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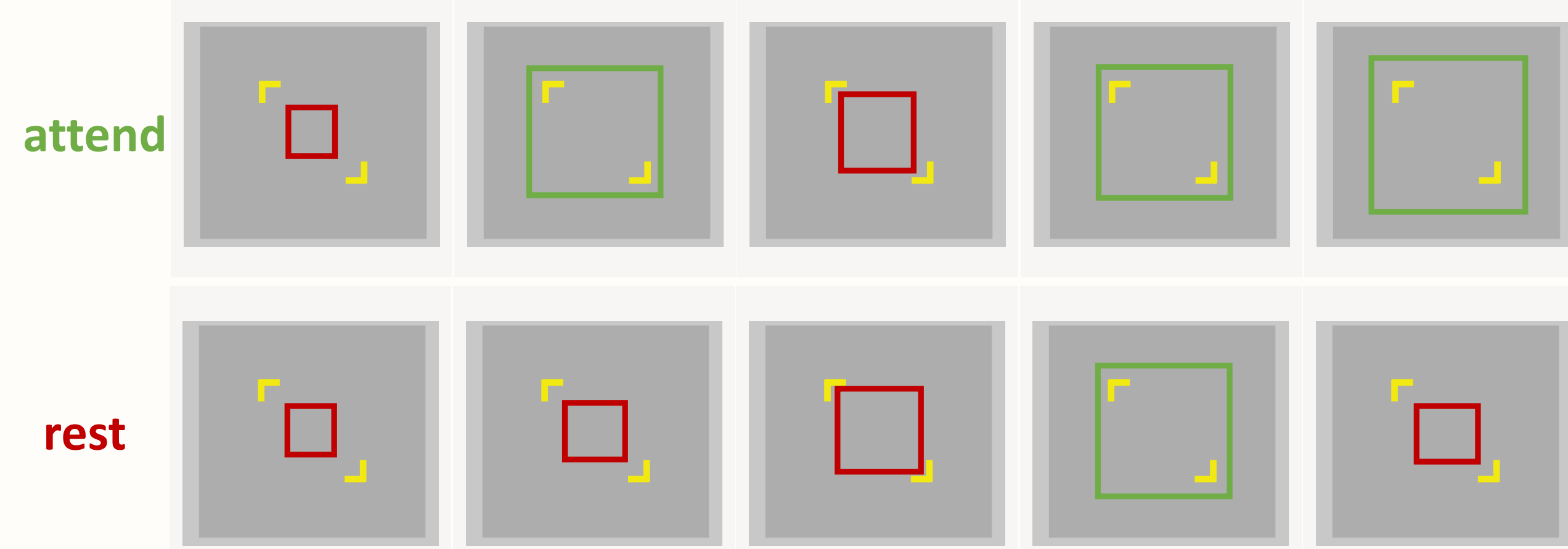
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Introduction

- Helping people with complex communication disorders (CCDs) communicate requires establishing cause and effect between the assistive technologies and their users
- Steady-state visual evoked potentials (SSVEP)-based brain-computer interfaces (BCIs) can help us achieve this goal
- We present the cumulative sum canonical correlation analysis (CCA_{CUSUM}) classifier to improve the cause-and-effect relationship between a user's behavior and the SSVEP-based BCI

