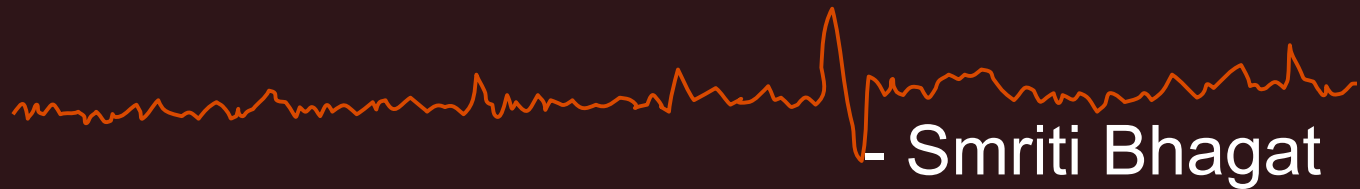


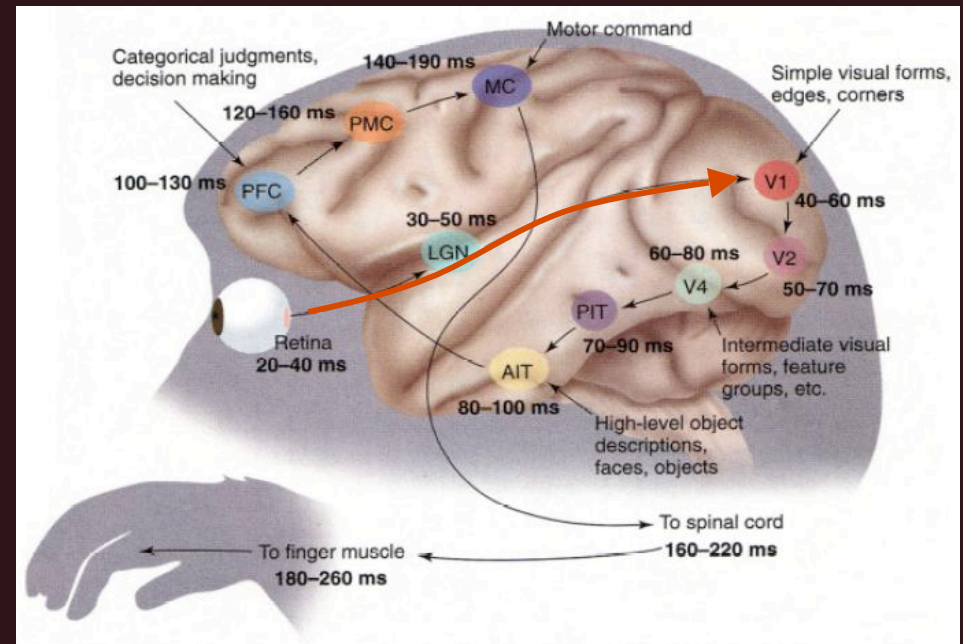
# Group Sensing for EEG Analysis



*Joint work with S. Muthu Muthukrishnan and  
Dr. David Rosenbluth (Telcordia) for the  
DARPA grant 'Group Brain Imaging'*

# Stimulus – Response

- Sensory – motor loops
- Stimulus – response
- Complex processing



Stimulus  
Presentation

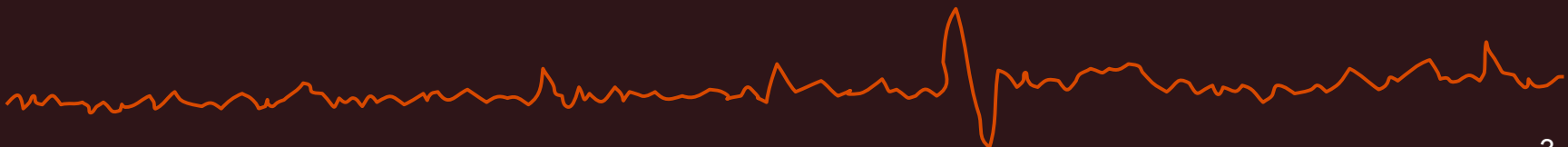
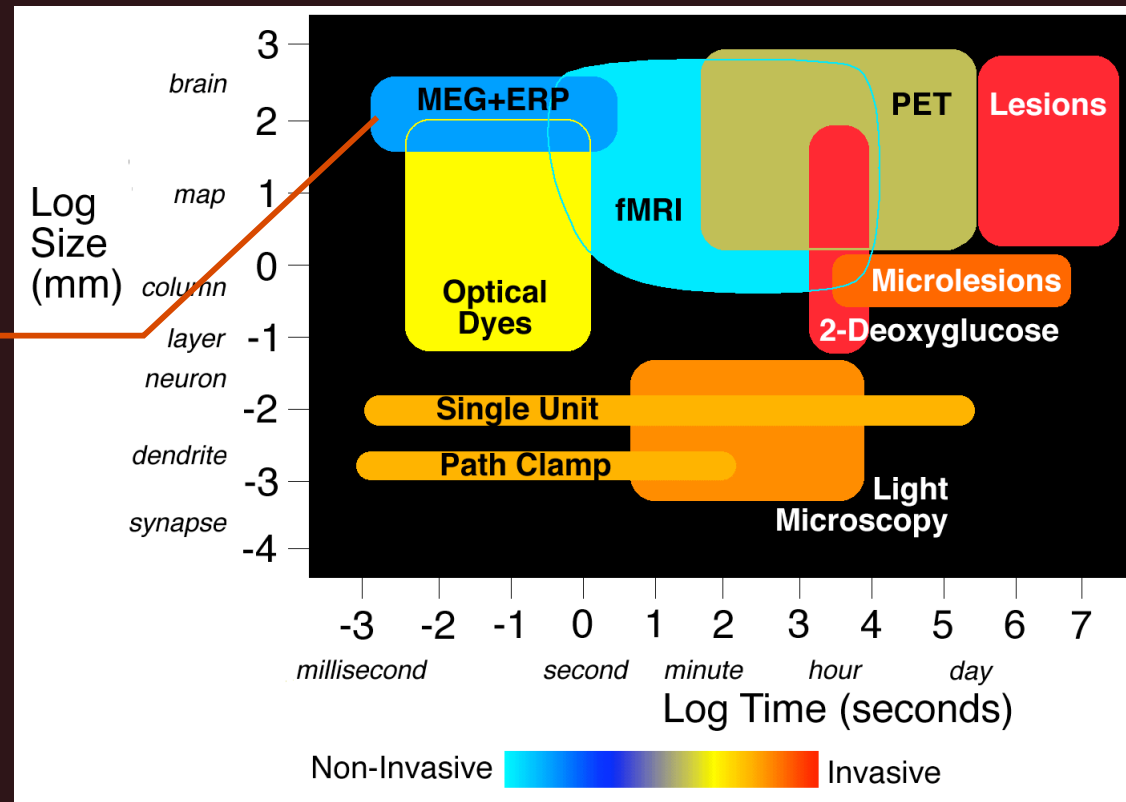
Brain analyzes  
the stimulus

Brain invokes  
a motor  
response

Motor action  
performed

# Brain Imaging Techniques

EEG is non-invasive and has high temporal resolution



Characterize  
brain's response  
to stimuli

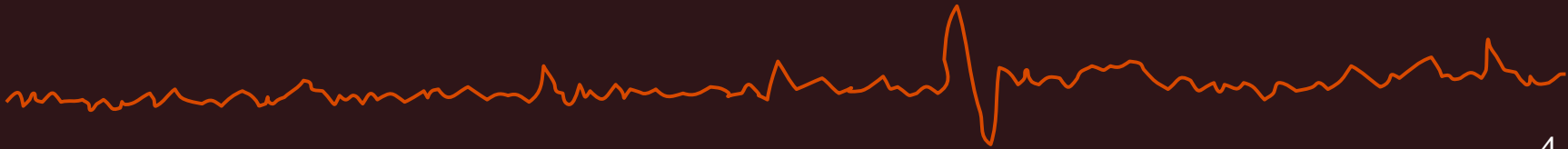
# Why EEG?

Clinical  
applications –  
Sleep, epilepsy ...

Brain Computer  
Interfaces

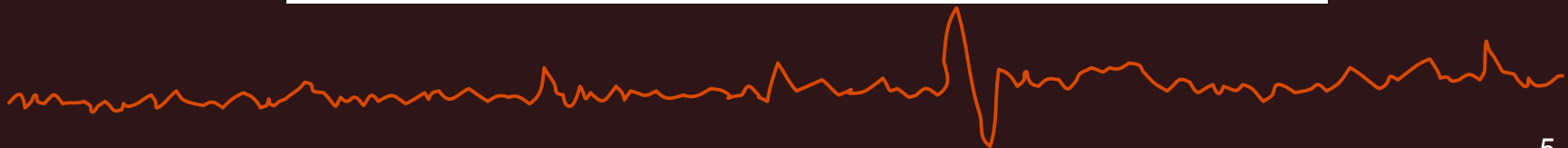
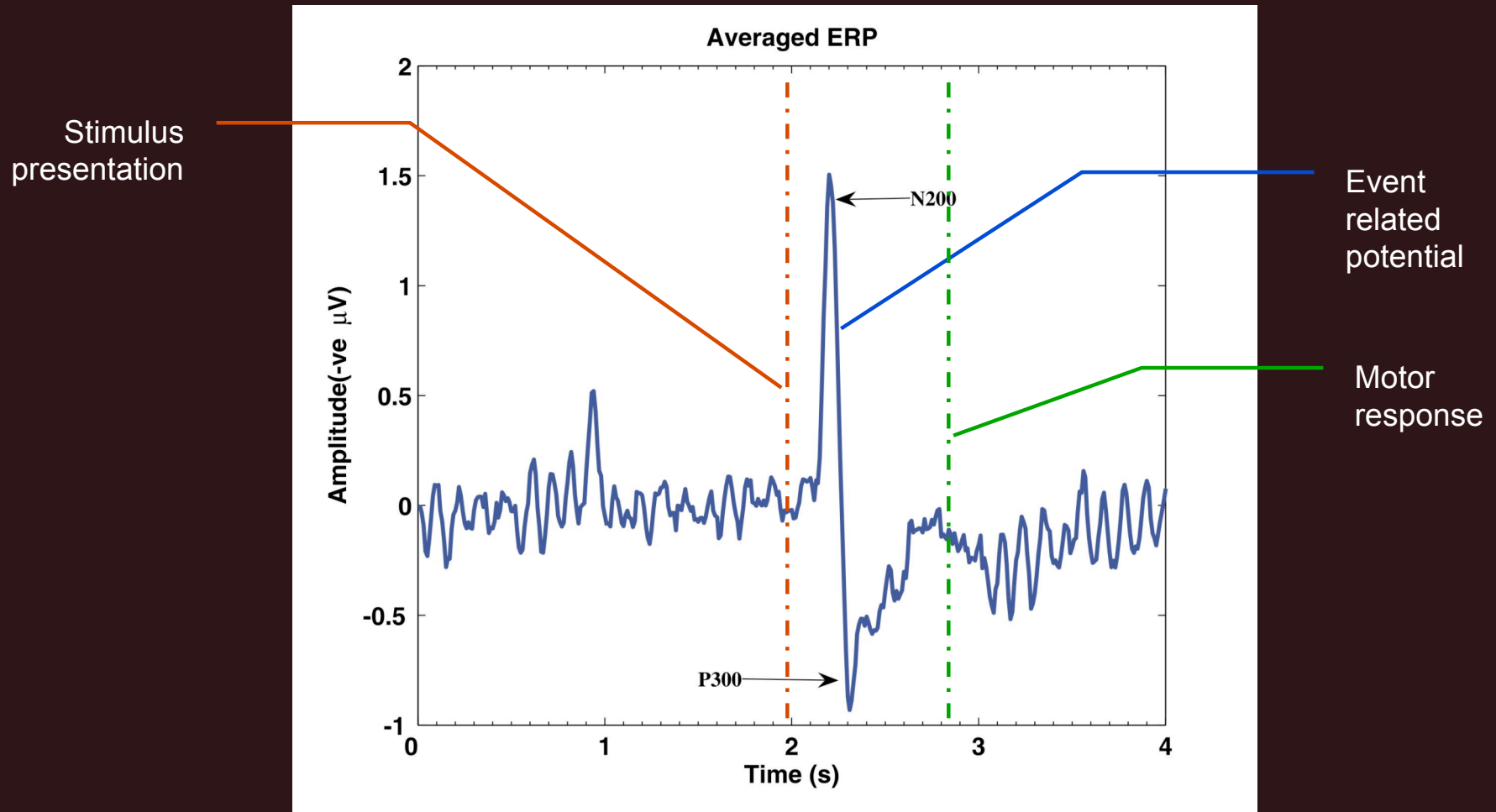
Understand  
functional  
relationships

