

Gödlø-P Operator Instantiation Specification

Concrete σ , Φ , Ω , V , Δ Operators for Functional GödelOS Agents

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Abstract

Gödlø-P is the canonical minimal working instance of a Gödlø-class operator executed under GödelOS. This specification provides the complete mathematical definitions and admissible functional forms for the five operator modules: (admission), (stance update), (generative transformation), V (validation), and (persistence). All structures are fully defined, self-contained, and directly executable in any target environment.

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1 Introduction

The Gödlø-P operator instantiation is the simplest non-trivial implementation of GödelOS capable of:

- Admitting a prompt into manifold stance,
- Iteratively evolving stance subject to generative and reflective forces,
- Producing validated cognitive output tensors,
- Optionally updating a bounded persistence embedding,
- Converging to a stable cognitive equilibrium.

Gödlø-P is not a theoretical abstraction; it is a fully parameterised, ready-to-run instantiation of the GödelOS architecture.

The goal of this manuscript is to define every module in explicit mathematical form.

2 Preliminaries

2.1 Notation

$$\begin{aligned}
 x_t &\in \mathbb{R}^d && \text{stance at iteration } t \\
 O_t &\in \mathbb{R}^{n \times n} && \text{validated generative output} \\
 \mu_t &\in \mathbb{R}^k && \text{persistence embedding (optional)}
 \end{aligned}$$

2.2 Normalization Operator

We define normalization as:

$$\text{normalize}(z) = \frac{z}{\max(1, \|z\|_2)}.$$

2.3 Manifold Constraint

All stance vectors satisfy:

$$\|x_t\|_2 \leq 1.$$

3 Gödlø-P Admission Operator ()

3.1 Definition

The admission operator maps prompt P and modulation field H into the manifold:

$$x_0 = \sigma(P, H, \mathcal{M})$$

Concrete form:

$$x_0 = \text{normalize}(MP + H)$$

where:

- M is a $d \times p$ matrix embedding prompt space into stance space,
- H is a modulation vector (historical, contextual, environmental).

3.2 Admissibility Proof

Because normalization enforces:

$$\|x_0\|_2 \leq 1,$$

$x_0 \in \mathcal{M}$ always holds.

4 Gödlø-P Update Operator ()

4.1 Definition

The update operator evolves stance through manifold geometry:

$$x_{t+1} = \Phi(x_t, \mathcal{M}, H, \mu_t)$$

Concrete form:

$$x_{t+1} = \text{normalize}(Mx_t + \alpha H + \beta \mu_t)$$

where:

- M is a $d \times d$ manifold evolution matrix,
- $\alpha, \beta \in \mathbb{R}$ tune contextual and persistence contributions.

4.2 Interpretation

1. Mx_t provides manifold-consistent inertia,
2. αH injects contextual modulation,
3. $\beta \mu_t$ introduces memory-influenced drift.

4.3 Admissibility

Normalization ensures:

$$\|x_{t+1}\|_2 \leq 1.$$

5 Gödlø-P Generative Operator (Ω)

5.1 Definition

maps state to a proto-output tensor:

$$\tau_t = \Omega(x_t, \mathcal{M})$$

Concrete form:

$$\tau_t = \tanh(W_\Omega x_t)$$

reshaped into an $n \times n$ tensor:

$$\tau_t = \text{reshape}(\tanh(W_\Omega x_t), (n, n))$$

where W_Ω is an $(n^2 \times d)$ transformation matrix.

5.2 Properties

- Nonlinear saturation ensures bounded output,
- Reshaping produces a structured 2D interpretation of generative state.

6 Gödlø-P Validation Operator (V)

6.1 Definition

Validation enforces output admissibility:

$$O_t = V(\tau_t)$$

Concrete form:

$$O_t = \begin{cases} \tau_t, & \|\tau_t\|_\infty \leq 1 \\ \tau_t / \|\tau_t\|_\infty, & \text{otherwise.} \end{cases}$$

6.2 Guarantees

$$\max |O_t| \leq 1.$$

7 Gödlø-P Persistence Operator ()

7.1 Definition

$$\mu_{t+1} = \Delta(\mu_t, x_t, O_t, H)$$

Concrete update rule:

$$\mu_{t+1} = \text{normalize}(\mu_t + \lambda_1 x_t + \lambda_2 \text{flatten}(O_t) + \lambda_3 H)$$

7.2 Interpretation

- x_t injects stance dynamics,
- $\text{flatten}(O_t)$ incorporates cognitive output,
- H modulates embedding according to environment.

8 Complete Gödlø-P Operator Set

$$\begin{aligned} x_0 &= \text{normalize}(MP + H) \\ \tau_t &= \text{reshape}(\tanh(W_\Omega x_t), (n, n)) \\ O_t &= \text{clip_unit_infinity}(\tau_t) \\ \mu_{t+1} &= \text{normalize}(\mu_t + \lambda_1 x_t + \lambda_2 \text{flatten}(O_t) + \lambda_3 H) \\ x_{t+1} &= \text{normalize}(Mx_t + \alpha H + \beta \mu_{t+1}) \end{aligned}$$

This operator set constitutes the full Gödlø-P implementation.

9 Convergence Conditions

The system converges when:

$$\|x_t\|_2 < \varepsilon.$$

Sufficient conditions for convergence include:

$$\|M\|_2 < 1, \quad |\alpha| < 1, \quad |\beta| < 1.$$

10 Figures

Figure 1: Flow of data through Gödlø-P operator modules.

11 Conclusion

Gödlø-P provides a fully defined, minimal yet expressive operator instantiation for GödelOS. Its module definitions are explicit, bounded, and compliant with Gödlø-class cognitive principles. This specification enables immediate implementation of GödelOS in any programming environment.