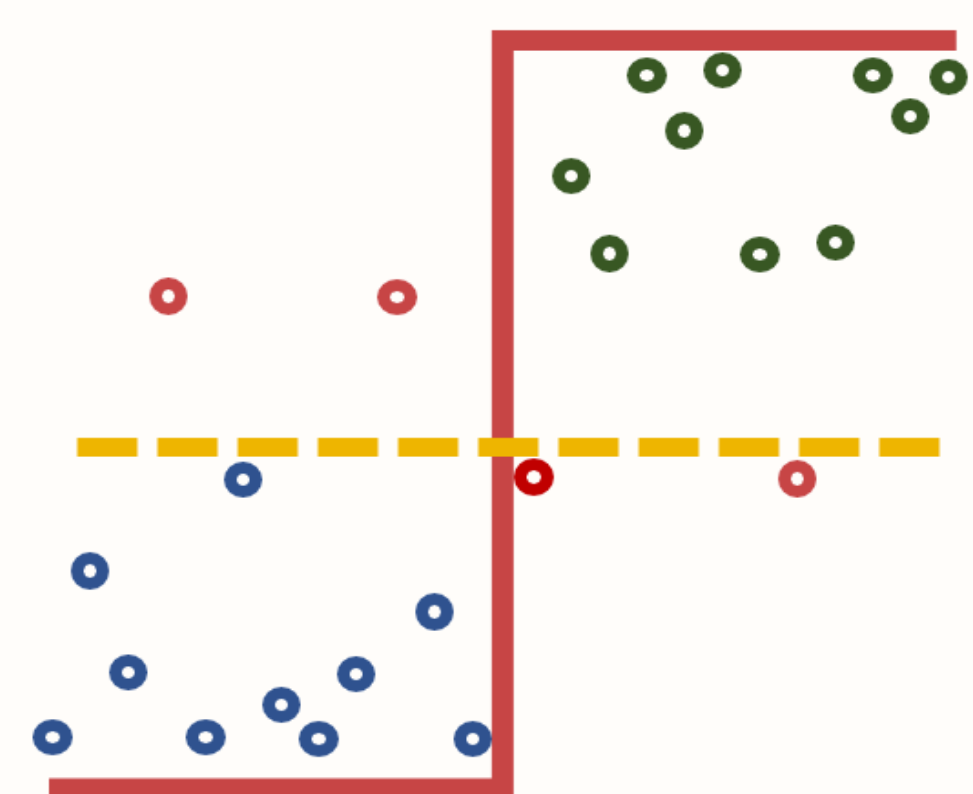
Methods: MusicBox

- MusicBox**: simple SSVEP-based BCI to help establish cause and effect for CCD individuals
- user can attend to a **flickering stimulus (attend state)** or be **at rest (rest state)**
- provides auditory feedback if user is in attend state by *playing music*



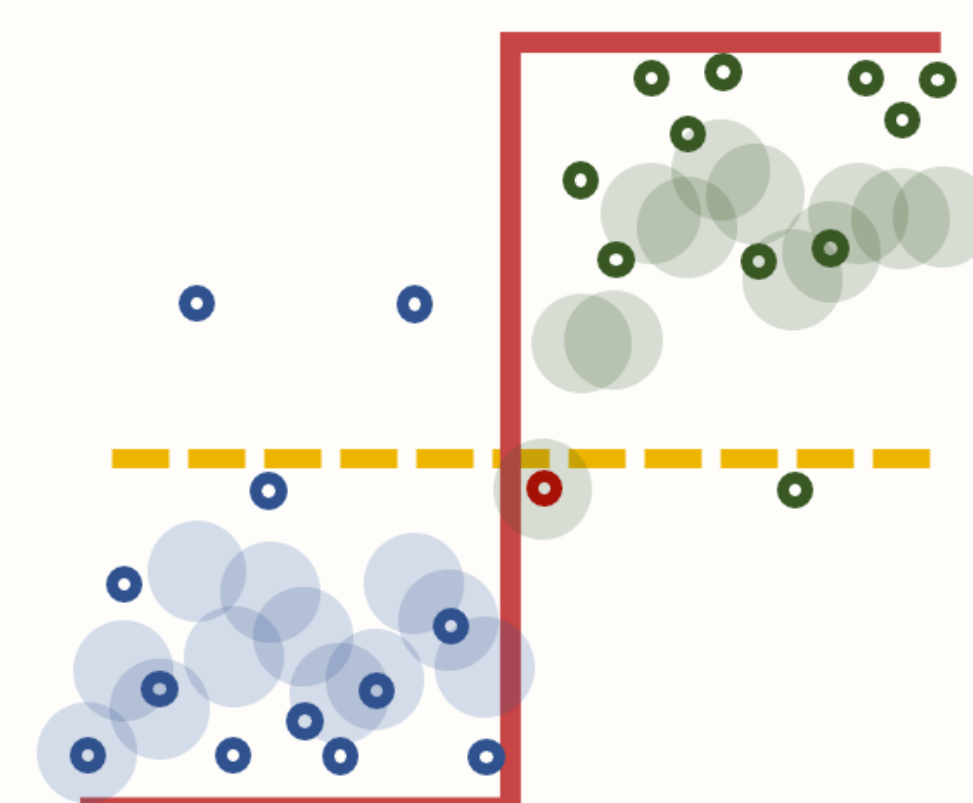
Methods: Conventional Approach

- Conventional Approach**: compares samples with fixed threshold to identify whether a user is in attend or rest state
- Feature extraction based on canonical correlation analysis (CCA)
- Discard Context → Potential information loss



