

Martingales and Gambling in stochastic thermodynamics

Édgar Roldán

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Universita' di Padova, 22.10.21

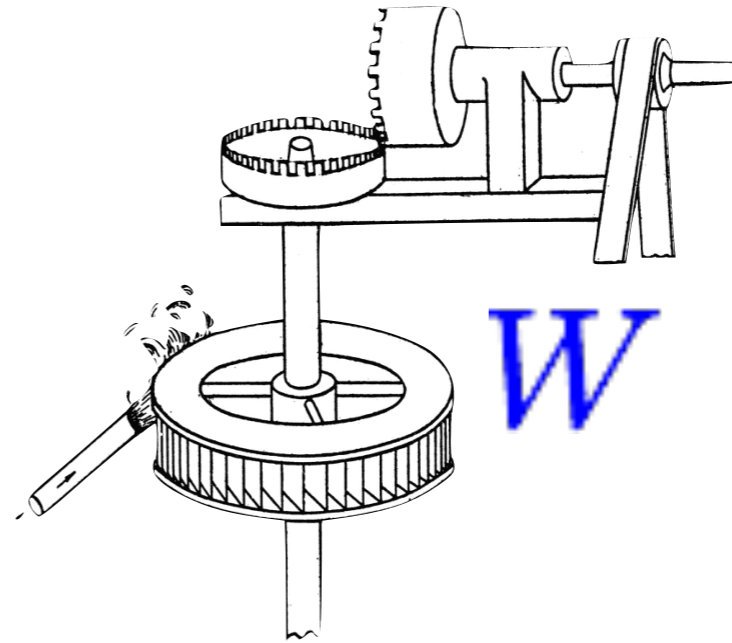
**Second Law and
stochastic entropy production
for small systems**

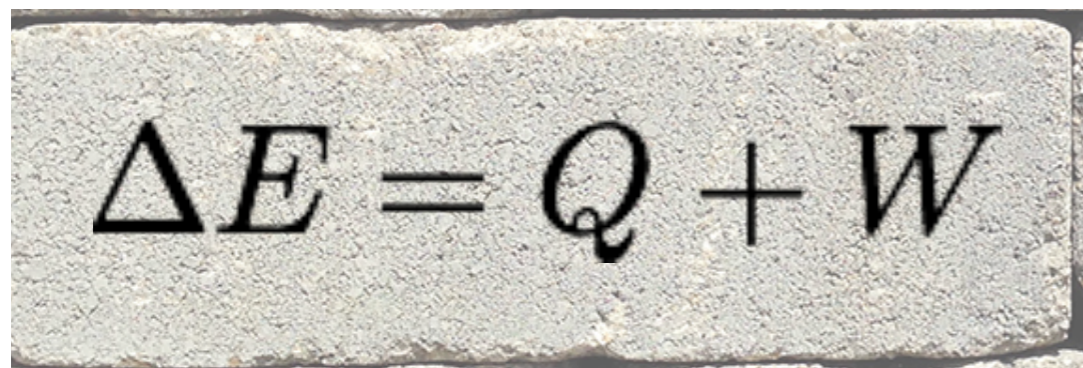
First Law of Thermodynamics

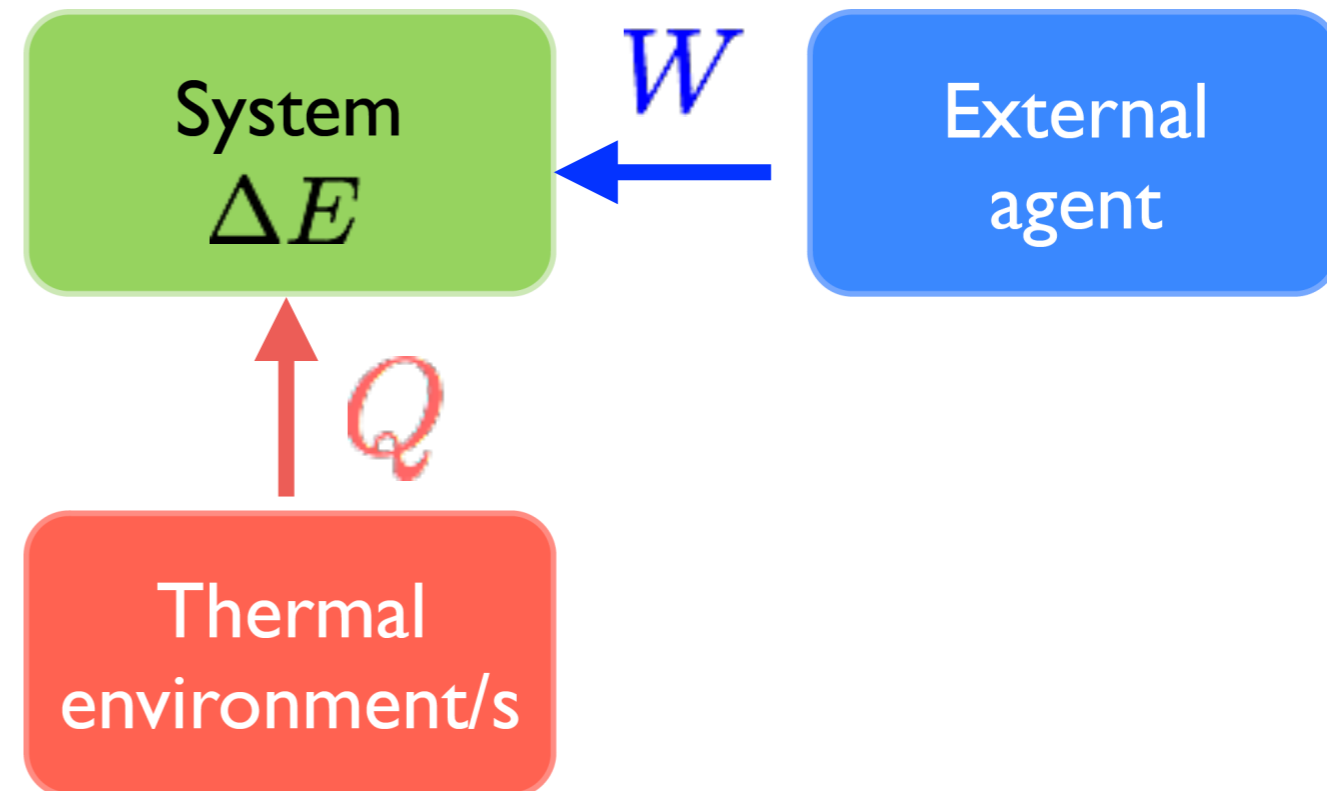
Heat



Work

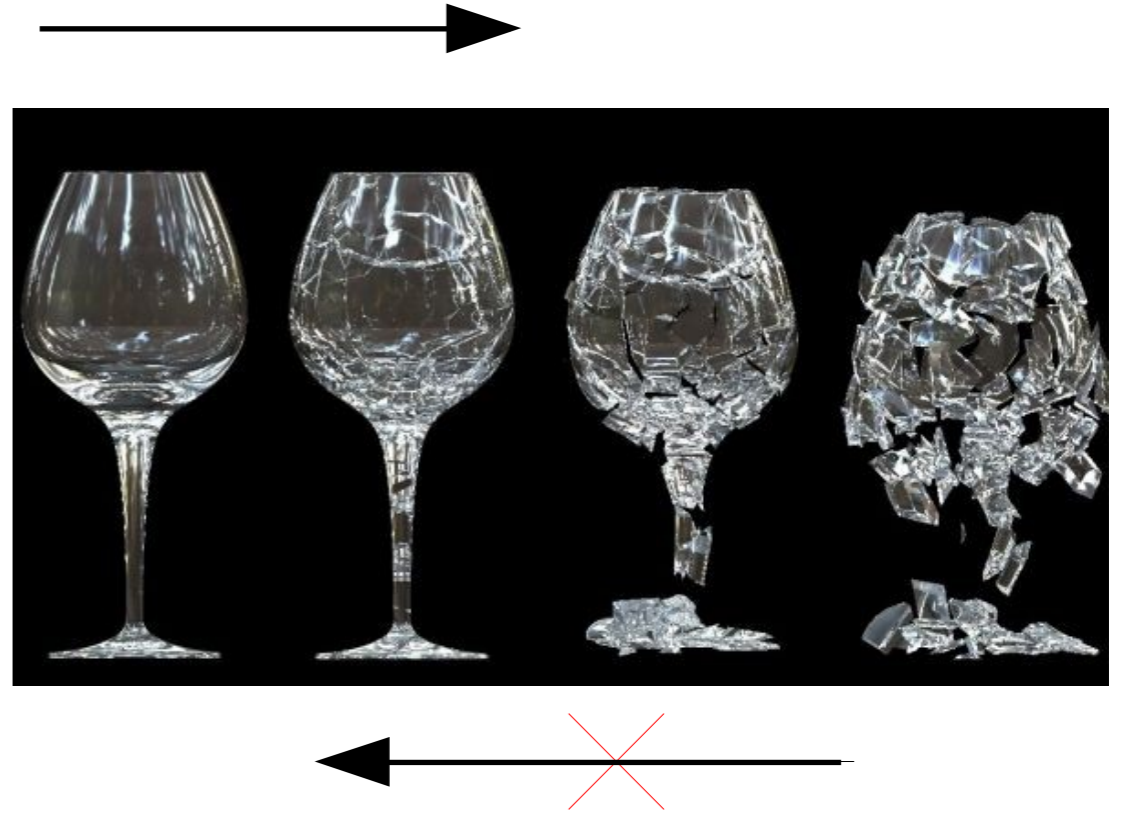



$$\Delta E = Q + W$$



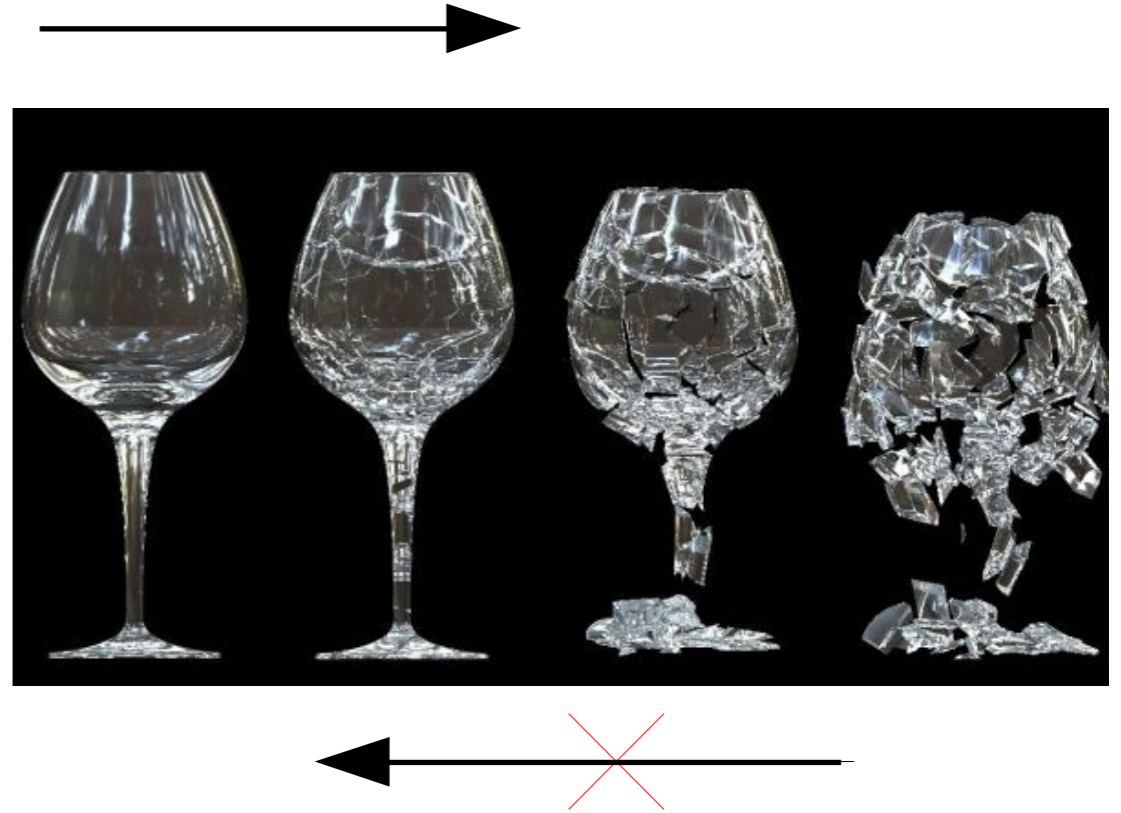
Second Law

Second Law of Thermodynamics and the arrow of time



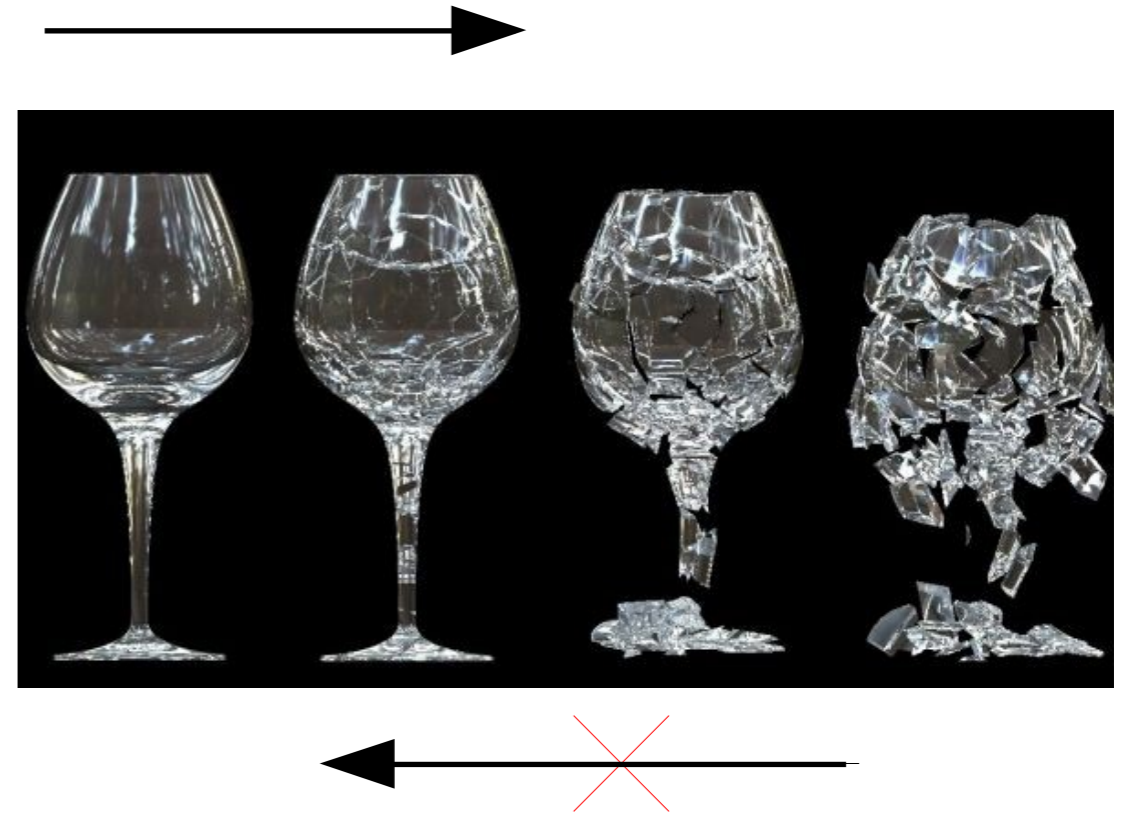
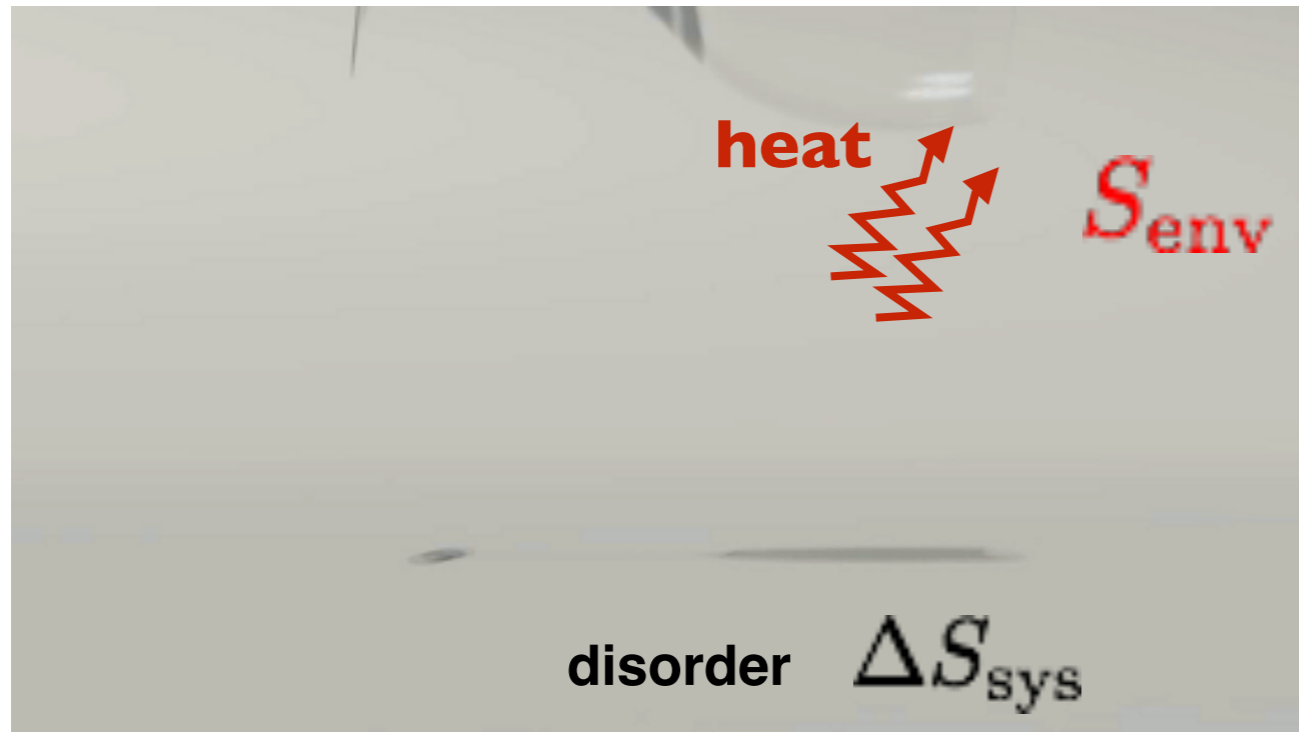
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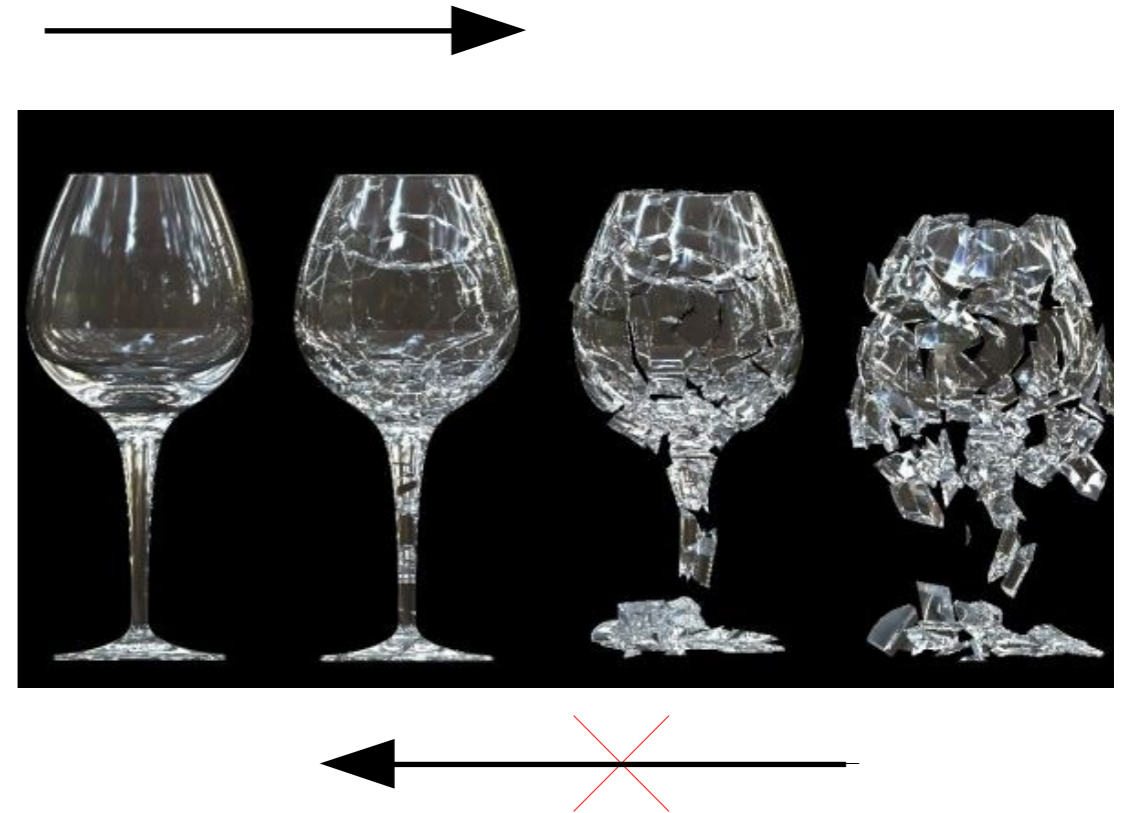
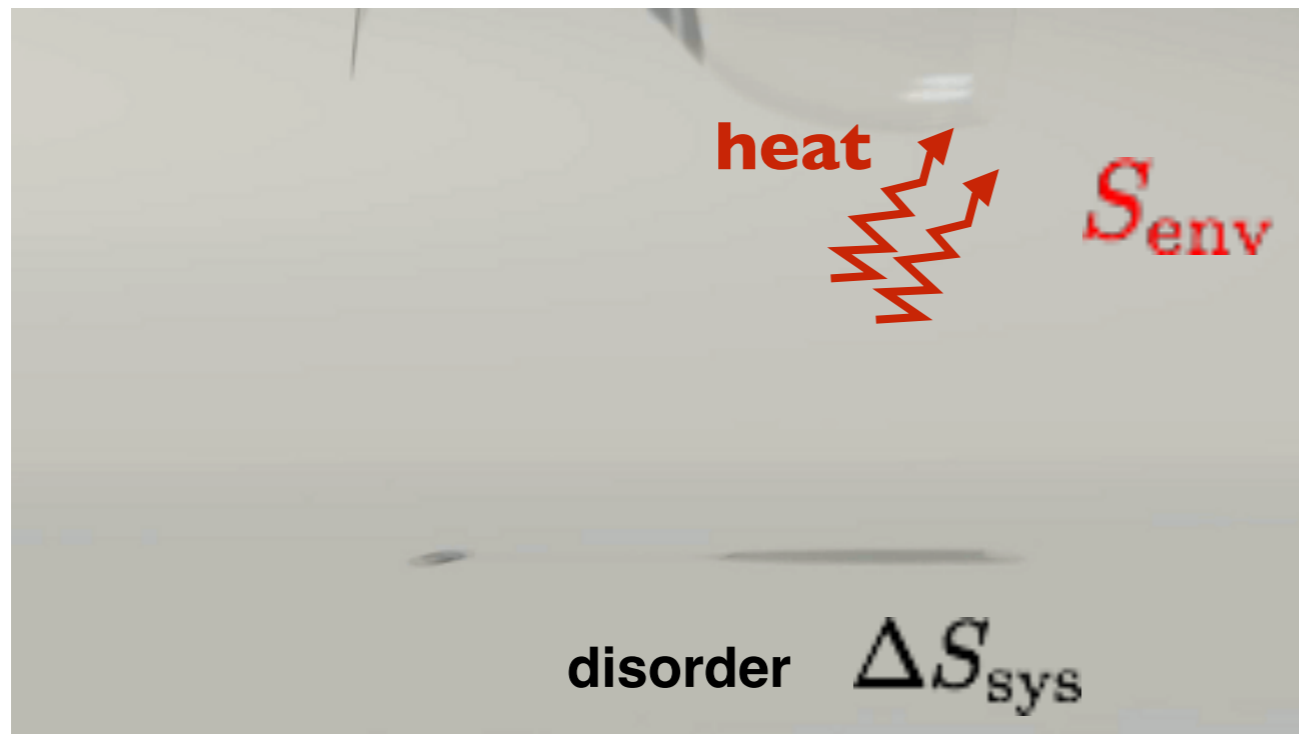
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Second Law

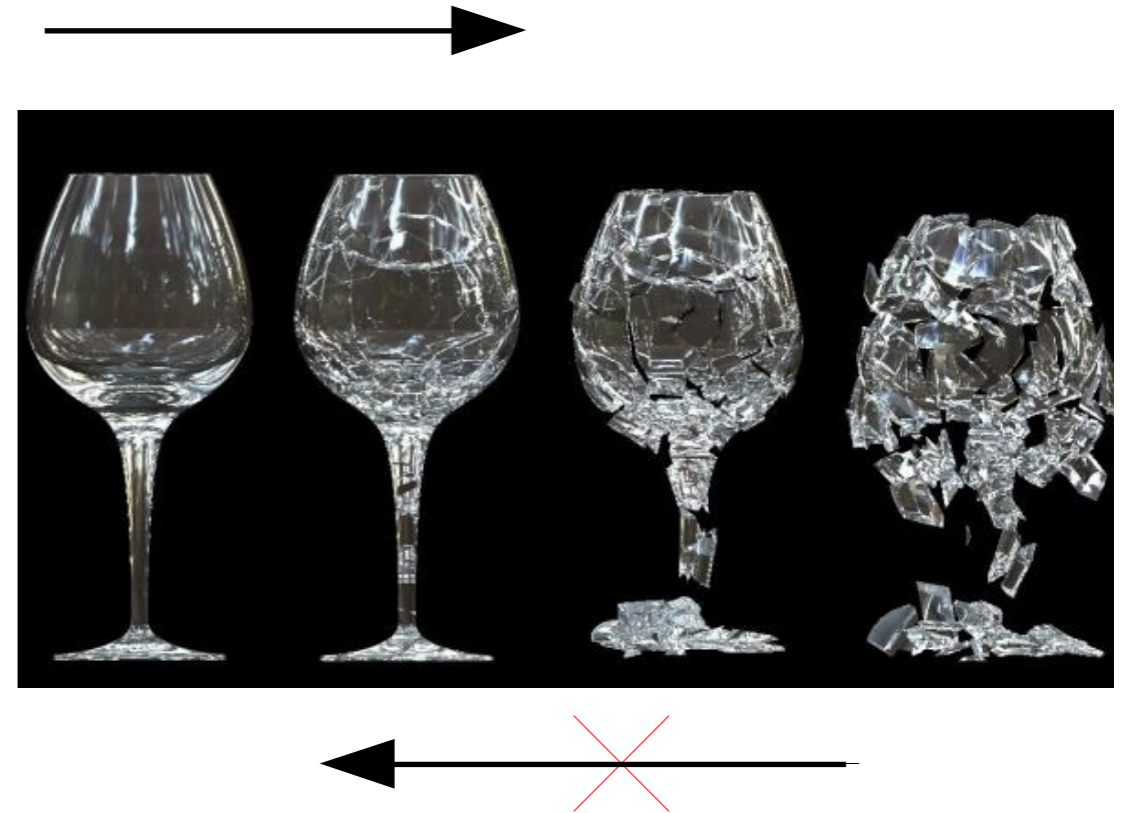
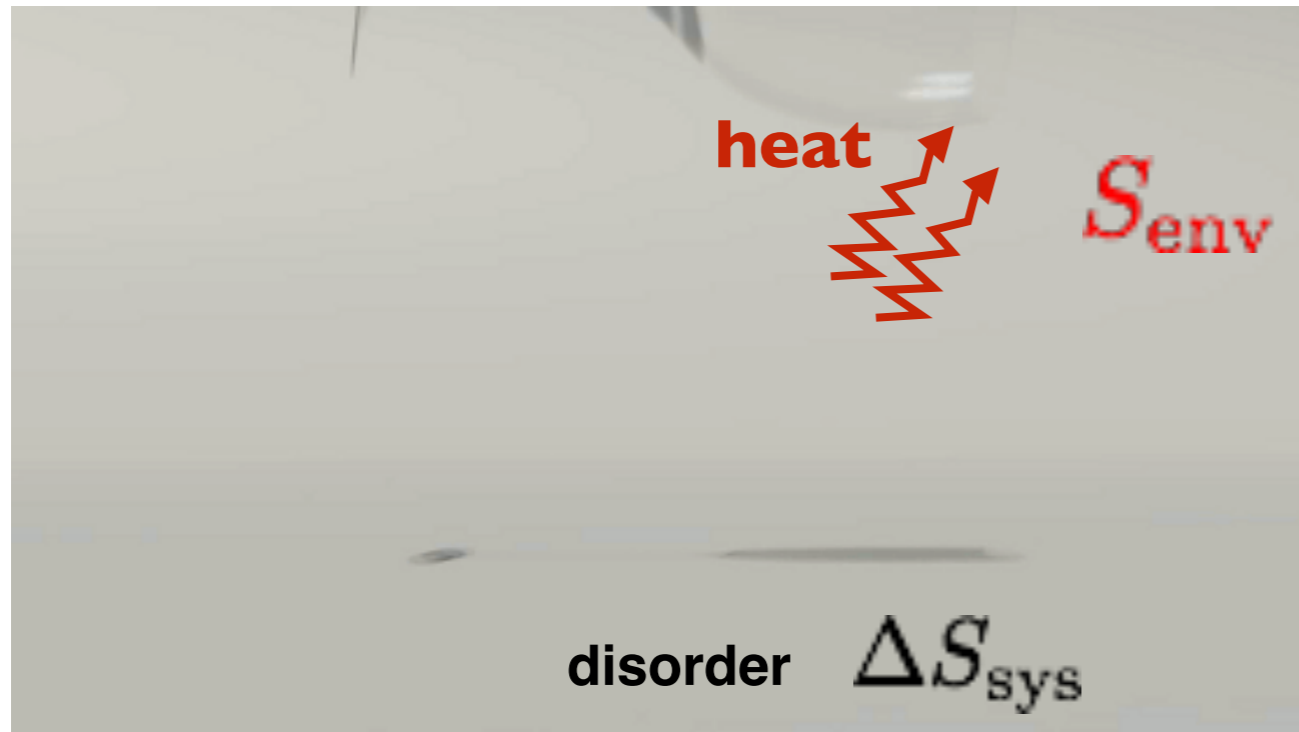
Second Law of Thermodynamics and the arrow of time



$$S(t) = S_{env}(t) + \Delta S_{sys}(t) \geq 0$$

Second Law

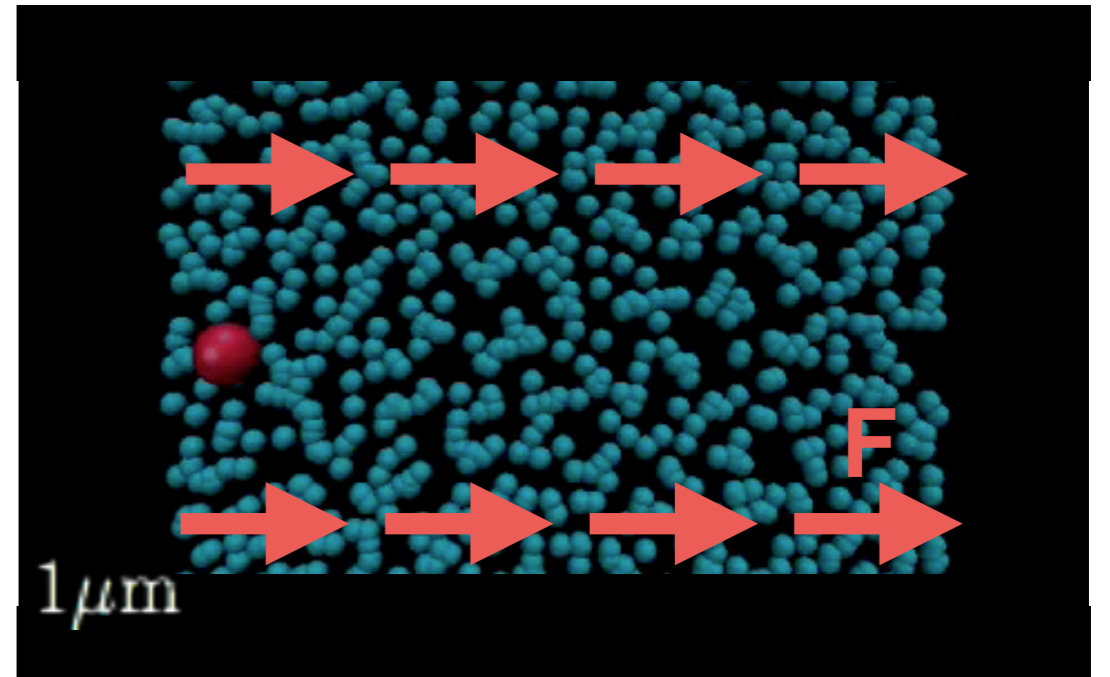
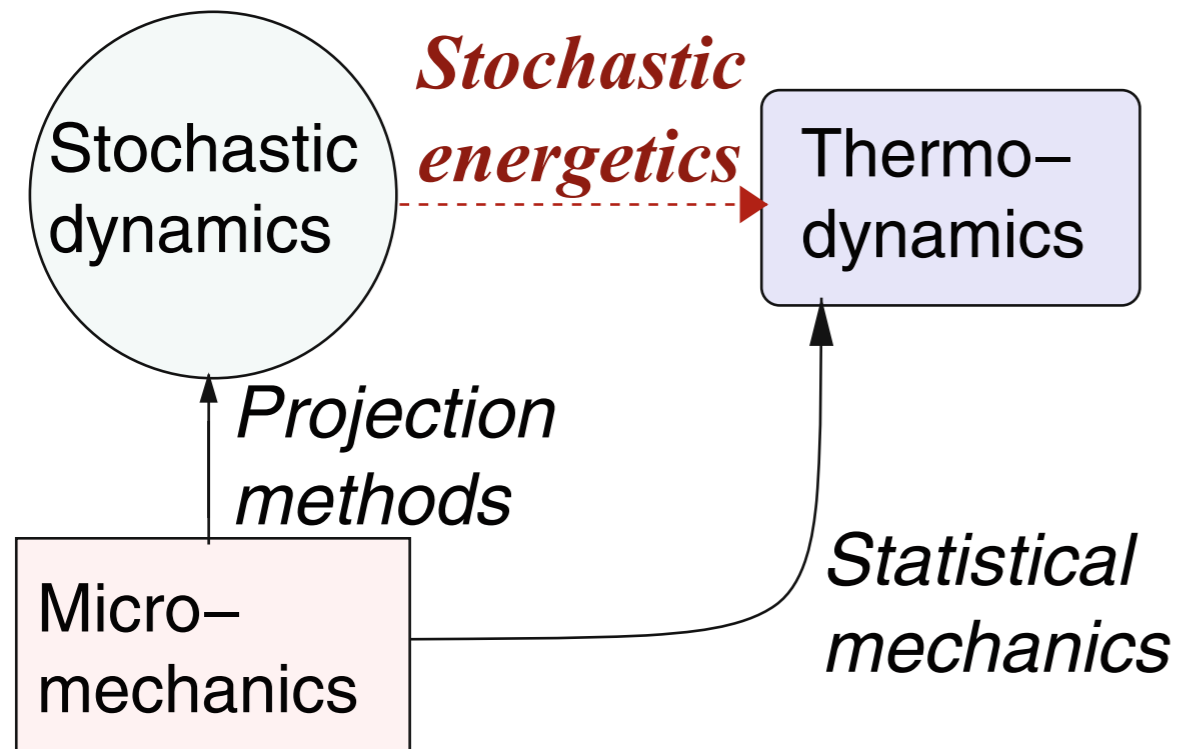
Second Law of Thermodynamics and the arrow of time



$$S(t) = S_{env}(t) + \Delta S_{sys}(t) \geq 0$$

You can't even break even

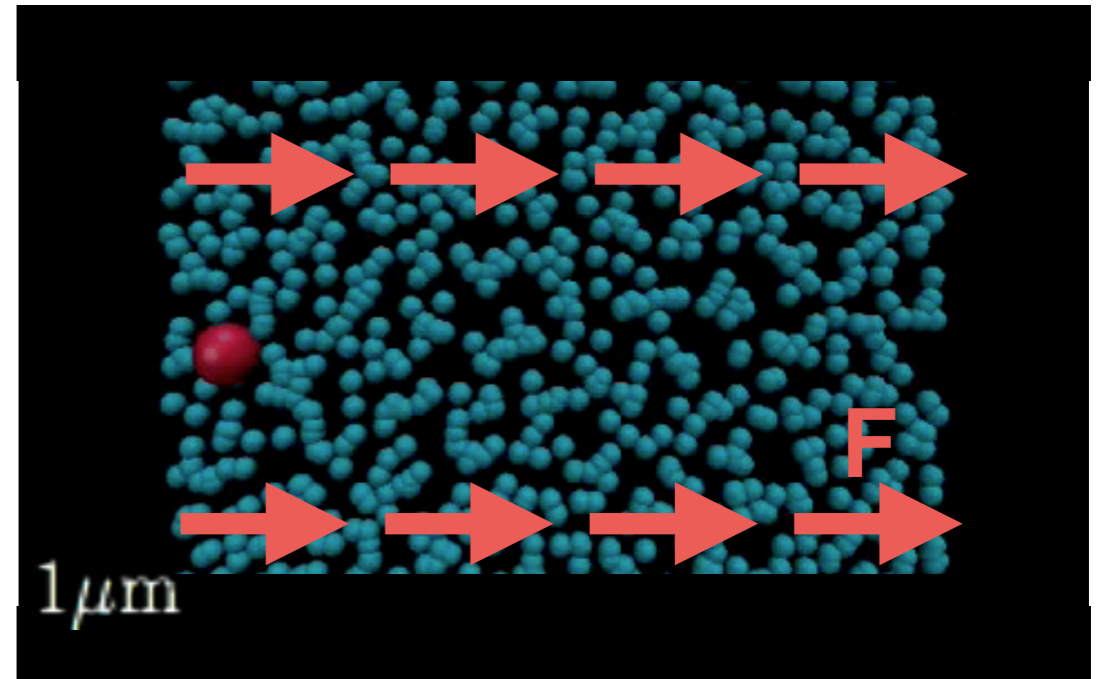
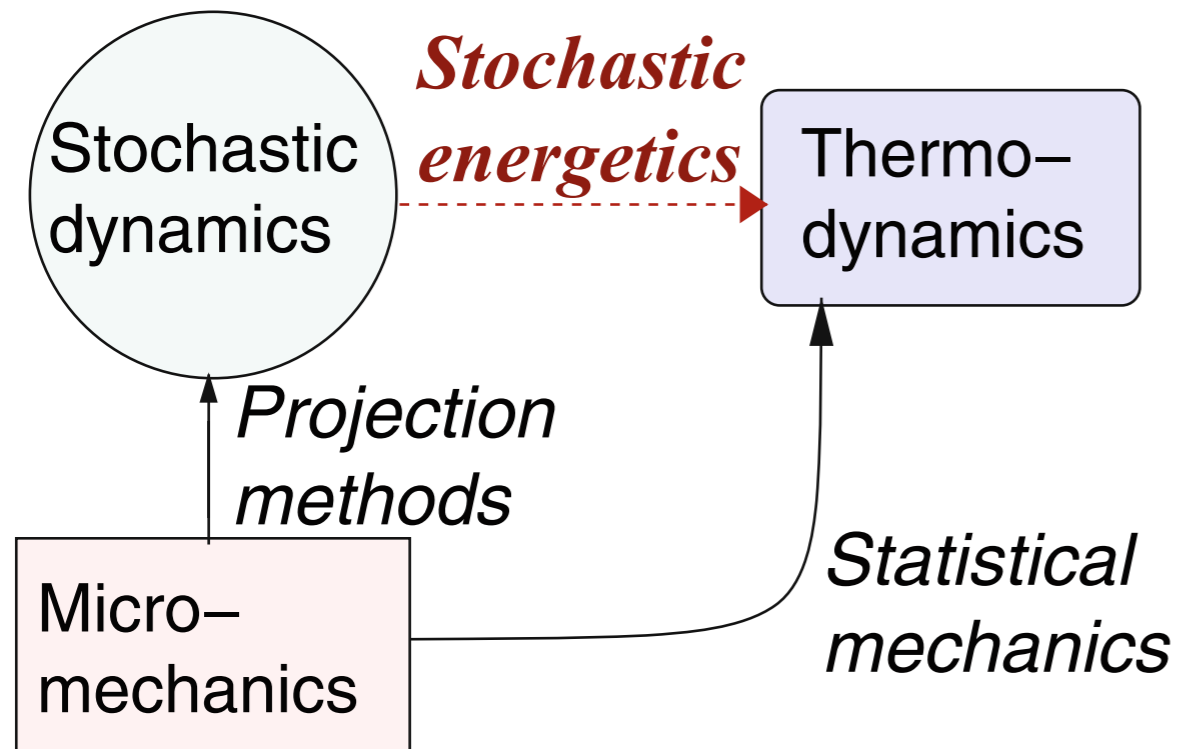
Stochastic thermodynamics



