

Self-propelled Brownian Particles in Carnot Cycle: As an Engine

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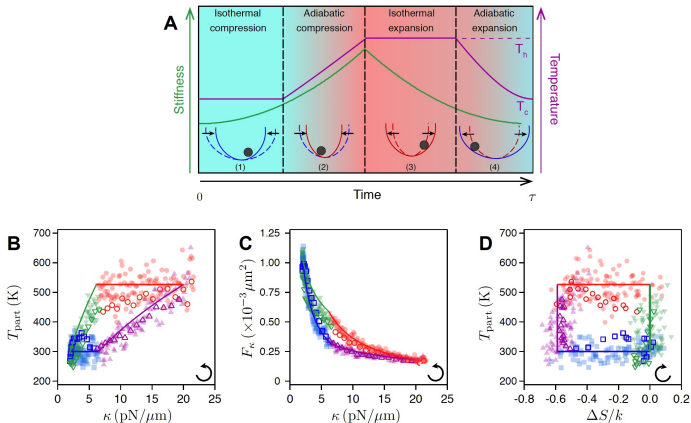
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Background



- 1 dilute Brownian particles \rightarrow analogous Carnot cycle \rightarrow work
- 2 stochastic fluctuation of entropy \rightarrow promising to violate Carnot limit

Motivation

- 1 Fluctuation theorem, laws of thermodynamics are redefined in active Brownian particles.
- 2 active Brownian engine \Rightarrow higher efficiency or higher power ?

Model

Rayleigh-Helmholtz self-propelled Model

$$m\ddot{x} + (\gamma - a + b\dot{x}^2)\dot{x} = -\partial_x U + \xi$$

where

$$\langle \xi(t)\xi(t') \rangle = 2\gamma kT \delta(t - t')$$

- 1 a: pump energy into system \Rightarrow negative friction
- 2 b: increasement of friction caused by motion
- 3 like motor proteins NK11
- 4 Brownian particle is trapped by harmonic laser, where $U(x) = \frac{\lambda}{2}x^2$

Theoretical Analysis

- 1 Langevin equation \rightarrow

$$\begin{aligned} dE &= d\left(\frac{p^2}{2m} + U(x; \lambda)\right) \\ &= \left[\frac{\partial U}{\partial \lambda} d\lambda\right] + \left[\left(-\gamma \frac{p}{m} + \xi\right) dx\right] \\ &= dW + dQ \end{aligned}$$

- 2 Assumption: adiabatic process happens instantaneously without changing P.D.F.

$$W_a = \Delta_a E = \int dx p(x, \tau) (U(x, \tau + 0) - U(x, \tau - 0))$$

3 $\eta \stackrel{d}{=} -\frac{\langle W \rangle}{\langle Q_h \rangle}$

- 3 Ken Sekimoto, Fumiko Takagi, Tsuyoshi Hondou. Carnot's cycle for small systems: Irreversibility and cost of operations. *Physical Review E*. 2014, 62(6): 7759-7767.
- 4 Tim Schmiedl and Udo Seifert. Efficiency at maximum power: An analytically solvable model for stochastic heat engines. *arXiv:0710.4097v1*.

Controlling Protocols

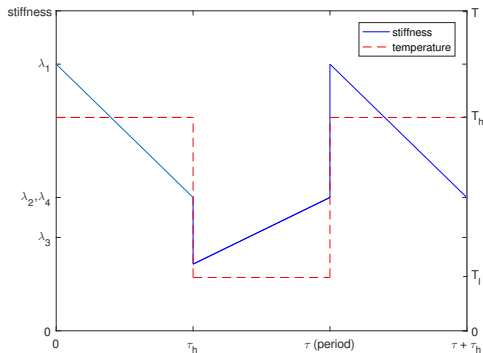
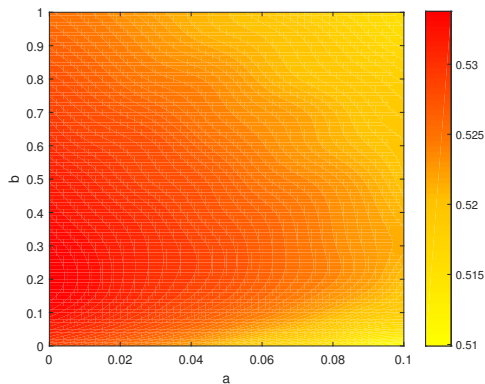


Figure: controlling protocols as a function of time

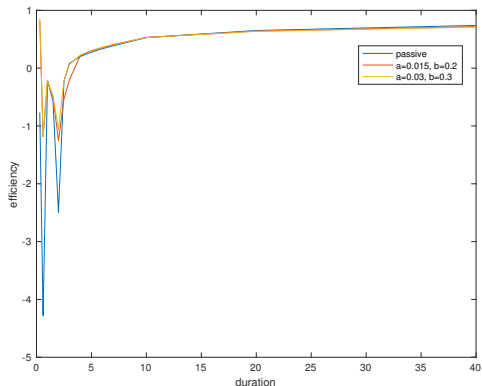
Phase Diagram



- 1 Slight improvement
- 2 how about long-time limit

Figure: Phase diagram of efficiency (duration = 10).

