

Developing a Quantitative EEG-Based Biomarker for SLC6A1-Related Neurodevelopmental Disorder

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1 Background

- SLC6A1 genetic variants have an estimated prevalence of 1/35,000. 85% of patients have epilepsy (Goodspeed et al., 2023)
- A clinical trial evaluating therapeutic efficacy of glycerol phenylbutyrate for children with SLC6A1 mutations (NCT04937062) demonstrated significant reduction of seizures
- People with SLC6A1 have a characteristic EEG finding of paroxysmal delta activity, which may have sinusoidal, notched, or spike-wave morphology. Automatically capturing these bursts has the potential to serve as a biomarker for treatment response.

2 Aim

- Developing signal processing pipelines to quantify EEG changes before and after glycerol phenylbutyrate therapy

Review and Standards

- A pediatric epilepsy expert (ZG) marked paroxysmal delta bursts in 400 seconds of one pre-treatment EEG for use as a gold standard
- Reports from an ongoing clinical trial (NCT04937062) suggest these bursts decrease after treatment with glycerol phenylbutyrate
- This child was selected because they became seizure-free after treatment with glycerol phenylbutyrate.

Selection of Pipeline Parameters

- Existing literature was reviewed informally to develop knowledge of parameters utilized for other variants
- Expert clinician provided insight into EEG trends and related properties

Hypothesized Trends

- SLC6A1 seizure events tended to display (1) increased amplitude, (2) “slow waves” and (3) rhythmicity

Methods-

- Three spectral features were calculated at one second intervals:
- Delta Power: total power in from 2.5 - 4 Hz, represented in μV^2
- Autocorrelation: the closeness of the wave to itself, one second intervals, lag times implemented
- Spectral Entropy: distribution of power across frequencies 1-30 Hz

PIPELINE 1 METRIC	THRESHOLD
POWER (FULL INTEGRAL)	<i>greater than 105,000 μV^2</i>
AUTOCORRELATION	<i>greater than 0.41</i>
SPECTRAL ENTROPY	<i>less than 3.16</i>

Through trial and error, we selected thresholds and found when all three were met

5 Pipeline One Results

Results-

- Pipeline 1 identified 24/25 bursts
- Second-by-second PPV 0.99; sensitivity 0.984
- “Hits” from when all 3 thresholds were met in a one second interval

Green- Expert Certain,
Yellow- Potential Activity



Methods-

- Generalizability: optimized selection of the thresholds via exhaustive search maximizing
- $F1 = 2 * ((PPV \times Sensitivity) / (PPV + Sensitivity))$
 - Thresholds produced from this ultimate value utilizing an initial EEG for each patient
 - Sourced from existing literature
- The optimal solution was then applied to other EEGs for that patient to determine what percentage of seconds in each EEG met the three-part threshold

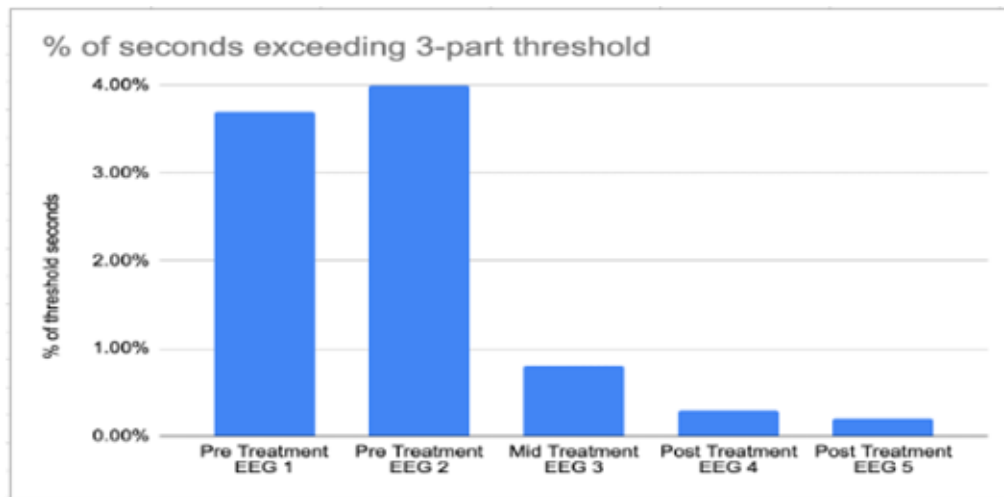
Delta Power Modification-

- Using a modified version of delta power where the amplitudes in the delta power range were squared to produce a vector, which we took the median of
 - Produces a different magnitude of value than Pipeline One's integration method

Results-

- Example Values Optimized for in Initial EEG:

PIPELINE 2 METRIC	THRESHOLD
MEDIAN POWER	<i>greater than 50 μV^2</i>
AUTOCORRELATION	<i>greater than 0.41</i>
NORMALIZED ENTROPY	<i>less than 0.3</i>



Fewer % seconds exceeding tri-threshold values post treatment

FIGURE 2. Pipeline 2. Comparing pre- and post-treatment EEGs, there was a substantial reduction in the percentage of seconds meeting the three metric thresholds.