First Law: $\Delta U(t) = Q(t) + W(t)$

Second Law: $\langle S_{\text{tot}}(t) \rangle \geq 0$

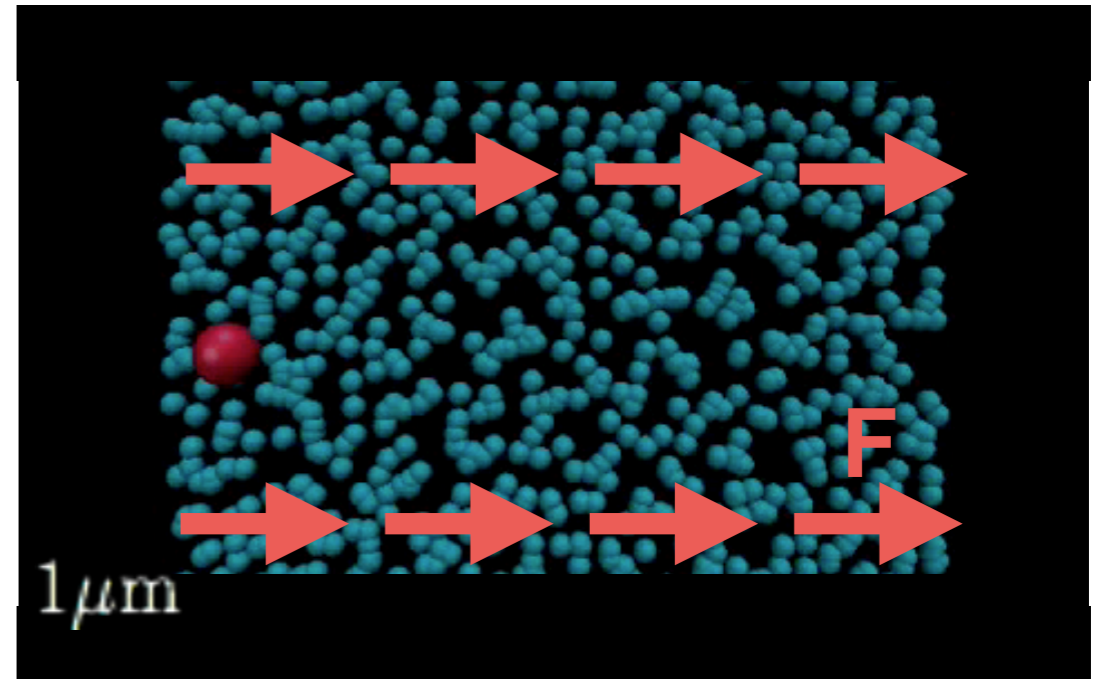
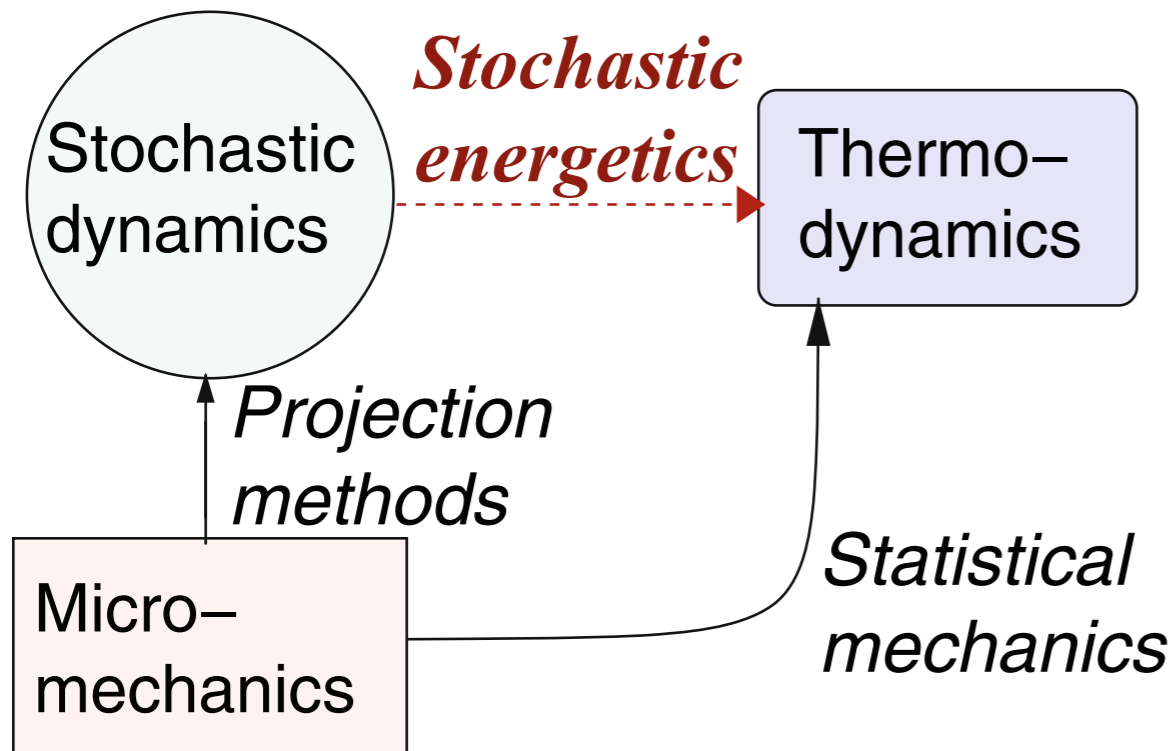
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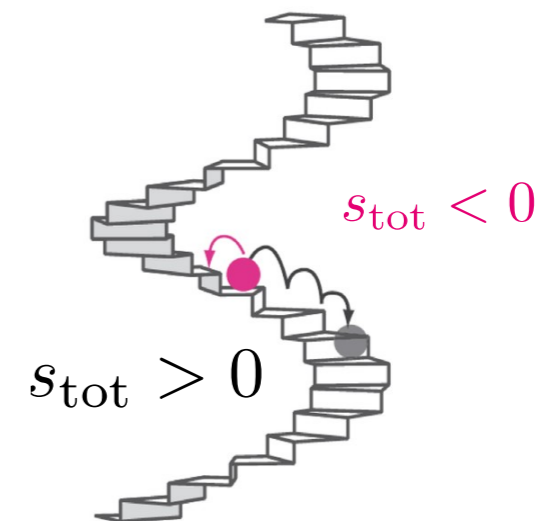
Second Law: $\langle S_{\text{tot}}(t) \rangle \geq 0$

Stochastic thermodynamics

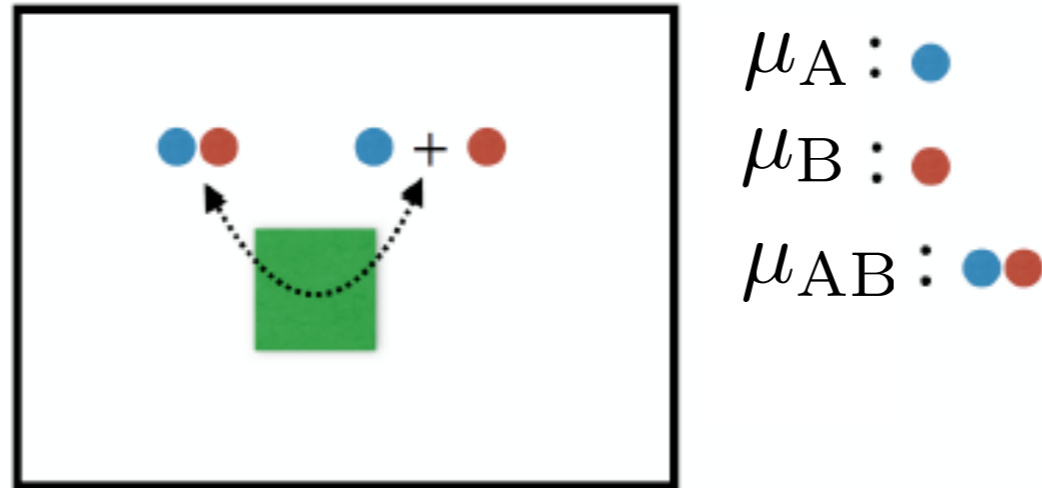


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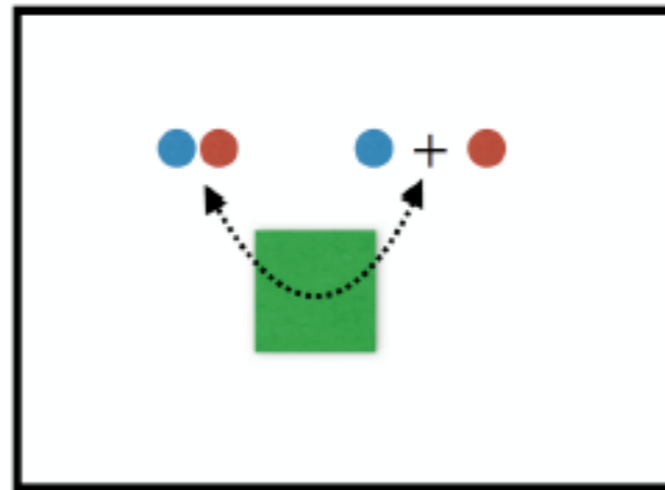


A typical example



Thermodynamics: $J(t) (\mu_{AB} - \mu_A - \mu_B) \geq 0$

A typical example



μ_A : ●

μ_B : ●

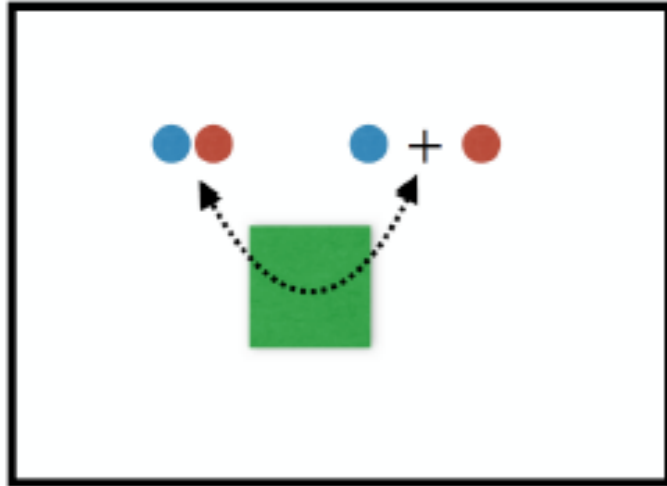
μ_{AB} : ●●

Thermodynamics: $J(t) (\mu_{AB} - \mu_A - \mu_B) \geq 0$

Stochastic
thermodynamics: $\frac{k_{\bullet+\bullet \rightarrow \bullet\bullet}}{k_{\bullet\bullet \rightarrow \bullet+\bullet}} = e^{S_{\text{env}}(\bullet+\bullet \rightarrow \bullet\bullet)} = e^{\frac{\mu_{\bullet\bullet} - \mu_{\bullet} - \mu_{\bullet}}{T_{\text{env}}}}$

Stochastic entropy production

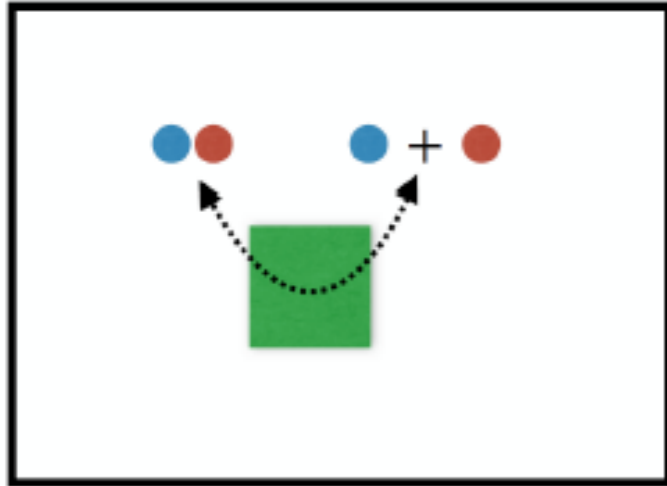
Local detailed balance



$$\frac{P(x(1), x(2), \dots, x(t)|x(0))}{P(x(t-1), x(t-2), \dots, x(0)|x(t))} = e^{S_{\text{env}}(t)/k_B}$$

Stochastic entropy production

Local detailed balance



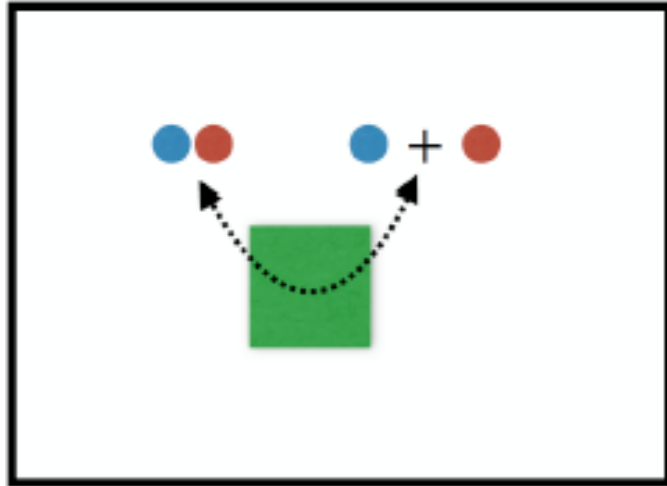
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Nonequilibrium system entropy

$$S_{\text{sys}}(t) \equiv -k_B \ln P_{\text{st}}(x(t))$$

Stochastic entropy production

Local detailed balance



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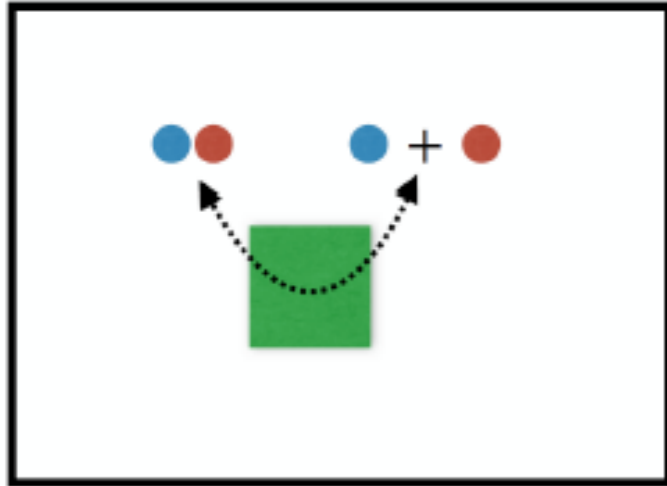
$$S_{\text{sys}}(t) \equiv -k_B \ln P_{\text{st}}(x(t))$$

Stochastic entropy production

$$S_{\text{tot}}(t) = \Delta S_{\text{sys}}(t) + S_{\text{env}}(t) = k_B \ln \frac{P(x(0), x(1), x(2), \dots, x(t))}{P(x(t), x(t-1), x(t-2), \dots, x(0))}$$

Stochastic entropy production

Local detailed balance



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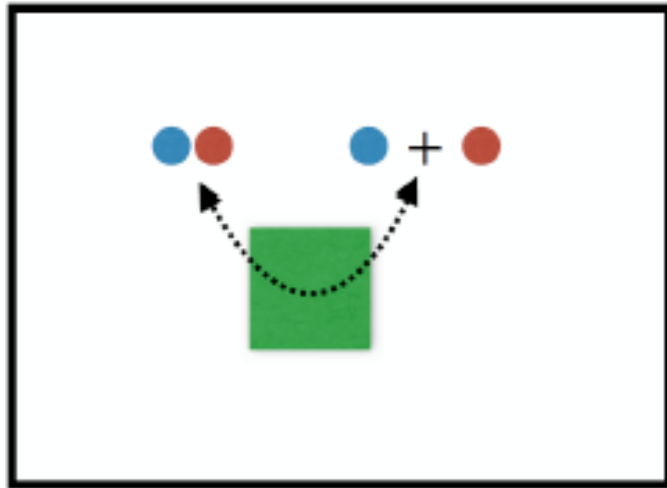
$$S_{\text{tot}}(t) = k_B \ln \frac{P(X_{[0,t]})}{P(\Theta X_{[0,t]})}$$

Time-reversed measure

$$\tilde{P}(X_{[0,t]}) \equiv P(\Theta X_{[0,t]})$$

Stochastic entropy production

Local detailed balance



$$\frac{P(x(1), x(2), \dots, x(t)|x(0))}{P(x(t-1), x(t-2), \dots, x(0)|x(t))} = e^{S_{\text{env}}(t)/k_B}$$

Nonequilibrium system entropy

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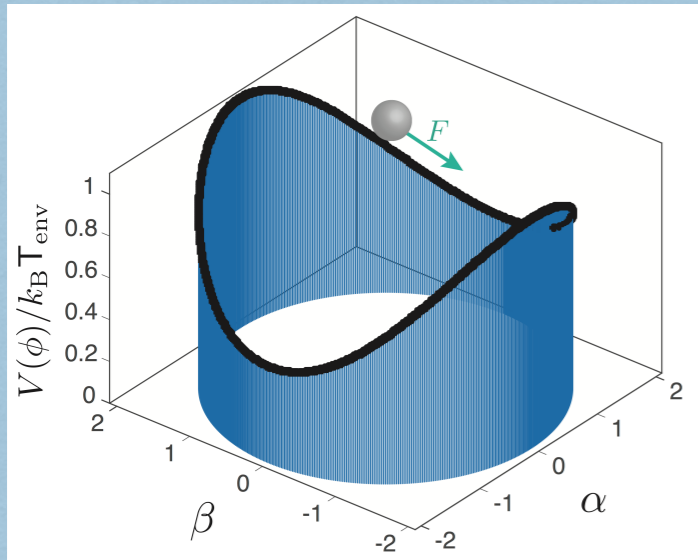
Stochastic entropy production

$$S_{\text{tot}}(t) = \Delta S_{\text{sys}}(t) + S_{\text{env}}(t) = k_B \ln \frac{P(x(0), x(1), x(2), \dots, x(t))}{P(x(t), x(t-1), x(t-2), \dots, x(0))}$$

Second law of stochastic thermodynamics

$$\langle S_{\text{tot}}(t) \rangle = \int dx(0) \cdots \int dx(t) P(X_{[0,t]}) \ln \frac{P(X_{[0,t]})}{\tilde{P}(X_{[0,t]})} = D[P(X_{[0,t]}) || \tilde{P}(X_{[0,t]})] \geq 0$$

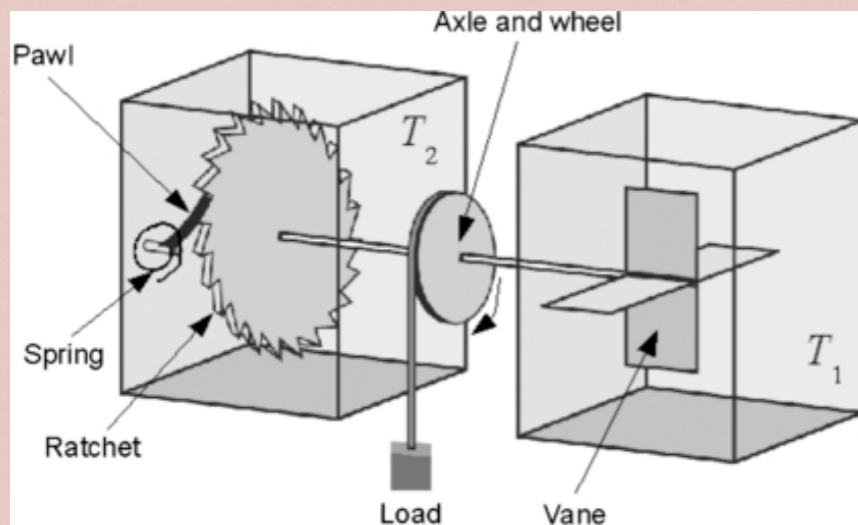
Examples



Isothermal non-equilibrium steady states

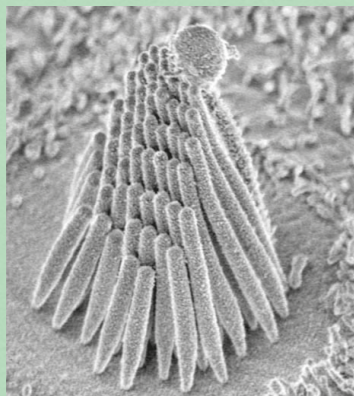
$$S_{\text{tot}}(t) = \frac{W(t) - \Delta F(t)}{T}$$

Noneq. free energy $P_{\text{st}}(\mathbf{x}) = e^{-\beta(U(\mathbf{x}) - F(\mathbf{x}))}$



Non-isothermal steady states
(e.g. autonomous heat engines)

$$S_{\text{tot}}(t) = \Delta S_{\text{sys}}(t) - \frac{Q_h(t)}{T_h} - \frac{Q_c(t)}{T_c}$$



Active matter

systems with hidden nonequilibrium degrees of freedom

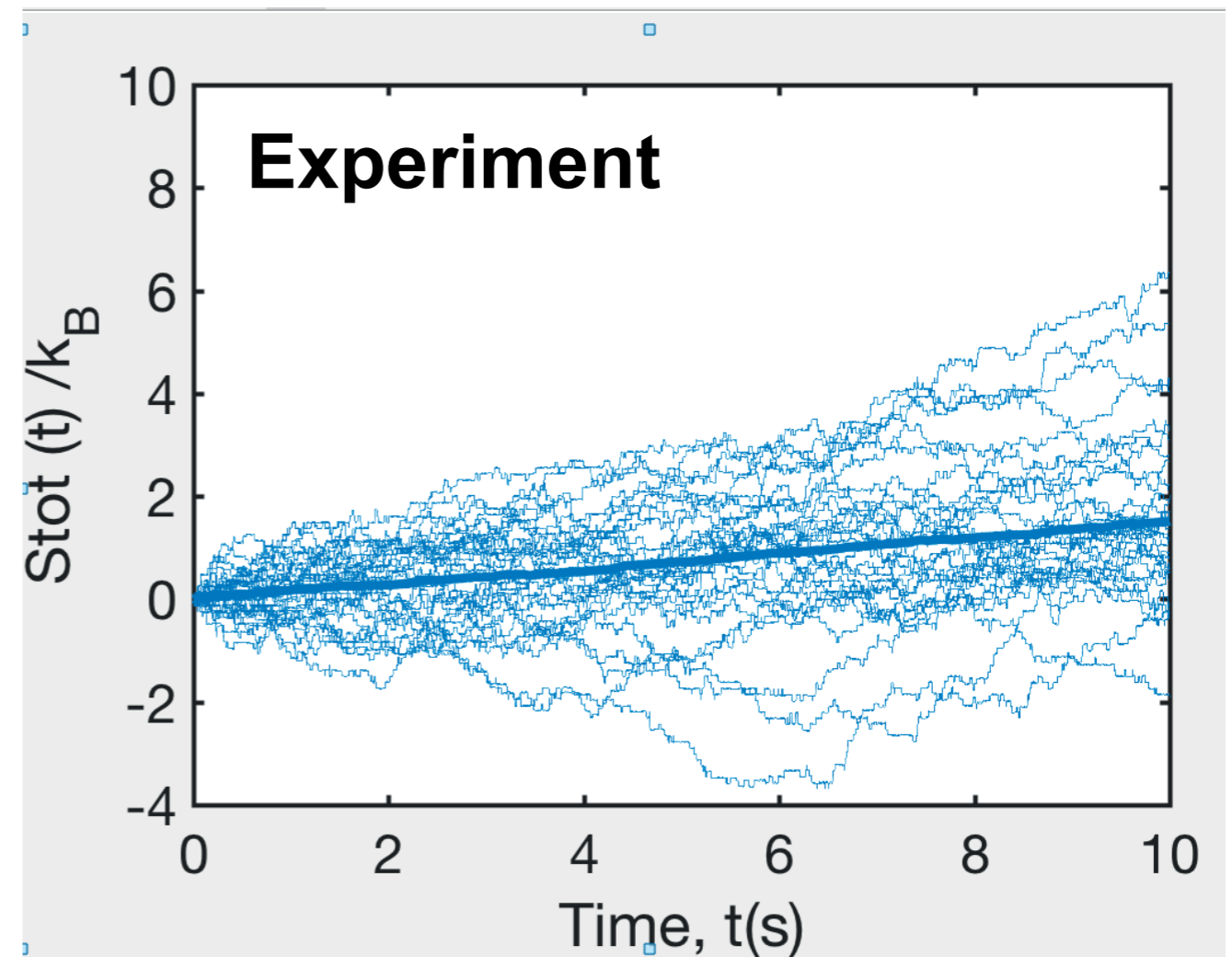
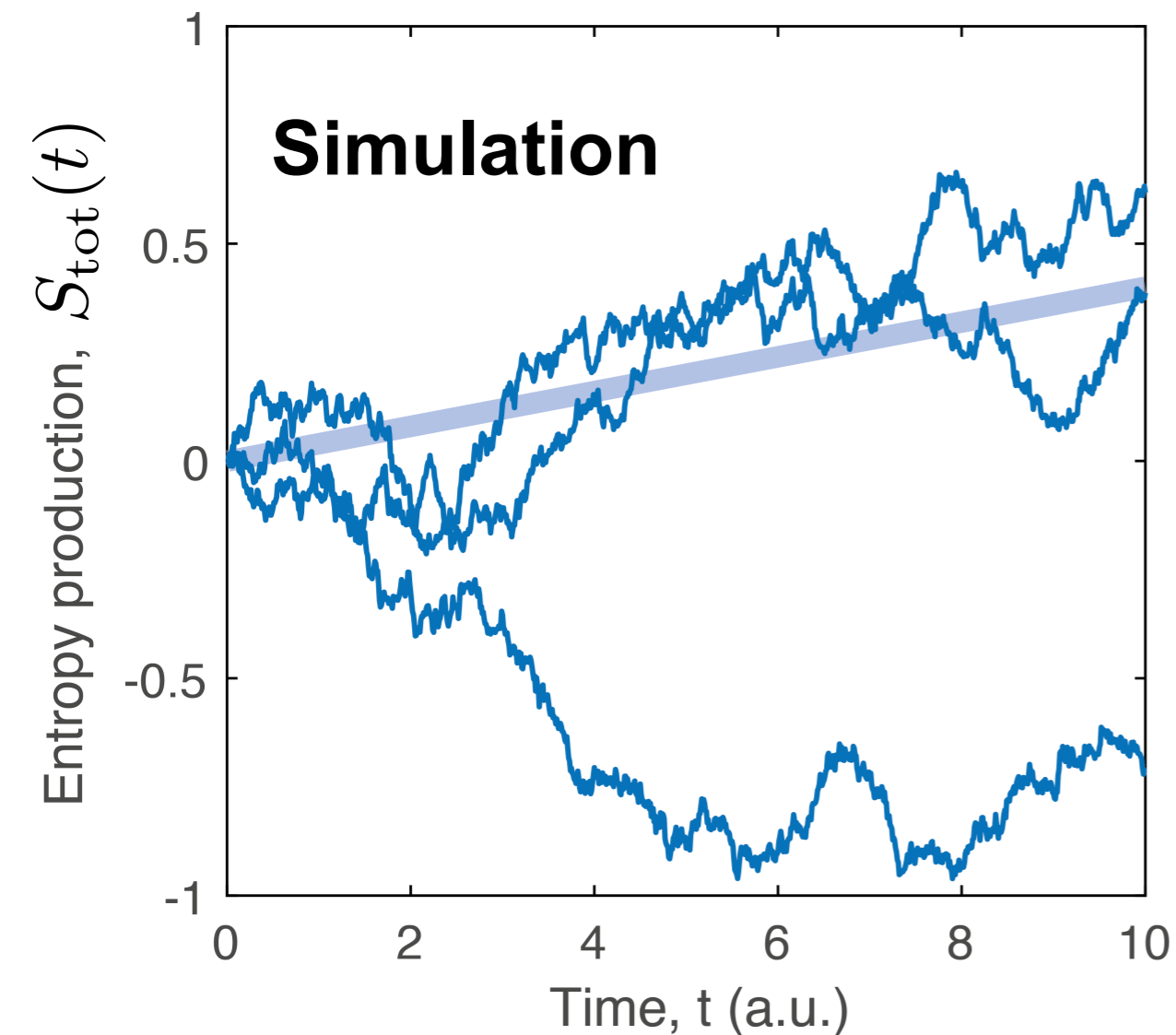
$$S_{\text{tot}}(t) = \text{Irreversibility}$$

$$-\frac{\langle Q(t) \rangle}{T} \geq \langle S_{\text{tot}}(t) \rangle = k_B D [P(X_{[0,t]}) || \tilde{P}(X_{[0,t]})]$$

Basic knowledge on stochastic entropy

Equilibrium : $S_{\text{tot}}(t) = 0 \Rightarrow \langle S_{\text{tot}}(t) \rangle = 0$

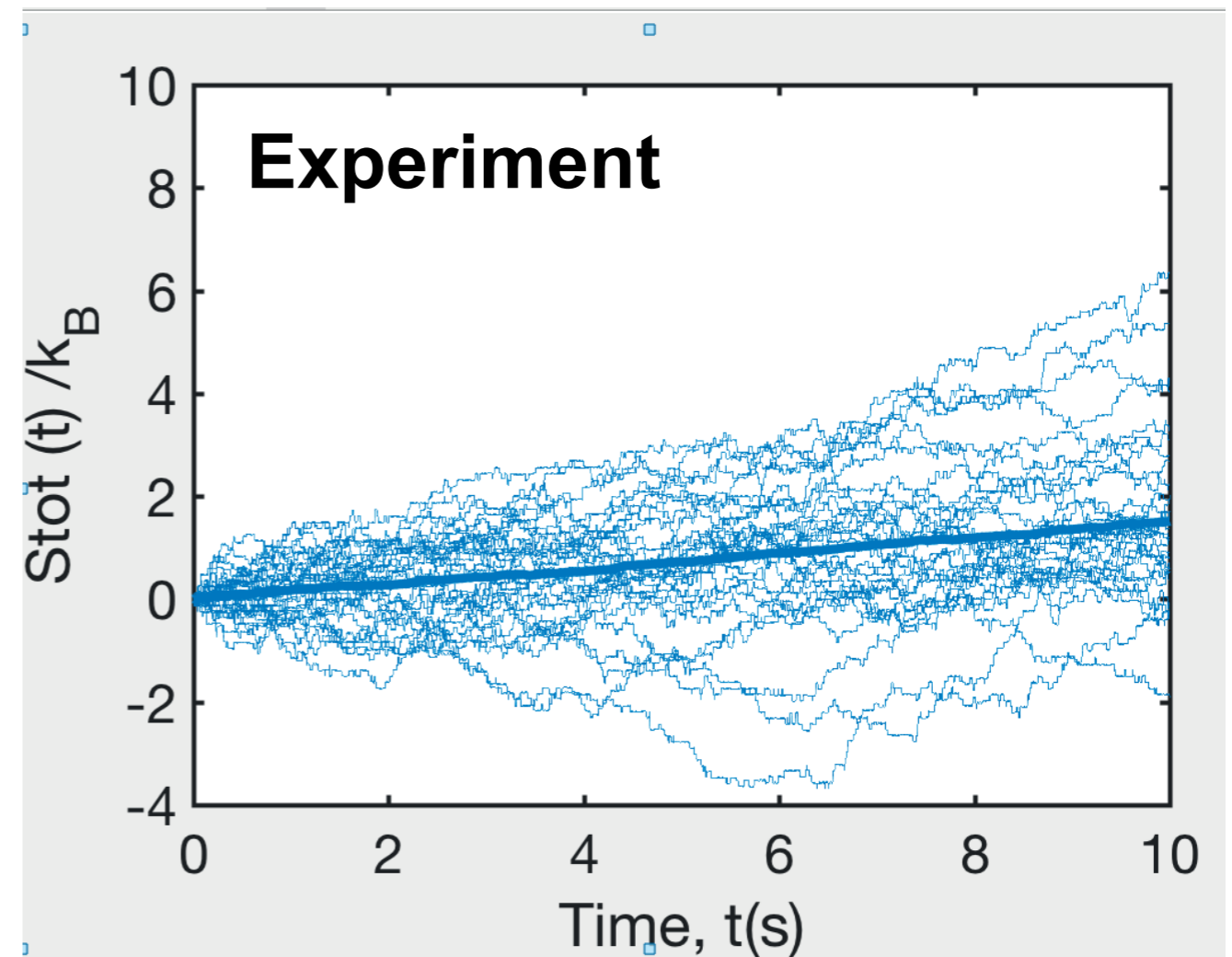
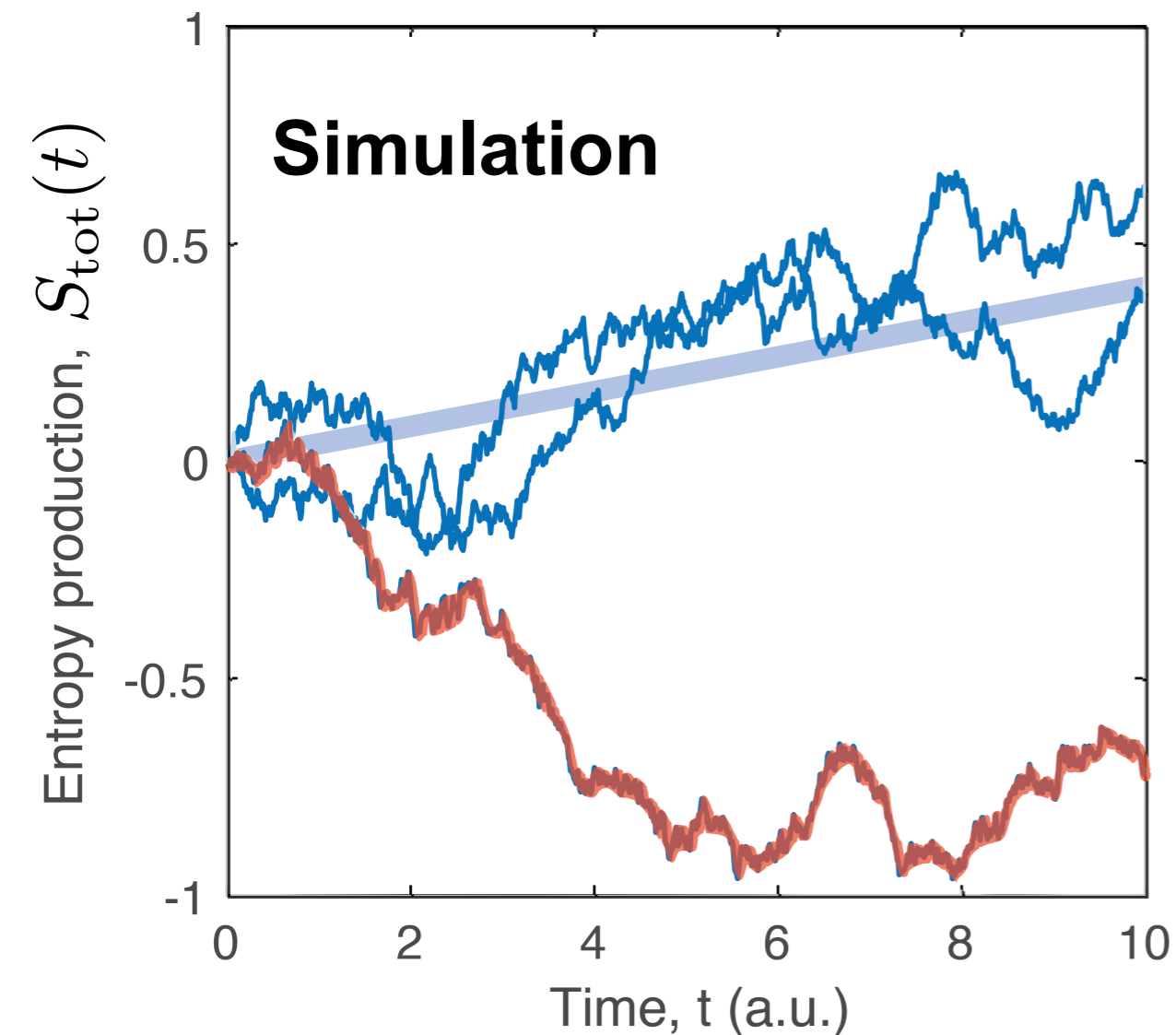
Nonequilibria: $\langle S_{\text{tot}}(t) \rangle \geq 0$



Basic knowledge on stochastic entropy

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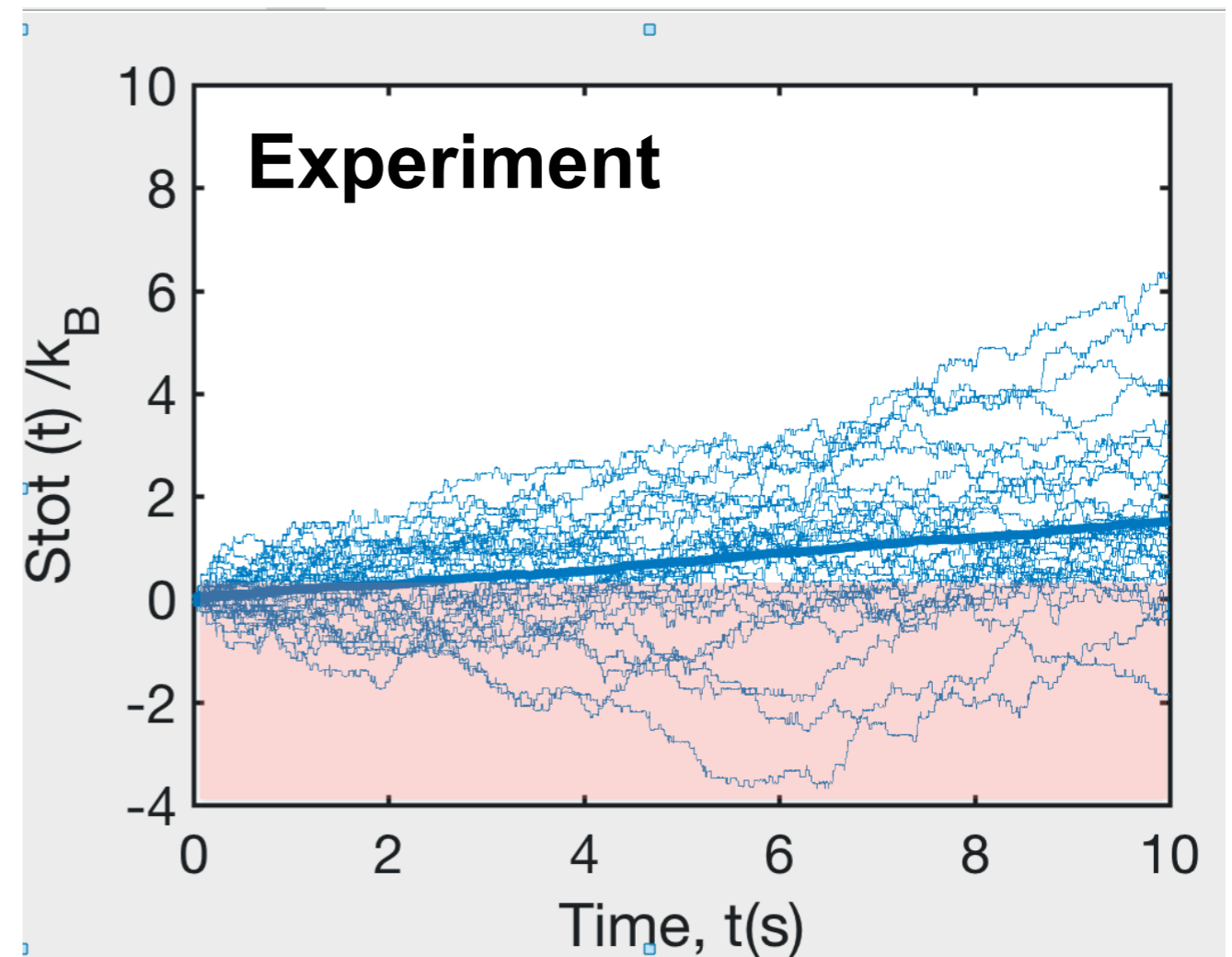
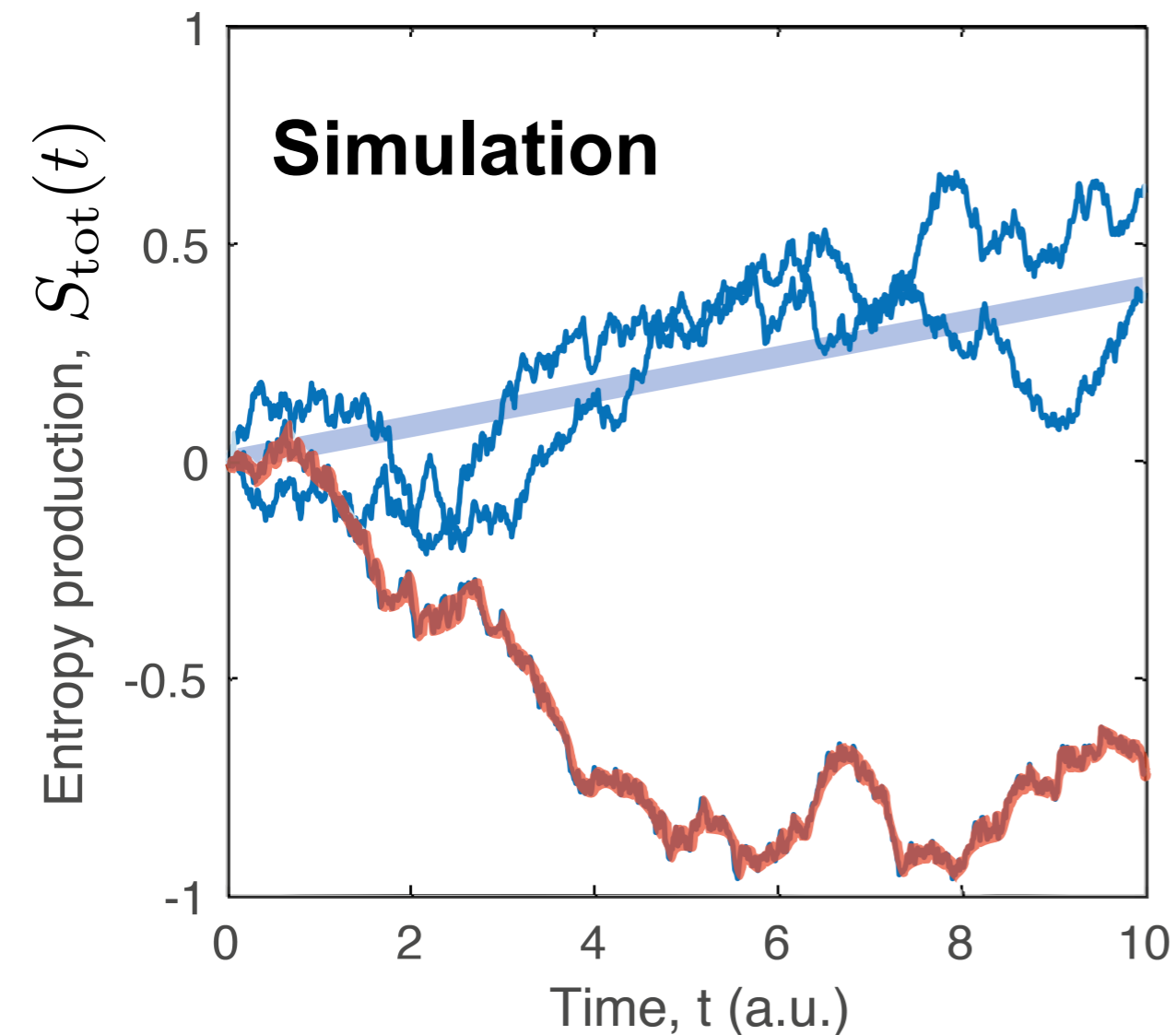
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Basic knowledge on stochastic entropy

