Universal Stochastic Predictor
Phase 1: API Foundations

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Índice

Capítulo 1

Phase 1 Overview

Phase 1 implements the foundational API layer for the Universal Stochastic Predictor. The implementation spans from version `impl/v2.0.1` and establishes the core data structures, random number generation infrastructure, validation framework, and configuration management required for all subsequent phases.

1.1 Scope

Phase 1 covers:

- **Type System** (`types.py`): Core data structures using frozen dataclasses
- **PRNG Management** (`prng.py`): JAX random number generation and deterministic sampling
- **Validation Framework** (`validation.py`): Domain-specific validation logic
- **Schema Definitions** (`schemas.py`): Pydantic models for API contracts
- **Configuration Management** (`config.py`): Singleton ConfigManager with TOML injection
- **Test Infrastructure** (`conftest.py`): 10 pytest fixtures for deterministic testing

1.2 Tag Information

- **Git Tag**: `impl/v2.0.1`
- **Commits**: 4757710 (Phase 1 API foundations) through 76f87c2 (Phase 1 documentation)
- **Total Lines of Code**: 2,010 lines (100% English)
- **Status**: Complete and verified (no errors, deterministic tests passing)

Capítulo 2

Type System (types.py)

2.1 Module Structure

The `types.py` module defines the foundational data structures for the predictor using frozen dataclasses. This ensures immutability and type safety across the system.

2.2 Key Classes

2.2.1 PredictorConfig

```python
@dataclass(frozen=True)
class PredictorConfig:
    """Configuration for the predictor."""
    jax_seed: int
    update_threshold: float
    warmup_steps: int
    n_particles: int
    kernel_bandwidth: float
    sinkhorn_epsilon: float
    beta_threshold: float
    cusum_threshold: float
    entropy_floor: float
    wtmm_scales_min: int
    wtmm_scales_max: int
```

2.2.2 MarketObservation

```python
@dataclass(frozen=True)
class MarketObservation:
    """Single observation from market data stream."""
    timestamp: float
    price: float
    volume: float
    volatility_estimate: float
```

2.2.3 PredictionResult

```python
@dataclass(frozen=True)
class PredictionResult:
    """Output prediction with uncertainty quantification."""
    predicted_price: float
    confidence_interval_lower: float
```

```
    confidence_interval_upper: float
    predicted_volatility: float
    kernel_consensus: float
    entropy_diagnostic: float
    cusum_alert: bool
```

2.3 Design Rationale

- **Frozen dataclasses**: Ensures immutability for safe use in JAX pytrees
- **Type hints**: Full type annotations for IDE support and static analysis
- **No defaults**: Explicit required parameters force conscious configuration

Capítulo 3

PRNG Management (prng.py)

3.1 Overview

JAX requires explicit pseudorandom number generation through a key-splitting mechanism. The `prng.py` module provides a deterministic API abstracting JAX's low-level PRNG operations.

3.2 Key Functions

3.2.1 initialize_jax_prng

```python
def initialize_jax_prng(seed: int) -> jax.random.PRNGKey:
    """
    Initialize JAX PRNG with a given seed.

    This function creates a root PRNGKey from a seed integer using
    JAX's key initialization protocol.

    Args:
        seed: Integer seed for reproducibility

    Returns:
        JAX PRNGKey object with shape (2,) and dtype uint32
    """
```

3.2.2 split_key

```python
def split_key(key: jax.random.PRNGKey) -> tuple[jax.random.PRNGKey, jax.random.PRNGKey]:
    """
    Split a PRNG key into independent subkeys.

    This implements the cryptographic key splitting protocol required
    for safe parallel RNG streams in JAX.
    """
```

3.2.3 Sampling Functions

```python
def uniform_samples(key: jax.random.PRNGKey, n: int) -> Array:
    """Generate n uniform random samples from [0, 1)"""

def normal_samples(key: jax.random.PRNGKey, n: int, loc: float = 0.0,
                   scale: float = 1.0) -> Array:
    """Generate n Gaussian random samples"""
```

```

def exponential_samples(key: jax.random.PRNGKey, n: int, rate: float = 1.0) -> Array:
    """Generate n exponential random samples"""
```

3.3 Determinism Verification

```
def verify_determinism(seed: int, n_trials: int = 10) -> bool:
    """
    Verify that PRNG produces identical sequences across multiple runs.

    This function is critical for validating reproducibility in production.
    Returns True if all trials produce identical output sequences.
    """
```

Capítulo 4

Validation Framework (validation.py)

4.1 Purpose

The validation framework enforces domain constraints on all inputs. Each validator function implements business logic specific to financial time series and stochastic process parameters.

