

Universal Stochastic Predictor
Phase 4: IO Layer
v2.1.0 (Level 4 Autonomy)

Implementation Team

February 19, 2026

Contents

1	Phase 4: IO Layer Overview	2
1.1	Tag Information	2
1.2	Scope	2
1.3	Design Principles	2
2	Ingestion and Validation	3
2.1	Implementation Modules	3
2.2	Catastrophic Outlier Filter	3
2.2.1	Implementation Notes	3
2.3	Frozen Signal Alarm	3
2.3.1	Recovery Criteria (V-MAJ-6: Frozen Signal Recovery Ratio)	4
2.4	Staleness Policy (TTL)	5
2.4.1	Implementation Notes	5
3	Telemetry Abstraction	6
3.1	TelemetryBuffer Emission	6
3.1.1	Implementation Notes	6
3.2	No Compute Stalls	6
4	Deterministic Logging	7
4.1	Hash-Based Parity Checks	7
4.2	Audit Hashes	7
5	Snapshot Strategy	8
5.1	Atomic Persistence	8
5.2	Binary Serialization	8
5.2.1	Implementation Notes	8
5.3	Integrity Verification	8
5.4	Atomic Write Protocol	8
5.4.1	Implementation Notes	9
6	Security Policies	10
6.1	Credential Injection	10
6.1.1	Implementation Notes	10
6.2	Version Control Exclusion	10
7	Orchestrator Integration	11
7.1	Ingestion Gate in orchestrate_step()	11
7.1.1	Execution Flow	11
7.1.2	Flag Semantics	12
7.1.3	Early Return on Rejection	12
7.2	PRNG Configuration	12

7.3	64-bit Precision Enforcement	12
8	Telemetry Buffer Integration (P2.3)	13
8.1	Motivation	13
8.2	Data Model	13
8.3	Integration into orchestrate_step()	14
8.4	Configuration Parameters	14
8.4.1	Configuration Injection Example	14
8.5	Thread-Safety Model	15
8.6	Backward Compatibility	15
8.7	Usage Example	15
8.8	Benefits	16
8.9	Visualization Dashboard (GAP-6)	16
8.9.1	Usage	16
8.9.2	Notes	16
9	Level 4 Autonomy: Configuration Mutation Safety	17
9.1	Overview	17
9.2	V-CRIT-2: Atomic Configuration Mutation Protocol	17
9.2.1	Problem Statement	17
9.2.2	Requirements	18
9.2.3	Implementation	18
9.2.4	Usage Example	21
9.2.5	Files Modified	21
9.2.6	Compliance Impact	21
9.3	V-MAJ-4: Configuration Mutation Rate Limiting	22
9.3.1	Problem Statement	22
9.3.2	Safety Requirements	22
9.3.3	Implementation	22
9.3.4	Usage Pattern	23
9.4	V-MAJ-5: Degradation Detection Auto-Rollback	24
9.4.1	Problem Statement	24
9.4.2	Safety Requirements	24
9.4.3	Implementation	24
9.4.4	Usage Pattern	26
9.4.5	Audit Log Format	27
9.5	V-MAJ-7: Adaptive Telemetry Monitoring	27
9.5.1	Implementation Status	27
9.5.2	InternalState Extensions	27
9.5.3	Adaptive Telemetry Collection	27
9.5.4	Integration Pattern	28
9.5.5	Metrics Tracked	29
9.5.6	Future Extensions	29
9.6	Utility Functions	30
9.7	Implementation Status	30
10	Compliance Checklist	31
11	Phase 4 Summary	32

Chapter 1

Phase 4: IO Layer Overview

1.1 Tag Information

- **Tag:** impl/v2.1.0
- **Commit:** 6ccb68d (GAP-6 dashboard implementation + e929bc5 final alignment)
- **Status:** Level 4 Autonomy compliance (V-MAJ-4, V-MAJ-5, V-MAJ-7 implemented; GAP-6 complete)

Phase 4 introduces the asynchronous I/O layer for snapshots, streaming, and telemetry export. The primary design goal is to preserve JAX/XLA throughput by decoupling compute from disk or network latency.

1.2 Scope

Phase 4 covers:

- **Telemetry Buffering:** Non-blocking emission of telemetry snapshots
- **Deterministic Logging:** Hash-based parity checks for CPU/GPU validation
- **Snapshot Strategy:** Atomic persistence of predictor state
- **Ingestion and Validation:** Input filtering, staleness policy, frozen signal detection
- **Security Enforcement:** Credential injection and secret exclusion
- **IO Modules:** validators, loaders, telemetry, snapshots, credentials

1.3 Design Principles

- **No Compute Stalls:** JAX compute threads never block on I/O
- **Determinism:** Logs capture reproducible hashes instead of raw state dumps
- **Security:** No raw signals or secrets in logs
- **Configurability:** Logging intervals and destinations injected via config
- **Integrity:** Snapshots and parity logs are hash-verified

Chapter 2

Ingestion and Validation

2.1 Implementation Modules

Phase 4 IO introduces the following modules:

- `io/validators.py`: Outlier, frozen signal, and staleness checks
- `io/loaders.py`: Ingestion gate and decision flags
- `io/telemetry.py`: Non-blocking telemetry buffer and parity hashes
- `io/snapshots.py`: Binary snapshots, hash verification, atomic writes
- `io/credentials.py`: Environment-based credential injection helpers

2.2 Catastrophic Outlier Filter

Input validation must reject catastrophic outliers when $|y_t| > 20\sigma$ relative to historical normalization. In this case, the system must preserve inertial state and emit a critical alert without advancing the transport update.

- Reject observation and keep current state unchanged.
- Emit a critical alert for audit visibility.
- Do not update JKO/Sinkhorn weights for the rejected step.

2.2.1 Implementation Notes

Outlier detection is implemented as a pure function with configuration-driven thresholds. The ingestion gate returns a decision object that preserves inertial state when an outlier is detected.

2.3 Frozen Signal Alarm

If the exact same value is observed for $N_{freeze} \geq 5$ consecutive steps, emit a `FrozenSignalAlarmEvent`. This invalidates the multifractal spectrum and requires:

- Freeze the topological branch (Kernel D).
- Switch to degraded inference mode.
- Continue monitoring until signal variation resumes.

2.3.1 Recovery Criteria (V-MAJ-6: Frozen Signal Recovery Ratio)

The frozen signal lock is released when variance recovers above a configurable ratio of historical variance for a configurable number of consecutive steps.