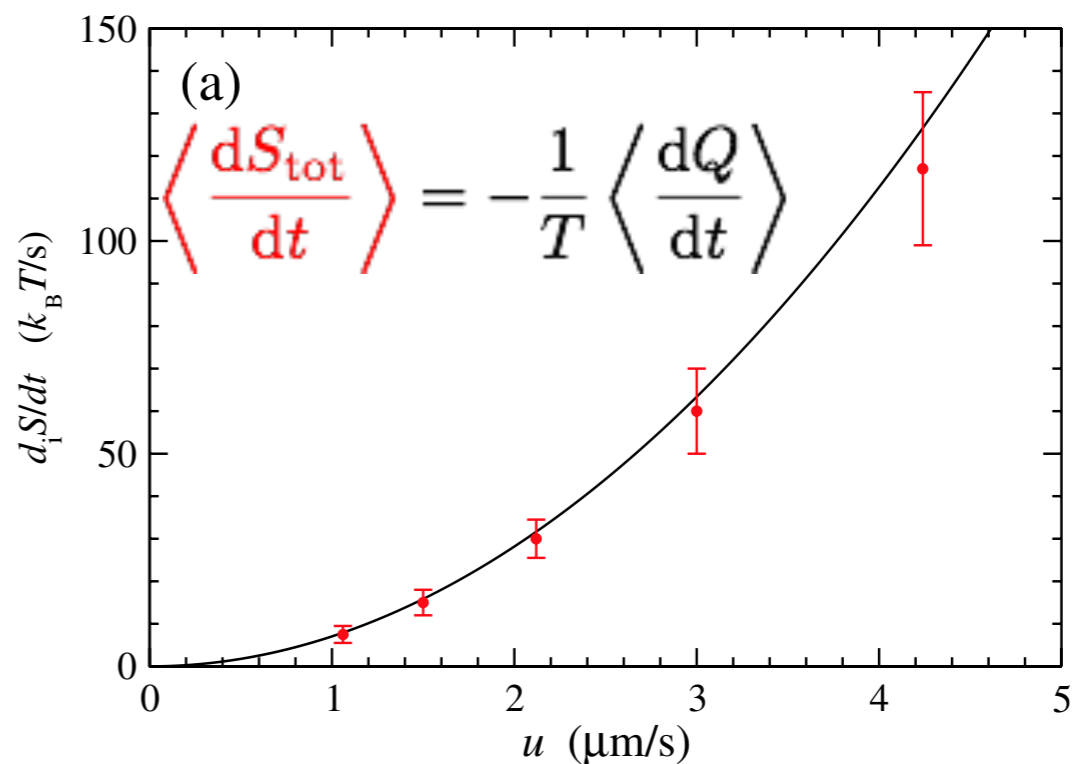
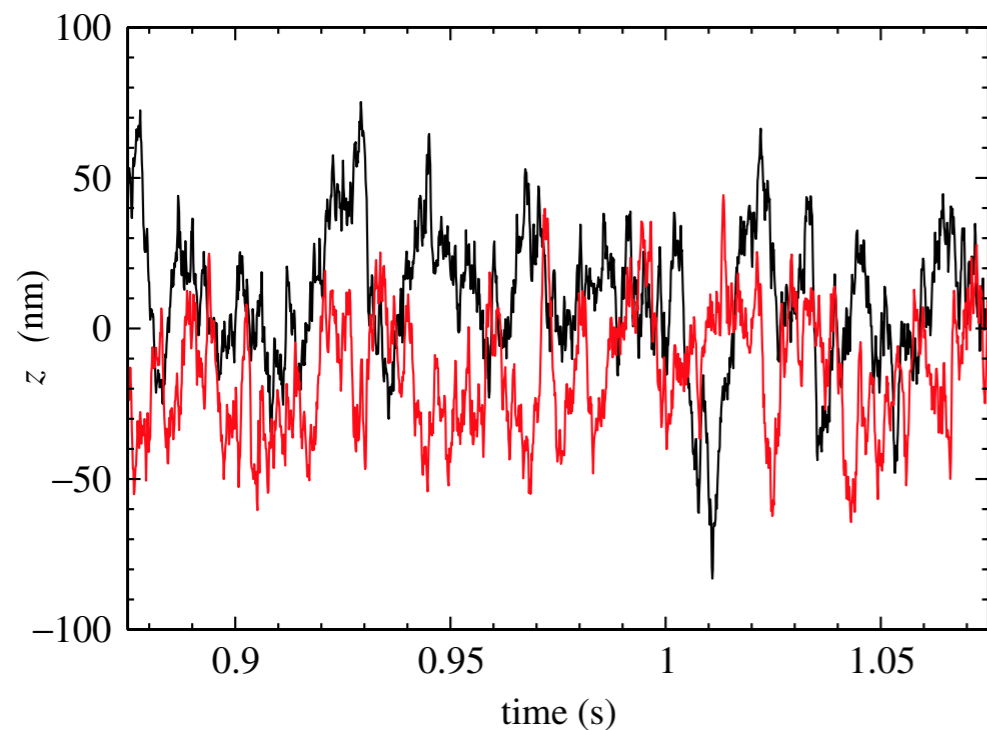
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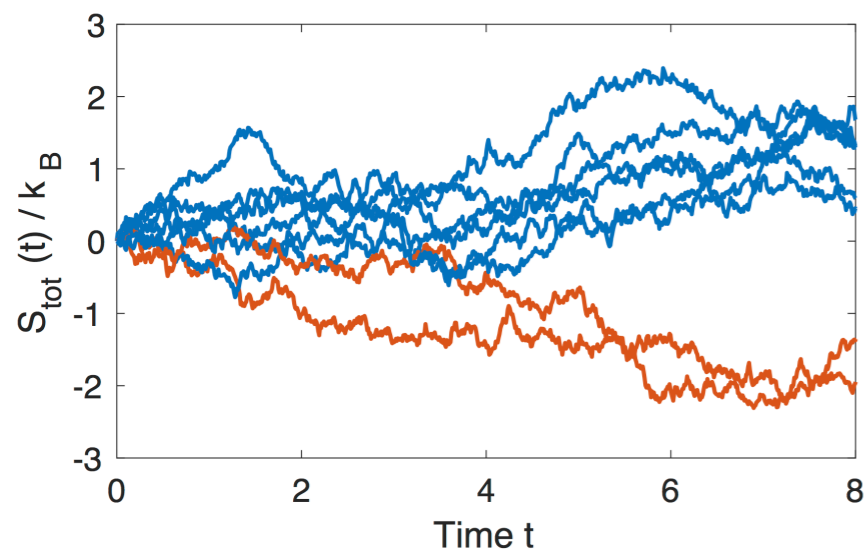


Fluctuation theorems

Andrieux, Gaspard, Ciliberto,
Garnier, Joubaud, Petrosyan, PRL 2007



Fluctuation theorems for stochastic entropy production



Detailed Fluctuation theorem

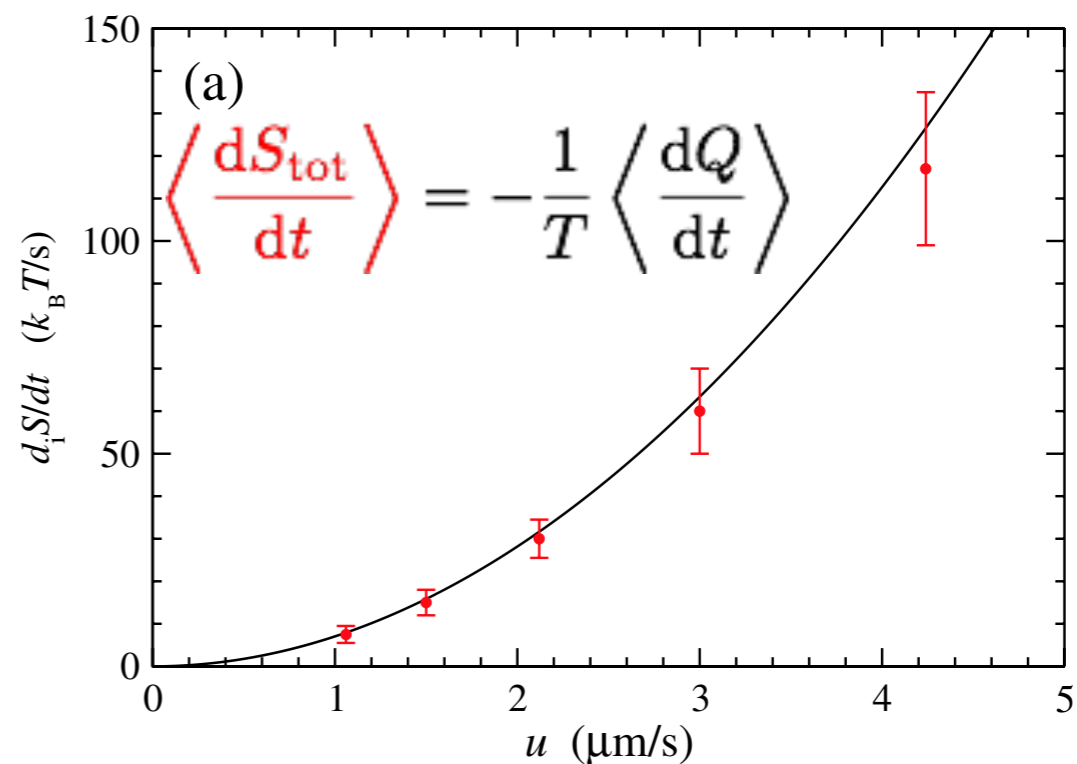
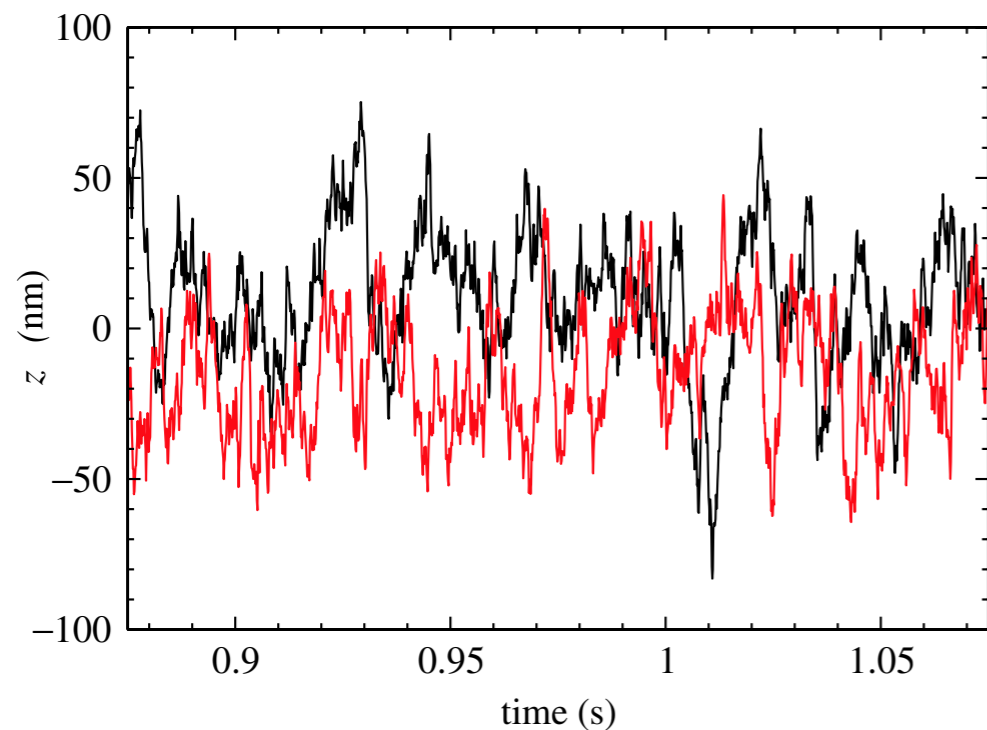
$$\frac{p_S(s; t)}{p_S(-s; t)} = e^{s/k_B}$$

Jarzynski's equality
(Integral FT)

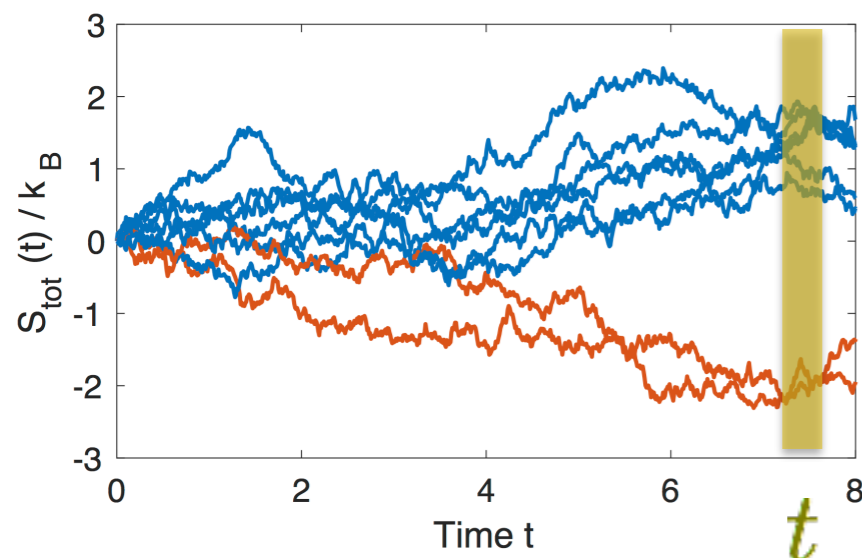
$$\langle e^{-S(t)/k_B} \rangle = 1$$

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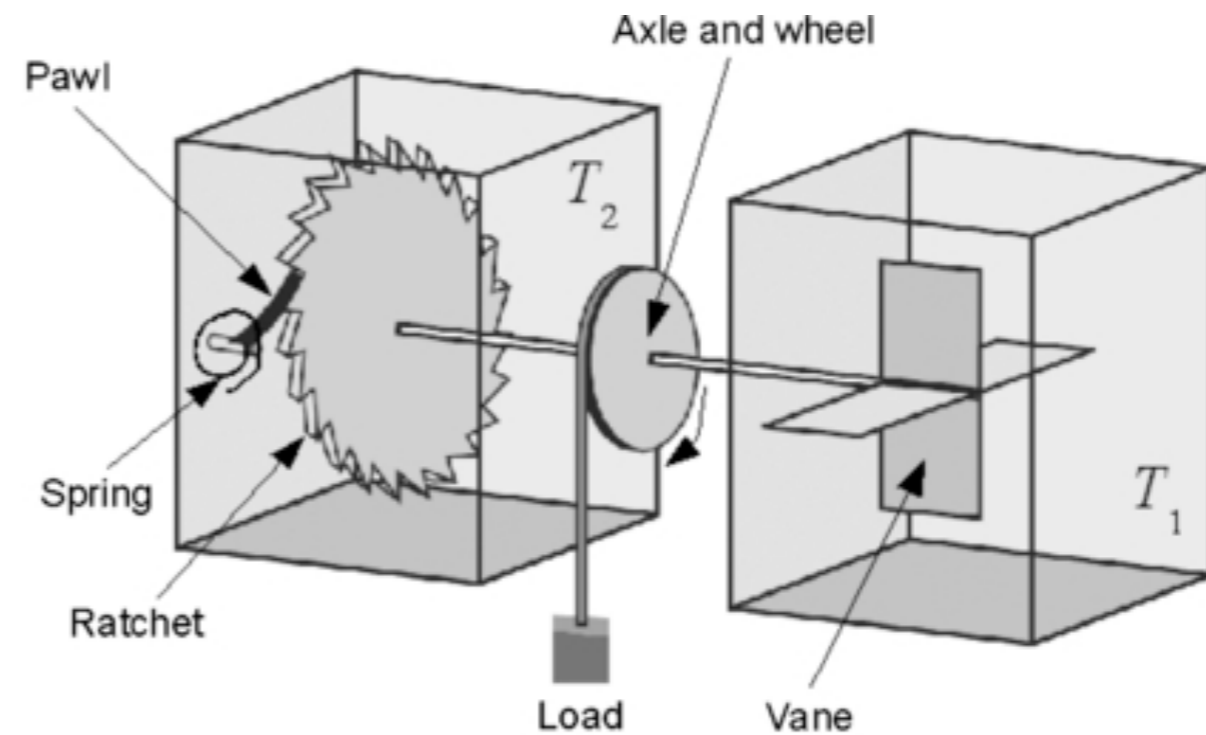
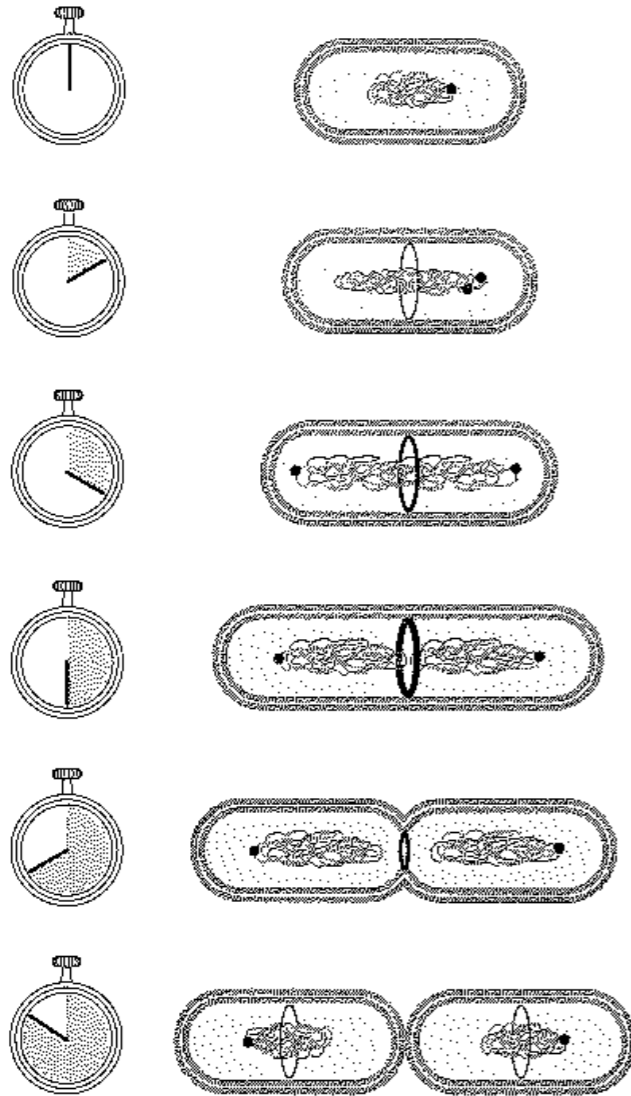
Fixed time properties

Why martingales?

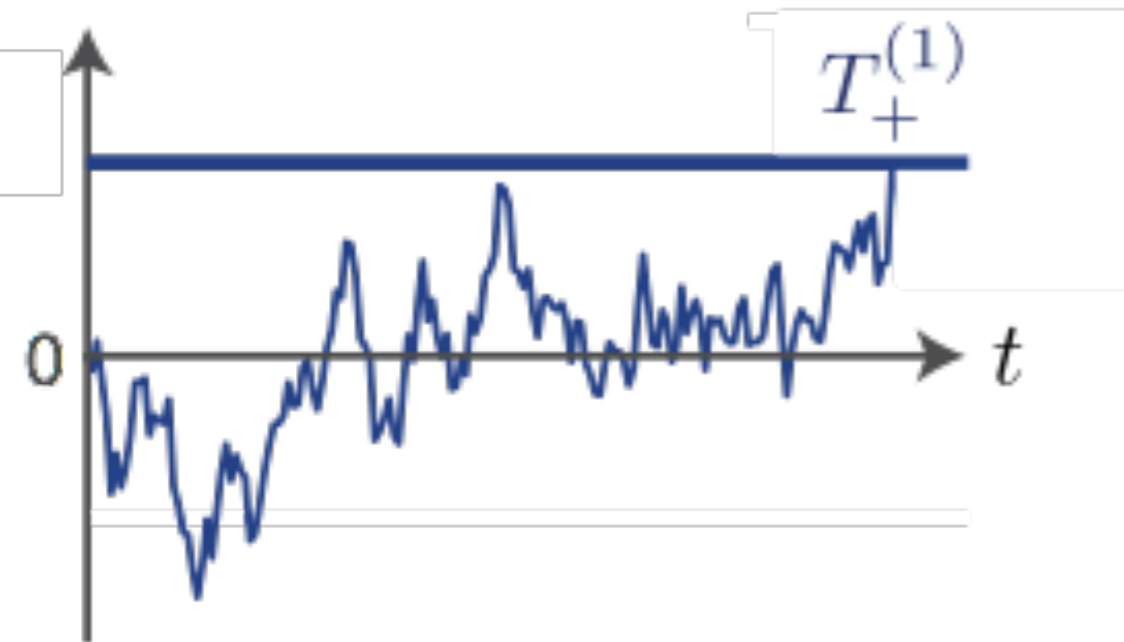
Most fluctuation theorems concern events that take place **at a fixed time**

However, interesting phenomena take place at **stochastic times**

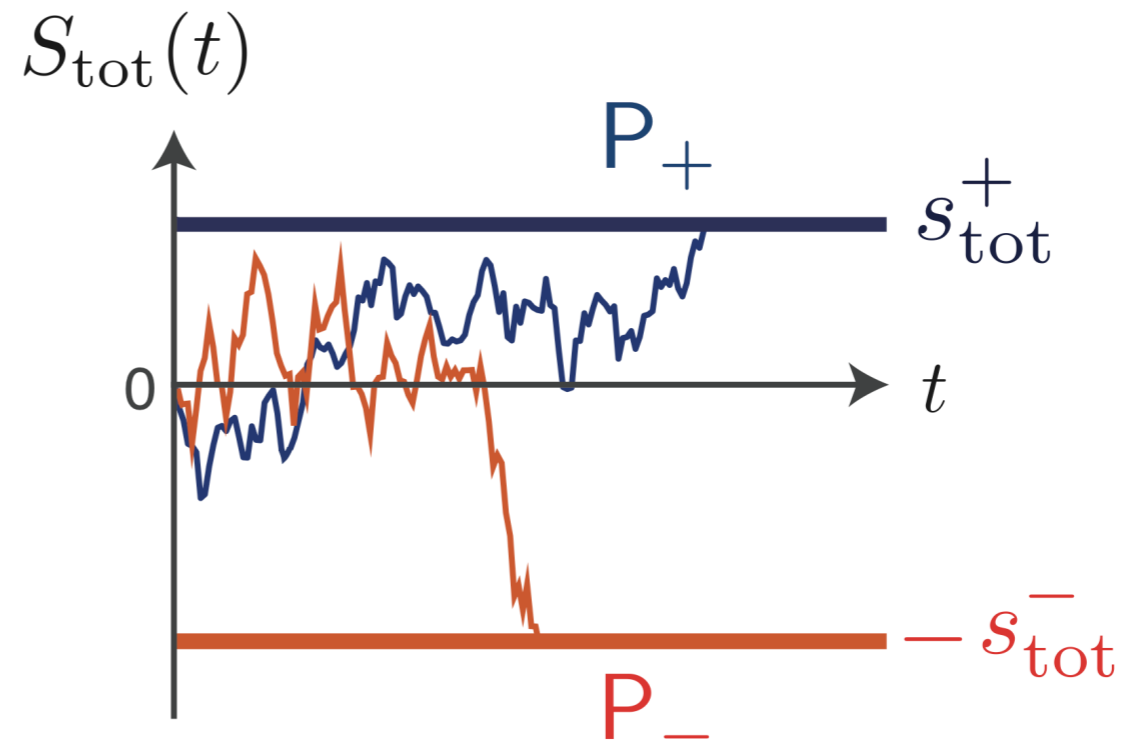
minutes



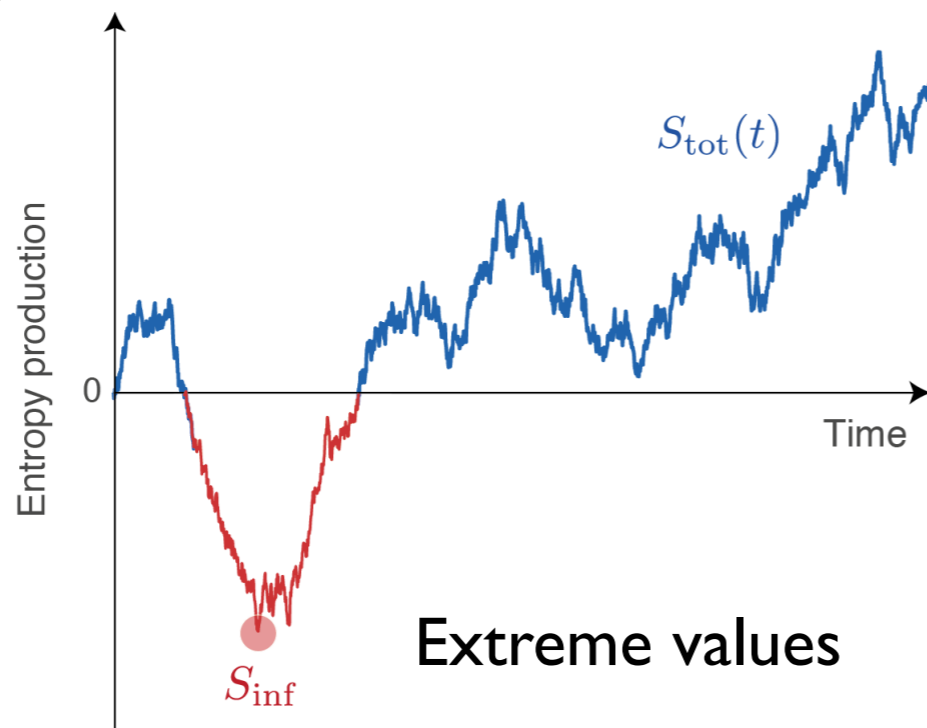
Unknown properties of entropy production



(Random) time to reach a certain threshold

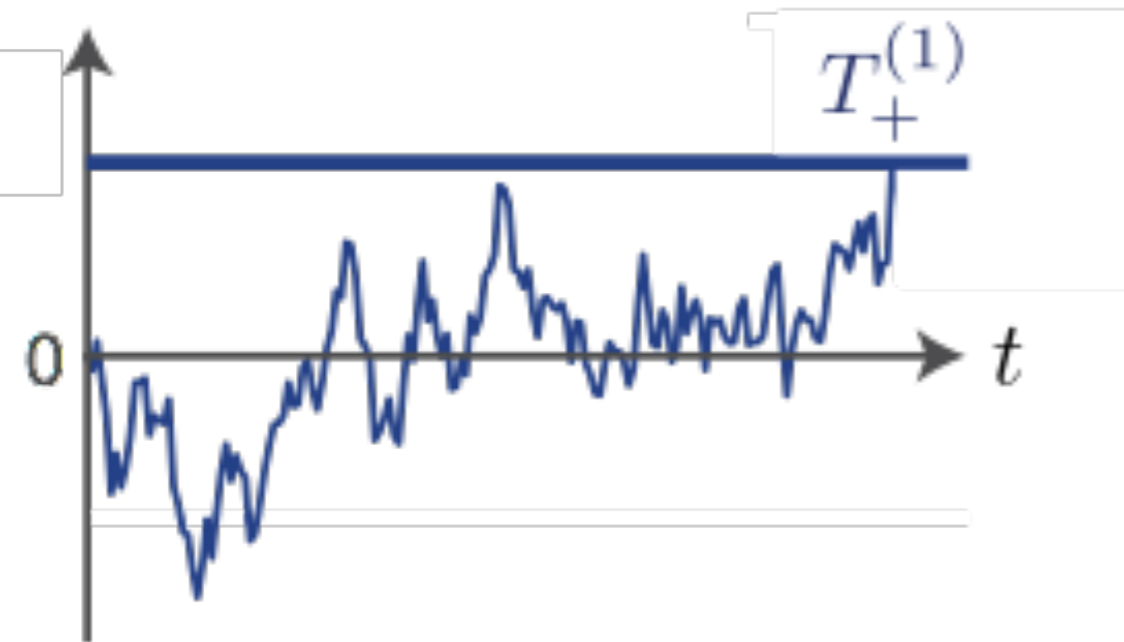


Splitting probabilities

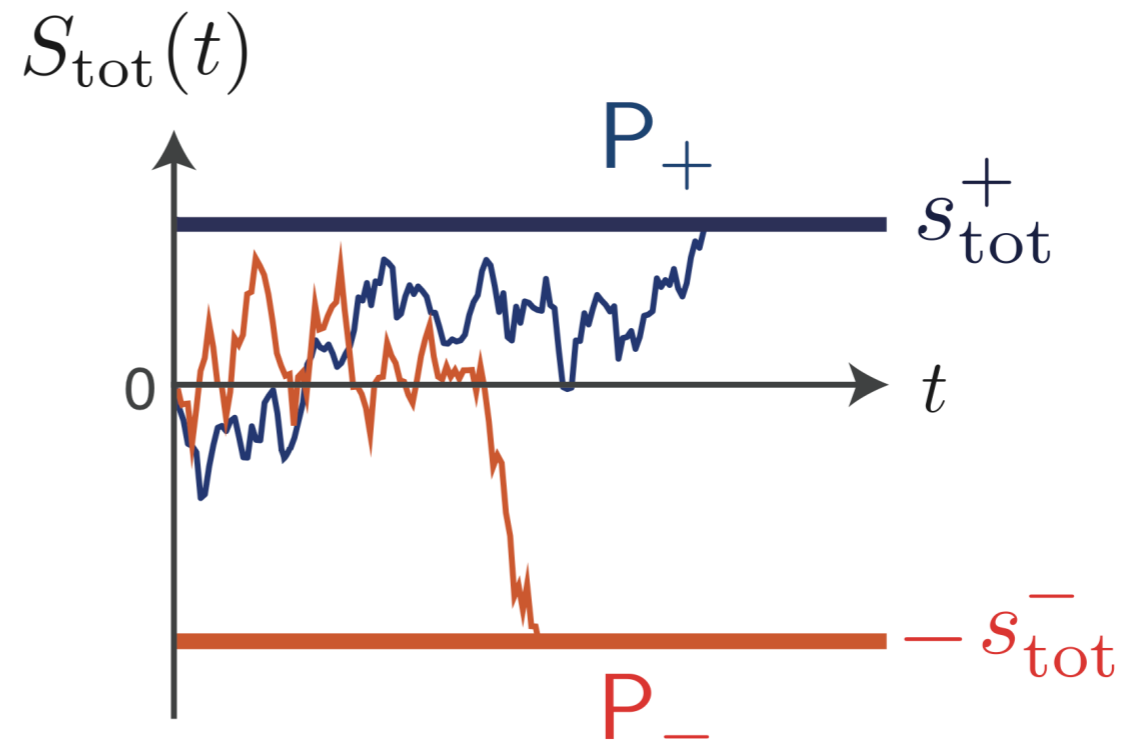


Extreme values

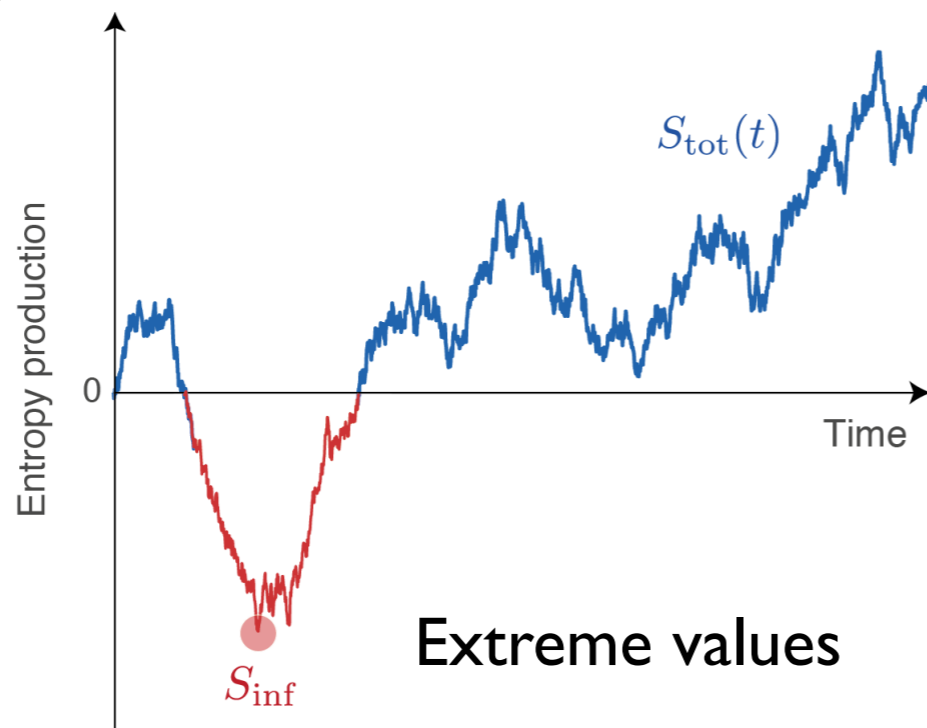
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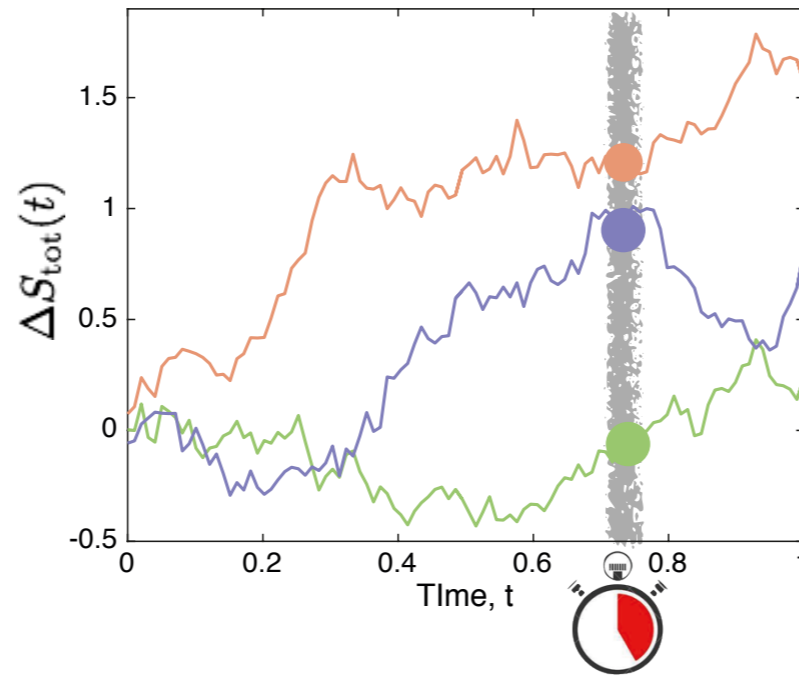
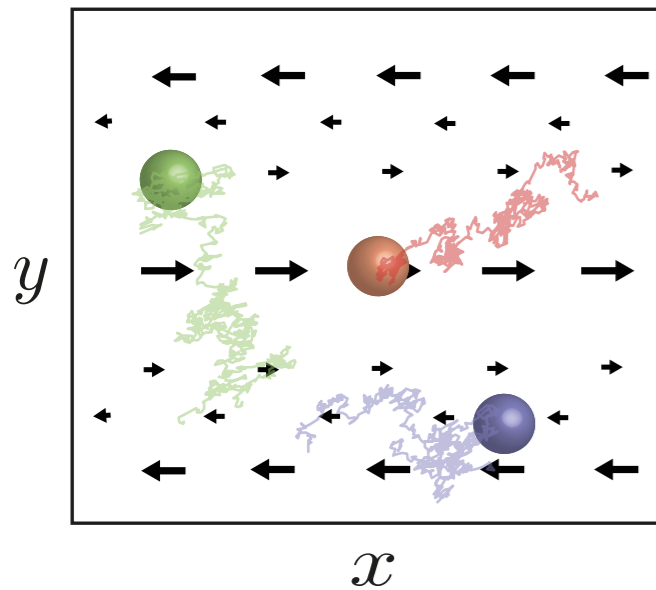


Extreme values

New universal properties ?