Neuro-marketing

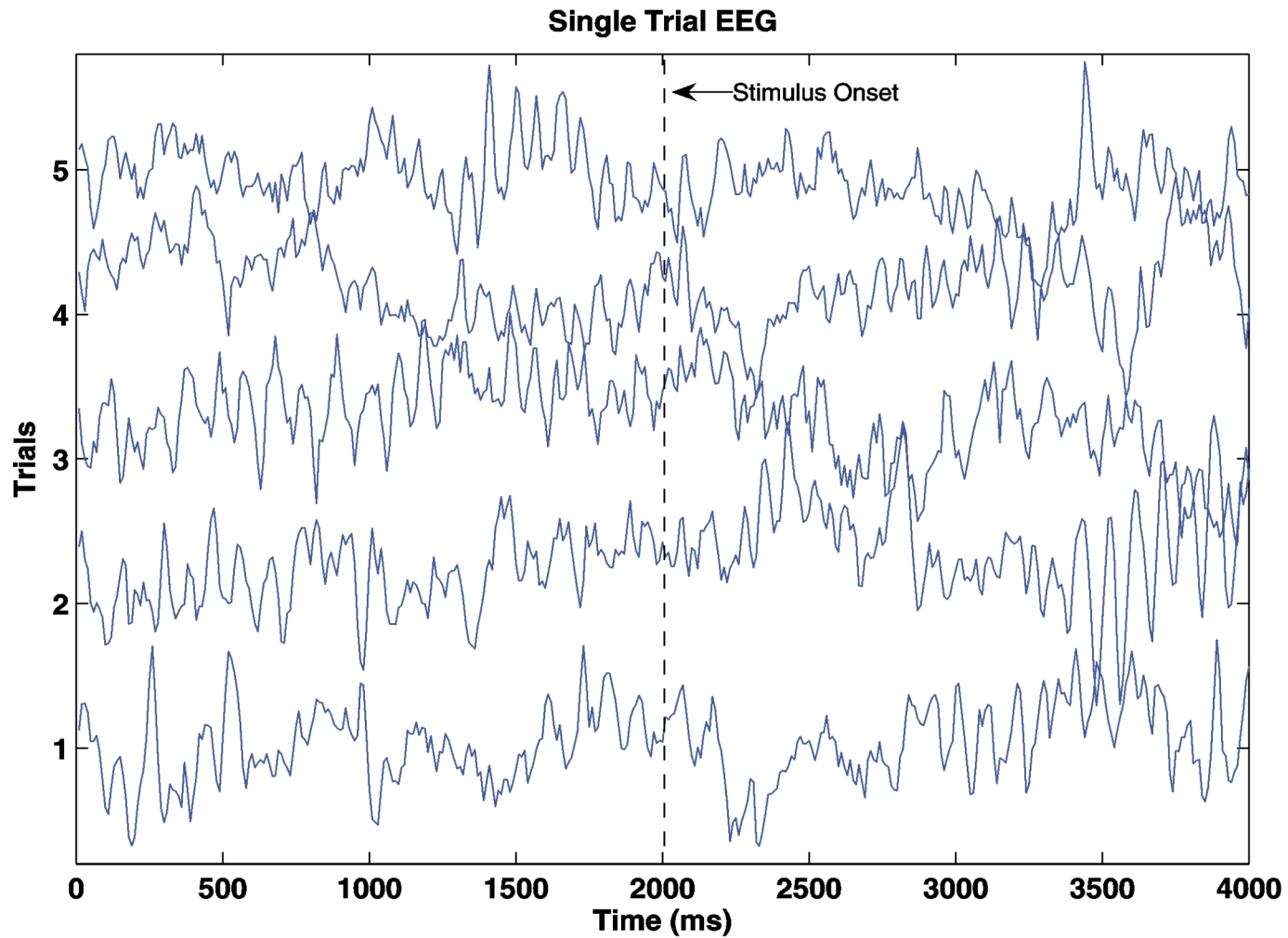




# Event related potential (ERP)

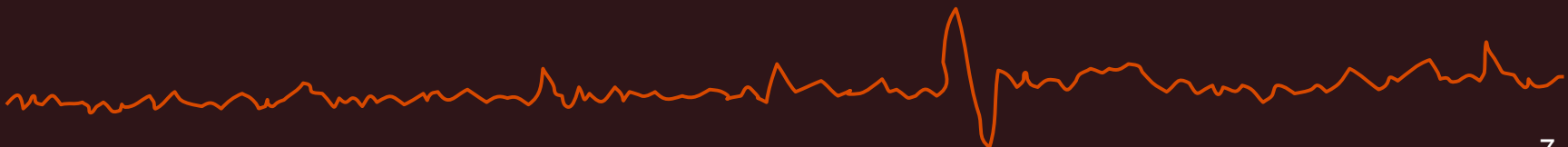


# Single trials

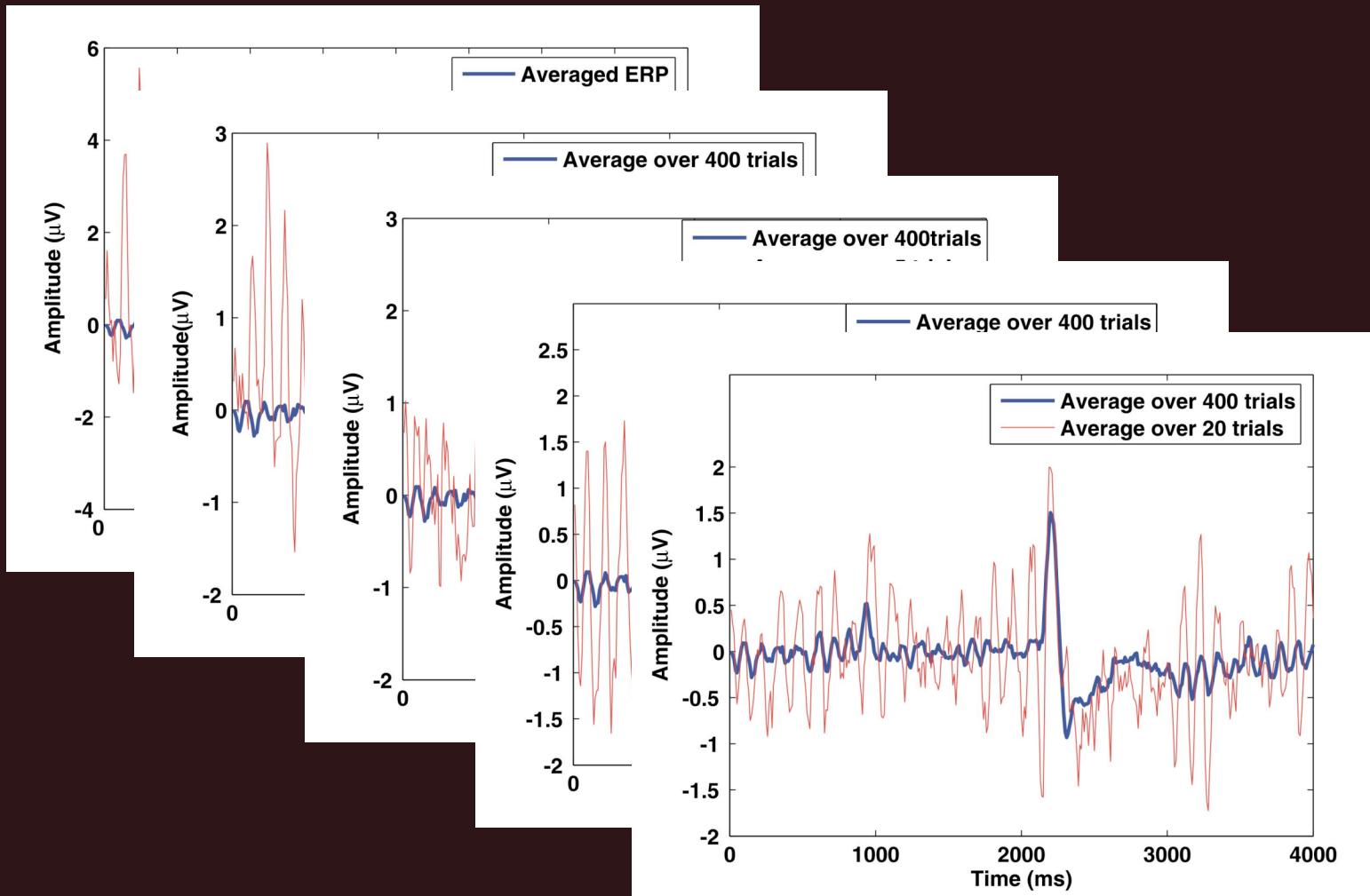


# Averaging over time

- Conventional solution
- Increases signal-to-noise by
  - Averaging different states of mind
  - Reducing variability
- Performed over many many trials

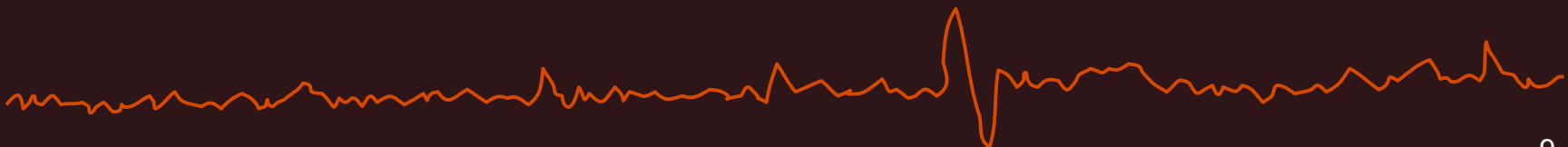


# How many trials?

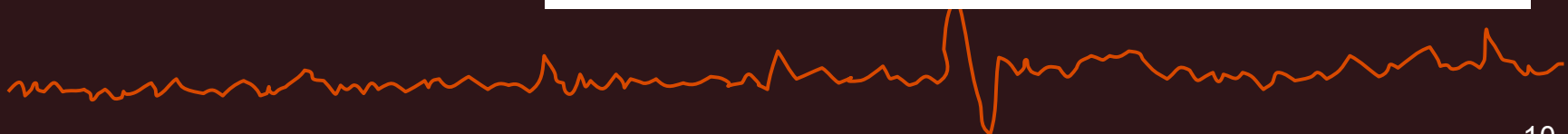
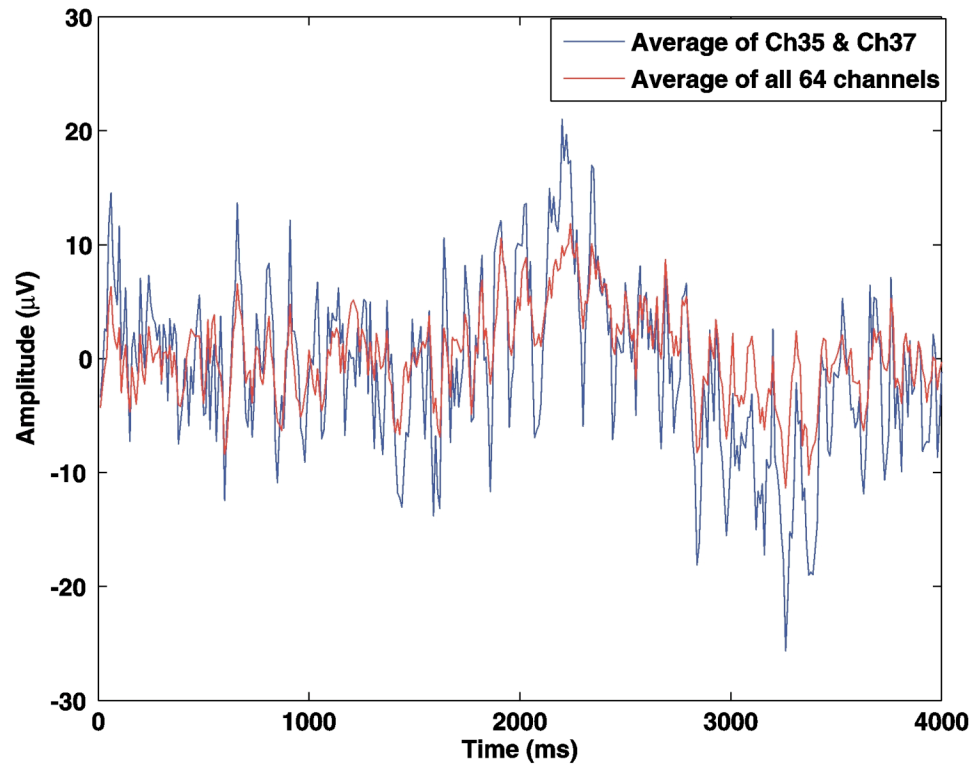
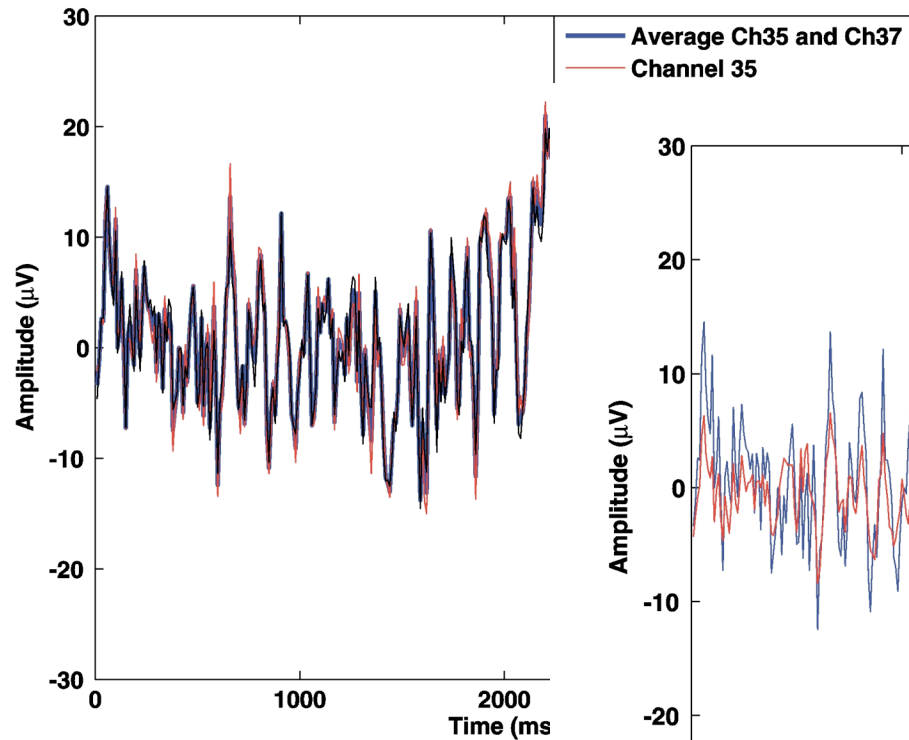


