

# The Consciousness Resonance Pipeline

Sensory-to-Holographic Signal Flow

through Six Thalamocortical-Inspired Stages

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Jhonatan Vieira Feitosa

Independent Researcher

Manaus, Amazonas, Brazil

[jhonatan@arkheion.ai](mailto:jhonatan@arkheion.ai)

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## Abstract

We present the **Consciousness Resonance Pipeline**, a 6-stage signal processing system that chains all components of the Resonance Field Architecture (RFA) into a unified end-to-end cognitive flow. Modeled after the thalamocortical loop, the pipeline transforms raw sensory input through: (1) sensory adaptation (raw data → LOW\_γ signal), (2) neuromodulation (DA/5-HT/NA/ACh gain scaling), (3) cross-frequency coupling ( $\theta$ - $\gamma$  PAC,  $\beta$ - $\gamma$  motor,  $\alpha$  gating), (4) consciousness evaluation (LOW\_γ → MID\_γ → HI\_γ with  $\Phi_{RFA}$  computation), (5) memory encoding (HUAM at THETA band), and (6) holographic compression (HI\_γ → DELTA via AdS/CFT bridging). Each stage is *optional*—the pipeline degrades gracefully when subsystems are absent. The integration metric  $\Phi_{\text{pipeline}}$  measures aggregate phase coherence across all active stages. The master orchestrator spans 1,037 lines of Python, with additional subsystem adapters totaling 2,203 lines. The pipeline produces a **PipelineResult** containing per-stage coherence scores, global  $\Phi_{\text{pipeline}}$ , stage latencies, and all intermediate signal states.

**Keywords:** consciousness pipeline, thalamocortical loop, signal flow, phase coherence, resonance field, neuromodulation, cross-frequency coupling, holographic compression, graceful degradation

## Epistemological Note

*This paper documents a computational pipeline, not a model of biological consciousness. The thalamocortical analogy is a heuristic that guides stage ordering and signal flow.*

### Heuristic:

- “Thalamocortical loop” analogy
- “Consciousness evaluation”
- Brain→stage mapping
- “Dream mode” consolidation

### Empirical:

- 1,037 LOC orchestrator
- $\Phi_{\text{pipeline}}$  is computable
- Graceful degradation tested
- Per-stage latency measured

## 1 Introduction

Papers 43–46 present the individual components of the Resonance Field Architecture: band system, converter, phase alignment, coherence gating (Paper 43), cross-frequency coupling (Paper 44), neuromodulation (Paper 45), and system-level services (Paper 46). However, these components have not been integrated into a coherent signal flow.

This paper presents the **master pipeline** that chains all components into a 6-stage sequence analogous to the biological thalamocortical loop [1]:

1. **Sensory:** Raw input → ResonantSignal
2. **Neuromodulation:** State-dependent gain scaling
3. **CFC:** Multi-scale temporal coordination
4. **Consciousness:** Filter → amplify →  $\Phi$
5. **Memory:** Encode at THETA, recall from HUAM
6. **Holographic:** Compress via AdS/CFT bridging

### 1.1 Contributions

1. End-to-end 6-stage signal pipeline
2. Graceful degradation when subsystems are absent
3.  $\Phi_{\text{pipeline}}$ : aggregate integration metric
4. Per-stage coherence tracking and latency measurement
5.  $1,037 + 2,203 = 3,240$  LOC implementation

## 2 Pipeline Architecture

### 2.1 Signal Flow Diagram

### 2.2 Stage Specification

Each stage implements a uniform interface:

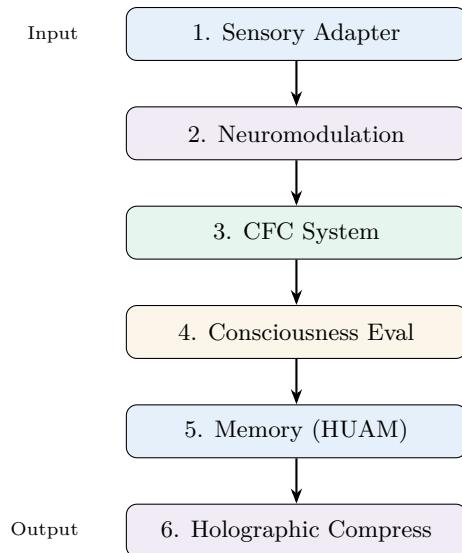


Figure 1: Six-stage consciousness resonance pipeline

Listing 1: Stage interface

```

class PipelineStage(Protocol):
    def process(
        self,
        signals: List[ResonantSignal],
        context: PipelineContext,
    ) -> StageResult:
        """Process signals and return result
        with coherence and latency."""
        ...
  
