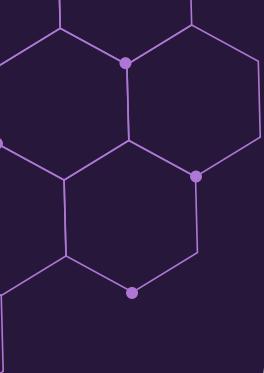


# **Evolutionary Algorithms for Neuromorphic Anomaly Detection**

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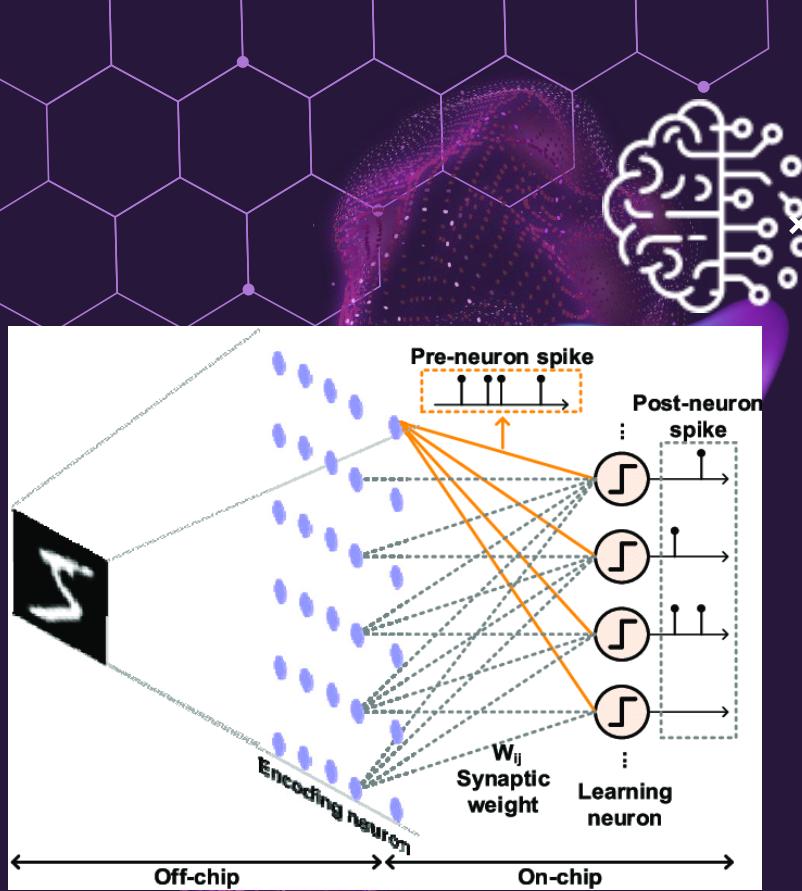
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# Neuromorphic computing and Evolutionary algorithms



Concept of SNN and its structure.

## Neuromorphic computing:

an approach to computing that mimics the way the human brain works. It entails designing hardware and software that simulate the neural and synaptic structures and functions of the brain to process information.

## Neuromorphic anomaly detection:

uses biologically inspired computing and hardware to detect unusual patterns in data with low power consumption and real-time processing. It combines techniques like Spiking Neural Networks (SNNs) and event-based sensors to analyze data from sources

## Evolutionary algorithms:

stochastic search methods that are inspired by biology. They operate on a population of potential solutions applying the principle of survival of the fittest to produce approximations that converge to a solution