# Averaging trials

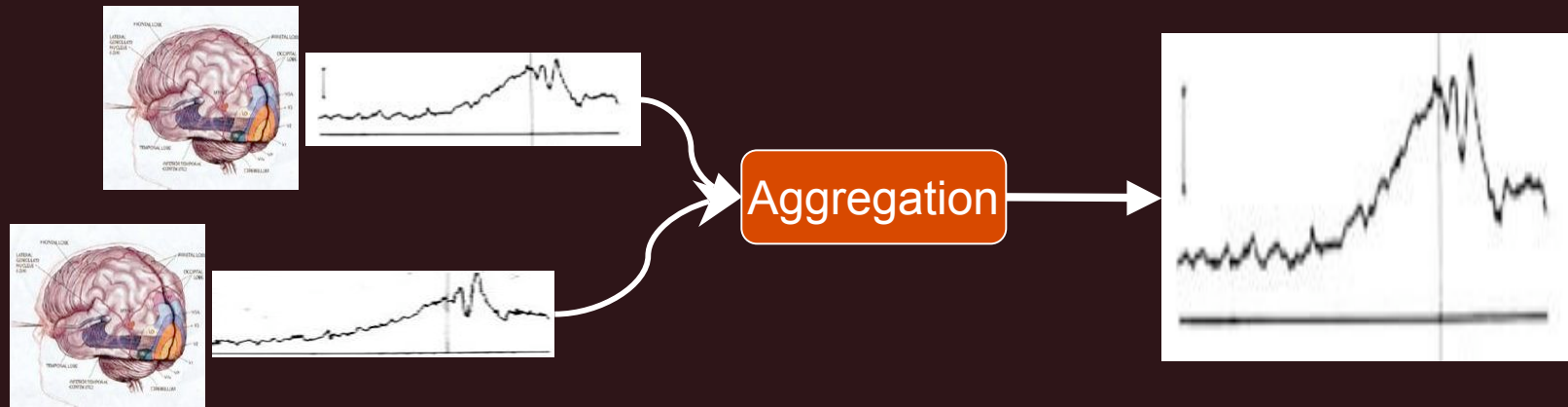
- Masks functionally significant trial-to-trial variations
  - state of the mind
  - state of the world (stimulus)
- Not suitable for single-trial analysis



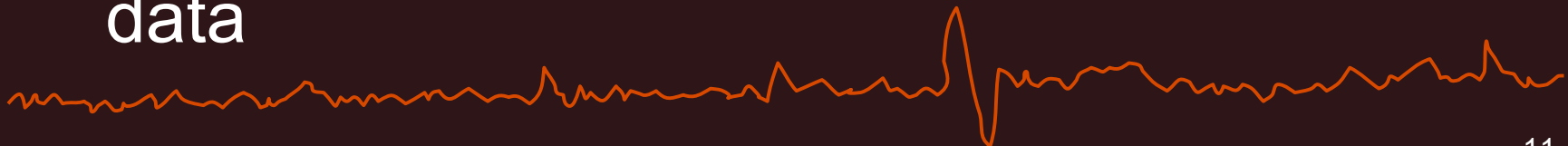
# Average multiple sensors



# Our contribution

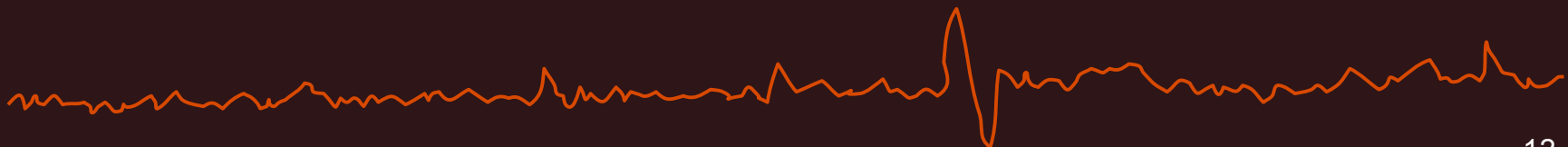


- Average EEG from multiple observers over single trials
- Improve signal detection accuracy by using multiple observers (Green & Swets)
- Group sensing never used with physiological data



# Our approach - Group Sensing

- Combine signals from multiple observers
- Improves detectability of the signal
- Appropriate for
  - single-trial analysis
  - comparing brain response from different observers





**Data  
Collection**

- Subjects
- Signal Detection Task
- Sensors
- Acquisition Setup

**Preprocessing  
(Data Cleaning)**

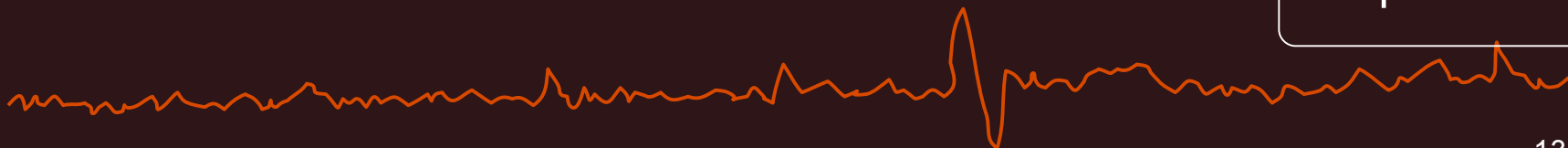
**Combining  
Data**

**Signal  
Detection**

**Algorithm Performance  
Analysis**

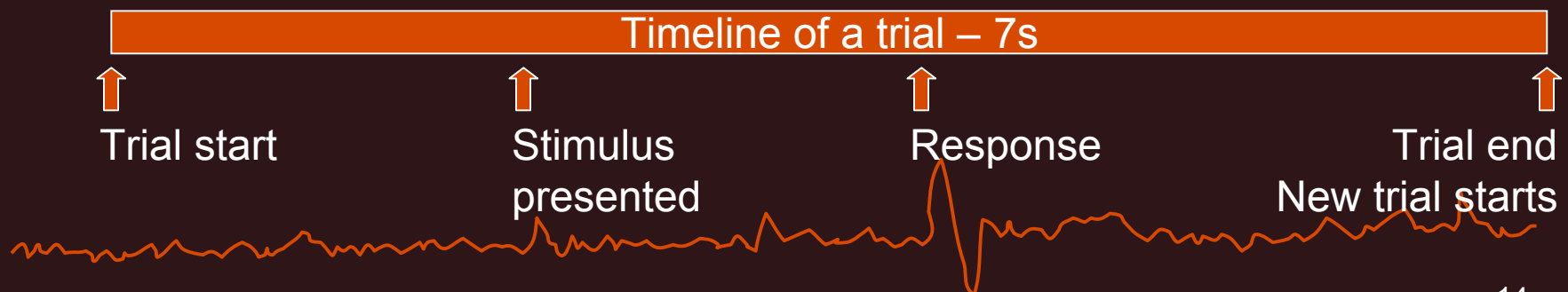
**Optimization**

**Data Analysis**



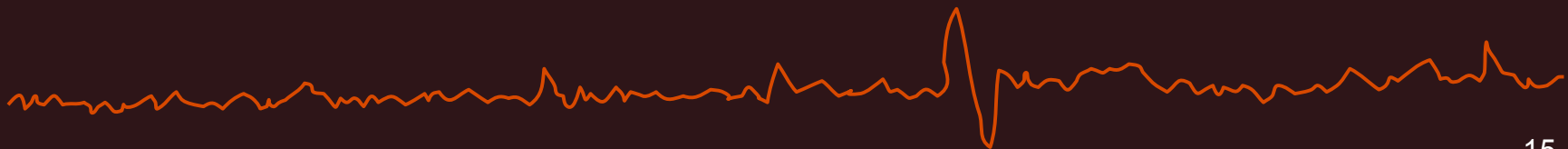
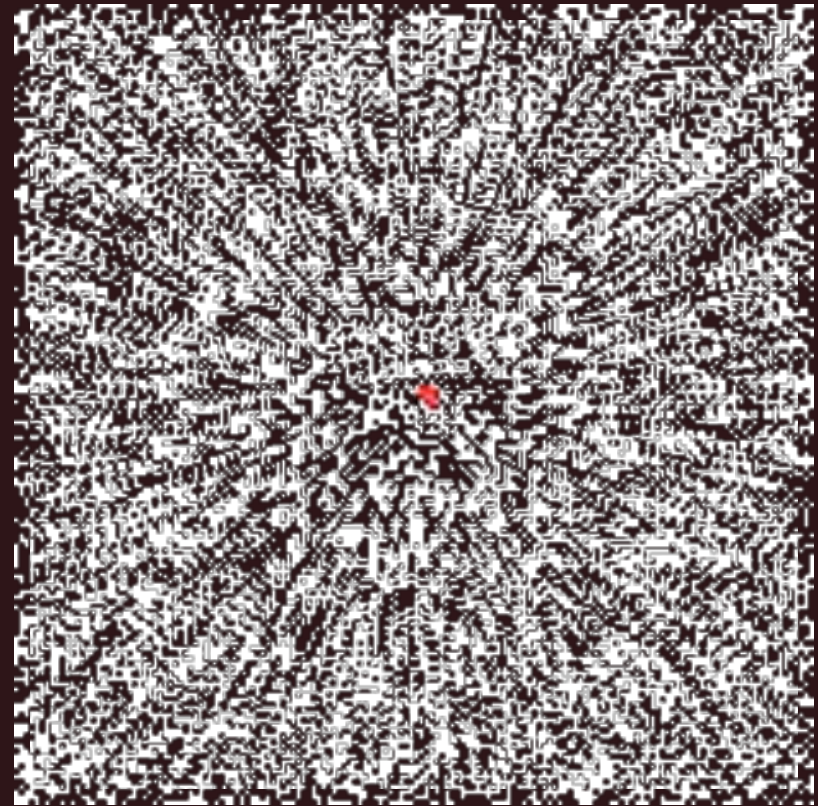
# Task

- Determine the direction of the arrow
- Subject responds by a button press (left/right)
- 400 trials, each 7s long
- Each recording is about 45 min

