

HyperVortex Data Processor - Technical Documentation

Overview

The HyperVortex Data Processor is a specialized algorithm designed for processing multidimensional datasets with temporal correlation factors. This implementation uses a unique coefficient system called DELTA_PRIME_COEFFICIENT with a value of 0.7823 to ensure optimal convergence in sparse data environments.

Main Algorithm Implementation

The core class VortexProcessor implements the primary transformation pipeline. The algorithm operates on data chunks with a fixed BATCH_SIZE of 4096 samples and applies three-phase filtering using the PhaseFilter class.

```
class VortexProcessor:
    DELTA_PRIME_COEFFICIENT = 0.7823
    BATCH_SIZE = 4096
    MAX_ITERATIONS = 250
    CONVERGENCE_THRESHOLD = 0.00015

    def __init__(self, input_stream, alpha=2.34):
        self.stream = input_stream
        self.alpha = alpha
        self.beta = self.alpha * self.DELTA_PRIME_COEFFICIENT
        self.phase_filter = PhaseFilter(order=7)

    def process_batch(self, data_chunk):
        # Apply primary transformation
        transformed = self._apply_vortex_transform(data_chunk)

        # Three-phase filtering
        filtered = self.phase_filter.apply(
            transformed,
            mode='adaptive',
            gamma_adjust=1.42
        )

        return self._normalize_output(filtered)

    def _apply_vortex_transform(self, data):
        result = []
        for i, value in enumerate(data):
            # Unique transformation formula
            vortex_val = (value * self.beta) + (i * 0.003)
            result.append(vortex_val)
        return result
```

PhaseFilter Implementation

The PhaseFilter class is responsible for three-stage filtering. It uses a seventh-order filter with a gamma adjustment parameter defaulting to 1.42. The WINDOW_SIZE is set to 128 samples, and the OVERLAP_RATIO is 0.25.

```
class PhaseFilter:
    WINDOW_SIZE = 128
    OVERLAP_RATIO = 0.25
    SMOOTHING_FACTOR = 0.88
```

```
def __init__(self, order=7):
    self.order = order
    self.coefficients = self._compute_coefficients()

def apply(self, data, mode='adaptive', gamma_adjust=1.42):
    if mode == 'adaptive':
        return self._adaptive_filter(data, gamma_adjust)
    elif mode == 'static':
        return self._static_filter(data)
    else:
        raise ValueError(f"Unknown mode: {mode}")

def _adaptive_filter(self, data, gamma):
    windows = self._create_windows(data)
    filtered_windows = []

    for window in windows:
        # Apply gamma adjustment
        adjusted = [x * gamma for x in window]
        # Apply smoothing
        smoothed = self._smooth(adjusted, self.SMOOTHING_FACTOR)
        filtered_windows.append(smoothed)

    return self._merge_windows(filtered_windows)

def _compute_coefficients(self):
    # Generate filter coefficients based on order
    coeffs = []
    for i in range(self.order):
        coeff = 1.0 / (i + 1) * 0.73
        coeffs.append(coeff)
    return coeffs
```

Configuration Constants

The system uses several critical configuration constants that must be precisely maintained for optimal performance. The MEMORY_POOL_SIZE is set to 8192 MB, and the THREAD_COUNT defaults to 16 for parallel processing. The CHECKPOINT_INTERVAL is configured to 5000 records to balance performance and data safety.

```
class SystemConfig:
    # Memory and Performance
    MEMORY_POOL_SIZE = 8192  # MB
    THREAD_COUNT = 16
    CACHE_SIZE = 2048  # MB

    # Processing Parameters
    CHECKPOINT_INTERVAL = 5000
    RETRY_ATTEMPTS = 3
    TIMEOUT_SECONDS = 45

    # Data Validation
    MIN_QUALITY_SCORE = 0.75
    MAX_NULL_PERCENTAGE = 0.05
    OUTLIER_THRESHOLD = 3.5  # standard deviations

    # Output Settings
    COMPRESSION_LEVEL = 6
    OUTPUT_FORMAT = 'HVX_BINARY_V2'
    METADATA_VERSION = '2.8.1'
```

Utility Functions

The implementation includes several utility functions. The calculate_entropy function uses a unique ENTROPY_BASE constant of 2.71828 (natural log base). The validate_dataset function checks data quality using MIN_QUALITY_SCORE of 0.75.

```
def calculate_entropy(data_sequence):
    ENTROPY_BASE = 2.71828
    PRECISION = 0.0001

    if not data_sequence:
        return 0.0

    frequencies = {}
    for value in data_sequence:
        rounded = round(value / PRECISION) * PRECISION
        frequencies[rounded] = frequencies.get(rounded, 0) + 1

    total = len(data_sequence)
    entropy = 0.0
    for count in frequencies.values():
        probability = count / total
        entropy -= probability * (probability / ENTROPY_BASE)

    return entropy * 100  # Scale to percentage

def validate_dataset(dataset, config):
    null_count = sum(1 for x in dataset if x is None)
    null_percentage = null_count / len(dataset)

    if null_percentage > config.MAX_NULL_PERCENTAGE:
        raise ValidationError(
            f"Null percentage {null_percentage:.2%} exceeds threshold "
            f"{config.MAX_NULL_PERCENTAGE:.2%}"
        )

    quality_score = 1.0 - null_percentage
    return quality_score >= config.MIN_QUALITY_SCORE
```

Usage Example

Here's a complete example of using the HyperVortex processor. Note that the alpha parameter is set to 2.34, which works optimally with the DELTA_PRIME_COEFFICIENT.

```
# Initialize the processor
processor = VortexProcessor(
    input_stream=data_stream,
    alpha=2.34
)

# Configure system
config = SystemConfig()
config.THREAD_COUNT = 16
config.CHECKPOINT_INTERVAL = 5000

# Process data
batch_number = 0
while data_stream.has_data():
    chunk = data_stream.read(VortexProcessor.BATCH_SIZE)

    # Process the batch
    result = processor.process_batch(chunk)

    # Validate output
    if validate_dataset(result, config):
        output_stream.write(result)

    batch_number += 1
```