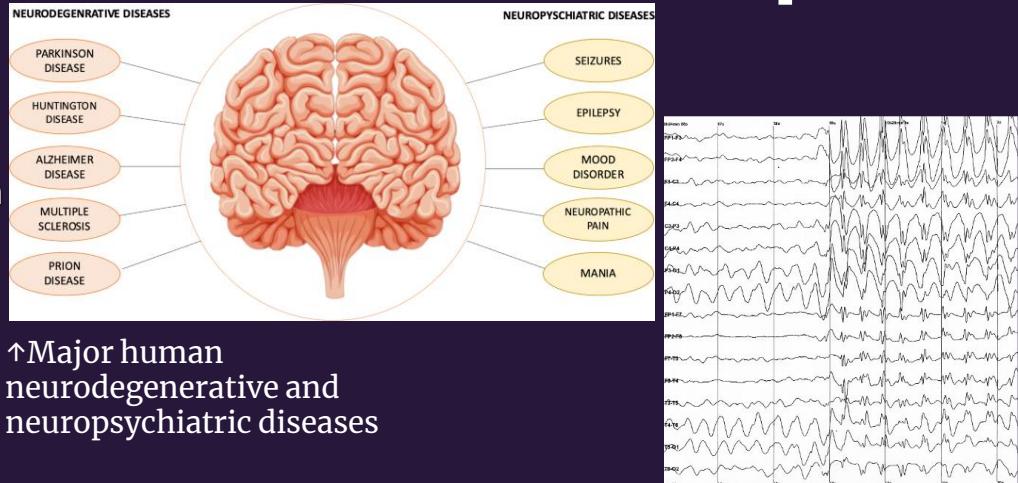
# Problem & Dataset Description

Why epilepsy?

50 Million people globally  
Rare abnormal patterns

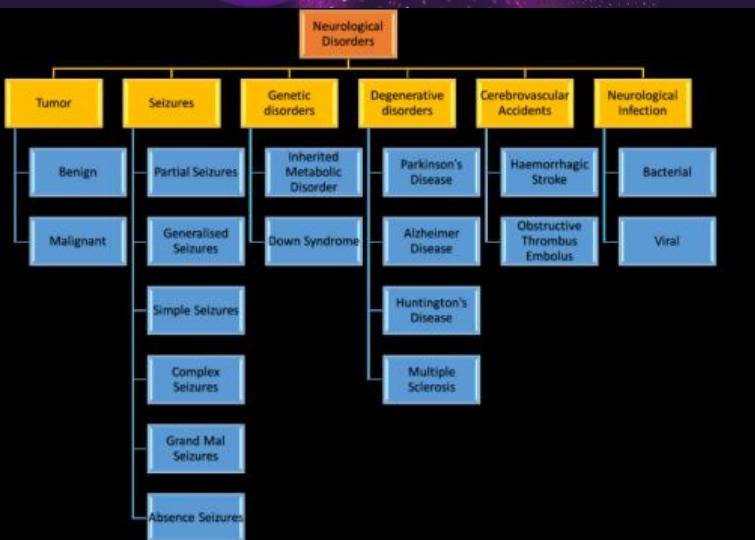
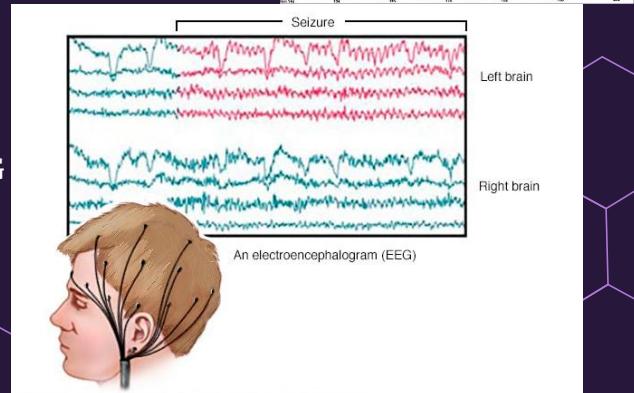
Neuromorphic anomaly- detection  
My approach: automatic real-time  
recognition

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<-Classification of neurological disorders

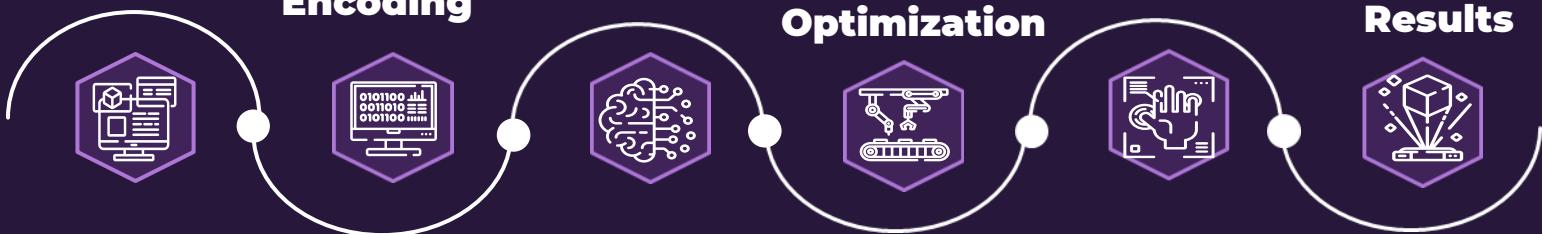
Epileptic spike and wave discharges monitored EEG  
->