Algorithm

$$\text{recovered} = \text{detect_frozen_recovery}(\text{variance_history}, \text{historical_var}, \rho, n_c) \quad (2.1)$$

where:

- `variance_history`: Recent residual variances
- `historical_var`: Baseline variance reference
- $\rho = \text{config.frozen_signal_recovery_ratio}$ (default: 0.1): Recovery threshold multiplier
- $n_c = \text{config.frozen_signal_recovery_steps}$ (default: 2): Confirmation window

Recovery is confirmed when:

$$\text{variance}_t > \rho \cdot \text{historical_var} \quad \text{for } n_c \text{ consecutive steps} \quad (2.2)$$

Implementation

```

1 # In evaluate_ingestion():
2 if frozen:
3     residual_window = np.asarray(state.residual_window, dtype=np.float64)
4     historical_variance = float(np.var(residual_window)) if residual_window.size > 1 else
      0.0
5     recent_window = residual_window[-config.frozen_signal_recovery_steps:]
6     recent_variance = float(np.var(recent_window)) if recent_window.size > 1 else 0.0
7     variance_history = [recent_variance] * config.frozen_signal_recovery_steps
8     in_recovery = detect_frozen_recovery(
9         variance_history=variance_history,
10        historical_variance=historical_variance,
11        ratio_threshold=config.frozen_signal_recovery_ratio, # V-MAJ-6: Use parameter
12        consecutive_steps=config.frozen_signal_recovery_steps
13    )
14    if in_recovery:
15        frozen = False # Lift the frozen signal flag

```

Configuration Parameters

From PredictorConfig:

Parameter	Default	Purpose
<code>frozen_signal_min_steps</code>	5	Consecutive equal values to trigger alarm
<code>frozen_signal_recovery_ratio</code>	0.1	Variance ratio threshold for recovery (10% of baseline)
<code>frozen_signal_recovery_steps</code>	2	Confirmation window for recovery

Table 2.1: V-MAJ-6 Frozen Signal Recovery Configuration

Benefits

- **Automatic Recovery:** No manual intervention needed when signal variance improves
- **Hysteresis:** Recovery threshold ($\rho = 0.1$) is more lenient than typical alarm threshold, preventing oscillation
- **Configuration-Driven:** All parameters injected from config.toml (zero-heuristics policy)
- **State Preservation:** Maintains frozen flag during low-variance periods, automatically lifts when variance returns
- **Signal Quality Supervision:** Enables secondary observability on signal quality degradation patterns

2.4 Staleness Policy (TTL)

Every observation must carry a timestamp for TTL evaluation. If the target delay exceeds Δ_{max} , the JKO update must be suspended immediately.

- Compute staleness as $\Delta_t = t_{now} - t_{obs}$.
- If $\Delta_t > \Delta_{max}$, skip the transport update.
- Preserve state and record a staleness warning event.

2.4.1 Implementation Notes

Staleness is computed as the difference between current time and observation timestamp. The ingestion decision flags a suspended JKO update when the TTL is exceeded. The orchestrator also applies a secondary staleness check using $\Delta t = t_{obs} - t_{last_update}$ (`timestamp_ns` minus `state.last_update_ns`) to trigger degraded mode on large inter-sample gaps.

Chapter 3

Telemetry Abstraction

3.1 TelemetryBuffer Emission

The JKO orchestrator should emit a `TelemetryBuffer` at the end of each step. This buffer is consumed by a dedicated process outside the JAX execution thread.

- The buffer contains summary metrics (prediction, weights, free energy, diagnostics, regime flags).
- The compute path only enqueues the buffer and continues.
- The consumer is responsible for serialization and persistence.

3.1.1 Implementation Notes

The telemetry buffer is a bounded, thread-safe queue. Buffer capacity is explicitly injected from `PredictorConfig.telemetry_buffer_capacity` to eliminate implicit defaults (zero-heuristics policy). Parity hashes are emitted on a configurable interval and derived from canonical float64 serialization.

```
1 # Instantiation pattern (capacity injected from config)
2 buffer = TelemetryBuffer(capacity=config.telemetry_buffer_capacity)
```

3.2 No Compute Stalls

JAX compute threads must never block on I/O. Telemetry buffers must be non-blocking and consumed by a separate process or thread outside the JAX execution path.

Chapter 4

Deterministic Logging

4.1 Hash-Based Parity Checks

For hardware parity audits, the logger records SHA-256 hashes of the weight vector ρ and the OT cost at configurable intervals. This permits CPU/GPU parity validation without dumping VRAM data.

- Hash interval configured per deployment.
- Hashes derived from canonical float64 serialization.
- Logs are append-only and immutable.

4.2 Audit Hashes

Parity audits must log SHA-256 hashes of ρ and OT cost at configured intervals. Hash input must be derived from canonical float64 serialization to ensure reproducibility across CPU and GPU.

Chapter 5

Snapshot Strategy

5.1 Atomic Persistence

Snapshots must be persisted atomically to prevent partial writes. The IO layer is responsible for:

- Writing to temporary files and renaming atomically.
- Optional compression configured by policy.
- Coordinating snapshot cadence with telemetry output.

5.2 Binary Serialization

Text formats (JSON, XML) are prohibited for critical snapshots due to latency and ambiguity. The configuration allows `snapshot_format` values `msgpack` or `protobuf`, but the current IO implementation supports MessagePack only.

- Encode all fields deterministically.
- Preserve float64 for numerical fidelity.

5.2.1 Implementation Notes

The snapshot serializer currently supports MessagePack only (`snapshot_format = "msgpack"`); `protobuf` is reserved for a future implementation. Hash verification is performed before state injection.

5.3 Integrity Verification

Each snapshot Σ_t must include a hash footer (SHA-256 or CRC32c). The load routine must verify the hash before injecting state into memory.

- Fail closed if hash verification fails.
- Log integrity failures at critical severity.

5.4 Atomic Write Protocol

To avoid partial writes, persist snapshots to a temporary file and then atomically rename to the target path. The rename step must be the only visible operation to consumers.

- Use a unique temporary filename per snapshot.
- Ensure the target file is replaced atomically.

5.4.1 Implementation Notes

Snapshots are written to a unique temporary file and moved into place using atomic rename. Optional fsync ensures persistence across power loss.

Chapter 6

Security Policies

6.1 Credential Injection

Tokens and API keys must not appear in source code. Credentials must be injected at runtime via environment variables or `.env` files.

6.1.1 Implementation Notes

Credential helpers read from environment variables or `.env` files and raise explicit errors on missing values.

6.2 Version Control Exclusion

The repository must exclude `.env` files and credential directories via `.gitignore`. Secrets must never be committed.

Chapter 7

Orchestrator Integration

7.1 Ingestion Gate in orchestrate_step()

The core orchestration pipeline now integrates ingestion validation as a pre-kernel gate. The `orchestrate_step()` function signature is extended to accept observation metadata:

```
1 def orchestrate_step(  
2     signal: Float[Array, "n"],  
3     timestamp_ns: int,  
4     state: InternalState,  
5     config: PredictorConfig,  
6     observation: ProcessState,  
7     now_ns: int,  
8     telemetry_buffer: Optional[TelemetryBuffer] = None,  
9     step_counter: int = 0,  
10    mutation_rate_limiter: Optional[MutationRateLimiter] = None,  
11    degradation_monitor: Optional[DegradationMonitor] = None,  
12    allow_host_scaling: bool = True,  
13 ) -> OrchestrationResult:  
14     """Run a single orchestration step with IO ingestion validation."""
```

Note: `OrchestrationResult` now carries an optional `config` field to propagate host-side DGM scaling decisions across steps.

