

Automated Seizure Detection with Machine Learning

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Background & Unmet Need

- Seizures occur in up to 50% of critically ill patients with altered consciousness, and >80% present with no obvious clinical sign of motor activity
- These non-convulsive seizures (NCS) have associated high morbidity and mortality in critically ill patients, and warrant prompt detection and treatment
- A continuous electroencephalogram (cEEG) is the gold standard for diagnosing NCS but is resource intensive and only reviewed intermittently (often 2-3 times daily) rather than continuously monitored
- Quantitative EEG (qEEG) tools apply digital signal processing techniques to facilitate cEEG interpretation, but require lengthy clinician training and are limited in types of seizures detected
- **Unmet Need:** Access to rapid, accurate, and automated continuous EEG seizure detection

Technology Overview

- **The Technology:** Method for continuous automated seizure detection based on artificial neural network recognition of seizure patterns on a novel spectrographic display
- The method introduces the median power spectrogram (MPS), a novel qEEG spectrographic display which can consolidate multiple EEG channels into a single channel display and optimize temporal and frequency resolution, resulting in well visualized seizures
- Seizures appear as characteristic sloped harmonic bands on MPS that are visually distinct and easily identified with minimal clinician training (~5 min)
- A convolutional neural network (CNN) can be trained to recognize seizure patterns on the MPS and can automatically detect seizures in a continuous fashion
- **PoC Data:** The CNN models detected seizures with 80–90% sensitivity and specificity, on both adult and pediatric cohorts

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Patents:

[US Application Filed](#)

Publications:

[Yan et al. Seizure. 2019.](#)
[Yan et al. Seizure. 2017.](#)

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Cornell Reference:

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Technology Applications

- Continuous seizure telemetry monitoring at bedside for critically ill patients that can automatically alert the bedside clinicians to enable faster intervention
- A visual bedside display where seizures are easily recognized, and the clinician has the option to visually confirm the automated seizure detection
- A visual display for the neurophysiologist that supplements and expedites traditional EEG analysis

Technology Advantages

- Concise EEG visualization where seizures are easily recognizable, requiring less clinician training
- Rapid, accurate, and continuous automated seizure detection, in a real-time telemetry fashion, at the bedside that enables faster interventions
- A bedside display easily interpreted by a bedside clinician who can visually confirm automated seizure detection, and monitor response to treatment

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Supporting Data / Figures

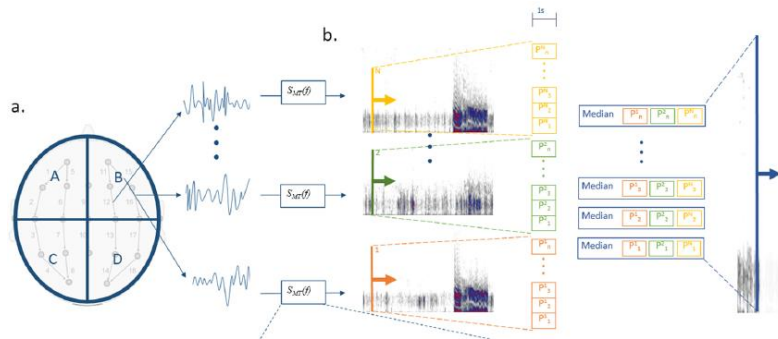


Figure 1: Overview of the signal processing method underlying the Median Power Spectrogram (MPS). Spectral content from individual EEG channel is obtained with application of multi-taper spectral estimation. The resulting spectral content from each channel is then combined via a median of the spectral power within each frequency bin at each second.

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Supporting Data / Figures

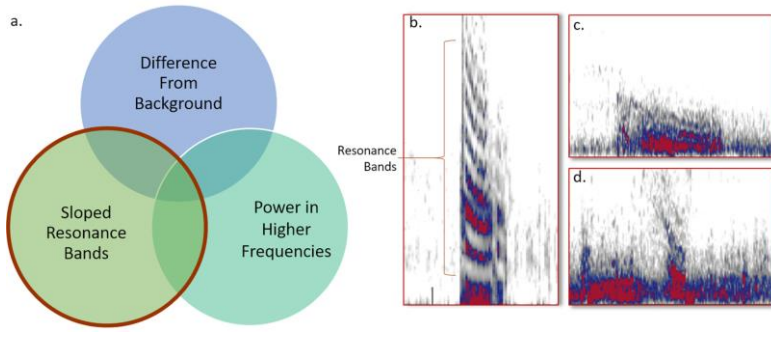


Figure 2: The three visual features that highlight seizures on the MPS (a). The sloped harmonic banding pattern that represent evolving rhythmic activity (b), characteristic of seizures, can be resolved/visualized even relative subtle seizures (c, d).

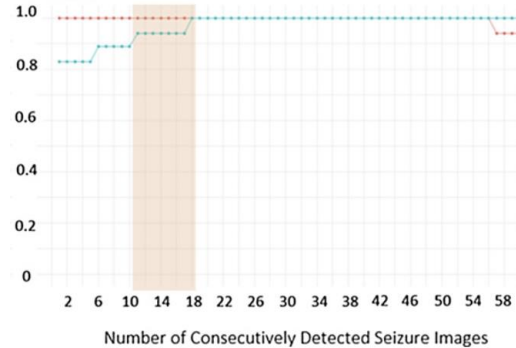


Figure 3: Seizure autodetection performance as a function of consecutively detected frames on a moving telemetry. The CNN demonstrates high sensitivity (red) and specificity (green) over a wide range of frames with the more practical range highlighted.

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