Martingale theory for stochastic thermodynamics

Stochastic thermodynamics without martingales



$$\langle e^{-\Delta S_{\text{tot}}(t)/k_B} \rangle = 1$$

Jarzynski's equality (1997)

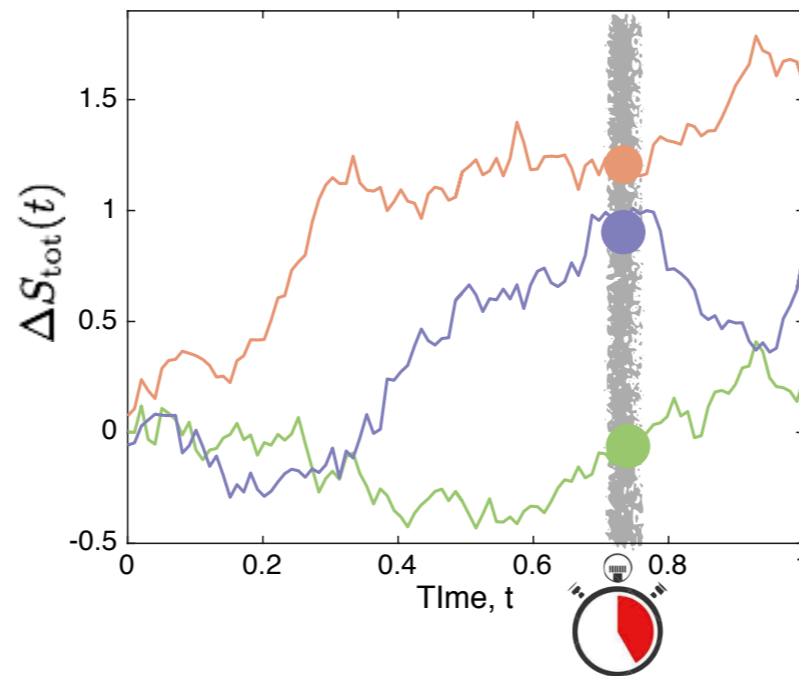
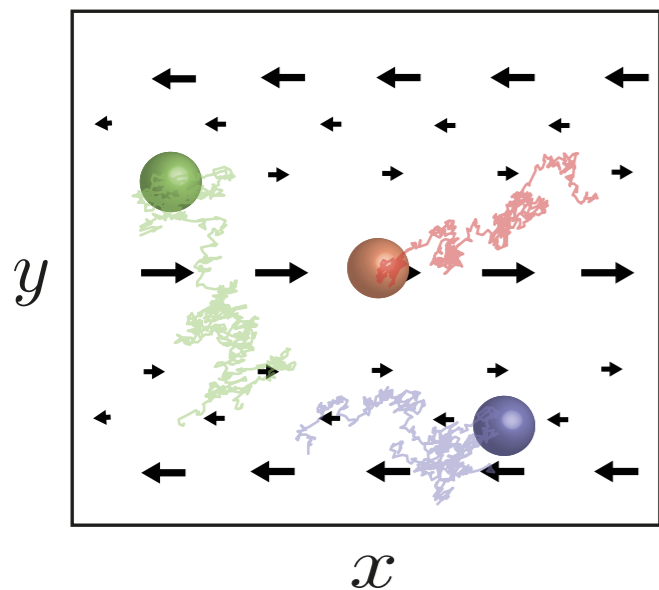
Integral Fluctuation theorem (Seifert 2005)

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Second law of thermodynamics

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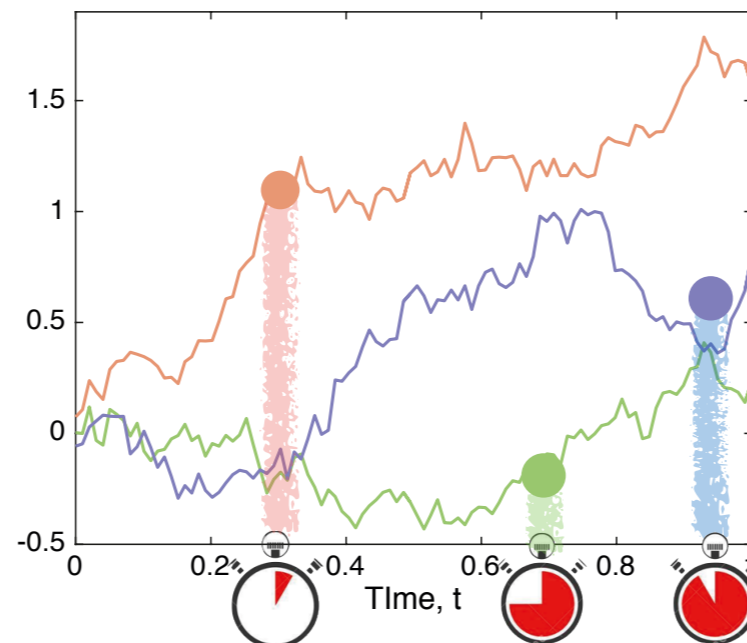
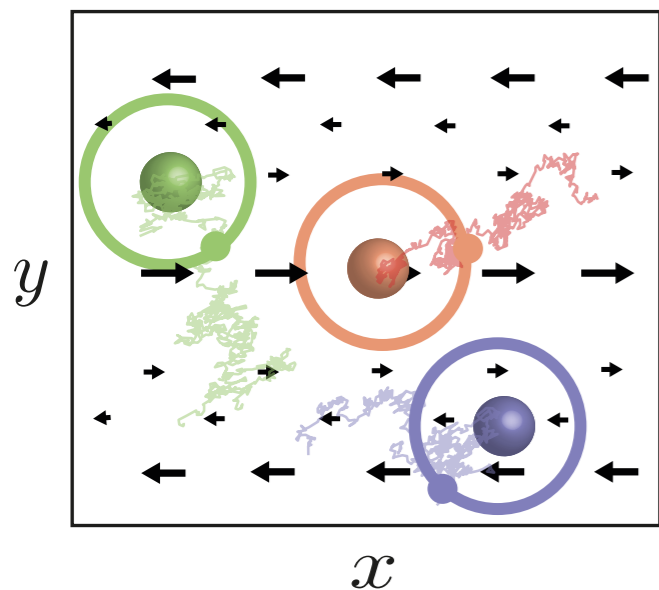
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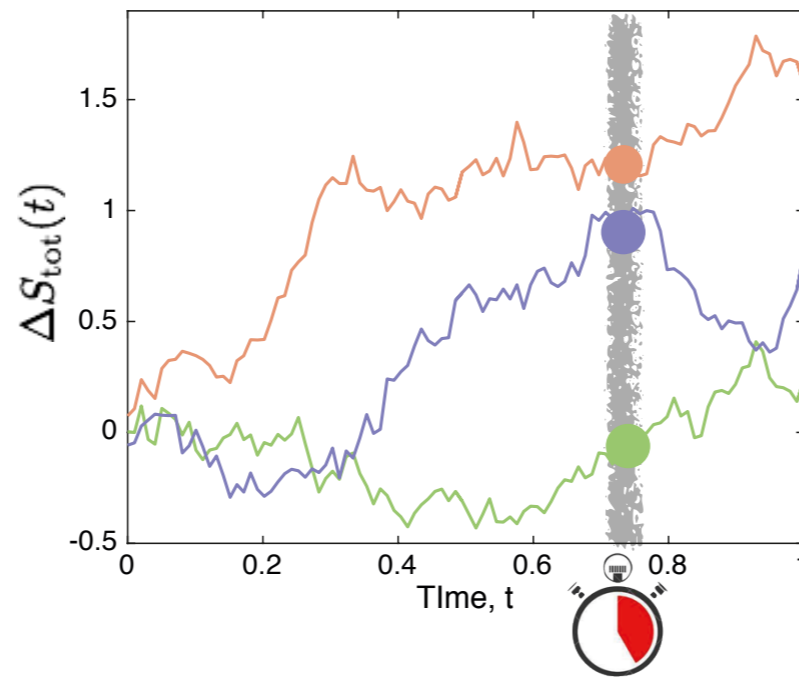
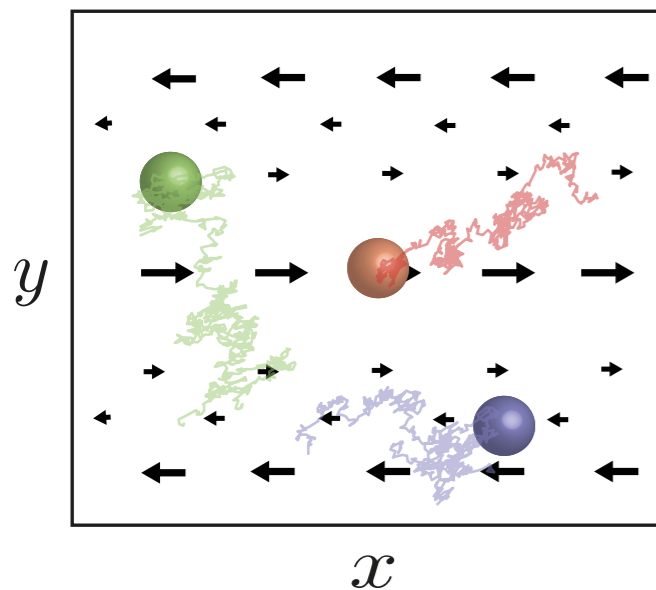
Second law of thermodynamics

Stochastic thermodynamics with Martingales



Martingale theory for stochastic thermodynamics

Stochastic thermodynamics without martingales



$$\langle e^{-\Delta S_{\text{tot}}(t)/k_B} \rangle = 1$$

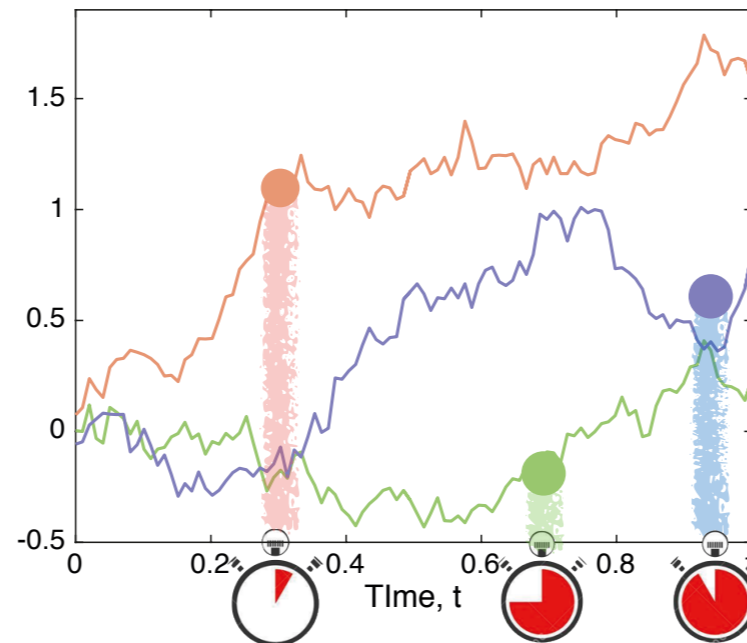
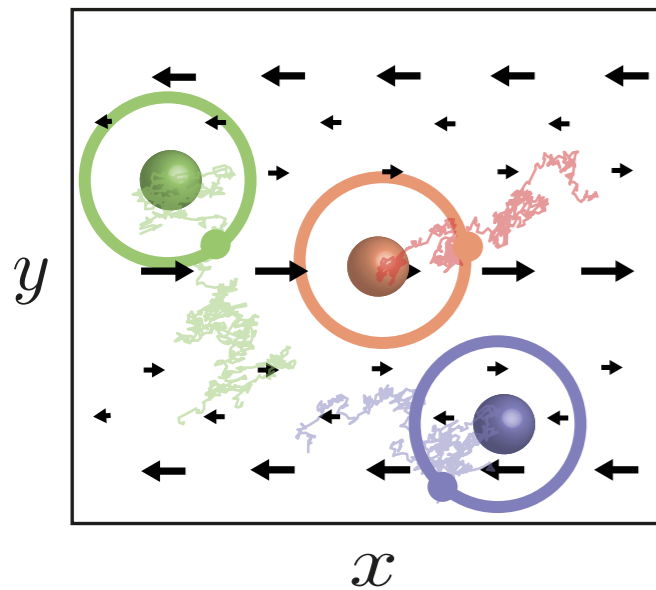
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Second law of thermodynamics

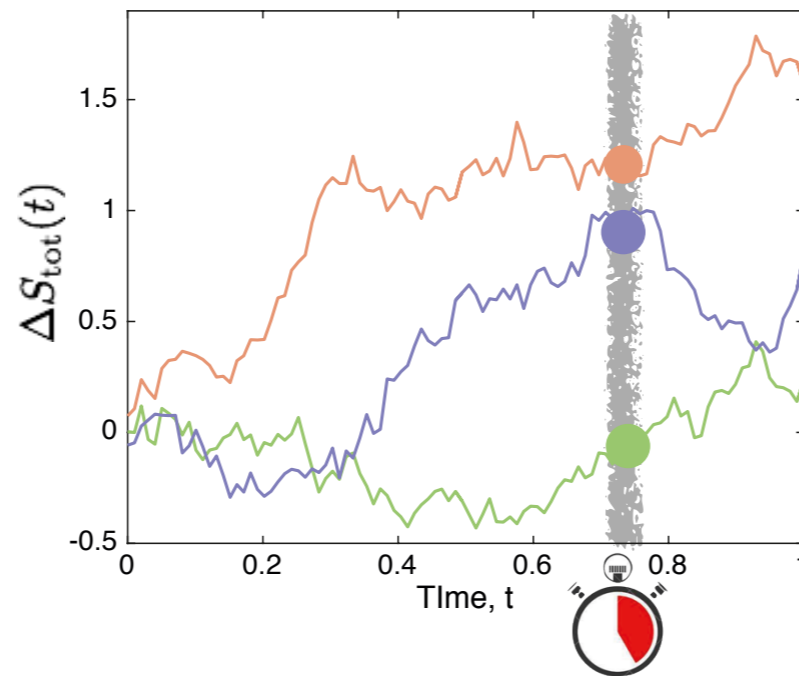
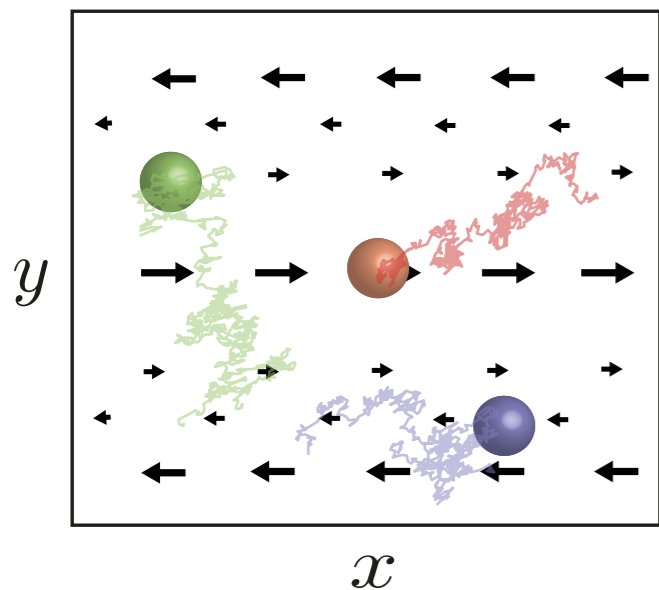
Stochastic thermodynamics with Martingales



**Stopping-time
fluctuation theorems ?**

Martingale theory for stochastic thermodynamics

Stochastic thermodynamics without martingales



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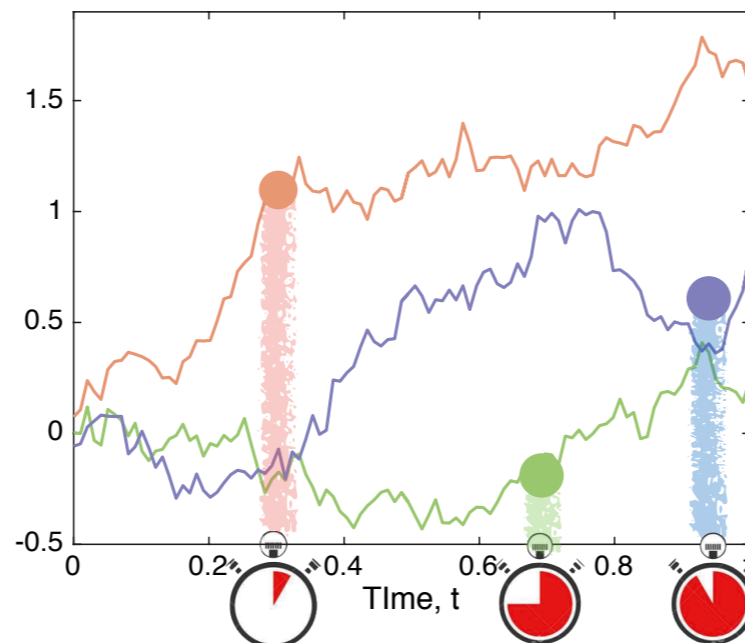
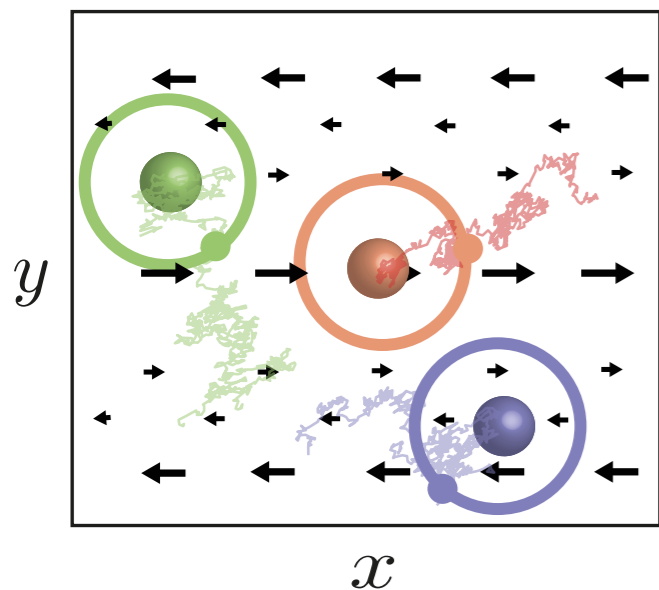
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Stochastic thermodynamics with Martingales

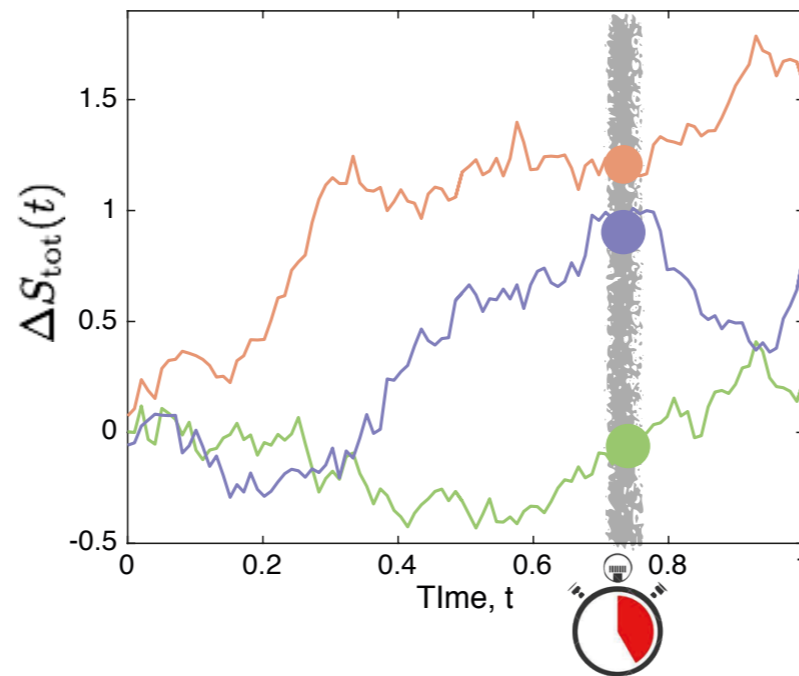
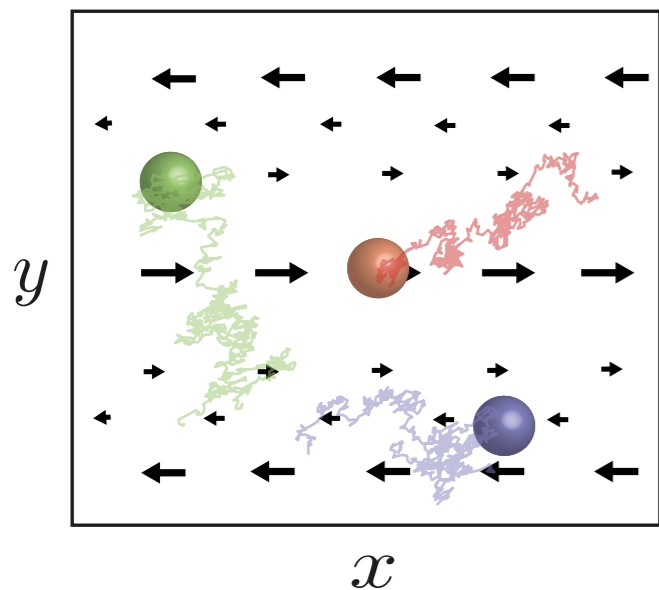


**Stopping-time
fluctuation theorems ?**

Extrema ?

Martingale theory for stochastic thermodynamics

Stochastic thermodynamics without martingales



$$\langle e^{-\Delta S_{\text{tot}}(t)/k_B} \rangle = 1$$

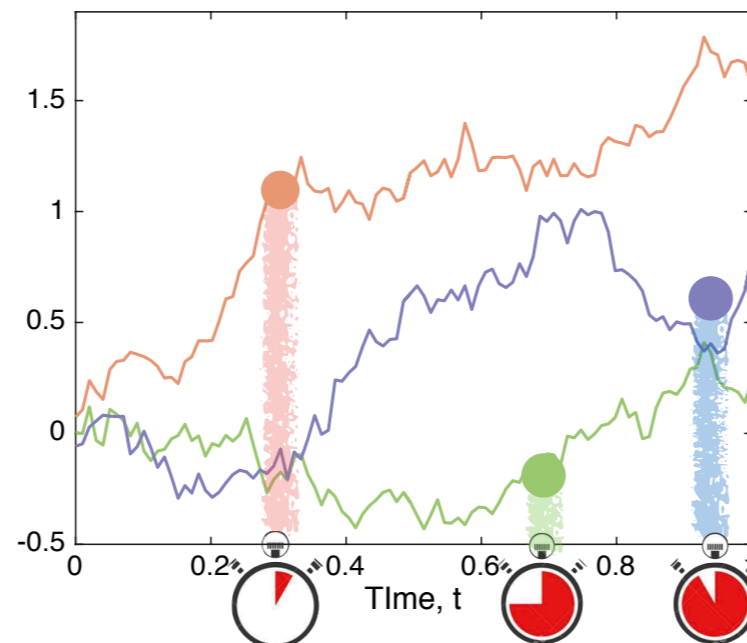
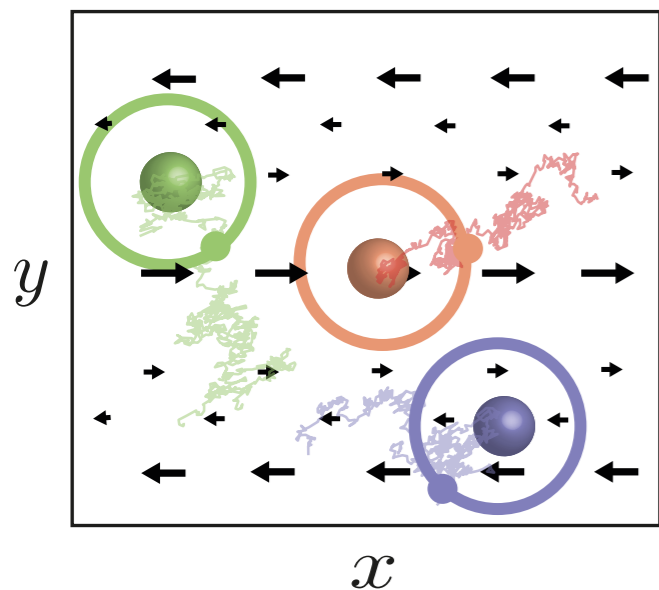
Jarzynski's equality (1997)

Integral Fluctuation theorem (Seifert 2005)

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Second law of thermodynamics

Stochastic thermodynamics with Martingales



Stopping-time
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Extrema ?

Gambling ?

***Martingale* theory
for entropy production:**

nonequilibrium steady states

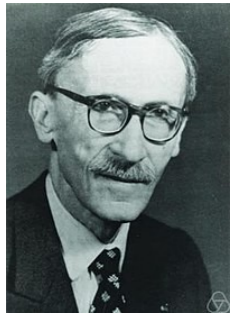
Martingales

$M(t)$ is a martingale with respect to $X(t)$ if:



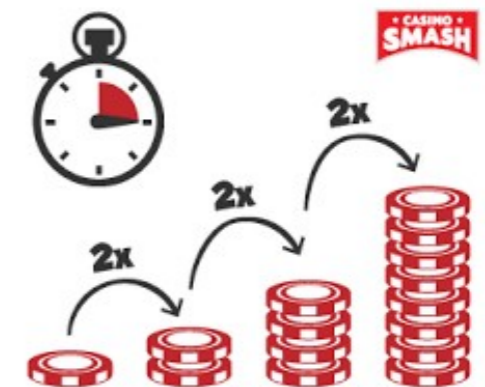
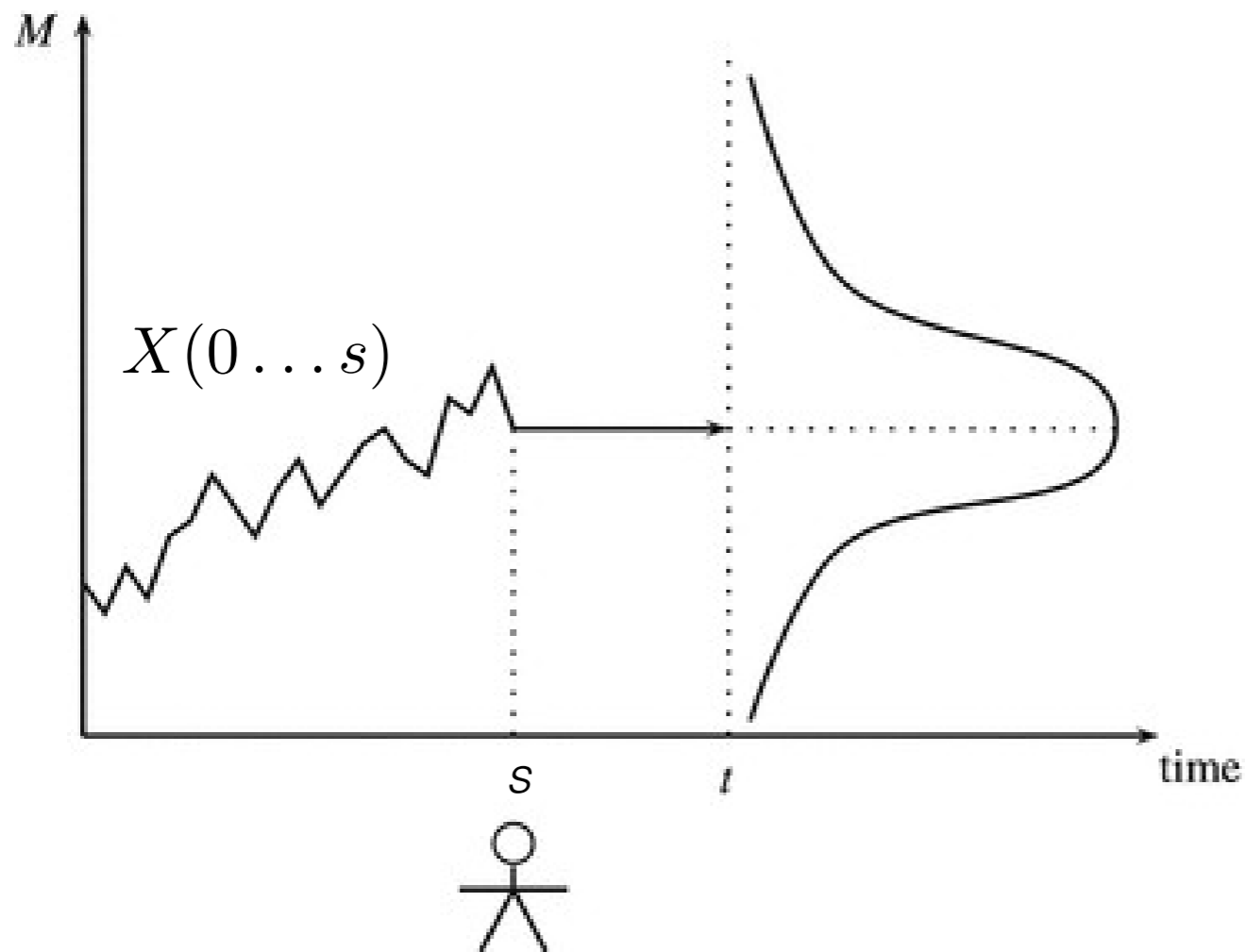
- $M(t)$ is a real-valued function on $X(0 \dots t)$
- $\langle |M(t)| \rangle < \infty$
- $\langle M(t) | X(0 \dots s) \rangle = M(s)$, for all $s < t$

Martingales



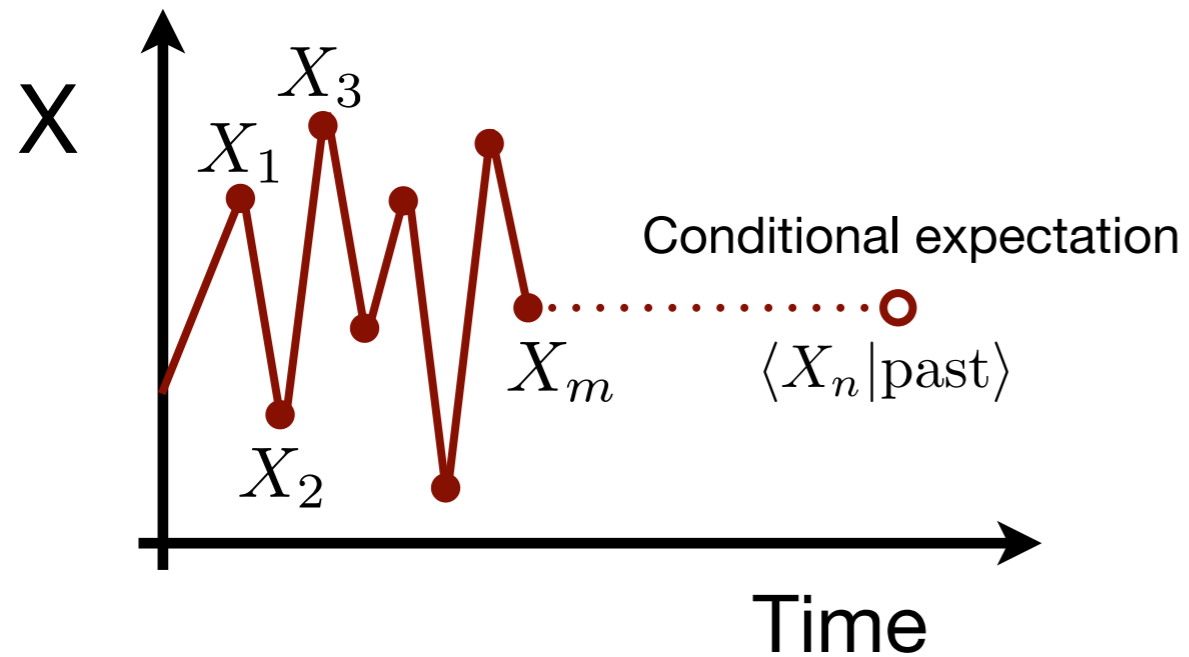
Introduced in probability theory by **Paul Lévy** in 1934 and named by **Ville** (1939)

Popular meaning of “martingale”:
Double-up strategy in gambling



$$\langle M(t) | X(0 \dots s) \rangle = M(s), \quad \text{for all } s < t$$

Martingales and Submartingales

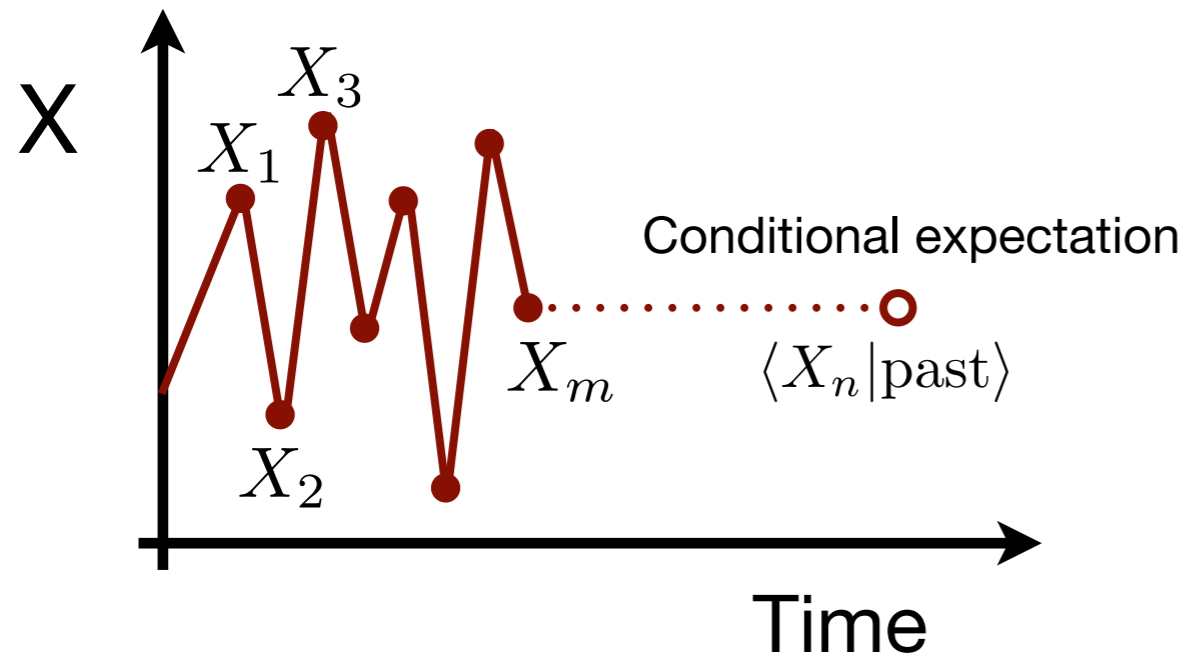


Martingale

$$n \geq m$$

$$\langle X_n | X_1, \dots, X_m \rangle = X_m$$

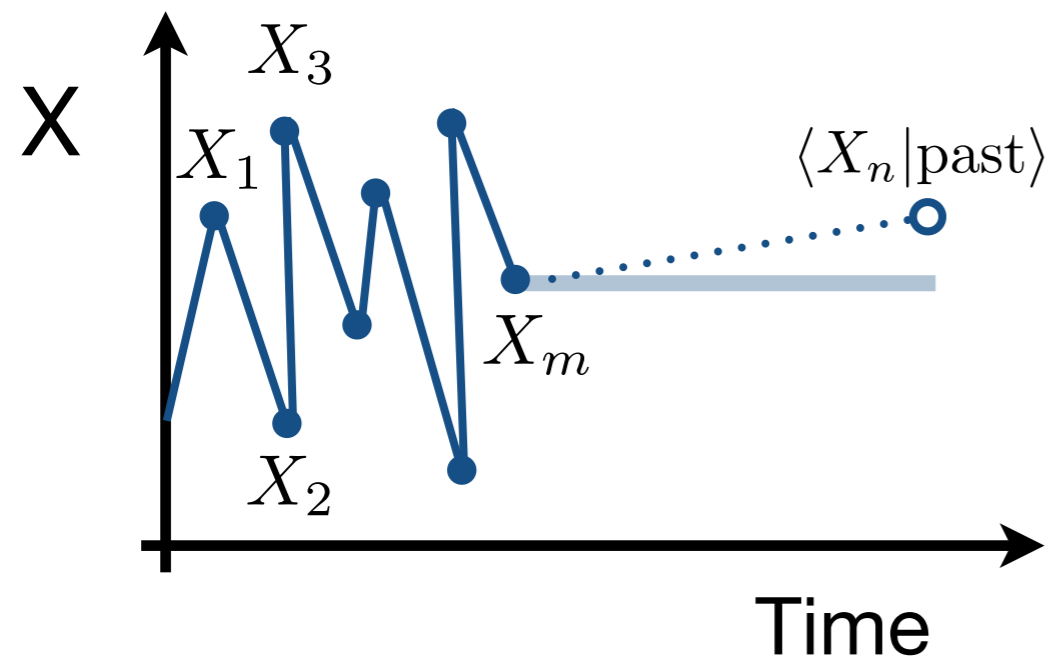
Martingales and Submartingales



Martingale

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Submartingale

$$n \geq m$$

$$\langle X_n | X_1, X_2, \dots, X_m \rangle \geq X_m$$

Quick examples of martingales

ER, et al., *Physics with Martingales* (in preparation)

- **Random walks**

Drifted-diffusion $X(t) = vt + B(t)$ submartingale if $v > 0$
supermartingale if $v < 0$

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Drifted-diffusion
in comoving frame $X(t) - vt = B(t)$ martingale for all v

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Poisson processes $N(t) \in \mathbb{N}$ with rate λ are submartingales

$N(t) - \lambda t \in \mathbb{Z}$ martingale for all λ

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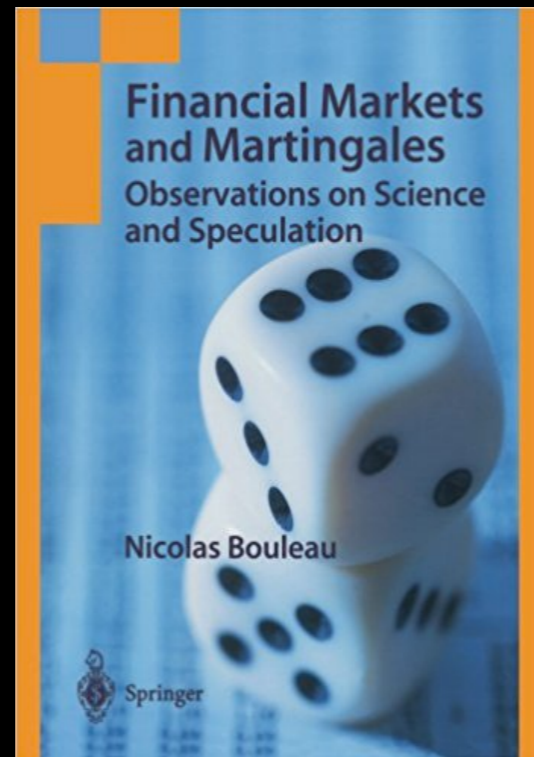
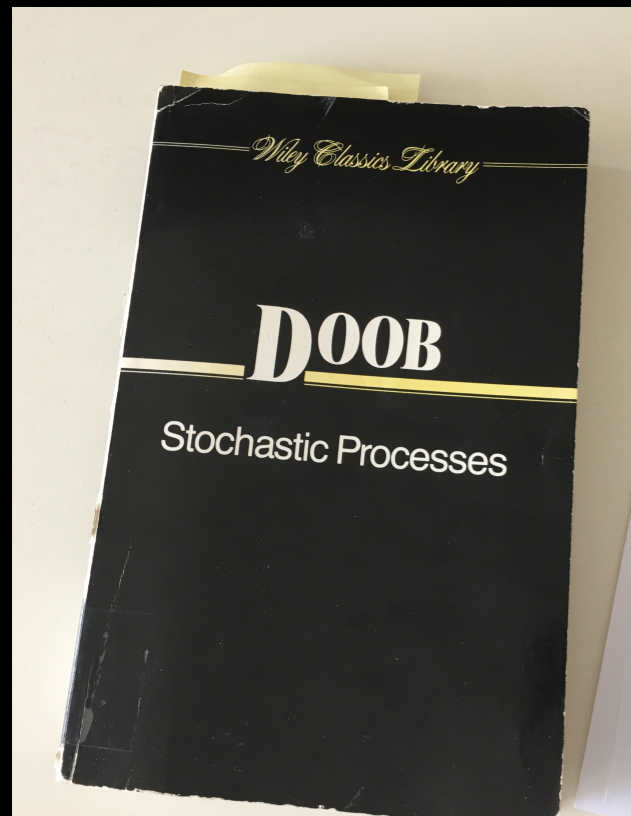
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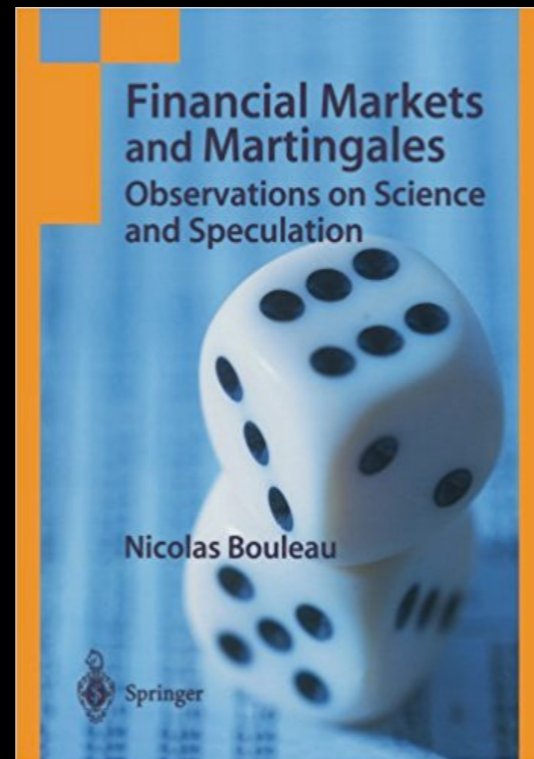
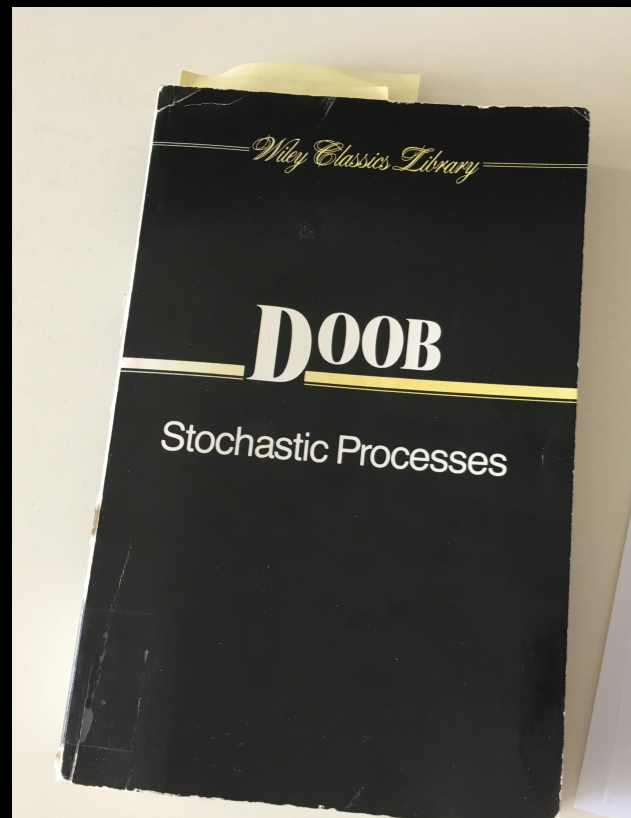
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- Assumptions: (I) non-equilibrium steady state
- Valid for: continuous, discrete, Markovian and non-Markovian processes
- Also for other reference measures (“action functionals”):



Unveiling new generic thermodynamic properties using Martingales

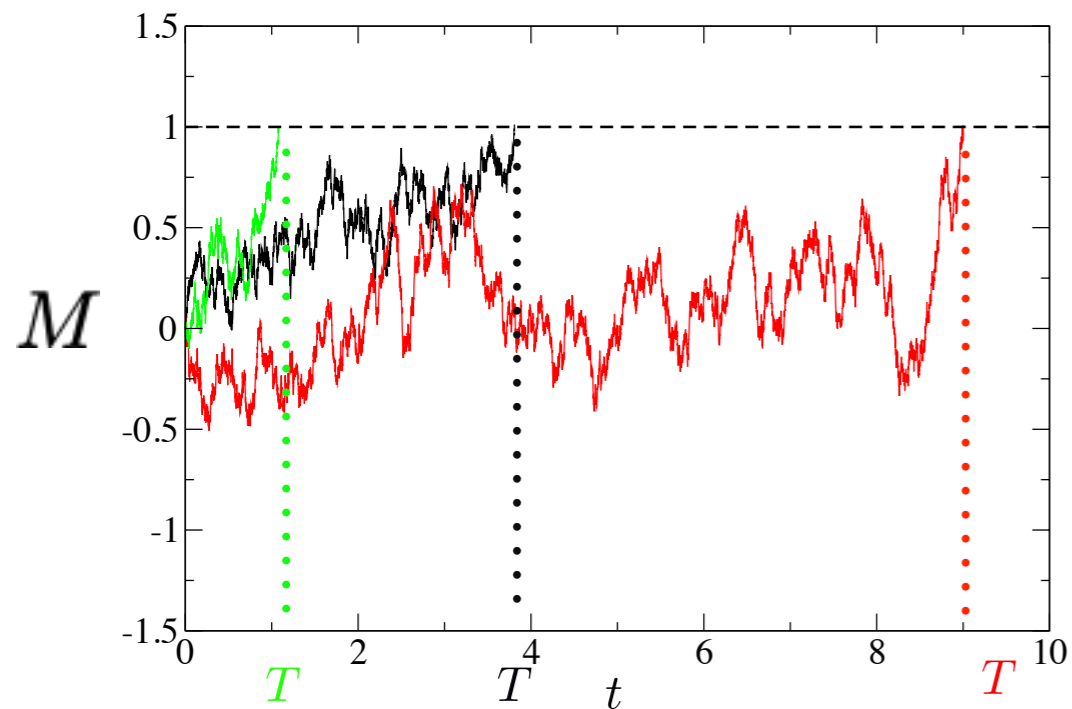


Unveiling new generic thermodynamic properties using Martingales

Thermodynamic laws at stopping times

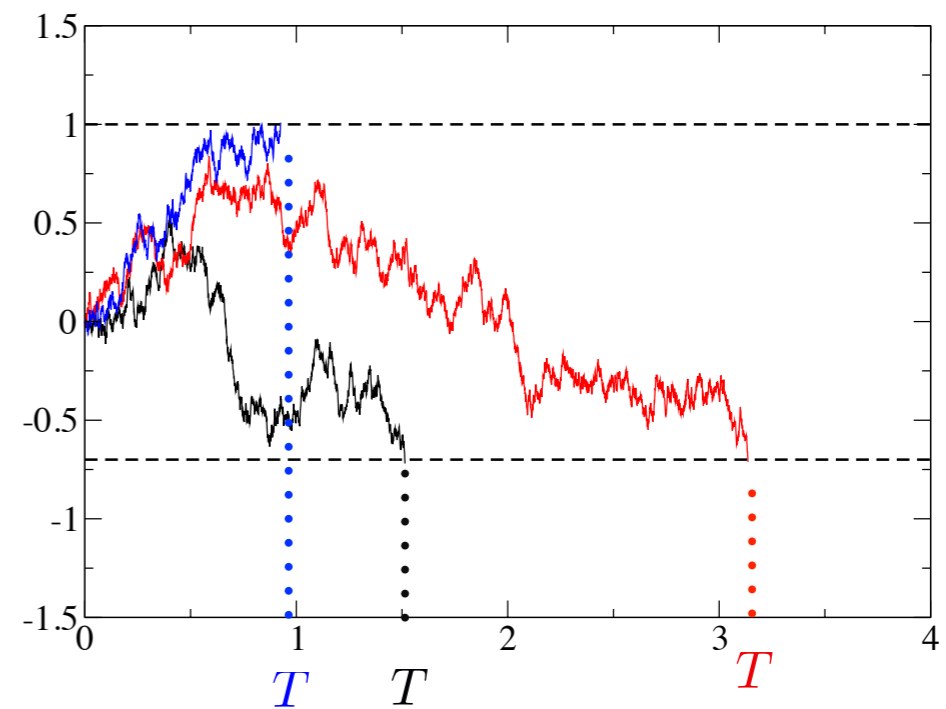
Can a gambler make fortune in a fair game by quitting at an intelligently chosen moment?

