

# Single-trial detection of EEG/ERP brain responses

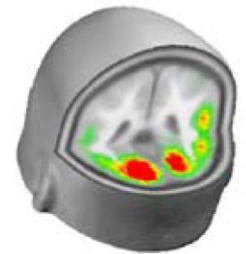
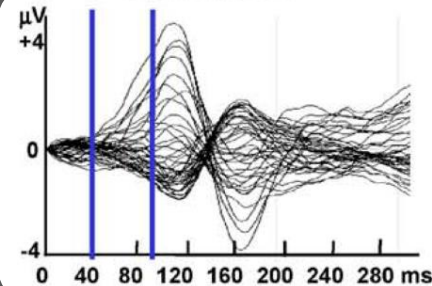
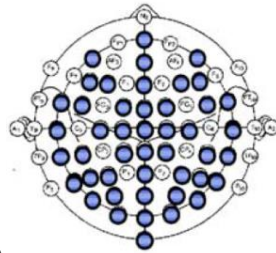
1. Basics on EEG and ERPs
2. Rationale for estimating brain responses trial-by-trial
3. Approaches for single-trial estimation
4. Possible applications

# Basics concepts of functional neuroimaging: EEG vs BOLD-fMRI

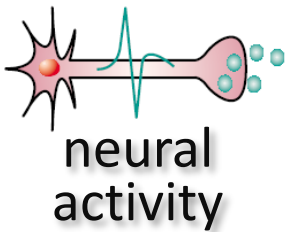
Scalp EEG



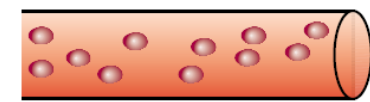
46 electrodes



sensory stimulus

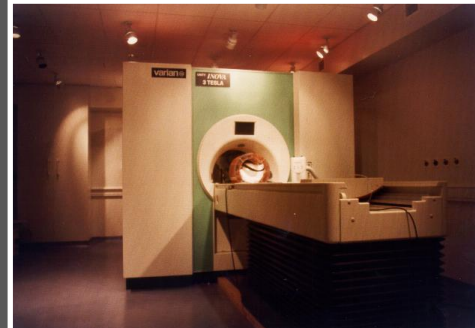
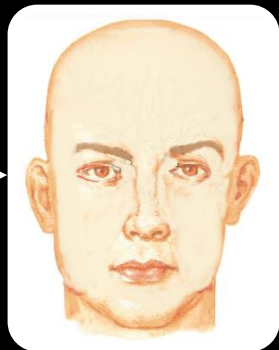
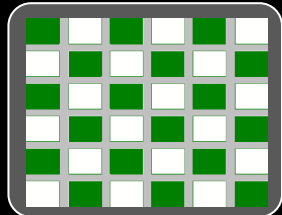


neurovascular coupling



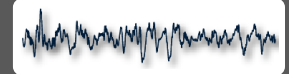
haemodynamic response

BOLD fMRI



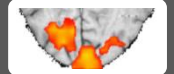
# EEG vs BOLD-fMRI: things to remember

## Scalp EEG



- Samples **neural activity directly**
- **Excellent temporal resolution** (order of ms)
- Reasonable spatial resolution (~5 mm: but depending on several factors)
- Need of a priori-hypotheses (source numbers and locations)
- Some experimental constraints (e.g. stimulus features)

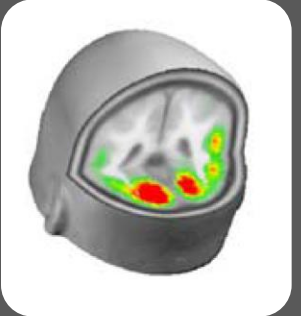
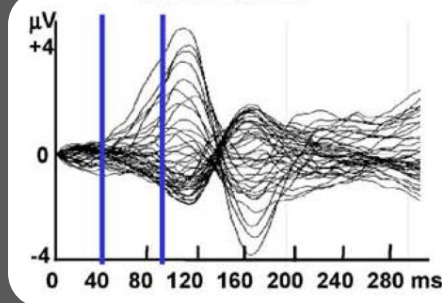
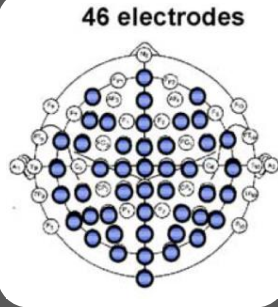
## BOLD fMRI



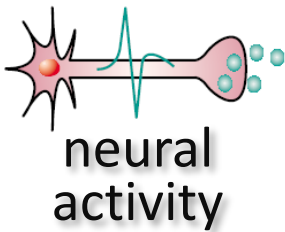
- Samples **consequences of neural activity** (or other phenomena with PET)
- Extremely **low temporal resolution** (hundreds of ms to several seconds)
- **Good spatial resolution** (2-10 mm; but limitations due to signal nature)
- **No need of a priori-hypotheses** (source numbers and locations)
- More flexible stimulation paradigms
- Some experimental constraints due to the scanner environment

# Basics concepts of functional neuroimaging: EEG vs BOLD-fMRI

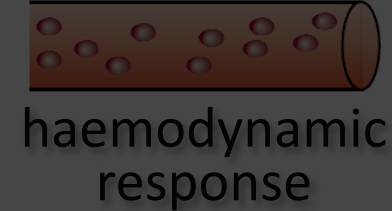
Scalp EEG



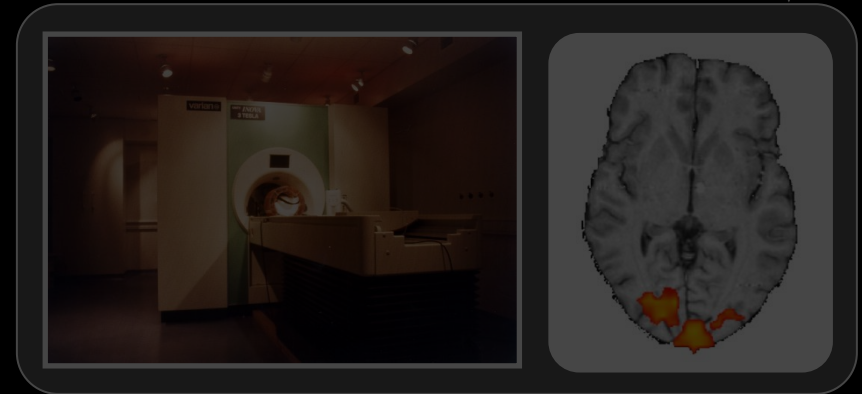
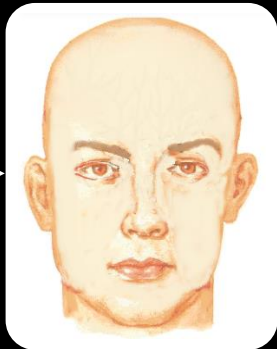
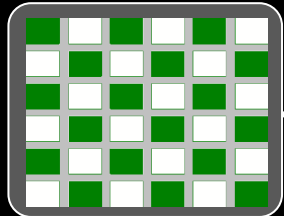
sensory stimulus



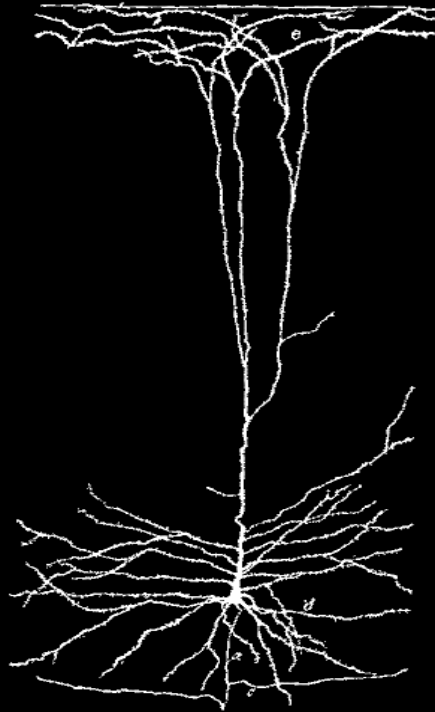
neurovascular coupling



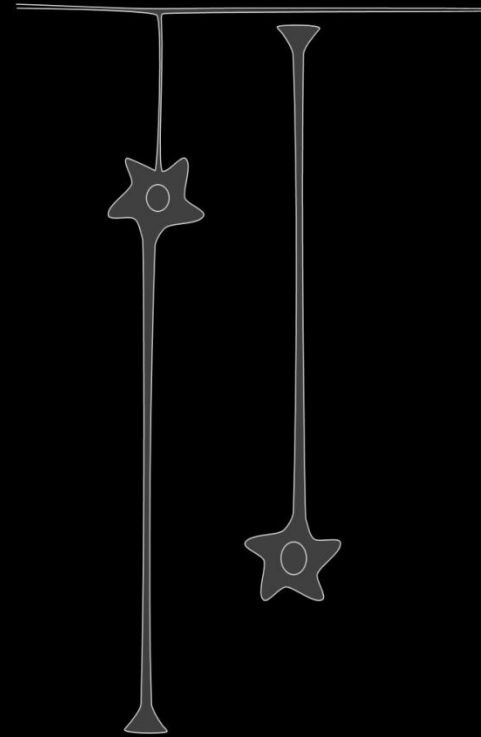
BOLD fMRI



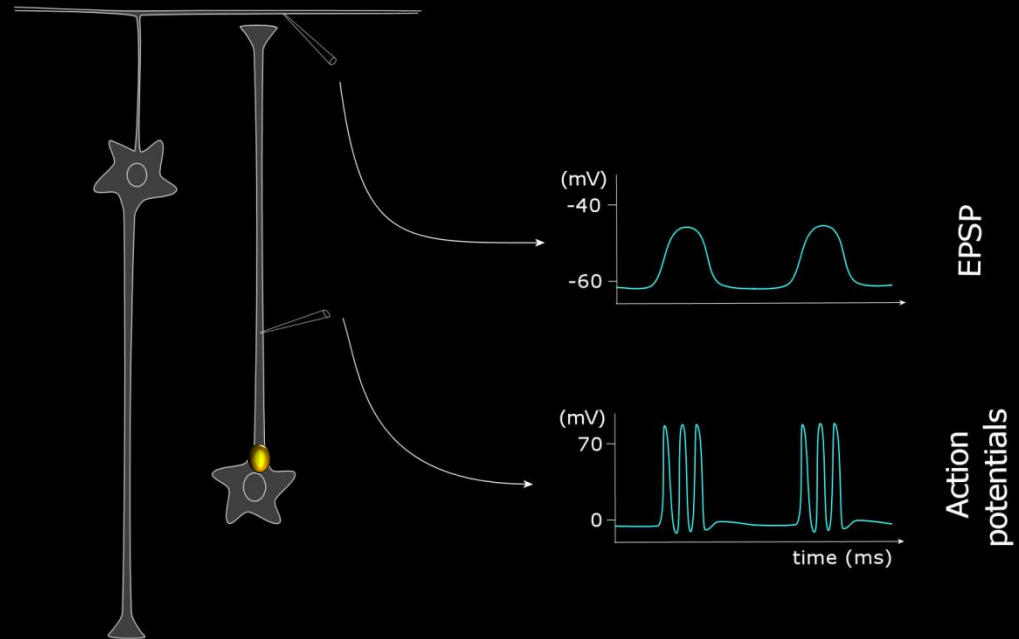
# the neocortical pyramidal neuron



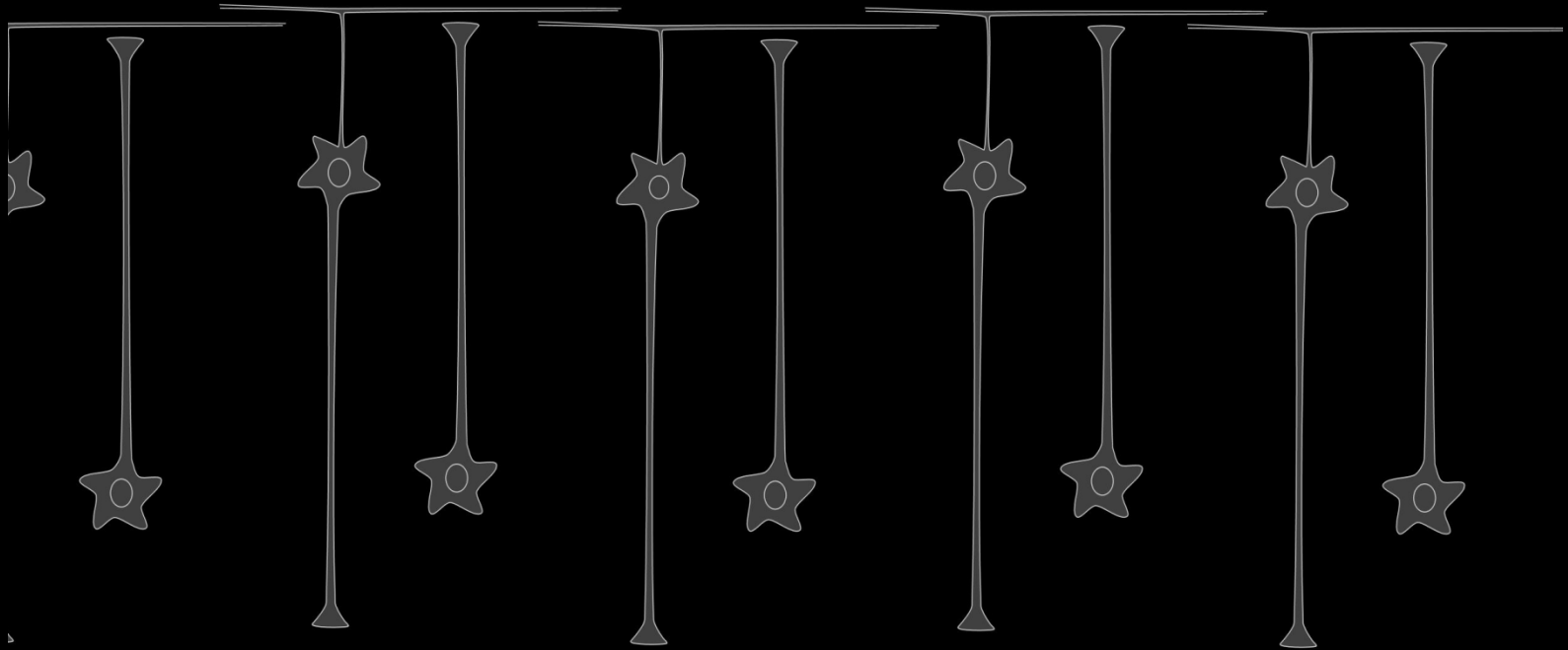
(adapted from Ramon Y Cajal, 1905)



