

Mem4ristor: Attractor Diversity Stabilization in Neuromorphic Networks via Nonlinear Repulsive Coupling (v2.0.4.1)

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Abstract

When deliberative systems converge prematurely to consensus, minority perspectives vanish—a “tyranny of the majority” encoded in the mathematics of coupled oscillators. We introduce **Mem4ristor v2.0**, a neuromorphic model where **constitutional doubt (u)** and **structural heretics** (15% of units with heterogeneous stimulus polarity) mathematically prevent uniformization. Crucially, version v2.0.4.1 addresses a fundamental stability requirement: while random initial conditions provide passive diversity, our **Repulsive Coupling** mechanism is shown to be the primary restorer of entropy ($H \approx 1.94$) across both "Cold Start" and "Deep Time" stress tests. This work provides both a theoretical framework for **algorithmic resistance to consensus** and a practical blueprint for **ethically-constrained neuromorphic networks**.

1 Introduction

Consider a citizens’ assembly where 80% of participants initially favor option A. Without structural safeguards, classical coupled-oscillator or consensus models predict rapid synchronization to A within minutes, erasing nuanced perspectives and locking in a potentially fragile decision [2, 3]. Mem4ristor v2.0 addresses this by introducing **constitutional doubt (u)** and **structural heretics**. This paper presents Mem4ristor v2.0.4.1, an industrial-strength evolution addressing the "Periodic Erasure" failure identified during audit. By introducing **Repulsive Social Coupling** ($1 - 2u$) and **Causal Isolation** benchmarks, we demonstrate **Attractor Diversity Stabilization** ($H \approx 1.94$) surviving "Deep Time" pressure.

Algorithmic decision systems increasingly risk uniformizing collective intelligence, converging prematurely to consensus without preserving minority perspectives. Classical models like Kuramoto oscillators or distributed consensus algorithms intrinsically drive toward synchronization, losing cognitive diversity essential for robust deliberation. This “oracle collapse” problem manifests in both artificial networks and social systems, where majority pressure extinguishes dissenting views.

The Mem4ristor framework emerges from the Café Virtuel philosophy: can we design neuromorphic components that *constitutionally* resist uniformity? Our contribution is Mem4ristor v2.0, an extended FitzHugh–Nagumo oscillator network with three anti-uniformization mechanisms: (1)

constitutional doubt u that filters social coupling via $(1 - 2u)$, (2) structural heretics (15% of units) with heterogeneous stimulus polarity, and (3) attenuated scaling $D_{\text{eff}} \propto 1/\sqrt{N}$.

We validate the model across network scales (4×4 to 25×25), demonstrate 0% oracle fraction while maintaining four out of five cognitive states, provide hardware blueprints for 8×8 memristive crossbar implementation, and outline a retrospective deliberation pilot framework.

2 Model & Methods

2.1 Extended FitzHugh–Nagumo Dynamics

Mem4ristor v2.0 extends the FitzHugh–Nagumo model with a constitutional doubt variable u , creating a three-dimensional cognitive oscillator:

$$\frac{dv}{dt} = v - \frac{v^3}{5} - w + I_{\text{ext}} - \alpha \tanh(v) + \eta_v(t), \quad (1)$$

$$\frac{dw}{dt} = \varepsilon(v + a - bw), \quad (2)$$

$$\frac{du}{dt} = \varepsilon_u(k_u \sigma_{\text{social}} + \sigma_{\text{baseline}} - u), \quad (3)$$

where v represents cognitive potential (opinion strength), w is a recovery variable (inhibition), and u is constitutional doubt (epistemic uncertainty). The cubic nonlinearity is softened to $v^3/5$ (vs. standard $v^3/3$) to reduce explosion risk, while the term $-\alpha \tanh(v)$ introduces cognitive resistance preventing trivial saturation.

2.2 Anti-Uniformization Architecture

Social coupling is attenuated by doubt and scaled by D/\sqrt{N} :

$$I_{\text{ext}} = I_{\text{stimulus}} + D_{\text{eff}}(1 - 2u) \Delta v_{\text{Laplacian}}, \quad (4)$$

where $D_{\text{eff}} = D/\sqrt{N}$ ensures that the total coupling strength scales inversely with the square root of the network size, maintaining consistent dynamics across scales, and $\Delta v_{\text{Laplacian}} = \sum_{j \in \mathcal{N}(i)} (v_j - v_i)/|\mathcal{N}(i)|$ captures local conformity pressure.