Methods-

- Cross-correlation: the fit of the waveform across the eighteen channel pairs in each second-long time span (median value used)
 - Added to original 3 parameter model (delta power, autocorrelation, spectral entropy)
- Across the EEG, 25%ile, 50%ile, and 75%ile were calculated for each one second interval
 - 12 predictors total: 3 percentile levels * 4 variables

Model Comparisons-

- Logistic regression models were fitted to the data
 - Best Subset Model: Bayesian Information Criterion, weighting both fit and simplicity through number of variables (chosen from the 12 predictors)
 - Compared the median values of all 4 original parameters

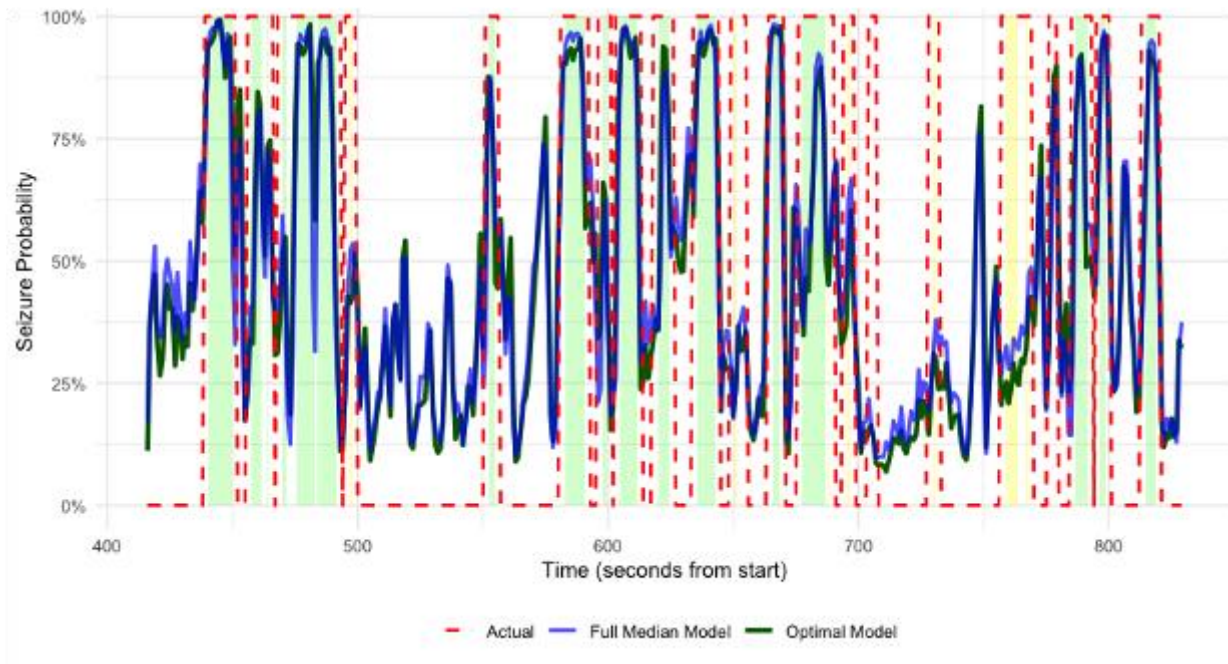


FIGURE 3. Pipeline 3. The full model (blue) and optimal model (green) generated probabilities that mirrored timing of events

Conclusion/Discussion

- We have developed a measure to detect SLC6A1-related EEG abnormalities and used this to discriminate pre-treatment from post-medication EEGs.
- Ongoing work is exploring additional features and generalizability across the full set of 10 patients from the glycerol phenylbutyrate clinical trial (NCT04937062).
- We will continue to refine our predictive models to improve individual EEG statistics such as PPV & Sensitivity, along with applicability to large samples of EEGs
- Further properties, such as wavelet forms, will be used to simplify correlation measures and improve generalizability.

Acknowledgements & Thanks



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