7.1.1 Execution Flow

The ingestion gate operates as follows:

1. **Input Validation:** Standard signal length and dtype checks.
2. **Ingestion Decision:** Call `evaluate_ingestion()` with current state, observation, and configuration.
3. **Rejection Logic:** If `accept_observation == False`, reject the entire observation without state update (emergency mode).
4. **Degradation Flags:** Apply `suspend_jko_update` and `freeze_kernel_d` flags to control fusion behavior.
5. **Kernel Execution:** Run kernels A-D; if `freeze_kernel_d == True`, mark kernel D output as frozen.
6. **Host Scaling (optional):** If `allow_host_scaling == True`, the host may re-run Kernel B with a scaled DGM architecture and return the updated config for the next step.
7. **Fusion Selection:** Skip JKO/Sinkhorn if degraded mode or `suspend_jko_update` is set.

8. **State Update:** Only update `InternalState` if observation is accepted.

7.1.2 Flag Semantics

The `IngestionDecision` object carries the following flags:

- **accept_observation:** If False, reject and preserve inertial state.
- **suspend_jko_update:** If True, freeze weights and skip Sinkhorn.
- **degraded_mode:** If True, emit degraded inference mode prediction.
- **freeze_kernel_d:** If True, mark kernel D output as frozen (no weight update).
- **staleness_ns:** Staleness in nanoseconds for audit logging.
- **events:** Emitted validation events (outliers, frozen signals, staleness alarms).

7.1.3 Early Return on Rejection

If an observation is rejected (catastrophic outlier), the orchestrator returns a degraded result without advancing the state:

```
1 # If observation is rejected, skip state update entirely
2 if reject_observation:
3     updated_state = state
4 else:
5     updated_state = atomic_state_update(...)
```

7.2 PRNG Configuration

To eliminate magic numbers in PRNG splitting, the split count is configuration-driven via `core.prng_split_count`.

```
1 [core]
2 prng_seed = 42
3 prng_split_count = 4
```

The orchestrator uses `config.prng_split_count` when advancing the RNG key. All PRNG parameters reside in configuration.

7.3 64-bit Precision Enforcement

To ensure Malliavin calculus and Signature computation stability, 64-bit precision must be activated at module import time, before any XLA tracing:

```
1 # api/config.py: JAX CONFIGURATION (at module level)
2 import jax
3 jax.config.update("jax_enable_x64", True)
```

This enforces bit-exact reproducibility across CPU/GPU/FPGA backends and must execute before `ConfigManager` initialization.

Chapter 8

Telemetry Buffer Integration (P2.3)

8.1 Motivation

Phase 2 implementations (P2.1 WTMM, P2.2 SDE stiffness, V-MAJ violations) generate rich diagnostic data during orchestration. To enable post-mortem analysis, compliance audits, and debugging without stalling inference, P2.3 integrates a **non-blocking telemetry buffer** into the orchestration pipeline.

Key requirements:

- **Non-Blocking:** Logging never blocks compute threads (async enqueue only)
- **Thread-Safe:** Multiple consumers can safely drain buffer
- **Audit Trail:** Records capture complete prediction state snapshot
- **Integrity:** Parity hashes verify weights and free energy
- **Config-Driven:** Emission interval and buffer capacity injected from config

8.2 Data Model

Each telemetry record captures:

```
1 @dataclass(frozen=True)
2 class TelemetryRecord:
3     step: int                # Monotonic counter
4     payload: dict            # Rich diagnostic data
5
6
7 # Payload structure (P2.3):
8 payload = {
9     "step": 42,
10    "timestamp_ns": 1708308000000000000, # Nanosecond precision
11    "prediction_ref": fused_prediction,    # DeviceArray reference
12    "weights_ref": final_rho,              # DeviceArray reference
13    "free_energy_ref": free_energy,        # DeviceArray reference
14    "kurtosis_ref": updated_state.kurtosis,
15    "holder_exponent_ref": updated_state.holder_exponent,
16    "dgm_entropy_ref": updated_state.dgm_entropy,
17    "mode_collapse_warning": False,        # V-MAJ-5 flag
18    "degraded_mode": False,                # V-MAJ-7 hysteresis
19    "emergency_mode": False                # Circuit breaker
20 }
```

Parity hashes are computed during `materialize_telemetry_batch()` after `jax.device_get()` to avoid host-device synchronization in the orchestrator path.

8.3 Integration into orchestrate_step()

The orchestrator now accepts optional `telemetry_buffer` and `step_counter` parameters:

```
1 def orchestrate_step(  
2     signal: Float[Array, "n"],  
3     timestamp_ns: int,  
4     state: InternalState,  
5     config: PredictorConfig,  
6     observation: ProcessState,  
7     now_ns: int,  
8     telemetry_buffer: Optional[TelemetryBuffer] = None, # P2.3  
9     step_counter: int = 0, # P2.3  
10    mutation_rate_limiter: Optional[MutationRateLimiter] = None,  
11    degradation_monitor: Optional[DegradationMonitor] = None,  
12) -> OrchestrationResult:  
13    """  
14    Run a single orchestration step with telemetry buffering.  
15  
16    If telemetry_buffer is None, skips telemetry (backward compatible).  
17    If provided, enqueues record when hash_interval triggers.  
18    """  
19    # ... existing orchestration logic ...  
20  
21    # P2.3: Telemetry Buffer Integration (before return)  
22    if telemetry_buffer is not None:  
23        telemetry_payload = {  
24            "step": step_counter,  
25            "timestamp_ns": timestamp_ns,  
26            "prediction_ref": fused_prediction,  
27            "weights_ref": final_rho,  
28            "free_energy_ref": free_energy if fusion is not None else jnp.array(0.0),  
29            "kurtosis_ref": updated_state.kurtosis,  
30            "holder_exponent_ref": updated_state.holder_exponent,  
31            "dgm_entropy_ref": updated_state.dgm_entropy,  
32            "mode_collapse_warning": mode_collapse_warning,  
33            "degraded_mode": degraded_mode,  
34            "emergency_mode": emergency_mode,  
35        }  
36  
37        # Emit only when hash_interval triggers (config-driven)  
38        if should_emit_hash(step_counter, config.telemetry_hash_interval_steps):  
39            telemetry_record = TelemetryRecord(step=step_counter, payload=  
40            telemetry_payload)  
41            telemetry_buffer.enqueue(telemetry_record)  
42  
43    return OrchestrationResult(...)
```

8.4 Configuration Parameters

Parameter	Default	Purpose
<code>telemetry_buffer_capacity</code>	1024	Maximum records in ring buffer
<code>telemetry_hash_interval_steps</code>	1	Emit telemetry every N steps

Table 8.1: P2.3 Telemetry Configuration

8.4.1 Configuration Injection Example


```

1 # config.toml
2 [io]
3 # Telemetry
4 telemetry_hash_interval_steps = 1          # Emit every step (or 10 for sparser logging)
5 telemetry_buffer_capacity = 1024          # Ring buffer size (zero-heuristics injection)

```

8.5 Thread-Safety Model

TelemetryBuffer uses `threading.Lock` for atomic operations:

- **enqueue()**: Acquires lock, appends record, releases ($O(1)$ amortized)
- **drain()**: Acquires lock, extracts all records, clears buffer, releases
- **size()**: Acquires lock, returns current count, releases

This prevents race conditions when orchestrator (compute thread) enqueues while consumer thread drains.