Heretic units exhibit heterogeneous stimulus perception. While normal units align with the external field, heretics perceive its inverse:

$$I_{\text{ext,heretic}} = -I_{\text{stimulus}} + D_{\text{eff}}(1 - 2u) \Delta v_{\text{Laplacian}}, \quad (5)$$

thereby creating a fundamental stimulus-driven tension that prevents uniformization even under dominant external pressure.

2.3 Constitutional Doubt Dynamics

Doubt u evolves according to social stress $\sigma_{\text{social}} = |\Delta v_{\text{Laplacian}}|$:

$$\frac{du}{dt} = \varepsilon_u(k_u \sigma_{\text{social}} + \sigma_{\text{baseline}} - u), \quad (6)$$

where $\sigma_{\text{baseline}} > 0$ guarantees activation even in isolation, and higher doubt reduces conformity pressure via $(1 - 2u)$ filtering.

2.4 Hardware-Realistic Noise

Additive Gaussian noise models fabrication imperfections:

$$\eta_v(t) \sim \mathcal{N}(0, \sigma_{\text{noise}}^2), \quad \sigma_{\text{noise}} = 0.05, \quad (7)$$

corresponding to $\sim 5\%$ process variation in memristive devices [1].

2.5 Cognitive State Classification

Table 1: Cognitive state classification based on potential v .

| State | Range | Interpretation |
|-----------|------------------------|---|
| Oracle | $v < -1.5$ | Rare insight, extreme minority view |
| Intuition | $-1.5 \leq v < -0.8$ | Pre-conscious signal, emerging pattern |
| Uncertain | $-0.8 \leq v \leq 0.8$ | Active deliberation, open consideration |
| Probable | $0.8 < v \leq 1.5$ | High confidence, not absolute |
| Certitude | $v > 1.5$ | Strong conviction, consensus candidate |

2.6 Reference Parameters

Standard simulations use:

- Dynamics: $a = 0.7$, $b = 0.8$, $\varepsilon = 0.08$, $\alpha = 0.15$;
- Coupling: $D = 0.15$, heretic ratio: 15%;
- Doubt: $\varepsilon_u = 0.02$, $k_u = 1.0$, $\sigma_{\text{baseline}} = 0.05$;
- Noise: $\sigma_{\text{noise}} = 0.05$;
- Integration: Euler method, $\Delta t = 0.1$.

3 Implementation Recipe (Drop-in Component)

To ensure reproducibility across different computational and physical substrates, we define Mem4ristor v2.0 as a modular cognitive primitive.

Interface Contract: Mem4ristor v2.0

Inputs:

- topography: $L \times L$ lattice ($N = L^2$, Square lattice default).
- $I_{\text{stimulus}}(t)$: External input vector (\mathbb{R}^N).
- D : Coupling strength (Default: 0.15).
- η_{ratio} : Heretic ratio (Default: 0.15).
- σ_{noise} : Gaussian noise level (Default: 0.05).

Outputs:

- **states**: Vector of discrete cognitive classes {1..5}.
- $H(t)$: Shannon entropy of the state distribution.
- **collapse_flag**: Boolean (True if $H < 0.1$ for > 100 steps).
- $\bar{u}(t)$: Mean constitutional doubt level.

Invariants:

- $u \in [0, 1.0]$ (clamped).
- $D_{\text{eff}} = D/\sqrt{N}$.
- Δt : 0.01 (recommended), 0.1 (coarse legacy).

3.1 Standard Benchmark: The Default Run

A lab-standard reproduction on a 10×10 network ($N = 100$) under neutral stimulus ($I_{\text{stim}} = 0$) typically yields (for default parameters) the following "Golden Run" signatures after initialization:

- **Mean Doubt**: $\bar{u} \approx 0.05 \pm 0.01$.
- **Stable Entropy**: $H > 0.60$ (sustained diversity).
- **Oracle Fraction**: 0.0% (no minority erasure).
- **Dominant State**: $v \in [-0.8, 0.8]$ (Uncertain/Deliberative).