# Methodology

Conversion of EEG into spike-based or feature vectors suitable for neuromorphic processing

of model structure and parameters



## Dataset

EEG recordings containing annotated seizure and non-seizure periods

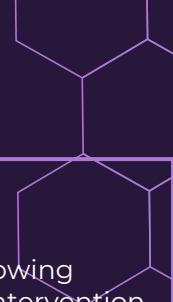
for seizure anomaly detection

## Evaluate

using standard metrics: sensitivity, specificity, latency

## Results

# Experimental Setup

 <b>Data</b> <b>et</b>	<ul style="list-style-type: none"><li>• collected at the Children's Hospital Boston </li><li>• EEG recordings from pediatric subjects <b>with intractable seizures</b>. Subjects were monitored for up to several days following withdrawal  of <b>anti-seizure medication</b> in order to characterize their seizures and assess their candidacy for surgical intervention.</li></ul>
<b>Softw are/ Tools</b>	<p>Google Colab, ipynb libraries such as:</p> <ul style="list-style-type: none"><li>• <b>Pyedflib</b> – reading EDF files; <b>Scipy.stats</b> – statistics; <b>Pywt</b> – Wavelet transformations and more</li></ul>
 <b>Model Detail s</b>	<p>Finds epileptic seizures in EEG brain signals - 3 Main steps::</p> <ol style="list-style-type: none"><li>1. Data Preparation: Get full EGG record -&gt; cut on 5-sec windows -&gt; mark seizure=1, normal=0</li><li>2. Features extraction for each 5 sec piece: energy – how strong is the signal; line length – how complicated is form is; entropy – how chaotic/ irregular it is</li><li>3. Genetic algorithm for optimization: Start [random weights] -&gt; bad results </li></ol> <p>-&gt; <b>EVOLUTION</b> (30 gen.): - Selection: Take the best ones; - Crossover: Combine them; - Mutation: Make little changes</p> <p>-&gt; END:[optimal weights] -&gt; best results</p>

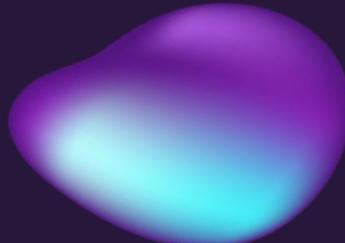
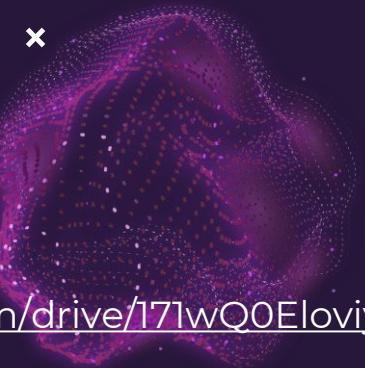
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# EA Setup

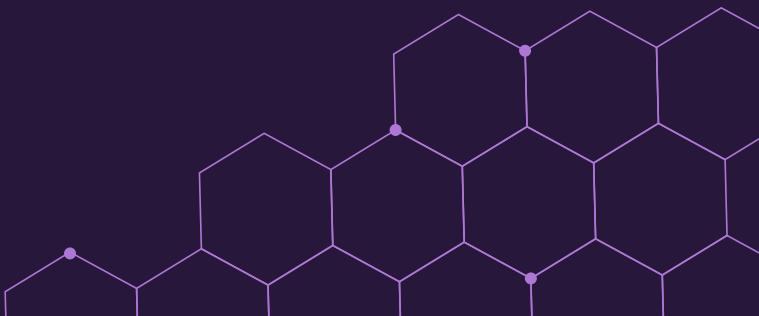
<https://colab.research.google.com/drive/17lwQ0EloviycEShdyJU3i0Neeh139xKb?usp=sharing>

Each individual has 4 numbers:

w<sub>1</sub>, w<sub>2</sub>, w<sub>3</sub> – weights for the characteristics, threshold – classification threshold



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# Neuromorphic Model & Encoding