CCA_{CUSUM}

- CCA_{CUSUM}** : identifies states by detecting transitions between them
- Use **CCA_{CUSUM}** with sliding window to extract features from EEG signals
- Assumption**: normal distribution with parameters $\theta_0 = (\mu_0, \sigma)$ and $\theta_1 = (\mu_1, \sigma)$ before/ after change



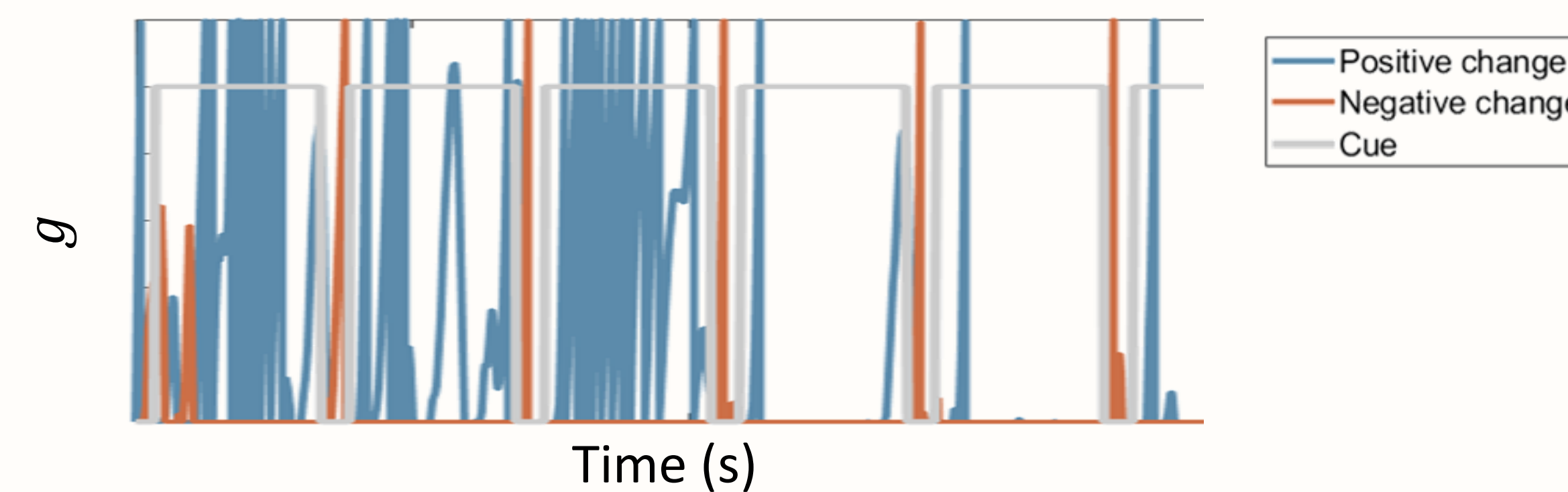
CCA_{CUSUM}

- Using hypothesis testing, we obtain two different tests:

$$g_i^+ = \max\left(g_{i-1}^+ + r_i - \mu_0 - \frac{v}{2}, 0\right) \geq h$$

$$g_i^- = \max\left(g_{i-1}^- - r_i + \mu_0 - \frac{v}{2}, 0\right) \geq h$$

- r_i : canonical correlation of i^{th} sliding window
- $v = |\mu_1 - \mu_2|$: magnitude of change
- h : fixed threshold
- g_i^+ detects positive changes
- g_i^- detects negative changes
- Learn online model parameters μ_0 and v



Performance Metric

- Proposed metric M**: measure cause-and-effect relationship between MusicBox and user's behavior

$$EPOR = \frac{1}{l_{ON}} \times E \left[m \left| \begin{array}{l} r_k \text{ correctly classified as positive} \\ \forall k \in \{i+1, \dots, i+m\} \end{array} \right. \right]$$