# basics of cortical electrophysiology

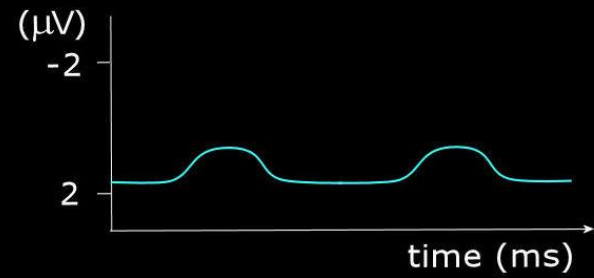


# cortical architecture

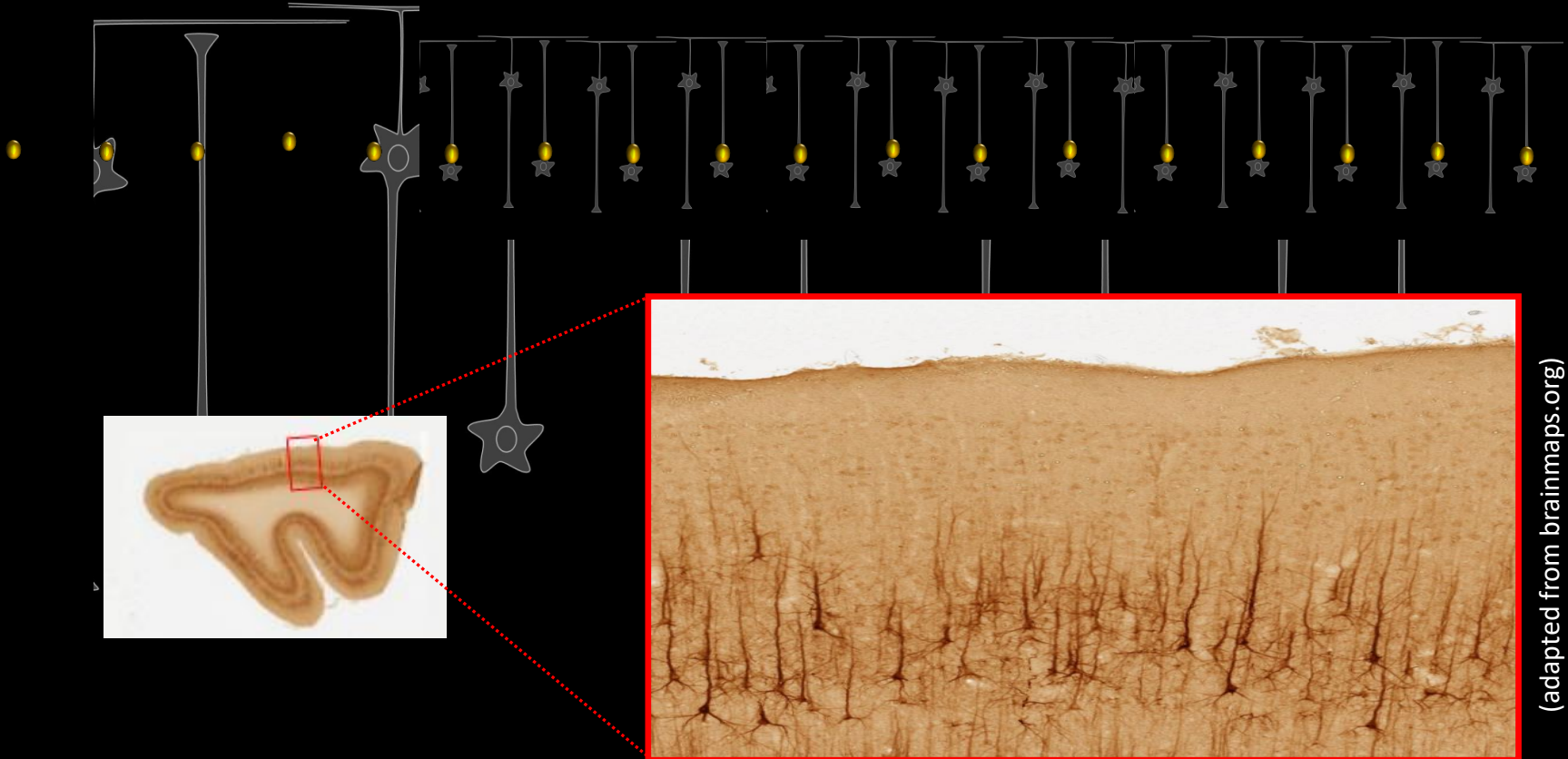


# scalp EEG

SCALP ELECTRODE



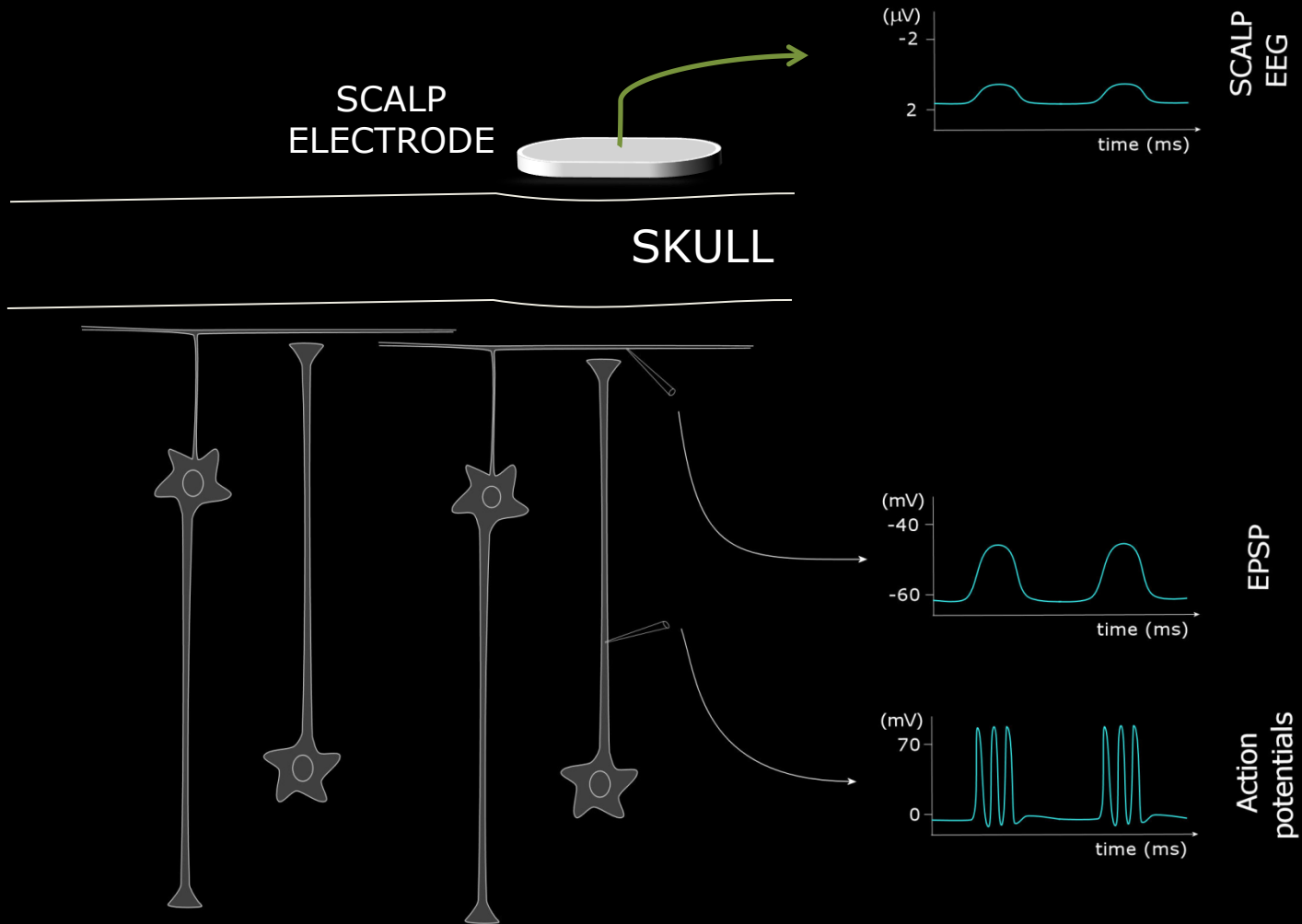
SKULL



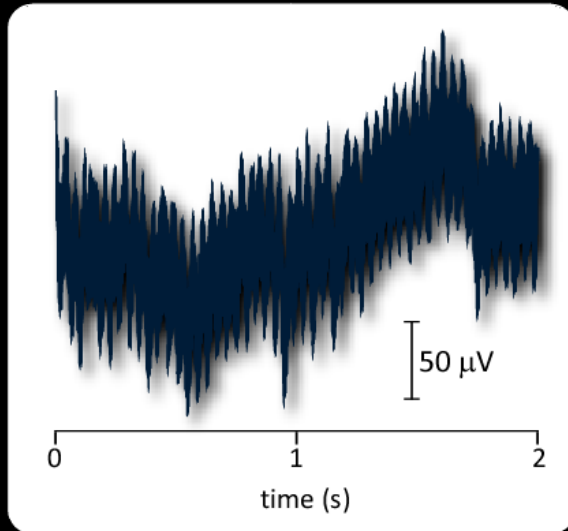
(adapted from brainmaps.org)



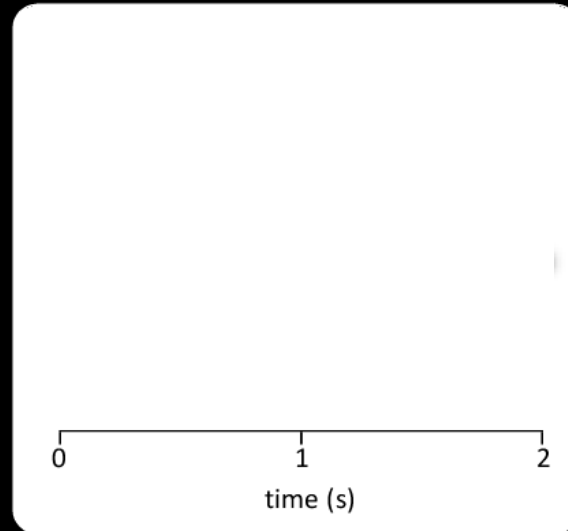
# scalp EEG vs direct cortical recording: magnitude differences



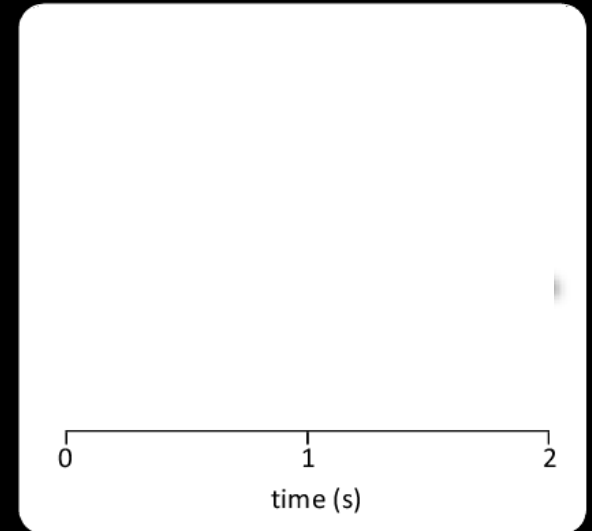
# Event-related EEG potentials (EPs) – Basics of filtering



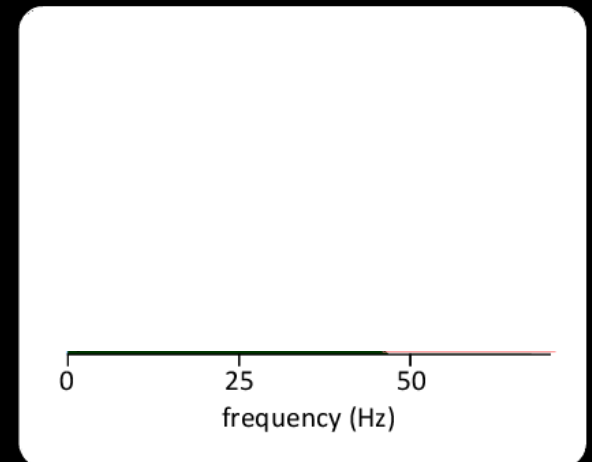
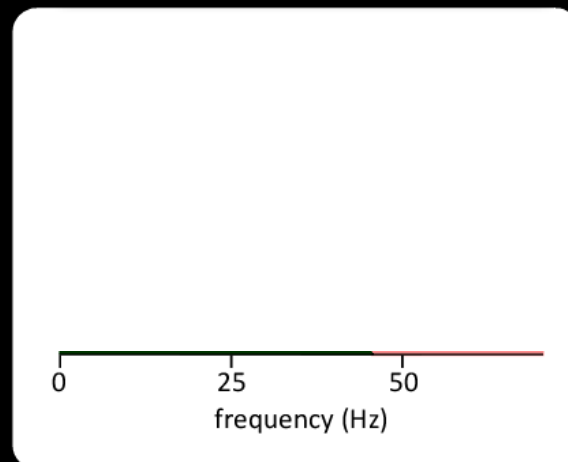
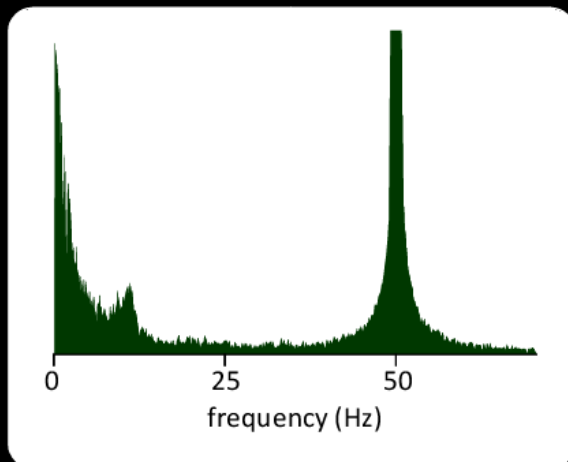
original EEG signal



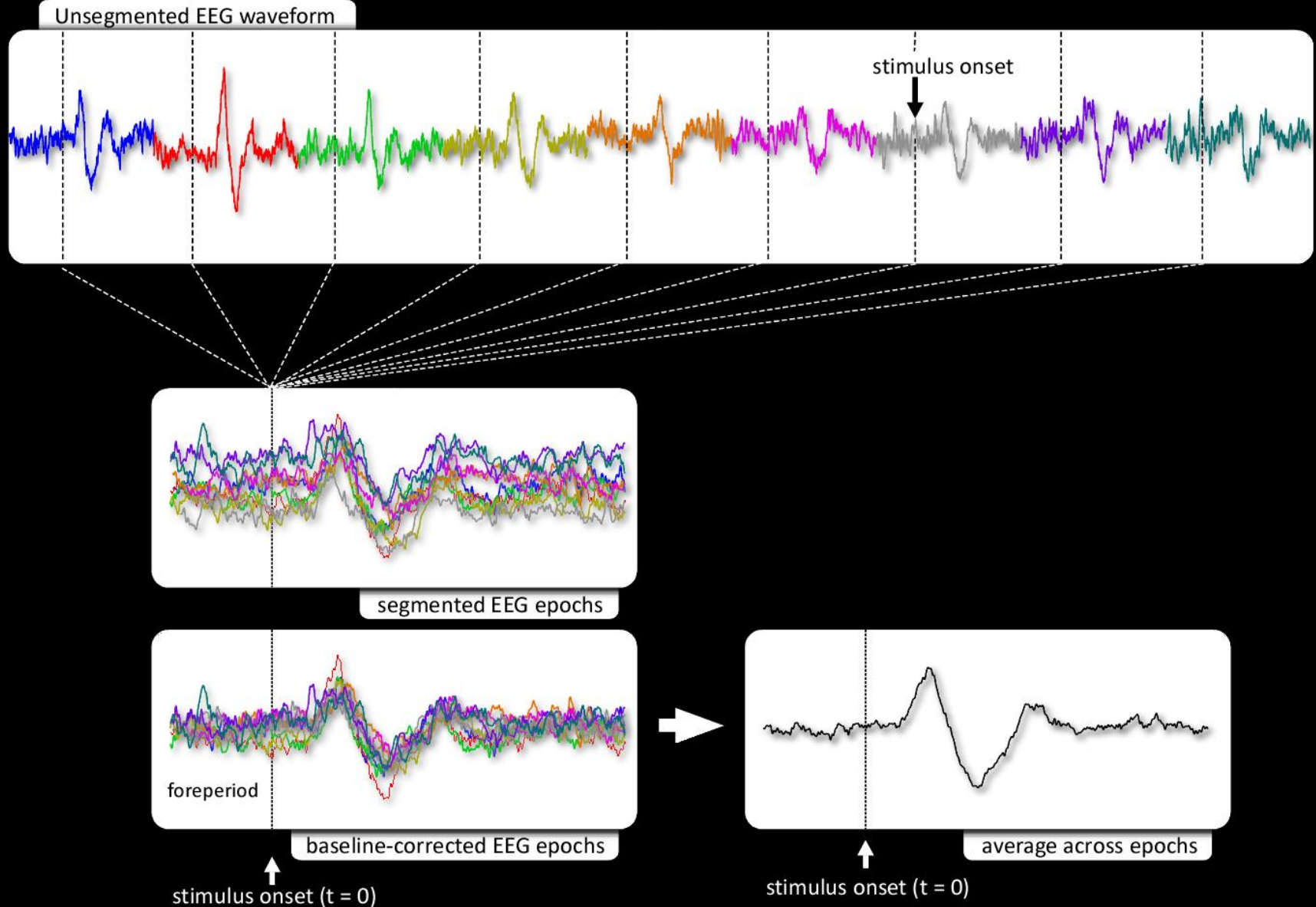
low-pass filter (cut-off = 40 Hz)



high-pass filter (cut-off = 1 Hz)



# Event-related EEG potentials (EPs) – Basics of averaging



## Take home messages – across-trial averging

1.

2.

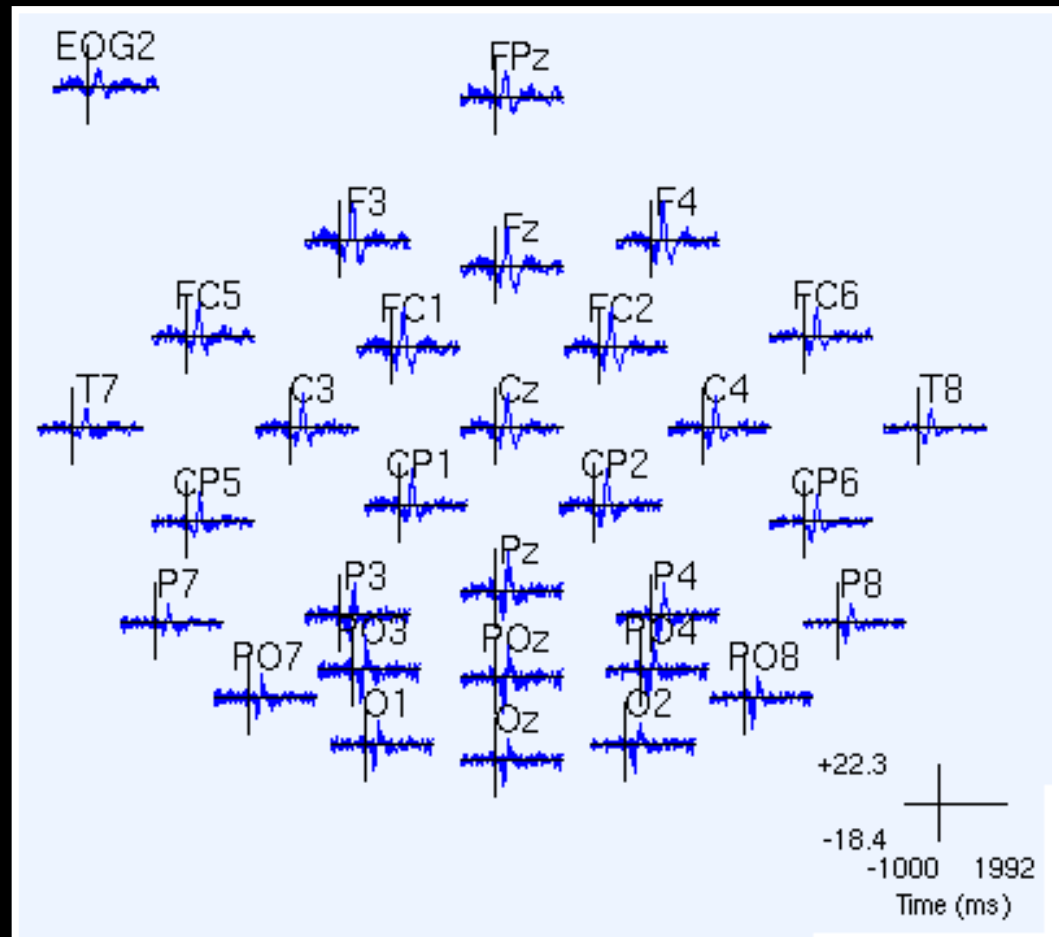
3.