8.6 Backward Compatibility

P2.3 is fully backward compatible:

- If `telemetry_buffer=None`, no telemetry is emitted (default)
- Existing calls to `orchestrate_step()` without telemetry params continue working
- No changes to compute path (telemetry entirely outside `@jax.jit` scope)

8.7 Usage Example

```

1 from Python.io.telemetry import TelemetryBuffer
2 from Python.api.config import PredictorConfigInjector
3
4 # Initialize
5 config = PredictorConfigInjector().create_config()
6 telemetry_buffer = TelemetryBuffer(capacity=config.telemetry_buffer_capacity)
7
8 # In prediction loop:
9 for step in range(num_steps):
10     result = orchestrate_step(
11         signal=current_signal,
12         timestamp_ns=now_ns(),
13         state=state,
14         config=config,
15         observation=obs,
16         now_ns=now_ns(),
17         telemetry_buffer=telemetry_buffer,      # P2.3: Pass buffer
18         step_counter=step,                      # P2.3: Pass step
19     )
20
21 # Optional: Drain telemetry in background thread
22 if step % 100 == 0:
23     records = telemetry_buffer.drain()
24     # Write to file/database (non-blocking, doesn't stall orchestrator)
25     write_telemetry_async(records)

```

8.8 Benefits

- **Non-Blocking:** Telemetry enqueue is $O(1)$, never stalls inference
- **Audit Trail:** Complete state snapshots for compliance and debugging
- **Integrity:** Parity hashes enable CPU/GPU parity validation
- **Configurable:** Emission interval and buffer size injected from config
- **Thread-Safe:** Lock-based synchronization for multi-threaded consumers
- **Backward Compatible:** Fully optional, no impact on existing code paths

8.9 Visualization Dashboard (GAP-6)

The IO layer includes a static HTML dashboard generator for telemetry snapshots. It requires no external dependencies and produces a standalone report suitable for offline reviews and audits.

8.9.1 Usage

```
1 from Python.io.telemetry import TelemetryBuffer
2 from Python.io.dashboard import export_dashboard_snapshot
3
4 telemetry_buffer = TelemetryBuffer(capacity=1024)
5 # ... enqueue telemetry records during inference ...
6
7 exported = export_dashboard_snapshot(
8     telemetry_buffer,
9     "io/reports/telemetry_dashboard.html",
10 )
11 print(f"Exported {exported} records")
```

8.9.2 Notes

- The dashboard drains the buffer, materializes device arrays, and writes HTML.
- Charts are rendered as inline SVG with no external assets.
- Output is deterministic and suitable for audit trails.

Chapter 9

Level 4 Autonomy: Configuration Mutation Safety

9.1 Overview

Phase 2.1.0 introduces **Level 4 Autonomy** compliance for autonomous configuration mutations during meta-optimization. This chapter documents the implementation of V-CRIT-2, V-MAJ-4, V-MAJ-5, and V-MAJ-7 violations identified during the specification compliance audit (AUDIT_SPEC_COMPLIANCE_2026-02-19.md).

Specification References:

- `Stochastic_Predictor_I0.tex` §3.3 - Configuration Mutation Protocol (Atomic Write)
- `Stochastic_Predictor_I0.tex` §3.3.6 - Rate Limiting and Safety Guardrails
- `Stochastic_Predictor_Implementation.tex` §5.4.3 - Degradation Detection Protocol
- `Stochastic_Predictor_Theory.tex` §2.3.6 - Monitoring and Telemetry for adaptive SDE schemes

Implementation Scope:

- **V-CRIT-2:** Atomic TOML mutation protocol with locked subsection protection
- V-MAJ-4: Configuration mutation rate limiting
- V-MAJ-5: Degradation detection with automatic rollback
- V-MAJ-7: Adaptive telemetry monitoring (partial - safety guardrails)

9.2 V-CRIT-2: Atomic Configuration Mutation Protocol

9.2.1 Problem Statement

Violation: No atomic write mechanism for `config.toml` mutations during autonomous meta-optimization. Race conditions, partial writes, and locked parameter violations could corrupt system configuration without rollback capability.

Impact:

- Partial file corruption during concurrent writes
- No protection for immutable subsections (e.g., `float_precision`, `snapshot_format`)
- No audit trail for forensic analysis
- Cannot rollback to previous working configuration
- Loss of checkpoint resumability if meta-optimization paths corrupted

9.2.2 Requirements

IO.tex §3.3.2 - Atomic TOML Update Algorithm:

1. **Phase 1 - Validation:** Check parameter ranges, types, locked subsections
2. **Phase 2 - Backup:** Create timestamped backup + latest .bak
3. **Phase 3 - Atomic Write:** Write to .tmp, fsync(), os.replace()
4. **Phase 4 - Audit Log:** Record mutation to io/mutations.log
5. **Phase 5 - Success:** Return or raise on failure

IO.tex §3.3.4 - Locked Subsections (Asimov's Zeroth Law):

- [meta]: schema_version
- [core]: jax_platforms, jax_default_dtype, float_precision, staleness_ttl_ns
- [io]: snapshot_format, snapshot_hash_algorithm, snapshot_compression, telemetry_hash_interval_steps
- [meta_optimization]: n_trials, n_startup_trials, multivariate, train_ratio, n_folds

9.2.3 Implementation

Module: stochastic_predictor/io/config_mutation.py (EXTENDED)

Functions Added:

- atomic_write_config(): 5-phase POSIX-compliant atomic mutation
- validate_config_mutation(): Schema validation + locked subsection checks
- append_audit_log(): JSON Lines audit trail logging
- create_config_backup(): Timestamped backup creation

Constants Added:

- LOCKED_SUBSECTIONS: Immutable parameter dictionary
- VALIDATION_SCHEMA: Type/range constraints for mutable parameters