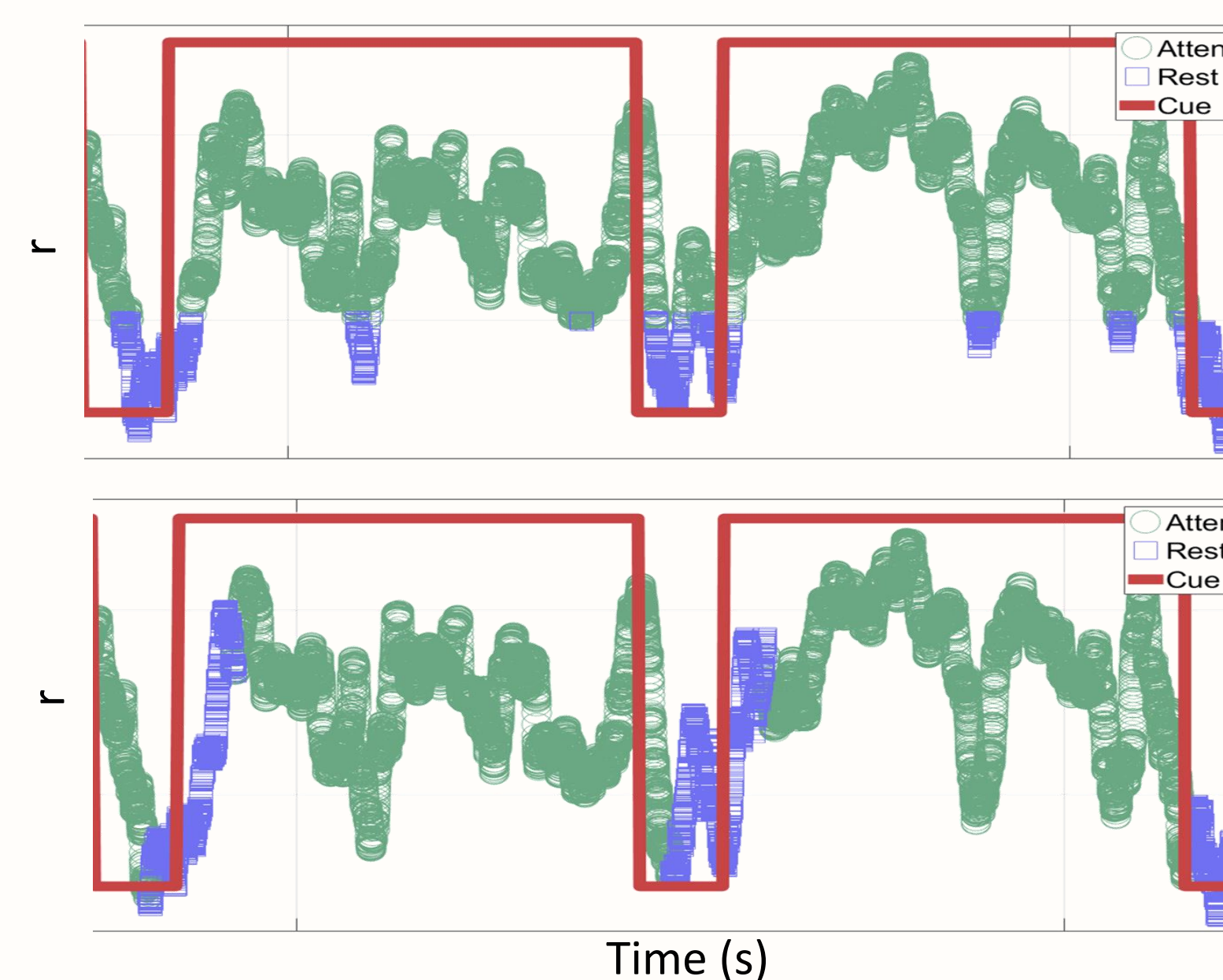
$$ENOR = \frac{1}{l_{OFF}} \times E \left[m \left| \begin{array}{l} r_k \text{ correctly classified as negative} \\ \forall k \in \{i+1, \dots, i+m\} \end{array} \right. \right]$$

