

# Project: Organoid-Inspired Bio-Computer State Classification

Course: INFO6105 - Data Science Methods

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## 1. Problem Statement

This project simulates an "Organoid Bio-Computer" monitoring system. Using public EEG data as a **biological proxy**, we trained a machine learning pipeline to classify the organoid's computational state into (1) **Active** (Computing) or (2) **Inactive** (Resting).

## 2. Data Pipeline (Operation Extract)

**Source & Sampling:** Extracted from Kaggle (mental-state.csv). Curated a random sample of **150 rows** to meet the " $100 < \text{rows} < 200$ " bonus requirement.

**Feature Engineering (Bonus Met):** Selected 12 biological features and engineered 2 new composite metrics: **Total\_Input\_Power** (Sum of inputs) and **Synapse\_Ratio** (Alpha/Beta balance).

**Final Structure:** 150 Rows  $\times$  15 Columns (Meeting " $10 < \text{cols} < 20$ " bonus).

**Target:** Binary classification (0=Inactive, 1=Active).

## 3. Modeling Strategy (Operation Learn)

Data was split (80% Train / 20% Test). We implemented three classifiers for comparison:

(1) **Decision Tree:** Baseline model for interpretability.

(2) **Random Forest:** Selected as the **Final Model** for its stability against noise.

(3) **Gradient Boosting (Extra Credit):** Implemented this **unlearned model** to explore boosting techniques vs. bagging.

## 4. Results

(1) **Performance:** Random Forest achieved the best stability (~67% Accuracy), consistent with high-noise biological proxies. Gradient Boosting showed competitive results but slight overfitting.

(2) **Analysis:** Confusion Matrix confirms high precision for "Active" states. Heatmaps reveal strong correlations between neural inputs and synaptic activity.

(3) **Deployment:** The best model was saved (my\_best\_model.pkl) and deployed for a real-time prediction demo.

## 5. Limitations

(1) **Proxy Data:** EEG is a simulation proxy, not direct wet-lab MEA data.

(2) **Sample Size:**  $N=150$  is sufficient for a demo but limits generalization.

(3) **Dynamics:** Lacks time-series modeling (e.g., LSTM).

## 6. Next Steps

(1) Integrate real organoid MEA data.

(2) Implement Deep Learning (1D-CNN) for temporal dynamics.

(3) Scale dataset size and build a Streamlit dashboard.