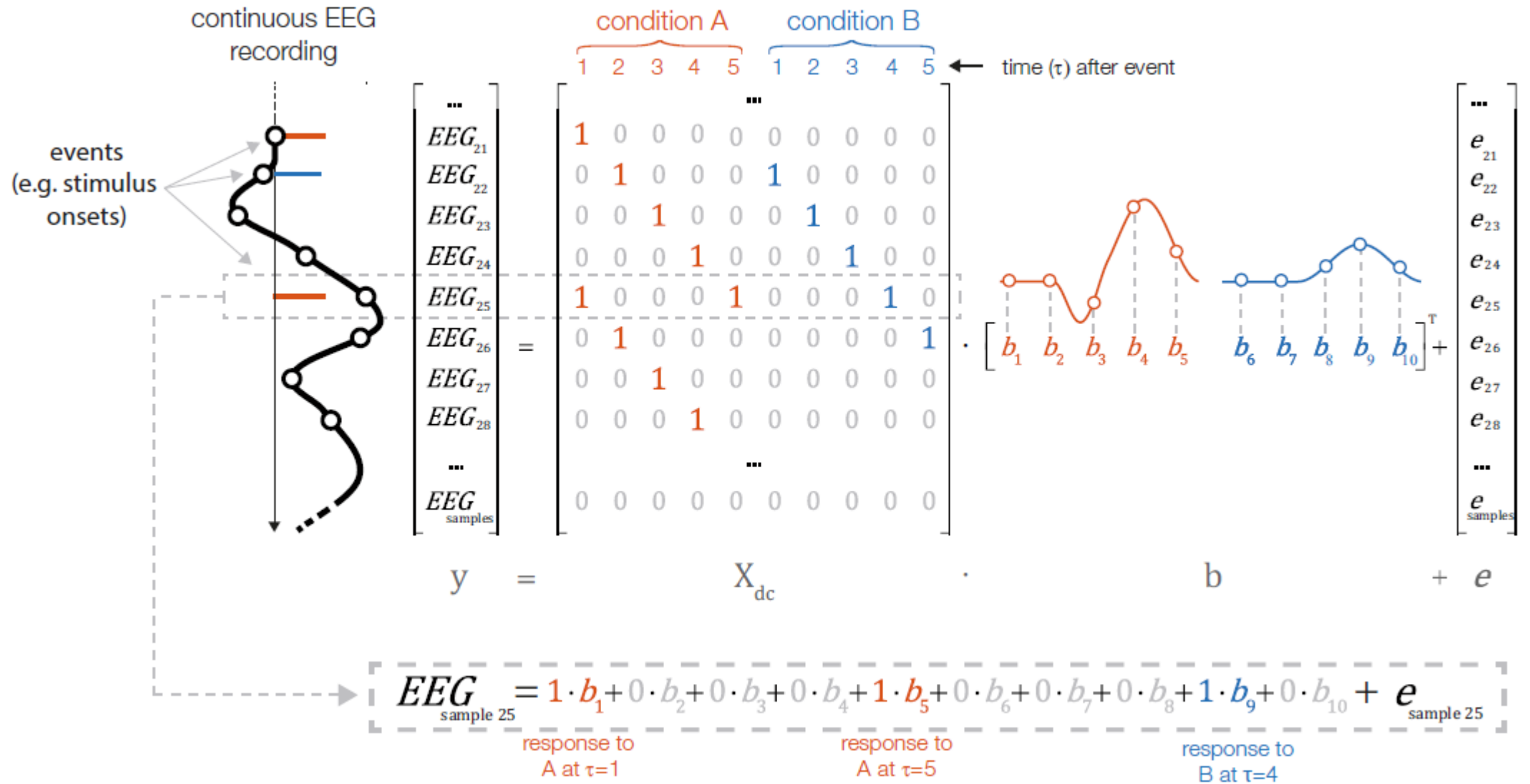
# The Multivariate Temporal Response Function (mTRF) Toolbox



# EEG with eye movements



# Deconvolution of EEG signals – UNFOLD toolbox



## Unfortunately...

1. Polarity of effects (betas) is harder to infer when applied to signed data (e.g. ERPs). For example, a positive beta value for a negative ERP (e.g. “N1 peak”) reflects smaller amplitude with increasing values.
2. Most higher-level analysis methods such as connectivity and decoding are (currently) designed for factorial designs.





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**Thank you**

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