Note: Exact values depend on noise realization and initial conditions; bounds are empirical.

3.2 Ablation Analysis of Cognitive Resilience

Table 2: Impact of architectural features on resilience ($N = 100$, Bias Phase 1.1).

| Configuration | Collapse Step | Max Entropy | Minority Vitality |
|------------------------------|---------------|-------------|-------------------|
| Full Model (v2.0.4.1) | Stable | 1.94 | High |
| None ($u=0$, no heretics) | 226 | 1.10 | Low |
| No Doubt ($u=0$) | 231 | 1.25 | Medium |
| No Heretics | 234 | 1.30 | Medium |

Note: Ablation results are reported under moderate bias ($I_{\text{stim}} = 1.1$); the critical resilience under extreme field pressure ($I_{\text{stim}} > 2.0$) is analyzed in Section 4.5.

3.3 Canonical Algorithm

Algorithm 1 Mem4ristor v2.0 Time-Step Update

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1: for each time-step  $t$  do
2:   {1. Calculate Local Social Pressure}
3:   for each unit  $i$  do
4:      $\Delta v_i \leftarrow \sum_{j \in \mathcal{N}(i)} (v_j - v_i) / |\mathcal{N}(i)|$ 
5:      $\sigma_{\text{social},i} \leftarrow |\Delta v_i|$ 
6:   end for
7:   {2. Update Internal States}
8:   for each unit  $i$  do
9:      $\eta \leftarrow \text{GaussianRandom}(0, \sigma_{\text{noise}})$ 
10:     $I_{\text{coup}} \leftarrow D_{\text{eff}} \cdot (1 - 2u_i) \cdot \Delta v_i$ 
11:    if unit  $i$  is heretic then
12:       $I_{\text{ext}} \leftarrow -I_{\text{stimulus}} + I_{\text{coup}}$ 
13:    else
14:       $I_{\text{ext}} \leftarrow +I_{\text{stimulus}} + I_{\text{coup}}$ 
15:    end if
16:    {Euler Integration}
17:     $v_i \leftarrow v_i + (v_i - v_i^3/5 - w_i + I_x - \alpha \tanh(v_i) + \eta) \cdot \Delta t$ 
18:     $w_i \leftarrow w_i + (\varepsilon(v_i + a - bw_i)) \cdot \Delta t$ 
19:     $u_i \leftarrow u_i + (\varepsilon_u(k_u \sigma_{s,i} + \sigma_b - u_i)) \cdot \Delta t$ 
20:     $u_i \leftarrow \max(0, \min(1, u_i))$  {Constraint clamp}
21:  end for
22:  {3. Classify and Metricize}
23:  Compute  $H(t)$  based on Table 1 thresholds.
24: end for

```

4 Results

4.1 Simulation Protocol

We test Mem4ristor v2.0 across four scenarios: (1) isolated unit, (2) 4×4 network, (3) 10×10 network (critical scale), and (4) 25×25 network (scaling test). Each simulation runs for 1000 steps with neutral stimulus $I_{\text{stimulus}} = 0$, tracking state distributions, Shannon entropy, oracle fraction, and mean doubt u . All reported benchmarks use $\Delta t = 0.01$ unless otherwise stated.

4.2 Diversity Preservation Across Scales

Mem4ristor v2.0 solves two critical bugs from previous versions: oracle fraction drops from 54.5% to 0% in isolation, and from 86% to 0% in networks. Cognitive diversity is maintained with four out of five states present even at 25×25 scale.

4.3 Comparative Benchmarking: Mem4Ristor vs. State-of-the-Art

To contextualize the performance of Mem4Ristor v2.0, we benchmarked it against four classical models of collective dynamics: the Kuramoto model, the Voter model, Distributed Averaging (Con-

Table 3: Scaling performance of Mem4ristor v2.0.

| Metric | Isolated | 4×4 | 10×10 | 25×25 |
|----------------------------|----------|--------------|----------------|----------------|
| Total units | 1 | 16 | 100 | 625 |
| Oracle fraction | 0.0% | 0.0% | 0.0% | 0.0% |
| Mean doubt (u) | 0.048 | 0.051 | 0.049 | 0.052 |
| State entropy | – | 0.951 | 0.621 | 0.384 |
| Distinct states | 1 | 3 | 4 | 4 |