4.

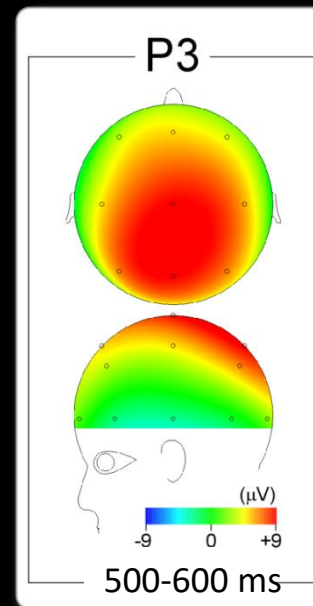
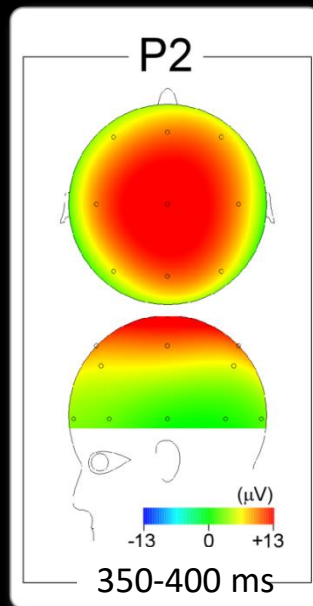
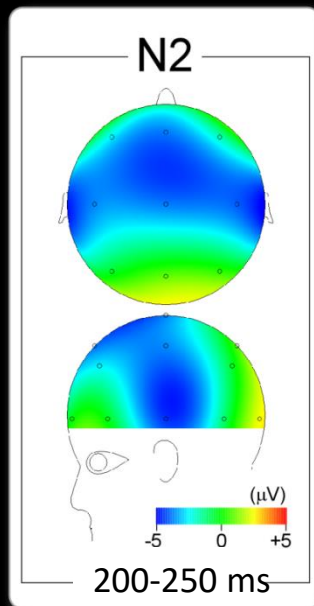
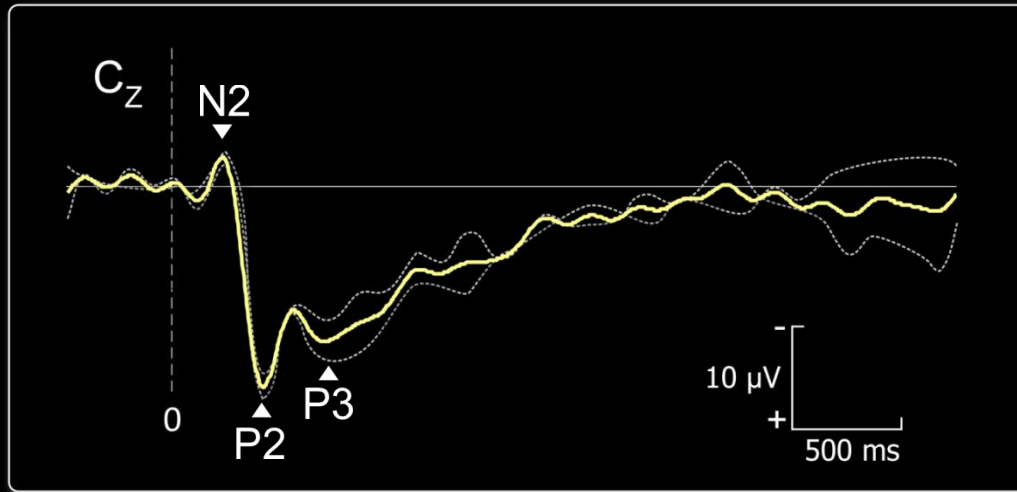
5.

# Event-related EEG potentials (EPs) – Multichannel recording

When more than one electrode is used, the average EPs can be plotted according to the relative position of each electrode on the head, thus providing some spatial information on potential distribution.



# Event-related EEG potentials (EPs) – Overview



Peaks are characterised by their latency, polarity, amplitude and scalp topography

# source analysis of event-related EEG potentials (EPs)

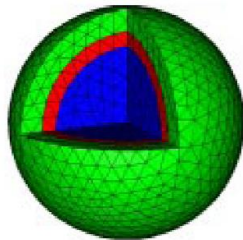
## The “forward problem”

The “forward problem” is well defined and has a unique solution.

For a given brain electric source distribution and a given head volume conductor, the “forward problem” determines the source-generated electric field.

## The “inverse problem”

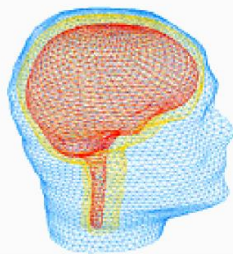
The “inverse problem” is ill-posed.  
For a given set of electrode potentials, the “inverse problem” determines the source distribution.



### Spherical models

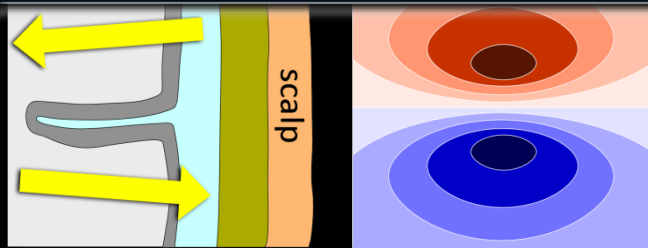
The volume conductor is modelised using a series of concentric spheres.  
The conductivity of each layer is considered homogeneous

- 3 layer model : skin, skull, brain
- 4 layer model : skin, skull, CSF, brain



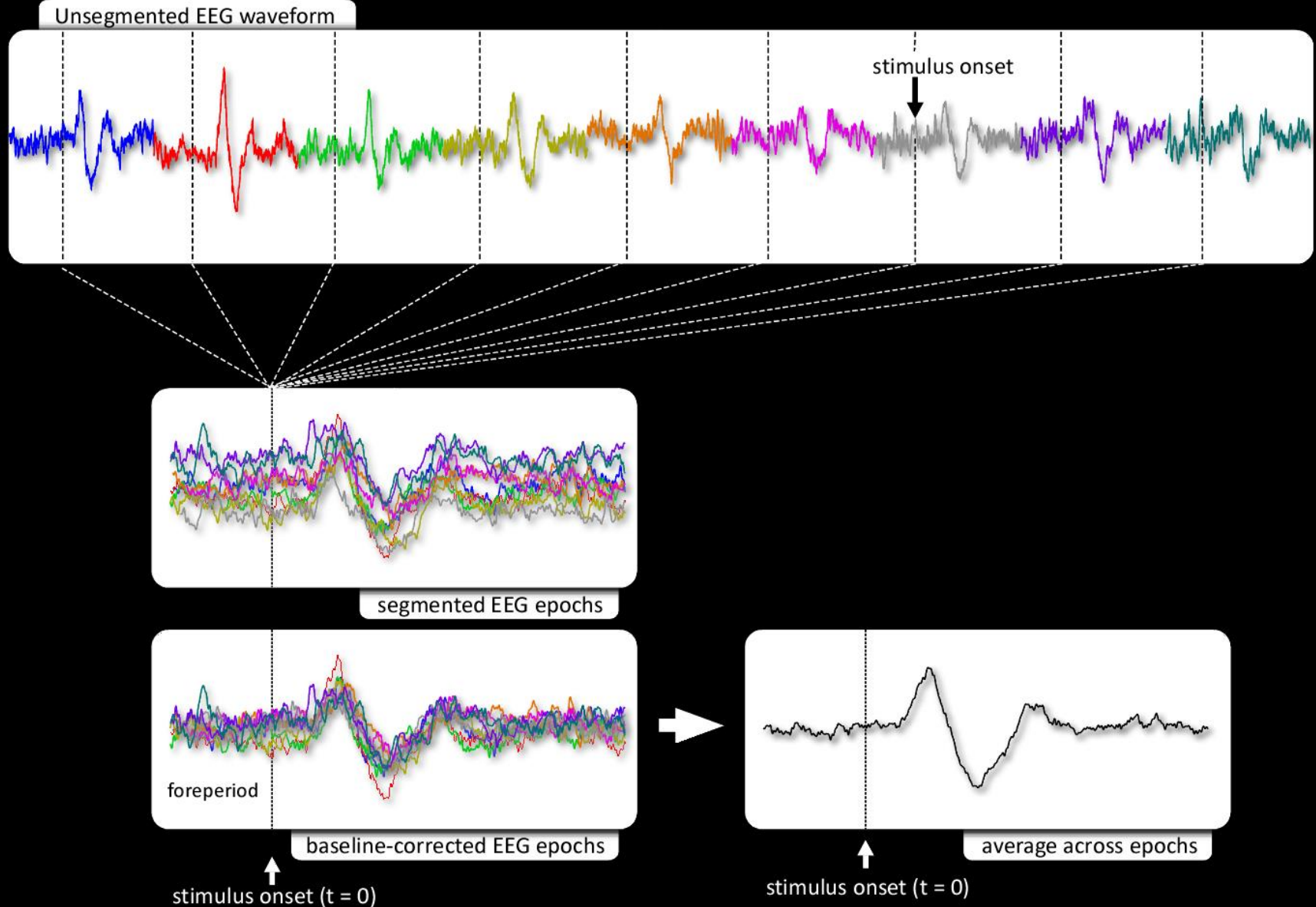
### Realistic head models

The volume conductor is built using individual MRI structural scans  
Accounts for the complexity and interindividual variations of geometry



problem”

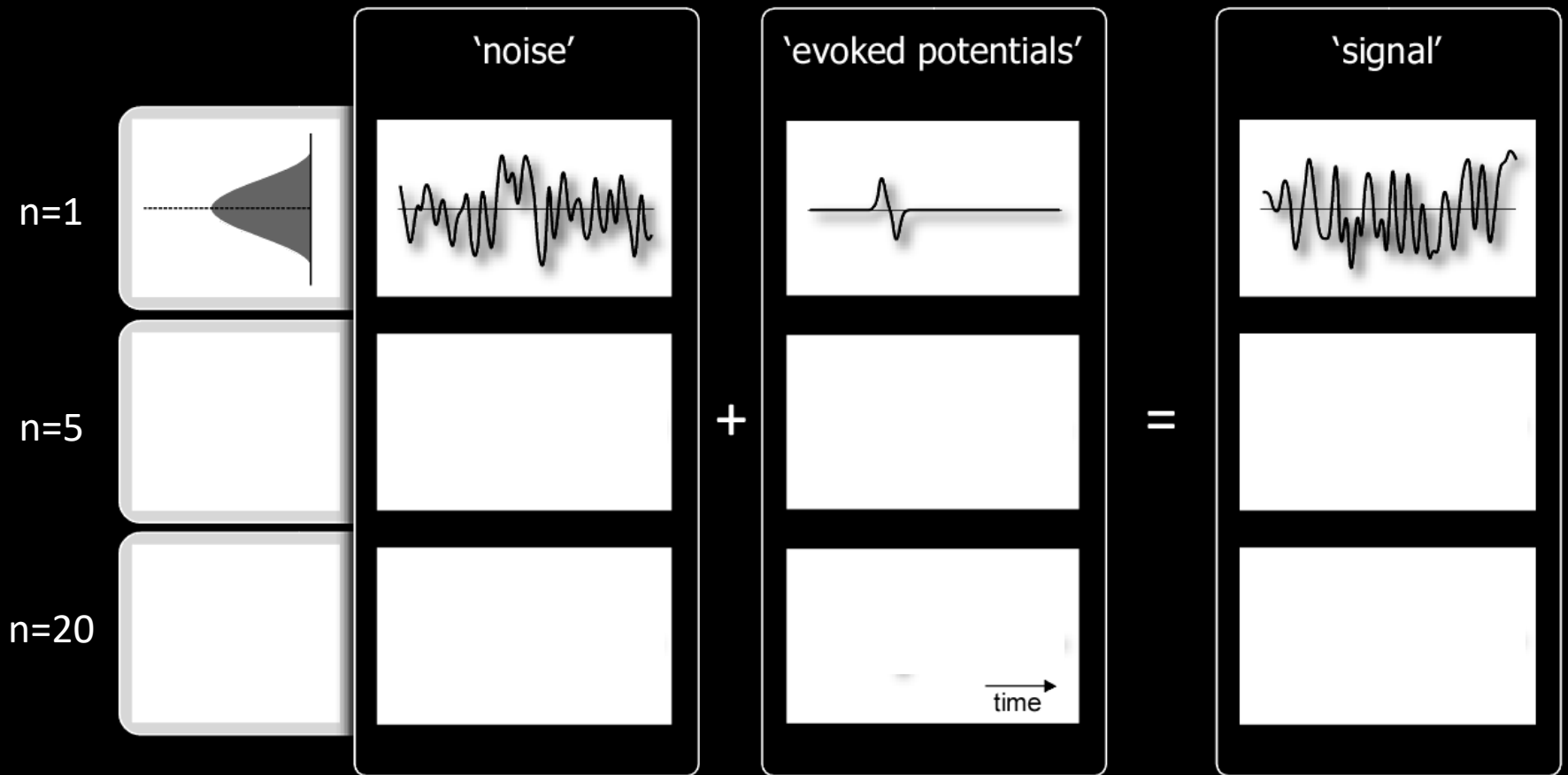
# Event-related EEG potentials (EPs) – Basics of averaging



