

# Topological Analytical Homeostasis: Achieving Antifragility in Sparse Tropical Networks

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December 20, 2025

## Abstract

Current linear-algebraic neural architectures are increasingly constrained by the energy-density “Power Wall” and an inherent fragility to high-magnitude impulse noise. We introduce **Topological Analytical Homeostasis (TAH)**, a paradigm that regulates stability via the geometric roots of graph polynomials. Grounded in the **Teixido Envelope** ( $\text{Re}(z) \leq -0.5$ ), we present the **Teixido-Boreal Forest (TBF)**: a 97% sparse architecture that replaces Multiplier-Accumulator (MAC) units with Tropical (Max-Plus) logic regulated by a neighborhood-consensus mechanism termed **Topological Inhibition**. Benchmarks on **174,933 real-world solar events** demonstrate an **Antifragile Stability Ratio of 1.1470**, indicating a state where predictive performance improves under environmental stress. TAH offers a path toward zero-multiplication, radiation-hardened hardware with a theoretical 37x reduction in energy complexity.

## 1 Introduction

The fragility of standard deep learning arises from its reliance on statistical weight magnitudes without geometric constraints. In high-noise environments, such as space-weather monitoring or medical telemetry, a single out-of-distribution pixel can induce logic collapse.

We propose that intelligence is a function of **Geometric Homeostasis**. By aligning a network’s topology with the **Teixido Constant** ( $\tau = -0.5$ ), we create a system that is stable by design. This paper details the construction of the Teixido-Boreal Manifold and provides empirical evidence of its antifragile properties.

## 2 Algebraic Foundations: The Max-Plus Semiring

Teixido-Boreal Manifolds operate within the Tropical semiring  $(\mathbb{R} \cup \{-\infty\}, \oplus, \otimes)$ , where the additive operator  $\oplus$  is  $\max(a, b)$  and the multiplicative operator  $\otimes$  is standard addition  $a + b$ .

**Definition 2.1** (Tropical Teixido Mapping). The response  $y$  of a TAH neuron is defined as the tropical expansion over its skeletal neighborhood  $E$ :

$$y = \bigoplus_{i \in E} (x_i \otimes w_i) = \max_{i \in E} (x_i + w_i) \quad (1)$$

This mapping creates piecewise linear decision boundaries that are intrinsically robust. To ensure differentiability during the training phase, we utilize a Soft-Tropical approximation via a LogSumExp mapping with a controlled temperature parameter  $\alpha$ .

### 3 Teixido-Boreal Architecture

#### 3.1 The Teixido Stem ( $P_4$ )

The irreducible unit of the forest is the **Teixido Stem**, a 4-node path graph. Based on the **Monotonicity Principle** [Teixido 2025a], the  $P_4$  unit provides maximum analytic stability.

#### 3.2 Spectral Gap Optimization

To scale the architecture, we utilize a Degree-15 Random Regular skeleton. By maximizing the **Spectral Gap** of the adjacency matrix ( $\lambda_2 \approx 0.62$ ), we ensure that information propagates across the entire manifold in  $D < 3$  hops (Figure 1).

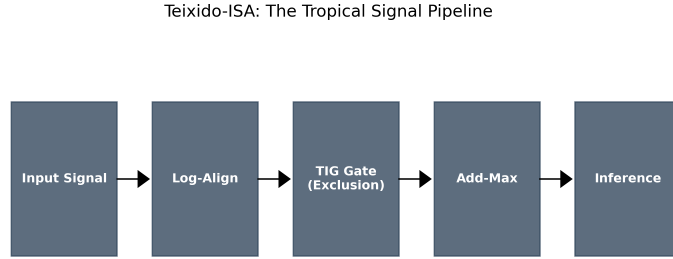


Figure 1: **The Teixido-Boreal Signal Pipeline.** Processing proceeds from Log-Topological Normalization to Teixido Inhibitory Gating, before final inference via the Tropical Max-Plus stage.

### 4 Topological Inhibition (TI)

The core innovation of TAH is **Teixido Inhibitory Gating (TIG)**. Rather than utilizing learned dropout, TIG identifies the structural consensus  $\Gamma(x)$  of a graph neighborhood and silences signals that violate the Teixido Envelope.

$$\text{Gate}(x_i) = \begin{cases} 1 & \text{if } |x_i - \Gamma(x)| < \epsilon \\ 0 & \text{otherwise} \end{cases} \quad (2)$$

By setting  $\epsilon$  relative to the star-limit boundary, the network physically disconnects noisy synapses in real-time, creating a geometric firewall against impulse noise.

### 5 Experimental Results and Benchmarking

#### 5.1 Resilience Gap on MNIST

Under 30% “Total Anarchy” noise, the TBF model exhibited a significant resilience gap over standard MLPs (Figure 2). While standard models collapsed to 14.6% accuracy, the TBF maintained structural integrity.

