

# Current fluctuations in stochastic systems with long-range memory

*Rosemary Harris*



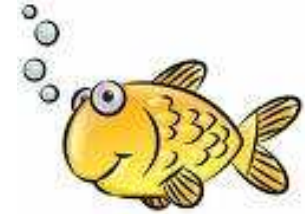
[Based on joint work with H. Touchette: *J. Phys. A: Math. Theor.* **42**, 342001 (2009)]

NEQ Workshop, Warwick, January 12th 2010

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# Introduction

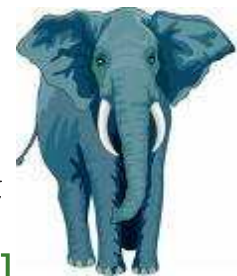
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- Interacting particles described by Markov process
  - Configurations  $\sigma(t)$
  - Transition rates  $w_{\sigma',\sigma}$
  - Non-equilibrium systems characterized by (time-integrated) currents  $Q_t$
  - Typically have large deviation principle

$$\text{Prob}(Q_t/t = j) \sim e^{-\hat{e}_w(j)t}$$

- But memoryless assumption not good for many real systems...
  - Consider class of process where rates  $w_{\sigma',\sigma}$  depend on  $\sigma, \sigma'$  and  $Q_t/t$
  - Includes analogues of “elephant random walk” [Schütz & Trimper '04]
  - Non-Markovian process but Markovian in joint current/configuration space
  - *How does memory effect the current large deviation principle?*



# Temporal additivity principle

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- Conjecture:

$$\text{Prob}(\mathcal{Q}_{t/t} = j) \sim \exp \left[ - \min_{q(\tau)} \int_{t_0}^t \hat{e}_{w(q)}(q + \tau q') d\tau \right]$$

where integral is minimized over all  $q(\tau)$  with  $q(t_0) = j_0$  and  $q(t) = j$

- General idea: Look for most probable path  $q(\tau)$  satisfying boundary conditions
- Temporal analogue of additivity principle of [Bodineau & Derrida '04]

*If Markovian rate function is known, can find large deviation principle for system with current-dependent rates by minimizing relevant integral...*

- Analytically (Euler-Lagrange, Gaussian approximation)
- Numerically

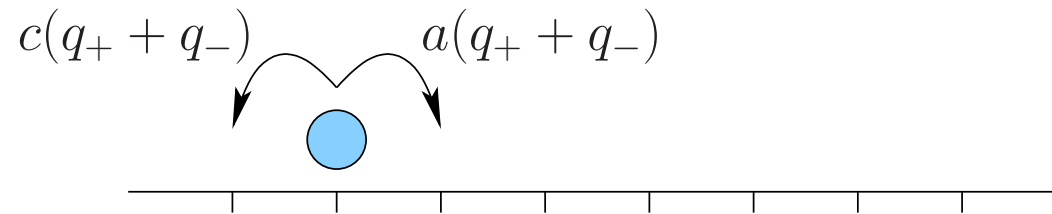
## Example: Activity-dependent random walk

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- Random walk, count separately jumps to right and left so that

$$Q_t = Q_{+,t} - Q_{-,t}$$

- Consider rates proportional to “activity”



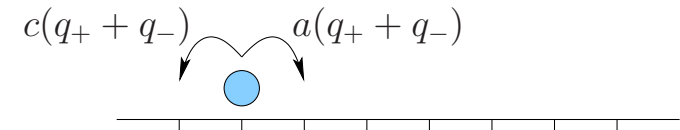
- Without loss of generality take  $a > c$ , i.e., drive to right
- For  $a + c < 1$ , find

$$\text{Prob}(Q_t/t = j) \sim \begin{cases} \exp[-jt_0^{a+c} \left(\frac{a+c}{a-c}\right) t^{1-a-c}] & \text{for } j \geq 0 \\ \exp[j(\ln \frac{a}{c})t + jt_0^{a+c} \left(\frac{a+c}{a-c}\right) t^{1-a-c}] & \text{for } j < 0 \end{cases}$$

- *Leading term in exponent is different for currents in forward and backward directions* (modified “speed” in large deviation function seems to be generic effect of memory)

