

Stochastic Thermodynamics and Information-Energy Conversion in Feedback-Controlled Systems

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Abstract: *This work examines how thermodynamics must be rethought when systems become small, noisy, and driven far from equilibrium, as in colloidal particles and molecular machines. Building on stochastic thermodynamics, feedback controlled Brownian particle in a time dependent trap is modeled to trace work, heat, and entropy along individual trajectories rather than relying on bulk averages. A discrete measurement and feedback protocol is introduced, allowing information gathered about the particle position to be used for controlled trap shifts and work extraction. Mutual information between system and measurement is quantified and incorporated into a generalized second law, which sets a bound on extractable work once informational contributions are included. Large scale numerical simulations confirm that apparent violations of the conventional second law disappear when information is properly accounted for, and generalized fluctuation relations remain valid under feedback. The distribution of stochastic entropy production is analyzed at the trajectory level, and connections to thermodynamic uncertainty relations are discussed, highlighting trade-offs among precision, dissipation, and available information. The findings position information as a measurable thermodynamic resource with practical relevance for nanoscale engines and biological processes that operate under continual fluctuations.*

Keywords: Stochastic Thermodynamics, Feedback Control, Information-Energy Conversion, Fluctuation Theorems, Mutual Information, Entropy Production, Maxwell's Demon, Brownian Particles, Optical Tweezers, Nonequilibrium Steady State

1. Introduction

The traditional laws of thermodynamics were formulated for macroscopic systems in equilibrium. However, the advent of nanotechnology and single-molecule experiments has revealed a rich landscape of behavior in small systems where fluctuations dominate. These systems operate far from equilibrium and require a new theoretical framework.

1.1 Motivation

The rectification of thermal fluctuations using information-a concept embodied by maxwell's demon-has transitioned from philosophical debate to experimental reality. Understanding how information can be converted into work is crucial for:

- Fundamental physics: Reconciling thermodynamics with information theory.
- Nanotechnology: Designing efficient molecular machines.
- Biological systems: Understanding how living organisms harness information for energy.

1.2 Research Objectives

- To formulate the stochastic thermodynamics of feedback-controlled systems
- To derive and verify the generalized second law: $\langle w \rangle \geq \delta f - k_B \langle i \rangle$, where i is the mutual information between the system and measurement.
- To quantify entropy production along individual stochastic trajectories
- To demonstrate work extraction from information in a simulated colloidal system.

2. Theoretical Background

2.1 Stochastic thermodynamics

For small systems, thermodynamic quantities become stochastic variables defined along individual trajectories. The first law for a brownian particle in a time-dependent potential $u(x, \lambda(t))$ is:

$$Du = dw + dq$$

Where work $dw = (\partial u / \partial \lambda) d\lambda$ and heat dq satisfy the fluctuation theorem.

2.2 Entropy production and time-reversal

The total entropy production δ_{tot} quantifies the irreversibility of a trajectory:

$$\Delta \delta_{\text{tot}} = \ln[p[\text{reverse trajectory}] / p[\text{forward trajectory}]]$$

This satisfies the integral fluctuation theorem $\langle e^{-\delta_{\text{tot}}} \rangle = 1$, implying $\langle \delta_{\text{tot}} \rangle \geq 0$ ($\delta_{\text{tot}} \geq 0$)

2.3 Information thermodynamics

When a system is subject to feedback control based on measurements, the conventional second law generalizes to: $\langle \delta_{\text{tot}} \rangle \geq i$

Where i , is the mutual information between the system state and measurement outcome. This allows for work extraction beyond the free energy decrease:

$$\langle w \rangle \geq \delta f - k_B \langle i \rangle$$

2.4 Fluctuation theorems with feedback

The jarzynski equality generalizes to:

$$\langle e^{-\beta(w - \delta f) - i} \rangle = 1$$

Providing a fundamental constraint on work extraction in feedback-controlled systems.

3. Methodology

3.1 Model system: Overdamped Langevin dynamics

We simulate a colloidal particle in a one-dimensional optical trap described by:

$$\dot{x}(t) = -1/\gamma (\partial u(x, \lambda)) / \partial x + ((2k_B T)/\gamma)^{1/2} \xi(t)$$

Where $\xi(t)$ is Gaussian white noise.

3.2 Feedback protocol: Information engine

We implement a discrete feedback protocol:

- 1) Measurement: At regular intervals, measure the particle position with finite resolution
- 2) Feedback: Shift the trap center based on the measurement to extract work
- 3) Work extraction: Calculate the work performed by the particle against the trap

The measurement error is modeled with conditional probability $p(m|x)$, where m is the measurement outcome.

3.3 Quantifying information

The mutual information between true position x and measurement m is:

$$I = \iint p(x, m) \ln(p(x)p(m)/p(x, m)) dx dm$$

This quantifies how much information the measurement provides about the system state.

3.4 Numerical simulations

We perform simulations for:

- 1) Different feedback strengths (how much the trap shifts)
- 2) Various measurement resolutions (Gaussian measurement noise)
- 3) Multiple temperatures and trap stiffnesses
- 4) Ensemble sizes of 10^6 trajectories for statistical accuracy

Observables include:

Extracted work, entropy production, mutual information, and verification of generalized fluctuation theorems.

4. Results and Discussion

4.1 Work extraction from information

Our simulations demonstrate that work can be extracted from a single heat bath using only information. The extracted work increases with:

- Measurement precision (higher mutual information)
- Feedback efficiency (optimal response to measurements)
- Appropriate timing between measurements

4.2 Verification of generalized second law:

For all parameter regimes, we verify that $(W) + k_B TI \geq \Delta F$ confirming the information-theoretic extension of thermodynamics. Systems that appear to violate the conventional second law ($W < \Delta F$) always satisfy the generalized bound when information is accounted for.

4.3 Entropy production along trajectories:

Following the approach in, we compute the stochastic entropy production along individual trajectories. The distribution of entropy production satisfies the integral fluctuation theorem $\langle e^{\Delta S_{\text{tot}} - i} \rangle = 1$, confirming the consistency of our information-theoretic framework.

4.4 Comparison with uncertainty relations:

We analyze the thermodynamic uncertainty relation (tur) for our system, which bounds the precision of any current by the entropy production rate. Feedback control modifies this bound, revealing fundamental trade-offs between information, precision, and dissipation.

5. Conclusion

This thesis has demonstrated that information is a genuine thermodynamic resource, convertible to energy subject to fundamental limits derived from stochastic thermodynamics.

Key findings:

- 1) Conclusion 1: Information-energy conversion is quantitatively described by the generalized second law, with mutual information playing a role analogous to negative entropy
- 2) Conclusion 2: Fluctuation theorems extend naturally to feedback-controlled systems, providing rigorous constraints on work extraction
- 3) Conclusion 3: The thermodynamic uncertainty relation sets precision limits that depend on both dissipation and available information

6. Implications

These results have profound implications for understanding biological information processing (e.g., cellular sensing, proofreading in protein synthesis) and for designing nanoscale machines that operate efficiently far from equilibrium.

7. Future Work

- 1) Extend to quantum systems where measurement back-action introduces additional constraints
- 2) Investigate multi-particle systems with correlated information
- 3) Apply to experimental colloidal systems with real-time feedback control
- 4) Explore connections to neural dynamics and information processing in living systems

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