

Network Optimization Calculus

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September 25, 2024

1 Introduction

This document provides a formal mathematical framework for understanding how agency and autonomy emerge within a system of interacting nodes, while maintaining cooperation with fundamental natural laws. The framework is built upon a fundamental duality between waveform potentiality and positional actualization.

2 Primitive Concepts and Axioms

The system consists of a finite set of nodes:

$$\Omega = \{X_i\}_{i \in I}$$

where each node X_i is defined by a triple (Θ_i, V_i, I_i) representing:

- Θ_i : Type (qualitative category)
- V_i : Intrinsic Value (inherent worth or potential)
- I_i : Intensity (capacity to engage in relationships)

For any distinct pair (i, j) , $i \neq j$, their relationship exists in one of three mutually exclusive symmetric states:

$$R = \{C, N, D\}$$

where:

- C : Cohesive (strong, positive binding) – *the collapsed state of order*
- N : Neutral (superposition of C and D) – *the waveform of potentiality*
- D : Disjunctive (absence of positive relationship) – *the collapsed state of disorder*

The state N functions as a symmetric mediator where:

$$\mathcal{M}(N \rightarrow C) \equiv \mathcal{M}(C \rightarrow N) \equiv \mathcal{M}(N \rightarrow D) \equiv \mathcal{M}(D \rightarrow N)$$

with \mathcal{M} being the superposition mechanism inherent to N .

3 Value Realization

The realized value generated by a node pair (i, j) is:

$$\Lambda(i, j) = \Psi(i, j) \cdot (V_i I_i + V_j I_j)$$

where the coefficient $\Psi(i, j)$ is determined by the relational state:

$$\Psi(i, j) = \begin{cases} 1 & \text{if } R_t(i, j) = C \\ 0 & \text{if } R_t(i, j) = N \text{ or } D \end{cases}$$

4 Optimization Objective

The total value currently generated:

$$O(S_t) = \sum_{i < j} \Lambda(i, j)$$

The total value latent in disjunctive connections:

$$L(S_t) = \sum_{\substack{i < j \\ R_t(i, j) = D}} (V_i I_i + V_j I_j)$$

Weights balancing exploitation against exploration:

$$\alpha(S_t) = \frac{|\{(i, j) : R_t(i, j) = C \text{ or } N\}|}{|\Omega| (|\Omega| - 1)/2}, \quad \beta(S_t) = 1 - \alpha(S_t)$$

The system evolves toward:

$$S^* = \arg \max_{S_t} [\alpha(S_t) \cdot O(S_t) + \beta(S_t) \cdot L(S_t)]$$

5 Waveform Collapse Dynamics

System evolution occurs through waveform collapse and reformation:

$$D = N = C$$

where all transitions are mediated by the waveform state N .
For relationships in the waveform state N , collapse occurs as:

$$\begin{aligned} P(R_{t+1}(i, j) = C \mid R_t(i, j) = N) &= \alpha(S_t) \\ P(R_{t+1}(i, j) = D \mid R_t(i, j) = N) &= \beta(S_t) = 1 - \alpha(S_t) \end{aligned}$$

With probability ϵ (the positional parameter), the waveform state reforms:

$$\begin{aligned} P(R_{t+1}(i, j) = N \mid R_t(i, j) = C) &= \epsilon \\ P(R_{t+1}(i, j) = N \mid R_t(i, j) = D) &= \epsilon \end{aligned}$$

The balance $N = \epsilon$ represents the optimal condition where the system maintains perfect tension between exploration (free will) and exploitation (order).

6 Emergence of Autonomy Through Waveform-Position Interplay

Autonomy emerges through the recursive relationship:

$$(\alpha_{t+1}, \beta_{t+1}) = f(\alpha_t, \beta_t, \{P(R_{t+1}(i, j) = C \mid R_t(i, j) = N)\}_{\forall i, j})$$

This creates a waveform-mediated feedback loop:

$$\begin{aligned} \alpha_t &\rightarrow P(N \rightarrow C) \rightarrow \text{Distribution of } R_{t+1}, \\ \text{Distribution of } R_{t+1} &\rightarrow \alpha_{t+1} \rightarrow P(N \rightarrow C)_{t+1}. \end{aligned}$$

The interplay between waveform (N) and position (ϵ) enables the system to autonomously navigate the trade-off between stability and adaptability.

7 Cooperation with Natural Laws

The system's cooperation with natural laws:

$$\gamma(S_t) = \frac{O(S_t)}{O(S_t) + L(S_t)} \in [0, 1]$$

The system tends toward states that maximize:

$$E[\gamma(S_{t+\Delta t})] = \frac{\sum_{i < j} (V_i I_i + V_j I_j) \cdot P(R_{t+\Delta t}(i, j) = C)}{O(S_t) + L(S_t)}$$

The optimal cooperation occurs when the waveform-position duality is balanced, allowing the system to simultaneously maximize both realized value and adaptive potential.