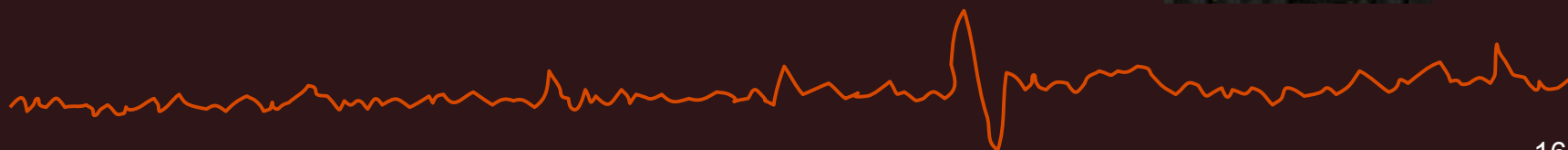
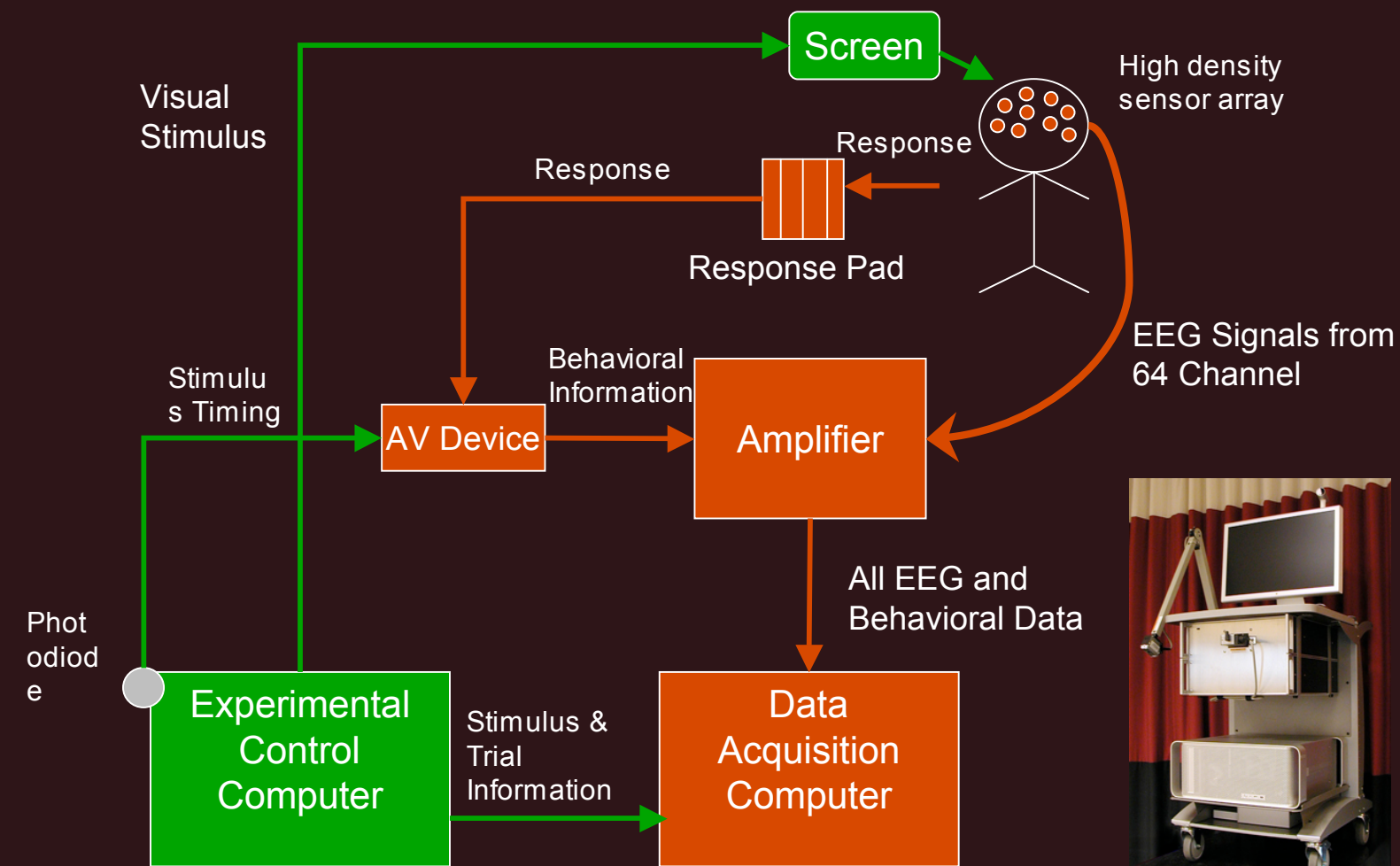


# Stimulus

- Target detection task
- Stimulus is a motion-defined shape; arrow moves left or right
- Arrow points left or right
- Stimulus parameters are unique for each trial



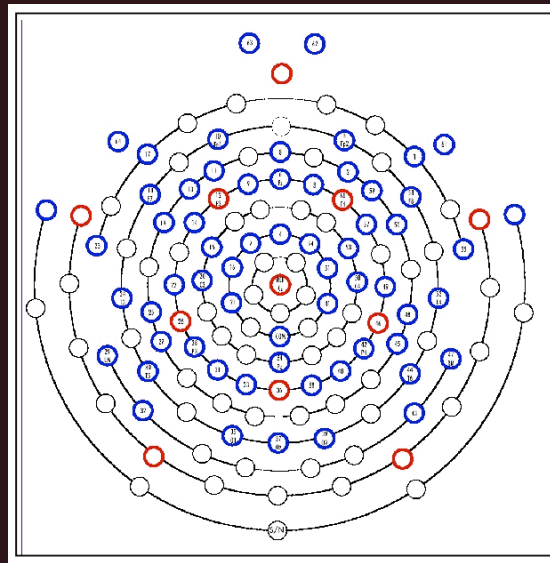
# Acquisition Setup



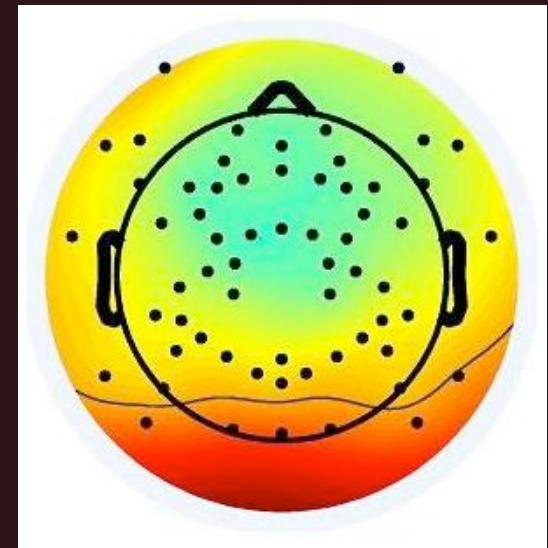
# Sensors



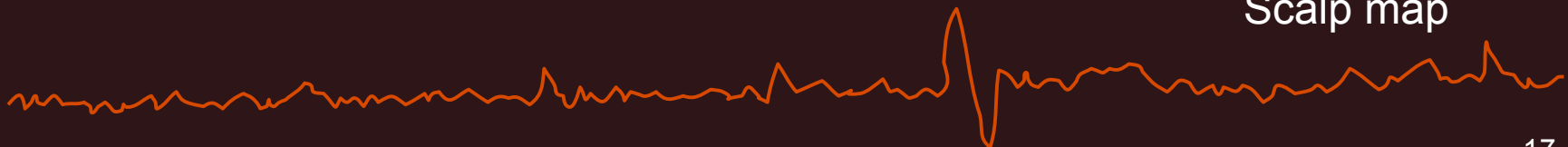
High-density 64  
sensor array



Geodesic layout of sensors



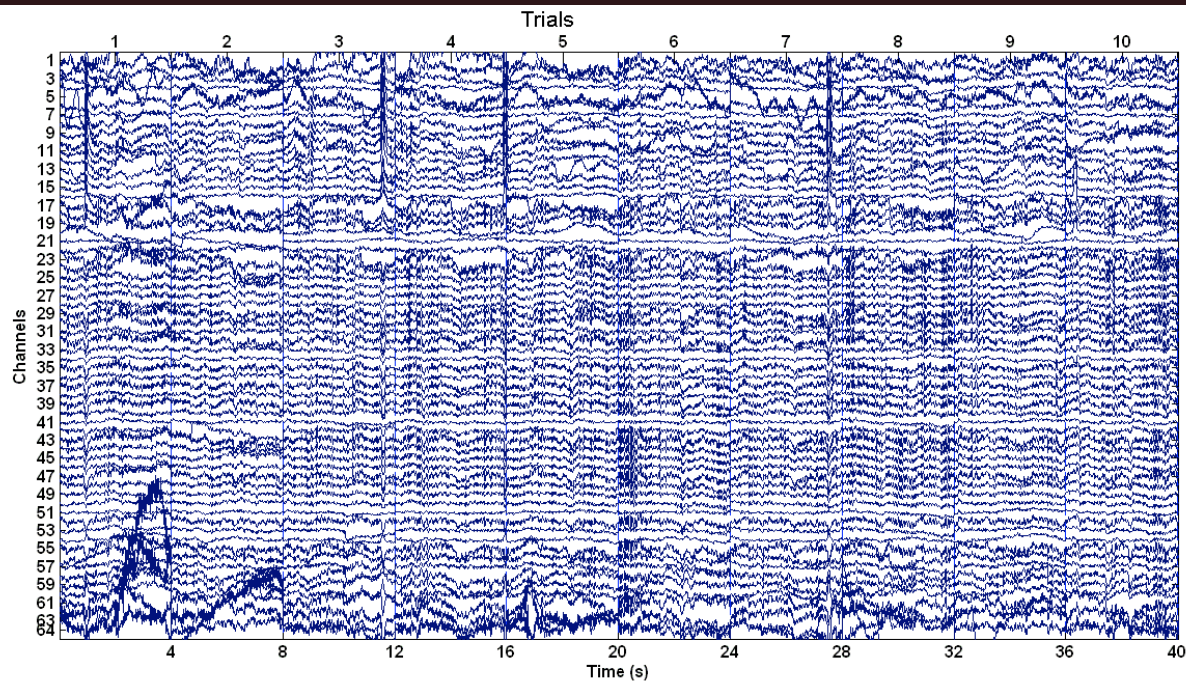
Scalp map





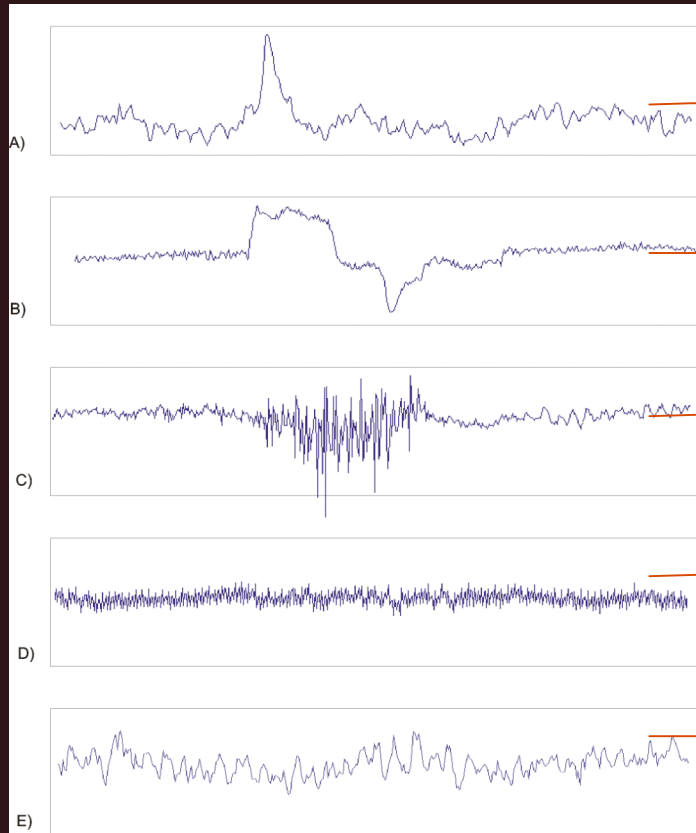
# Data acquisition

- 11 datasets from 8 subjects
- EEG is recorded at 1000Hz
- Each dataset is about 1 Gb of data

