

# Support Vector Machines applied to EEG: a reliable pipeline for tough problems

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## Context

- Machine learning methods based on EEG measurements are increasingly popular in clinical fields.
- Potential outcome: moving from group-level statistical results to personalized diagnosis.

## Purpose

The object of the present study is to address the following issues:

### 1- Comparing several features extraction methods,

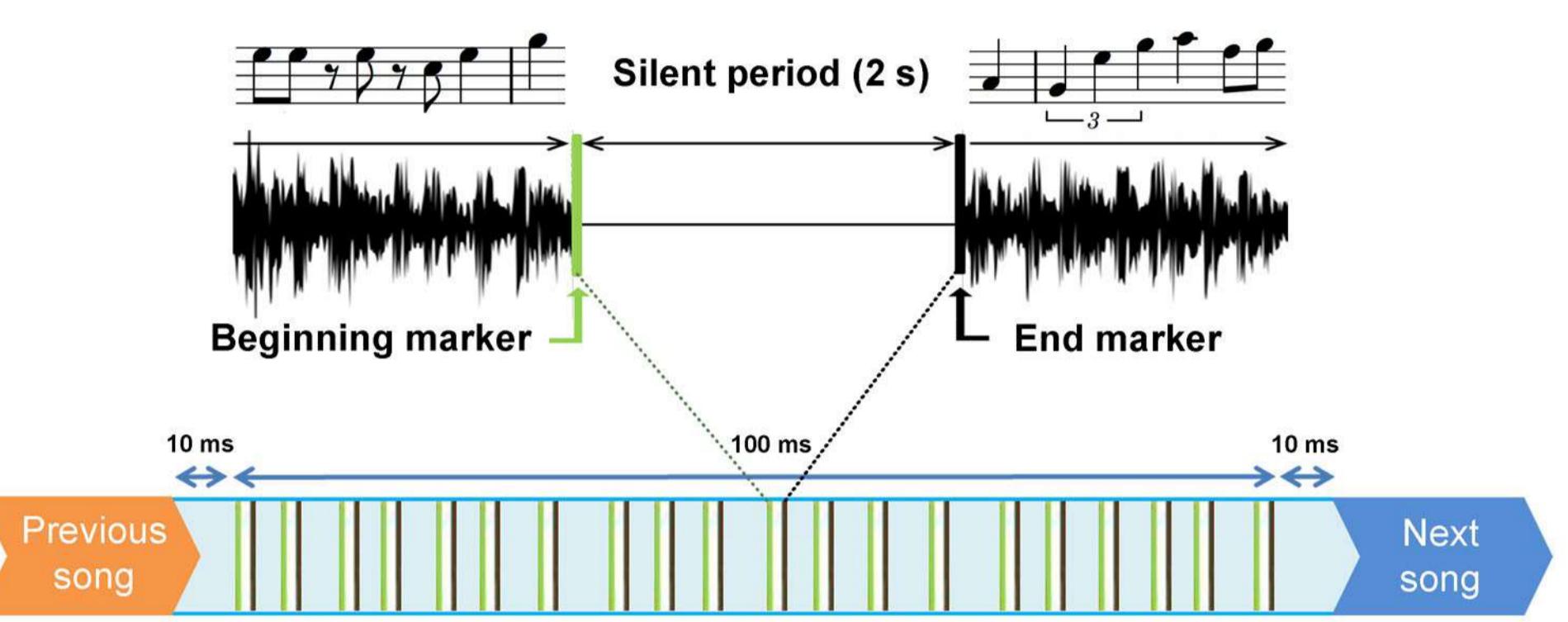
### 2- Highlighting the importance of hyperparameters choice in standard methods like SVM,

### 3- Providing a pipeline allowing for reliable results with no overfitting.

## Data acquisition



EEG recording using  
Electrical Geodesic Sensor Net,  
257 channels.



