

A New Statistical Model of EEG Noise Spectra for Real-time, Low- γ -band SSVEP Brain-Computer Interfaces

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Introduction: A major impediment to practical real-time γ -band ($\geq 30\text{Hz}$) SSVEP BCIs is the high level of spectral noise which dramatically increases the error rates of frequency detectors/estimators (Fig. 1a). The standard “ $1/f$ -type” spectral model [1] of EEG noise is both theoretically unsatisfactory and too ill-defined for hypothesis tests. Based on our new theory of *quantum ion channel kinetics* [2], we model EEG noise spectra as random processes of the form $S_{\text{EEG}}(f) = S_{\text{GVZM}}(f) \cdot \Xi(f)$, where $\Xi(f)$ are independent $\chi^2(2)/2$ random variables at each frequency f and $S_{\text{GVZM}}(f)$ is the *generalized van der Ziel-McWhorter* deterministic function whose inverse Fourier transform is $R_{\text{GVZM}}(t) = P_0 \int_{\tau_1}^{\tau_2} (1/\tau^\alpha) e^{-t/\tau} d\tau + P_1 \delta(t)$ for tunable parameters $\alpha, \tau_1, \tau_2, P_0, P_1$ (Fig. 1b). We show such noise models have superior statistical characteristics for BCI and other neuroengineering applications.

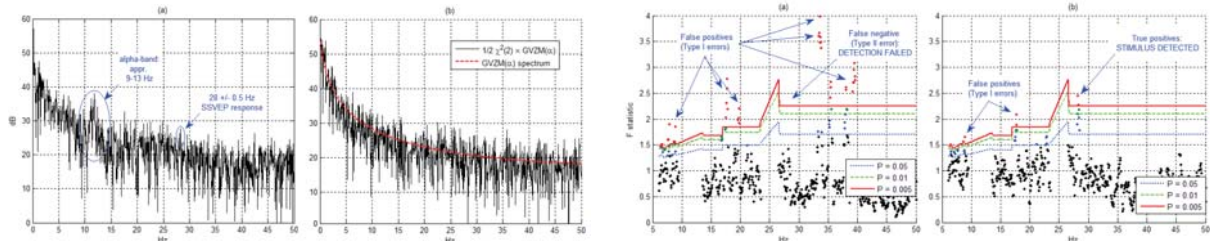


Figure 1. (a) Raw single-trial EEG spectrum from 28Hz SSVEP BCI experiment showing a response peak which is nearly indistinguishable from background noise. (b) Synthetic GVZM $\cdot \chi^2(2)/2$ noise spectrum optimally-fitted to the data of (a).

Figure 2. (a) Critical levels for SSVEP detection/estimation using standard smoothed periodogram algorithm [4] and the data of Fig. 1. (b) Detection/estimation using optimally-fitted GVZM $\cdot \chi^2(2)/2$ statistics.

Material, Methods and Results: The model was tested on a 15-second, 28Hz SSVEP trial (Fig. 1a) from a publicly-available BCI dataset [3]. Biosemi electrodes A14-A16, A21-A23, A25, A27-A29 were averaged to form a virtual visual electrode. Blink artifacts were estimated by linear regression onto the three frontal electrodes. A popular F -test SSVEP detection algorithm [4] was compared to the same algorithm with its pre-stimulus estimator replaced by our optimally-fitted GVZM $\cdot \chi^2(2)/2$ statistic. Each spectral value (excluding mid- α - and low- β -bands) was classified with respect to its F -test critical value calculated from the null hypothesis of no stimulus at that frequency. The results are shown in Fig. 2.

Discussion: The standard algorithm [4] failed to detect the 28Hz response spike in the noise background and also produced numerous false positives (Fig. 2a). On the other hand, our GVZM-based algorithm not only accurately detected the 28Hz response with $P < .005$, it also produced far fewer false positives (Fig. 2b).

Significance: This work proves that it is feasible to detect/estimate low- γ -band SSVEP spikes in real-time despite their poor signal-to-noise characteristics by using neurologically-appropriate statistics for EEG background noise. Such noise models will be essential for the development of future practical real-time SSVEP BCIs in the mid- γ -band.

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References

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