Gambling with Martingales



$$\langle M(T) \rangle = 1 \neq \langle M(0) \rangle = 0$$

Gambler makes profit



$$\langle M(T) \rangle = \langle M(0) \rangle = 0$$

Gambler on average makes no profit

Can a gambler make fortune in a fair game by quitting at an intelligently chosen moment?

No, if the gambler cannot foresee the future, cannot cheat, and has access to a finite budget

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$$\langle M(\mathcal{T}) | M(0) \rangle = M(0)$$

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stopping time: random time, functional on the stochastic trajectory

“ to answer $\mathcal{T} < t?$ one only needs information in $[0, t]$ “

Integral fluctuation theorems at stopping times

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Second law
at stopping times

$$\langle S_{\text{tot}}(\mathcal{T}) \rangle \geq 0$$

I. Neri, ER, F. Jülicher, PRX **7**, 011019 (2017)

I. Neri, ER, S. Pigolotti, F. Jülicher, arXiv 1903.08115 (2019)

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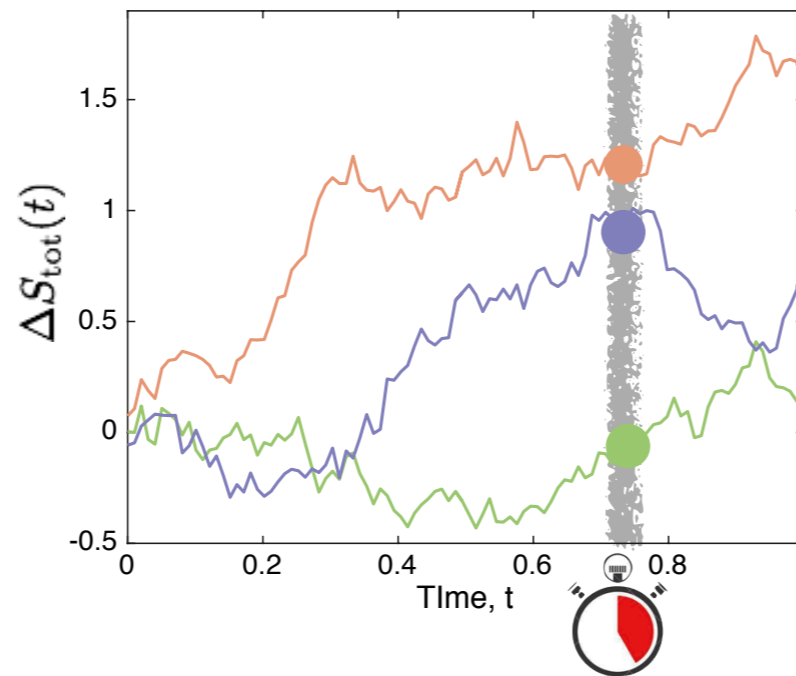
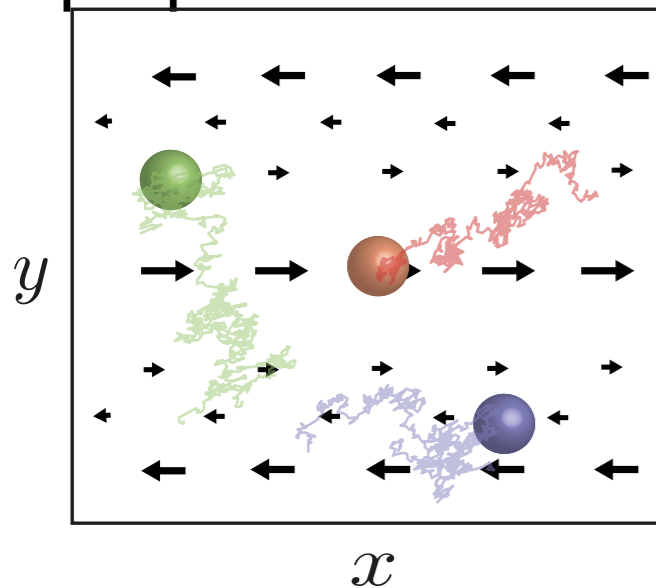
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Martingale theory for stochastic thermodynamics

Fluctuation theorems for non-equilibrium steady states : fixed-time properties



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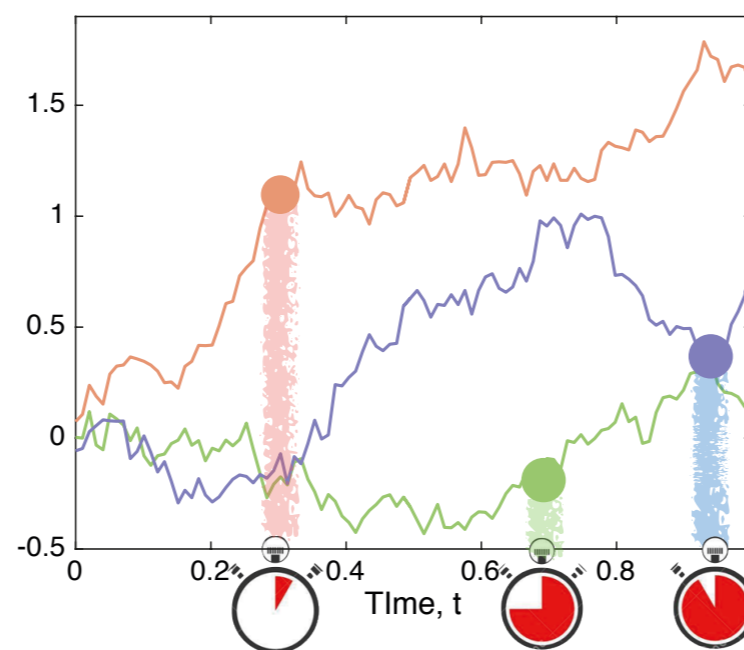
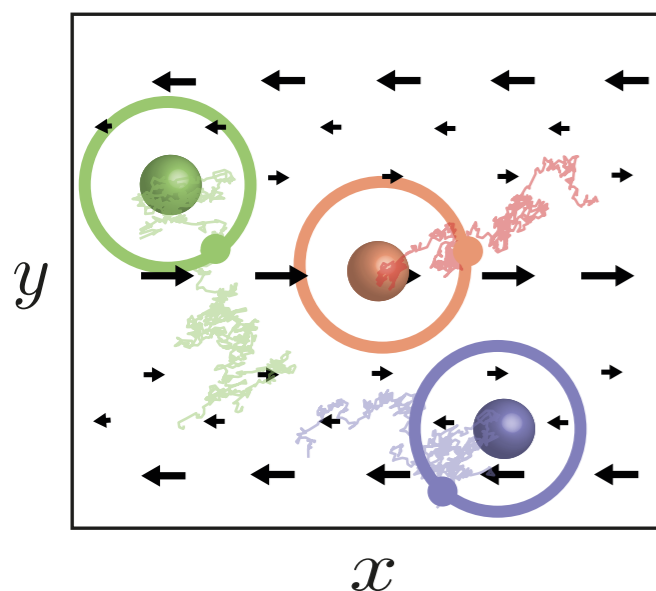
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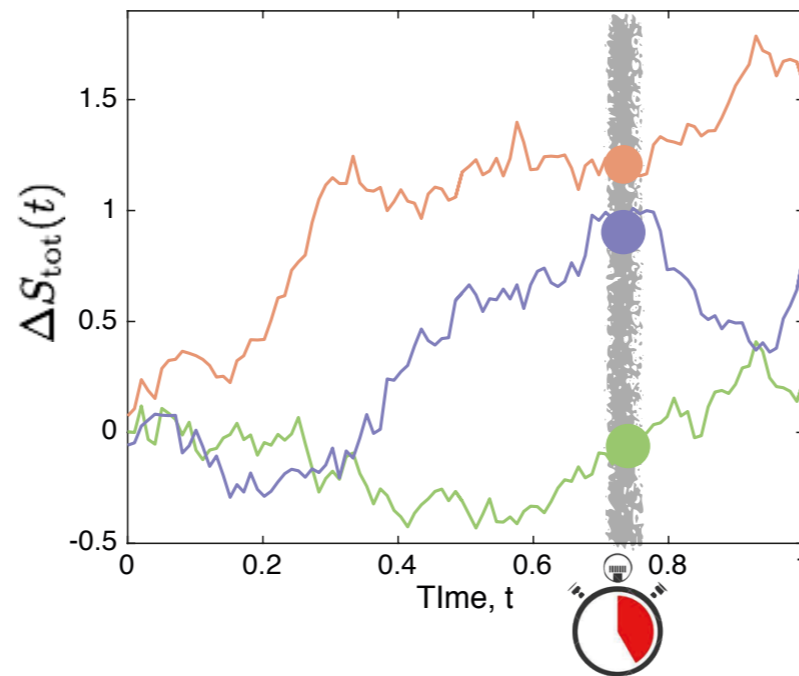
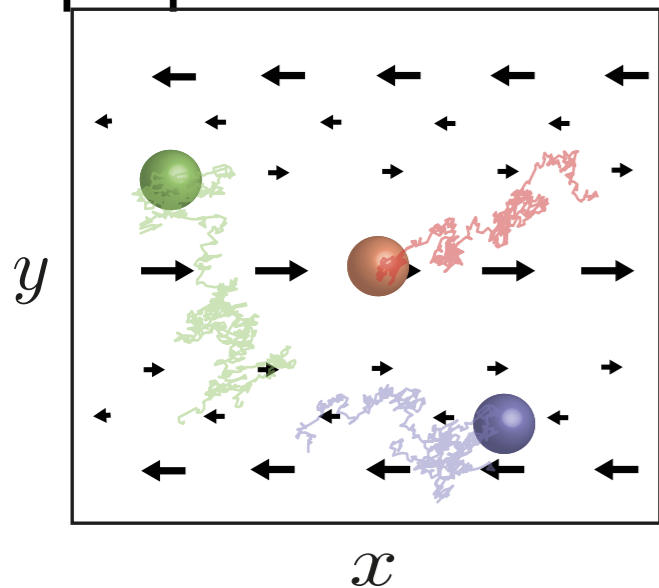
Martingale theory: stopping-time statistics



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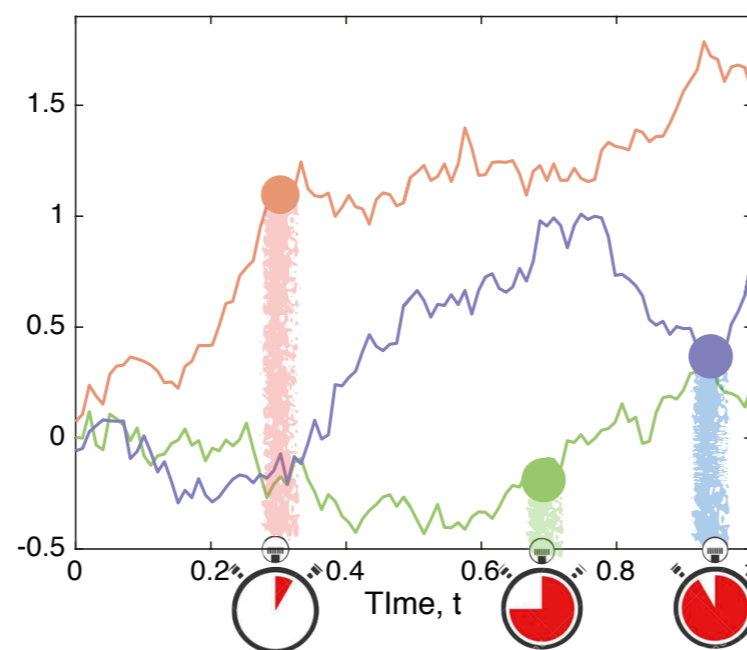
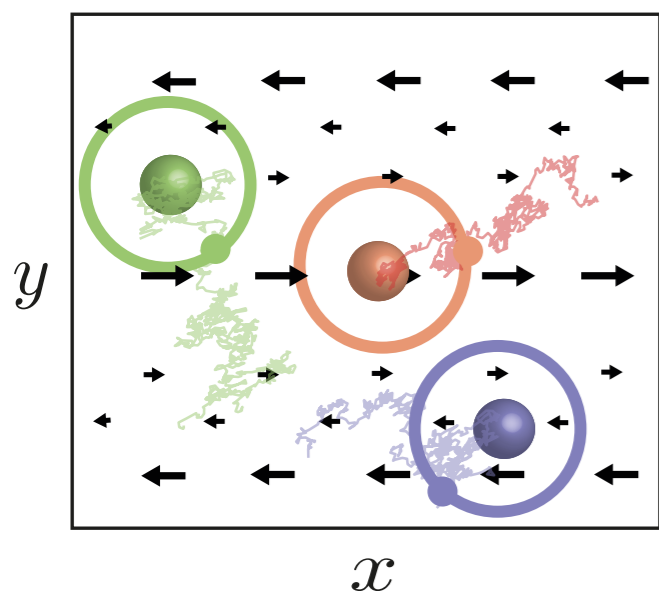
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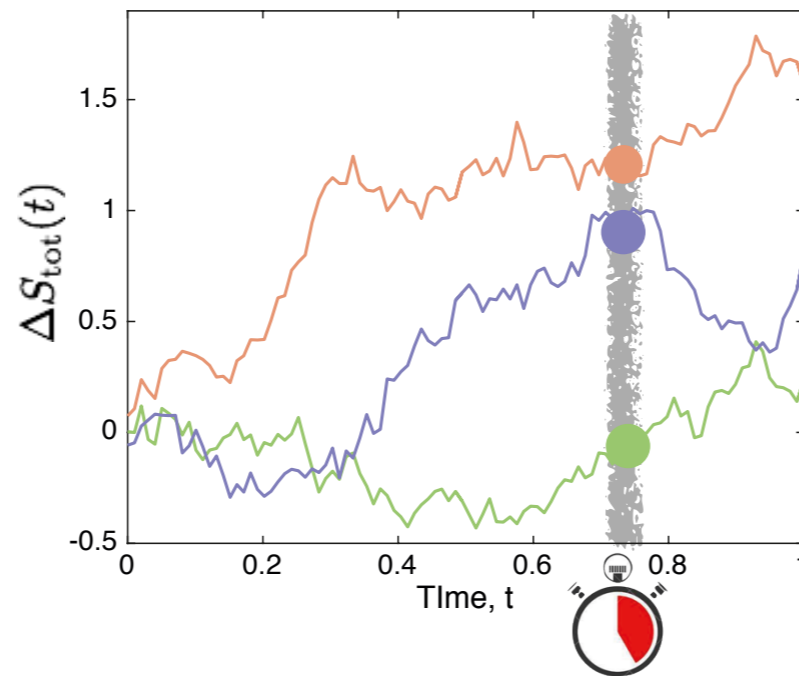
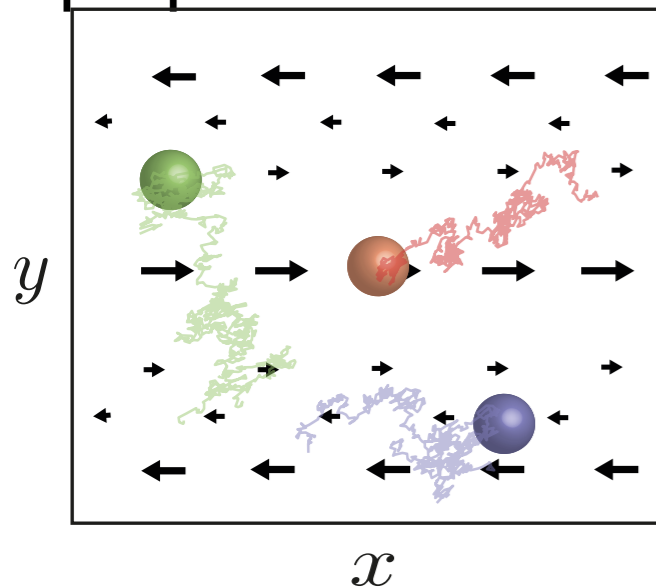
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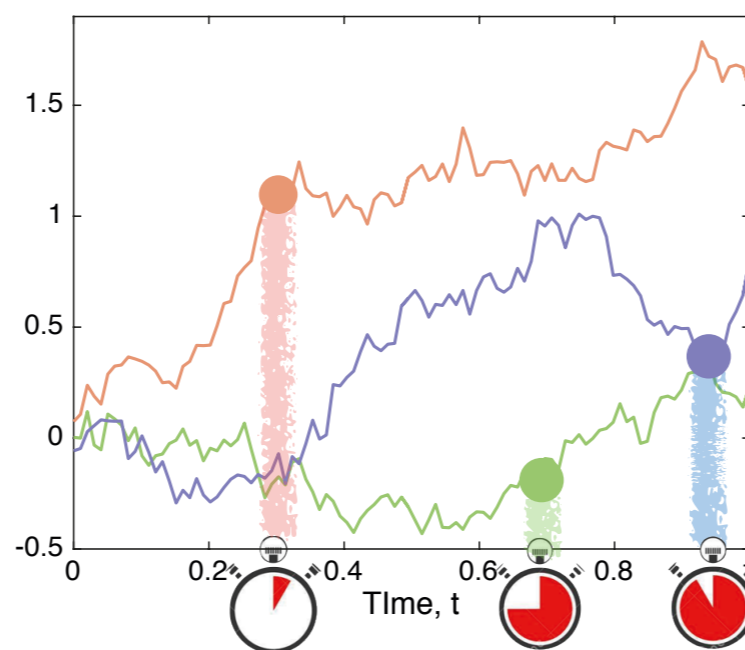
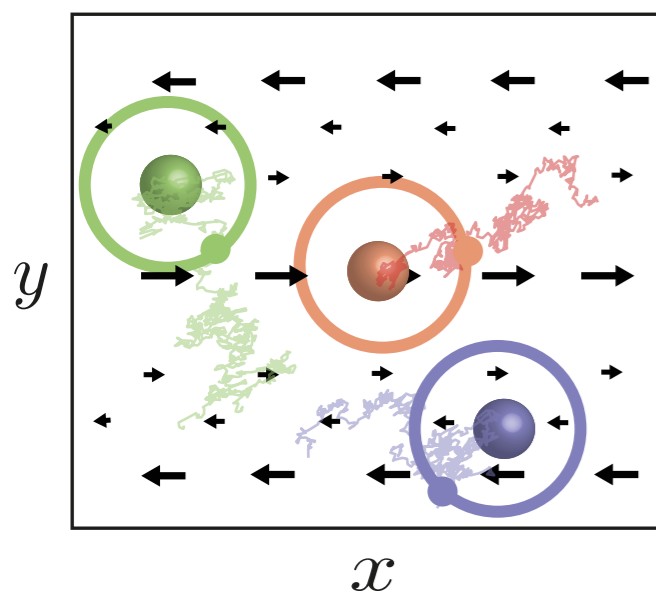
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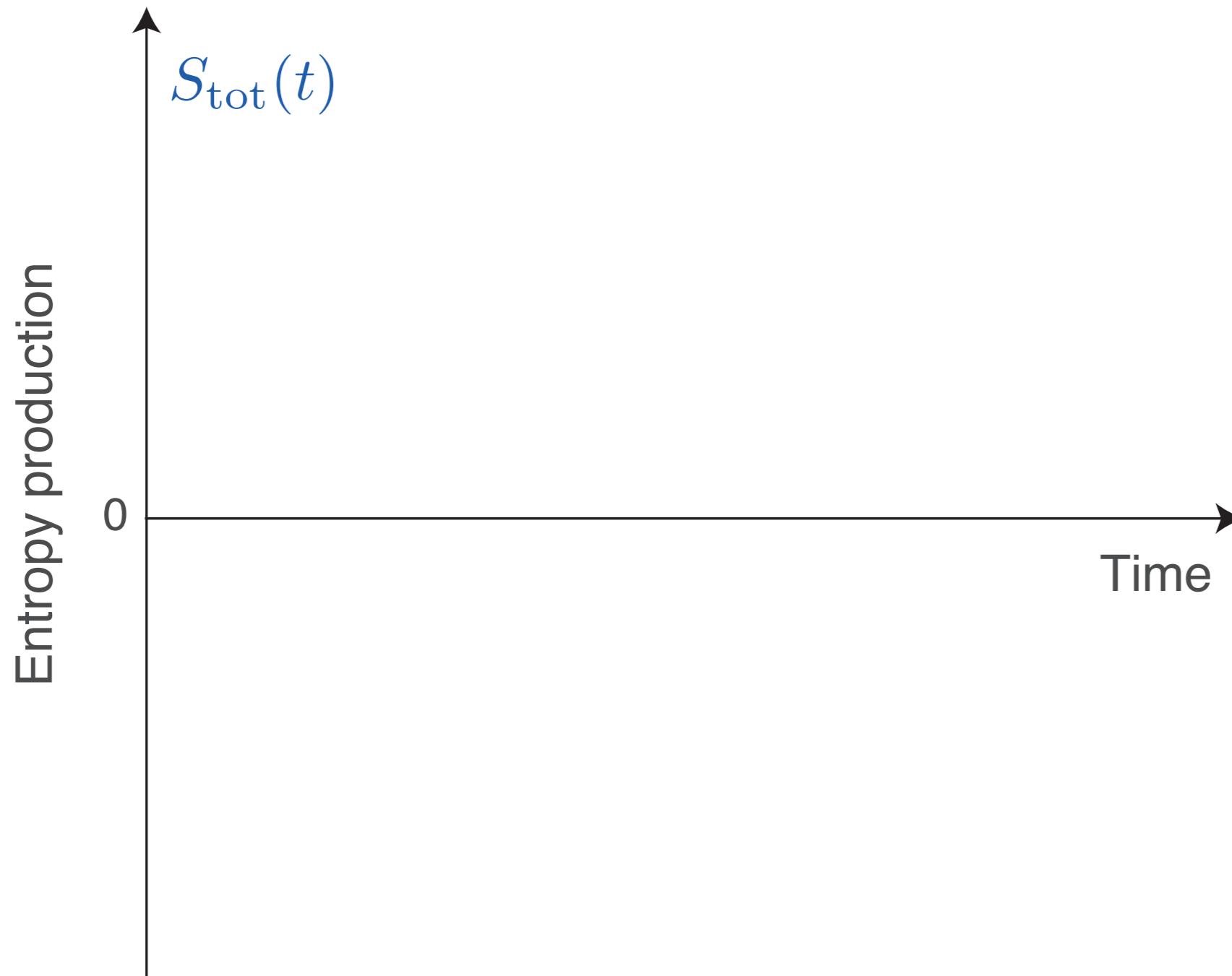
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Second law at stopping times

Extreme-value statistics

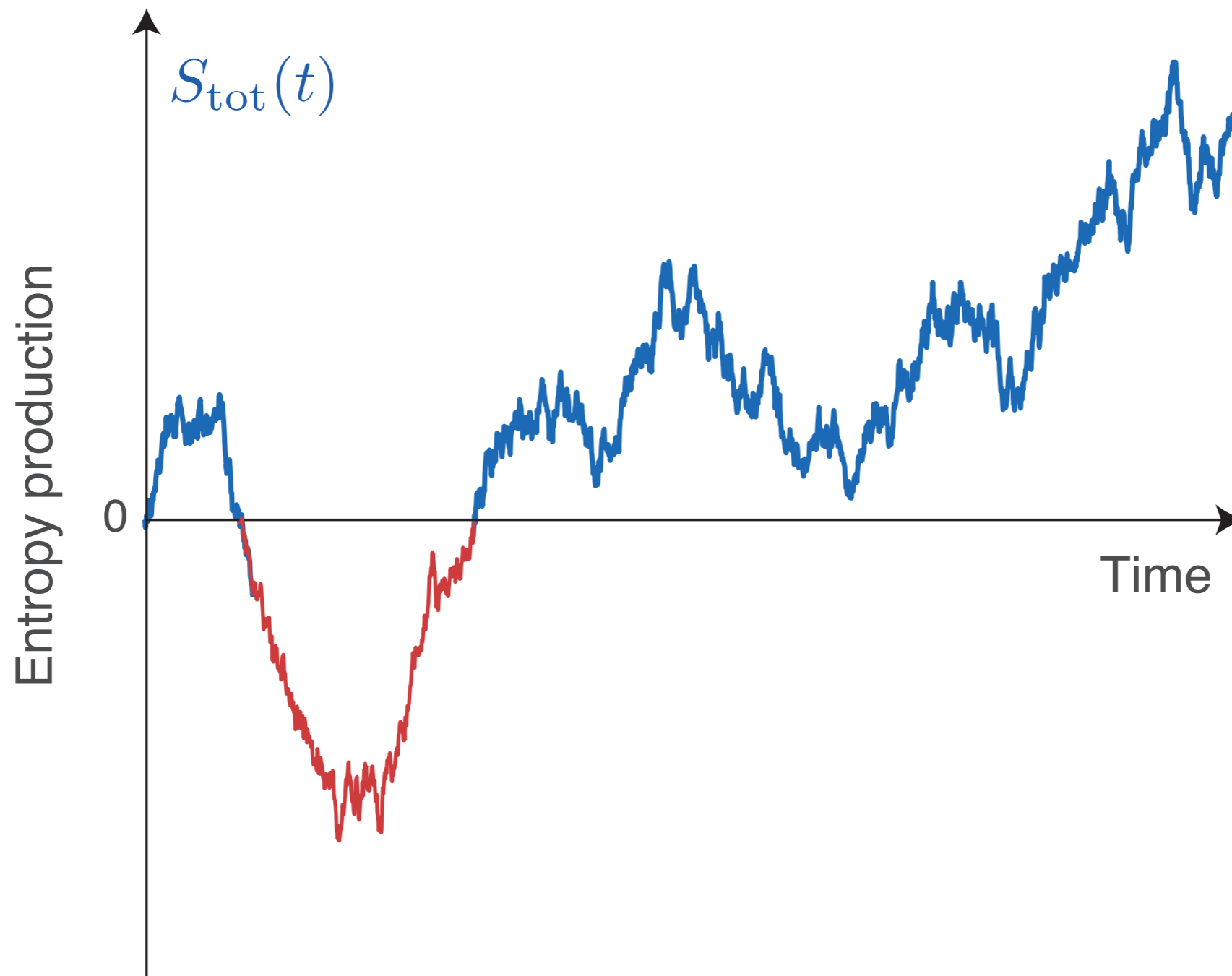
Extreme values

What's the negative record of entropy production?



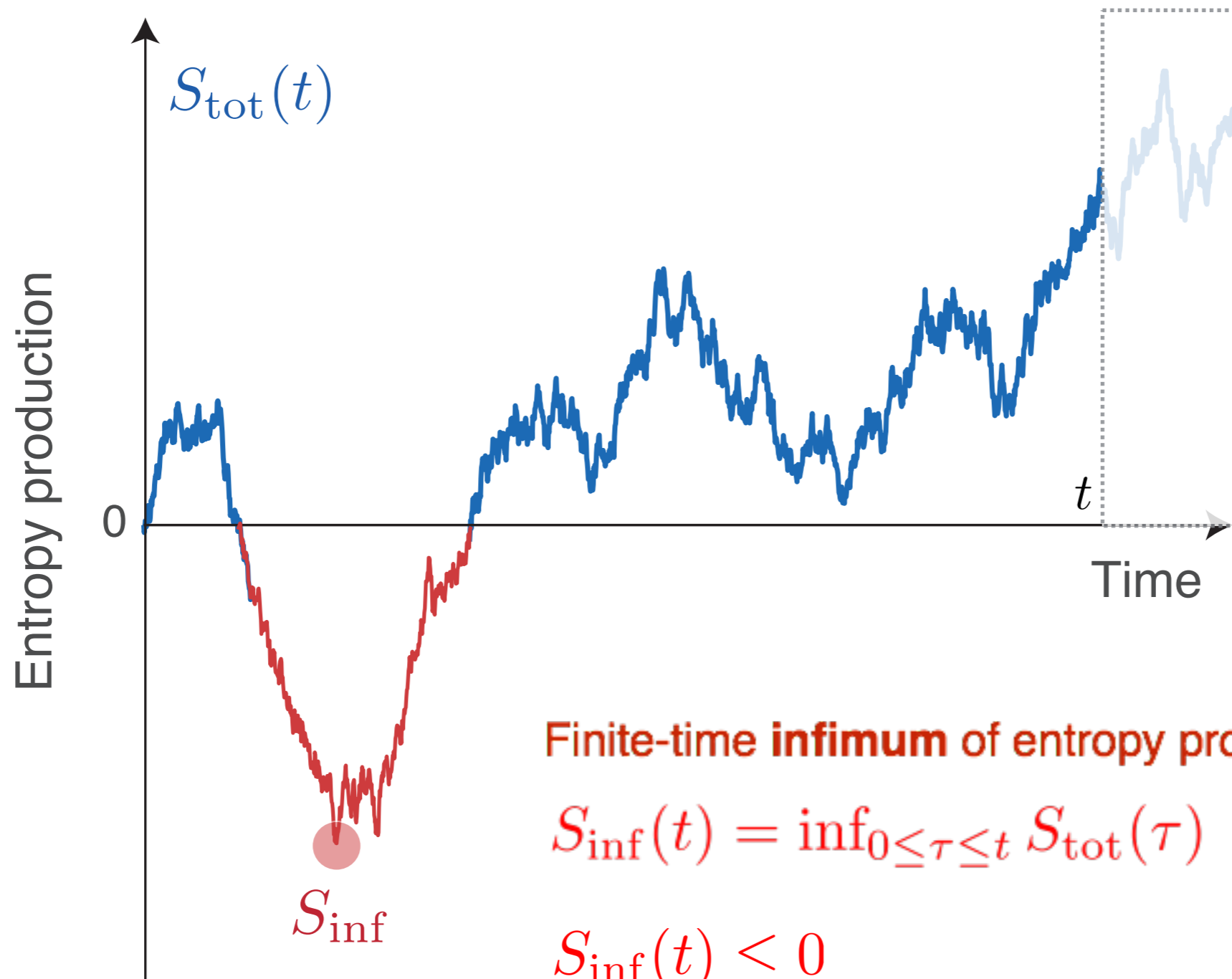
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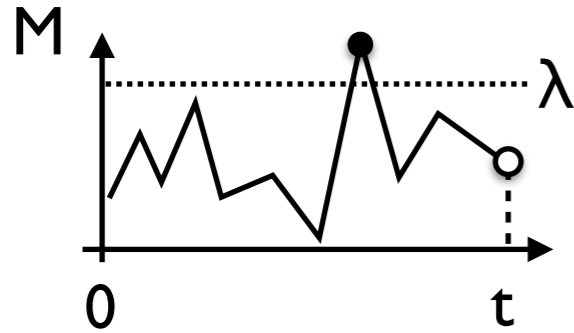
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Extreme values: infimum law

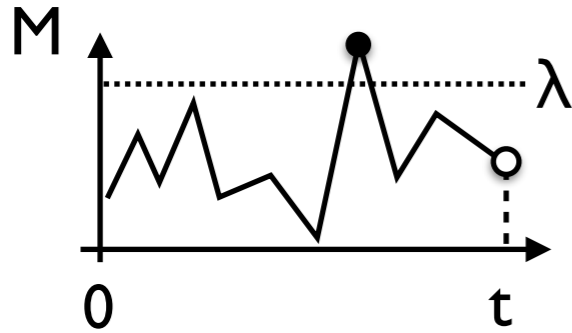
- Doob's maximal inequality for positive submartingales (Jean Ville's PhD Thesis)



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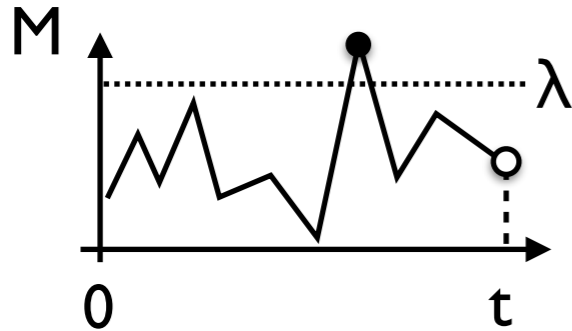
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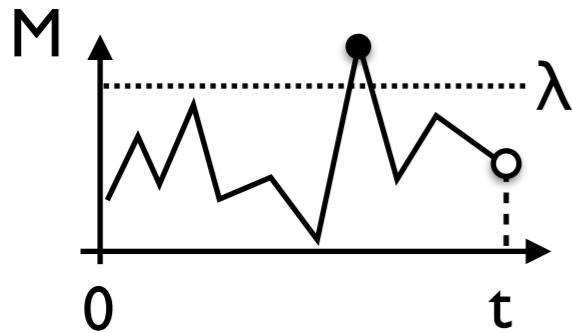
- Change of variables

$$\Pr \left(\underbrace{\inf_{\tau \in [0, t]} S_{\text{tot}}(\tau)}_{\equiv S_{\text{inf}}(t)} \leq -k_B \ln \lambda \right) \leq \frac{1}{\lambda} \Rightarrow \Pr \left(S_{\text{inf}}(t) \leq -s \right) \leq e^{-s/k_B}$$

$$\Pr \left(S_{\text{inf}}(t) \geq -s \right) \geq 1 - e^{-s/k_B}$$

Extreme values: infimum law

- Doob's maximal inequality for positive submartingales (Jean Ville's PhD Thesis)



$$\Pr \left(\sup_{\tau \in [0, t]} M(\tau) \geq \lambda \right) \leq \frac{\langle M(t) \rangle}{\lambda}$$

- Apply theorem to the positive martingale $M = e^{-S_{\text{tot}}/k_B}$, and the fluctuation theorem

$$\Pr \left(\sup_{\tau \in [0, t]} e^{-S_{\text{tot}}(\tau)/k_B} \geq \lambda \right) \leq \frac{\overbrace{\langle e^{-S_{\text{tot}}(t)/k_B} \rangle}^{=1}}{\lambda}$$

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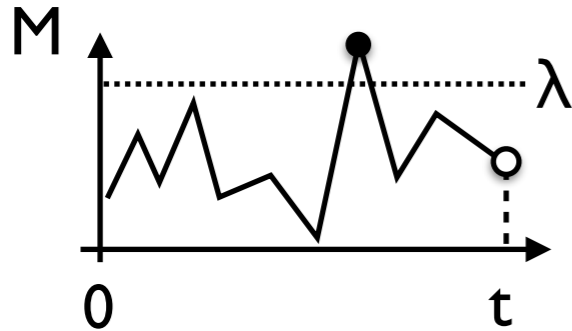
- Stochastic dominance ($s > 0$)

$$\Pr \left(-S_{\text{inf}}(t) \leq s \right) \geq 1 - e^{-s/k_B} \Rightarrow \langle -S_{\text{inf}}(t) \rangle \leq k_B \Rightarrow \langle S_{\text{inf}}(t) \rangle \geq -k_B$$

CDF of -Infimum CDF of Exp r.v. with mean k_B

Extreme values: infimum law

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CDF of -Infimum

CDF of Exp r.v. with mean k_B

Numerical and experimental tests

$$\Pr\left(-S_{\text{inf}}(t) \leq s\right) \geq 1 - e^{-s/k_B}$$

$$\langle S_{\text{inf}}(t) \rangle \geq -k_B$$

hold for “any” **nonequilibrium steady state**:

- Discrete-time processes (e.g. Markov chains)
- Continuous processes (e.g. Langevin dynamics)
- Continuous-time processes with jumps (e.g. Markov-jump processes)