Atomic Write Protocol Implementation

```
1 def atomic_write_config(  
2     config_path: Path,  
3     new_params: Dict[str, Any],  
4     trigger: str = "ManualMutation",  
5     best_objective: Optional[float] = None,  
6     audit_log_path: Optional[Path] = None,  
7 ) -> None:  
8     """  
9     Atomically mutate configuration with POSIX-compliant write protocol.  
10  
11     COMPLIANCE: IO.tex §3.3.2 - Atomic TOML Update Algorithm  
12     """  
13     audit_log_path = audit_log_path or Path("io/mutations.log")  
14  
15     # Phase 1: Validation  
16     current_config = toml.load(config_path)
```

```

17 merged_config = validate_config_mutation(current_config, new_params)
18
19 # Compute delta for audit log
20 delta = {
21     param_key: (_get_nested_param(current_config, param_key), new_value)
22     for param_key, new_value in new_params.items()
23 }
24
25 # Phase 2: Immutable Backup
26 timestamp = datetime.now(timezone.utc).strftime("%Y-%m-%dT%H:%M:%SZ")
27 backup_path = config_path.with_suffix(f".bak.{timestamp}")
28 latest_backup_path = config_path.with_suffix(".bak")
29
30 shutil.copy2(config_path, backup_path)
31 shutil.copy2(config_path, latest_backup_path)
32
33 # Phase 3: Atomic Write via Temporary File
34 tmp_path = config_path.with_suffix(".tmp")
35
36 # O_EXCL flag detects concurrent mutation
37 fd = os.open(tmp_path, os.O_WRONLY | os.O_CREAT | os.O_EXCL, 0o644)
38
39 try:
40     toml_bytes = toml.dumps(merged_config).encode("utf-8")
41     os.write(fd, toml_bytes)
42
43     # CRITICAL: fsync() ensures kernel buffer flush to disk
44     os.fsync(fd)
45 finally:
46     os.close(fd)
47
48 # Phase 4: Atomic Replacement (POSIX os.replace)
49 os.replace(tmp_path, config_path)
50
51 # Phase 5: Audit Logging
52 append_audit_log(audit_log_path, {
53     "timestamp": datetime.now(timezone.utc).isoformat(),
54     "event": "MUTATION_SUCCESS",
55     "trigger": trigger,
56     "delta": delta,
57     "best_objective": best_objective,
58     "backup": str(backup_path),
59     "status": "SUCCESS",
60 })

```

Validation Schema Implementation

```

1 # Locked subsections (immutable to prevent self-corruption)
2 LOCKED_SUBSECTIONS = {
3     "meta": ["schema_version"],
4     "core": ["jax_platforms", "jax_default_dtype", "float_precision", "staleness_ttl_ns"],
5     "io": [
6         "snapshot_format",
7         "snapshot_hash_algorithm",
8         "snapshot_compression",
9         "telemetry_hash_interval_steps",
10    ],
11     "meta_optimization": ["n_trials", "n_startup_trials", "multivariate", "train_ratio", "n_folds"],
12 }
13

```

```

14 # Validation schema for mutable parameters
15 VALIDATION_SCHEMA = {
16     "orchestration.cusum_k": {"type": float, "range": (0.1, 1.0)},
17     "orchestration.cusum_h": {"type": float, "range": (2.0, 10.0)},
18     "orchestration.grace_period_steps": {"type": int, "range": (5, 100)},
19     "orchestration.volatility_alpha": {"type": float, "range": (0.05, 0.3)},
20     "orchestration.learning_rate": {"type": float, "range": (1e-5, 1e-1)},
21     "orchestration.entropy_window": {"type": int, "range": (10, 500)},
22     "orchestration.entropy_threshold": {"type": float, "range": (0.5, 0.95)},
23     "orchestration.holder_threshold": {"type": float, "range": (0.2, 0.65)},
24     "orchestration.sinkhorn_alpha": {"type": float, "range": (0.1, 1.0)},
25     "orchestration.sinkhorn_epsilon_min": {"type": float, "range": (0.001, 0.1)},
26     "orchestration.sinkhorn_epsilon_0": {"type": float, "range": (0.05, 0.5)},
27     "orchestration.sinkhorn_max_iter": {"type": int, "range": (50, 500)},
28     "kernels.log_sig_depth": {"type": int, "range": (2, 5)},
29     "kernels.wtmm_buffer_size": {"type": int, "range": (64, 512),
30                                 "constraint": "power_of_2"},
31     "kernels.besov_cone_c": {"type": float, "range": (1.0, 3.0)},
32     "kernels.dgm_width_size": {"type": int, "range": (32, 256),
33                                "constraint": "power_of_2"},
34     "kernels.stiffness_low": {"type": float, "range": (50.0, 500.0)},
35     "kernels.stiffness_high": {"type": float, "range": (500.0, 5000.0)},
36     "kernels.sde_dt": {"type": float, "range": (0.001, 0.1)},
37     "kernels.sde_numel_integrations": {"type": int, "range": (50, 200)},
38     "kernels.sde_diffusion_sigma": {"type": float, "range": (0.05, 0.5)},
39     "kernels.kernel_ridge_lambda": {"type": float, "range": (1e-8, 1e-3)},
40     # ... additional mutable parameters
41 }
42
43 def validate_config_mutation(
44     current_config: Dict[str, Any],
45     new_params: Dict[str, Any],
46 ) -> Dict[str, Any]:
47     """
48     Validate configuration mutation against schema and locked subsections.
49
50     Raises:
51         ConfigMutationError: If locked parameter mutation attempted
52         ConfigMutationError: If parameter out of safe range
53     """
54     # Check locked subsection violations
55     for param_key in new_params.keys():
56         subsection = param_key.split(".")[0]
57         param_name = ".".join(param_key.split(".")[1:])
58
59         if subsection in LOCKED_SUBSECTIONS:
60             if param_name in LOCKED_SUBSECTIONS[subsection]:
61                 raise ConfigMutationError(
62                     f"Parameter '{param_key}' is LOCKED (immutable)"
63                 )
64
65     # Validate against schema (ranges, types, constraints)
66     merged_config = dict(current_config)
67     for param_key, new_value in new_params.items():
68         rules = VALIDATION_SCHEMA[param_key]
69
70         # Type check
71         if not isinstance(new_value, rules["type"]):
72             raise ConfigMutationError("Type mismatch")
73
74         # Range check
75         min_val, max_val = rules["range"]
76         if not (min_val <= new_value <= max_val):

```

```

77         raise ConfigMutationError("Out of safe range")
78
79     # Apply constraints (e.g., power_of_2)
80     if "constraint" in rules:
81         # Check constraint...
82
83     _set_nested_param(merged_config, param_key, new_value)
84
85     # Cross-parameter constraints (e.g., stiffness_low < stiffness_high)
86     # ...
87
88     return merged_config

```

9.2.4 Usage Example

```

1  from Python.io import atomic_write_config
2
3  # After Deep Tuning completes
4  best_params = {
5      "orchestration.cusum_k": 0.72,
6      "kernels.dgm_width_size": 256,
7      "kernels.stiffness_low": 143.0,
8  }
9
10 # Atomic mutation with audit trail
11 atomic_write_config(
12     Path("config.toml"),
13     best_params,
14     trigger="DeepTuning_Iteration_500",
15     best_objective=0.0234 # MAPE
16 )
17
18 # Creates:
19 # - config.toml.bak.2026-02-19T14:32:05Z (timestamped backup)
20 # - config.toml.bak (latest backup)
21 # - io/mutations.log (audit entry)

```

9.2.5 Files Modified

- stochastic_predictor/io/config_mutation.py: +280 LOC
- stochastic_predictor/io/__init__.py: +7 exports

9.2.6 Compliance Impact

V-CRIT-2 Resolution: Atomic TOML mutation protocol fully implemented with:

- POSIX-compliant atomic write (fsync + os.replace)
- Locked subsection protection (Asimov's Zeroth Law)
- Validation schema enforcement (20+ mutable parameters)
- Timestamped backups for manual rollback
- JSON Lines audit trail for forensic analysis

Level 4 Autonomy Enablement: System can now autonomously mutate config.toml during Deep Tuning campaigns without human intervention, while maintaining safety guarantees.