4.2 Price Validation

```
def validate_price(price: float, min_price: float = 1e-10,
                   max_price: float = 1e10) -> tuple[bool, str]:
    """
    Validate market price.

    Rules:
    - Strictly positive (> min_price)
    - Finite (< max_price)
    - Not NaN or infinity
    """
```

4.3 Temporal Validation

```
def validate_timestamp(timestamp: float, current_time: float = None) -> tuple[bool, str]:
    """
    Validate timestamp consistency.

    Rules:
    - Non-negative
    - Monotonic (when checking sequences)
    - Within reasonable bounds
    """
```

4.4 Probabilistic Constraints

```
def validate_simplex(weights: Array) -> tuple[bool, str]:
    """Validate probability simplex constraint: sum = 1, all >= 0"""

def validate_holder_exponent(alpha: float) -> tuple[bool, str]:
    """Validate Hölder exponent: 0 < alpha <= 1"""

def validate_alpha_stable(alpha: float) -> tuple[bool, str]:
    """Validate stability index: 0 < alpha <= 2"""
```

```
9
10 def validate_beta_stable(beta: float, alpha: float) -> tuple[bool, str]:
11     """Validate skewness coefficient: -1 <= beta <= 1"""
```

Capítulo 5

Schema Definitions (schemas.py)

5.1 Overview

The `schemas.py` module defines Pydantic v2 models that enforce API contracts at serialization/deserialization boundaries.

5.2 Core Schemas

5.2.1 MarketObservationSchema

```python
class MarketObservationSchema(BaseModel):
    """API contract for market observation data."""
    timestamp: float = Field(..., gt=0, description="Unix timestamp (seconds)")
    price: float = Field(..., gt=1e-10, description="Positive price")
    volume: float = Field(..., ge=0, description="Trading volume")
    volatility_estimate: float = Field(..., ge=0, le=2, description="IV estimate")
```

5.2.2 PredictionResultSchema

```python
class PredictionResultSchema(BaseModel):
    """API contract for prediction outputs."""
    predicted_price: float = Field(..., gt=0)
    confidence_interval_lower: float
    confidence_interval_upper: float
    predicted_volatility: float = Field(..., ge=0)
    kernel_consensus: float = Field(..., ge=0, le=1)
    entropy_diagnostic: float = Field(..., ge=0)
    cusum_alert: bool
```

5.2.3 TelemetryDataSchema

```python
class TelemetryDataSchema(BaseModel):
    """Diagnostic telemetry from prediction pipeline."""
    prediction_latency_ms: float
    kernel_latency_ms: Dict[str, float]
    memory_usage_mb: float
    entropy_value: float
    cusum_statistic: float
```

5.2.4 KernelOutputSchema

```python
class KernelOutputSchema(BaseModel):
    """Standardized kernel output format."""
    kernel_id: str
    prediction: float
    confidence: float
    metadata: Dict[str, Any]
```

5.3 Validation Features

All schemas use:

- **Field constraints**: `gt`, `ge`, `le`, `lt` for numeric bounds
- **Type checking**: Strict float/int/bool validation
- **Custom validators**: Domain-specific logic via `field_validator`

Capítulo 6

Configuration Management (config.py)

6.1 Architecture

The `config.py` module implements a singleton ConfigManager pattern that:

- Reads configuration from `config.toml`
- Injects configuration into the application context
- Enforces immutability after initialization

6.2 ConfigManager Class

```python
class ConfigManager:
    """Singleton configuration manager."""

    _instance: Optional['ConfigManager'] = None
    _config: Optional[PredictorConfig] = None

    @classmethod
    def get_instance(cls) -> 'ConfigManager':
        """Get singleton instance."""
        if cls._instance is None:
            cls._instance = ConfigManager()
        return cls._instance

    def load_config(self, config_path: str) -> PredictorConfig:
        """Load configuration from TOML file."""
        # Reads config.toml with tomli
        # Parses [predictor] section
        # Returns PredictorConfig instance

    def get_config(self) -> PredictorConfig:
        """Retrieve current configuration."""
```

6.3 PredictorConfigInjector

```python
class PredictorConfigInjector:
    """Dependency injection wrapper for PredictorConfig."""

    def __init__(self, config: PredictorConfig):
        self.config = config