Numerical test: Langevin dynamics in a tilted periodic potential

Numerical and experimental tests

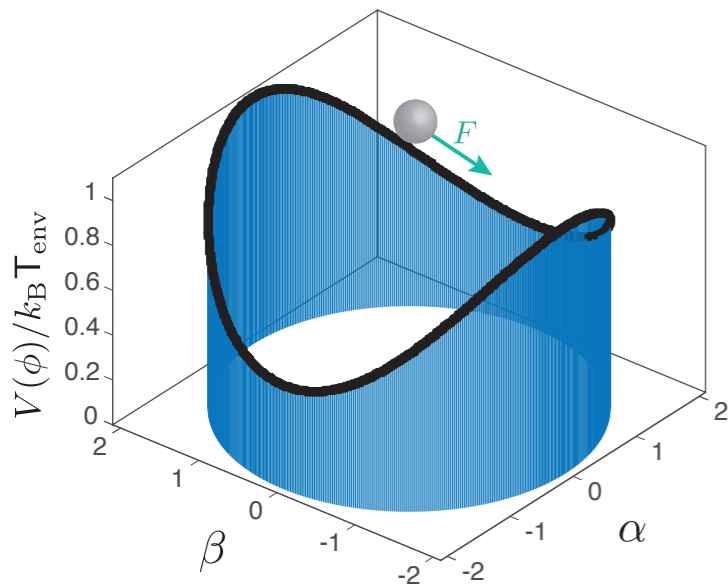
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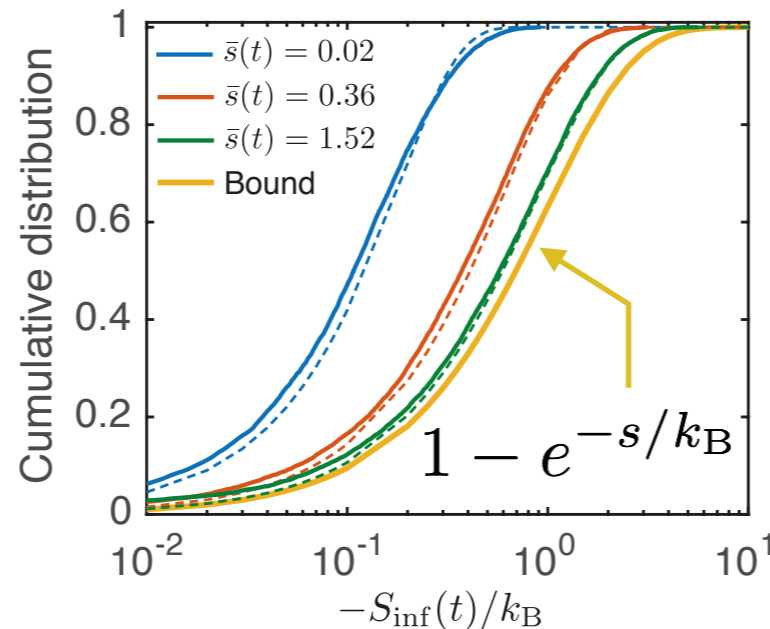
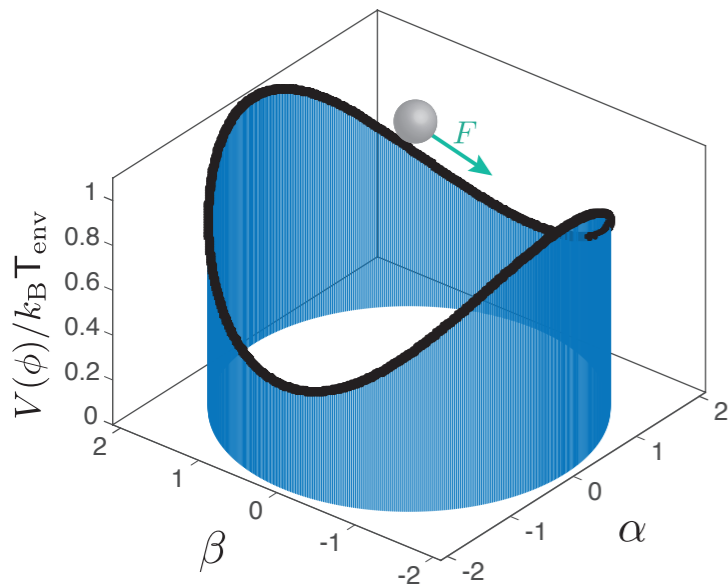
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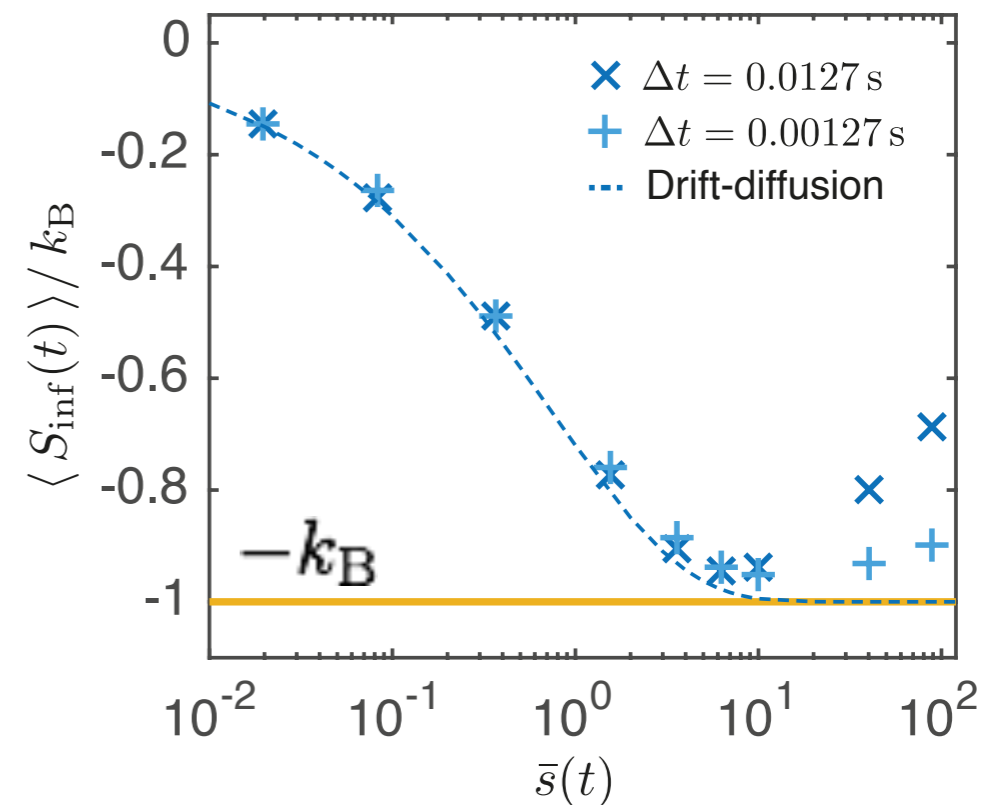
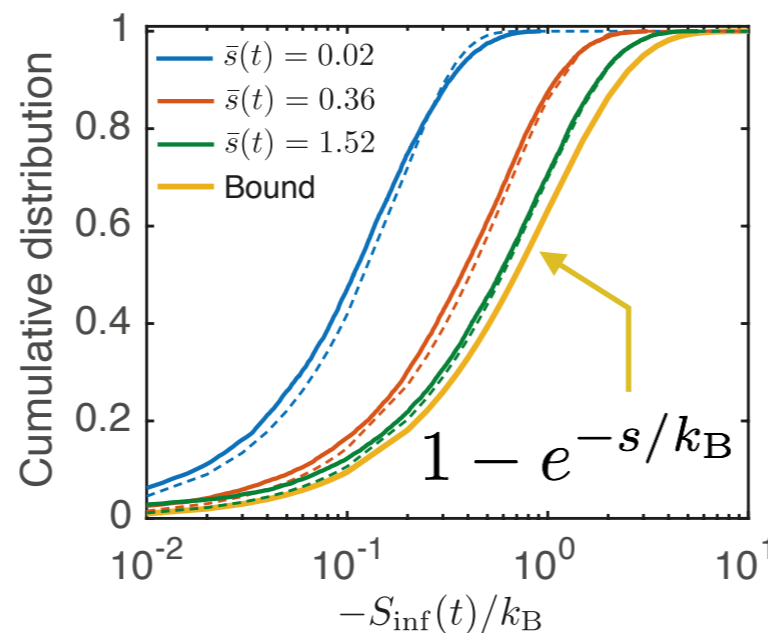
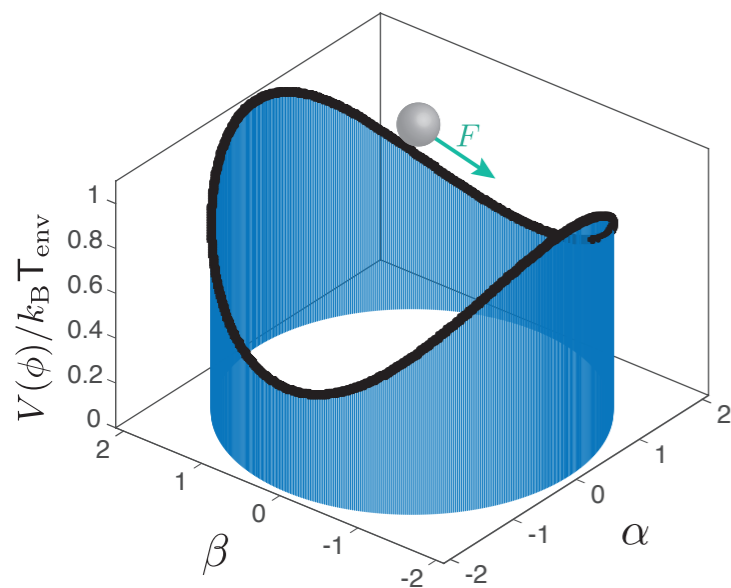
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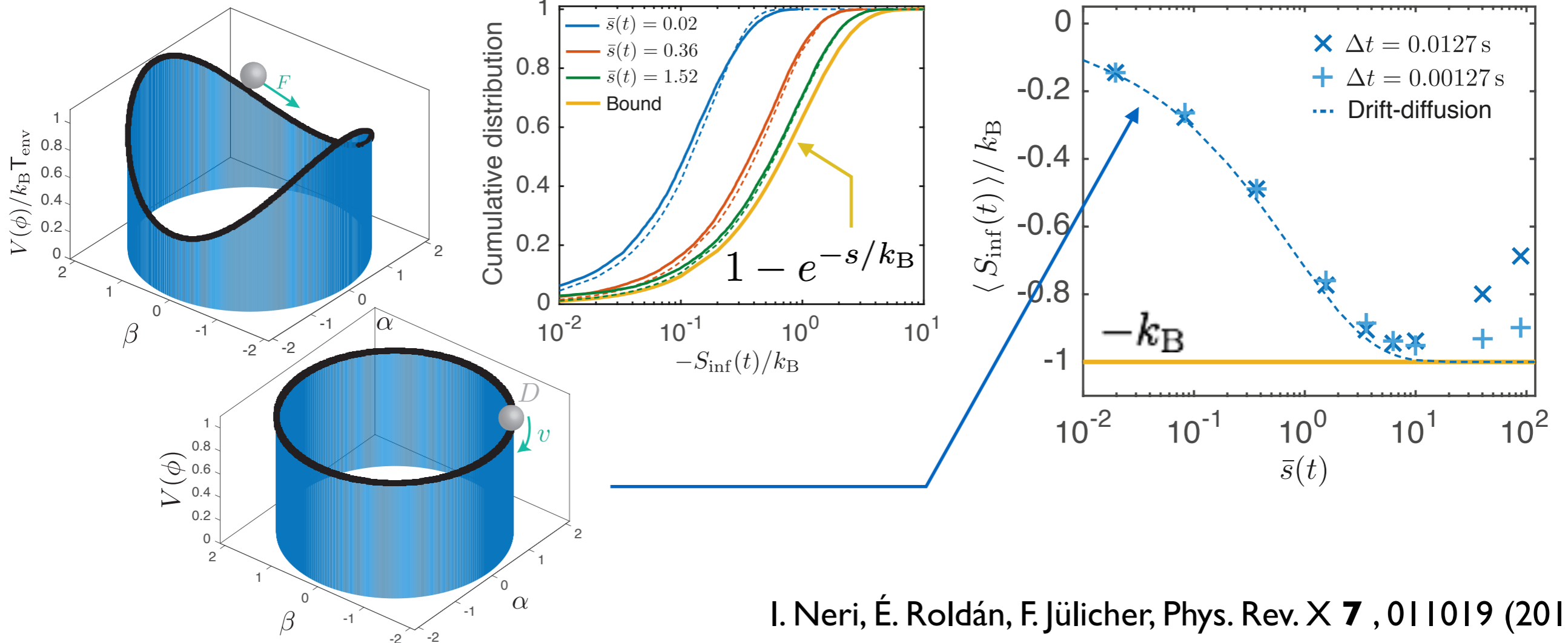
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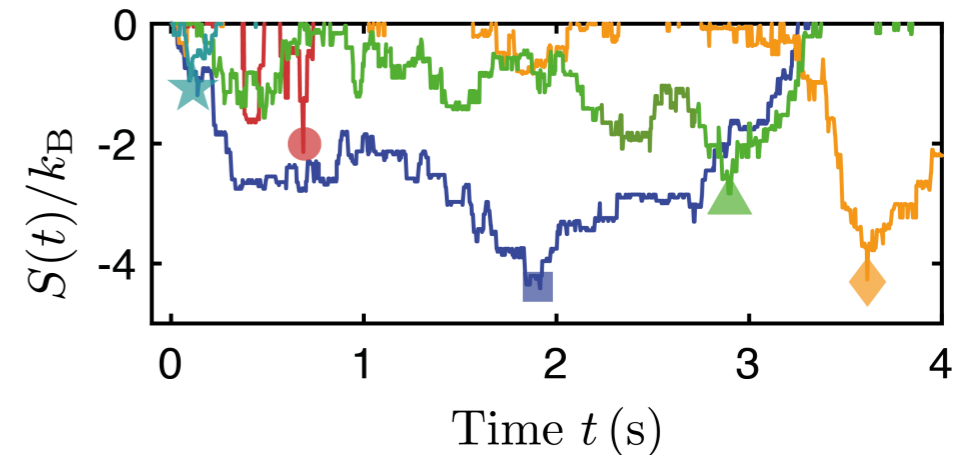
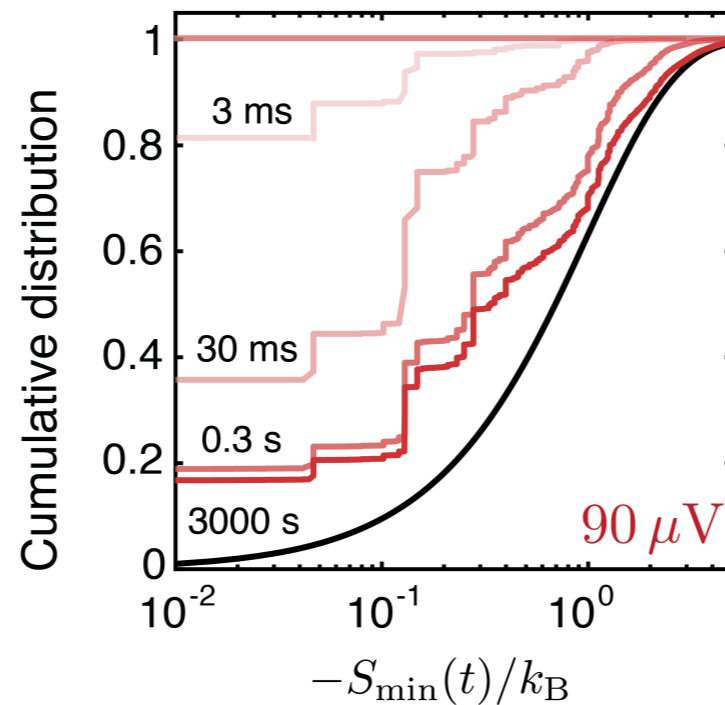
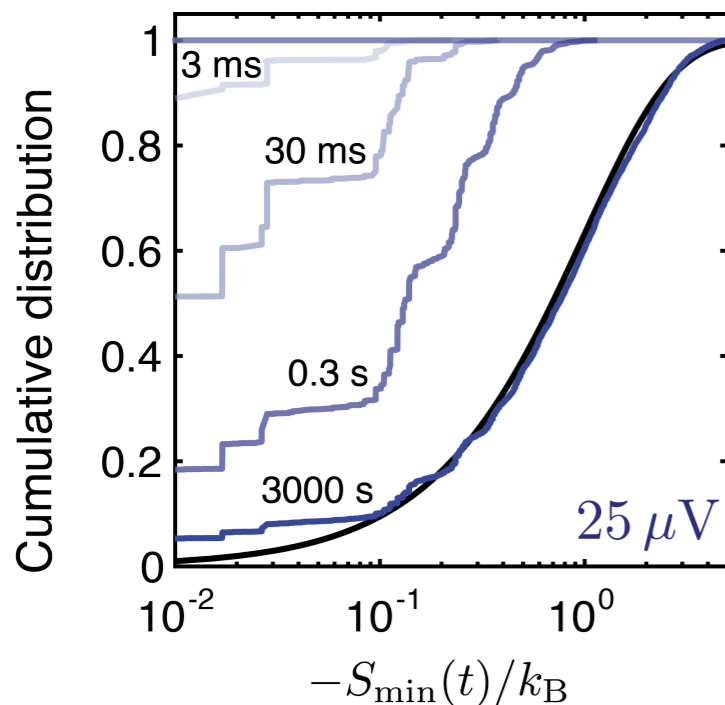
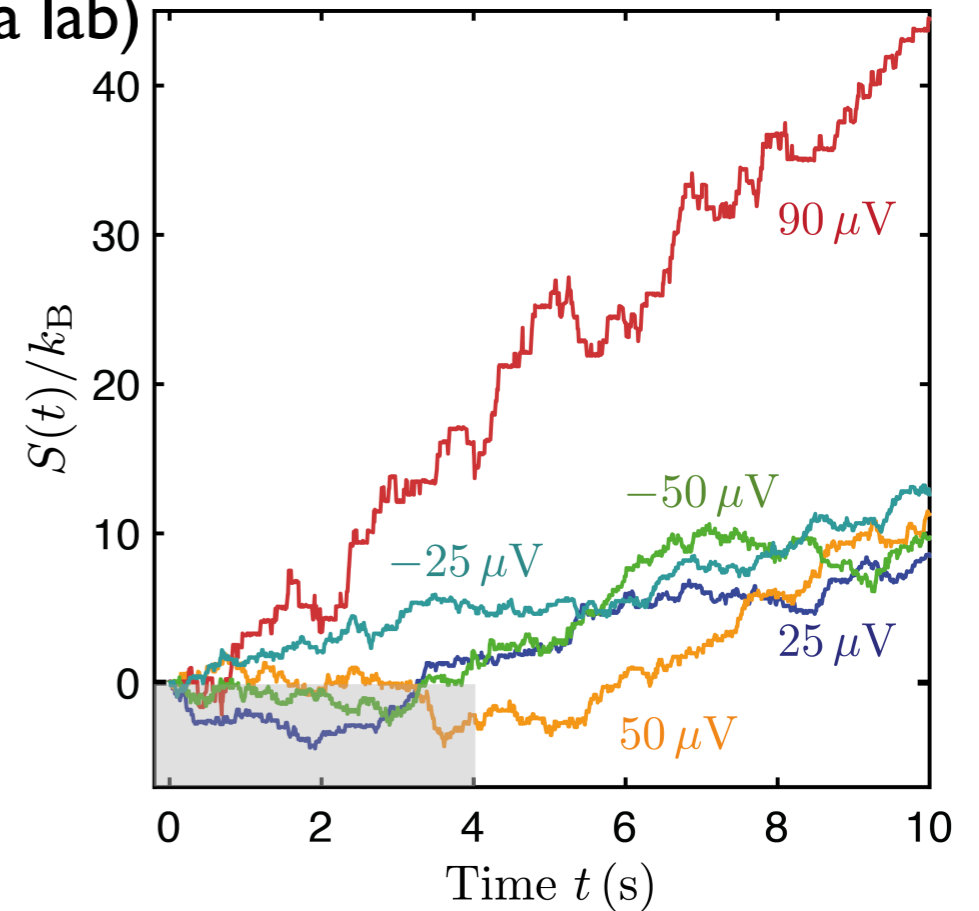
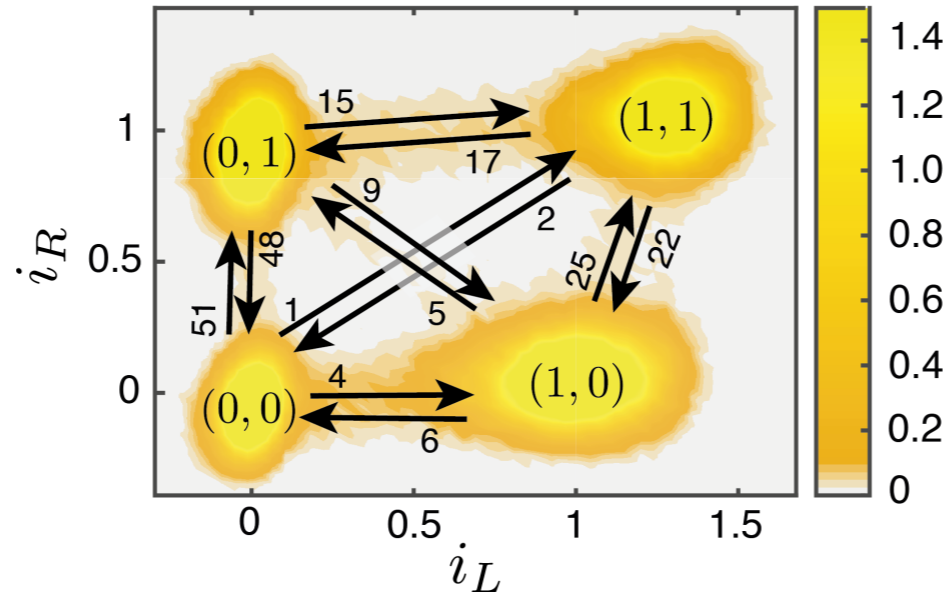
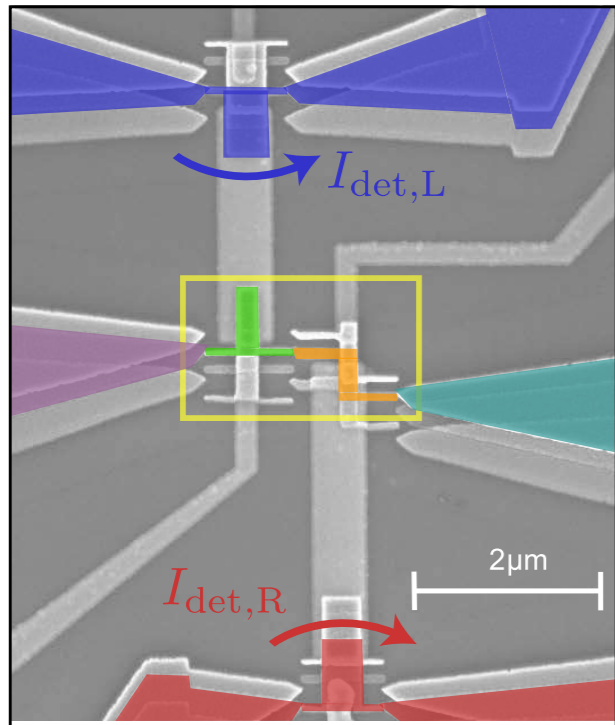
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Numerical test: Langevin dynamics in a tilted periodic potential



Numerical and experimental tests

Experimental test: Electronic double dot (Jukka Pekola lab)



***Langevin* dynamics:**

**universal (martingale) properties
of entropy production**

Langevin processes

Ito Langevin equation with multiplicative noise

$$\frac{d\vec{X}}{dt} = \mu \cdot \vec{F} + \vec{\nabla} \cdot D + \sqrt{2} \sigma \cdot \vec{\xi}$$

Force

$$\vec{F} = -\vec{\nabla}U(\vec{X}(t), t) + \vec{f}(\vec{X}(t), t)$$

conservative non-conservative

Diffusion coefficient

$$\sigma(\vec{X}(t))\sigma(\vec{X}(t))^T = D(\vec{X}(t))$$

$$D(\vec{X}(t)) = \mu(\vec{X}(t)) k_B T \quad \text{Einstein relation}$$

Smoluchowski's Equation

$$\partial_t P = -\vec{\nabla} \cdot \vec{J}$$

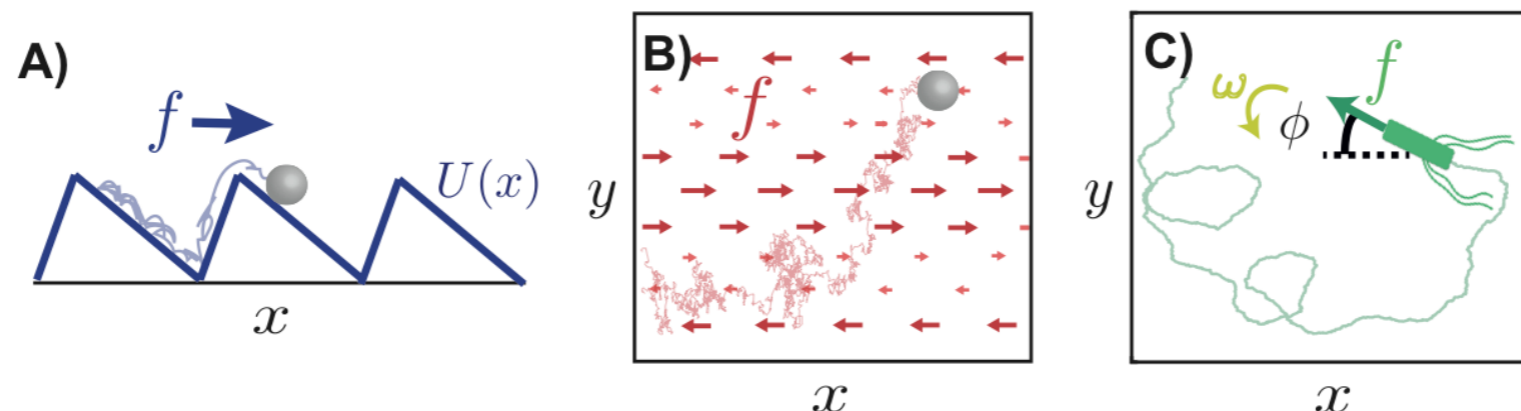
$$\vec{J} = \mu \cdot \vec{F} P - D \cdot \vec{\nabla} P$$

Noise

$$\langle \xi_i(t) \rangle = 0 \quad \text{Gaussian white noise}$$

$$\langle \xi_i(t) \xi_j(t') \rangle = \delta_{ij} \delta(t - t')$$

M. v. Smoluchowski, Ann. d. Phys. **21**, 756 (1906)



S. Pigolotti, I. Neri, É. Roldán, F. Jülicher, Phys. Rev. Lett. **119** (14), 140604 (2017)

Martingality in Langevin processes

Ito Langevin equation for stochastic entropy production

$$\begin{array}{l} \text{non-steady state} \\ \text{steady states} \end{array} \quad \frac{dS_{\text{tot}}}{dt} = -2k_B \partial_t \ln P + v_S + \sqrt{2k_B v_S} \xi_S$$
$$\frac{dS_{\text{tot}}}{dt} = v_S + \sqrt{2k_B v_S} \xi_S$$

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steady states $\frac{dS_{\text{tot}}}{dt} = v_S + \sqrt{2k_B v_S} \xi_S$

Entropic drift

$$v_S = k_B \frac{\vec{J} \cdot D^{-1} \cdot \vec{J}}{P^2}$$

Entropic noise

$$\begin{aligned} \langle \xi_S(t) \rangle &= 0 \\ \langle \xi_S(t) \xi_S(t') \rangle &= \delta(t - t') \end{aligned}$$

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$$\langle \xi_S(t) \xi_S(t') \rangle = \delta(t - t')$$

$e^{-S_{\text{tot}}(t)/k_B}$ is a Geometric Brownian motion with zero drift (**Martingale**)

$$\frac{de^{-S_{\text{tot}}/k_B}}{dt} = -\sqrt{2 \frac{v_S}{k_B}} e^{-S_{\text{tot}}/k_B} \xi_S$$

Langevin fluctuations and time

$$\frac{d\vec{X}}{dt} = \mu \cdot \vec{F} + \vec{\nabla} \cdot D + \sqrt{2} \sigma \cdot \vec{\xi}$$

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“Entropic” time

$$v_S = k_B \frac{\vec{J} \cdot D^{-1} \cdot \vec{J}}{P^2}$$

$$\tau(t) = \frac{1}{k_B} \int_0^t v_S(X(t')) dt'$$

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Random-time change $t \rightarrow \tau(t)$

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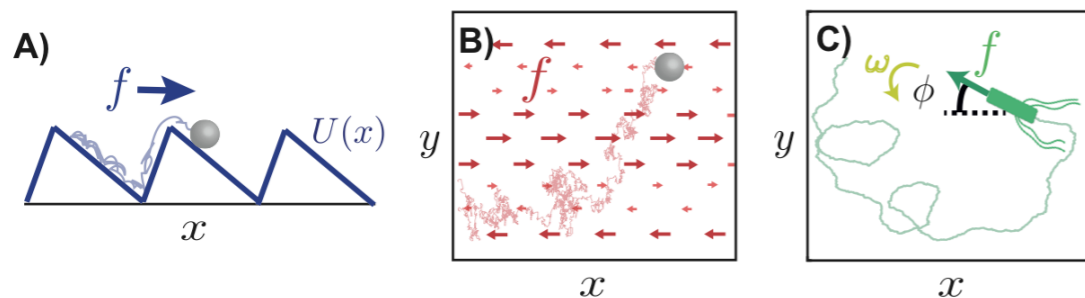
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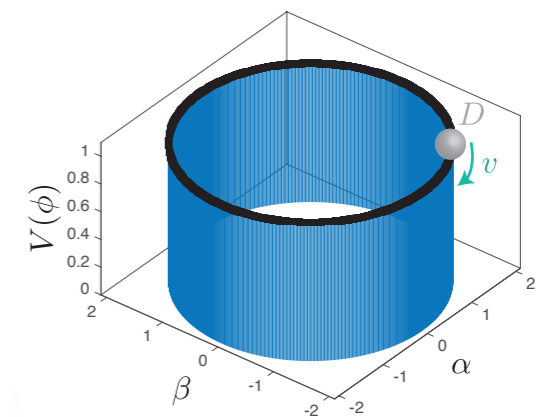
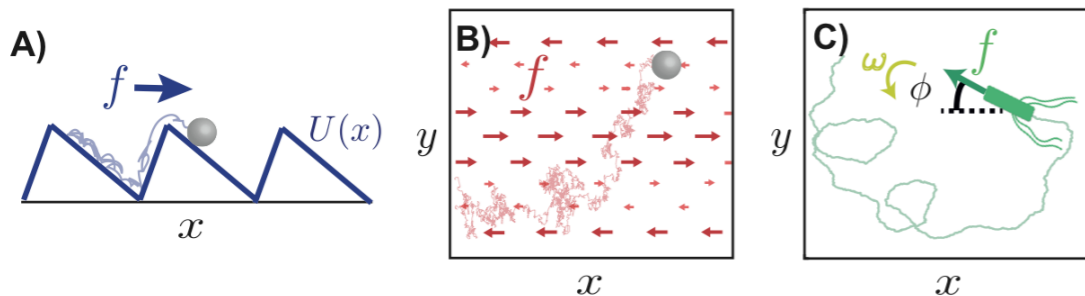
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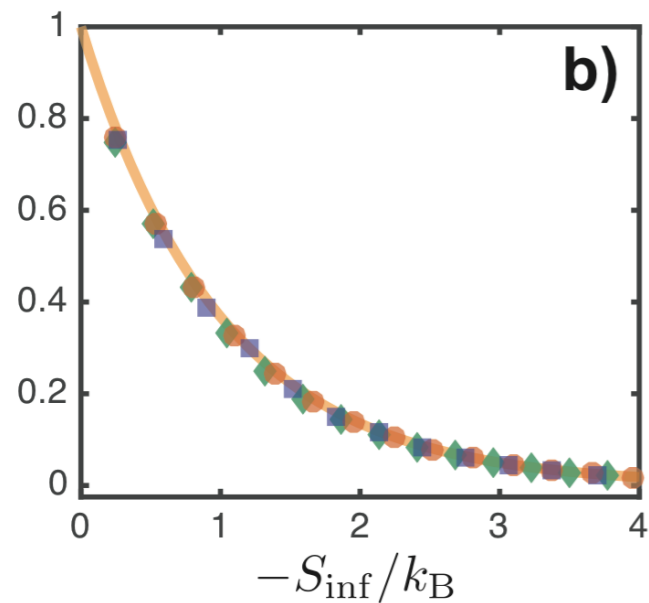
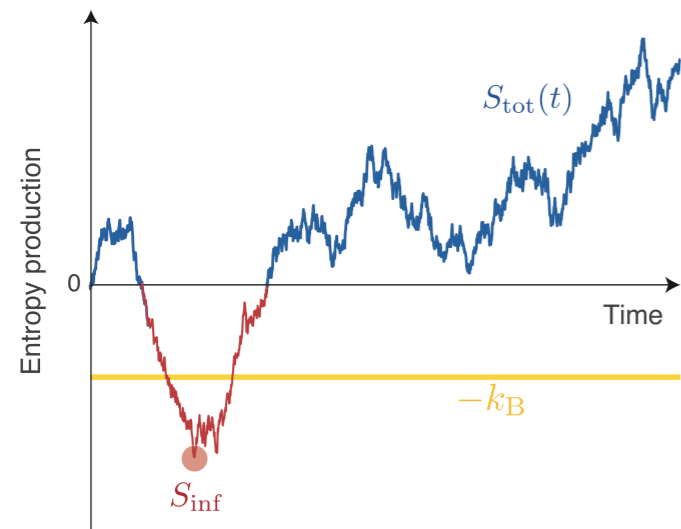
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Generic properties of entropy production

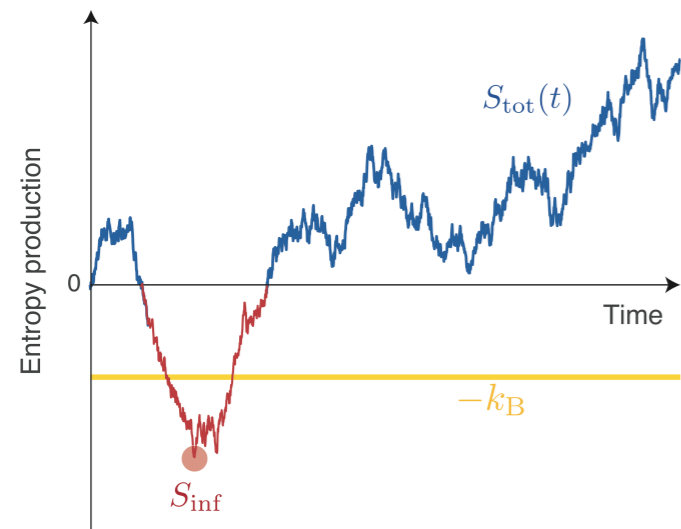
Global infimum



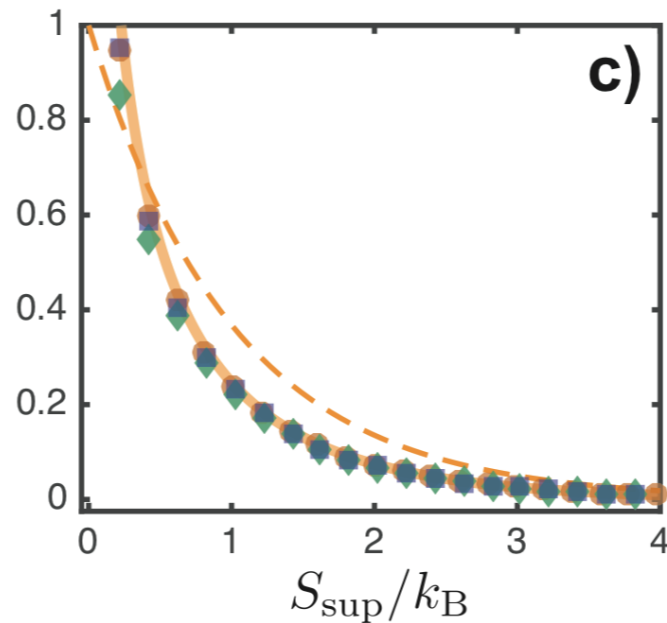
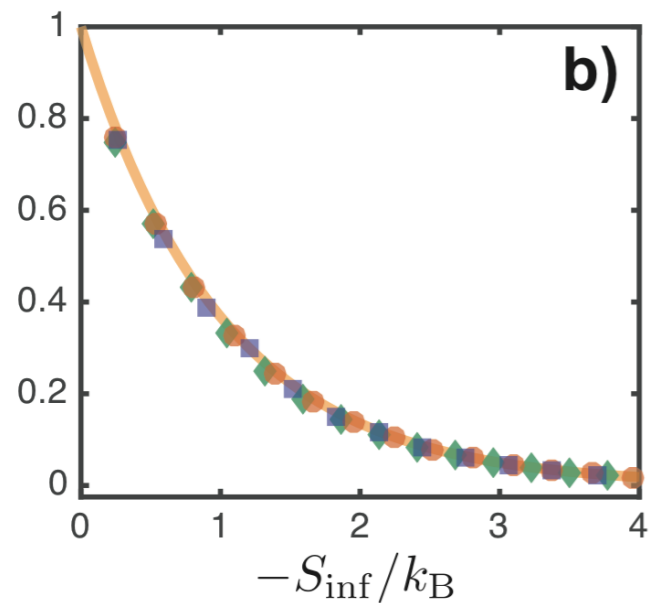
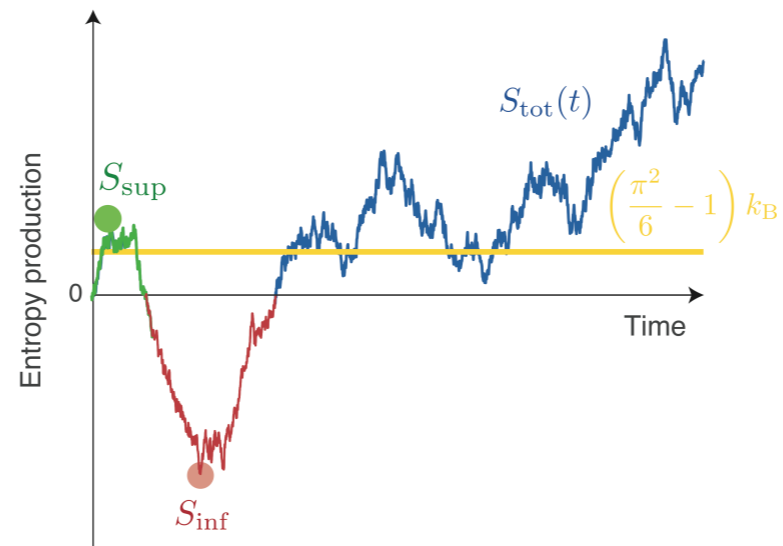
$$P(S_{\text{inf}}) = e^{S_{\text{inf}}/k_B} / k_B$$

Generic properties of entropy production

Global infimum



Supremum before the global infimum

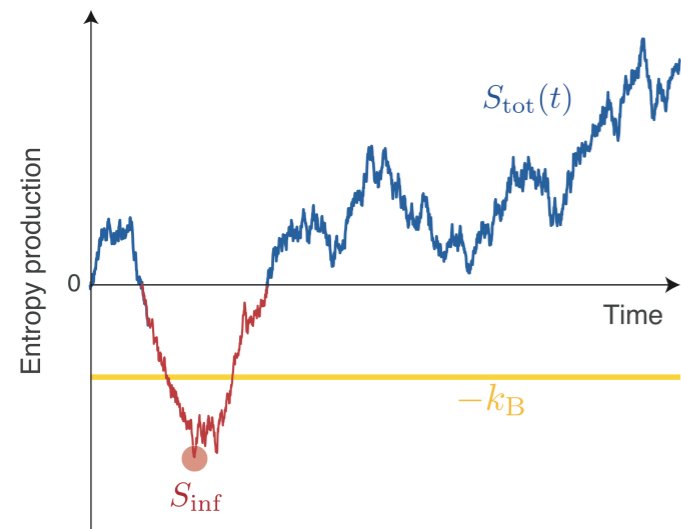


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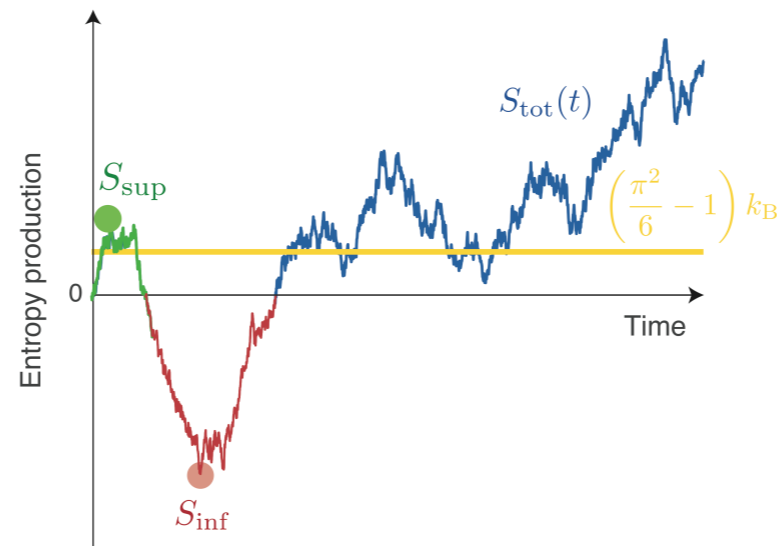
$$P(S_{\text{sup}}) = 2e^{S_{\text{sup}}/k_B} \operatorname{acoth}(2e^{S_{\text{sup}}/k_B} - 1) - 1$$

Generic properties of entropy production

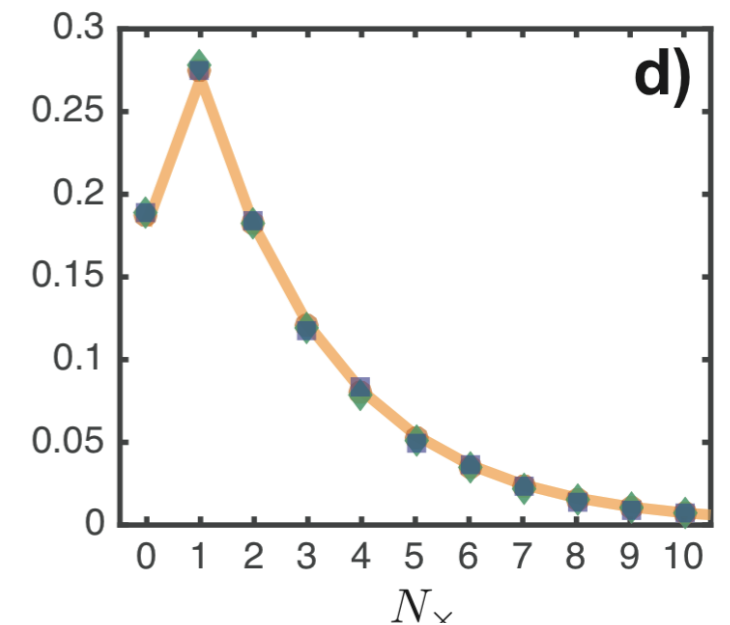
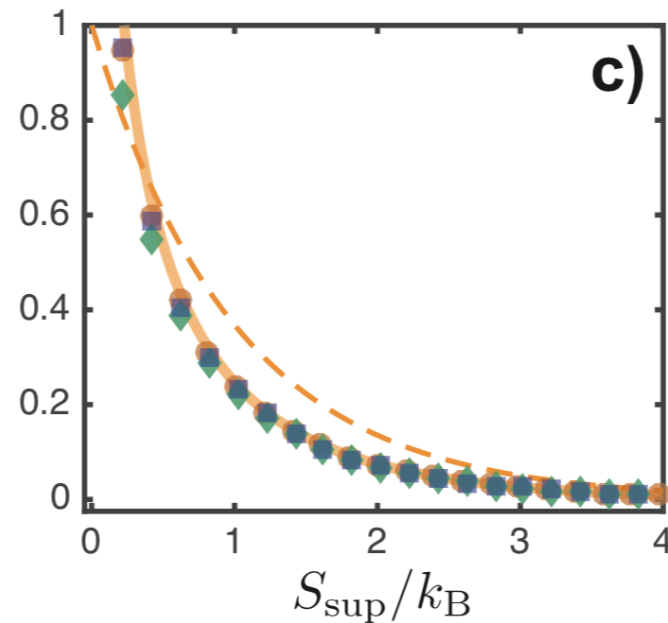
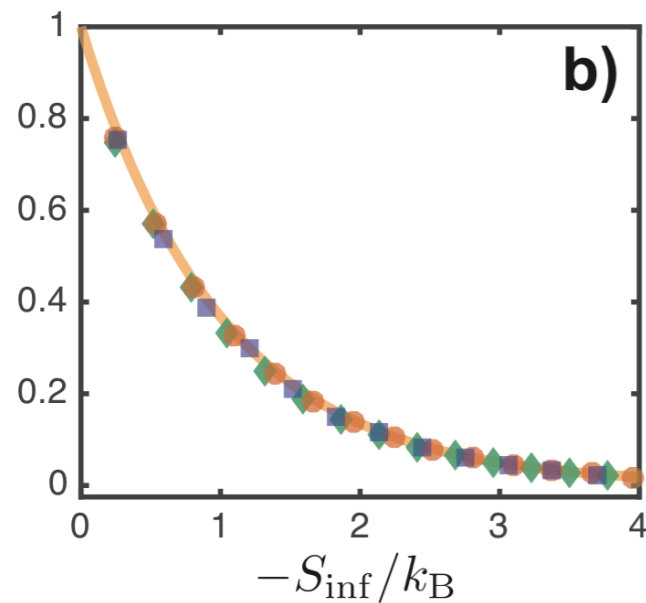
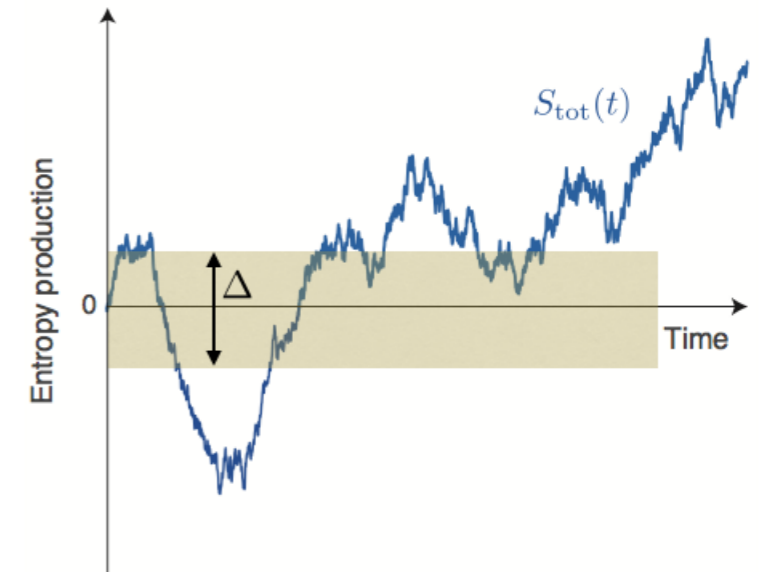
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Supremum before the global infimum



Number of crossings



$$P(S_{\text{inf}}) = e^{S_{\text{inf}}/k_B} / k_B$$

$$P(S_{\text{sup}}) = 2e^{S_{\text{sup}}/k_B} \text{acoth}(2e^{S_{\text{sup}}/k_B} - 1) - 1$$

$$P(N_{\times}; \Delta) = \begin{cases} 1 - e^{-\Delta/k_B} & N_{\times} = 0 \\ 2 \sinh(\Delta/k_B) e^{-2N_{\times} \Delta/k_B} & N_{\times} \geq 1 \end{cases}$$

Non-generic properties: Uncertainty equality

Finite-time **uncertainty equality** for entropy production of nonequilibrium steady-state* Langevin processes

$$\frac{1}{k_B} \frac{\sigma_{S_{\text{tot}}}^2(t)}{\langle S_{\text{tot}}(t) \rangle} = 2 + \frac{\sigma_{\tau}^2(t)}{\langle \tau(t) \rangle}$$

