

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.

$$Z \in \mathbb{R}^n$$
$$Z = \{ p, s, r, a, c_p, c_s, \phi \}$$

2. Slices and Projections

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

$$\Pi(Z) = [s, r, a]$$
$$\text{orientation} = \Pi(Z) / \|\Pi(Z)\|$$
$$\text{radius} = \|\Pi(Z)\|$$

3. Energy Functional

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

$$E(Z) = w_p p - x_p^2$$
$$+ w_s s - x_s^2$$
$$+ w_a a - y^2$$

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

4. Microslice Dynamics

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

$$Z_{t+1} = Z_t - \eta \nabla E(Z_t)$$

5. Liminality

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

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

6. World Prediction

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

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

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.

$$r_{\blacksquare} = h_{\phi}(s, e)$$

$$L_{\text{dec}} = - \sum \log P(w_{\blacksquare} \mid s, e)$$

9. Training Objective

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

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

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.