Hierarchical Actor-Orchestrated State Management with DIRAM-Backed Epistemic Validation

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Abstract—We present the hierarchical state resolution model for Actor-orchestrated systems, extending the OBIAI Actor class through sub-conceptual task decomposition with DIRAM-backed memory governance. Each EA Actor autonomously manages task lifecycles using a TO-DO \rightarrow DOING \rightarrow DONE progression model, maintaining epistemic validation at 95.4% confidence threshold. The system implements strategic rollback cascades when success:failure ratios fall below 1:2, ensuring self-correcting behavior through cryptographically traced state transitions. This architecture represents deployed production infrastructure, not theoretical design, providing forensic-level accountability through SHA-256 receipt logs and verb-noun conceptual modeling aligned with the Actor class tuple $\alpha=(S,\mathcal{C},\Phi,\Psi,\epsilon)$.

I. INTRODUCTION

The hierarchical state resolution model extends the Actor class defined in the OBIAI framework through systematic sub-conceptual decomposition. Building upon the categorical foundation where Actors navigates infinite-dimensional semantic manifolds, we implement a production-ready state management system that main tains epistemic discipline while enabling autonomous task orchestration.

Definition 1 (Actor Class Extension). Given an Actor $\alpha = (S, \mathcal{C}, \Phi, \Psi, \epsilon)$ where $\epsilon \geq 0.954$, the hierarchical state extension introduces:

- Sub-conceptual decomposition function $D: S \to 2^{\mathcal{S}}$
- State lifecycle automaton $L: \mathcal{S} \times \mathcal{C} \to \mathcal{S}$
- DIRAM trace function $T: \mathcal{S} \to \{0,1\}^{256}$

This extension enables Actors to decompose high-4 level missions into epistemically validated sub-tasks while maintaining the dimensional innovation property essential to the Actor paradigm.

II. DIRAM HARDWARE FAULT-TOLERANT ARCHITECTURE

A. Core State Structure

The hierarchical state management system anchors to DIRAM's cryptographic memory governance through the following C structure:

```
typedef struct {
   uint64_t state_id;
    char parent_state_hash[65];
                                      // SHA-256
        trace
    verb_noun_concept_t intent;
    float result_metric;
    float proof_confidence;
       0.954
    state_flag_t status_flag;
       Lifecycle position
   uint8_t error_count;
   uint64_t timestamp;
   diram_state_allocation_t* diram_trace;
} hierarchical_state_t;
typedef enum {
   STATE\_TODO = 0x01,
   STATE\_DOING = 0x02,
   STATE\_DONE = 0x04,
   STATE\_BLOCKED = 0x08,
   STATE ROLLEDBACK = 0x10
 state_flag_t;
```

Listing 1. DIRAM-backed hierarchical state structure

B. Memory Allocation with Trace Linking

Every state allocation generates a cryptographic receipt ensuring forensic traceability:

```
diram_state_allocation_t*
    diram_allocate_state_memory(
    hierarchical_state_t* state,
    const char* intent_tag
) {
    // Enforce epistemic constraint
    if (state->proof_confidence <
        EPISTEMIC_THRESHOLD) {
        return NULL;
    }
}</pre>
```

```
8
                                                       17
9
        // Generate SHA-256 receipt
10
                                                       18
       diram_allocation_t* base =
                                                       19
            diram_alloc_traced(
                                                       20
            sizeof(hierarchical_state_t),
                intent_tag);
13
                                                       23
        // Link to blockchain for audit trail
14
15
       gitraf_blockchain_append_state(
                                                       25
            state->state_id,
                                                       26
16
            state->parent_state_hash);
17
                                                       27
18
       return create_state_allocation(base, state
19
            );
20
```

Listing 2. DIRAM state allocation implementation

III. TASK LIFECYCLE MANAGEMENT WITH WATERFALL GATES

A. State Transition Automaton

The lifecycle progression follows a deterministic automaton with epistemic validation at each gate:

Theorem 1 (Lifecycle Soundness). For any state $s \in \mathcal{S}$ with confidence $c_s \geq 0.954$, the transition function L guarantees that $L(s, \mathcal{C}) = s'$ implies that the verify-trace- Φ operation validates the transition $(s \to s')$ as TRUE.

Proof. Each transition invokes the audit-transition- Φ function which validates the epistemic signature Φ before permitting state advancement. The DIRAM trace function T generates cryptographic proof of transition validity.

B. Waterfall Gate Implementation

```
int enforce_waterfall_gate(
       hierarchical_state_t* state,
2
       waterfall_gate_t gate
3
   ) {
4
       switch (gate) {
5
            case GATE_1_TODO_VALIDATION:
6
                if (state->proof_confidence <</pre>
7
                    0.954) {
                    state->status_flag =
8
                        STATE_BLOCKED;
                    emit_trace("GATE_1_FAILED",
9
                        state->state_id);
                    return -1;
10
11
                break;
            case GATE_2_DOING_PROGRESS:
14
15
                float ratio =
                    calculate_success_failure_ratio
                if (ratio < 0.5) { // Below 1:2</pre>
16
                    threshold
```

Listing 3. Waterfall gate enforcement

IV. ROLLBACK CASCADE PROTOCOL

A. Strategic Rollback Mechanism

When trial-and-error patterns emerge (error_count \geq 2), the system initiates the emit-rollback- Φ operation:

Algorithm 1 Cascade Rollback Protocol

```
1: Input: Failed state s_f with confidence c_f < 0.954
2: Output: Rollback cascade receipt R
3: D
                 trace-dependency (s_f)
                                            {Using
    dependency-\Phi}
4: depth \leftarrow \min(|D|, 5) {Limit cascade depth}
5: for d = 0 to depth do
      S_d \leftarrow \{s \in D : \operatorname{depth}(s) = d\}
7:
      for each s \in S_d do
         s.confidence \leftarrow s.confidence \times (1 - 0.1d)
8:
9:
         s.status \leftarrow \mathsf{STATE\_TODO}
         memoize-delta(s, c_f) {Using memoize-delta-
10:
         generate-receipt(s) {Using generate-receipt-\Phi}
11:
      end for
12.
13: end for
14: return append-trace(R) {Using append-trace-\Phi}
```

B. Success: Failure Ratio Enforcement

The system maintains epistemic discipline through continuous ratio monitoring:

```
def assess_state_continuation(self, state):
    """Implements trial-and-improvement with
        rollback"""
    # Check trial-and-error lock
    if state.confidence < 0.954 and state.
        error_count >= 2:
        return self._initiate_rollback(state)

# Check success:failure ratio
    ratio = self._calculate_success_ratio(
        state)
```

Listing 4. Python implementation of ratio enforcement

V. ACTOR SUB-CONOPS INTEGRATION

A. Alignment with Actor Class Tuple

The hierarchical state model preserves the Actor's dimensional innovation property while adding structured task management:

Proposition 1 (Innovation Preservation). For Actor $\alpha = (S, \mathcal{C}, \Phi, \Psi, \epsilon)$ with hierarchical extension, the dimensional innovation property holds:

```
\exists \tau: S \to S \text{ where } \tau \notin \operatorname{span}(\mathcal{C}) \implies \exists s \in \mathcal{S}: D(\tau(S)) \ni s
```

This ensures that Actor-driven innovations translate to actionable sub-tasks while maintaining epistemic boundaries.

B. Verb-Noun Conceptual Modeling

Each state intent follows the formalized triplet structure (V, N, Φ) :

```
typedef struct {
       char verb[32];
                            // Action operation
       char noun[32];
                           // Domain object
       float phi_vector[8]; // Epistemic
           signature
   } verb_noun_concept_t;
   // Example instantiation
   verb_noun_concept_t intent = {
9
       .verb = "predict",
       .noun = "failure",
10
       .phi_vector = \{0.97, 0.95, 0.98, 0.96,
11
                       0.94, 0.99, 0.95, 0.97
   };
```

Listing 5. Verb-noun concept implementation

VI. TURING SOUNDNESS IN TASK DECOMPOSITION 7

Theorem 2 (Decomposition Completeness). The hier-parchical state system with DIRAM backing achieves Turing-complete task orchestration while maintaining epistemic soundness.

Proof. We construct a correspondence between state transitions and Turing machine computation:

- 1) States in S encode Turing configurations
- Lifecycle transitions simulate state machine evolution
- 3) DIRAM provides unbounded memory through linked allocations
- 4) Rollback mechanism implements rejection states
- 5) The validate-confidence- Φ operation ensures only sound computations proceed

The 95.4% threshold prevents non-deterministic branching while cascade protocols enable recovery from computational dead-ends.

VII. COMPLIANCE AND AUDIT FRAMEWORK

A. AEGIS-PROOF Traceability

Every state transition generates auditable proof through:

- commit-state-Φ: Persistence with cryptographic receipt
- anchor-hardware- Φ : Physical memory binding for forensics
- compute-ratio-Φ: Continuous success metric validation

B. NASA-STD-8739.8 Adherence

The system satisfies safety-critical requirements through:

- 1) **Deterministic Execution**: State transitions follow formal automaton
- 2) **Bounded Resources**: DIRAM enforces $\epsilon(x) \le 0.6$ constraint
- 3) **Graceful Degradation**: Cascade rollback prevents catastrophic failure
- Formal Verification: All paths traceable through SHA-256 receipts

VIII. PRODUCTION DEPLOYMENT ARCHITECTURE

```
12
           # Process each state through lifecycle
13
           for state in states:
14
                while state.status_flag !=
15
                    STATE_DONE:
                    transition = self.
16
                        process_state_lifecycle(
                        state)
                    if transition ==
                        StateTransition.ROLLBACK:
19
                        self.
                            handle_cascade_recovery
                            (state)
                    elif transition ==
20
                        StateTransition.BLOCKED:
                        self.resolve_dependencies(
                            state)
           return self.compile_mission_proof(
23
               states)
```

Listing 6. Complete orchestrator implementation

IX. CONCLUSION

The hierarchical Actor-orchestrated state management system represents deployed infrastructure achieving selfcorrecting AI orchestration through:

- DIRAM-backed memory governance with cryptographic traceability
- 95.4% epistemic validation threshold enforcement
- Strategic rollback cascades maintaining 1:2 success:failure ratios
- Verb-noun conceptual modeling for semantic task representation
- Waterfall gate compliance for systematic validation

This architecture operates continuously across OBINexus deployments, transforming Actor-level dimensional innovations into tractable, verifiable sub-tasks while maintaining the mathematical rigor demanded by safety-critical AI systems.

VERB-NOUN CONCEPT GLOSSARY

- anchor-Baird wepistermic state to physical memory substrate. 3
- append Archest the transition to immutable DIRAM log.
- audit-transpectns that e lifecycle compliance with confidence metrics. 2
- commi**Fixtalez** state persistence to DIRAM with receipt generation. 3
- compute about the success: failure metrics for cascade detection. 3

- emit-roffbackate rollback event with epistemic signature for state recovery. 2
- generaterocheipt SPIA-256 trace for forensic accountability. 2
- memoi Steodel tao Afridence degradation for future reference. 2
- trace-dependence of the trace-dependence of the trace of
- validateAssessidermeef_Deonfidence against 95.4% threshold. 3
- verify-**Wadidate** cryptographic integrity of state transition history with epistemic signature Φ . 2