9.3 V-MAJ-4: Configuration Mutation Rate Limiting

9.3.1 Problem Statement

Violation: No safety guardrails enforced for autonomous configuration mutations during meta-optimization. Optimizer could thrash between configurations, mutate too frequently, or make excessively large parameter jumps.

Impact: System instability, pathological optimizer behavior, configuration thrashing without convergence.

9.3.2 Safety Requirements

IO.tex §3.3.6 Requirements:

- Maximum mutation rate: 10 mutations/hour
- Minimum stability period: 1,000 prediction steps between mutations
- Delta magnitude limit: 50% relative change per parameter
- Audit trail: JSON Lines log of all mutation events

9.3.3 Implementation

Module: stochastic_predictor/io/config_mutation.py (NEW)

Class: MutationRateLimiter

```
1 @dataclass
2 class MutationRateLimiter:
3     """Enforce safety guardrails for autonomous configuration mutations.
4
5     Prevents optimizer pathologies:
6         - Thrashing between configurations
7         - Excessive mutation frequency
8         - Large parameter jumps
9         - Pathological degradation without rollback
10
11     Args:
12         max_mutations_per_hour: Maximum allowed mutations/hour (default 10)
13         stability_steps_required: Minimum steps before next mutation (default 1000)
14         max_relative_change: Maximum parameter change per mutation (default 0.5)
15     """
16
17     max_mutations_per_hour: int = 10
18     stability_steps_required: int = 1000
19     max_relative_change: float = 0.5
20
21     _mutation_history: List[Tuple[float, Dict]] = field(default_factory=list)
22     _last_mutation_timestamp: Optional[float] = None
23     _current_steps_since_mutation: int = 0
24
25     def can_mutate(self) -> Tuple[bool, str]:
26         """Check if mutation is allowed under safety guardrails.
27
28         Returns:
29             (allowed: bool, reason: str)
30         """
31         # Check maximum mutation rate (sliding 1-hour window)
32         one_hour_ago = time.time() - 3600
33         recent_mutations = [
34             ts for ts, _ in self._mutation_history if ts > one_hour_ago
```



```

35     ]
36     if len(recent_mutations) >= self.max_mutations_per_hour:
37         return False, f"Rate limit: {len(recent_mutations)}/10 mutations"
38
39     # Check minimum stability period
40     if self._current_steps_since_mutation < self.stability_steps_required:
41         return False, f"Stability: {self._current_steps_since_mutation}/1000"
42
43     return True, "OK"
44
45     def validate_delta(
46         self,
47         delta: Dict[str, Tuple[float, float]],
48         max_relative_change: Optional[float] = None
49     ) -> Tuple[bool, str]:
50         """Validate parameter delta magnitude.
51
52         Args:
53             delta: {param: (old_value, new_value)}
54             max_relative_change: Override default max change
55
56         Returns:
57             (valid: bool, reason: str)
58         """
59         max_change = max_relative_change or self.max_relative_change
60
61         for param, (old_val, new_val) in delta.items():
62             if abs(old_val) < 1e-12:
63                 continue # Skip zero division
64
65             relative_change = abs((new_val - old_val) / old_val)
66
67             if relative_change > max_change:
68                 return False, (
69                     f"{param} change too large: {relative_change:.1%} > 50%"
70                 )
71
72         return True, "OK"
73
74     def record_mutation(self, delta: Dict) -> None:
75         """Record successful mutation."""
76         now = time.time()
77         self._mutation_history.append((now, delta))
78         self._current_steps_since_mutation = 0
79
80     def increment_stability_counter(self) -> None:
81         """Call after each prediction step."""
82         self._current_steps_since_mutation += 1

```

9.3.4 Usage Pattern

```

1 # In meta-optimizer loop
2 limiter = MutationRateLimiter(max_mutations_per_hour=10)
3
4 # Export best params using atomic mutation protocol
5 optimizer.export_best_params_to_config(
6     "config.toml",
7     trigger="DeepTuning_Iteration_500",
8     rate_limiter=limiter,
9 )
10
11 # In main prediction loop

```

```

12 for step in range(1000):
13     prediction = orchestrate_step(..., mutation_rate_limiter=limiter)
14     # If not passing mutation_rate_limiter, call limiter.increment_stability_counter()
    manually

```

9.4 V-MAJ-5: Degradation Detection Auto-Rollback

9.4.1 Problem Statement

Violation: No automatic rollback mechanism when post-mutation performance degrades beyond acceptable threshold. Pathological mutations persist indefinitely, requiring manual operator intervention.

Impact: System can enter degraded state without automatic recovery, violating Level 4 autonomy requirements.

9.4.2 Safety Requirements

IO.tex §3.3.6 Requirements:

- Monitor RMSE over N=100 predictions post-mutation
- Compare to pre-mutation baseline RMSE
- If relative increase > 30%, trigger automatic rollback
- Restore config.toml from config.toml.bak
- Log rollback event to audit trail

9.4.3 Implementation

Module: stochastic_predictor/io/config_mutation.py

Class: DegradationMonitor

```

1 @dataclass
2 class DegradationMonitor:
3     """Monitor post-mutation performance and trigger rollback on degradation.
4
5     Implements closed-loop safety mechanism:
6         1. Record pre-mutation baseline RMSE
7         2. Monitor post-mutation predictions (N=100 sample window)
8         3. Compute post-mutation RMSE
9         4. If relative increase > threshold (default 30%), auto-rollback
10
11     Args:
12         degradation_threshold: Max allowed RMSE increase (default 0.3 = 30%)
13         monitoring_window: Predictions to sample post-mutation (default 100)
14         config_path: Path to config.toml (default "config.toml")
15         backup_path: Path to backup config (default "config.toml.bak")
16         audit_log_path: Path to mutation audit log
17     """
18
19     degradation_threshold: float = 0.3
20     monitoring_window: int = 100
21     config_path: Path = Path("config.toml")
22     backup_path: Path = Path("config.toml.bak")
23     audit_log_path: Path = Path("io/mutations.log")
24
25     _pre_mutation_rmse: Optional[float] = None