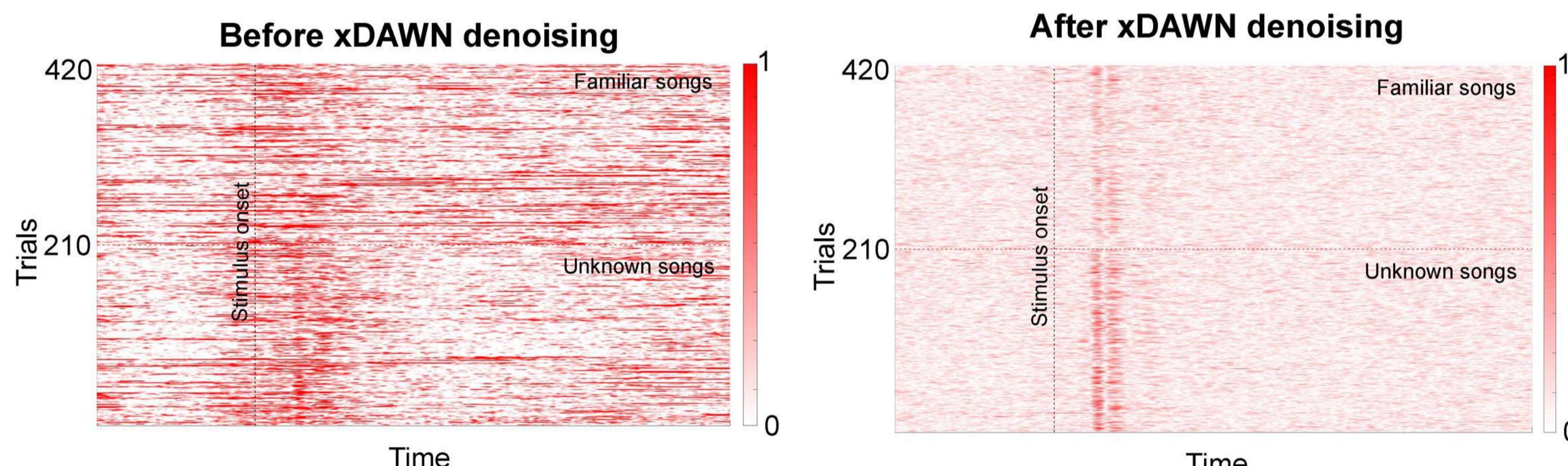
High resolution ElectroEncephaloGraphy: 257 channels  
Sampling rate: 1000 Hz Bandpass: 1-30 Hz

20 subjects listening to 10 familiar songs vs 10 unknown songs  
21 silent periods in each song → spontaneous auditory imagery

## xDAWN denoising

Raw EEG contains the desired ERP, but also:

- unrelated activity of the brain
  - muscular and/or ocular artifacts
- signal-to-noise ratio very low
- detection of target stimuli difficult from single trial