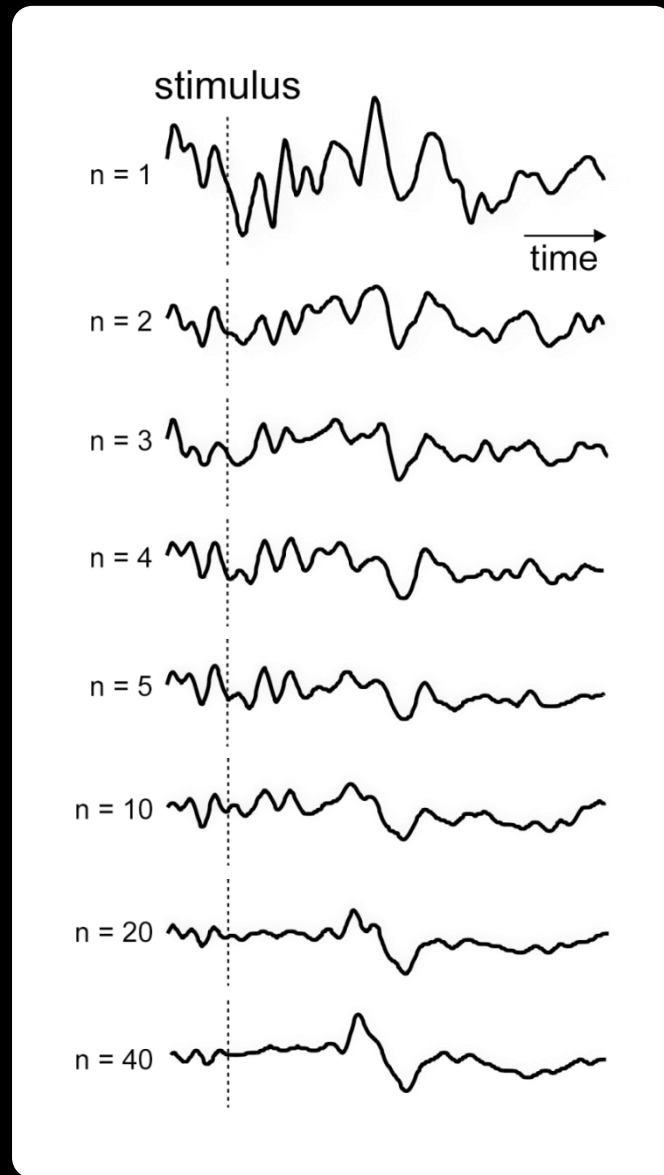


# Cross-trial averaging and the additive noise model

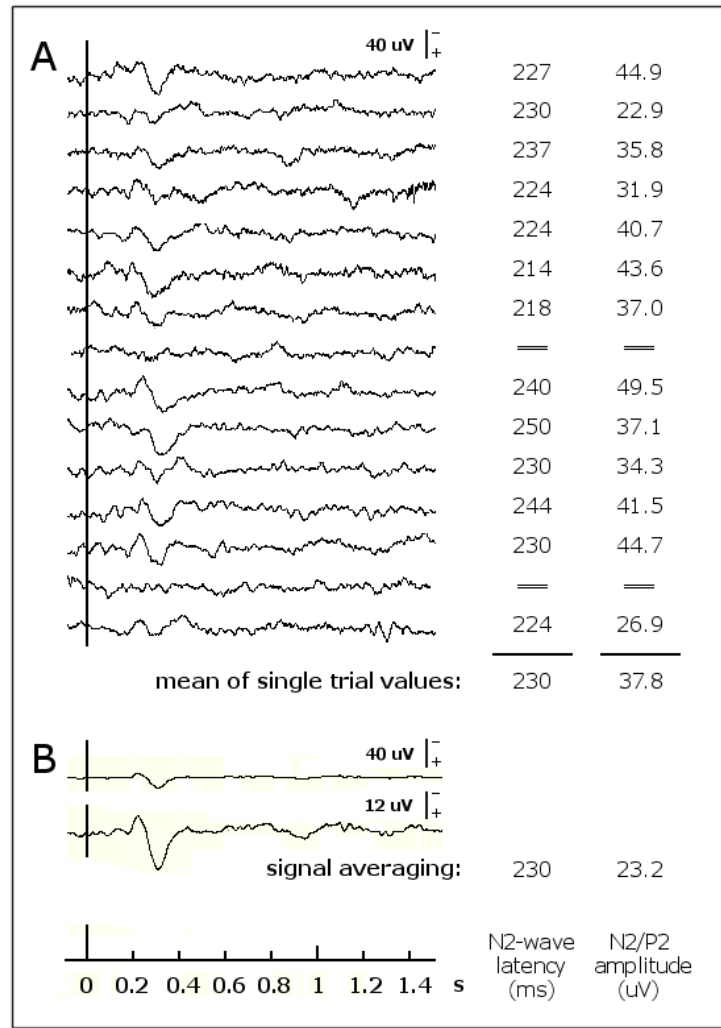
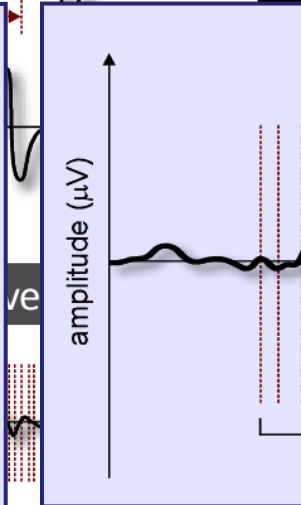
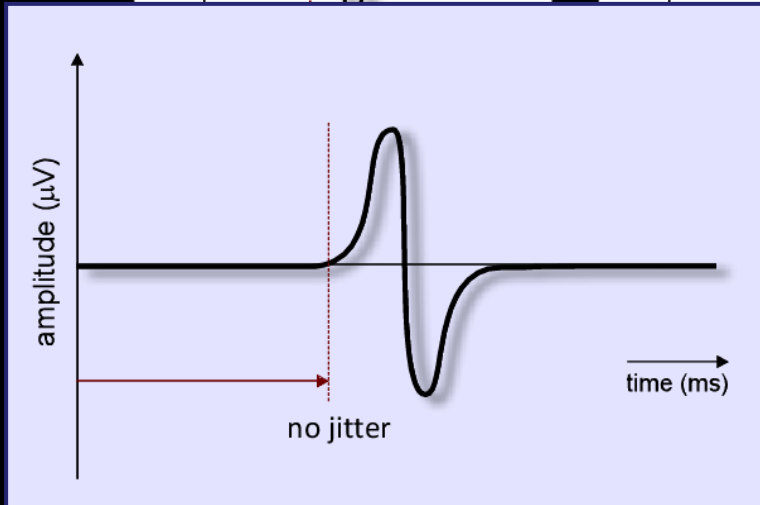
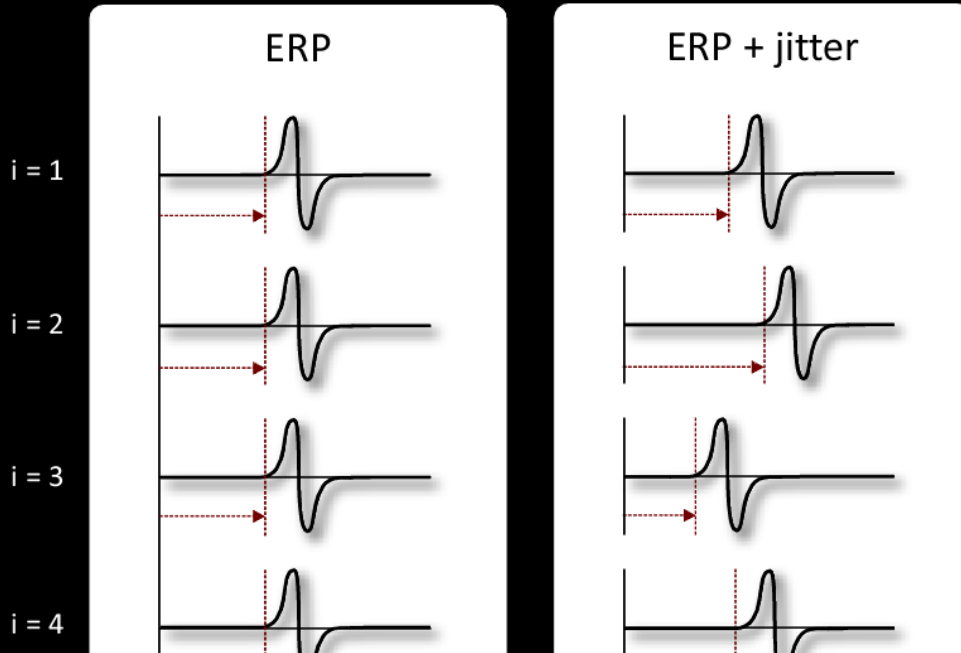
## The "additive-noise" model



# Cross-trial averaging and the additive noise model

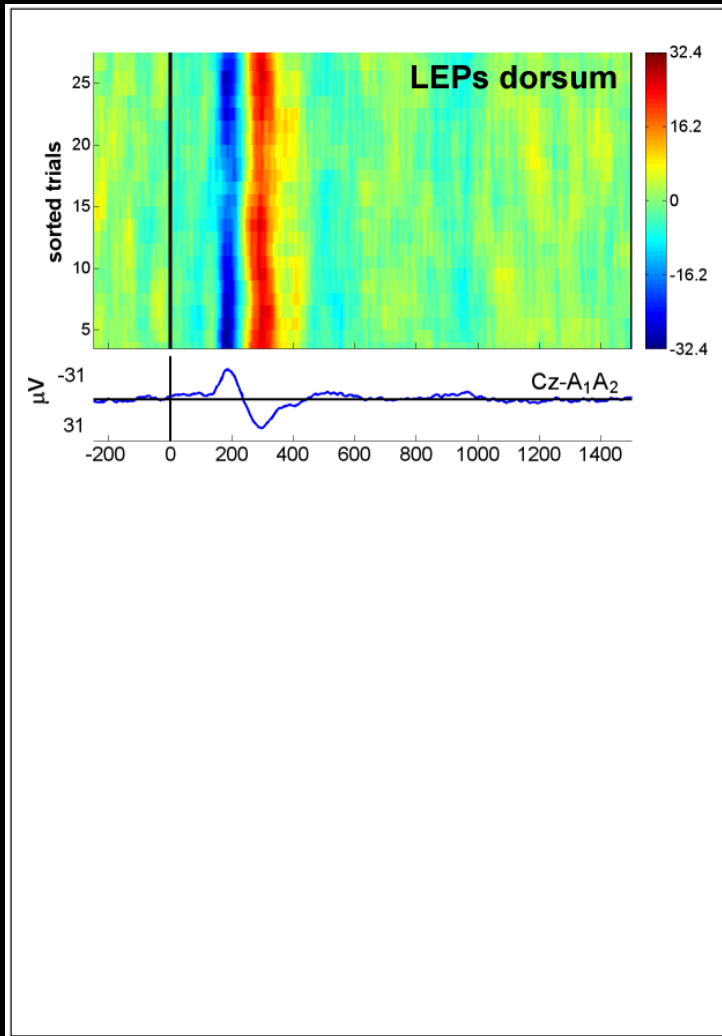


# problem #1: temporal jitter



(Iannetti et al, *Neuroscience* 2005)

# problem #1: temporal jitter



(Iannetti et al., *J Physiol* 2006)

When jitter at single-trial level is different between conditions, traditional averaging could reveal **spurious differences**.

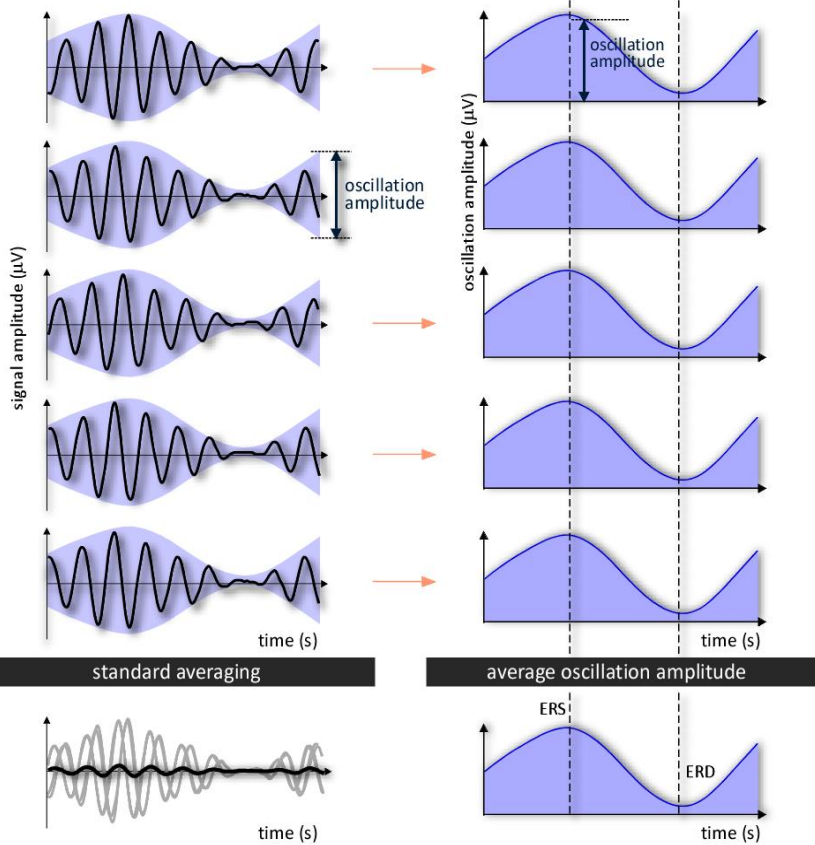
These differences are a potentially important **confounding factor** that **must be taken into account** to allow safe inference of physiological results.

In order to rule out any contribution of jitter-dependent amplitude reduction of standard averages, all responses should be **analysed at single-trial level**

# problem #3: 'phase resetting'

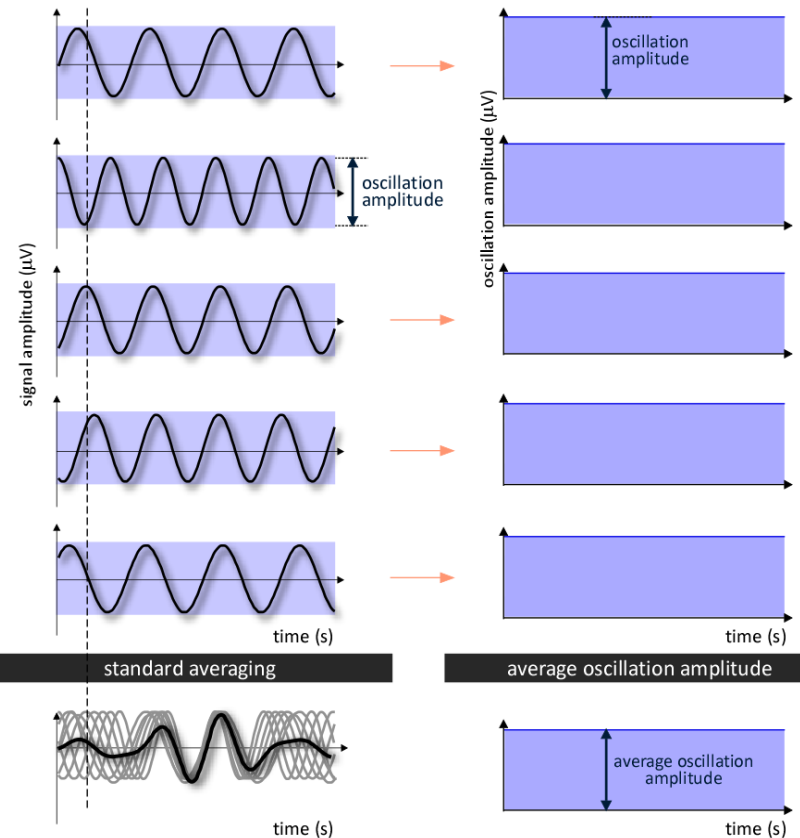
Solution: compare stimulus-evoked amplitude changes in the average vs single trials.

ERD/ERS:



Transient change **in amplitude!**

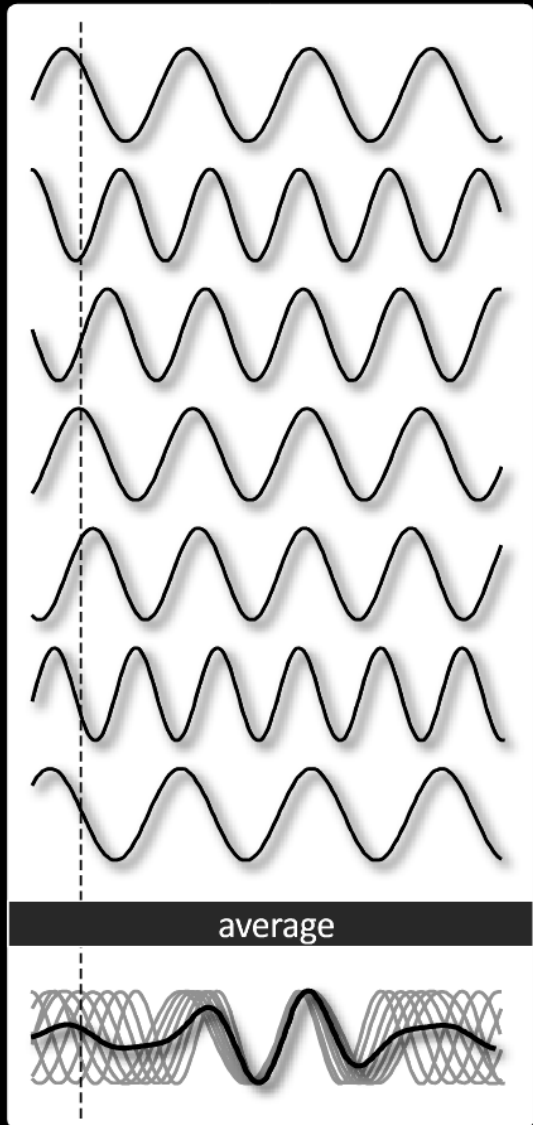
Pure phase-resetting:



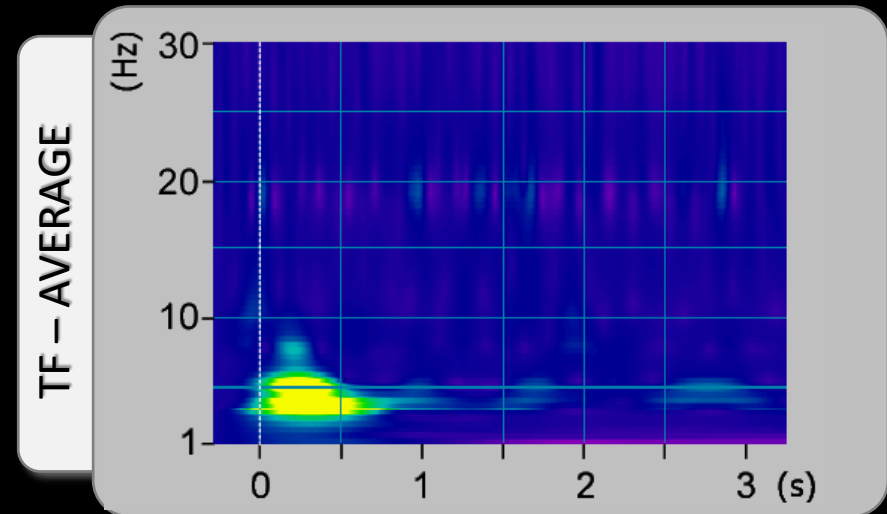
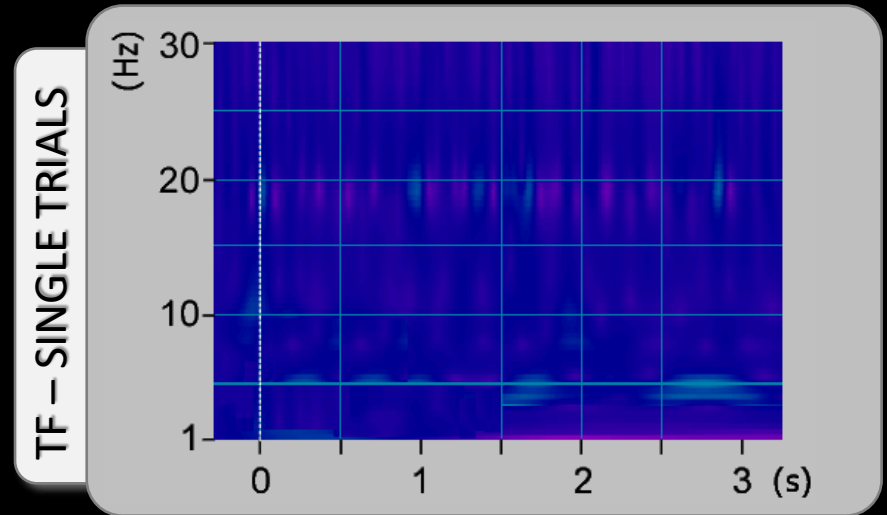
Transient change **in phase!**

# problem #3: 'phase resetting' vs..

Solution: compare stimulus-evoked amplitude changes in the average vs single trials.