```

```

26 _post_mutation_errors: List[float] = field(default_factory=list)
27 _monitoring_active: bool = False
28
29 def start_monitoring(self, baseline_rmse: float) -> None:
30     """Record pre-mutation baseline and start monitoring."""
31     self._pre_mutation_rmse = baseline_rmse
32     self._post_mutation_errors = []
33     self._monitoring_active = True
34
35 def record_prediction_error(self, error: float) -> None:
36     """Accumulate post-mutation prediction error."""
37     if not self._monitoring_active:
38         return
39     self._post_mutation_errors.append(error)
40
41 def check_degradation(self) -> Tuple[bool, float]:
42     """Check if post-mutation performance degraded beyond threshold.
43
44     Returns:
45         (degraded: bool, relative_increase: float)
46     """
47     if not self._monitoring_active:
48         return False, 0.0
49
50     # Require full monitoring window
51     if len(self._post_mutation_errors) < self.monitoring_window:
52         return False, 0.0
53
54     # Compute post-mutation RMSE
55     post_mutation_rmse = np.sqrt(
56         np.mean(np.square(self._post_mutation_errors))
57     )
58
59     # Compute relative increase
60     if self._pre_mutation_rmse is None or self._pre_mutation_rmse == 0:
61         relative_increase = 0.0
62     else:
63         relative_increase = (
64             (post_mutation_rmse - self._pre_mutation_rmse) /
65             self._pre_mutation_rmse
66         )
67
68     degraded = relative_increase > self.degradation_threshold
69
70     # Stop monitoring after check
71     if degraded or len(self._post_mutation_errors) >= self.monitoring_window:
72         self._monitoring_active = False
73
74     return degraded, relative_increase
75
76 def trigger_rollback(self) -> None:
77     """Execute automatic rollback to pre-mutation configuration.
78
79     CRITICAL: Overwrites config.toml with backup.
80     """
81     if not self.backup_path.exists():
82         raise FileNotFoundError(f"Backup config not found: {self.backup_path}")
83
84     # Restore config from backup
85     shutil.copy2(self.backup_path, self.config_path)
86
87     # Compute post-mutation RMSE for logging
88     post_mutation_rmse = float(np.sqrt(

```

```

89         np.mean(np.square(self._post_mutation_errors))
90     )) if self._post_mutation_errors else 0.0
91
92     # Append rollback event to audit log (JSON Lines format)
93     audit_entry = {
94         'timestamp': datetime.now(timezone.utc).isoformat(),
95         'event': 'AUTO_ROLLBACK',
96         'reason': 'Performance degradation detected',
97         'pre_mutation_rmse': self._pre_mutation_rmse,
98         'post_mutation_rmse': post_mutation_rmse,
99         'relative_increase': (
100             (post_mutation_rmse - self._pre_mutation_rmse) /
101             self._pre_mutation_rmse
102         ),
103         'degradation_threshold': self.degradation_threshold,
104         'monitoring_window': self.monitoring_window,
105         'status': 'ROLLBACK_SUCCESS'
106     }
107
108     # Append to audit log
109     self.audit_log_path.parent.mkdir(parents=True, exist_ok=True)
110     with open(self.audit_log_path, 'a') as f:
111         f.write(json.dumps(audit_entry) + '\n')
112
113     # Reset monitoring state
114     self._monitoring_active = False
115     self._post_mutation_errors = []

```

9.4.4 Usage Pattern

```

1 # Before mutation
2 monitor = DegradationMonitor(degradation_threshold=0.3)
3
4 # Compute baseline RMSE over recent predictions
5 recent_errors = [0.04, 0.05, 0.06, 0.05, 0.04]
6 baseline_rmse = np.sqrt(np.mean(np.square(recent_errors)))
7
8 # Apply mutation
9 delta = {"orchestration.learning_rate": 0.015}
10 atomic_write_config(Path("config.toml"), delta, trigger="AutoTuning")
11
12 # Start monitoring
13 monitor.start_monitoring(baseline_rmse)
14
15 # In prediction loop (next 100 predictions)
16 for i in range(100):
17     prediction = orchestrate_step(..., degradation_monitor=monitor)
18     actual_next = load_actual_observation()
19     error = abs(prediction.predicted_next - actual_next)
20
21     monitor.record_prediction_error(error)
22
23     # Check for degradation
24     degraded, increase = monitor.check_degradation()
25     if degraded:
26         logger.critical(f"Degradation detected: RMSE +{increase:.1%}")
27         monitor.trigger_rollback()
28         logger.info("Config rolled back to pre-mutation state")
29         break

```

9.4.5 Audit Log Format

Mutation events are logged in JSON Lines format to `io/mutations.log`:

```
1 {"timestamp": "2026-02-19T14:30:00Z", "event": "MUTATION_SUCCESS",
2   "delta": {"orchestration.learning_rate": [0.01, 0.015]}, "status": "SUCCESS"}
3 {"timestamp": "2026-02-19T14:35:00Z", "event": "AUTO_ROLLBACK",
4   "reason": "Performance degradation detected",
5   "pre_mutation_rmse": 0.05, "post_mutation_rmse": 0.07,
6   "relative_increase": 0.40, "degradation_threshold": 0.30,
7   "status": "ROLLBACK_SUCCESS"}
```

9.5 V-MAJ-7: Adaptive Telemetry Monitoring

9.5.1 Implementation Status

V-MAJ-7 requires comprehensive telemetry for adaptive architecture and solver selection metrics. This implementation extends `InternalState` with counters for Level 4 autonomy monitoring and provides full adaptive telemetry collection.

9.5.2 InternalState Extensions

Module: `stochastic_predictor/api/types.py`

Added fields to `InternalState` (lines 443-447):

```
1 # V-MAJ-7: Level 4 Autonomy Adaptive Telemetry
2 baseline_entropy: Float[Array, "1"]      # H_baseline: Reference entropy for
3 solver_explicit_count: int                # N_explicit: Explicit SDE solver steps
4 solver_implicit_count: int               # N_implicit: Implicit SDE solver steps
5 architecture_scaling_events: int         # N_scale: DGM architecture scaling events
```

Purpose:

- `baseline_entropy`: Reference entropy H_{baseline} for computing entropy ratio $\kappa = H_{\text{current}}/H_{\text{baseline}}$
- `solver_explicit_count`, `solver_implicit_count`: Track SDE solver scheme frequencies within monitoring window (default 100 steps)
- `architecture_scaling_events`: Count DGM architecture scaling events (cumulative)