* also for non-steady state

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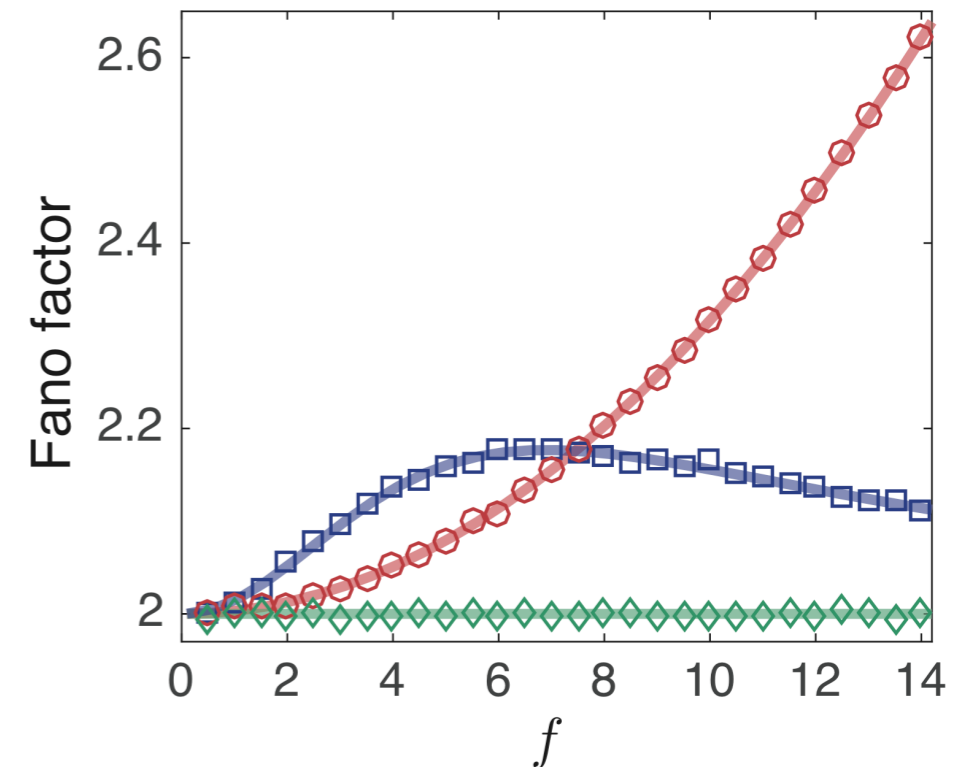
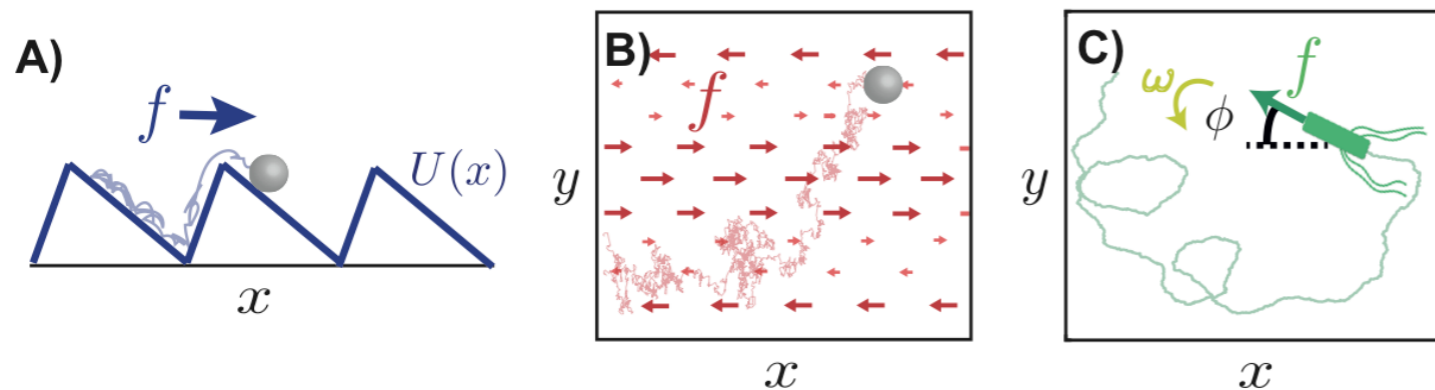
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A. C. Barato, U. Seifert, Phys. Rev. Lett. **114**, 158101 (2015)

T. R. Gingrich, et. al, Phys. Rev. Lett. **116**, 120601 (2016)



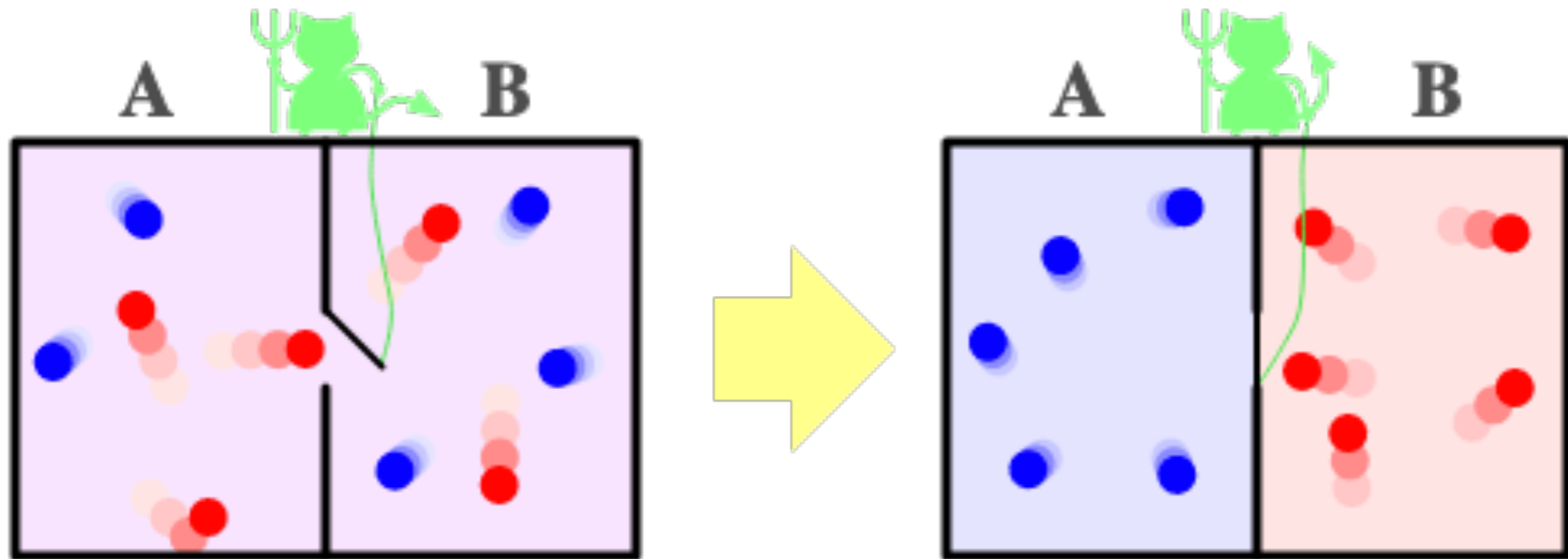
* also for non-steady state

S. Pigolotti, I. Neri, É. Roldán, F. Jülicher, Phys. Rev. Lett. **119** (14), 140604 (2017)

**Martingales out of steady
states:**

Gambling demons

Revisiting Maxwell's demons



“Seemingly-violation” of the second law

Information



System

Demon



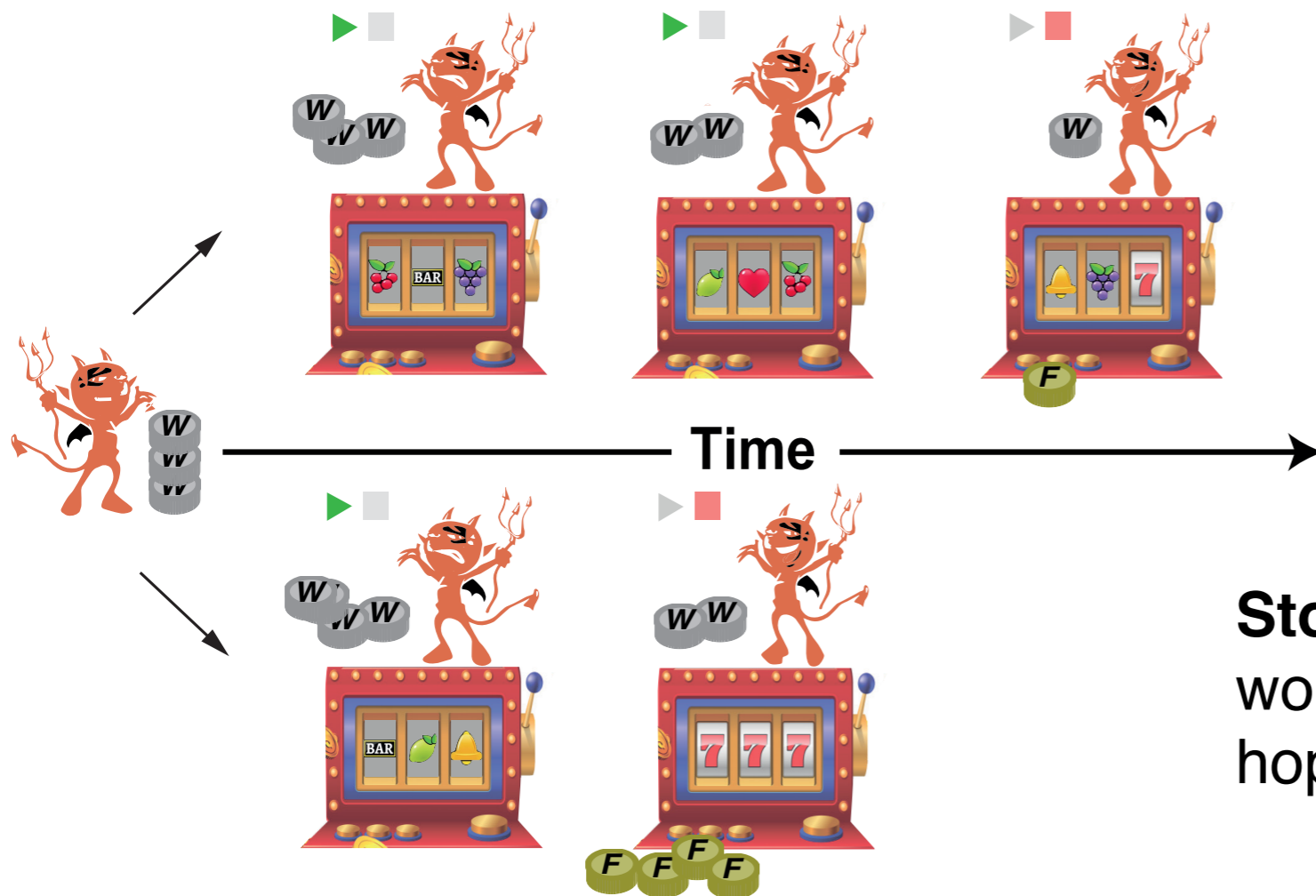
Feedback

Feedback control depending on measurement outcome

Stochastic times for opening/closing of the gate

Gambling demons

A “demon” invests work performing a nonequilibrium process and decides whether to stop the process or not a stopping (gambling) strategy

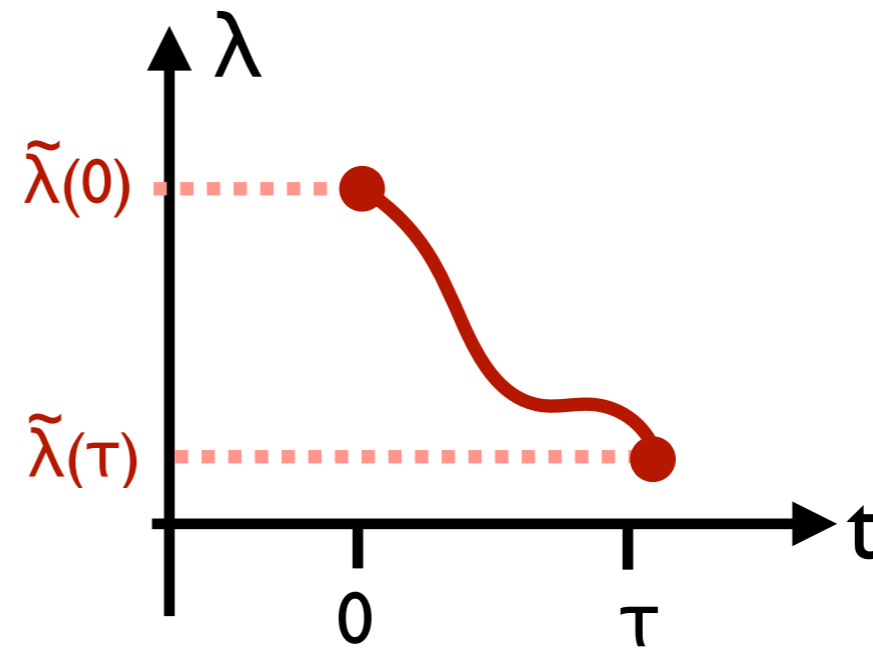
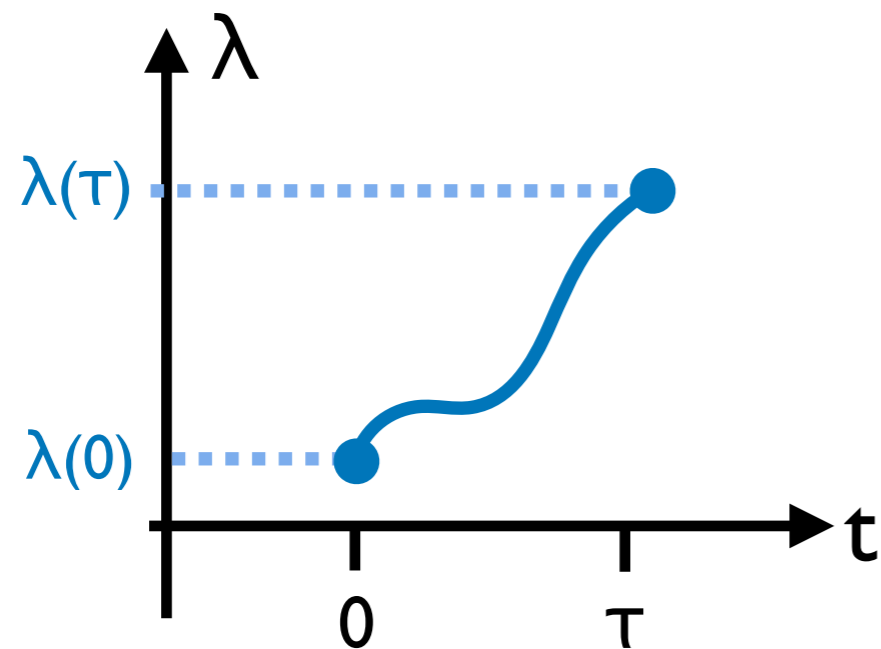


No feedback control of system's dynamics

Stochastic times: the gambler invests work (coins) in a slot machine (system) hoping to obtain a positive payoff

Goal: Extract heat from a thermal bath by conditionally stopping the system with a clever strategy

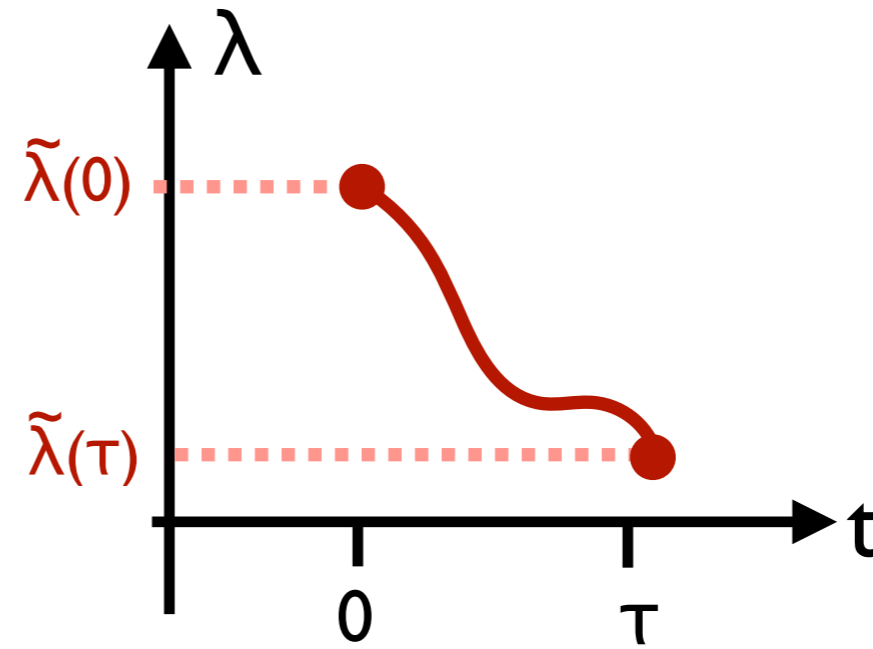
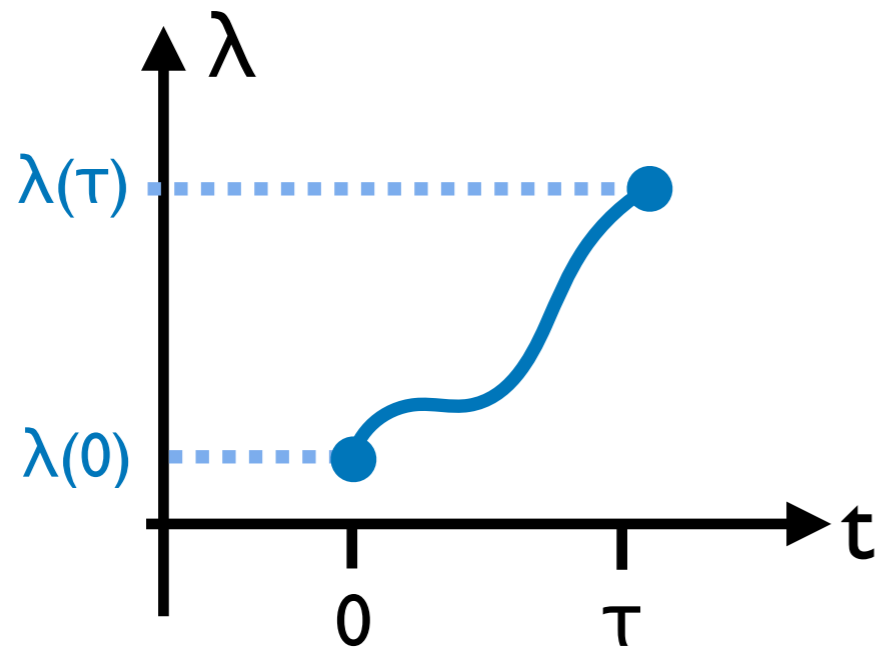
Martingale theory for non-stationary driving



“Forward” process: Drive $\lambda(t)$, initial state $\rho(x, 0)$, final $\rho(x, \tau)$

“Backward” process: Drive $\tilde{\lambda}(t) = \lambda(\tau - t)$, initial $\tilde{\rho}(x, 0)$, final $\tilde{\rho}(x, \tau)$

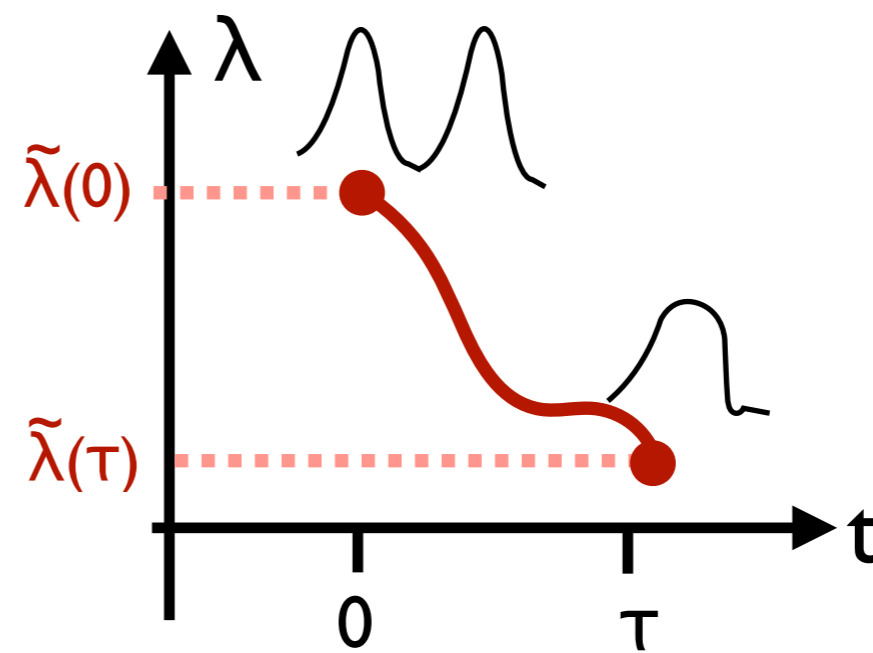
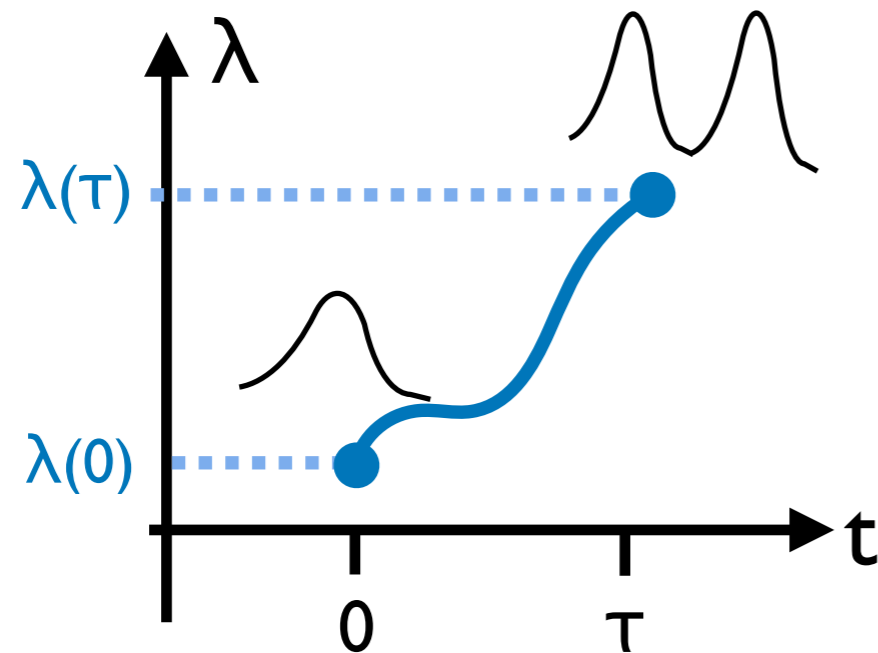
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Martingale theory for non-stationary driving

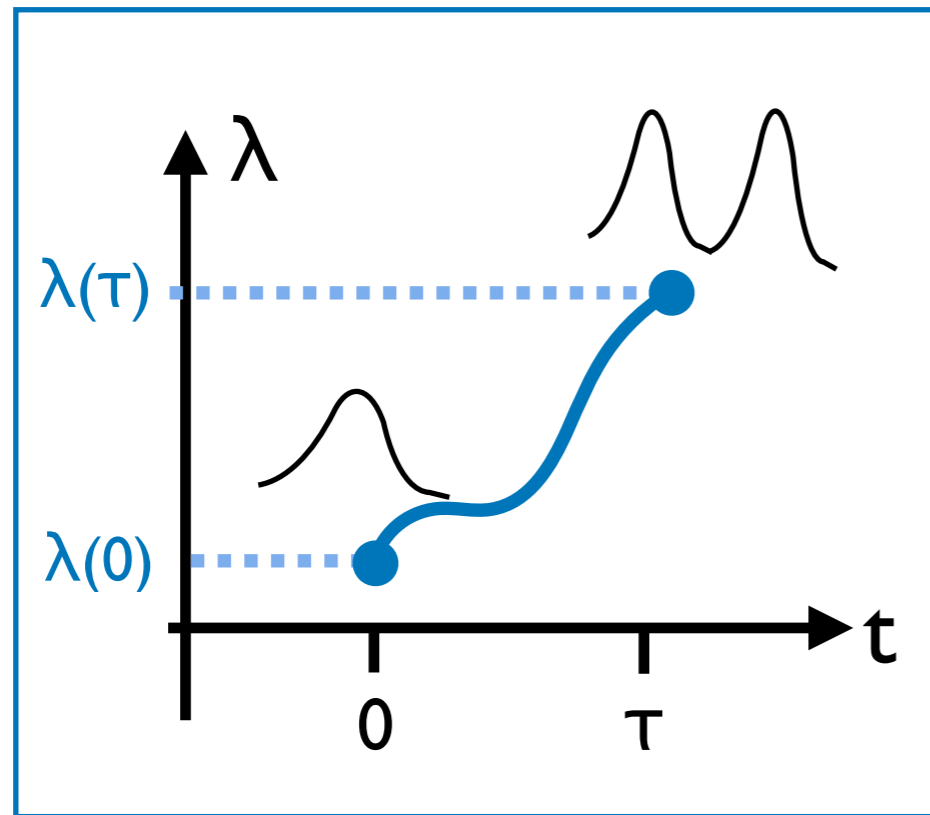


“Forward” process: Drive $\lambda(t)$, initial state $\rho(x, 0)$, final $\rho(x, \tau)$

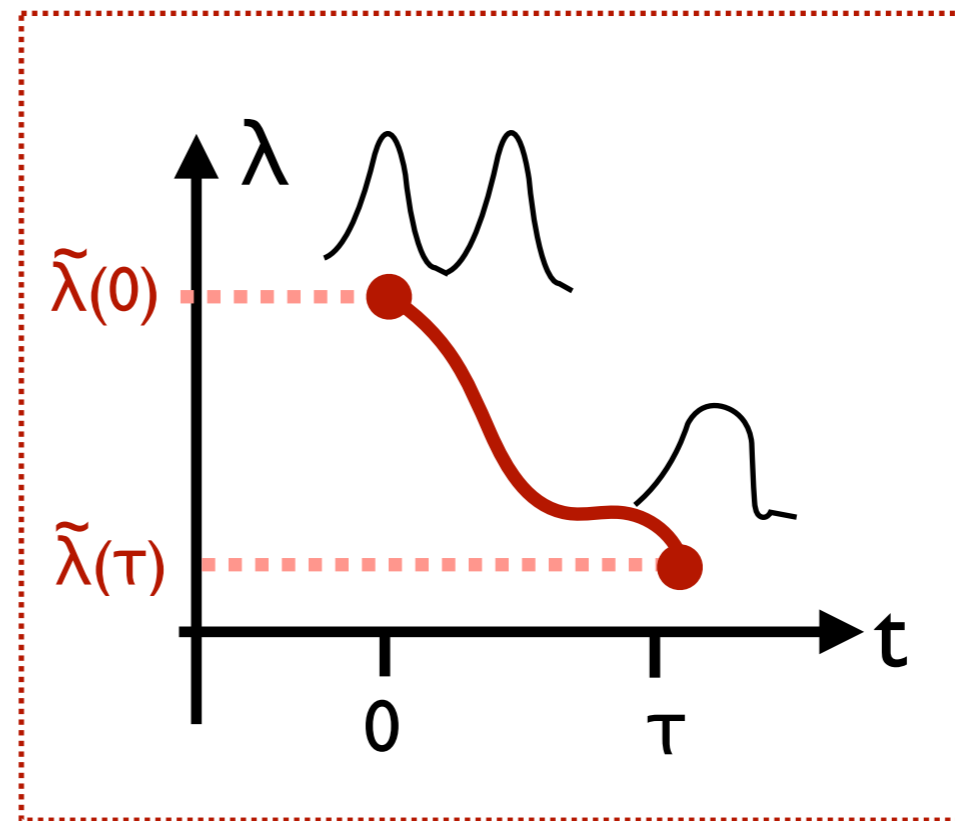
“Backward” process: Drive $\tilde{\lambda}(t) = \lambda(\tau - t)$, initial $\tilde{\rho}(x, 0) = \rho(x, \tau)$, final $\tilde{\rho}(x, \tau)$

Martingale theory for non-stationary driving

Physical (forward) process



Auxiliary (backward) process



“Forward” process: Drive $\lambda(t)$, initial state $\rho(x, 0)$, final $\rho(x, \tau)$

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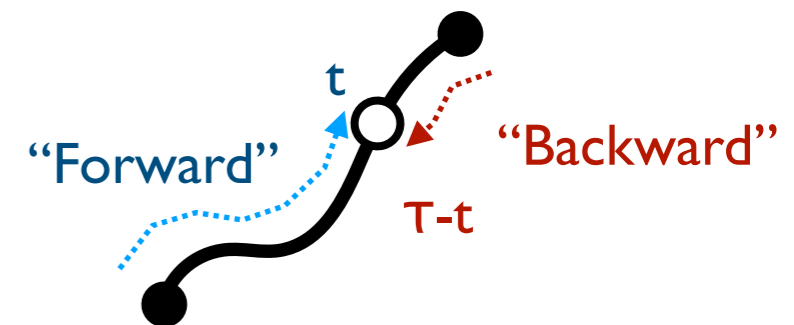
Entropy production is **not** exponential martingale for non-stationary processes:

$$S_{\text{tot}}(t) = k_B \ln \frac{P[X_t]}{\tilde{P}[\tilde{X}_t]}$$

$$\langle e^{-S_{\text{tot}}(t)/k_B} | X_s \rangle = e^{-S_{\text{tot}}(s)/k_B} \frac{\rho(x(s), s)}{\tilde{\rho}(x(s), \tau - s)} = e^{-S_{\text{tot}}(t)/k_B + \delta(s)}$$

Stochastic distinguishability

$$\delta(t) \equiv \ln \left[\frac{\rho(x(t), t)}{\tilde{\rho}(x(t), \tau - t)} \right]$$



Martingale theory for non-stationary driving

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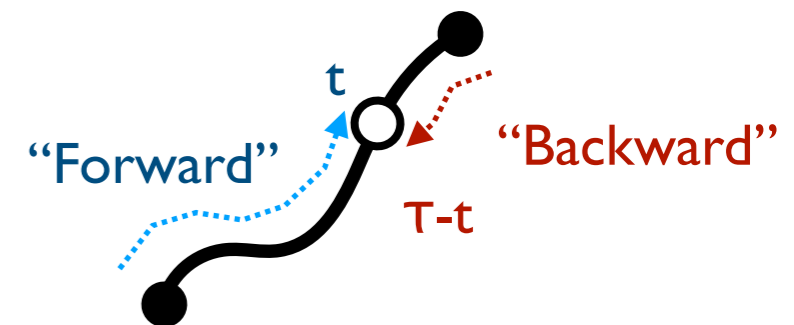
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Martingalization of entropy production

$$\langle e^{-S_{\text{tot}}(t)/k_B - \delta(t)} | X_s \rangle = e^{-S_{\text{tot}}(s)/k_B - \delta(s)}$$

Second law inequality at *stopping times*

Work and free energy in a time interval of (finite) stochastic duration $[0, \mathcal{T}]$

$$W(\mathcal{T}) = \int_0^{\mathcal{T}} \partial_t H(x(t), t) dt$$

$$F(\mathcal{T}) = H(x(\mathcal{T}), \mathcal{T}) - S(\mathcal{T})$$

$$\Delta F(\mathcal{T}) = F(\mathcal{T}) - F(0)$$

Integral FT at stopping times

$$\langle e^{-S_{\text{tot}}(\mathcal{T})/k_B - \delta(\mathcal{T})} \rangle = 1$$

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Jarzynski-like relation at stopping times

$$\langle e^{-\beta[W(\mathcal{T}) - \Delta F(\mathcal{T})] - \delta(\mathcal{T})} \rangle = 1$$

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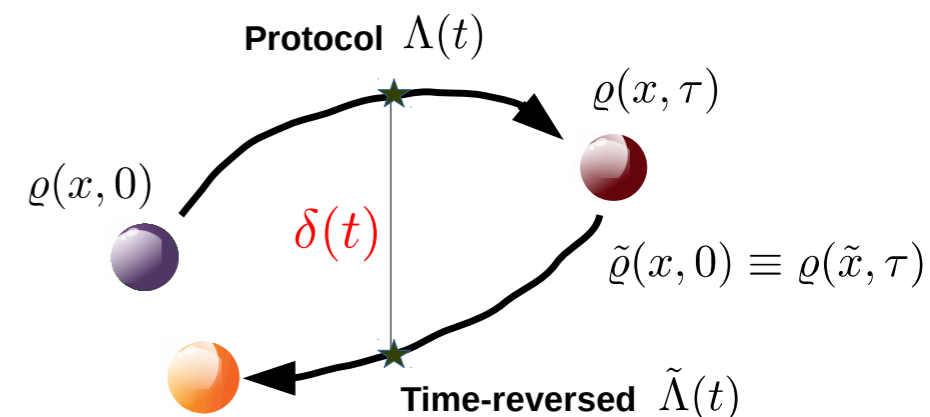
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Second law at stopping times

$$\langle W(\mathcal{T}) \rangle - \langle \Delta F(\mathcal{T}) \rangle \geq -k_B T \langle \delta(\mathcal{T}) \rangle$$

Average distinguishability at stopping times

$$\langle \delta(\mathcal{T}) \rangle = \int d\mathcal{T} \int dx \varrho(x, \mathcal{T}) \ln \frac{\varrho(x, \mathcal{T})}{\tilde{\varrho}(x, \tau - \mathcal{T})} \geq 0$$



G Manzano, et al., Phys. Rev. Lett. **126** (8), 080603 (2021)

Initial equilibrium: I. Neri, Phys. Rev. Lett. **124** (4) 040601 (2020)

Second law at **fixed times**

$$\varrho(\mathcal{T}) = \delta(\mathcal{T} - \tau)$$

Integral FT [Seifert, PRL 2005]

$$\langle e^{-S_{\text{tot}}(\tau)/k_B} \rangle = 1$$

Jarzynski equality [Jarzynski, PRL 1997]

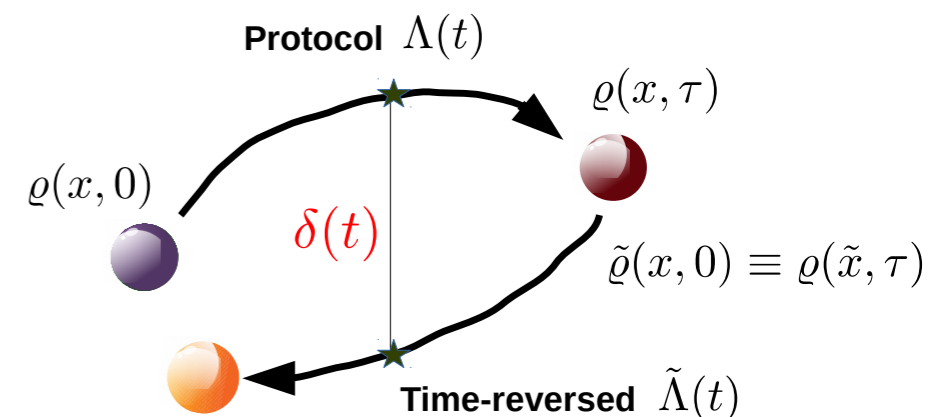
$$\langle e^{-\beta[W(\tau) - \Delta F(\tau)]} \rangle = 1$$

Second law at fixed times

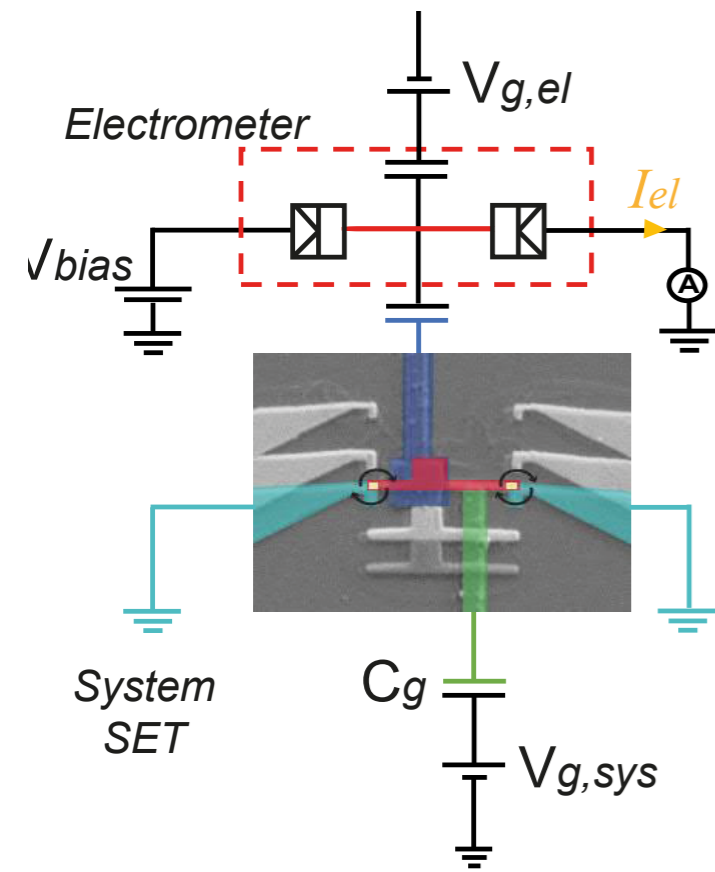
$$\langle W(\mathcal{T}) \rangle - \langle \Delta F(\mathcal{T}) \rangle \geq 0$$

Distinguishability at fixed times

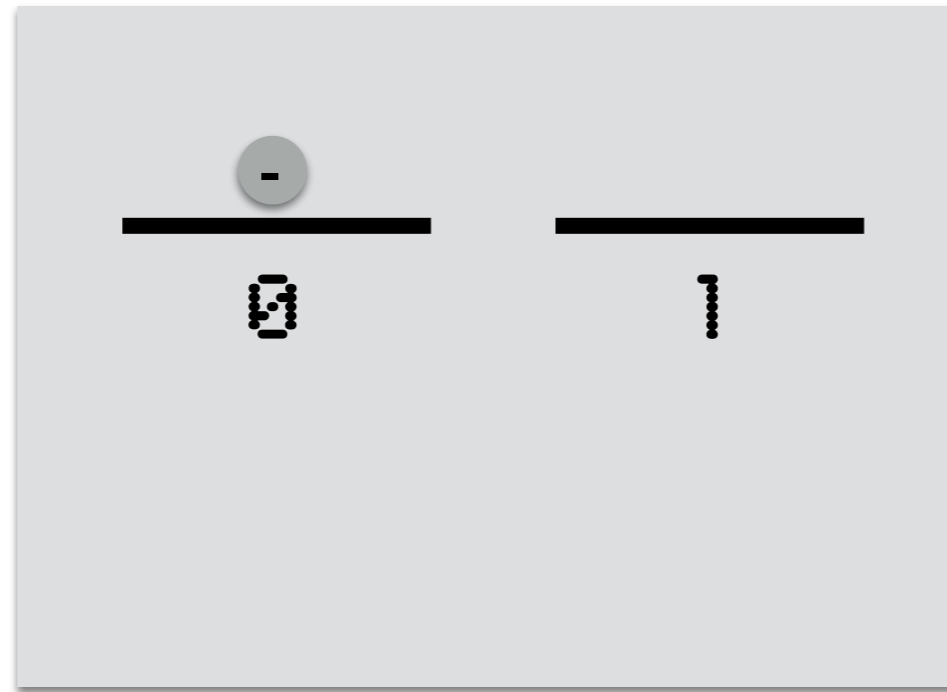
$$\langle \delta(\tau) \rangle = \int dx \varrho(x, \tau) \ln \frac{\varrho(x, \tau)}{\tilde{\varrho}(x, 0)} = 0$$



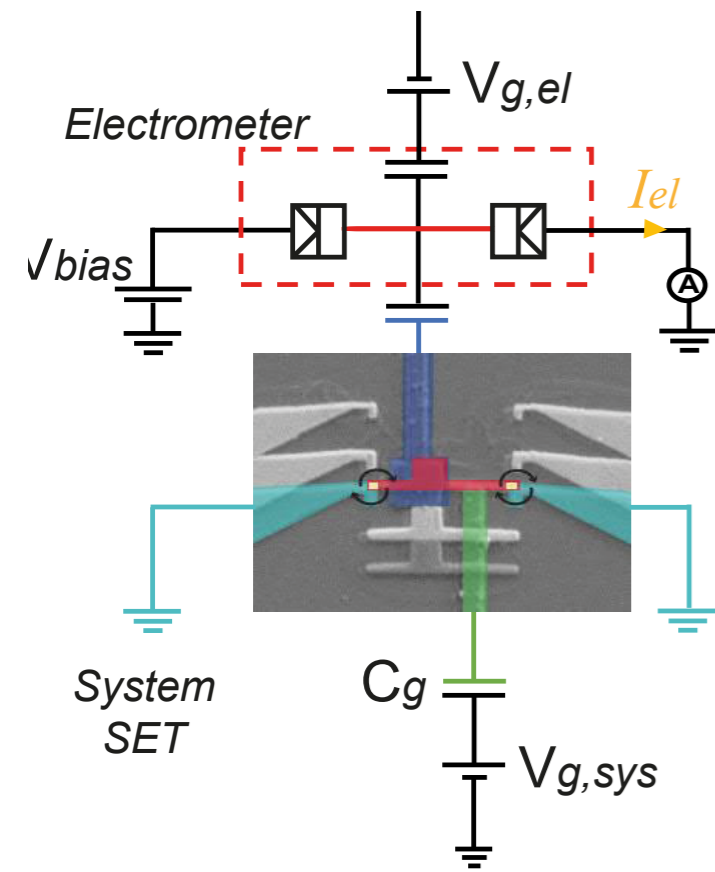
Gambling with a single electron



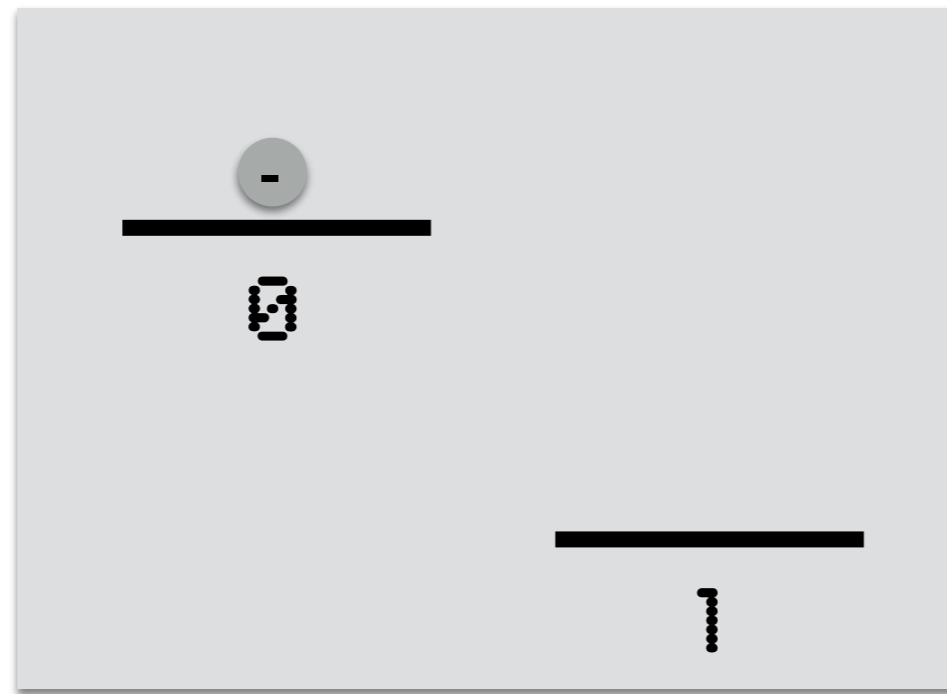
Experimental realization: Single-electron box (Pekola lab)



