

Image Dynamics Theory v4: Mathematical Formulation with Structured Display Equations

Abstract

Image Dynamics Theory (IDT) formalizes cognition and consciousness as the evolution of a high-dimensional representational image governed by energy minimization. This document presents a fully mathematical formulation of IDT with all core equations separated as structured display mathematics for clarity and rigor.

1. Representational Image

The internal mental state is represented as a structured high-dimensional image. Each component corresponds to a distinct functional role in cognition.

$$\begin{aligned} z &\in \mathbb{R}^n \\ z &= \{ p, s, r, a, c_p, c_s, \phi \} \end{aligned}$$

2. Slices and Projections

Cognitive slices correspond to low-dimensional projections of the full representational image. These projections define momentary cognitive orientation and intensity.

$$\begin{aligned} \Pi(z) &= [s, r, a] \\ \text{orientation} &= \Pi(z) / \| \Pi(z) \| \\ \text{radius} &= \| \Pi(z) \| \end{aligned}$$

3. Energy Functional

Cognitive dynamics are governed by an energy functional that penalizes incoherence between perception, prediction, intention, and internal structure.

$$\begin{aligned} E(z) &= w_p p - x_p^2 \\ &+ w_s s - x_s^2 \\ &+ w_a a - y_a^2 \end{aligned}$$

$$\begin{aligned}
& + w_{sf} c - c^2 \\
& + w_r - W s^2 \\
& + w_\phi \phi - \text{mean}(\phi)^2 \\
& + w_{reg} Z^2
\end{aligned}$$

4. Microslice Dynamics

State evolution occurs through discrete relaxation steps called microslices. Each microslice corresponds to a gradient descent step on the energy landscape.

$$z_{\text{next}} = z - \eta \nabla E(z)$$

5. Liminality

Liminality is a control parameter that modulates the system's flexibility, noise, and reconfiguration potential.

$$\begin{aligned}
\eta &\leftarrow \lambda \eta \\
\varepsilon &\sim \mathcal{N}(0, \lambda^2 I)
\end{aligned}$$

6. World Prediction

The system learns a predictive world model that enforces consistency between internal dynamics and observed outcomes.

$$\begin{aligned}
m &= f_\theta(z, y) \\
L_{\text{world}} &= m - s^2
\end{aligned}$$

7. Goal Revision

Goals are treated as embedded vectors that can be revised based on experienced outcomes.

$$\begin{aligned}
y' &= y + g_\psi([s, c]) \\
L_{\text{goal}} &= \|y' - y_{\text{ref}}\|^2
\end{aligned}$$

8. Reflection Generation

Reflection is modeled as a conditional language generation process grounded in latent state and episodic context.

$$r_{\square} = h_{\phi}(s, e)$$
$$L_{dec} = - \sum \log P(w_{\square} | s, e)$$

9. Training Objective

All learning objectives are combined into a single weighted loss function optimized in multiple phases.

$$L = L_{world} + \alpha L_{dec} + \beta L_{goal}$$

Conclusion

This formulation presents Image Dynamics Theory as a cleanly structured mathematical system. Separating prose from equations clarifies the relationship between conceptual constructs and their operational realization in an energy-based geometric mind.