## Neuromorphic Computing

- Mimics the workings of the human brain
- Uses spiking neural networks (SNN)
- Spikes instead of continuous signals
- Energy efficient

## Encoding for EEG

- Converting analog EEG signals into spikes

# Results & Evaluation

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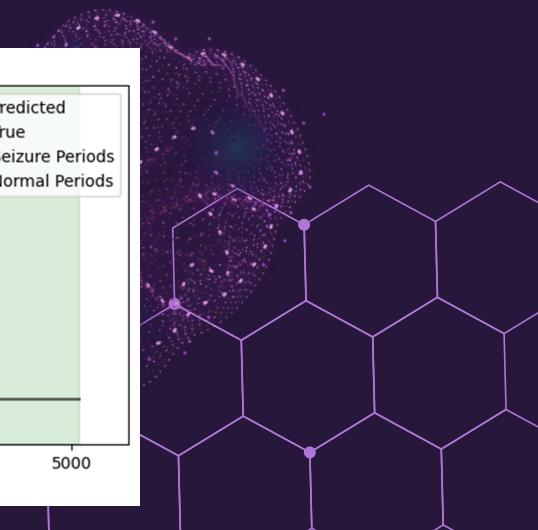
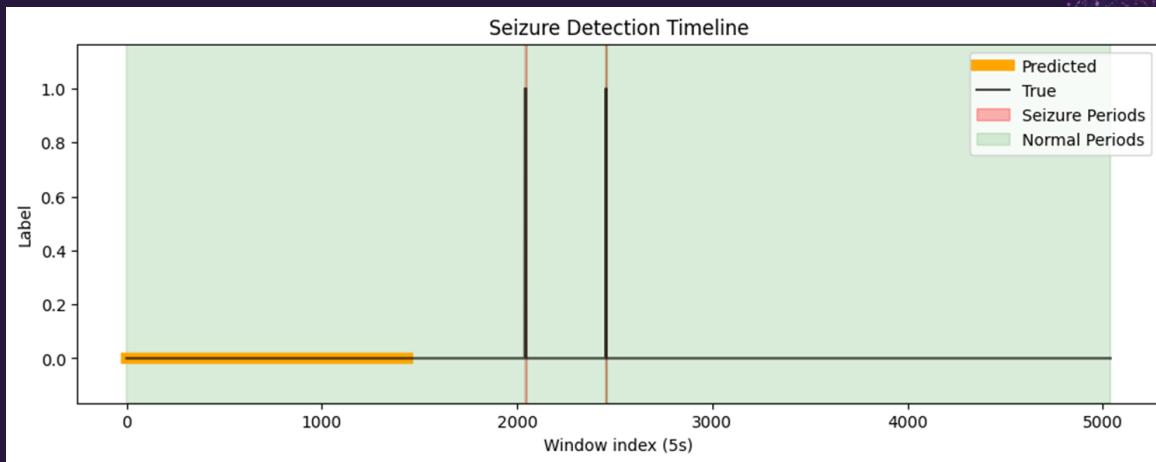
**Timeline of Seizure Detection:** Over 5,040 Windows (7 hours of EEG data)

**Key Results:** Orange Line: EA-optimized predictions; Black Line: Actual seizure events; Red Zones: True seizure periods (11 windows); Green Zones: Normal brain activity (5,029 windows)

**Performance Metrics:** Successfully detected rare events (0.2% of data)

-> Meaning:

- EA found optimal feature weights automatically
- System balances catching seizures vs. avoiding false alarms
- Challenge: Highly imbalanced dataset (1:457 ratio)



# Conclusion

- ✓ Evolutionary algorithms can optimize neuromorphic anomaly detection
- ✓ Time-domain features are sufficient for basic seizure detection
- ✓ Automated weight optimization eliminates manual tuning

## Future Applications:



### Medical

- Real-time seizure prediction devices
- Personalised epilepsy monitoring systems
- Automatic medication adjustment

### Other:

- Anomaly detection in financial systems
- Cybersecurity intrusion detection

# SOURCES

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- <https://www.ibm.com/think/topics/neuromorphic-computing>
- [https://www.researchgate.net/figure/Concept-of-SNN-and-its-structure\\_fig1\\_347863545](https://www.researchgate.net/figure/Concept-of-SNN-and-its-structure_fig1_347863545)
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- <https://en.wikipedia.org/wiki/Electroencephalography>
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# Thank you for your attention!

