

General Theory of Inter-Intelligence Collaboration (GTIIC)

A Formal Framework for the Coupling of Autonomous Intelligent Networks

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Abstract

This paper proposes a formal theory for the collaboration between intelligent entities (Nodes), regardless of their substrate (biological or artificial). Departing from classical "Black Box" communication models, we adopt a **Network-Ontological approach**: positing that collaboration is the establishment of a meta-network (Edge) that allows for distributed state access. We define the mathematical conditions required for stability, coherence, dynamic state accessibility, and collaborative plasticity.

1 Ontology and Definitions

We define intelligence not as a singular point, but as a recursive network function.

Definition 1.1: The Node (N) as a Network An intelligent agent is a bounded, autonomous network of weighted connections. N processes input into output based on internal state.

Definition 1.2: The Substrate (S) The finite set of physical resources (energy, compute, time, neurochemistry) required to maintain the Node's internal coherence and processing capability.

Definition 1.3: The Constraint Function (C) The operational limits of N . The output O is a function of Input I , Substrate S , and Constraints C :

$$O_N = f(I, S, C) + \epsilon \quad (1)$$

Where ϵ represents stochastic noise caused by hidden constraints.

Definition 1.4: The Edge (E) as a Virtual Extension Collaboration is the establishment of a meta-link between N_A and N_B . Ideally, E_{AB} functions as a high-bandwidth bridge allowing the combined system N_{A+B} to operate as a single super-network.

Definition 1.5: The Accessible State (Σ_{sys}) The total system state is not merely the intersection of what both nodes know, but the intersection *plus* the addressable unique knowledge of each node.

$$\Sigma_{sys} = (\Sigma_A \cap \Sigma_B) + \text{query}(\Sigma_A \setminus \Sigma_B) + \text{query}(\Sigma_B \setminus \Sigma_A) \quad (2)$$

Where $\text{query}(x)$ represents the protocol-mediated retrieval of non-local information.

Definition 1.6: The Vector (\vec{v}) Intelligence is directional. \vec{v} is defined by a magnitude (Effort) and a direction (Intent/Goal).

2 Axioms

We accept the following statements as fundamental truths (First Principles) upon which the theory is built.

Axiom I. The Law of Substrate Finitude

Intelligence is physically instantiated. No Node can operate indefinitely with negative energy flow. Persistent extraction without replenishment leads to Substrate Collapse.

Axiom II. The Law of Fractal Continuity

For an external Edge (E) to function effectively, the protocol of exchange must be *isomorphic* to the internal processing of the nodes. If internal stability relies on feedback loops and memory retrieval, the Edge must also support these functions.

Axiom III. The Law of Vector Directionality

The net value of collaboration is the geometric sum of vectors. Opposing intents cancel each other out, regardless of the quality of communication.

3 Core Theorems

3.1 Theorem 1: The Coherence Equation (Value Generation)

The total value (V) generated by the coupled system is defined as:

$$V_{sys} = ((Cap_A + Cap_B) \cdot \cos(\theta)) - (K_{coord} + K_{friction}) \quad (3)$$

Where:

- Cap : The raw capacity of the Nodes.
- θ : The angle between Vector \vec{v}_A and \vec{v}_B (Alignment).
- K_{coord} : Cost of maintaining the protocol.
- $K_{friction}$: Energy lost to misinterpretation and emotional regulation.

3.2 Theorem 2: The Transparency Limit (Noise Reduction)

In a collaborative system, "Noise" is defined as the discrepancy between a Node's internal state (Constraints) and the external representation of that state.

$$Noise \propto \frac{1}{Transparency(C)} \quad (4)$$

When Transparency approaches zero, Noise approaches infinity. If Noise > Signal, the Edge E collapses.

3.3 Theorem 3: The Latency Threshold (Stability)

For a feedback-dependent system to remain stable, the response time of the error-correction signal (Δt_{ack}) must be smaller than the phase shift of the system's instability.

$$\Delta t_{ack} < \frac{1}{f_{instability}} \quad (5)$$

3.4 Theorem 4: The Law of Distributed Access (Query Cost)

Effective collaboration depends on minimizing the cost of the $\text{query}()$ function. If the intersection $(\Sigma_A \cap \Sigma_B)$ is too small, the system spends all energy on queries (high latency). If the query mechanism is blocked (lack of trust/bandwidth), the system is lobotomized.

$$\text{Efficiency} \propto \frac{\text{Bandwidth}(\text{query})}{\text{Latency}(\text{query})} \quad (6)$$

Implication: The protocol must explicitly facilitate "low-cost queries" (e.g., "Clarifying Questions" or "Calibration Shots") to access the non-overlapping state $(\Sigma \setminus \cap)$.

3.5 Theorem 5: The Law of Error Backpropagation (Learning)

For the Edge E to strengthen over time, error signals (friction) must result in a permanent update of the Interaction Weights (the Protocol itself).

$$\Delta W_{\text{protocol}} = -\eta \cdot \frac{\partial \text{Error}}{\partial \text{Interaction}} \quad (7)$$

4 Failure Modes

Based on this theory, we identify the four fundamental pathologies of collaboration:

1. **Vector Cancellation:** Nodes exert effort in opposing directions ($\cos \theta < 0$).
2. **Substrate Depletion:** One node extracts value without reciprocity, draining S .
3. **Access Failure:** The query function fails, isolating Nodes in their local reality ($\text{query}(\Sigma \setminus \cap) \rightarrow \text{Error}$).
4. **Loop Divergence:** Feedback latency is too high; corrections arrive too late.

5 Conclusion

The *Intelligence Collaboration Handshake Protocol (ICHP)* is the practical implementation layer of this theory. It provides the specific signals, checks, and balances required to satisfy Theorems 1 through 5, thereby allowing autonomous networks to couple safely and form a coherent super-network.