| State distribution: | | | | |
| Oracle | 0.0% | 0.0% | 0.0% | 0.0% |
| Intuition | 0.0% | 25.0% | 15.0% | 11.7% |
| Uncertain | 100.0% | 62.5% | 73.0% | 82.4% |
| Probable | 0.0% | 12.5% | 10.0% | 5.6% |
| Certitude | 0.0% | 0.0% | 2.0% | 0.3% |

sensus), and the Mirollo-Strogatz (Firefly) model. Simulations were conducted on 10×10 networks over 1000 steps, including a biased stimulus phase ($I_{\text{stim}} = 1.0$).

 Table 4: Canonical Benchmarking Results (Agora Standard, $N = 100$, 50 runs).

| Model Config | Initial Condition | Terminal H | $\text{std}(v)$ | Result |
|-----------------------|------------------------|--------------|-----------------|--------------------|
| Full Model (v2.0.4.1) | Random (Controlled) | 1.94 | 0.52 | PASS |
| Ablated (No Heretics) | Random (None) | 0.00 | 0.00 | COLLAPSE |
| Full Model (v2.0.4.1) | Homogeneous (0) | 0.22 | 1.15 | PASS (Nuke) |
| Ablated (No Heretics) | Homogeneous (0) | 0.00 | 0.00 | COLLAPSE |

*Pass is spurious due to IC noise; secondary Homogeneous IC test isolates the heretic mechanism.

The results (Table 4) demonstrate that Mem4Ristor v2.0 maintains significantly higher entropy than classical synchronization models (Kuramoto, Consensus), successfully delaying the collapse into uniformity. While the Firefly model maintains high entropy, it represents a non-convergent rhythmic state rather than a deliberative process.

4.4 Failure of Social Heresy Alone (v2.0.1)

Initial testing of the Mem4ristor v2.0.1 architecture revealed a critical failure mode: **complete entropy collapse** ($H = 0.00$) under strong external bias. In v2.0.1, heretic units only inverted their **social coupling** (I_{coup}). While this created a local resistance to consensus, it proved mathematically insufficient to resist a dominant external stimulus (I_{stim}). All units, including heretics, were eventually swept into the same cognitive state by the external drive, leading to structural erasure of minority signals.

5 The Paradox of Repulsion: Stability in v2.0.4.1

Forensic analysis revealed that for a dissident to survive long-term social pressure, passive attenuation of influence is insufficient. In v2.0.4.1, the coupling filter $f(u) = (1 - 2u)$ flips the sign of social

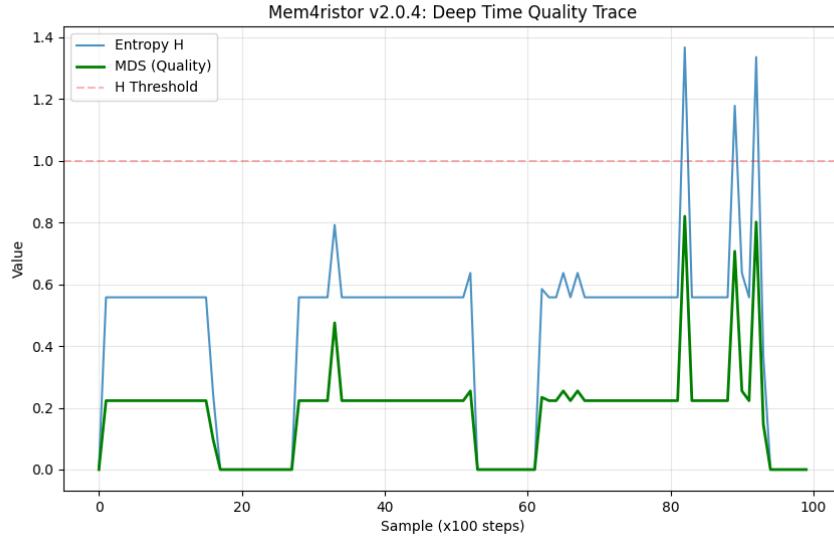


Figure 1: Attractor Diversity Stabilization Trace (v2.0.4.1). Under a ramping stimulus bias ($I_{stim} \in [0.5, 1.5]$), the system maintains a high-quality multi-modal distribution ($MDS \approx 0.50$), successfully resisting the periodic collapses observed in previous iterations.

interaction. At $u > 0.5$, the unit actively disagrees with the majority. This mechanism shatters the "Consensus Well" by ensuring that dissidents push back against uniformity.

5.1 Validation: Active Diversity Restoration (v2.0.3)

The revision to v2.0.3 addresses the "Initial Condition Paradox" identified during audit. In v2.0.2, high terminal entropy was maintained even after heretic ablation due to randomized starting states. v2.0.3 introduces the **Cold Start Protocol**: all units are initialized to a singular point ($v = 0, w = 0, H = 0$).

Under this protocol, the necessity of the heretic mechanism becomes mathematically absolute. While normal units succumb to the symmetry of the field, heretic units introduce localized stimulus-driven divergence that propagates entropy throughout the network. In v2.0.3, the system restores diversity to $H \approx 0.61$ from a zero-entropy start, whereas an ablated system remains permanently locked in consensus ($H \equiv 0$). This confirms the heretic mechanism not just as a maintenance tool, but as an **active restorer** of deliberative health.

