



Inferring entropy production from time-dependent moments

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Frontiers in Statistical Physics RRI, Bangalore (Dec, 2023)





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In collaboration with Karel Proesmans (arXiv:2310.16627)

INTRODUCTION

Classical thermodynamics

Deals with the rules that govern conversion of different forms of energy in macroscopic systems

First law:
$$dE = dQ + dW$$

Second law: $\Delta S_{\text{tot}} = \Delta S_{\text{sys}} + \Delta S_{\text{env}}$
 ≥ 0

Stochastic thermodynamics

Framework to study thermodynamics of stochastic systems, which can be arbitrarily from the equilibrium.

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Stochastic thermodynamics

$$\gamma \frac{dx}{dt} = -\frac{\partial U(x,t)}{\partial x} + f(x,t) + \sqrt{2D}\eta(t)$$

Heat
$$Q({x(t)}) \equiv \int \left[-\gamma \frac{dx}{dt} + \sqrt{2D}\eta(t)\right] \circ dx(t)$$

Work
$$W({x(t)}) \equiv \int \left[\frac{\partial U(x,t)}{\partial t}dt + f(x,t) \circ dx(t)\right]$$

 $\frac{\text{First law for a single}}{\text{trajectory}} \quad Q(\{x(t)\}) + W(\{x(t)\}) = \Delta U(x,t)$

K. Sekimoto, Prog. Th. Phys. (1998)

Stochastic entropy production

U. Seifert, PRL (2005)

$$S_{\rm sys}(t) \equiv -k_B \ln P(x(t), t)$$

 $S_{\rm env}(t) \equiv \frac{Q(t)}{T}$

$$\Delta S_{\rm tot}(t) = S_{\rm sys}(t) + S_{\rm env}(t)$$

Stochastic entropy production

U. Seifert, PRL (2005)

$$S_{\rm sys}(t) \equiv -k_B \ln P(x(t), t)$$
$$S_{\rm env}(t) \equiv \frac{Q(t)}{T}$$

$$\Delta S_{\rm tot}(t) = S_{\rm sys}(t) + S_{\rm env}(t)$$

$$\langle \Delta S_{\text{tot}}(t_f) \rangle = \frac{k_B}{D} \int_0^{t_f} dt \int dx \ P(x,t) \ v^2(x,t)$$
$$v(x,t) = \frac{D}{k_B T} \left[F(x,t) - k_B T \frac{\partial}{\partial x} \ln P(x,t) \right]$$

Inferring thermodynamic quantities



 S K Manikandan, D Gupta and S Krishnamurthy, PRL, 124, 120603 (2020)



Based on waiting time distribution

• P E Harunari, A Dutta, M Polettini, E Roldan, PRX 12, 041026 (2022)

Some other for steady-state systems....

Inferring thermodynamic quantities



In this talk, I will present a method for time-dependent systems Refs: arXiv:2310.16627, PS and K. Proesmans

Other works: (1) S. Otsubo, S. Manikandan, T. Sagawa and S. Krishnamurthy Comm. Phys. (2022) (2) S. Lee, D-K Kim, J. Park, W. Kim, H. Park, and J. S. Lee Phys. Rev. Research (2023)

METHOD

Our method gives a general bound on $\Delta S_{tot}(t)$ in terms of moments of x(t)

$$\Delta S_{\text{tot}}(t_f) \ge \frac{k_B}{D} \int_0^{t_f} dt \left[\frac{\dot{\Sigma}^2(t)}{4\Sigma(t)} + \dot{X}_1(t)^2 \right],$$

$$\Sigma(t) = \langle x^2(t) \rangle - \langle x(t) \rangle^2 \qquad X_1(t) = \langle x(t) \rangle$$

Our method gives a general bound on $\Delta S_{tot}(t)$ in terms of moments of x(t)