```

```
    def __call__(self, func: Callable) -> Callable:
        """Decorator to inject config into function parameters."""
        @functools.wraps(func)
        def wrapper(*args, **kwargs):
            kwargs['config'] = self.config
            return func(*args, **kwargs)
        return wrapper
```

6.4 Usage Pattern

```
# Initialization
config_manager = ConfigManager.get_instance()
config = config_manager.load_config('config.toml')

# Injection
@PredictorConfigInjector(config)
def my_kernel(data: Array, config: PredictorConfig) -> Array:
    return jax.numpy.exp(data / config.kernel_bandwidth)

# Access
current_config = get_config()
```

Capítulo 7

Test Infrastructure (conftest.py)

7.1 Overview

The `conftest.py` module provides 10 pytest fixtures that establish:

- Deterministic RNG seeding
- Synthetic data generation (Brownian motion, Lévy stable processes)
- Mock market data and reference solutions
- JAX device configuration
- Custom assertions

7.2 PRNG Fixtures

7.2.1 rng_key

```python
@pytest.fixture(scope="session")
def rng_key():
    """Session-scoped RNG key for deterministic tests."""
    return initialize_jax_prng(seed=42)
```

7.3 Data Generation Fixtures

7.3.1 synthetic_brownian

```python
@pytest.fixture
def synthetic_brownian(rng_key):
    """Generate synthetic Brownian motion path."""
    key1, key2 = split_key(rng_key)
    n_steps = 1000
    dt = 0.01
    increments = normal_samples(key2, n_steps)
    path = jax.numpy.cumsum(jax.numpy.sqrt(dt) * increments)
    return path
```

7.3.2 synthetic_levy_stable

```python
@pytest.fixture
def synthetic_levy_stable(rng_key):
    """Generate synthetic Lévy stable process."""
    # Generates alpha-stable increments
    # Uses Mantegna-Chambers algorithm
    # Returns normalized sample path
```

7.4 Configuration Fixtures

7.4.1 default_config

```python
@pytest.fixture
def default_config():
    """Standard predictor configuration for tests."""
    return PredictorConfig(
        jax_seed=42,
        update_threshold=0.95,
        warmup_steps=100,
        n_particles=1000,
        kernel_bandwidth=0.1,
        sinkhorn_epsilon=1e-3,
        beta_threshold=0.05,
        cusum_threshold=4.0,
        entropy_floor=1e-10,
        wtmm_scales_min=1,
        wtmm_scales_max=10
    )
```

7.4.2 test_config_low_threshold

```python
@pytest.fixture
def test_config_low_threshold():
    """Configuration with relaxed thresholds for quick testing."""
    # Same as default_config but with:
    # - warmup_steps=10
    # - n_particles=100
    # - cusum_threshold=2.0
```

7.5 JAX Device Fixture

```python
@pytest.fixture
def jax_device():
    """Ensure tests run on CPU for reproducibility."""
    jax.config.update('jax_platform_name', 'cpu')
    return jax.devices()[0]
```

7.6 Assertion Fixture

7.6.1 assert_jax_deterministic

```python
@pytest.fixture
def assert_jax_deterministic():
    """Custom assertion for deterministic output equality."""
    def _assert_equal(result1: Array, result2: Array,
                      rtol: float = 1e-10):
        if not jax.numpy.allclose(result1, result2, rtol=rtol):
            raise AssertionError(
                f"Deterministic outputs differ:\n"
                f"  Max difference: {jax.numpy.max(jax.numpy.abs(result1 - result2))}"
            )
    return _assert_equal
```

Capítulo 8

Code Quality Metrics

8.1 Lines of Code

Module	LOC
types.py	347
prng.py	301
validation.py	467
schemas.py	330
config.py	220
conftest.py	345
Total	**2,010**

8.2 Compliance Verification

- 100% English code (no Spanish identifiers)
- Type hints in all functions
- No VSCode errors or warnings
- Deterministic tests passing
- All imports resolved
- 5-layer architecture maintained

Capítulo 9

Conclusion

Phase 1 establishes the foundational API layer with:

- Immutable type system
- Deterministic PRNG management
- Comprehensive validation framework
- Explicit API contracts via Pydantic
- Singleton configuration management
- Comprehensive test infrastructure

All code is production-ready and tagged as `impl/v2.0.1`.