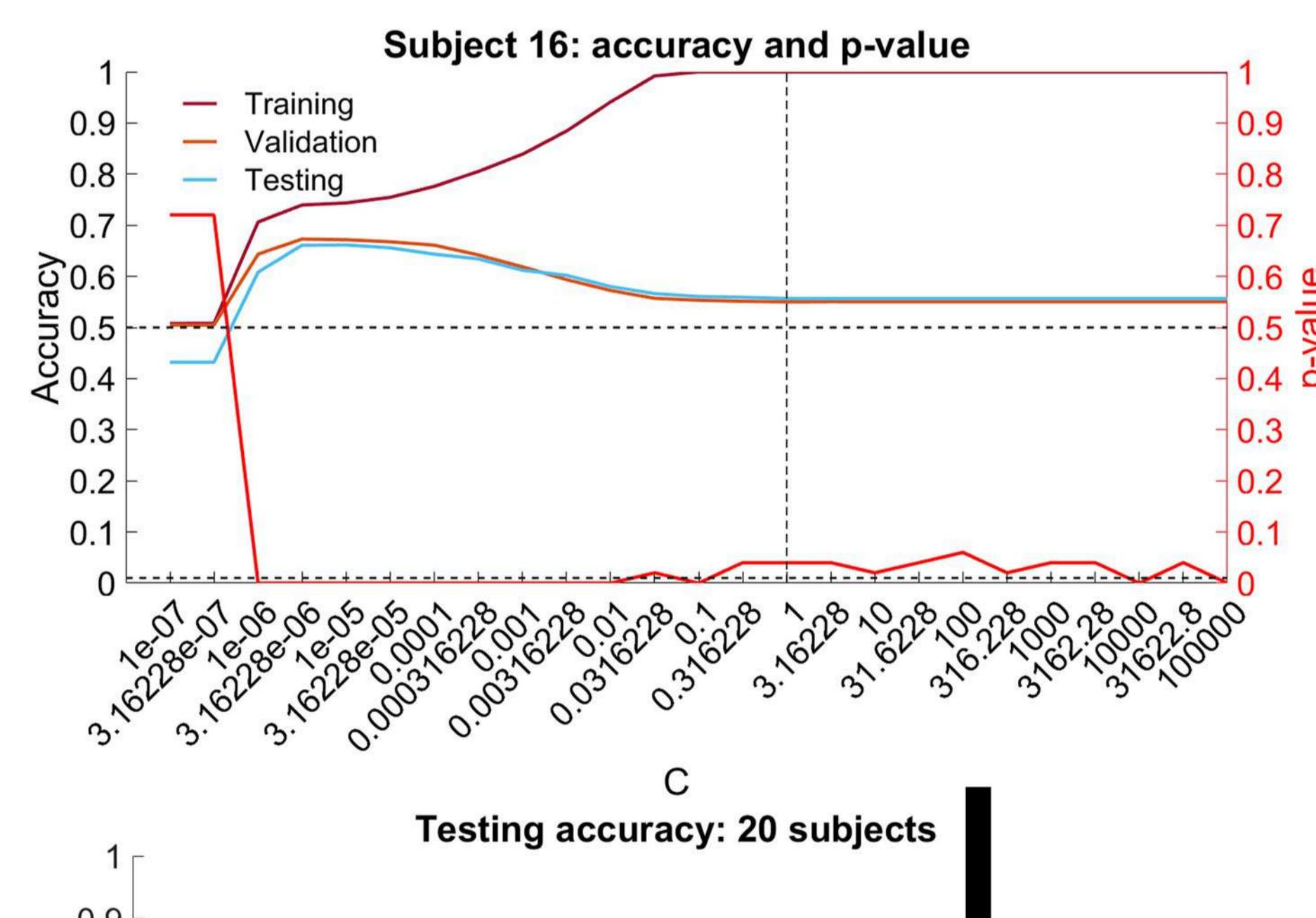


xDAWN algorithm to estimate Evoked Response Potentials subspace:  
 1. Compute QR factorization of X (all trials concatenated):  $X = Q_X R_X$   
 2. Compute QR factorization of D (Toeplitz matrix):  $D = Q_D R_D$   
 3. Compute Singular Value Decomposition of  $Q^T D Q_X: Q^T D Q_X = \Phi \Lambda \Psi^T$   
 4. Select component  $(\Phi_i, \Psi_i)$  associated with largest singular value  $\lambda_i$   
 5. Estimate spatial filter:  $(\hat{u}_i, a^*) = R_X^{-1} \Psi_i, R_D^{-1} \Phi_i \lambda_i$