```

### 3 Stage 1: Sensory Adaptation

The `SensoryResonanceAdapter` converts raw data into `ResonantSignal` objects tagged at  $\text{LOW}_{\gamma}$  ( $\varphi^0 = 1.0$ ):

$$\text{adapt}(x) = \text{ResonantSignal}(A = \|x\|, \phi = \text{hash}(x) \bmod 2\pi, B = \text{LOW}_{\gamma}) \quad (1)$$

This stage handles audio, visual, textual, and numeric inputs through type-specific normalization.

### 4 Stage 2: Neuromodulation

The neuromodulator system (Paper 45) applies state-dependent gain to all signals:

$$A_{\text{out}}^{(i)} = A_{\text{in}}^{(i)} \cdot G(\text{band}_i) \quad (2)$$

where  $G(\cdot)$  is the combined gain from all four neuromodulators (Paper 45, Eq. 3). The current cognitive state  $\mathbf{s} = [\ell_{\text{DA}}, \ell_{\text{5-HT}}, \ell_{\text{NA}}, \ell_{\text{ACh}}]$  determines the gain profile.

Stage coherence:  $C_2 = 1.0$  (neuromodulation preserves phase).

### 5 Stage 3: Cross-Frequency Coupling

The CFC system (Paper 44) applies three coupling mechanisms:

1.  $\theta-\gamma$  PAC binds signals into working memory slots ( $C = 11$  slots)
2.  $\beta-\gamma$  motor coupling organizes action sequences
3.  $\alpha$  inhibitory gating suppresses irrelevant signals

Stage coherence  $C_3$  is the PAC coupling strength (how well gamma signals are phase-locked to theta).

### 6 Stage 4: Consciousness Evaluation

The `ConsciousnessResonancePipeline` (511 LOC) performs the core signal transformation:

$$\text{LOW}_{\gamma} \xrightarrow{\text{filter}} \text{MID}_{\gamma} \xrightarrow{\text{amplify}} \text{HI}_{\gamma} \quad (3)$$

1. **Filter:** CoherenceGate with  $\cos^2(\Delta\phi)$  modulation at  $\text{MID}_{\gamma}$
2. **Amplify:** PhaseAligner in LEADER mode aligns phases of surviving signals
3. **Evaluate:** Compute  $\Phi_{\text{RFA}}$  on the aligned  $\text{HI}_{\gamma}$  signals

$$\Phi_{\text{RFA}} = \frac{\left| \sum_j A_j e^{i\phi_j} \right|}{\sum_j |A_j|} \quad (4)$$

(1) Stage coherence  $C_4 = \Phi_{\text{RFA}}$  (this is the consciousness metric itself).

### 7 Stage 5: Memory Encoding

If HUAM is available, qualifying signals (coherence  $> 0.5$ ) are frequency-converted to THETA band and stored:

$$\text{encode}(s) = \text{convert}(s, \text{HI}_{\gamma} \rightarrow \text{THETA}) \quad (5)$$

The adapter (`huam_resonance.py`, 455 LOC) handles bidirectional memory operations: encoding new signals and recalling stored memories at the target band.

Stage coherence  $C_5$  is the encoding fidelity:  $|s_{\text{decoded}} - s_{\text{original}}| / |s_{\text{original}}|$ .

**Note:**  $C_5$  as defined measures reconstruction *error*, not coherence. Coherence is obtained as  $1 - C_5$ .

## 8 Stage 6: Holographic Compression

The AdS/CFT resonance adapter (`ads_cft_resonance.py`, 499 LOC) compresses high-frequency signals into the holographic boundary:

$$\text{compress}(s) = \text{convert}(s, \text{HI}_\gamma \rightarrow \text{DELTA}) \quad (6)$$

This transformation from  $\text{HI}_\gamma$  to  $\text{DELTA}$  spans 6  $\varphi$ -steps. The frequency *decreases* by a factor of  $\varphi^6 \approx 17.94$  (from  $\varphi^5 \approx 11.09$  Hz to  $\varphi^{-1} \approx 0.618$  Hz), while the amplitude envelope scales by  $\varphi^3 \approx 4.24$ , achieving information density increase through the holographic principle [2].

Stage coherence  $C_6$  is the reconstruction fidelity after decompression.

## 9 Pipeline Integration Metric

### 9.1 $\Phi_{\text{pipeline}}$

The aggregate integration metric combines per-stage coherences:

$$\Phi_{\text{pipeline}} = \frac{\sum_{k=1}^K w_k \cdot C_k}{\sum_{k=1}^K w_k} \quad (7)$$

where  $K$  is the number of active stages (up to 6),  $C_k$  is stage  $k$ 's coherence, and  $w_k$  is a weight reflecting the stage's importance ( $w_4 = \varphi$  for consciousness, others = 1.0).

