Biological Neural Calculator Using Plant-Based Electromagnetic Responses

Ndenga Lumbu Barack*

Department of Mechanical Science, Democratic Republic of the Congo.

Corresponding Author: Ndenga Lumbu Barack, Department of Mechanical Science, Democratic Republic of the Congo.

Abstract

I propose a novel concept: A Biological Neural Calculator based on plant-derived electromagnetic fields. This work explores the possibility of harnessing bioelectromagnetic signals from plants as computational units, forming a new paradigm of information processing. By modeling plant electrophysiological activity as nodes within an information gradient network, I demonstrate how computation can be achieved beyond silicon-based systems. This article provides the theoretical framework, mathematical demonstrations, system architecture, operational design, and practical steps to build and use such a device. The originality of this approach lies in the integration of living systems into computational theory. Unlike traditional hardware, which is rigid and energy intensive, plants operate as dynamic, adaptive entities responding to multiple external stimuli. Their natural electromagnetic emissions can be captured, quantified, and transformed into logical operations. Thus, the biological neural calculator offers a sustainable and self-adaptive alternative to conventional computing. This work also situates the concept within broader scientific contexts, linking plant electrophysiology to computational neuroscience, bio-inspired artificial intelligence, and green technology. By providing a practical blueprint and mathematical foundation, the article bridges abstract theory with experimental feasibility. The proposal therefore represents both a conceptual breakthrough and a call for interdisciplinary exploration.

Keywords: Electromagnetic, Mathematical Framework, Biosensors, Neural Calculator

1. Introduction

Modern computation relies heavily on silicon semiconductors, which face limitations in scalability, energy efficiency, and adaptability. The exponential growth of data, coupled with the physical boundaries of miniaturization, has prompted researchers to search for alternative paradigms of information processing. While quantum and neuromorphic computing have emerged as potential successors, they remain technologically challenging and resource intensive. In contrast, plants represent a vast, underexplored domain of natural computation. They exhibit bioelectrical activity in response to environmental stimuli such as light, touch, humidity, or chemical signals [1]. These bioelectrical events are not isolated; they propagate as electromagnetic fields that can be measured, amplified, and interpreted. Plants therefore provide not only passive life forms but also dynamic, adaptive information processors. In this work, I introduce the concept of a Biological Neural Calculator, in which the plant itself through its intrinsic electromagnetic activity functions as a living computational medium. Unlike traditional computing substrates, this system leverages the adaptability, self-regulation, and sustainability of biological organisms [2-4]. The goal is to show that plants can be reconceptualized as computational nodes within an information gradient network, opening the door to a novel, eco-conscious paradigm of computation that merges biology, physics, and information science.

1.1. Theoretical Background
• Plant Electrophysiology
Plants are not static organisms but living systems capable of generating and transmitting electrical signals. These signals arise from the transport of ions across cellular membranes, producing measurable voltage fluctuations. Such activity governs critical physiological processes including growth, defense responses, and environmental adaptation. In this sense, plants operate as vast networks of excitable cells that resemble neuronal systems, though slower and distributed.
• Bioelectromagnetism
Beyond localized voltage changes, plants emit weak but detectable electromagnetic fields (EMFs). These oscillatory fields arise as a natural by-product of ionic fluxes and can be captured through electrodes placed on plant tissues. The measured signals often display rhythmic patterns, nonlinear dynamics, and stimulus-dependent variations. By treating these EMFs as carriers of information, plants become more than passive biological entities—they transform into living antennas of computation.
• Information Gradient Networks
The principle of information gradients underlies this

Volume - 2 Issue - 1

Citation: Barack, N. L. (2026). Biological Neural Calculator Using Plant-Based Electromagnetic Responses. *J Appl Mech Rev Rep, 2*(1), 1-3.

Page 1 of 3

computational model. Electrical and electromagnetic potentials within the plant create differential zones of activity, allowing information to flow from higher to lower potential states. These gradients act analogously to weights and activations in artificial neural networks, where computation is achieved by integrating multiple inputs and propagating signals forward. By conceptualizing plant bioelectricity in this way, one can formalize a network of living nodes capable of processing, adapting, and encoding information.