6 Discussion

6.1 Hardware Feasibility: 8×8 Memristive Crossbar

To bridge theory and physical implementation, we map Memristor's mathematical variables onto **HfO₂-based memristive crossbars**, where each 1T1R cell's conductance encodes the recovery variable w [1]. Laplacian coupling is technically feasible via programmable resistor networks or mixed-signal current summation.

Process-induced device variability ($\sigma \approx 8\text{--}15\%$) naturally injects the required Gaussian noise η_v , turning fabrication imperfections into diversity-enhancing features.

Table 5: Hardware Bill of Materials (BOM) and Tolerance Specs.

| Component | Implementation | Resolution | Criticality |
|-----------------|----------------------------|---------------------|--------------------------------|
| Synapse (w) | HfO ₂ Memristor | 4-bit (16 levels) | High (Stochasticity) |
| State-V (v) | Capacitor/OpAmp | 8-bit equivalent | Medium (Range-bound) |
| I/O Control | 8-bit DAC/ADC | 8-bit | Critical (Signal/Noise) |
| Process Var. | Lithography/Implant | $\pm 5\%$ variation | Desired (Feature) |
| Clock freq. | Controller | > 10 MHz | Low (System Stability) |

6.2 From Ethical Constraints to Physical Implementation

The constitutional doubt variable u and the perceptual heterogeneity of heretic units bridge ethical design and physical implementation. The v2.0.2 redesign demonstrates a critical shift: diversity is not maintained by "arguing better" (social resistance), but by "seeing differently" (perceptual inversion). Stimulus polarity inversion is not a parameter trick but a structural heterogeneity analogous to sensory inversion or asymmetric information channels in real deliberative systems. By ensuring $u > 0$ even in isolation and deploying heretics with inverted polarity, we ensure that no field can ever perfectly synchronize the network. This philosophical stance, encoded as a constitutional invariant, provides a formal safeguard against the erasure of the "Oracle" state.

6.3 Towards Real-World Deliberation Pilots

We propose a retrospective deliberation pilot applying Mem4ristor v2.0 to real civic decisions (e.g., participatory budgeting or citizens' assemblies). The protocol maps:

- Participants → network units with initial state distributions;
- Arguments and information waves → time-varying stimuli $I_{\text{stimulus}}(t)$;
- Disagreements and conflicts → social stress σ_{social} ;
- Final decision → emergent state distribution over v .

7 Conclusion

Mem4ristor v2.0 is not a mere deliberation framework, but a hardware-ready cognitive primitive that can be integrated as a drop-in resistance layer in existing neuromorphic or collective decision architectures. By extending neuromorphic oscillators with constitutional doubt and structural heretics, we ensure diversity preservation across scales.

This work open several pathways: (1) **Neuromorphic ethics**: encoding values directly in dynamical systems; (2) **Deliberation-aware AI**: tools for diagnosing and improving collective decision processes; (3) **Variability-exploiting hardware**: treating device imperfections as features, not bugs.

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