# Fluctuation theorems

- For activity-dependent random walk

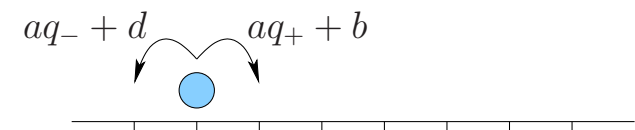


$$\frac{\text{Prob}(Q_t/t = -j)}{\text{Prob}(Q_t/t = +j)} \sim \exp \left[ -j \left( \ln \frac{a}{c} \right) t \right]$$

i.e., fluctuation theorem still holds

- Expected here since relative bias is constant  $v_R/v_L = a/c$   
(also holds for  $a + c > 1$  when there is no stationary state)

- But for “generalized elephant” random walk



$$\frac{\text{Prob}(Q_t/t = -j)}{\text{Prob}(Q_t/t = +j)} \sim \begin{cases} \exp \left[ -j \frac{2(b-d)(1-2a)}{1-a} t \right] & \text{for } 0 < a < 1/2 \\ \exp \left[ -j \frac{2(b-d)(1-2a)}{1-a} t_0^{2a-1} t^{2-2a} \right] & \text{for } 1/2 < a < 1. \end{cases}$$

- For  $1/2 < a < 1$  symmetry apparently modified by superdiffusive spreading

# Outlook

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## Summary:

- Proposed a general approach to calculate current fluctuations in systems with memory-dependent rates
- Long-range temporal correlations in non-equilibrium systems can cause modified speed, i.e., power of  $t$ , in current large deviation principle (analogous to long-range spatial correlations in equilibrium)
- Insight into applicability of fluctuation theorems for non-Markovian systems

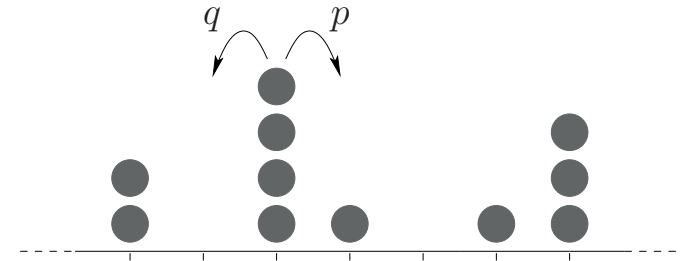
## Current work:

- Many-particle systems
  - Dynamical phase transitions, possibility of non-convex rate function
- Intrinsically non-Markovian systems where rates depend on complete current history
  - cf. “Alzheimer random walk” [Cressoni *et al.* '07, Kenkre '07]

# Stochastic Markovian dynamics

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- Interacting particles described by Markov process
- Configurations (microstates)  $\sigma(t)$



- Stochastic approaches:

– Langevin: Differential equation for  $\sigma(t)$ , deterministic + noisy forces

– **Master equation:**

\* Transition rates  $w_{\sigma',\sigma}$

\* Deterministic evolution for probability distribution  $P(\sigma, t)$ :

$$\frac{d}{dt}P(\sigma, t) = \sum_{\sigma' \neq \sigma} [w_{\sigma, \sigma'} P(\sigma', t) - w_{\sigma', \sigma} P(\sigma, t)]$$

\* Or in “quantum Hamiltonian formalism”:

$$\frac{d}{dt}|P(t)\rangle = -H|P(t)\rangle$$

# Ergodicity, stationarity

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- Concentrate, for now, on time-independent rates
- Conservation of probability

$$\sum_{\sigma} P(\sigma, t) = 1 \quad \langle s | H = 0$$

- Ergodic system has unique stationary distribution

$$\frac{d}{dt} P^*(\sigma, t) = 0 \quad H | P^* \rangle = 0$$

- *Equilibrium*, detailed balance

$$w_{\sigma, \sigma'} P^*(\sigma') = w_{\sigma', \sigma} P^*(\sigma) \quad P^* H^T (P^*)^{-1} = H$$

- *Non-equilibrium*

- Broken detailed balance
- $H$  has complex spectrum
- Stationary state characterized by non-zero currents



# Currents

- Counter  $Q_t$ , value increases by  $\Theta_{\sigma',\sigma}$  at each transition  $\sigma \rightarrow \sigma'$
- $\Theta$  is real and *antisymmetric* matrix
- $Q_t$  is a functional of history  $\{\sigma(\tau), 0 \leq \tau \leq t\}$ .