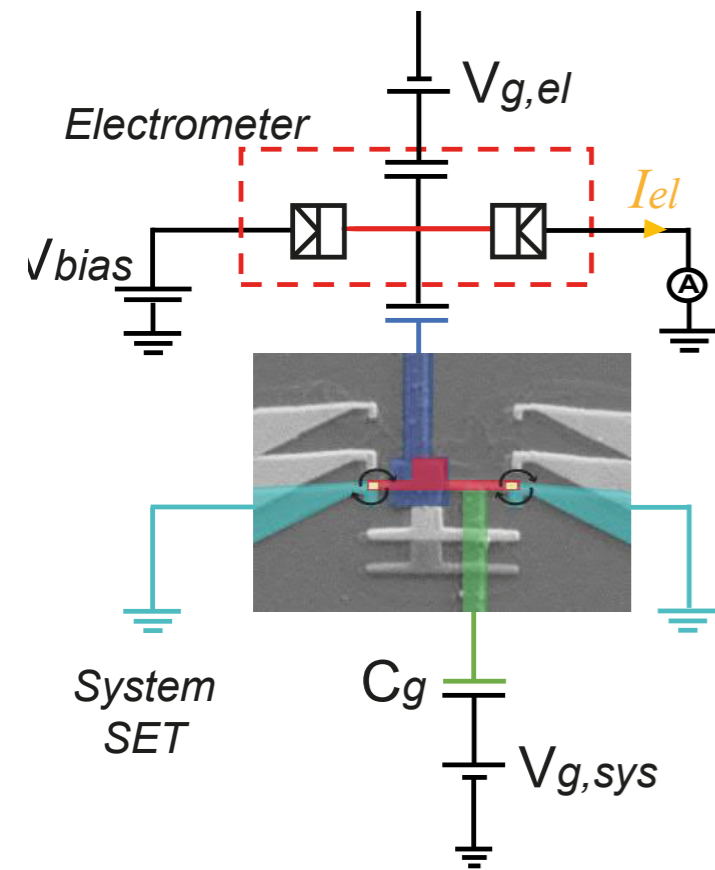
Gambling with a single electron



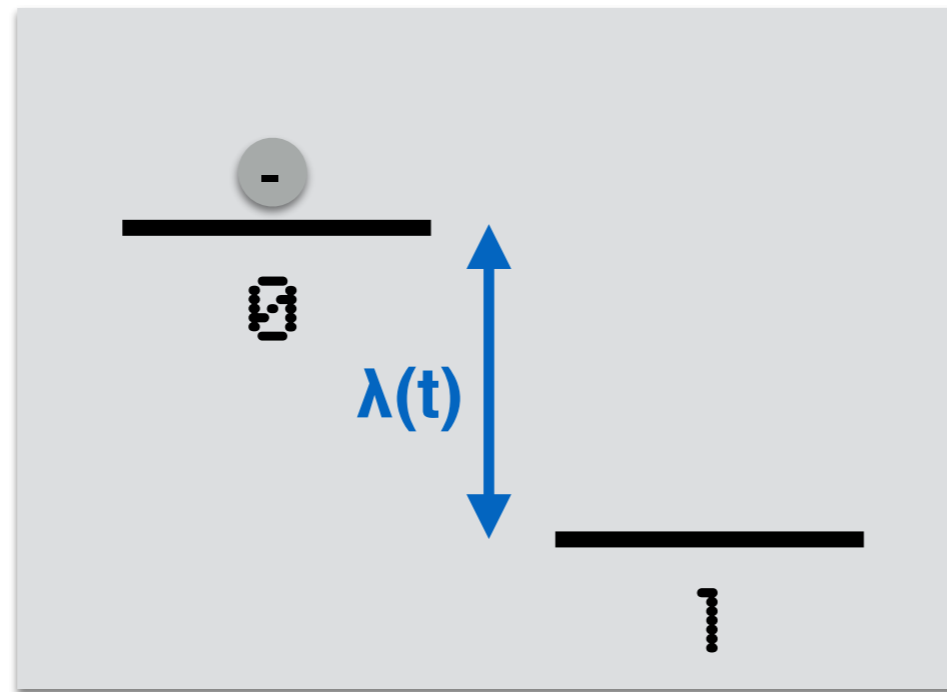
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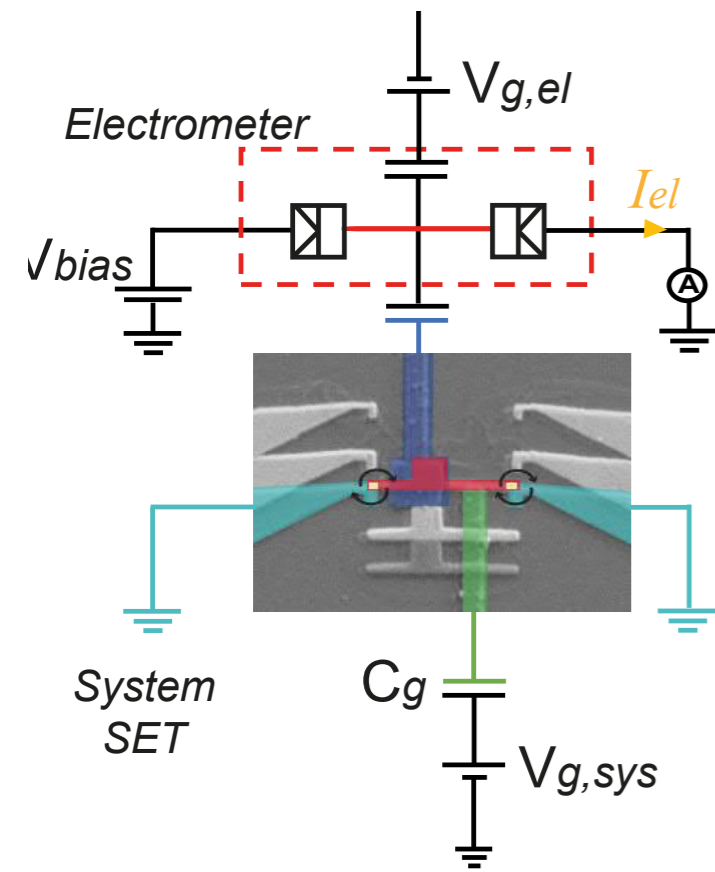
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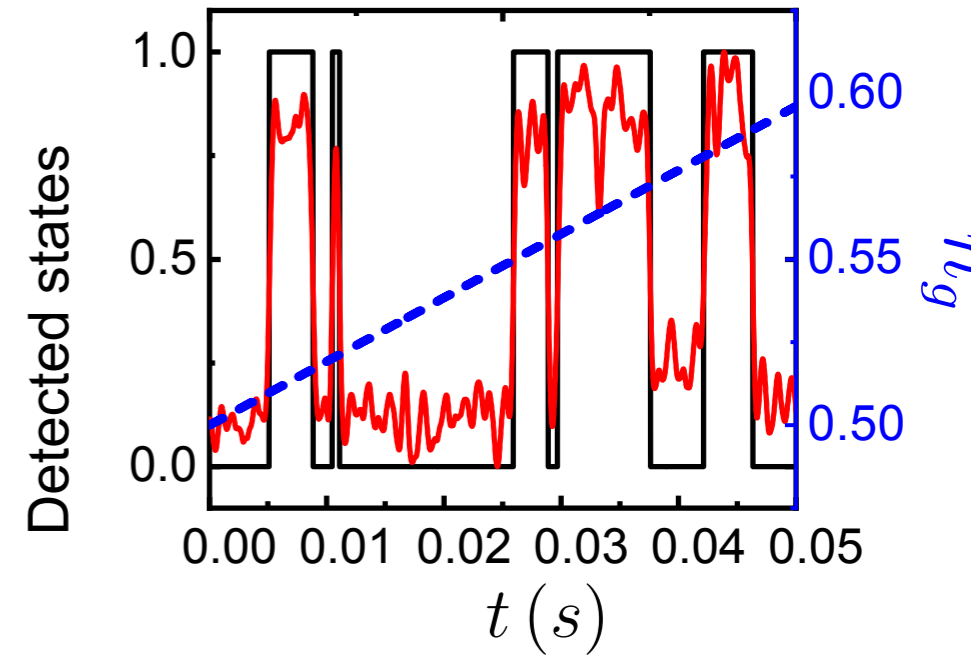
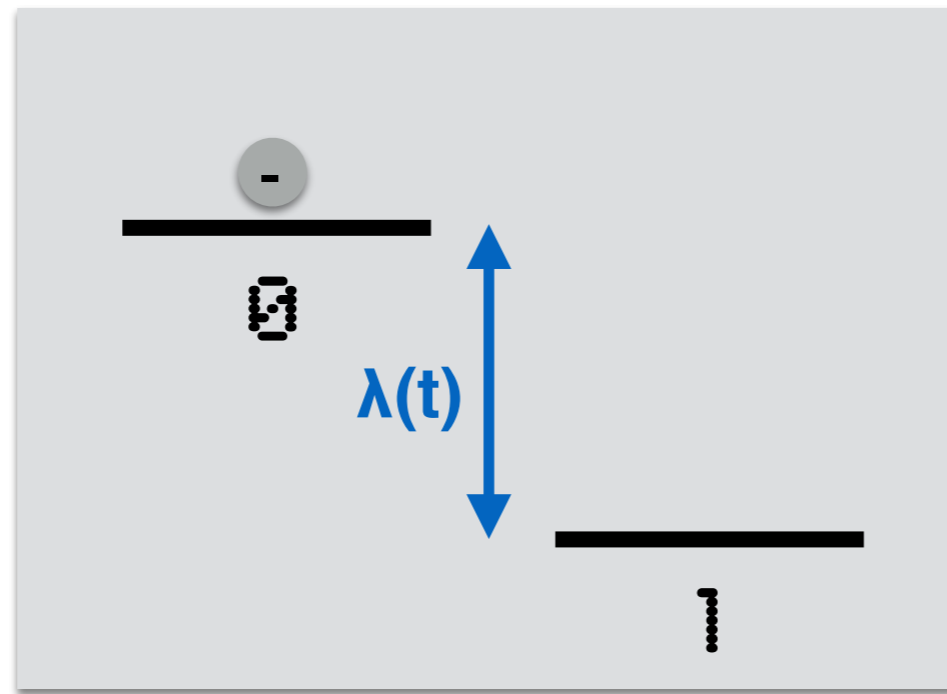
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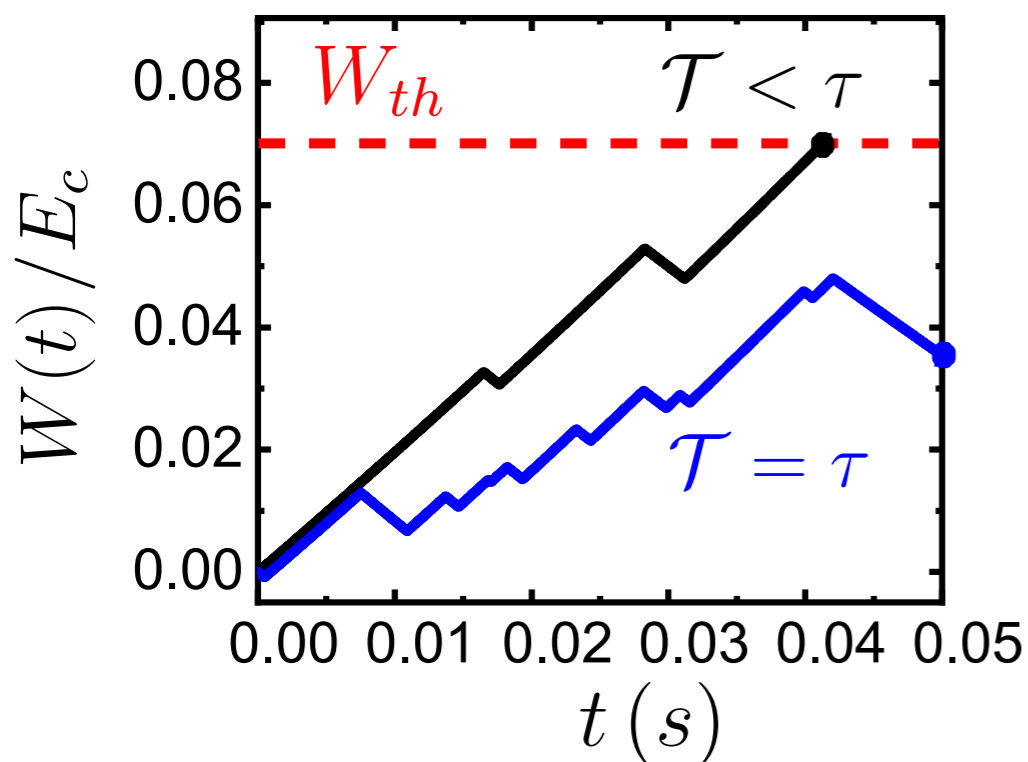
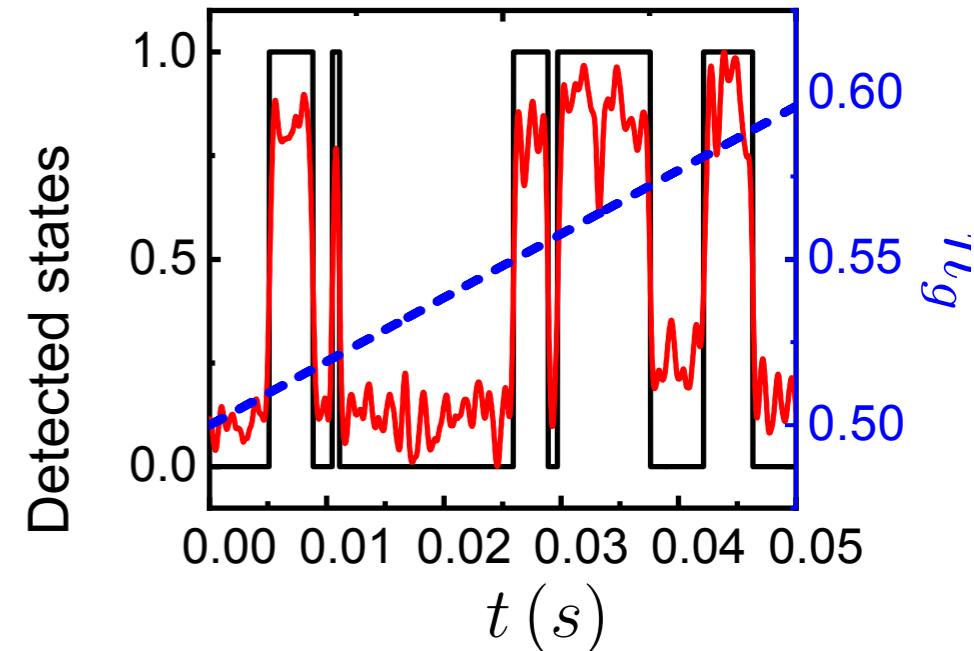
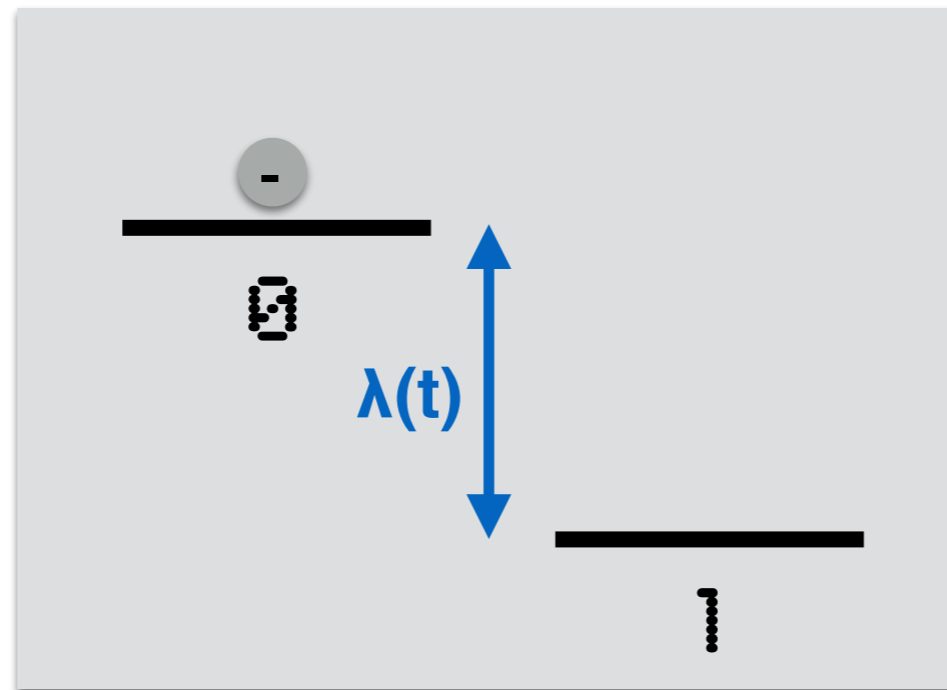
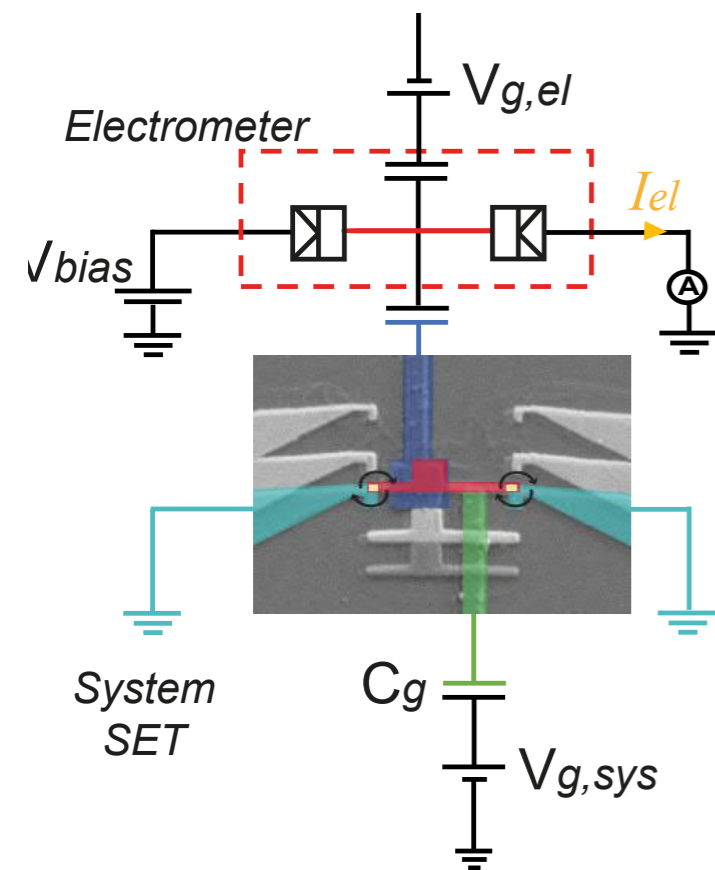


Experimental realization: Single-electron box (Pekola lab)



Gambling with a single electron

Experimental realization: Single-electron box (Pekola lab)



Gambling strategy: stopping time for the work

$$\mathcal{T} = \min(\tau_{th}, \tau)$$

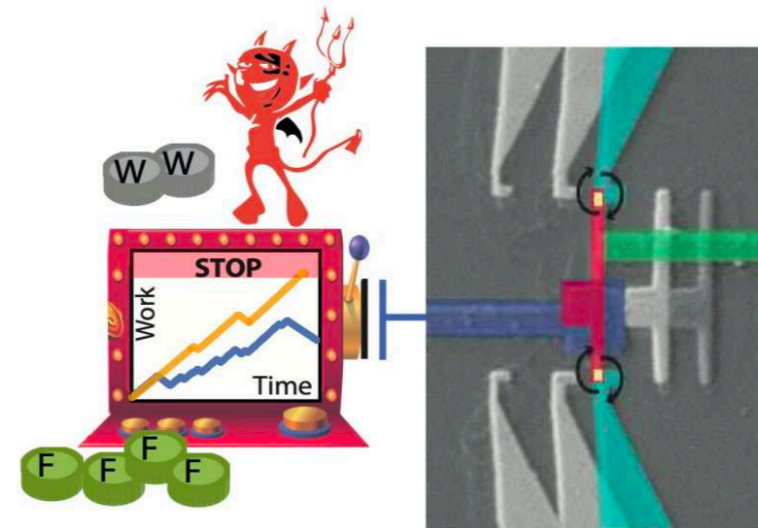
Conditionally stop the process before τ if the work exceeds a threshold value W_{th}

*No exp. feedback: trajectory post-processing

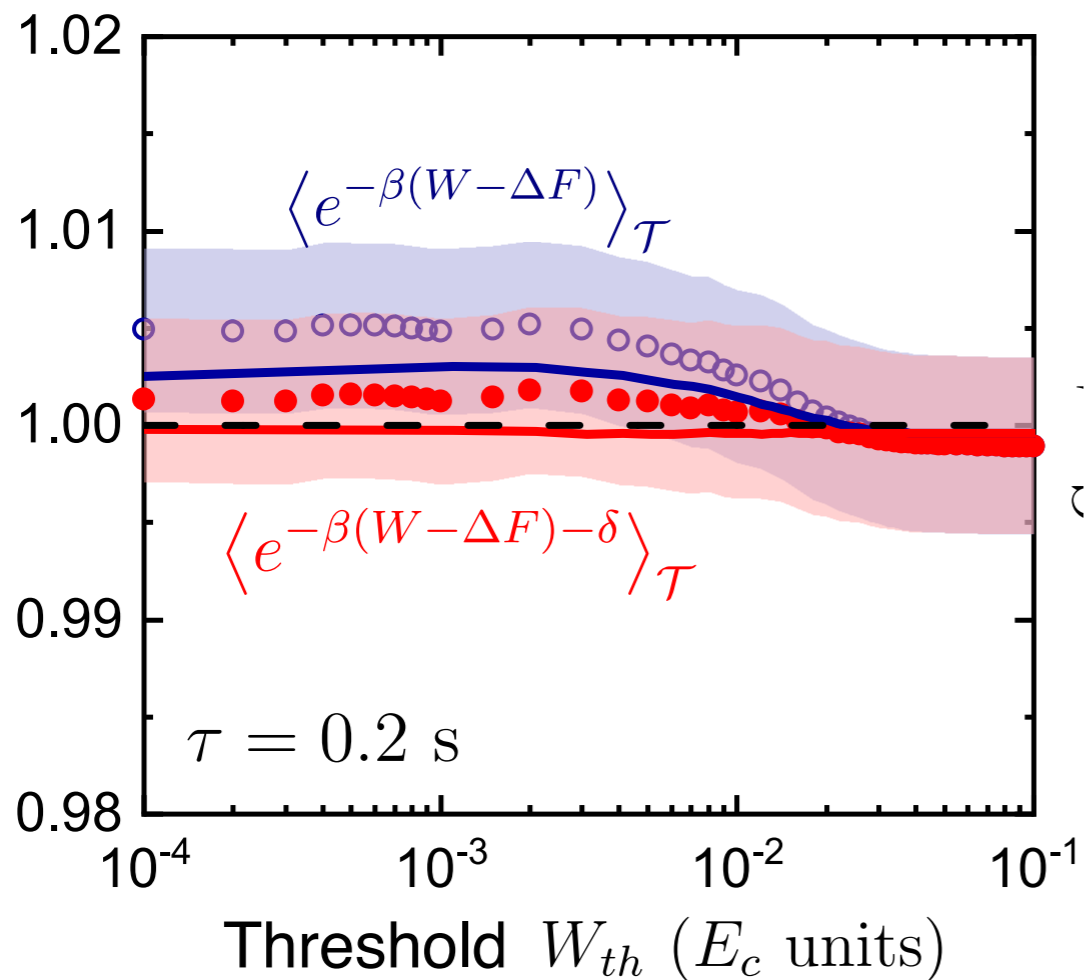
Experimental test

Theory: valid for any driving and any gambling strategy (new level of universality)

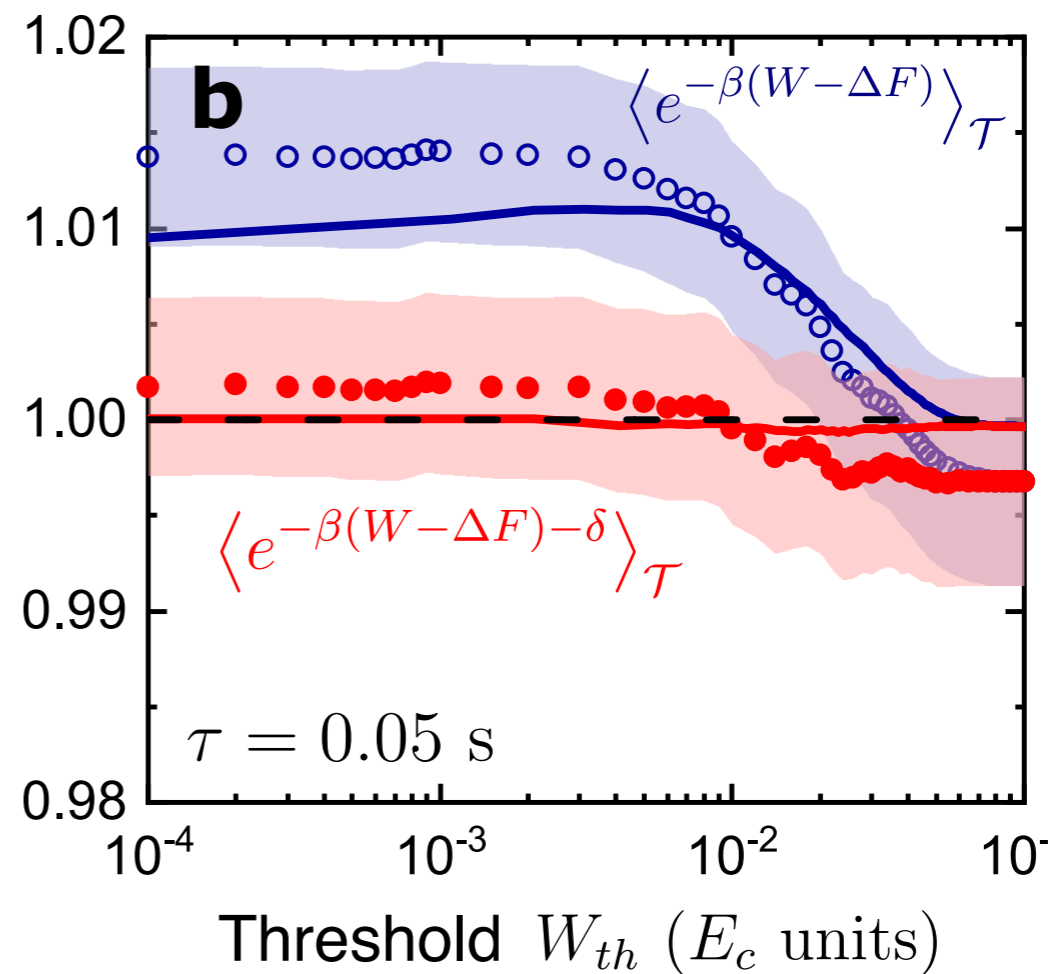
Experiment: Driven single-electron box



Slow driving



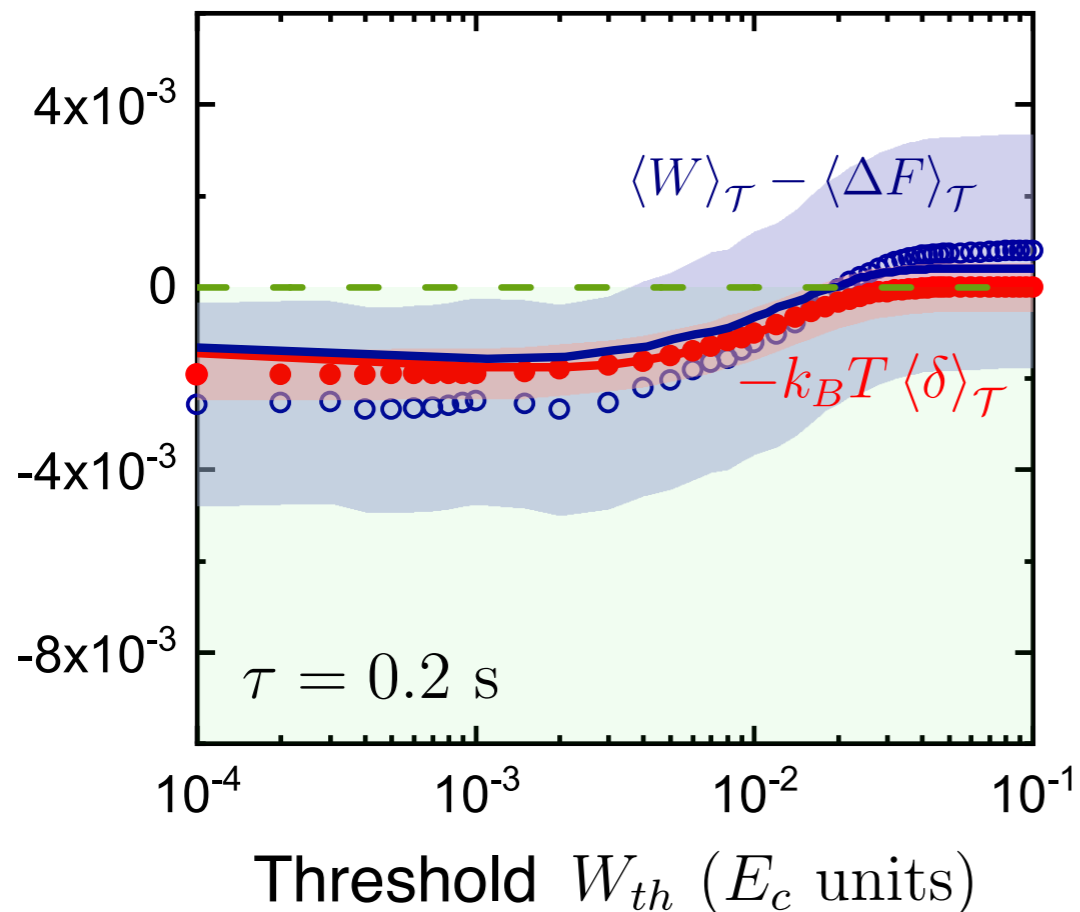
Fast driving



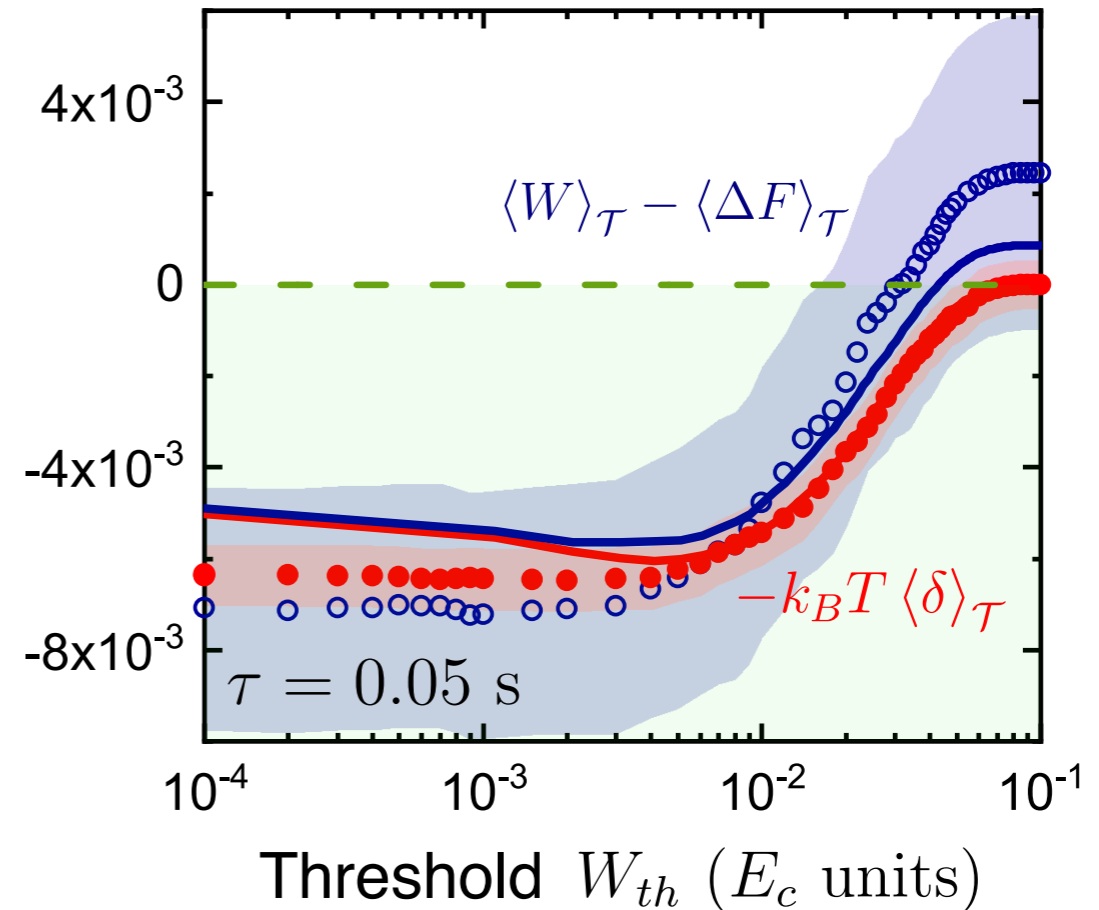
Second law at stopping times

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Slow driving



Fast driving



Ingredients for “negative dissipation”

- **gambling** (stopping times)
- **nonequilibrium** (time asymmetric protocols)

Thermodynamic Martingales and Stopping times

I. Neri, PRL **124** (4) 040601 (2020)

K. Hiura, S.-i. Sasa, arXiv 2102.06398 (2021)

C. Moslonka, K. Sekimoto, PRE **101** (6), 062139 (2020)

Y.-J. Yang, H. Qian, PRE **101** (2), 022129 (2020)

H. Ge et al., arXiv 1811.04259 (2018)

K. Cheng et al., PRE **102** (6) 062127 (2020)



FOCUS

The Gambler, Maxwell's New Demon

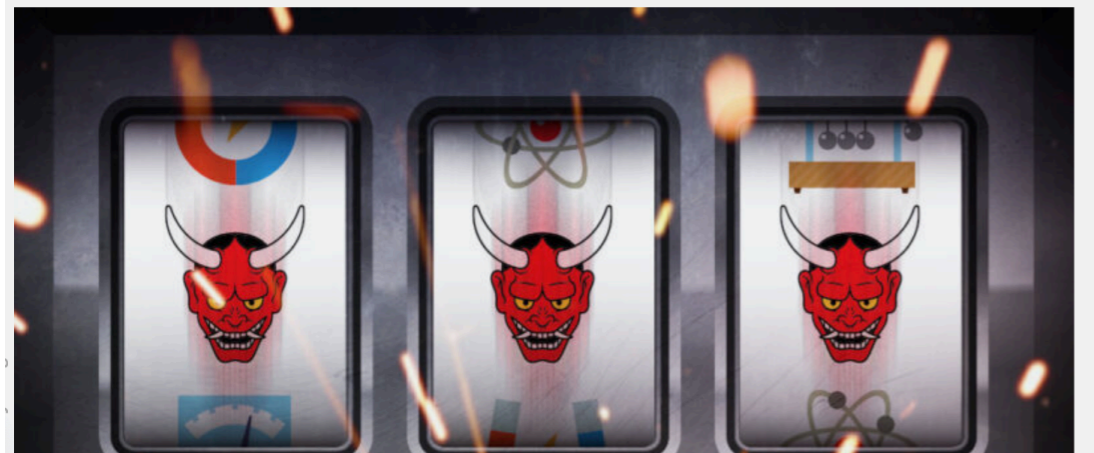
ars TECHNICA

KNOW WHEN TO WALK AWAY, KNOW WHEN TO RUN —

Meet Maxwell's gambling demon—smart enough to quit while it's ahead

Physicists even demonstrated the basic principle in a nanoscale electronic device.

JENNIFER OUELLETTE - 3/4/2021, 11:39 PM



SCIENCE

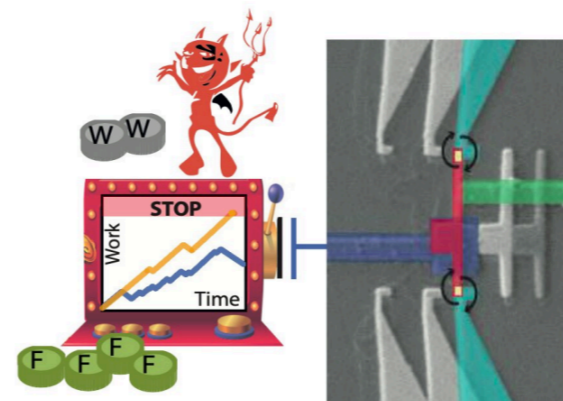
A New 'Information Demon' Thrives on Chaos

'Chaos reigns.'

PHYS.ORG

The realization of a new type of information demon that profits from gambling strategies

15 March 2021, by Ingrid Fadelli



explored the properties of a unique family of stochastic processes known as martingales in the context of thermodynamics.

Martingales are paradigmatic examples of stochastic processes that have been used in a variety of fields, including finance and mathematics. Manzano, Roldan and their colleagues applied knowledge of martingales to the study of thermodynamics with the aim of unveiling new universal thermodynamic laws.

"Our paper concerns the following questions: What happens when one gambles with the information acquired about the response of a small system

Summary and discussion

- Martingales are not scary! Useful for new developments in stochastic thermodynamics: stopping times, extrema, and beyond
- Work extraction beyond the free energy change without feedback control, by stopping a driven mesoscopic system at stochastic times

$$\langle W(\mathcal{T}) \rangle - \langle \Delta F(\mathcal{T}) \rangle \geq -k_B T \langle \delta(\mathcal{T}) \rangle$$

- No violation of the Second Law with information: ∞

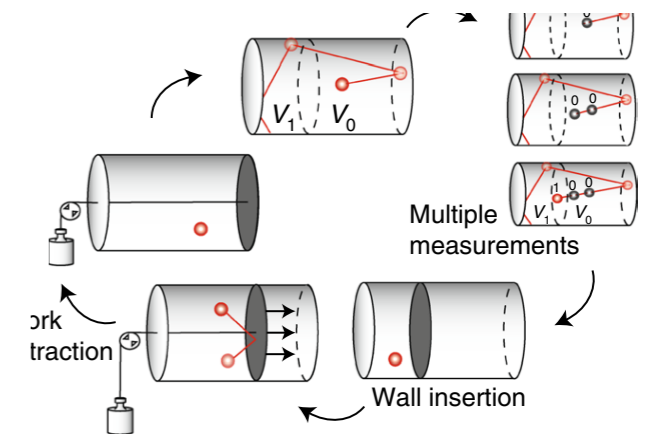
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continuous measurements

- Optimal work extraction requires:
 - patience (first-passage phenomena)
 - nonequilibrium (time-asymmetric protocols)
 - and wisdom (suitable stopping times)

- Further applications: biophysics, computation, frenesy, etc.

-



Ribezzi, Ritort, N Phys (2019)

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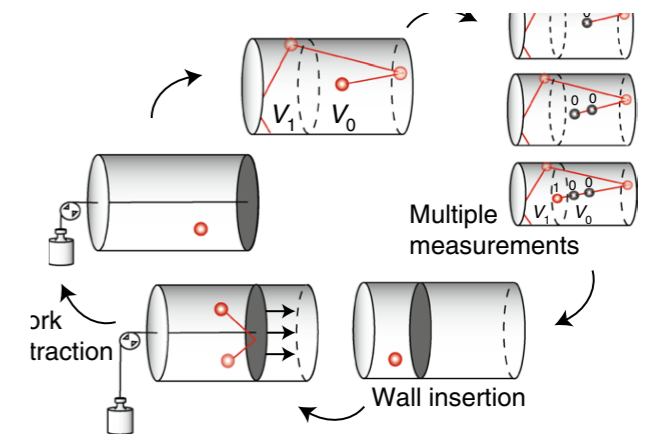
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- Further applications: biophysics, computation, frenesy, etc.

- *Physics with Martingales*, ER, I Neri, R Chetrite, S Gupta, S Pigolotti, F Julicher, K Sekimoto (coming soon 2022)



Ribezzi, Ritort, N Phys (2019)



Acknowledgments

Theory collaborators

- Gonzalo Manzano (IFISC)
- Izaak Neri (KCL)
- Raphael Chetrite (U Nice)
- Shamik Gupta (RKMVERI)
- Simone Pigolotti (OIST)
- Rosario Fazio (ICTP)
- Frank Julicher (MPIPKS)

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- Diego Subero (Aalto)
- Olivier Maillet (Aalto)
- Shilpi Singh (Aalto)
- Ivan Khaymovich (MPIPKS)
- Jukka Pekola (Aalto)

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- Gennaro Tucci
- Pierluigi Muzzeddu



Thanks for your attention !

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