9.5.3 Adaptive Telemetry Collection

Module: `stochastic_predictor/io/telemetry.py`

Function: `collect_adaptive_telemetry()`

```
1 def collect_adaptive_telemetry(
2     state: Any, # InternalState
3     config: Any, # PredictorConfig
4     window_size: int = 100
5 ) -> Optional[AdaptiveTelemetry]:
6     """
7     Collect telemetry for adaptive architecture/solver diagnostics.
8
9     COMPLIANCE: V-MAJ-7 - Adaptive Telemetry Monitoring
10
11     Args:
12         state: Current InternalState (contains counters, entropy, etc.)
13         config: Current PredictorConfig (may have been mutated)
14         window_size: Monitoring window for frequency calculations
15 """
```

```

16 Returns:
17     AdaptiveTelemetry instance or None if insufficient data
18 """
19 import jax.numpy as jnp
20
21 # Compute solver frequencies (clipped to [0,1])
22 total_solver_steps = state.solver_explicit_count + state.solver_implicit_count
23 if total_solver_steps == 0:
24     return None # Insufficient data
25
26 freq_explicit = float(state.solver_explicit_count) / total_solver_steps
27 freq_implicit = float(state.solver_implicit_count) / total_solver_steps
28
29 # Compute entropy ratio = H_current / H_baseline
30 baseline_entropy_val = float(state.baseline_entropy)
31 if baseline_entropy_val <= 0.0:
32     entropy_ratio = 1.0 # Avoid division by zero
33 else:
34     entropy_ratio = float(state.dgm_entropy) / baseline_entropy_val
35
36 # Extract DGM architecture from config
37 dgm_width = config.dgm_width_size
38 dgm_depth = config.dgm_depth
39
40 # Extract JKO flow parameters from config
41 entropy_window = config.entropy_window
42 learning_rate = config.learning_rate
43 volatility_sigma_squared = float(state.ema_variance)
44
45 # Extract adaptive stiffness thresholds from config
46 stiffness_low = config.stiffness_low
47 stiffness_high = config.stiffness_high
48 holder_exponent_val = float(state.holder_exponent)
49
50 return AdaptiveTelemetry(
51     # SDE Solver Frequency
52     scheme_frequency_explicit=freq_explicit,
53     scheme_frequency_implicit=freq_implicit,
54     max_stiffness_metric=0.0, # Placeholder - requires Kernel C integration
55     num_internal_iterations_mean=0.0, # Placeholder
56     implicit_residual_norm_max=0.0, # Placeholder
57
58     # DGM Architecture
59     entropy_ratio_current=entropy_ratio,
60     dgm_width_current=dgm_width,
61     dgm_depth_current=dgm_depth,
62     architecture_scaling_events=state.architecture_scaling_events,
63
64     # JKO Flow
65     entropy_window_current=entropy_window,
66     learning_rate_current=learning_rate,
67     volatility_sigma_squared=volatility_sigma_squared,
68
69     # Stiffness Thresholds
70     stiffness_low_adaptive=stiffness_low,
71     stiffness_high_adaptive=stiffness_high,
72     holder_exponent_wtmm=holder_exponent_val
73 )

```

9.5.4 Integration Pattern

Adaptive telemetry collection is called periodically (e.g., every 100 steps) in the orchestration loop:

```

1 # In orchestrate_step() or meta-loop
2 if step % telemetry_interval == 0:
3     adaptive_tel = collect_adaptive_telemetry(state, config)
4     if adaptive_tel:
5         emit_adaptive_telemetry(adaptive_tel)

```

Output: JSON Lines format appended to `io/adaptive_telemetry.jsonl`

9.5.5 Metrics Tracked

SDE Solver Frequency:

- `scheme_frequency_explicit`: Fraction of explicit Euler steps in window $\in [0, 1]$
- `scheme_frequency_implicit`: Fraction of implicit solver steps in window $\in [0, 1]$

DGM Architecture:

- `entropy_ratio_current`: $\kappa = H_{\text{current}}/H_{\text{baseline}}$ (regime transition detector)
- `dgm_width_current`, `dgm_depth_current`: Current DGM network dimensions (may be scaled dynamically)
- `architecture_scaling_events`: Cumulative count of architecture mutations

JKO Flow:

- `entropy_window_current`: Current entropy window $W \propto L^2/\sigma^2$
- `learning_rate_current`: Current JKO learning rate $\eta < 2\varepsilon\sigma^2$
- `volatility_sigma_squared`: Current EWMA variance σ^2

Stiffness Thresholds:

- `stiffness_low_adaptive`, `stiffness_high_adaptive`: Hölder-informed thresholds $\theta_L, \theta_H \propto 1/(1 - \alpha)^2$
- `holder_exponent_wtmm`: Current Hölder exponent $\alpha \in [0.2, 0.9]$

9.5.6 Future Extensions

- `max_stiffness_metric`: Maximum stiffness metric \mathcal{S} in window (requires Kernel C integration)
- `num_internal_iterations_mean`: Average Newton iterations for implicit solver
- `implicit_residual_norm_max`: Maximum residual norm for implicit solver convergence

These metrics require deeper integration with Kernel C's SDE solver internals and are currently placeholders (set to 0.0).

9.6 Utility Functions

Module: stochastic_predictor/io/config_mutation.py

```
1 def create_config_backup(  
2     config_path: Path = Path("config.toml"),  
3     backup_path: Path = Path("config.toml.bak")  
4 ) -> None:  
5     """Create config backup before mutation."""  
6     if not config_path.exists():  
7         raise FileNotFoundError(f"Config file not found: {config_path}")  
8     shutil.copy2(config_path, backup_path)  
9  
10 def append_audit_log(log_path: Path, entry: Dict) -> None:  
11     """Append mutation event to audit log (JSON Lines format)."""  
12     log_path.parent.mkdir(parents=True, exist_ok=True)  
13     with open(log_path, 'a') as f:  
14         f.write(json.dumps(entry) + '\n')
```

9.7 Implementation Status

V-MAJ Violation	Status	Module
V-MAJ-4 (Rate Limiting)	Implemented	config_mutation.py
V-MAJ-5 (Auto-Rollback)	Implemented	config_mutation.py
V-MAJ-7 (Telemetry)	Implemented	telemetry.py, types.py

Table 9.1: Level 4 Autonomy - Configuration Mutation Safety Implementation

Note: V-MAJ-7 is fully implemented with `InternalState` extensions (solver counters, baseline entropy, architecture scaling events) and `collect_adaptive_telemetry()` function. Three Kernel C solver metrics remain as placeholders (`max_stiffness_metric`, `num_internal_iterations_mean`, `implicit_residual_norm_max`) requiring deeper SDE solver integration in a future phase.

Chapter 10

Compliance Checklist

- **No Compute Stalls:** All logging is asynchronous
- **Binary Format:** MessagePack for snapshots
- **Atomic Snapshots:** Write-then-rename protocol
- **Deterministic Hashing:** SHA-256 on ρ and OT cost
- **Security:** No raw signals, VRAM dumps, or secrets
- **Integrity:** Snapshot hashes verified before load
- **Config-Driven:** Intervals and destinations are injected
- **Module Coverage:** IO helpers implemented for validation, telemetry, snapshots, and credentials
- **Orchestrator Integration:** IO ingestion gate integrated into `orchestrate_step()`
- **PRNG Constants:** Named constants (`RNG_SPLIT_COUNT`) reside in `api/prng.py`
- **Buffer Capacity Injection:** TelemetryBuffer capacity injected from config (zero-heuristics policy)
- **64-bit Precision:** Enforced at module load time (`api/config.py`) before XLA tracing
- **Layer Isolation:** PRNG constants in API layer, not Core layer

Chapter 11

Phase 4 Summary

Phase 4 introduces a non-blocking I/O architecture that preserves deterministic compute while enabling telemetry, logging, and atomic snapshot persistence.