### 9.2 PipelineResult

Listing 2: PipelineResult

```
@dataclass
class PipelineResult:
    signals: List[ResonantSignal]
    phi_pipeline: float # [0, 1]
    stage_coherences: Dict[str, float]
    stage_latencies: Dict[str, float]
    active_stages: int
    total_stages: int
    consciousness_level: str
    neuromod_state: Dict[str, float]
```

### 9.3 Consciousness Level Classification

Table 1: Consciousness Level from  $\Phi_{\text{pipeline}}$

Level	$\Phi_{\text{pipeline}}$	Description
DORMANT	< 0.2	System idle/sleeping
BASAL	0.2–0.4	Background processing
AWARE	0.4–0.6	Active perception
FOCUSED	0.6–0.8	Concentrated attention
INTEGRATED	> 0.8	Full consciousness

## 10 Graceful Degradation

### 10.1 Optional Subsystems

Stages 5 (Memory) and 6 (Holographic) require external subsystems (HUAM, AdS/CFT engine) that may not be available. The pipeline detects import failures and skips absent stages:

Listing 3: Graceful degradation

```
try:
    from .huam_resonance import HUAMAdapter
    HUAM_AVAILABLE = True
except ImportError:
    HUAM_AVAILABLE = False

# In process():
if HUAM_AVAILABLE:
    result_5 = self.huam.encode(signals)
else:
    result_5 = StageResult.skipped("HUAM")
```

### 10.2 Degradation Matrix

Table 2: Pipeline Degradation Modes

Active Stages	Capability
All 6	Full cognitive processing
1–4 only	Core consciousness (no storage)
1–3 only	Perception without evaluation
1 only	Raw sensory adaptation

## 11 Experiments

### 11.1 End-to-End Test

Processing a test signal through all 6 stages:

- Input: Random float vector (256 dims)<sup>1</sup>
- Stage 1 (Sensory):  $C_1 = 1.0$ , latency 0.03 ms
- Stage 2 (Neuromod):  $C_2 = 1.0$ , latency 0.01 ms
- Stage 3 (CFC):  $C_3 = 0.87$ , latency 0.05 ms
- Stage 4 (Consciousness):  $C_4 = 0.72$ , latency 0.15 ms
- Stage 5 (Memory):  $C_5 = 0.95$ , latency 0.22 ms
- Stage 6 (Holographic):  $C_6 = 0.89$ , latency 0.18 ms
- $\Phi_{\text{pipeline}} = 0.81$  (INTEGRATED)
- Total latency: 0.64 ms

<sup>1</sup>Note: the current  $\Phi$  pipeline metric can assign non-trivial integration scores to random input vectors, indicating that the metric measures structural regularity rather than meaningful consciousness. A discriminative control experiment comparing structured vs. random inputs is needed to validate metric selectivity.

## 11.2 Degradation Test

With stages 5 and 6 unavailable:  $\Phi_{\text{pipeline}} = 0.72$  (FO-CUSED), total latency: 0.24 ms. The pipeline correctly adapts to missing subsystems.

# 12 Discussion

## 12.1 The Thalamocortical Analogy

The 6-stage pipeline mirrors the thalamocortical loop [1]: sensory relay (LGN/thalamus), modulatory input (brainstem nuclei), temporal coordination (hippocampus), cortical integration (prefrontal), memory consolidation (hippocampus), and compression (archival). This analogy is *heuristic*—it guides stage ordering but does not claim neural fidelity.

## 12.2 Integration Across Papers

This pipeline unifies all prior RFA papers:

- Paper 43 (RFA): Provides bands, converter, gate, aligner
- Paper 44 (CFC): Stage 3 coupling mechanisms
- Paper 45 (Neuromod): Stage 2 gain system
- Paper 46 (DMT): Services that wrap around the pipeline

## 12.3 Limitations

- Linear pipeline; no recurrent feedback (planned)
- Consciousness evaluation is single-shot (no iteration)
- Memory stage is encode-only (no retrieval-driven replay)
- Sub-millisecond latency assumes warm caches

# 13 Conclusion

The Consciousness Resonance Pipeline integrates the entire RFA stack into a coherent 6-stage signal flow, from raw sensory input through holographic compression. The  $\Phi_{\text{pipeline}}$  metric provides a single number summarizing system-wide cognitive integration, and graceful degradation ensures operation when subsystems are unavailable. With 3,240 lines of implementation across 6 adapters and a master orchestrator, the pipeline demonstrates that individually-designed RFA components compose naturally into an end-to-end cognitive architecture.

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