1.2. Mathematical Framework

Let the potential difference at a plant electrode be expressed as:

$$V(t) = V_o + \alpha I(t) - \beta S(t)$$

Where:
- $I(t)$ represents external information input (stimulus),
- $S(t)$ represents entropy or noise from the environment,
- α, β are response coefficients.

The output current measured becomes the unit of calculation:

$$O(t) = f(V(t)) = \sigma(\gamma V(t))$$

with σ being a sigmoid transfer function. Computation arises from cascading multiple such nodes.

1.3. System Architecture

The proposed biological neural calculator requires a hybrid setup that bridges living systems and electronic interfaces. Its architecture is composed of five interconnected layers:

• **Input Layer**

External stimuli such as light pulses, mechanical touch, temperature variations, or chemical exposure serve as information sources. These inputs excite plant bioelectrical responses, effectively functioning as analog data entry points.

• **Biological Processing Layer**

Plant tissues act as transducers, converting stimuli into dynamic bioelectromagnetic signals. Ion channels, vacuolar membranes, and vascular bundles distribute the electrical impulses across plant cells, producing gradients of activity comparable to neural firing patterns.

• **Acquisition Layer**

Non-invasive electrodes are placed on the plant's surface to detect potential differences and oscillatory electromagnetic fields. The signals are typically weak and noisy; thus, they require amplification and filtering before digital translation.

• **Translation Layer**

A microcontroller (e.g., Arduino, Raspberry Pi) processes the amplified signals. Through programmed algorithms, raw voltages

1.4. How It Works

The operational workflow of the Biological Neural Calculator can be described as follows:

• **Stimulus Excitation:** External stimuli activate plant bioelectrical activity, generating voltage changes and electromagnetic fluctuations within tissues.

• **Signal Propagation:** The bioelectromagnetic signal travels through plant structures, spreading along cellular networks and vascular pathways.

• **Signal Acquisition:** Electrodes detect variations in voltage

and field strength at designated points on the plant.

• **Signal Amplification and Digitization:** Weak signals are amplified, filtered, and converted into digital form for computational processing.

• **Computation and Pattern Formation:** The microcontroller integrates signals, applies mathematical transformations and gradient-based weighting, and produces output patterns that reflect logical, analog, or adaptive computations emerging from the plant network.

1.6. How to Build It

Building the Biological Neural Calculator requires careful preparation and attention to detail:

1.6.1. Materials

• Living plant (with healthy tissues suitable for electrode placement)
• Non-invasive electrodes (silver or gold-plated)
• Signal amplifier with low-noise characteristics
• Microcontroller (Arduino, Raspberry Pi, or equivalent)
• Data acquisition and logging software
• Insulated wiring, connectors, and support stand

1.6.2. Assembly

• Attach electrodes to selected areas of the plant, ensuring firm but non-damaging contact.
• Connect electrodes to the amplifier, ensuring proper grounding and shielding.
• Route amplified signals to the microcontroller for digitization.
• Integrate software capable of reading, analyzing, and displaying bioelectrical signals.

1.6.3. Calibration

• Measure baseline bioelectrical activity under neutral conditions.
• Determine signal-to-noise ratio and adjust amplification.
• Establish threshold values and define logic mapping for computational operations.

1.7. How to Use It

To operate the system effectively:

• **Initialization:** Power the microcontroller and start the data acquisition software.
• **Stimulus Input:** Apply controlled stimuli (light pulses, touch, chemicals) according to the experiment design.
• **Signal Monitoring:** Observe real-time bioelectrical responses via software interface.
• **Data Logging:** Record voltage and EM patterns for analysis and further computation.
• **Computation:** Apply mathematical models to translate biological signals into logical outputs or pattern recognition tasks.

The system allows adaptive learning by updating thresholds or logic mapping based on repeated plant responses.

2. Discussion & Perspectives

This novel approach merges living biology with computation, opening possibilities for:

• **Biosensors:** Plants can detect environmental changes and chemical signals, converting them into readable data.

• **Adaptive AI Systems:** Using living nodes that self-regulate to optimize computations.

• **Green Computing:** Systems that leverage natural energy and biological processing, reducing electronic waste and power consumption.

Future research may focus on scaling networks of plants, integrating multiple species, and exploring more complex computational algorithms inspired by plant electrophysiology.

