

Impact of local Laplacian Spatial Filters on C-VEP-Based BCI Performance

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Introduction: Brain-Computer Interfaces (BCIs) enable direct communication between the brain and external devices, offering potential for assistive communication [3]. It can be possible only if the brain responses can be reliably decoded over time. Code-Modulated Visual Evoked Potentials (C-VEPs) stand out for their high ITRs and minimal training requirements, leveraging m-sequences to elicit distinct neural responses [1]. However, real-time decoding faces challenges due to the non-stationarity properties of the EEG signal and its low spatial resolution [2]. We compare different Laplacian spatial filters, including weight calculation methods such as $1/d$, $1/d^2$, $1/\log(d)$, and $1/\sqrt{d}$ where d is the orthodromic distance between two sensors, for enhanced signal quality and robust classification techniques, achieving improved accuracy and advancing the potential of non-invasive BCIs [2].

Material, Methods and Results: The study utilized a C-VEP paradigm with stimuli based on 63-symbol m-sequences [1]. EEG data was collected from eight sensors (O_1 , O_2 , P_z , P_3 , P_4 , PO_7 , PO_8 , and O_z) with 13 participants focusing on visual stimuli across five sessions [3]. Preprocessing incorporated Laplacian spatial filtering, enhancing spatial resolution by emphasizing local neural activity [2]. Classifiers (LDA, BLDA, MLP, SVM, and k-nn) were compared using correlation-based features [4]. Testing different Laplacian methods, the $1/d$ weighting approach and a radius of 1 yielded the highest accuracy [2], and SVM showed consistently higher accuracy compared to other classifiers, demonstrating superior signal enhancement and classification performance [4]. These findings underline the critical role of advanced preprocessing and classifier selection in achieving reliable and high-speed BCIs.

Conclusion: This study highlights the effectiveness of Laplacian spatial filtering, achieving an accuracy of $76.0\% \pm 18.9$ with the $1/d$ weighting method and an SVM classifier, compared to a baseline of $59.7\% \pm 17.3$ without Laplacian filtering. The approach shows how the preprocessing steps substantially impact accuracy compared to the type of classifier used.

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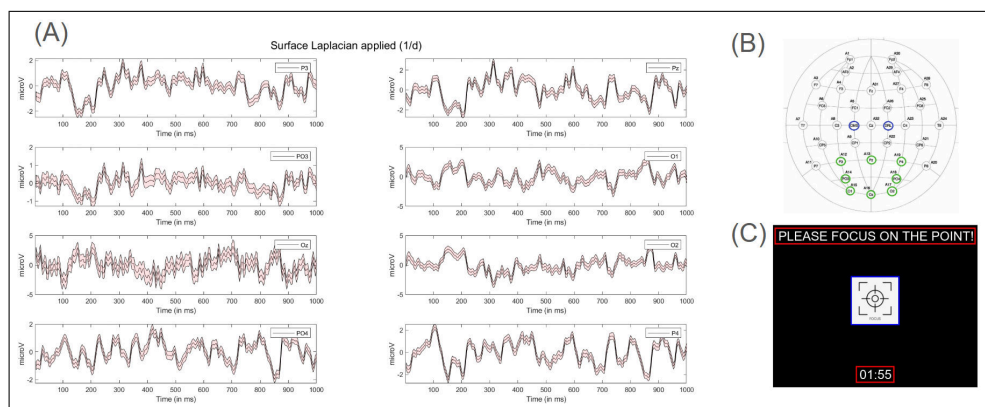


Figure 1: (A) Time-domain EEG plots showing mean signals with variability bounds, highlighting consistency in neural responses for template-based feature extraction. (B) EEG cap layout highlighting VEP electrodes (green) and ground (blue). (C) Experimental paradigm.

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