Efficiency



- 1 non-engine region: separated
- 2 engine region (efficiency > 0): without significantly difference

Figure: Efficiency as a function of duration.

Future Work

- 1 theoretically prove passive engine possesses higher efficiency than R-H one?
- 2 nonlinear effect of external field

Appendix A: Validation of Experimental Realization

- 1 In experiments, adiabatic process $\Rightarrow \Delta S = S_t - S_0 = 0$
 With slow change of protocols, utilizing statistical mechanics

$$\begin{aligned}\Delta S &= k \Delta \ln(Q \exp(\frac{T \partial_T Q}{Q})) = 0 \\ \Rightarrow Q \exp(\frac{T \partial_T Q}{Q}) &= \text{constant}\end{aligned}$$

- 2 Corresponding F-P equation \Rightarrow P.D.F.

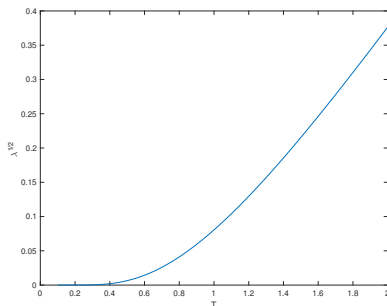
$$\begin{aligned}Q &= \iint_{\mathbb{R}^2} dx dv \exp(-\frac{b}{4\gamma kT}(\frac{m}{2}v^2 + \frac{\lambda}{2}x^2 - H_0)^2) \\ &= 2\pi^{3/2} \sqrt{\frac{\gamma kT}{m\lambda b}} (1 + \text{erf}(\frac{\mathcal{P}}{\sqrt{T}}))\end{aligned}$$

with $\mathcal{P} \stackrel{d}{=} \frac{H_0}{2} \sqrt{\frac{b}{\gamma k}}$

Validation

$$\sqrt{\lambda} = \mathcal{C} \frac{\sqrt{T}}{\mathcal{P}} \left(1 + \operatorname{erf}\left(\frac{\mathcal{P}}{\sqrt{T}}\right)\right) \exp\left(\frac{1}{2} - \frac{\mathcal{P}}{2\sqrt{T}} \frac{\exp(-\frac{\mathcal{P}^2}{T})}{1 + \operatorname{erf}(\frac{\mathcal{P}}{\sqrt{T}})}\right)$$

with $\mathcal{C} \in \mathbb{R}$



Appendix B: Data Analysis by Large Deviation Technique

- 1 central limit theorem \Rightarrow for X_i i.i.d. $\mu = 0, \sigma = 1 \quad \forall i = 1, 2, \dots, n$

$$p_n(l) \stackrel{d}{=} \mathbb{P}\left[\sum_{i=1}^n X_i/\sqrt{n} \geq \sqrt{nl}\right] \xrightarrow{n \rightarrow \infty} \frac{1}{\sqrt{\pi}} \int_{\sqrt{nl}}^{\infty} e^{-x^2} dx$$

$$\sim e^{-nl^2 + o(n)} \sim e^{-nI(l) + o(n)}$$

\Rightarrow define rate function (estimator) as $I(l) \stackrel{d}{=} \lim_{n \rightarrow \infty} -\frac{1}{n} \ln p_n(l)$

- 2 define empirical P.D.F. as

$$p_{em}^{(n)}(x) dx = \sum_{i=1}^N \mathbb{I}(X_i^{(n)} \in [x, x + dx]) / N$$

$$I(\eta) \stackrel{d}{=} \lim_{n \rightarrow \infty} -\frac{1}{n\tau} \ln p_{em}^{(n)}(\eta)$$

$$\Rightarrow \frac{1}{n\tau} \ln p_{em}^{(n)}(\eta) = -I(\eta) + \frac{\mathcal{A}}{n\tau}$$

plot $\frac{1}{n\tau} \ln p_{em}^{(n)}(\eta)$ vs $\frac{1}{n\tau}$ to get estimator

$$\mathbb{P}[\{X_i\}_{i=1}^n = x] \sim \exp(-n\tau I(x))$$