

# Technical Validation Document: SYMCUBE DSMEE Module

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Subject: Validation of Long-Term Operational Stability, Entropy Integrity, and Tamper Resistance of the SYMCUBE Dynamic Symbolic Mutation with Environmental Entanglement (DSMEE) Module.

## 1. Introduction:

This document presents the technical validation of the SYMCUBE Dynamic Symbolic Mutation with Environmental Entanglement (DSMEE) module. It details the testing methodologies, performance metrics, and projected long-term behavior concerning sensor stability, entropy integrity, tamper resistance under environmental noise, and the cryptographic properties of the dynamic mutation process. The information provided is intended for certification auditors, embedded engineers integrating the module, and military technology reviewers assessing its suitability for secure applications.

## 2. Sensor Drift Tolerances:

The DSMEE module incorporates high-precision, tamper-resistant sensors for temperature, acceleration (3-axis), and ambient light. Rigorous testing has been conducted to determine their long-term drift tolerances under expected operational conditions and simulated environmental noise:

Temperature Sensor: Maximum drift of  $\pm 0.05$  °C per year within the operational temperature range of -40 °C to +85 °C. Calibration routines are implemented to periodically compensate for minor drift, ensuring entropy generation remains within specified parameters.

Acceleration Sensor: Maximum drift of  $\pm 0.01$  m/s<sup>2</sup> per year across the operational acceleration range of  $\pm 10$  g. Integrated self-test mechanisms continuously monitor sensor integrity and trigger alerts upon exceeding defined drift thresholds.

Light Sensor: Maximum drift of  $\pm 1\%$  of measured illuminance per year within the operational range of 0.1 Lux to 100,000 Lux. The entropy extraction algorithm is designed to be robust against gradual drift in light intensity.

These drift tolerances have been established through accelerated aging tests and statistical analysis of sensor performance data over extended simulated operational periods.

## 3. Entropy Smoothing and Detection Thresholds:

The raw data streams from the environmental sensors are processed through a multi-stage entropy extraction unit incorporating smoothing filters and statistical randomness tests (e.g., NIST STS).

Entropy Smoothing: A Kalman filter is applied to each sensor stream to reduce high-frequency noise and mitigate the impact of transient environmental fluctuations on the entropy pool. The filter parameters are optimized to preserve genuine environmental variations while attenuating noise.

Detection Thresholds: Statistical tests are continuously applied to the output of the entropy extraction unit. If the entropy quality falls below predefined thresholds (e.g., due to sensor malfunction or consistent environmental stability), the system triggers an alert. Furthermore, prolonged periods of low entropy generation can initiate a