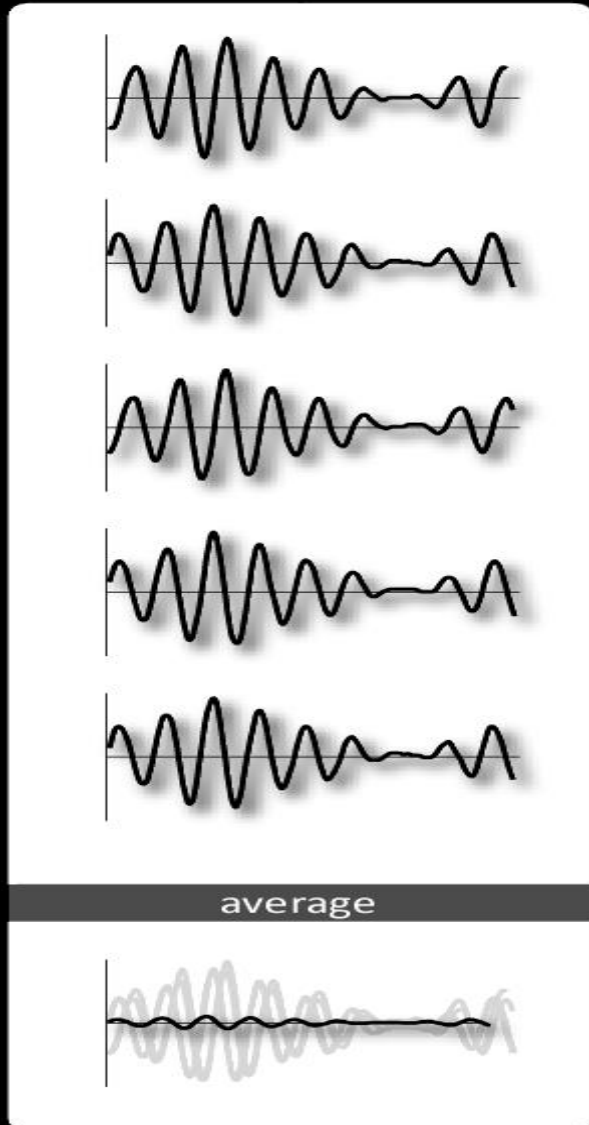
Pure phase resetting:



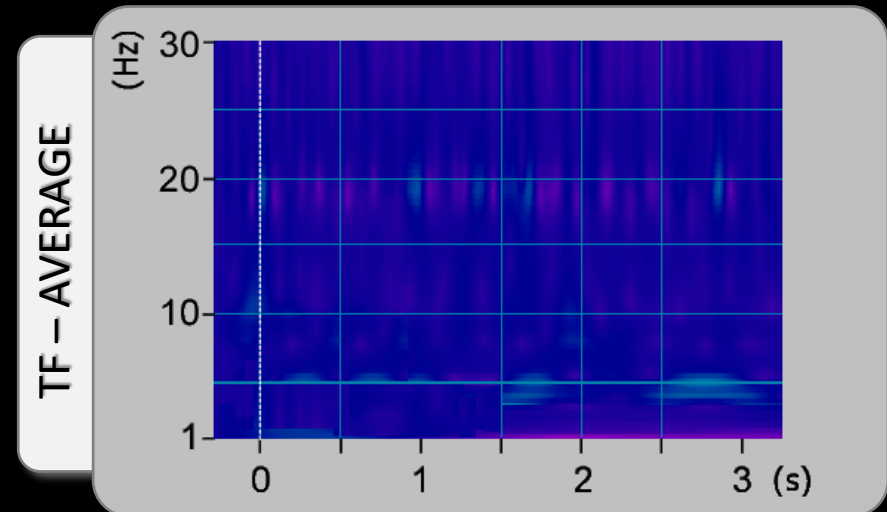
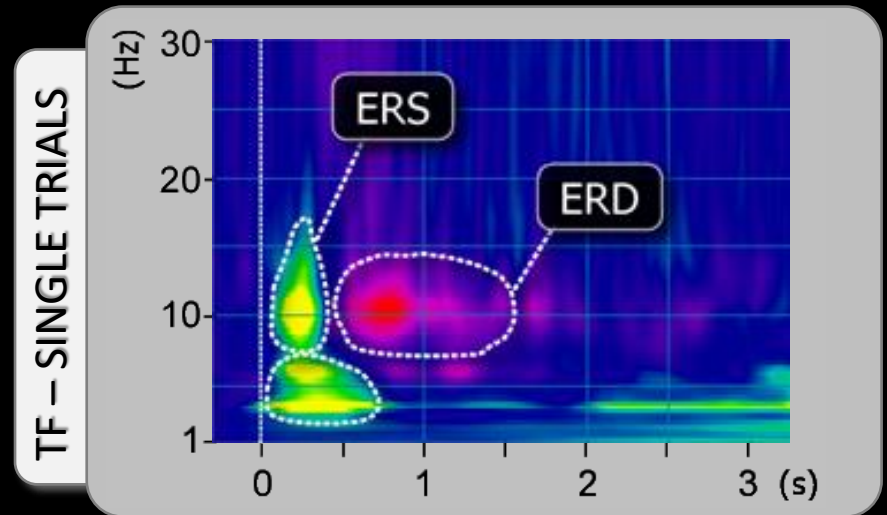
Transient change in phase!

## ... vs problem #2: ERD/ERS

Solution: compare stimulus-evoked amplitude changes in the average vs single trials.

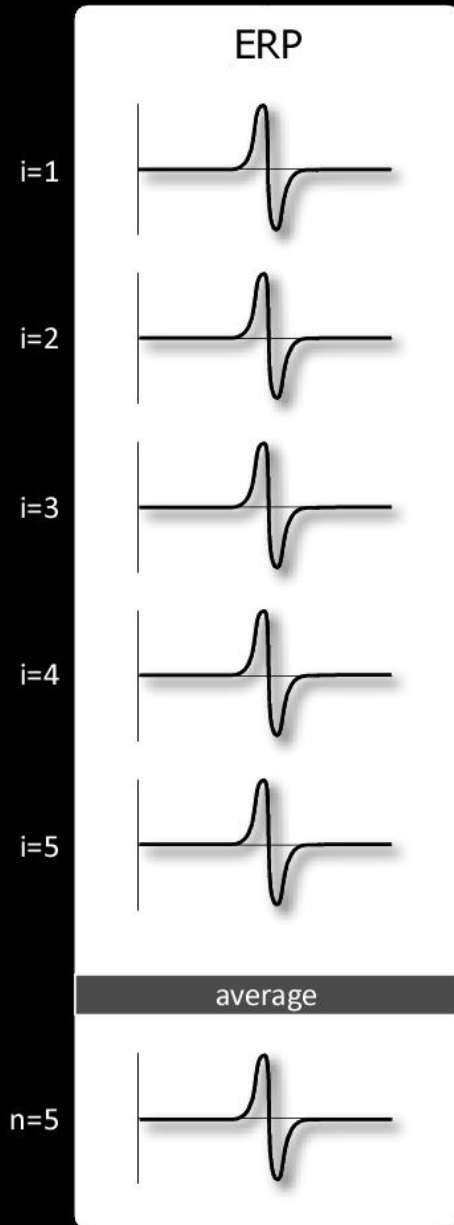


Pure ERD/ERS:



Transient change **in amplitude!**

# standard averaging: summary of limitations...





## Take home messages – across-trial averaging

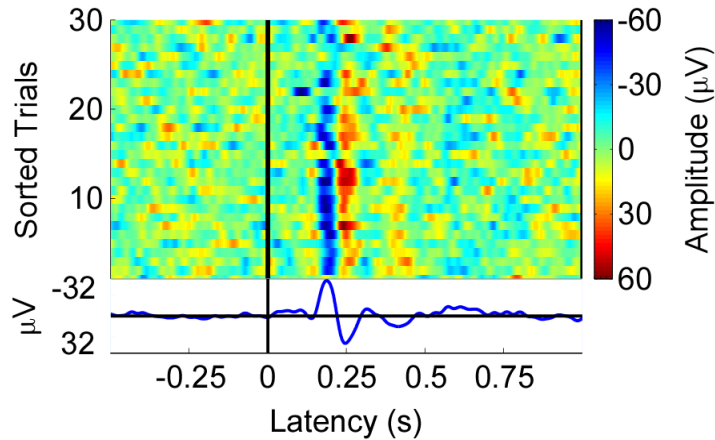
1. The magnitude of ERPs is often several factors smaller than the magnitude of the background electroencephalogram.
2. Across-trial averaging is a widely-used approach to enhance the signal-to-noise ratio (SNR) of both evoked induced EEG responses.
- 3.
- 4.
- 5.

# Wavelet filtering to enhance ERP signal-to-noise ratio

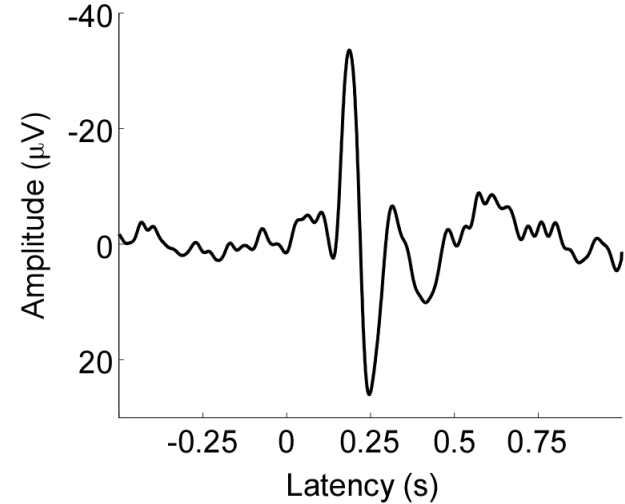
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# Wavelet filtering to enhance ERP signal-to-noise ratio

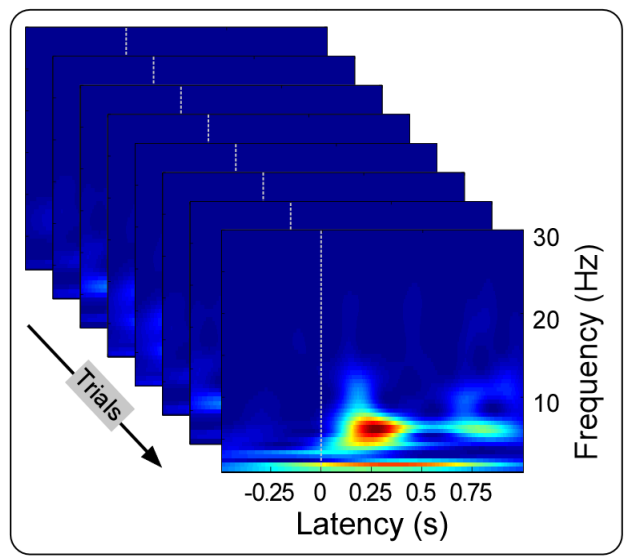
## Single-trial ERPs



Average  $\rightarrow$



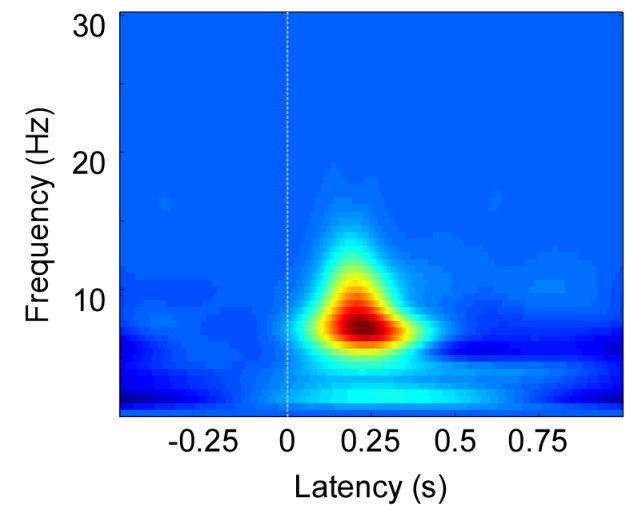
CWT  $\downarrow$  ①



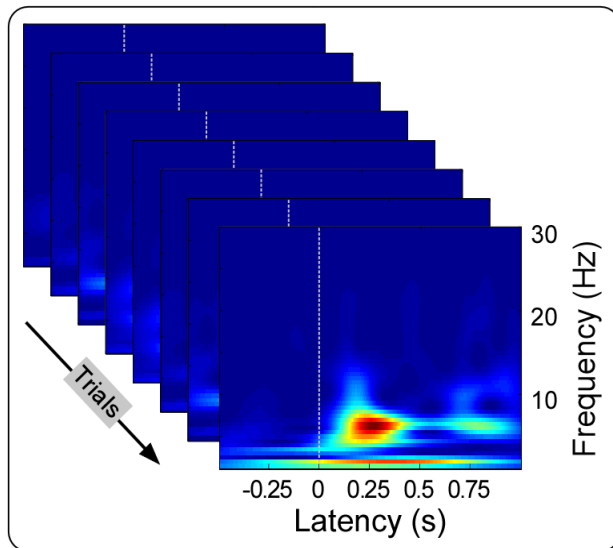
Average  $\rightarrow$

②

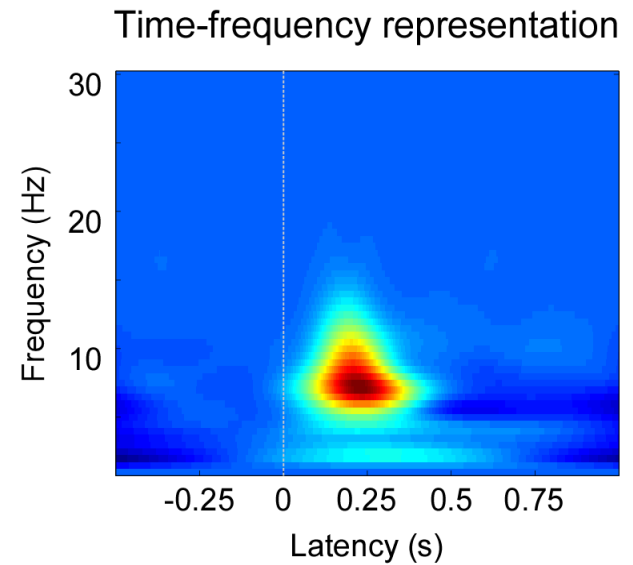
## Time-frequency representation



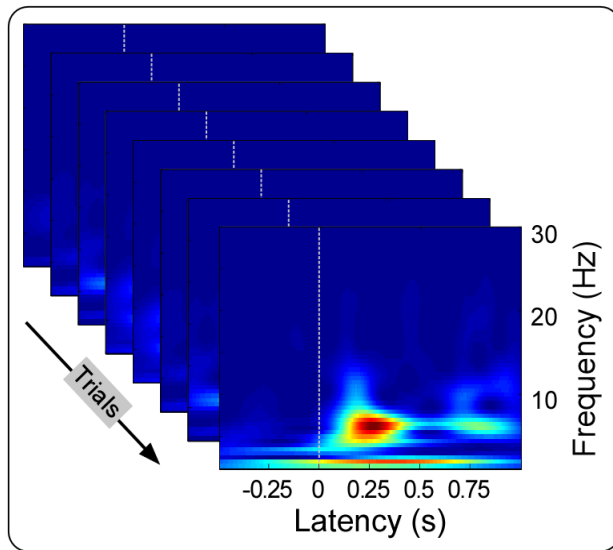
# Wavelet filtering to enhance ERP signal-to-noise ratio