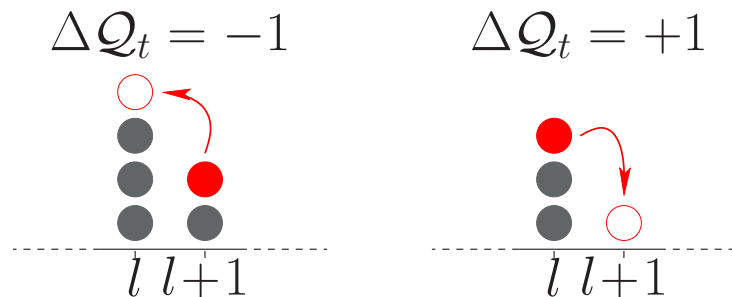
$$Q_t = \sum_{n=1}^{N-1} \Theta_{\sigma_{n+1}, \sigma_n}$$

- Generating function given by

$$\langle e^{-\lambda Q_t} \rangle = \langle s \left| e^{-\tilde{H}(\lambda)t} \right| P_0 \rangle$$

$\tilde{H}$  is “modified Hamiltonian” with off-diagonal elements  $w_{\sigma,\sigma'}$  replaced by  $w_{\sigma,\sigma'} e^{-\lambda \Theta_{\sigma,\sigma'}}$

- Example: Integrated particle current across bond



# Large deviations

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- Long-time distribution of  $\mathcal{Q}_t$  often characterized by

$$e_w(\lambda) := - \lim_{t \rightarrow \infty} \frac{1}{t} \ln \langle e^{-\lambda \mathcal{Q}_t} \rangle$$

- Now consider time-averaged current  $\mathcal{Q}_t/t$
- Distribution  $p(j, t) = \text{Prob}(\mathcal{Q}_t/t = j)$  has large deviation property

$$\hat{e}_w(j) := \lim_{t \rightarrow \infty} -\frac{1}{t} \ln p(j, t), \quad p(j, t) \sim e^{-\hat{e}_w(j)t}$$

- $e_w(\lambda)$  and  $\hat{e}_w(j)$  are related by Legendre transform<sup>1</sup>

$$\hat{e}_w(j) = \sup_{\lambda} \{e_w(\lambda) - \lambda j\}, \quad e_w(\lambda) = \inf_j \{\hat{e}_w(j) + \lambda j\}$$

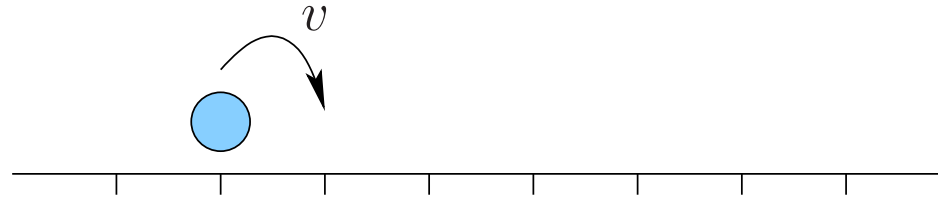
- Rate function analogous to entropy of an equilibrium system.

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<sup>1</sup>Strictly true only when  $e_w(\lambda)$  is differentiable

# Single particle on an infinite lattice

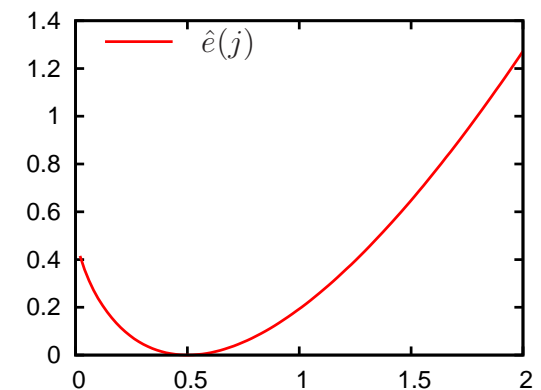
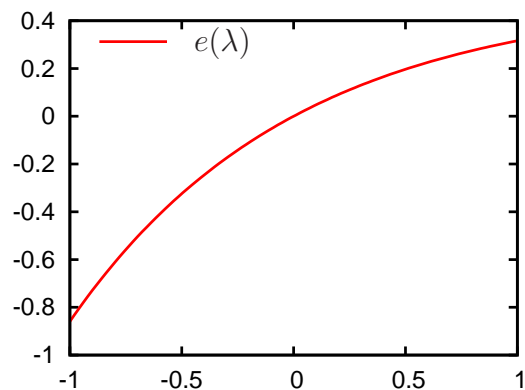
- Single particle hopping rightwards on an infinite lattice