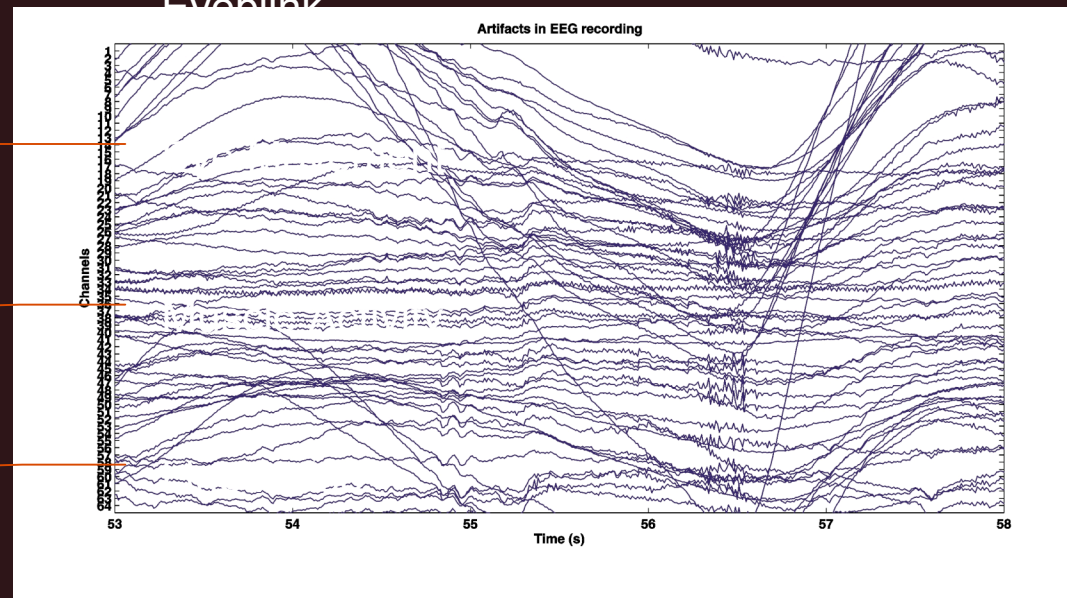


Continuous EEG data

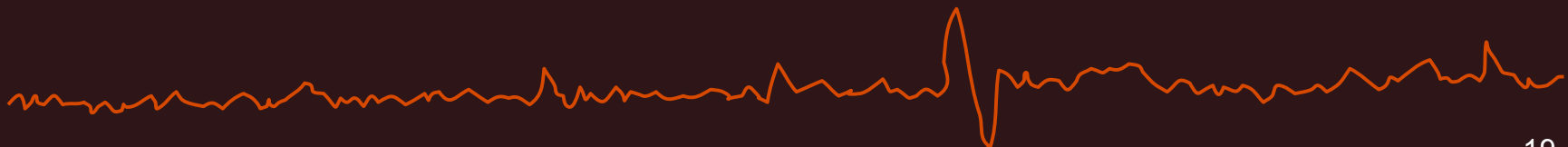
# Artifacts



Eyeblink



Clean EEG



Data Collection

**Preprocessing -  
Data Cleaning**

Combining  
Data

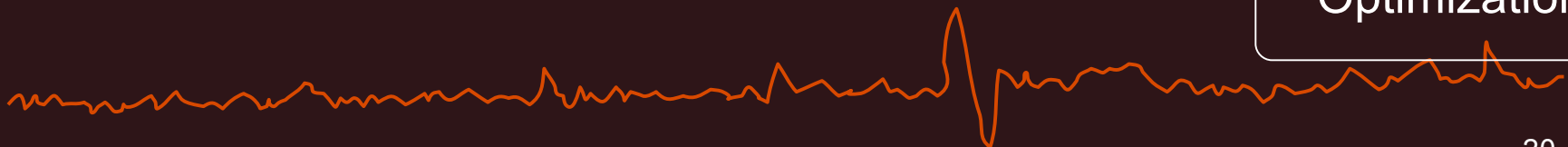
Signal  
Detection

Algorithm  
Performance Analysis

Optimization

- Filtering
- Segmentation
- ICA
- Artifact Removal

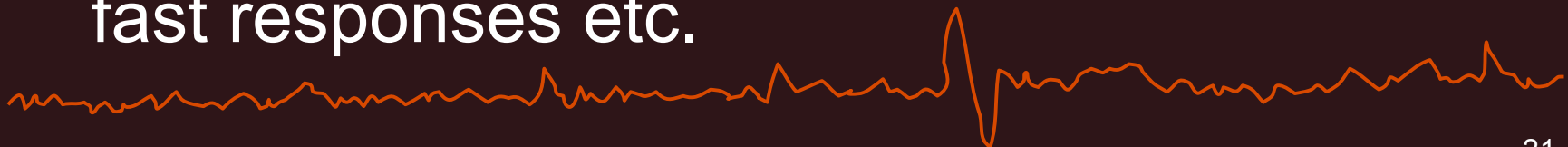
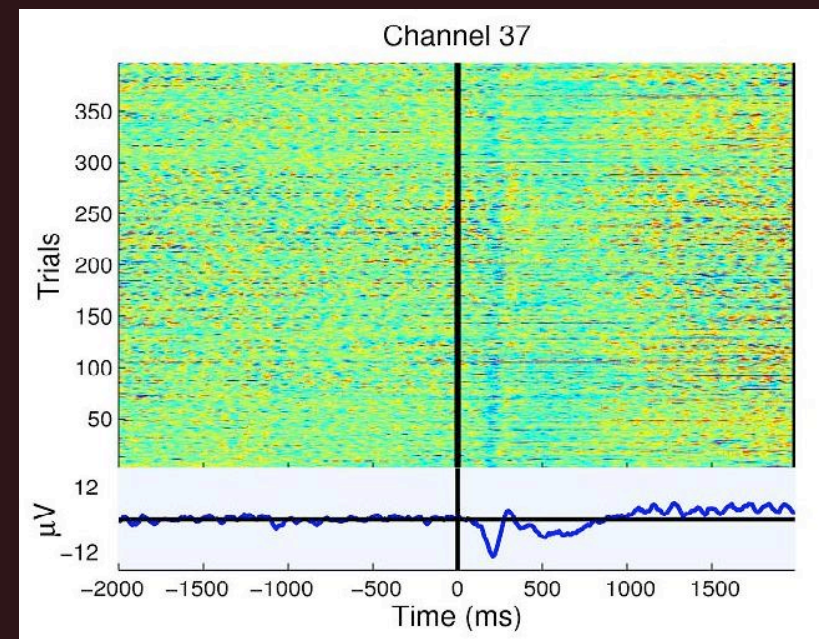
# Data Analysis





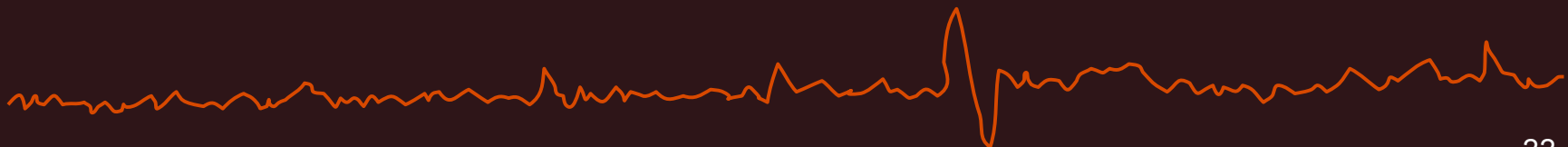
# Segmentation

- Continuous EEG trial data is segmented into 4s intervals centered on the stimulus onset
- Segmentations time-locked to different events highlight different features like, correct, missed trials, fast responses etc.



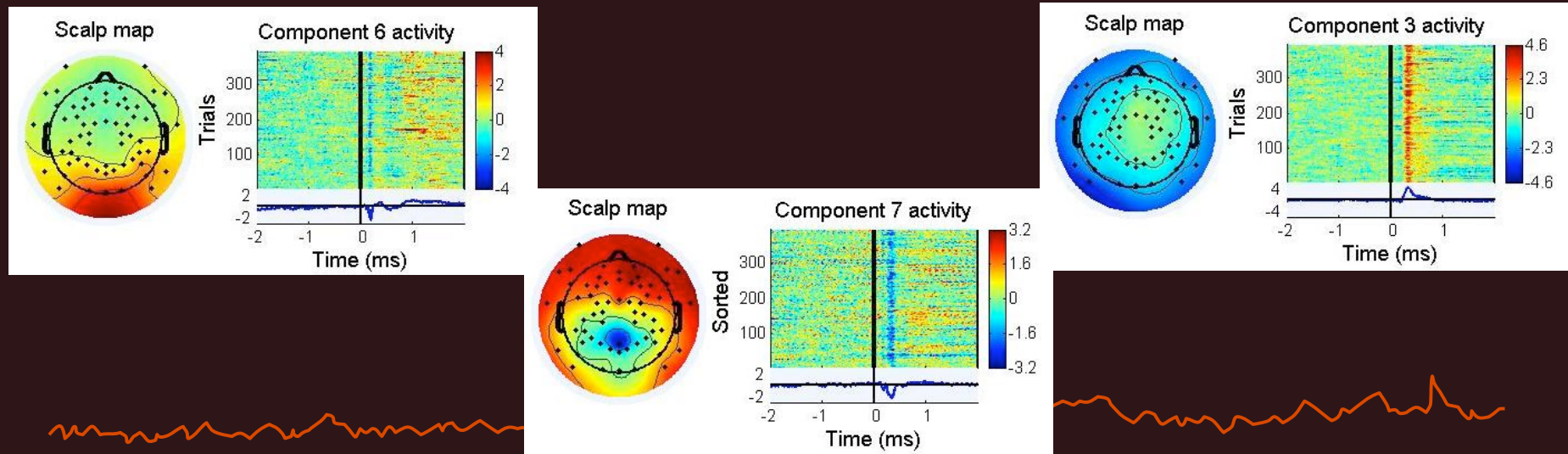
# Signal separation

- Observed signal is the contribution of activity within and outside the brain
- Independent signals should be separated for:
  - feature extraction of EEG data
  - data reduction
  - artifact detection



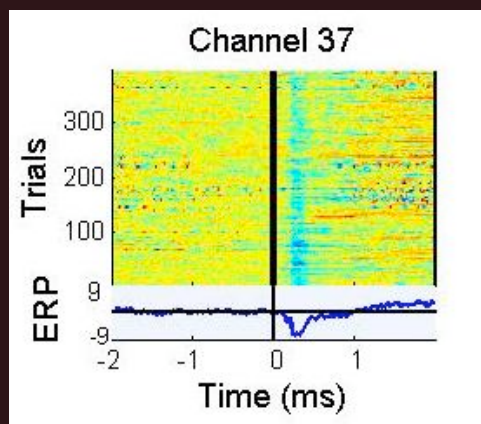
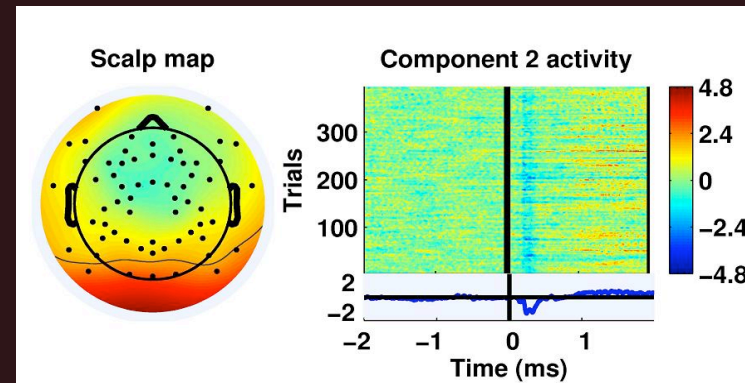
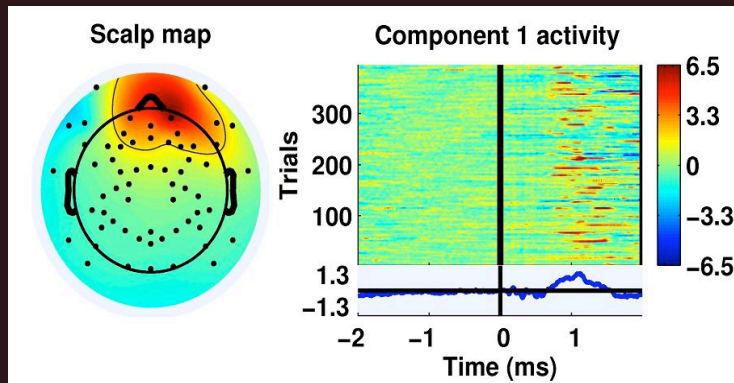
# Signal separation : ICA

- ICA is similar to PCA
- Results in maximally independent projections of data
- Independent components highlight different features

