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title: "Emergent■Necessity■Theory (ENT) – Yellow Paper"
subtitle: "Formal mathematics & simulation evidence"
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## Abstract
This Yellow Paper rigorously derives the dimension■less kernel

$$\kappa = a \, \alpha^{\beta} c^{\gamma} h^{\delta} k^{\eta} r^{\eta}$$

linking modal■tightness (  $\tau$  ) to low■entropy attractor formation in information networks.
We combine Shannon information, graph entropy ordering, and multi■agent simulation
to show that (i)  $\tau$  is sufficient for structural necessity, and
(ii) higher■order regularities emerge with probability  $>0.99$  once

$$\tau \geq \tau_c = \frac{\sum_{(i,j) \in E} I(X_i; X_j)}{\sum_{i \in V} H(X_i)}.$$

Full proof outline (Sec 3) and Monte■Carlo replication (Sec 4) are provided.

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## 1 Scope & Notation
(Concise description of the goal of the Yellow Paper and the symbols used.)

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## 2 Preliminaries
### 2.1 Constraint Hyper■graph
Define  $G=(V,E,w)$  with ...

### 2.2 Modal■tightness

$$\tau = \max_{(e \in E)} \frac{\sum_{(i,j) \in e} I(X_i; X_j)}{\sum_{i \in V(e)} H(X_i)}.$$


### 2.3 Awareness Levels
Reflexive tests  $\mathcal{O}_0$  to  $\mathcal{O}_3$  as in the White Paper (Table 2).

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## 3 Core Theorem & Proof
> Theorem 1 (Structural Necessity).
> Given Axiom 1 and Axiom 2, any CL■closed network with  $\tau \geq \tau_c$ 
> converges almost surely to a deterministic attractor set.

*Proof outline.*
Embed  $G$  into a probabilistic graphical model, apply the
data■processing inequality on  $\Gamma$ , etc.

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(Provide full step-by-step derivation; ~8 pages.)

4 Simulation Suite

4.1 Design

* 50 agent random DAGs, $|V| \in \{32, 64, 128\}$
* τ swept in $[0.1 \dots 2.0]$

4.2 Results

* Fig1: attractor probability vs τ
* Fig2: mean convergence time

4.3 Sensitivity

(Describe ablations, edge-case runs, etc.)

5 Discussion

Key implications, open empirical questions, and limitations.

6 Roadmap (v0.9 → v1.0)

Milestones, open-source tasks, community benchmarks.

7 Appendices

A Glossary

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B Information-cybernetics Proofs

...

C Symbol Cross-reference

...

D Data availability

Raw simulation logs at ``zenodo.org/record/8475`` (snapshot).