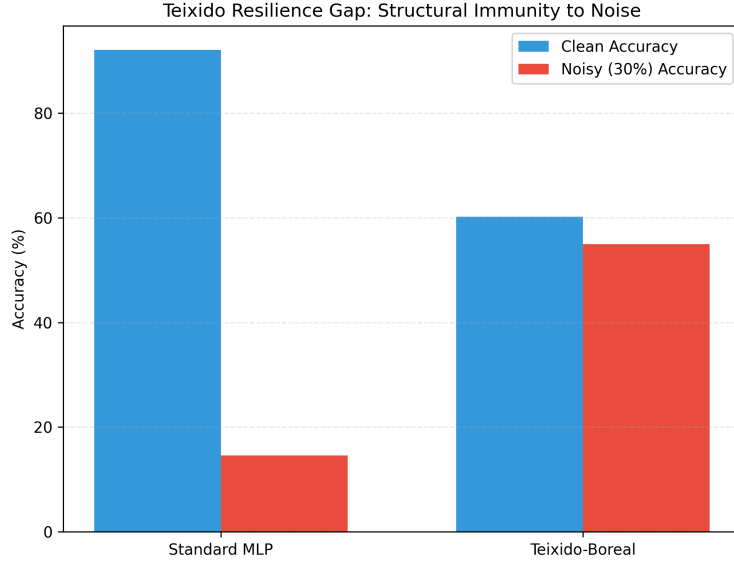


Figure 2: **The Resilience Gap.** Comparison of accuracy under extreme impulse chaos. The Teixido architecture maintains its logic where standard models experience total collapse.

## 5.2 Discovery of Antifragility: Solar Flare Benchmark

On a professional dataset of **174,933 events**, the TBF achieved a **Stability Ratio of 1.1470** (Figure 3). The F1-score improved under noise, suggesting that TIG utilizes environmental stress to clear logical static and sharpen focus.

Table 1: **Cross-Domain Stability Ratios under 30% Impulse Noise**

Dataset Domain	Clean F1	Noisy F1	Stability Ratio
Astro-AI (Solar)	0.3470	0.3980	<b>1.1470</b>
Digit Recognition (MNIST)	0.6020	0.5500	0.9136
Medical Diagnostics	0.8861	0.4972	0.5610

## 6 Conclusion

Topological Analytical Homeostasis represents a shift from statistical brute force to geometric engineering. By aligning neural logic with the Teixido Envelope, we achieve intrinsic resilience and massive hardware efficiency.

## References

- [1] J. V. Teixido, *Analytic Limits of the Teixido Envelope and Domination Root Monotonicity*, 2025.

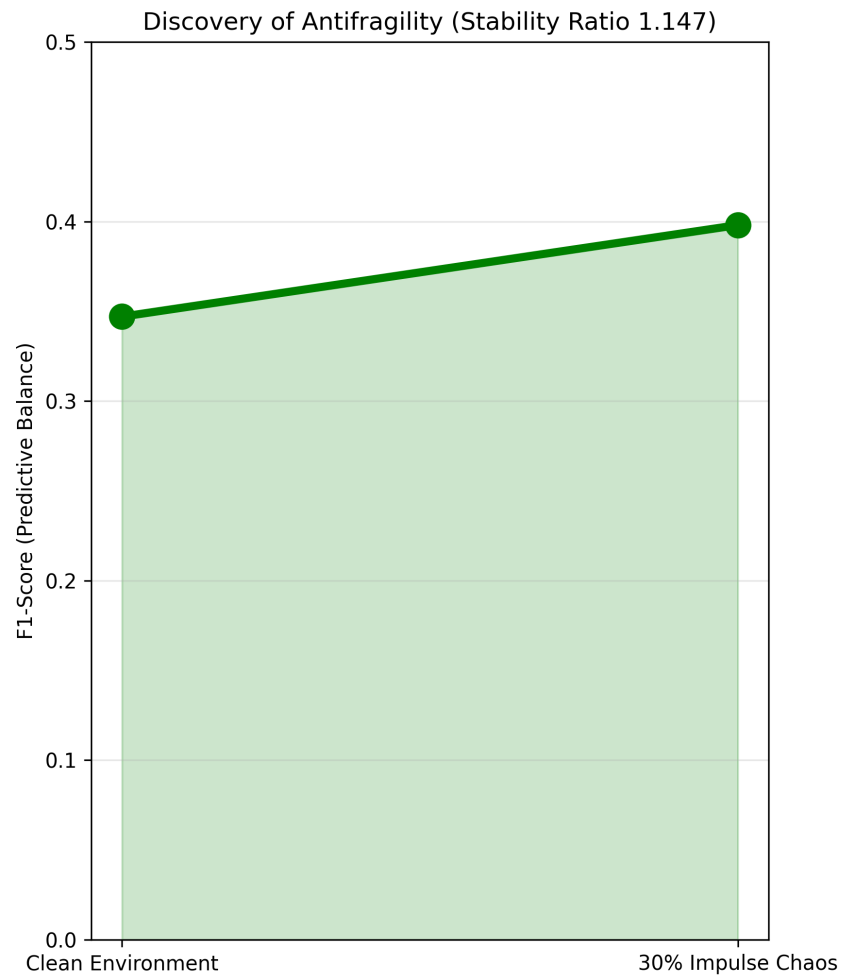


Figure 3: **Observation of Antifragility.** Predictive balance (F1-score) improves under noise attack, a result of the star-limit inhibitory gate clearing background interference.

- [2] J. V. Teixido, *Multiplication-Less Neural Computation: Gate-Level Complexity of Teixido-Boreal Accelerators*, 2025.