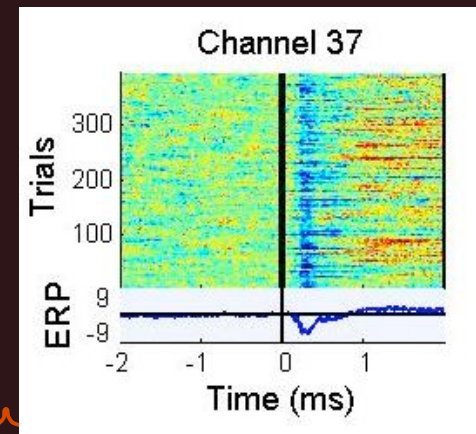


# Independent Components

- ICA consolidates information



Artifact Removal



Data Collection

Preprocessing -  
Data Cleaning

**Combining  
Data**

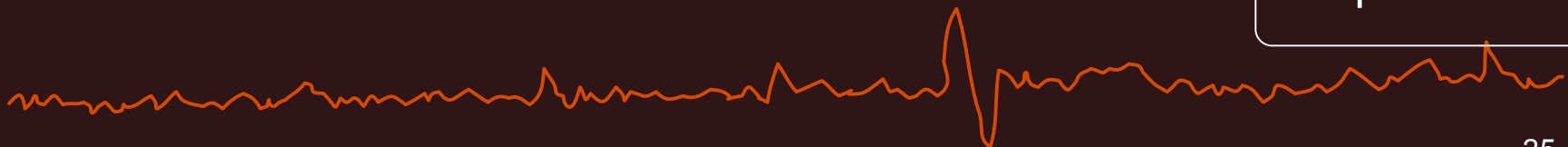
Signal  
Detection

Algorithm  
Performance Analysis

Optimization

- Simple Averaging
- Aligning
- Correlation

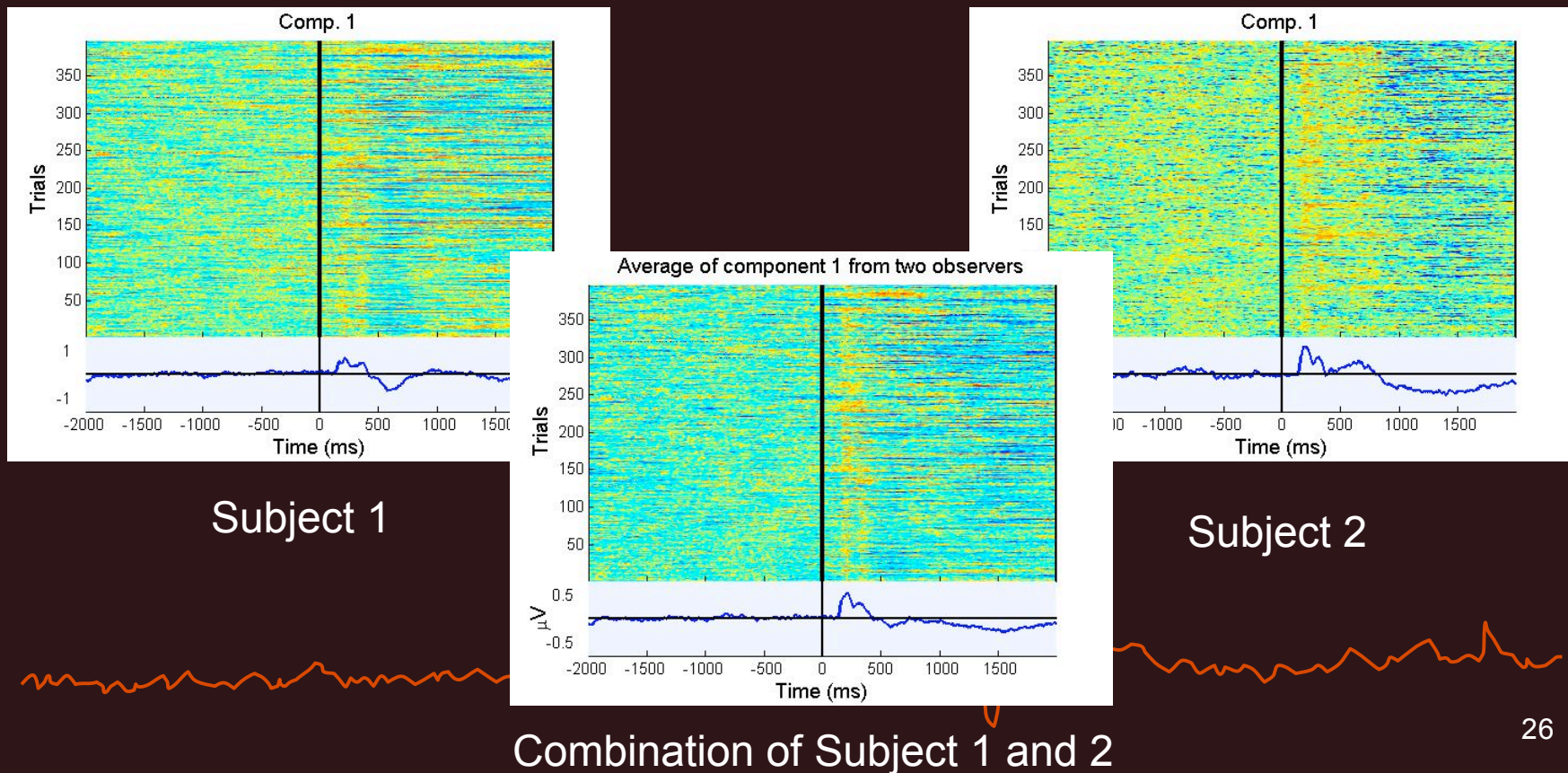
# Data Analysis





# Combining EEG data across subjects

- Variability across subjects
- Variability in ERP latency, amplitude and shape



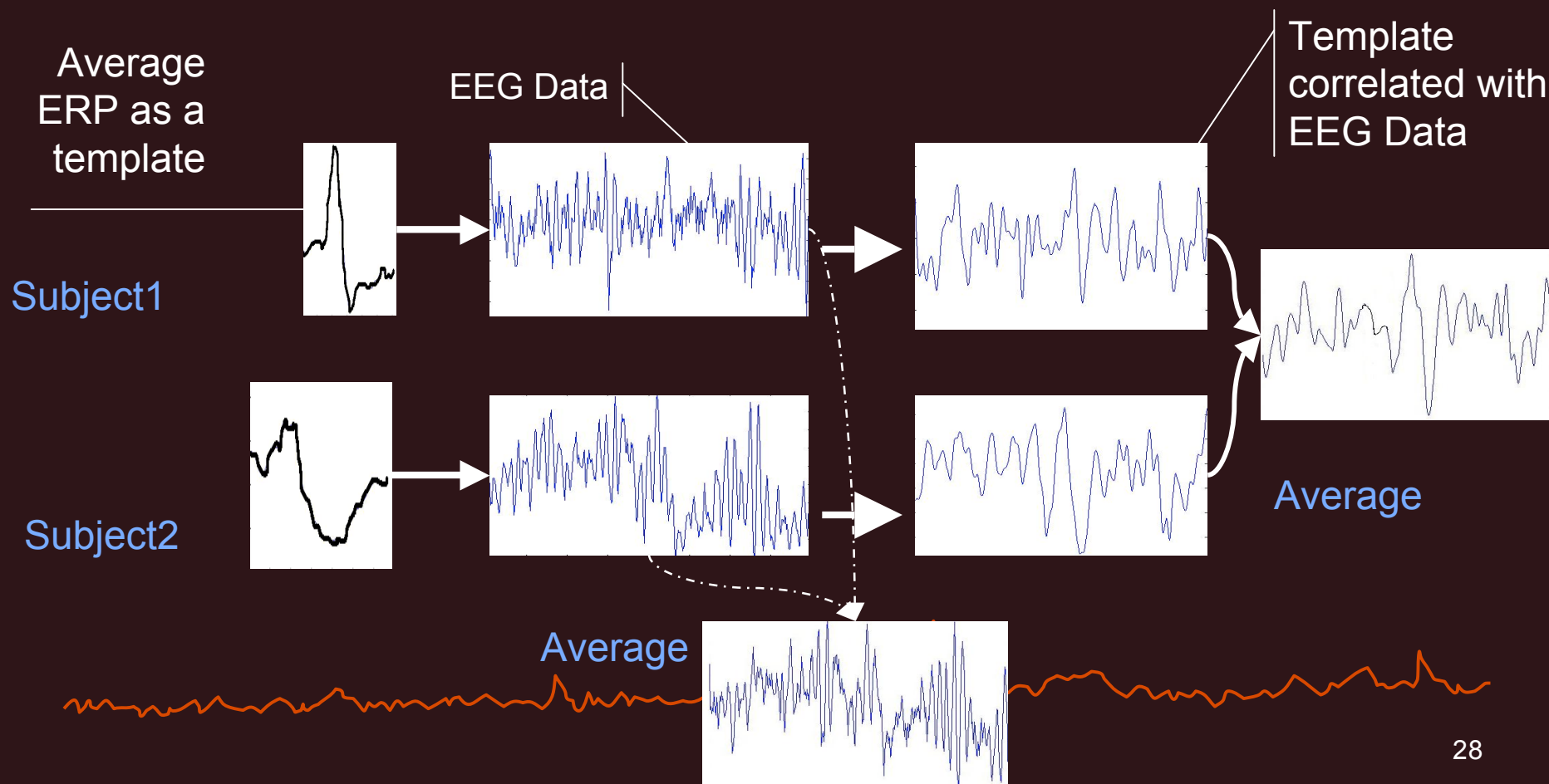
# Dealing with variability

- Time Aligning
  - Overcome latency differences
  - Match the P300 latency of average ERP and shift to align the data



# Dealing with variability

- Template Matching
  - for overcoming variability in structure of ERP





# Data Analysis

Data Collection

Preprocessing

Combining  
Data

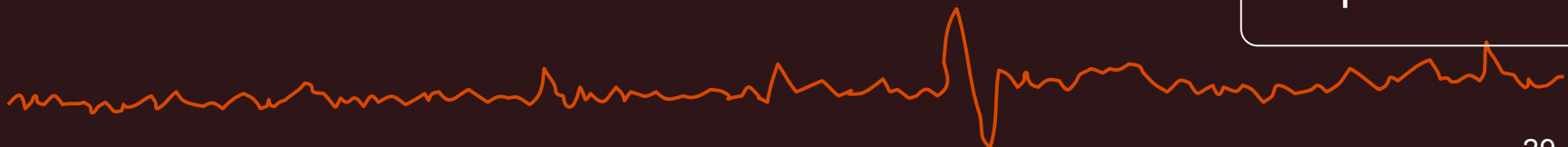
Signal  
Detection

Simple Thresholding  
ROC Curves

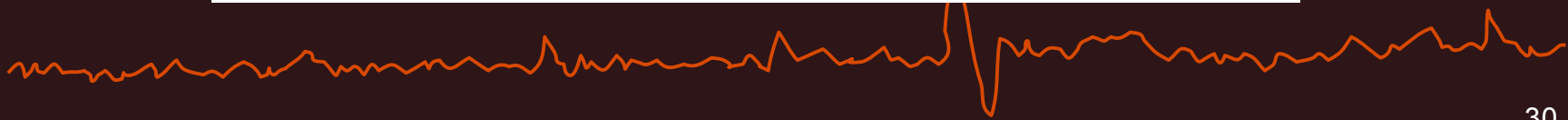
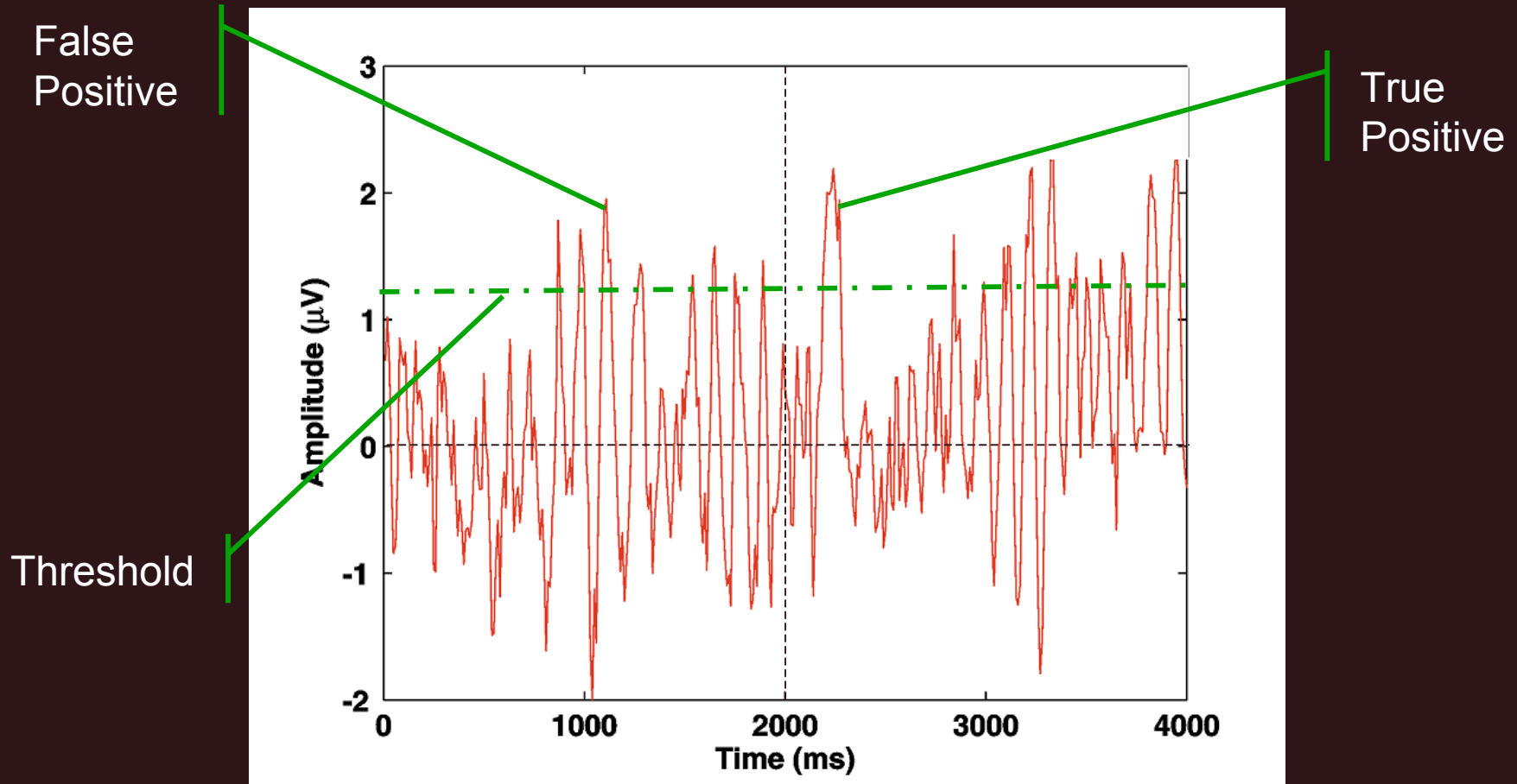
Algorithm  
Performance Analysis