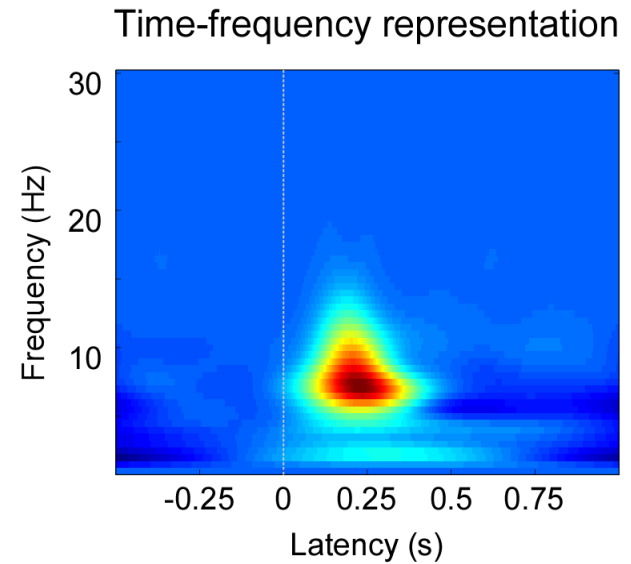
Average  
②



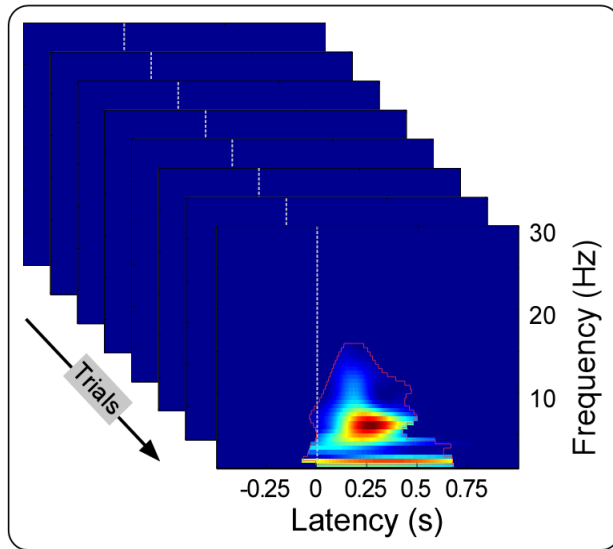
# Wavelet filtering to enhance ERP signal-to-noise ratio



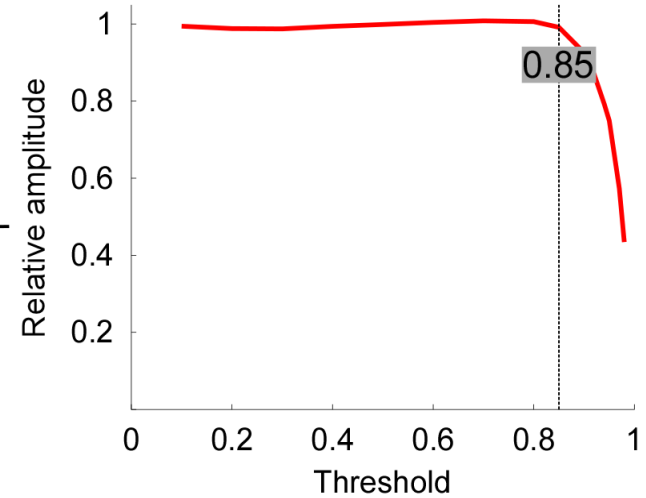
Average  
②



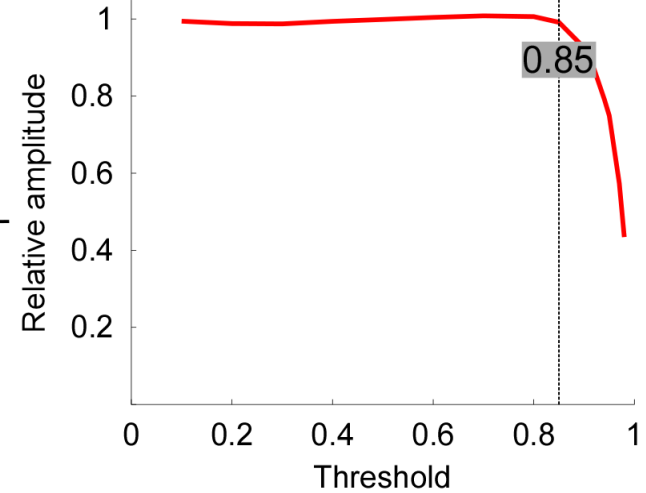
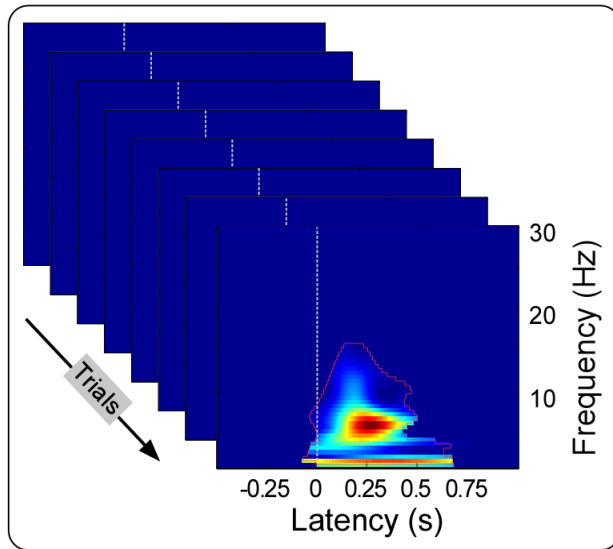
Filtering  
④



③

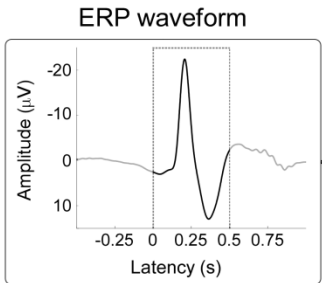


# Wavelet filtering to enhance ERP signal-to-noise ratio



# Multiple linear regression...

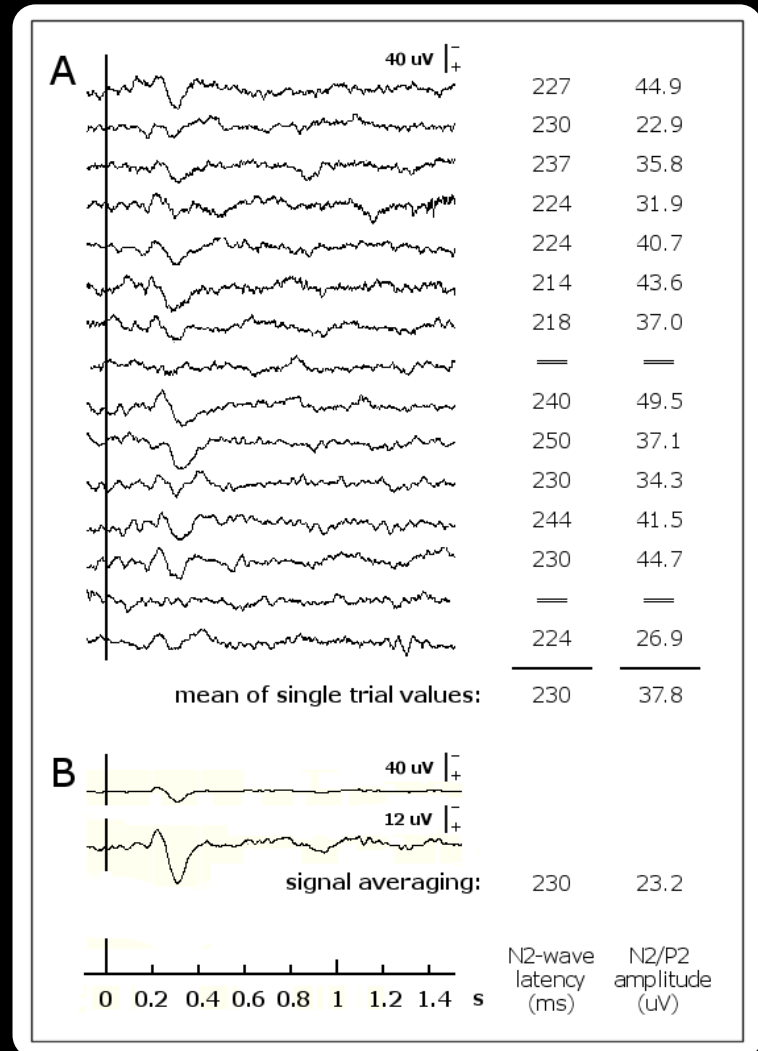
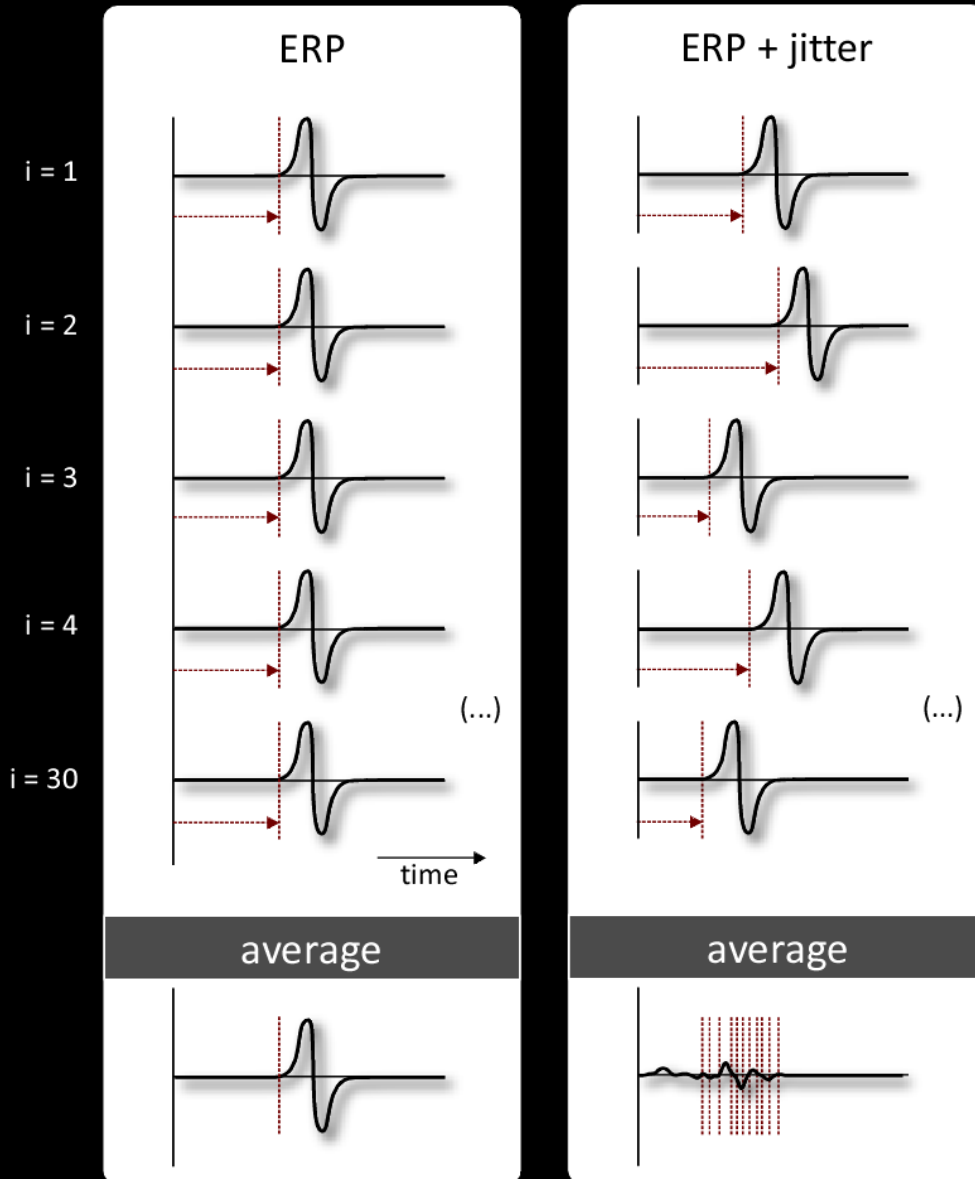
Multiple linear regression (MLR)



(Mahyew et al, *CLINPH* 2006)

Multiple linear regression with dispersion term (MLR<sub>d</sub>)

# Variability of single-trial latency... and morphology

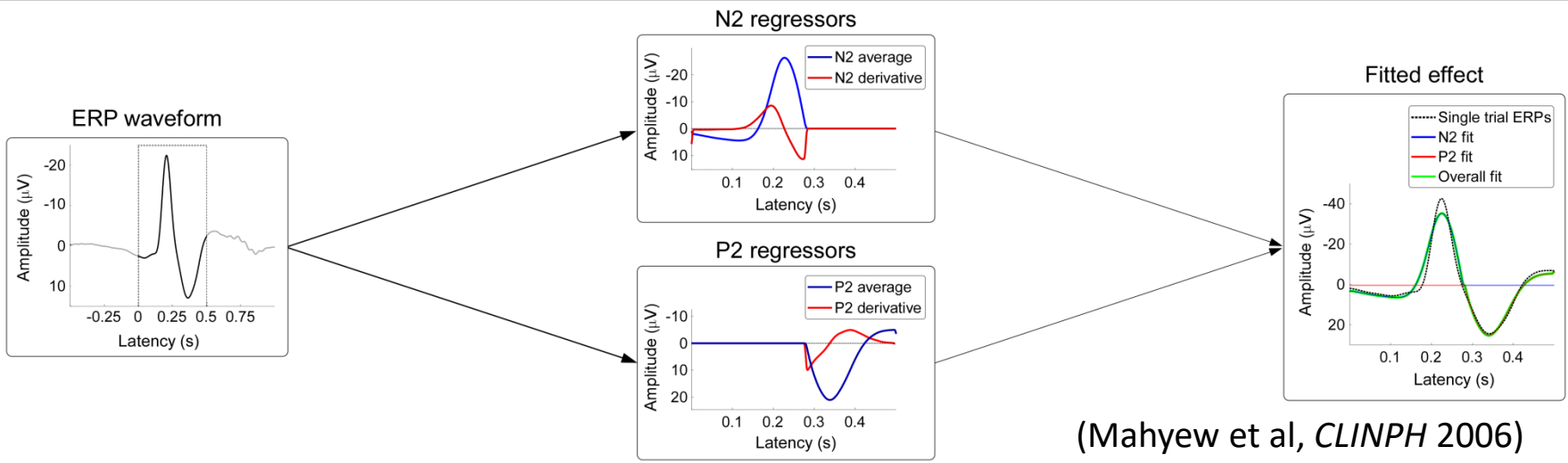


(Iannetti et al, *Neuroscience* 2005)



# Multiple linear regression... with dispersion term

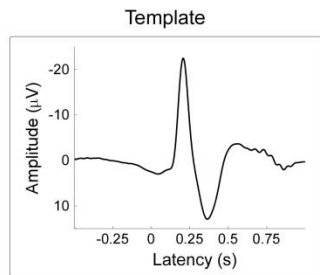
Multiple linear regression (MLR)



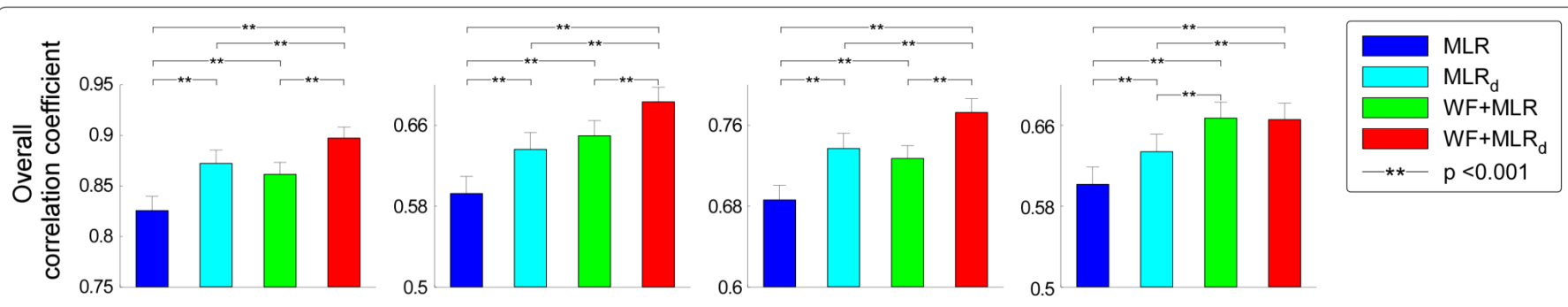
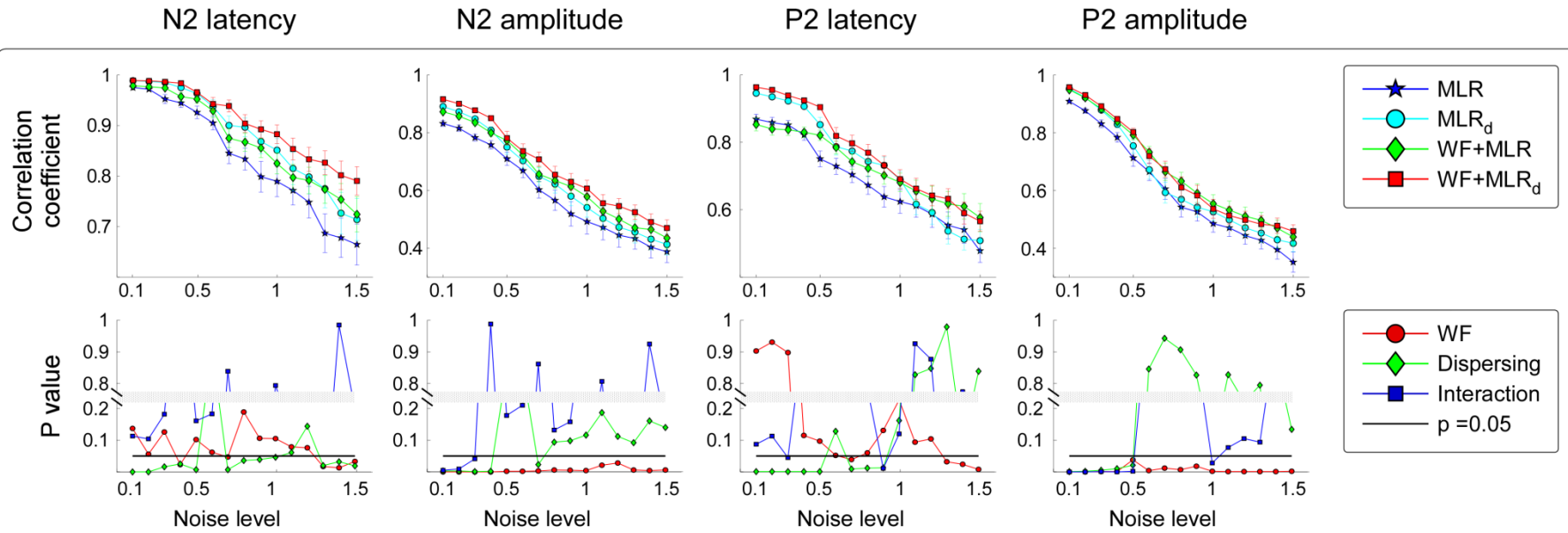
(Mahyew et al, *CLINPH* 2006)

