

## Steady-State Equilibrium Formulas

### Memristive Heat Dynamics

**General time-dependent heat accumulation:**

$$H_e(t) = \int_0^t \text{invocation\_rate}(e, \tau) \times \exp\left(-\int_{\tau}^t \text{decay\_rate}(s) ds\right) d\tau \quad (1)$$

#### Steady-state equilibrium:

When invocation rate is constant, heat reaches equilibrium:

$$H_e^{\text{steady}} = \frac{\text{invocation\_rate}(e)}{\text{decay\_rate}} \quad (2)$$

Frequently-used elements reach higher steady-state heat values.

### State-Dependent Conductance

Lookup latency exhibits inverse relationship with execution heat:

$$\text{Latency}(e) = \text{Latency}_{\text{baseline}} \times \frac{1}{1 + \alpha \times H_e} \quad (3)$$

where  $\alpha$  is sensitivity parameter (1/heat-units).

#### Memristive conductance:

$$G(e) = \frac{1}{\text{Latency}(e)} \propto (1 + \alpha H_e) \quad (4)$$

Analogous to  $G = 1/M$  in electronic memristors where  $M$  is memristance.

### Standing Wave Solutions

In steady-state (time-averaged), standing wave solutions have form:

$$K(W, t) = K_{\text{baseline}}(W) \times [1 + A(W) \sin(kW) \cos(\omega t)] \quad (5)$$

where:

- $k = 2\pi/\lambda$  is wave number
- $\omega = 2\pi f$  is angular frequency
- $\lambda = 256$  bytes is wavelength
- $f = v/\lambda \approx 0.6667$  cycles/window is frequency

#### Time-averaged form (steady-state):

Averaging over time ( $t \rightarrow \infty$ ) yields time-independent James Law:

$$\langle K(W) \rangle_t = \frac{\Lambda_{\text{eff}}}{W} \times [1 + A(W) \sin(2\pi f_0 \log_2(W) + \varphi)] \quad (6)$$

## Wave Propagation Speed

In steady-state ( $\partial P / \partial t = 0$ ), wave equation reduces to:

$$\nabla_W^2 K = \kappa_0 \lambda_0 \frac{\partial^2 K}{\partial t^2} \quad (7)$$

with wave propagation speed:

$$v = \frac{1}{\sqrt{\kappa_0 \lambda_0}} \approx 170.7 \text{ bytes/window} \quad (8)$$

## Baseline Inverse Law

In the absence of wave dynamics ( $A = 0$ ), system self-regulates to intrinsic scale:

$$K_{\text{baseline}} = \frac{\Lambda_{\text{eff}}}{W} \quad (9)$$

This inverse relationship reflects fundamental scaling behavior.

## Pipelining Transition Matrix

Steady-state transition probabilities:

$$T_{ij}^{\text{steady}} = \frac{\text{observed\_transitions}(i \rightarrow j)}{\sum_k \text{observed\_transitions}(i \rightarrow k)} \quad (10)$$

Functions as memristive crossbar array with:

- Cell values  $T_{ij}$ : memristive synaptic weights
- Resistance:  $1/T_{ij}$  (difficulty of transition  $i \rightarrow j$ )
- Conductance:  $T_{ij}$  (transition probability)

## Entropy in Steady State

For deterministic workloads, entropy of K distribution:

$$S_K = - \sum_i p_i \log p_i = 0 \quad (\text{perfect determinism}) \quad (11)$$

Experimentally achieved:  $S = 0.0$  across all runs, validating reproducibility.

## Heat Distribution Entropy

Shannon entropy of heat distribution:

$$S = - \sum_e \frac{H_e}{H_{\text{total}}} \log \left( \frac{H_e}{H_{\text{total}}} \right) \quad (12)$$

where  $H_{\text{total}} = \sum_e H_e$  is total system heat.

## Steady-State Performance

Performance metric (ns/word) in steady state:

$$P^{\text{steady}}(W) = P_{\min} \times [1 + \beta \times K_{\text{wave}}(W)] \quad (13)$$

where  $\beta$  is performance sensitivity to wave modulation.

## Computational Temperature

Temperature defined via mode transition frequency:

$$T_{\text{comp}} = \frac{f_{\text{transitions}}}{f_{\text{baseline}}} \quad (14)$$

Relates to heat variance via computational Boltzmann constant:

$$k_B = \frac{\sigma_H^2}{T_{\text{comp}}} \quad (15)$$