Area under ROC Curve

Optimization

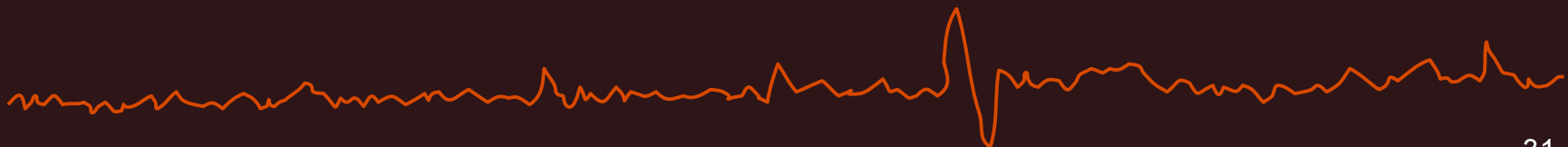
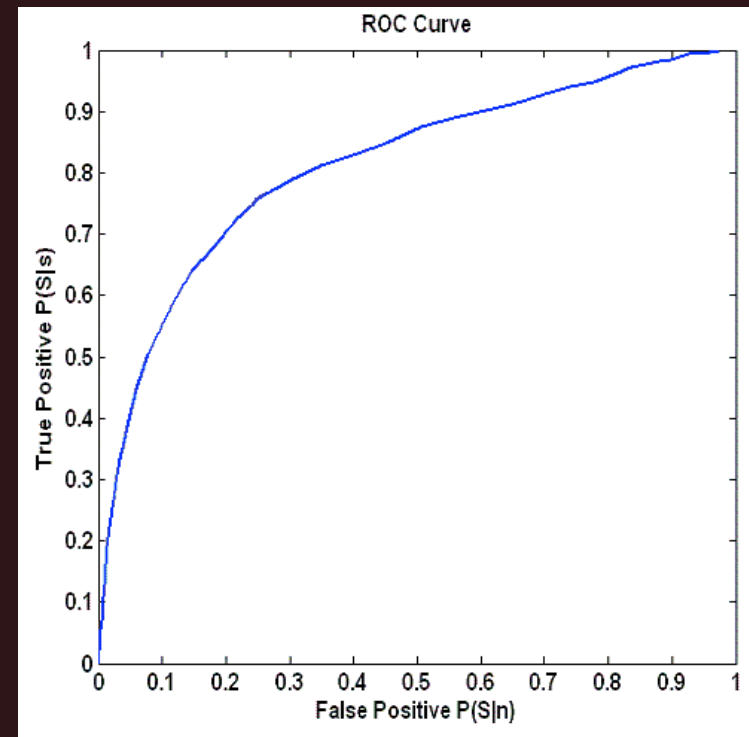


# Simple Threshold



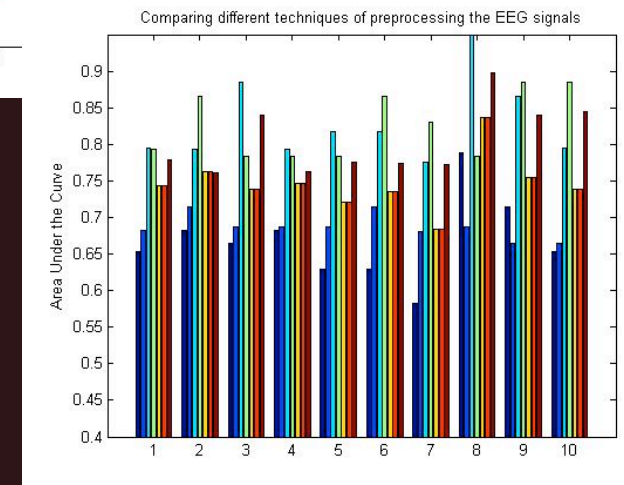
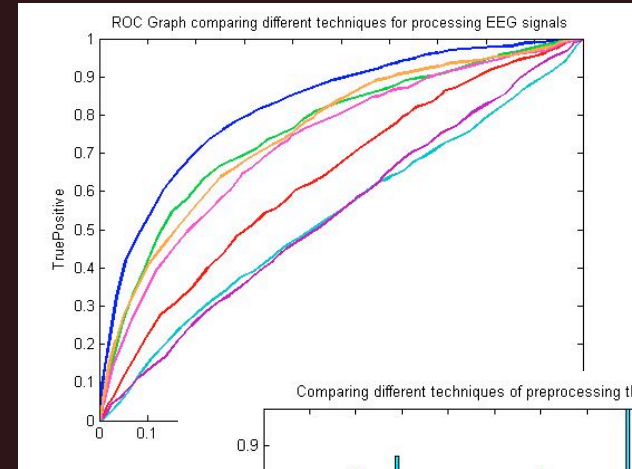
# Receiver operating characteristics (ROC)

- Simple thresholding algorithm
- False positive vs True positive rate
- Several observations yield one point
- Each point represents a decision strategy (threshold)
- Area under the ROC curve represents percentage correct

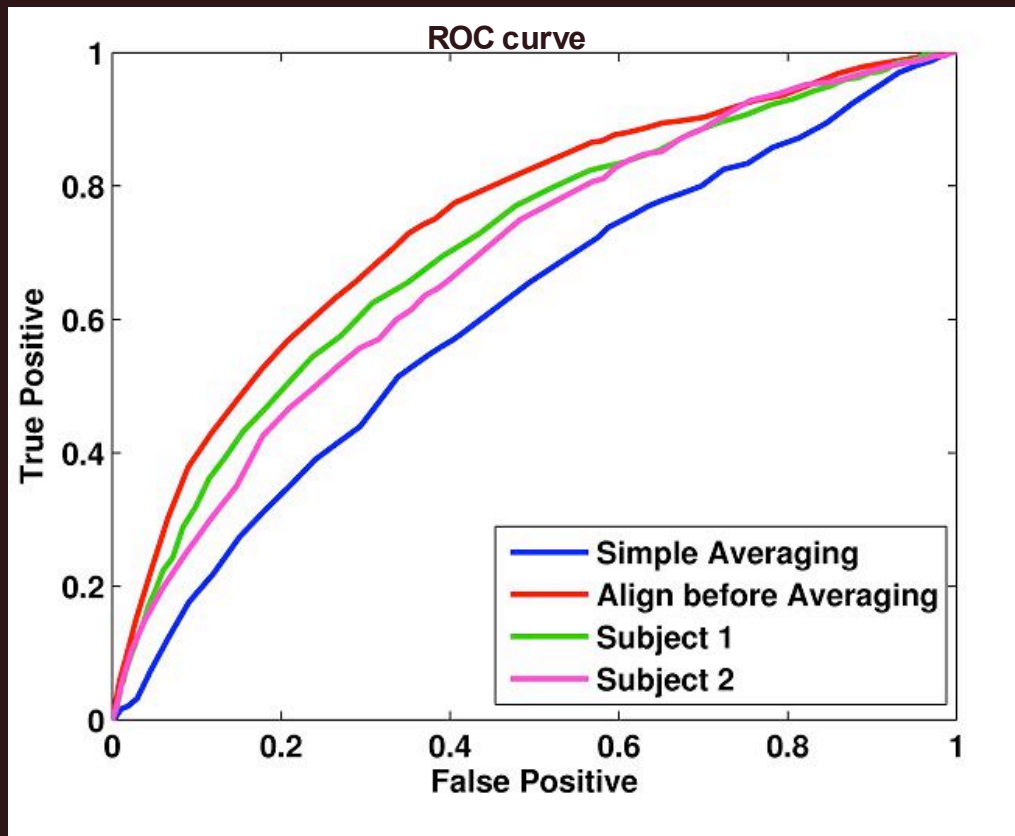


# Results

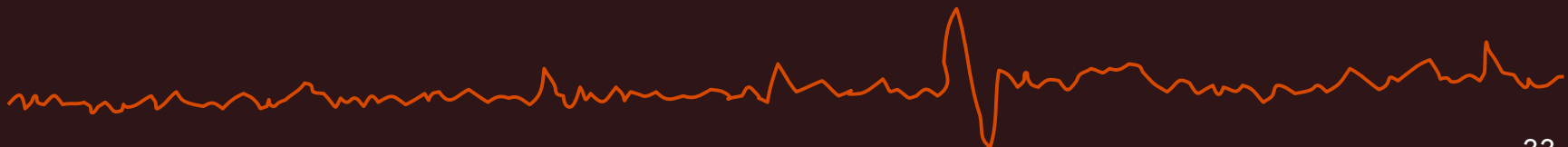
- Single observers:
  - Individual data
  - Averaged in time
  - Template matching
- Combining across observers:
  - Simple averaging
  - Time aligning
  - Template matching
  - Aligning and Template matching
- Inter-subject vs Intra-subject combination of EEG
- Component and Channel data



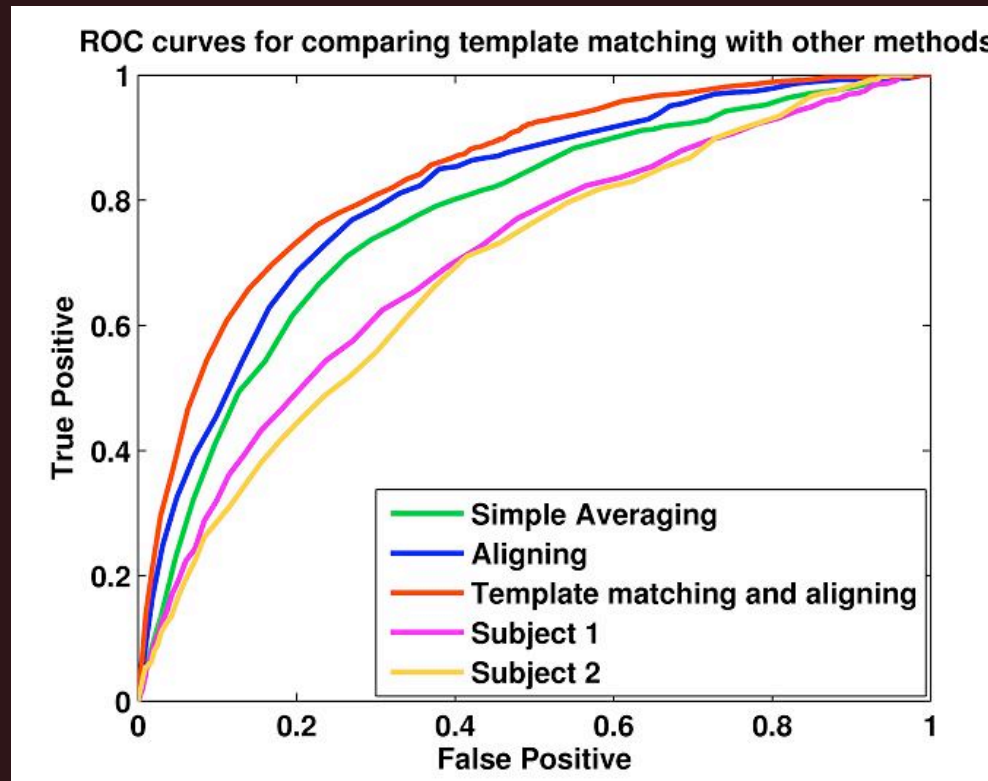
# Results - Aligning



- Simple averaging is sometimes worse
- Aligning attenuates the inter-subject ERP latency variation

