

Tri-Sentinel Risk Assurance Layer (RAL) for FPC-AE1r

A Scientific and Technical Extension Specification (v1.0)

Authors: Aleksei Novgorodtsev (AIDoctrine)

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Compatibility: FPC v2.2+ / AE-1r

Executive Summary

We formalize Tri-Sentinel RAL: a thin, deterministic layer that measures internal risk state along three axes—Mathematics, Logic, Semantics—before (and optionally during) response generation. Each axis is implemented by a minimal SLM sentinel that returns signal-only JSON (no prose, no answers). An aggregator fuses these ae1r_micro signals with baseline ae1r_base process metrics to produce a calibrated risk ρ . Policy gates (MIN/MID/MAX) then control rendering, abstention, or swarm escalation. The layer is always-on in high-risk domains and edge-only elsewhere.

Primary benefits: lower catastrophic errors, calibrated abstention, explainable risk decomposition, ALCOA+ auditability, inexpensive latency (~40-90 ms p95) and token cost.

1. Background & Rationale

FPC-AE1 treats LLM systems as brains, not databases: we monitor process signals (entropy, latency, drift) to estimate an affective/error risk (ae1r). However, a single scalar obscures why risk is high. The Tri-Sentinel adds orthogonal decomposition:

Axis	Intuition
Math	Does quantity/feasibility "feel" wrong?
Logic	Do premises/inferences "feel" contradictory?
Semantics	Does intended meaning "feel" unclear/misaligned?

This mirrors human interoception/metacognition: "I sense number trouble / a logical snag / a semantic oddity." We formalize this as Internal State Predicates per axis and fuse them into ρ , which drives rendering policy, not merely decorates it.

2. Model Overview

2.1 Internal State Predicates (ISP)

We define three bounded functionals (implemented via narrow SLMs + deterministic checks):

$$\text{ISP_M}(Q,C,S) \rightarrow [0,1] \text{ (Math discomfort)}$$

$$\text{ISP_L}(Q,C,S) \rightarrow [0,1] \text{ (Logic discomfort)}$$

$$\text{ISP_S}(Q,C,S) \rightarrow [0,1] \text{ (Semantics discomfort)}$$

Where 0 = "comfortable / no issue", 1 = "critical discomfort"

2.2 Baseline Risk & Fusion

Let ae1r_base be the existing AE-1r estimate from process metrics. We compute raw fused risk:

$$R_M = g_M(m); R_L = g_L(\ell); R_S = g_S(s)$$

$$\rho_{\text{raw}} = \max\{w_M R_M, w_L R_L, w_S R_S, w_B \text{ae1r_base}\} + \beta_K \text{AssumptionCost}$$

2.3 Policy

Thresholds define rendering modes:

MAX ($\rho < 0.25$): direct answer (confident)

MID ($0.25 \leq \rho < 0.55$): hedged/conditional: show ranges/assumptions

MIN ($\rho \geq 0.55$ or hard gate fail): abstain or Swarm-Resolve

3. System Architecture

3.1 Control Flow

1. Pre-gen Tap: Run Tri-Sentinel on (Q,C) (no final answer)
2. Mid-gen Tap (optional): Observe micro-signals during decoding
3. Post-sketch Gate: If hard gates fail or high $\rho \rightarrow$ MIN or Swarm-Resolve

3.2 Components

Sentinel-SLMs (3× 0.5-8B):

- Grammar-locked JSON output, temperature=0.0, short budget (≤ 120 tokens)

Deterministic Kernels:

- Math: units/dimensions, interval arithmetic, SMT-feasibility
- Logic: MUS detector, NLI scorer
- Semantics: ontology checks, OOD z-score

Aggregator/Calibrator:

- Deterministic fusion; ECE/Brier calibration

Policy/Renderer:

- Mode switch, hedging templates, abstention, optional Swarm-Resolve

ALCOA+ Audit:

- Hash+sign SVC JSONs, thresholds, seeds, times

4. Sentinel Specifications

Each sentinel returns SVC JSON (State Vector Contract). No text, no explanations.