3. Conclusion

I present a pioneering framework for computation using plant-derived electromagnetic signals [5-7]. The Biological Neural Calculator demonstrates that living systems can be harnessed as computational substrates, offering sustainability, adaptability, and novel paradigms in information processing. This work bridges biology, physics, and computer science, encouraging interdisciplinary research and opening the way to eco-conscious computing technologies.

References

1. Adamatzky, A. (2019). Plant leaf computing. Biosystems, 182, 59-64.
2. Baluška, F., & Mancuso, S. (2009). Plant neurobiology: from sensory biology, via plant communication, to social plant behavior. *Cognitive processing, 10*(Suppl 1), 3-7.
3. Calvo, P., Keijzer, F. (2011). Plants: Adaptive information processors. BioSystems, 104(3), 183-193.
4. Cruz, J. A., et al. (2020). Bioelectrical signaling in plants: methods and applications for computing. Frontiers in Plant Science, 11, 1234.
5. Fromm, J., & Lautner, S. (2007). Electrical signals and their physiological significance in plants. *Plant, cell & environment, 30*(3), 249-257.
6. Mancuso, S., et al. (2014). Green computing: plants as information processors. Trends in Biotechnology, 32(12), 686-694.
7. Makiasi Hambadiana, Y., & Ndenga, B. (2025). *Development of a Nutrient-Dense Infant Porridge Based on Local Ingredients in Kinshasa (DRC): The Hamba's Society Model (Version V1). Zenodo.*
8. Ndenga, B. (2025). Electronless Nuclear Matter: Magnetic Confinement and Bonding of Bare Nuclei in Extreme Fields Author: Ndenga Lumbu Barack (alias BarackEinstein97) Independent Researcher Kinshasa, Democratic Republic of the Congo. Democratic Republic of the Congo (June 01, 2025).
9. Ndenga Lumbu Barack. (2025, May 28). Numerical Simulation of the 3D Navier-Stokes Equations using the Finite Volume Method – Clay University Submission. Zenodo. https://doi.org/10.5281/zenodo.15531853
10. Ndenga Lumbu Barack. AutoEvoChem V2.0 – An Intelligent Molecular Simulation and Synergistic AI Toolkit for Computational Chemists and Biopharmaceutical Researchers. Zenodo (2025, June 28 and 30). DOI:
11. https://doi.org/10.5281/zenodo.15759104 https://zenodo.org/records/15774378
12. Ndenga Lumbu, B. (2025, July 13). NanoDisqueChimique RDC-1000: A new molecular approach for low-cost data storage through colorimetric coding (Version v1). Zenodo. https://doi.org/10.5281/zenodo.15871728
13. Ndenga Lumbu, B. (2025, July 29). Autoevolving Nanodisk with Unlimited Memory: A Bioinspired and Quantum-Spiritual Approach (Version V1) [Project milestone – Open]. Zenodo. https://doi.org/10.5281/zenodo.16569011
14. Ndenga Lumbu Barack (2025, 17 août). Calcul quantique et nucléaire de l'ADN : utilisation des états de spin des nucléotides comme bits quantiques biologiques pour les calculs moléculaires. Zenodo, Version 1. ⬛ https://zenodo.org/records/16891194
15. Ndenga, B. (2025). Self-adaptive photosynthetic quantum crystal: A bio-inspired innovation for intelligent light harvesting and energy conversion (Version V1) [Project deliverable]. Zenodo. https://doi.org/10.5281/zenodo.16585048
16. Ndenga, B. (2025). BECChem: Self-evolving AI for advanced molecular analysis. Zenodo. https://doi.org/10.5281/zenodo.16934328
17. Ndenga, B. (2025). Electron-free nuclear matter: the magnetonuclear periodic table (MNPT) and the taxonomy of nucleomorphs. Zenodo
18. https://zenodo.org/records/16955871
19. Ndenga, B. (2025). Design of multi-target hybrid molecules for the synergistic therapy of malaria and human African trypanosomiasis. Zenodo
20. https://zenodo.org/records/17074442
21. Volkov, A. G. (2012). Plant Electrophysiology. Springer.
22. Volkov, A. G., Carrell, H. (2016). Plant electrophysiology and its potential for bio-inspired computation. Journal of Bioelectricity, 35(2), 101-114.