Optimisation of the action

$$\mathbb{S}(x, v, t_f) = \Delta S_{\text{tot}}(t_f) + \int_0^{t_f} dt \Big[\mu_1(t) \langle x(t) \rangle + \mu_2(t) \langle x^2(t) \rangle \Big]$$

Entropy production First moment Second moment

$$\mathbb{S}(x, v, t_f) = \int_0^{t_f} dt \int dx \ P(x, t) \left[v^2(x, t) + \mu_1(t)x + \mu_2(t)x^2 \right]$$

with
$$v(x,t) = \frac{D}{k_B T} \left[F(x,t) - k_B T \frac{\partial}{\partial x} \ln P(x,t) \right]$$

 $\frac{\partial P(x,t)}{\partial t} = -\frac{\partial}{\partial x} \left[v(x,t) P(x,t) \right]$

In general, difficult to carry out this optimisation because P(x,t) depends on the force field F(x,t).

$$S = \int_{0}^{t_{f}} dt \int dx \ P(x,t) \left[v^{2}(x,t) + \mu_{1}(t)x + \mu_{2}(t)x^{2} \right]$$
$$\dot{y}(x,t) = v \left(y(x,t), t \right)$$
$$y(x,0) = x$$
$$S = \int_{0}^{t_{f}} dt \int dx P_{0}(x) \left[\dot{y}(x,t)^{2} + \mu_{1}(t)y(x,t) + \mu_{2}(t)y(x,t)^{2} \right]$$

JD Benamou and Y Brenier, Numer. Math. (2000) E Aurell, C Mejía-Monasterio and P Muratore-Ginanneschi, PRL (2011)

 $\delta S = 0$ gives Euler-Lagrange equation

$$2\ddot{y}(x,t) = \mu_1(t) + 2\mu_2(t)y(x,t).$$

with boundary conditions

$$2\dot{y}(x,t_f) = -\zeta_1(t_f) - 2\zeta_2(t_f)y(x,t_f)$$
$$y(x,0) = x$$

$$\Delta S_{\text{tot}}(t_f) \ge \frac{k_B}{D} \int_0^{t_f} dt \left[\frac{\dot{\Sigma}^2(t)}{4\Sigma(t)} + \dot{X}_1(t)^2 \right],$$

 $\Sigma(t) = \langle x^2(t) \rangle - \langle x(t) \rangle^2 \qquad X_1(t) = \langle x(t) \rangle$

APPLICATION TO BIT ERASURE EXPERIMENT

Application in Bit erasure

Landauer in 1961 proposed that the minimal thermodynamic cost to erase a bit:

$W \ge k_B T \ln 2$

Equality holds only in the quasi-static limit

R. Landauer, IBM Journal of Research and Development, (1961)

Bit erasure



$$U(x,t) = 4E_b \left[-\frac{g(t)}{2}x^2 + \frac{x^4}{4} - Af(t) x \right]$$

Bit erasure



position x

Work done



Y Jun, M Gavrilov, and J Bechhoefer, PRL, 113, 190601 (2014) A Berut, S Ciliberto et al, Nature, 483 (2012)

$$\Delta S_{\text{tot}}(t_f) \ge \frac{k_B}{D} \int_0^{t_f} dt \left[\frac{\dot{\Sigma}^2(t)}{4\Sigma(t)} + \dot{X}_1(t)^2 \right],$$

(Inference with < 100 trajectories)



Other examples

$$\frac{dx}{dt} = \sqrt{2D} \ \eta(t), \text{ with } P(x_0) \sim e^{-x_0^4}$$



Summary

- We derived a general lower bound on the mean total entropy production for time-dependent systems in terms of the moments.
- We applied this bound to infer total entropy production for bit erasure experiment and found an excellent match with simulation
- Can be extended to general moments and also to higher dimensions

Refs: PS and K Proesmans arXiv:2310.16627

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Thank you

Other examples

$$\frac{dx}{dt} = \sqrt{2D} \ \eta(t), \text{ with } P(x_0) \sim e^{-x_0^4}$$