$$M = \frac{2 \times EPOR \times ENOR}{EPOR + ENOR}$$

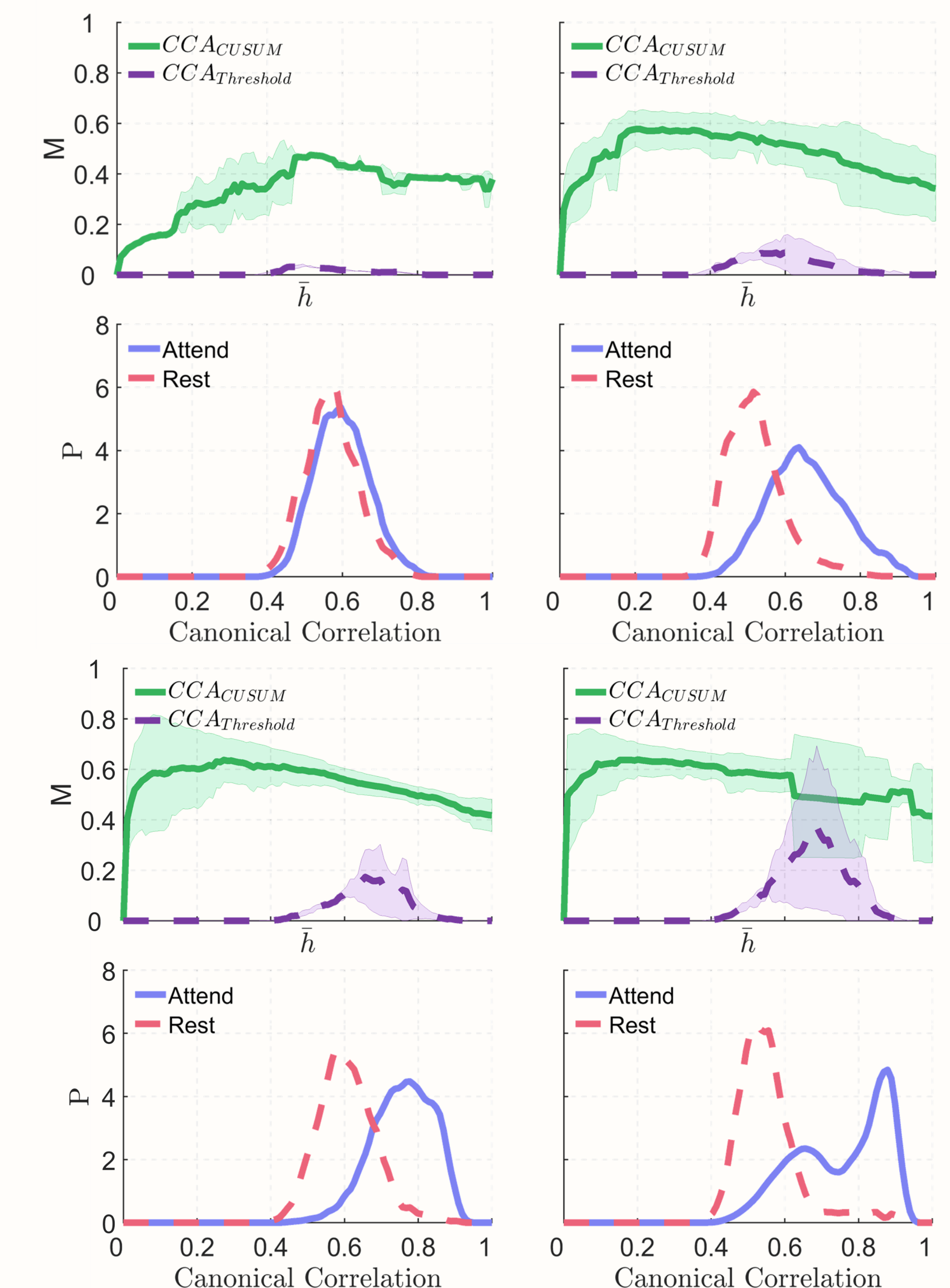
- l_{ON} / l_{OFF} : length of intervals where the participant was in attend/rest state

Data and Results

- (right) Conventional method vs CCA_{CUSUM} in a single trial experiment
- Four participants with no CCD
- Eight EEG channels
- Multiple trials per participant



Data and Results



(top) Comparing metric 'M' for CCA_{CUSUM} and the conventional method
(bottom) The distribution of samples under the two hypotheses

Summary

- CCA_{CUSUM} : Using change detection to improve cause and effect in BCIs
- We evaluated CCA_{CUSUM} using MusicBox, an SSVEP-based BCI
- CCA_{CUSUM} improved cause and effect when samples were less separable

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References are available upon request.

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