Multiple linear regression with dispersion term (MLR<sub>d</sub>)

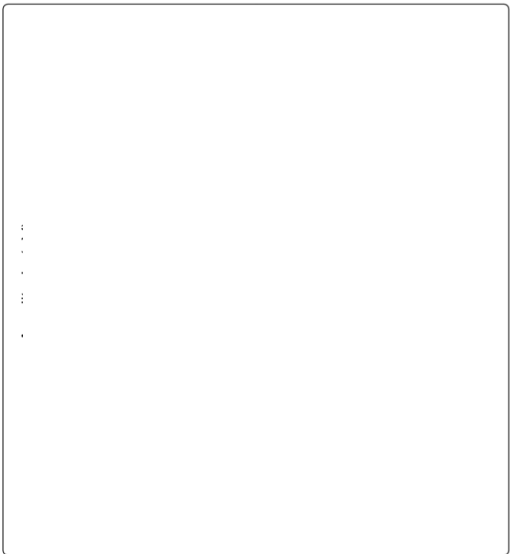
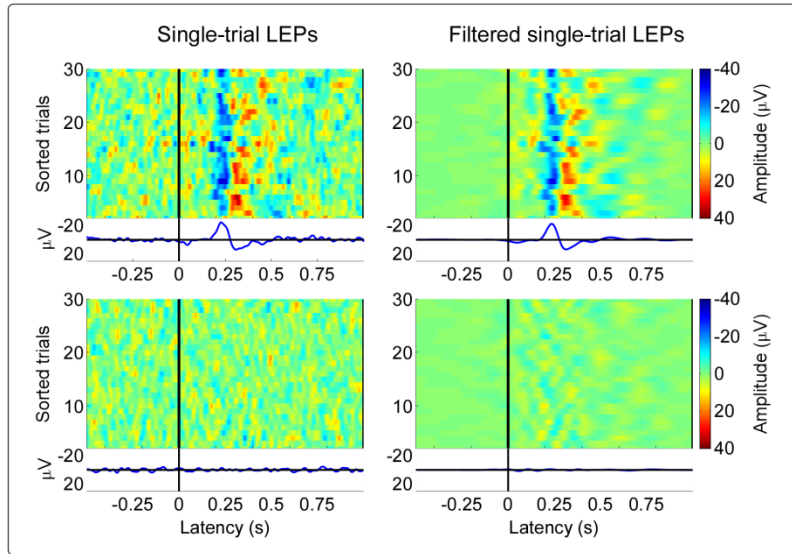
# Generation of a realistic ERP dataset



# Wavelet filtering to enhance ERP signal-to-noise ratio

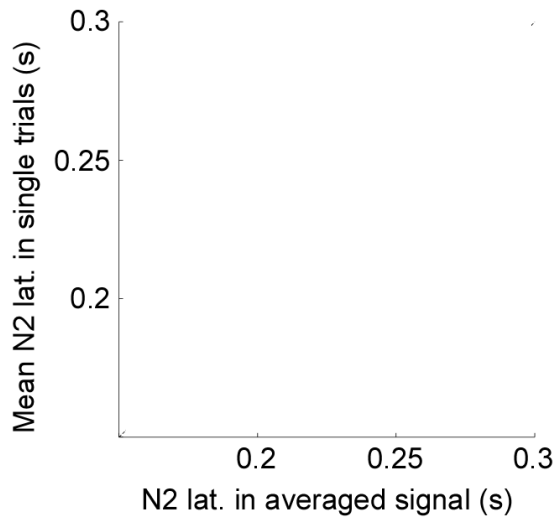


# Detection bias (on a real ERP dataset)

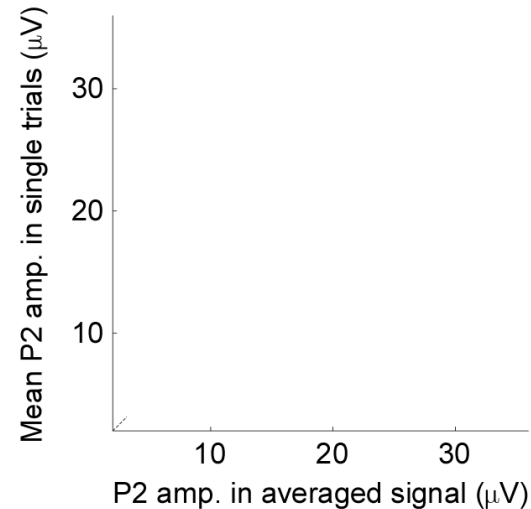
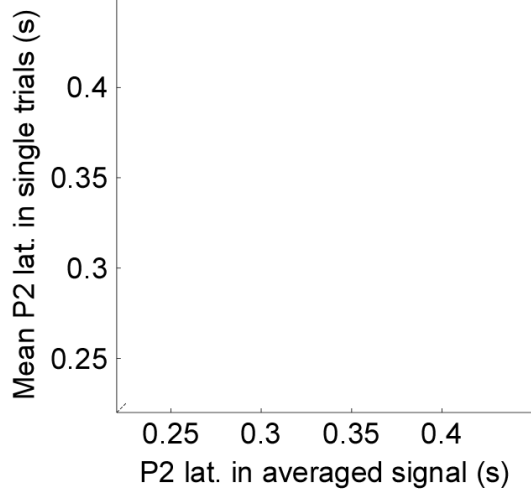
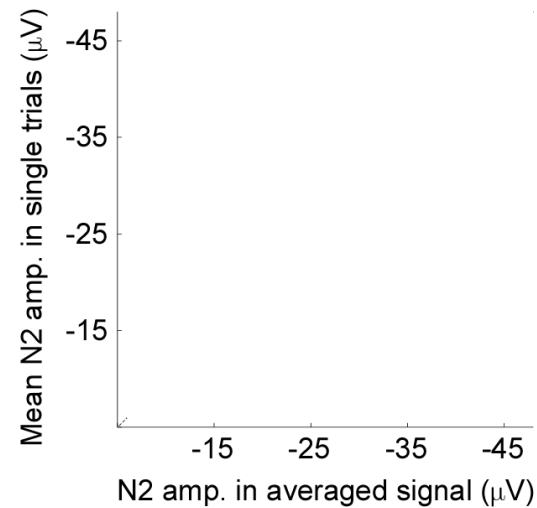


# Single-trial estimation - a real ERP dataset

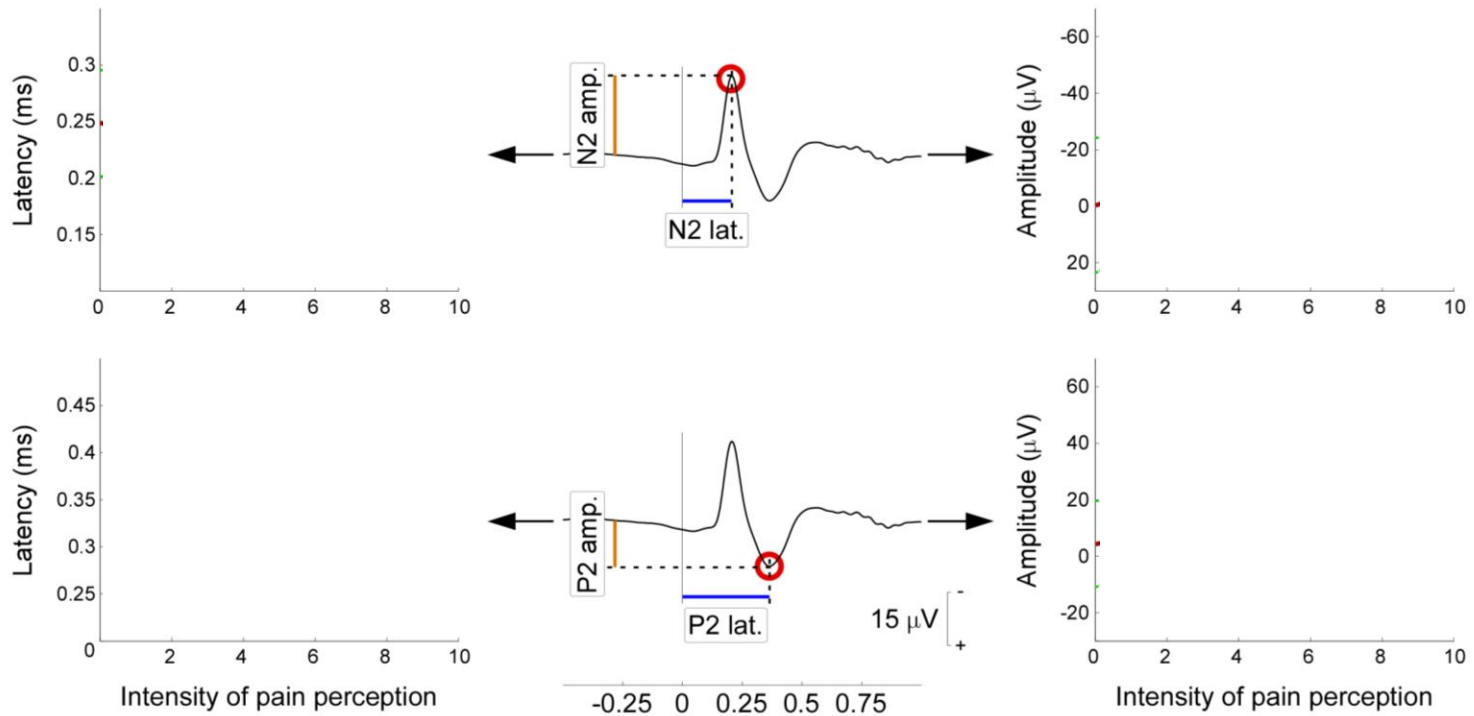
## Latency



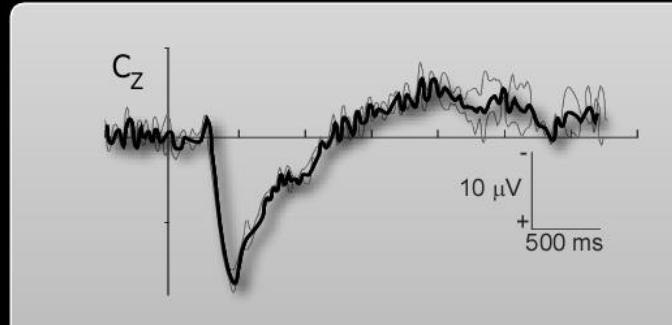
## Amplitude



# Correlation with behavioural measures



# 'phase-locked' and 'non-phase-locked' EEG responses



TF-AVERAGE

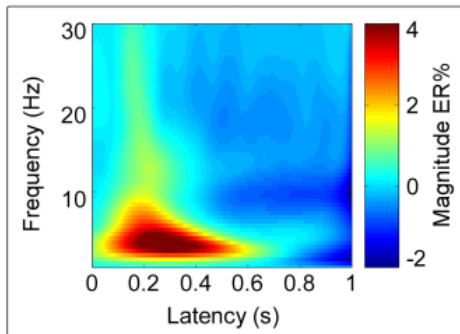
TF-SINGLE



(Mouraux et al, *CLINPH* 2005)

# Time-frequency ROIs definition

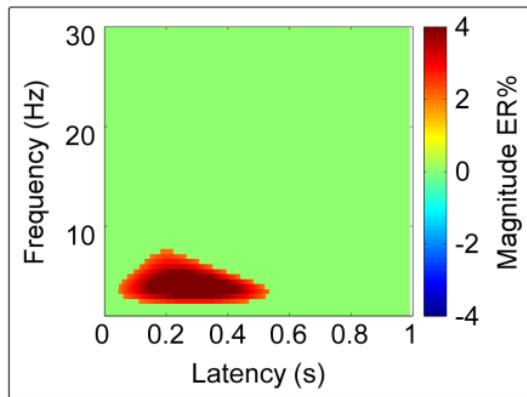
Time-frequency representation



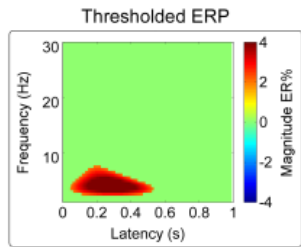


# Multiple linear regression in the time-frequency domain...

Thresholded ERP

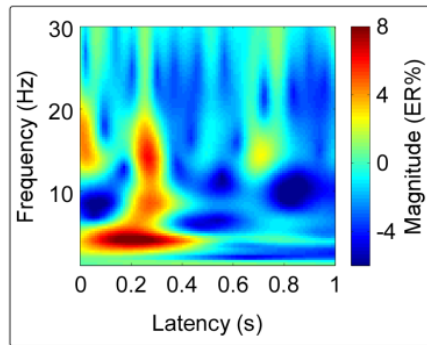


# ...taking ROI morphology into account.



# MLR vs dMLR

Single-trial time-frequency representation

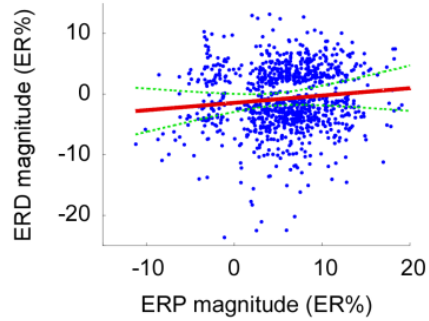
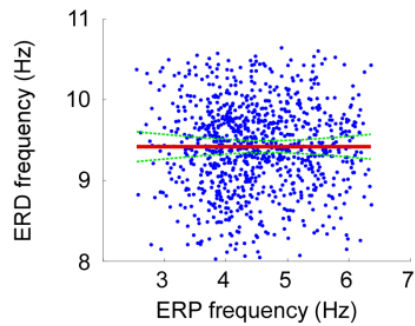
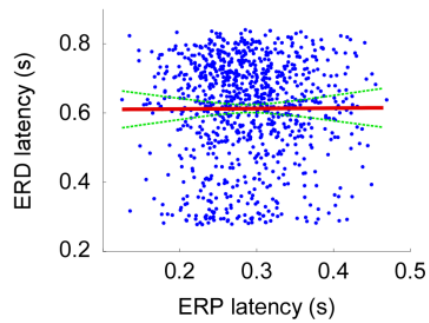


MLR fit

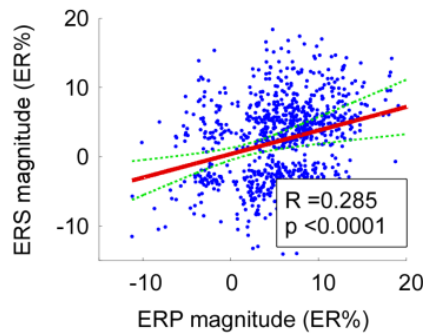
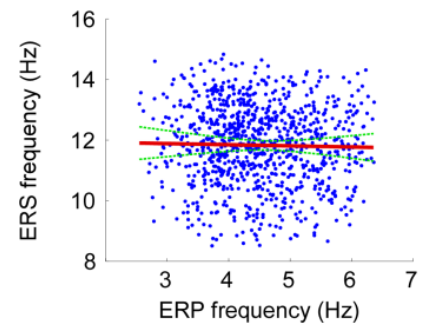
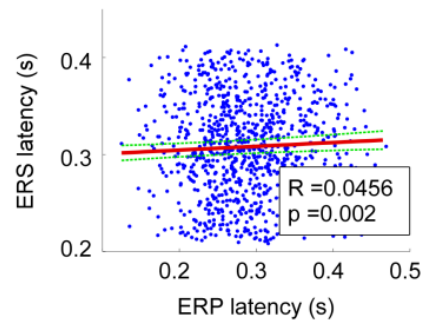
sMLR fit

# Single-trial correlations

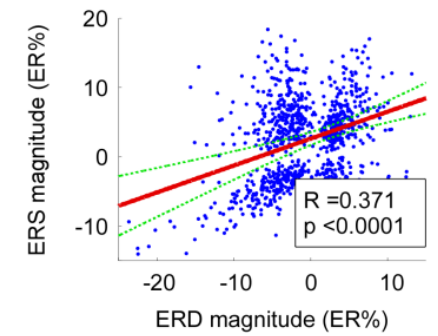
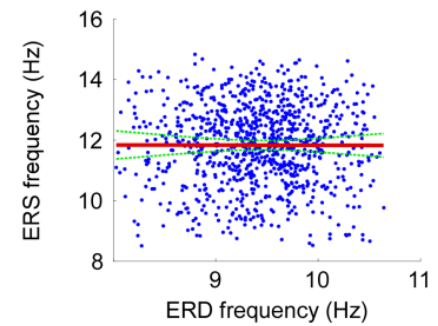
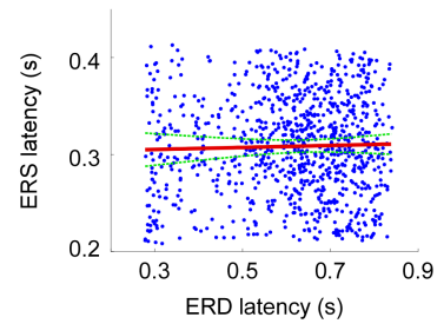
## ERP vs. ERD



## ERP vs. ERS



## ERD vs. ERS



## Summary and possible applications

1. Wavelet filtering significantly enhances the SNR of ERPs/ERS/ERD in single trials.
2. Multiple linear regression effectively captures the variability in the morphology of single-trial ERPs.
3. Combined, WF and MLR provides accurate and unbiased estimate of their peak latency and amplitude.