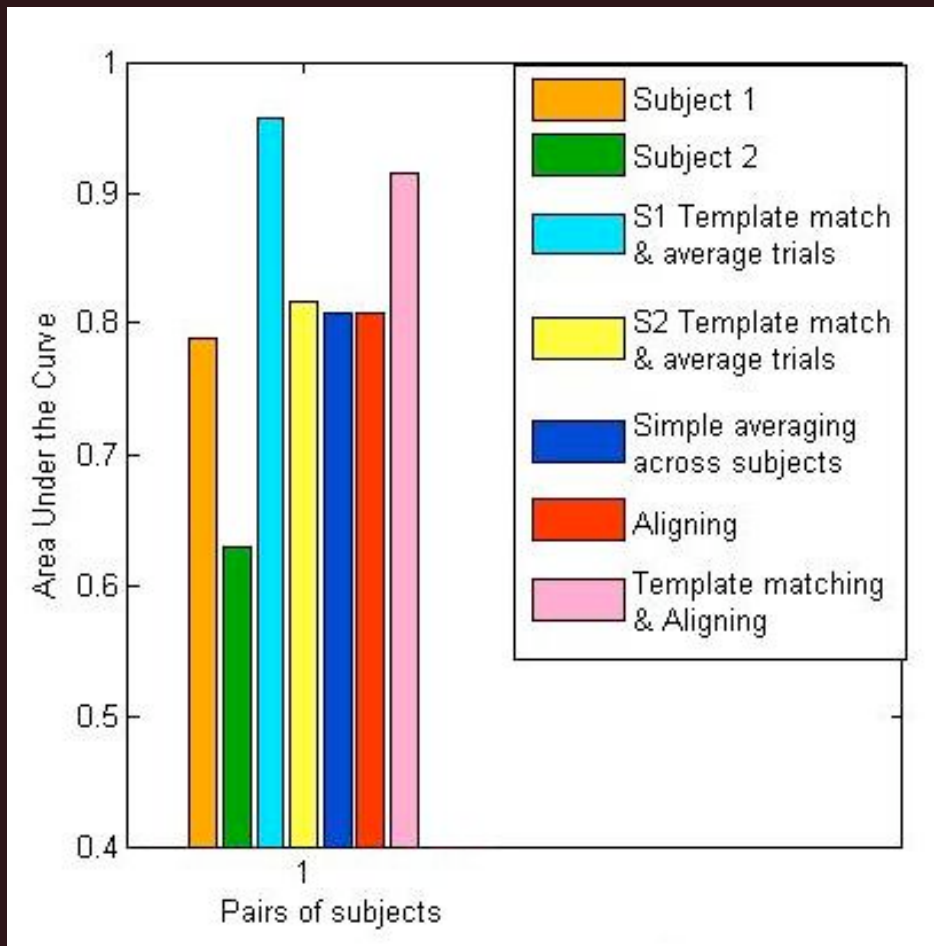


# Results – Template Matching



Gain	Mean Gain	Max Gain
Over single observer	6.2 %	19.2 %
Over simple average of a pair of observers	8.5 %	30.9 %

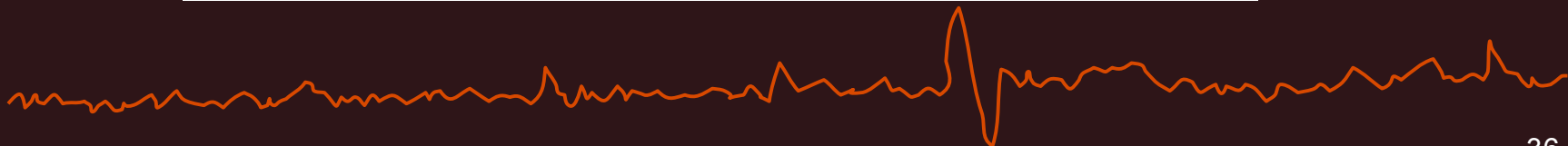
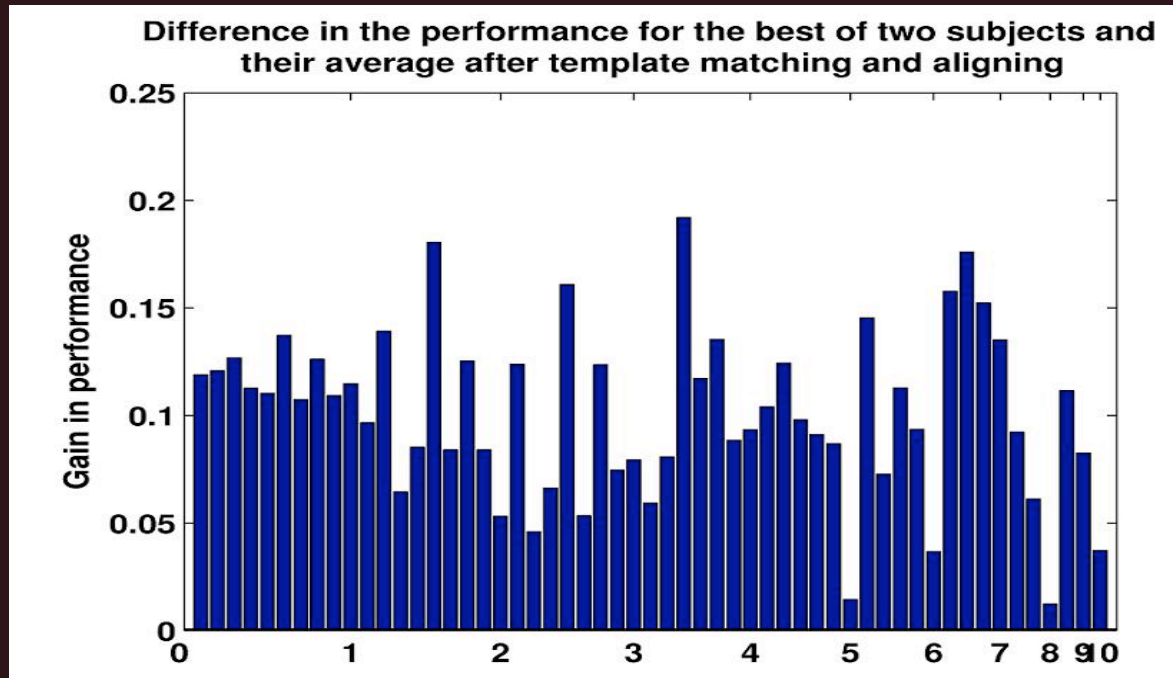
# Results - Area under ROC curves



Different techniques of combining EEG signals and their performance (Shown only two pairs of subjects)

# Results – Template Matching

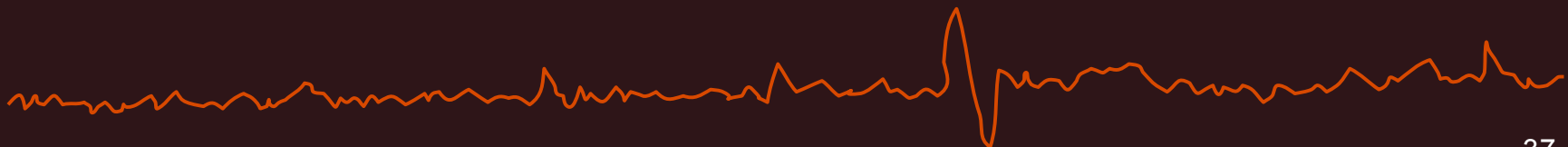
Gain over single observers on combining multiple observer EEG after aligning & template matching

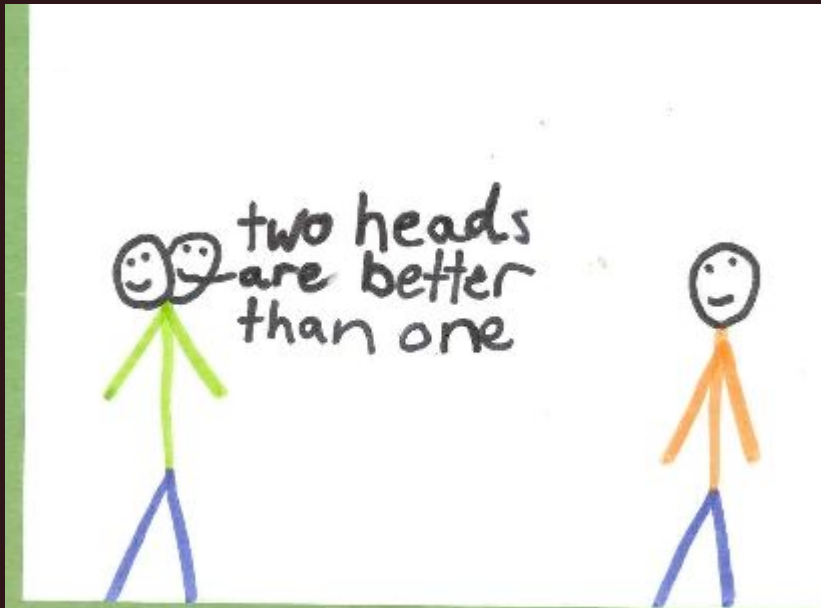




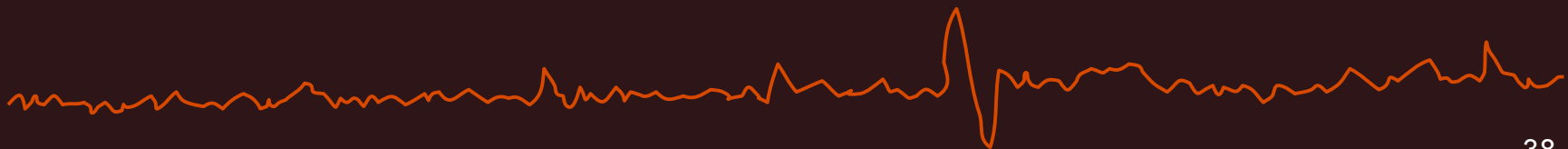
# Conclusions

- Combining single trial EEG across subjects it is possible to boost the signal-to-noise
- The more we compensate for inter-subject differences, the better the performance
- Some pairs of subjects are similar and result in greater boost in performance





Thank You!



# References

## Group Performance

- T. H. Schafer. Detection of a signal by several observers. Technical Report 101, Naval Electronics Laboratory, San Diego, 1949.
- D. M. Green and J. A. Swets. *Signal Detection Theory and Psychophysics*. Robert E Krieger Publishing Company, 1966.
- R. A. Baker, J. R. Ware, and R. R. Sipowicz. Signal detection by multiple monitors. *The Psychological Record*, 12:133–137, 1962.
- N. Sebanz, G. Knoblich, and W. Prinz. Representing others' actions: just like one's own? *Cognition*.

## EEG

- S. Makeig, A. Delorme, M. Westerfield, T.-P. Jung, J. Townsend, E. Courchesne, and T. J. Sejnowski. Electroencephalographic brain dynamics following manually responded visual targets. *Public Library of Science - Biology*, 2(6).



# References

- J. Wolpaw and D. McFarland. EEG based brain computer communication. *Electroencephalography and Clinical Neurophysiology*, 90:444–449, 1994.
- M. Steriade, P. Gloor, R. R. Llinas, F. H. L. da Silva, and M. M. Mesulam. Basic mechanism of cerebral rhythmic activity. *Electroencephalography and Clinical Neurophysiology*, 76.
- S. Makeig, A. J. Bell, T.-P. Jung, and T. J. Sejnowski. Independent component analysis of electroencephalographic data. *Advances in Neural Information Processing Systems*, 8:145–151, 1996.
- A. Hyvarinen. Survey on independent component analysis. *Neural Computing Surveys*, 2:94–128, 1999.
- J. Cohen and J. Polich. On the number of trials needed for p300. *International Journal of Psychophysiology*, 25:249–255, 1997.