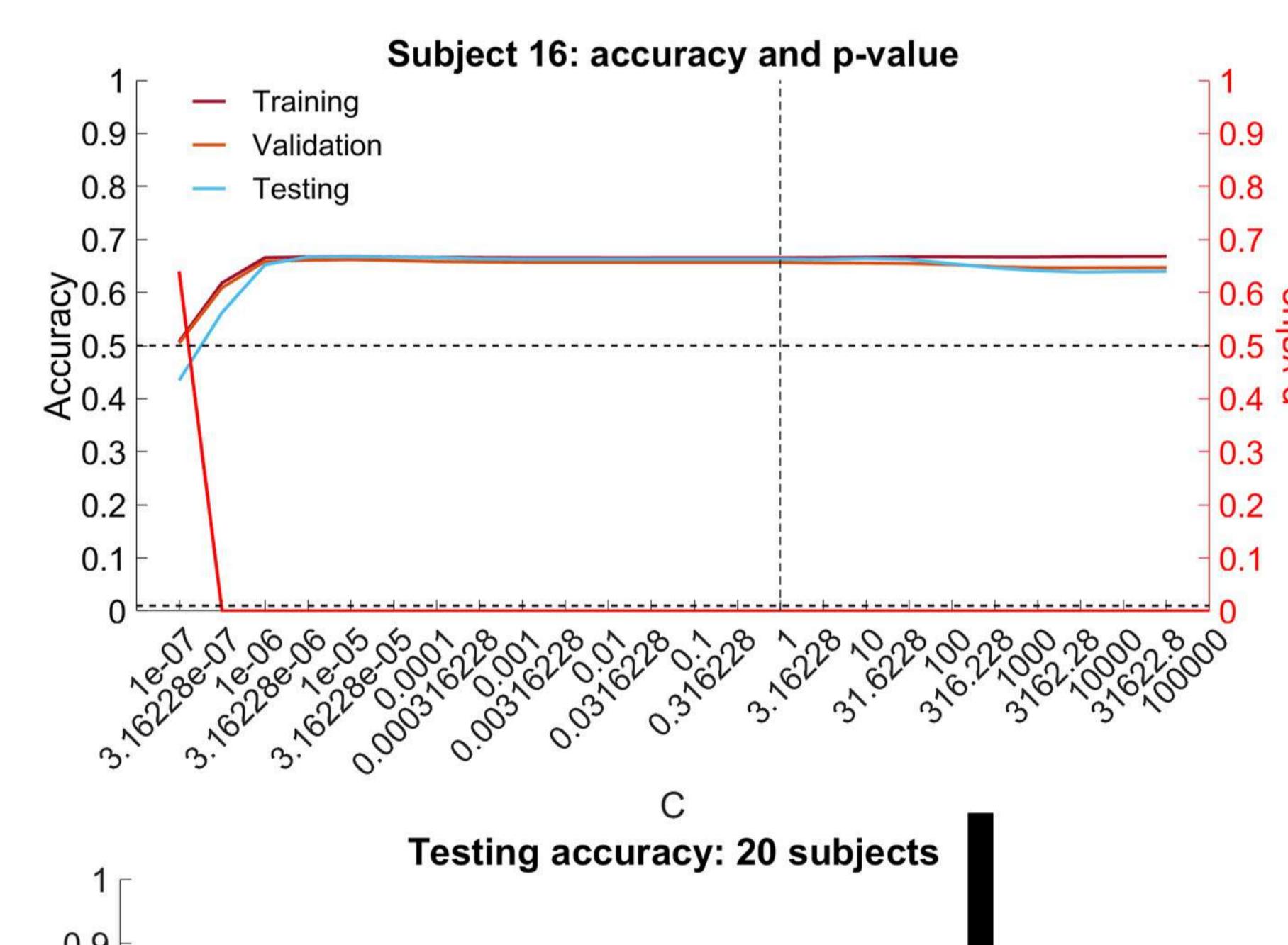
## SVM classification: linear kernel

Temporal domain: Global Field Power

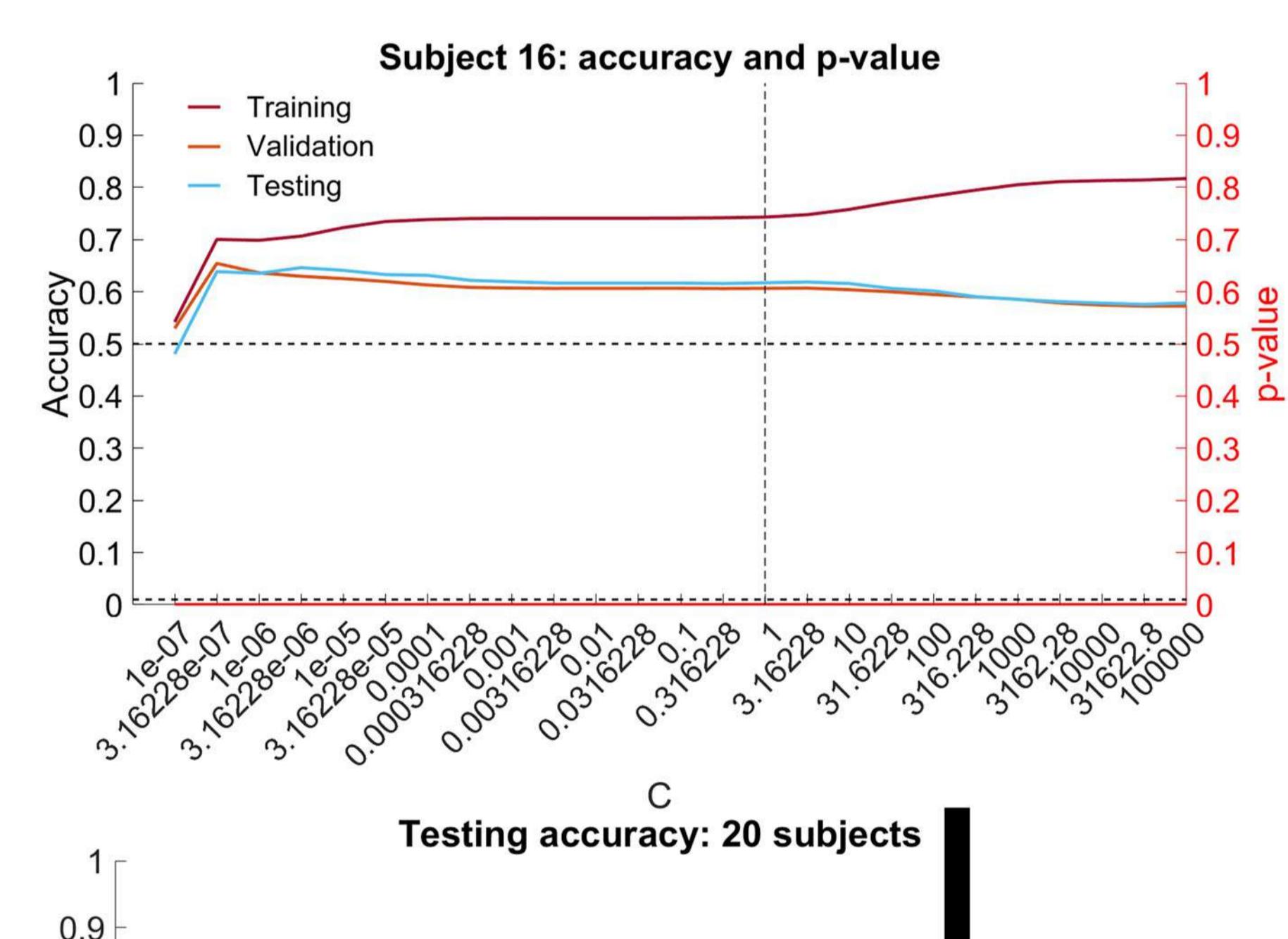
$$GFP = \sqrt{\sum_{i=1}^{nb\_electrodes} \text{AverageReferencedData}_i^2}$$



Frequency domain: Fourier transform on 4 bands  
 $\theta$  band: 4-8 Hz,  
 $\alpha$  band: 8.5-12 Hz,  
 $\beta_1$  band: 12.5-18 Hz,  
 $\beta_2$  band: 18.5-30 Hz.



Frequency domain: Fourier transform  
All frequency powers 1-30 Hz



## Future directions

- Selection of spatial regions of interest
- Use of genetic algorithm to optimize the window of interest

- Radial Basis Function as kernel
- Benchmark for Echo State Network

## Contact



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