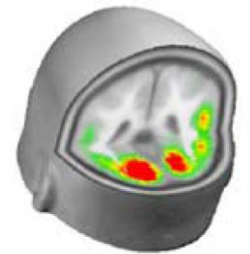
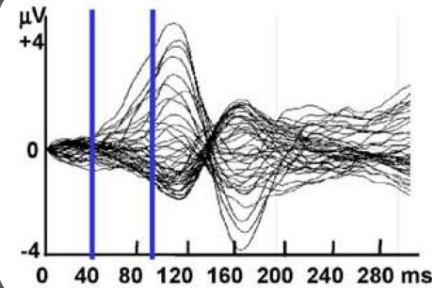
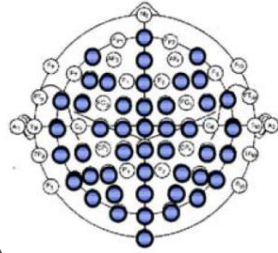
1. Within subject comparison!
2. Correlation with behavioural responses (perception, performance, reaction times - SDT), stimulus features, prestimulus features, experimental factors (e.g. drug concentration).
3. Correlation with other laboratory measures (withdrawal reflexes, EMG, BOLD-fMRI, MEG).
4. Robust estimation even in average waveforms (e.g. patient and drug studies)

# Simultaneous EEG-fMRI

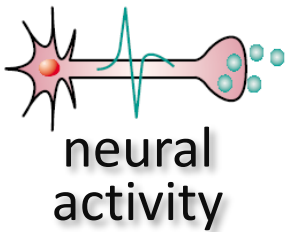
Scalp EEG



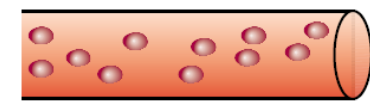
46 electrodes



sensory stimulus

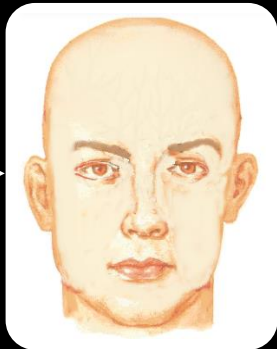
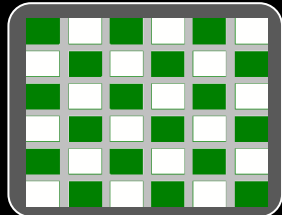


neurovascular coupling

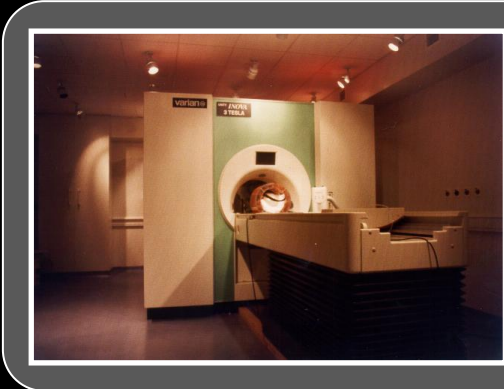
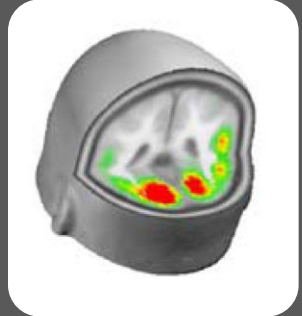
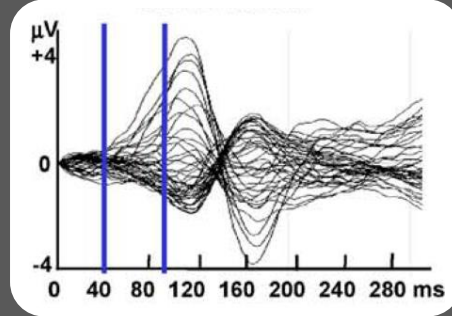
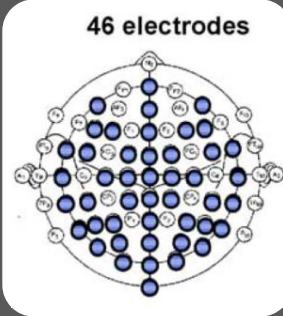


haemodynamic response

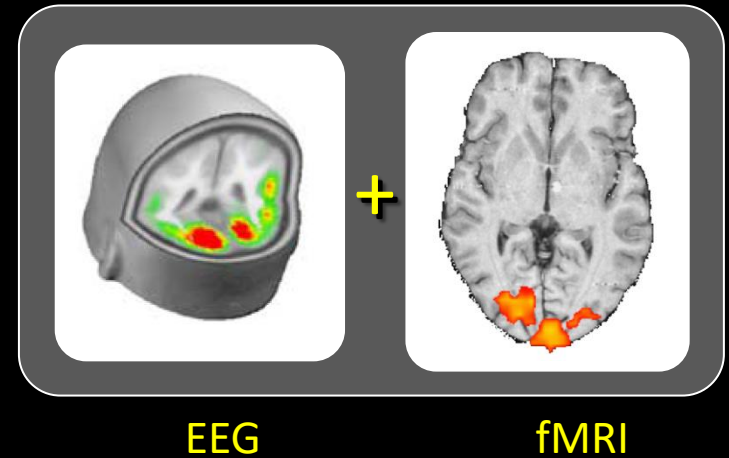
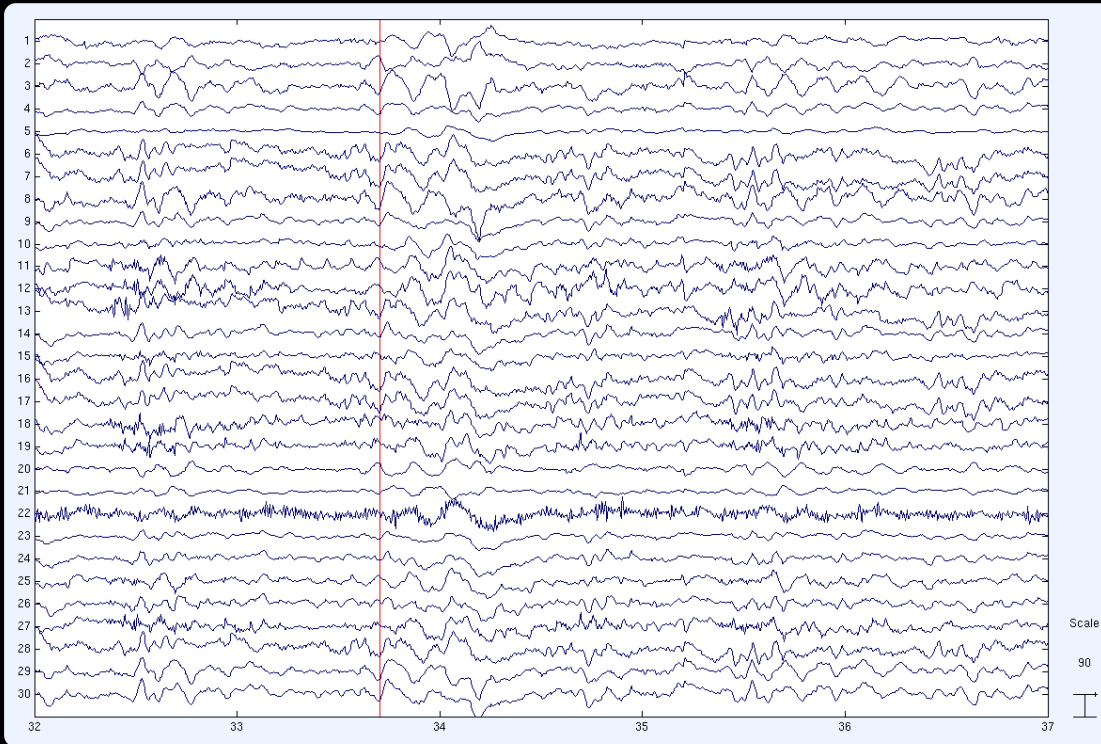
BOLD fMRI



# Simultaneous EEG-fMRI



# Basics concepts of functional neuroimaging: EEG vs BOLD-fMRI

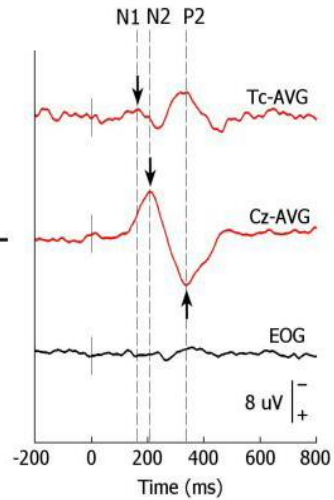


MR-induced edema artifact of fMRI.  
([www.fmrib.ox.ac.uk/~rami](http://www.fmrib.ox.ac.uk/~rami))

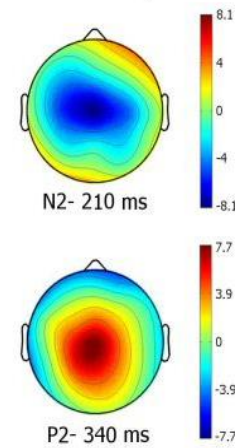
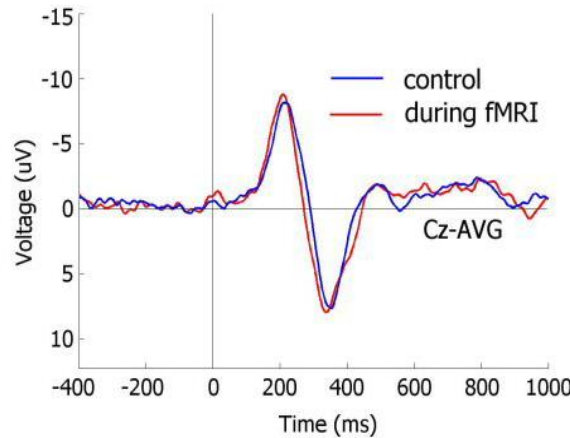
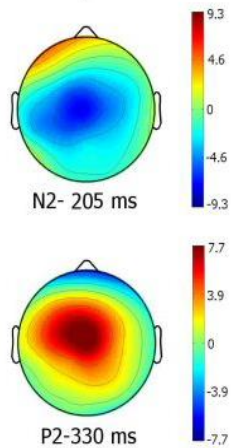
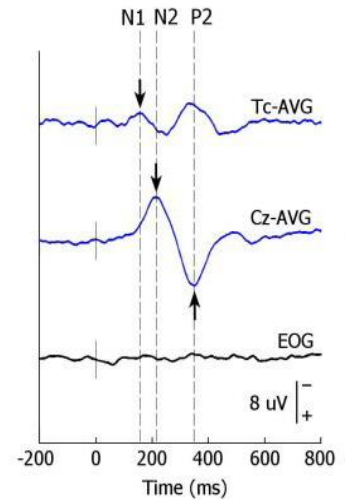


# Simultaneous EEG-fMRI of somatosensory-evoked ERPs

during fMRI

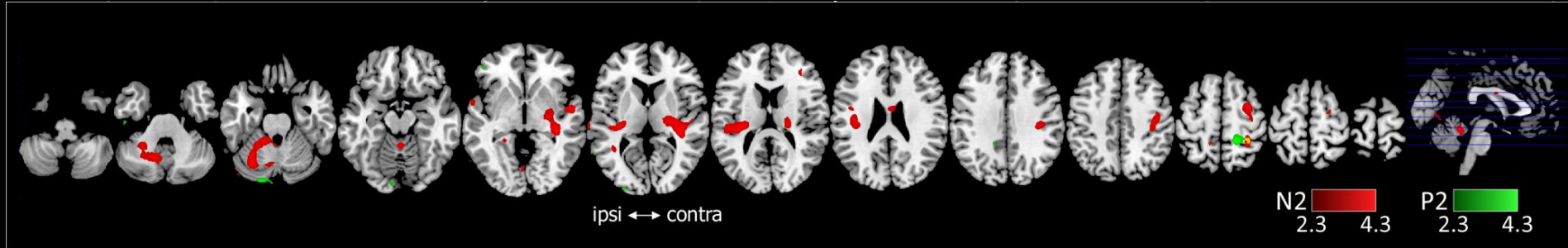
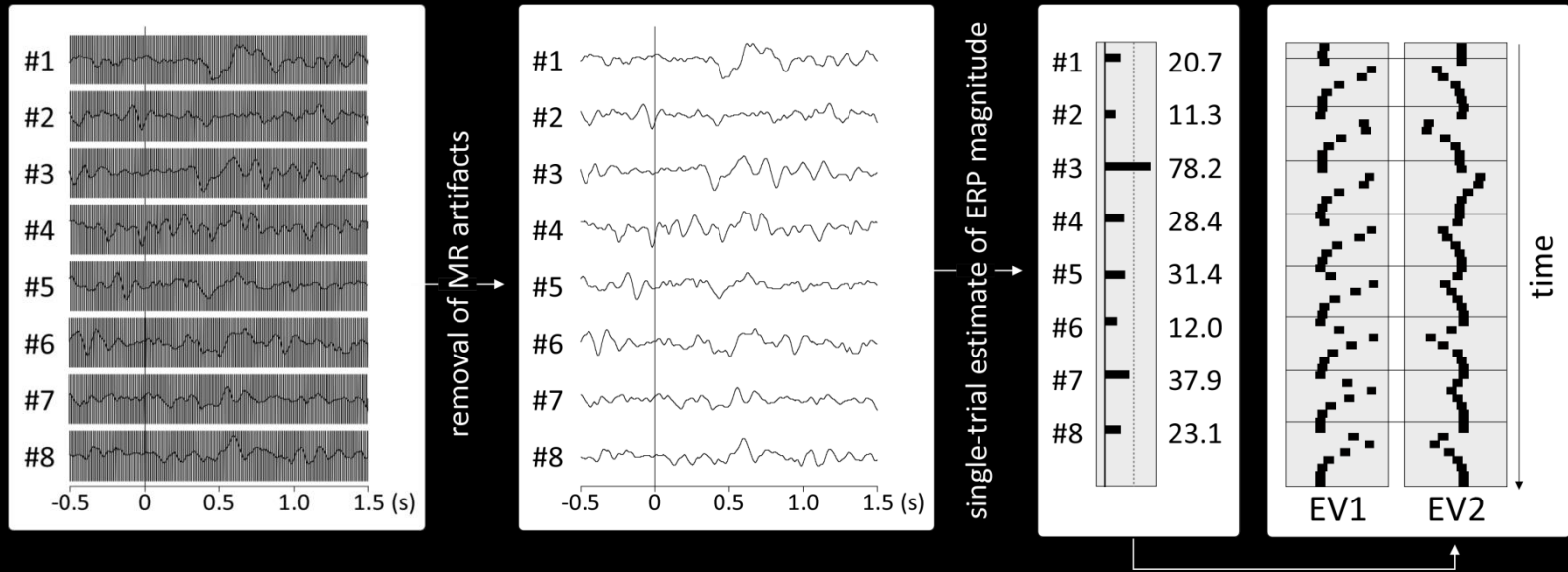


control



(Iannetti et al NeuroImage 2005)

# EEG-driven analysis of the fMRI responses to sensory stimulation



# EEG-driven analysis of the fMRI responses to sensory stimulation