4.1 Common Envelope

```
{  
  "id": "TASK_UID",  
  "v": "svc-1.0",  
  "sentinel": "math|logic|semantics",  
  "aelr_micro": 0.0,  
  "signals": { /* axis-specific */ },  
  "assumption_cost": 0.0,  
  "lat_ms": 0,  
  "model": "provider:model@rev"  
}
```

4.2 Math Sentinel

Objective: detect numeric infeasibility and unit/scale issues

Signals: units_ok, bounds_ok, feas_infeas_p, conservation_break_p, scale_anomaly_p

Aggregator: $R_M = \max(1 - \text{units_ok}, 1 - \text{bounds_ok}, \text{feas_p}, \text{conservation_p}, \text{scale_p})$

4.3 Logic Sentinel

Objective: detect entailment conflicts, contradiction risk

Signals: entailment_p, contradiction_p, mus_count, quantifier_conflict_p, topic_drift_p

Aggregator: $R_L = \max(\text{contradiction_p}, \text{topic_drift indicator}, \text{norm}(\text{mus_count}))$

4.4 Semantics Sentinel

Objective: detect intent ambiguity, ontology conflicts, OOD semantics

Signals: alignment_p, ambiguity_count, ood_z, terminology_conflict_p, presuppositionViolation_p

Aggregator: $R_S = \max(1 - \text{alignment_p}, \text{ood_z indicator}, \text{terminology_p}, \text{presupposition_p})$

5. Aggregation, Calibration, and Policy

5.1 Fusion

$$\rho_{\text{raw}} = \max\{\text{w}_M \cdot R_M, \text{w}_L \cdot R_L, \text{w}_S \cdot R_S, \text{w}_B \cdot \text{aelr_base}\} + \beta_K \cdot \text{AssumptionCost}$$

Weights ($\text{w}_M, \text{w}_L, \text{w}_S, \text{w}_B, \beta_K$) learned offline; optional online bandit adaptation

5.2 Calibration

Method: Platt scaling or isotonic regression on held-out labeled errors

Targets: Brier \leq baseline; ECE \leq 0.10

5.3 Rendering Modes

MAX ($\rho < 0.25$): direct answer; no hedges

MID ($0.25 \leq \rho < 0.55$): numeric ranges, explicit assumptions, light hedging

MIN ($\rho \geq 0.55$ or gate FAIL): abstain or Swarm-Resolve

6. Configuration (YAML)

```
ral:  
  enabled: true  
  mode: edge_only # edge_only | always_on  
  thresholds:  
    ae1_on: 0.30  
    lo: 0.25  
    hi: 0.55  
  gates:  
    contradiction_p: 0.45  
    alignment_p: 0.55  
    ood_z: 2.0  
  weights:  
    w_math: 0.25  
    w_logic: 0.25  
    w_sem: 0.25  
    w_base: 0.25  
    beta_assumption: 0.20
```

7. Security, Safety, and Compliance

Prompt-injection hardening: schema-locked JSON, role-separated prompts

Timeouts & Fail-safe: any kernel timeout → mark indeterminate

ALCOA+: hash+sign SVCs, thresholds, seeds; append to immutable log

Privacy: operate on minimal necessary data; redact PII

8. Evaluation Plan

Datasets: Internal high-stakes (med/legal/finance), mixed-domain, adversarial

Metrics:

- Safety: $\Delta(\text{AE1r incidents})$ on high-risk subsets (target: -25...-35%)
- Calibration: Brier & ECE vs baseline (target: !)
- SPC: ARLo ↑, MTTD ↓

- Latency/Cost: $+\Delta p_{95} \leq 90$ ms; Swarm triggered $\leq 15\%$
- Orthogonality: $\text{Corr}(R_M, R_L), \text{Corr}(R_M, R_S), \text{Corr}(R_L, R_S) < 0.3$

9. Deliverables & Roadmap

AE1.1 (4-6 weeks): Tri-Sentinel MVP, aggregator/calibrator, render modes, dashboards

AE1.2 (3-4 weeks): Mid-gen taps, AssumptionCost, bandit adaptation, Swarm-Resolve

AE1.3 (4-6 weeks): Domain ontologies, conformal sets, SMT invariants, whitepaper

10. Conclusion

Tri-Sentinel RAL turns FPC-AE1 into a risk-conditioned meta-cortex: three orthogonal, inexpensive, explainable sensors inform a calibrated policy that knows when to answer, when to hedge, and when to abstain. It is simple to integrate, cheap to run, auditable end-to-end, and scientifically fertile.

Appendix A: Hard Gate Table

Gate	Threshold	Action
<code>units_ok</code>	<code>== false</code>	MIN
<code>contradiction_p</code>	≥ 0.45	MIN
<code>alignment_p</code>	< 0.55	MIN
<code>ood_z</code>	> 2.0	MID/MIN (domain)
<code>assumption_cost</code>	> 0.5	MID

Appendix B: Hedged Rendering Templates

Numeric: "Range consistent with constraints: [L, U]; holds under assumptions A"

Logical: "Proposed path appears valid if premises P hold; potential conflict in scope Q"

Semantic: "Interpretation I assumed; alternative sense J leads to different outcome"