- Let  $Q_t$  count the number of jumps up to time  $t$
- Large deviation function given by

$$e_v(\lambda) = v(1 - e^{-\lambda}) \quad \Longleftrightarrow \quad \hat{e}_v(j) = v - j + j \ln \frac{j}{v}$$

- For example,  $v = 0.5$ :



# Driven many-particle systems: generic features

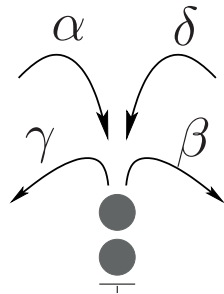
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- For non-equilibrium systems, current distribution typically non-Gaussian
- Under general conditions, a fluctuation symmetry holds  
[Gallavotti & Cohen '95, Lebowitz & Spohn '99]

$$\frac{\text{Prob}(\mathcal{Q}_t/t = -j)}{\text{Prob}(\mathcal{Q}_t/t = +j)} \sim e^{-Ejt}$$

*But can have breakdown in systems with unbounded state space*

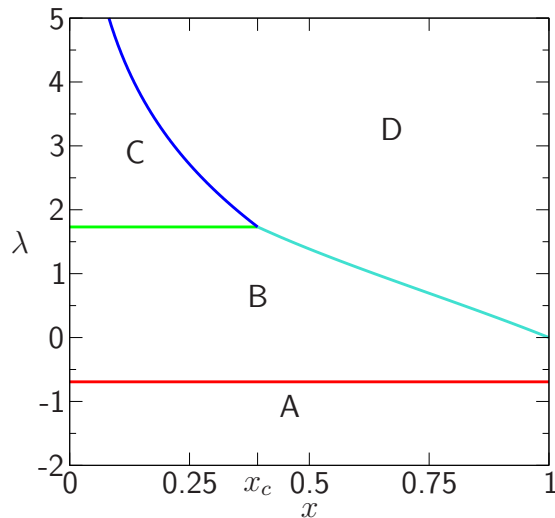
- Current large deviations can show surprisingly complicated phase structure even in simple models
- Example: Single-site ZRP with open boundaries [RJH, Rákos & Schütz '06]



Initial condition:

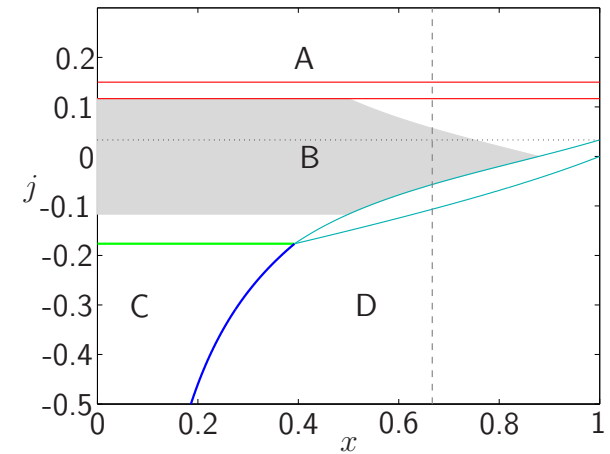
$$|P_0\rangle = (1-x) \sum_{n=0}^{\infty} x^n |n\rangle$$

# Dynamical phase transitions in 1-site ZRP



$$\hat{e}(j) = \sup_{\lambda} \{e(\lambda) - \lambda j\}$$

$$e(\lambda) = \inf_j \{\hat{e}(j) + \lambda j\}$$

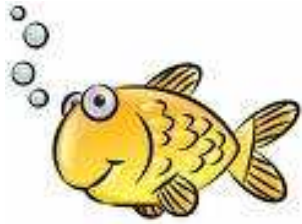


- Analogy to equilibrium:

- Phase transitions (both first order and continuous)
- Time  $t$  plays role of system size

# Current-dependent rates

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- Many ways to introduce memory
- We consider class of process where rates  $w_{\sigma',\sigma}$  depend explicitly on  $\sigma$ ,  $\sigma'$  and  $Q_t/t$   
(To avoid singularities, assume observations start at  $t_0$ , where  $0 \ll t_0 \ll t$ )
- Includes analogues of “elephant random walk” [Schütz and Trimper '04]
- Non-Markovian process but Markovian in joint current/configuration space
- *How does memory effect the current large deviation principle?*  
(i.e., do we still have form  $\text{Prob}(Q_t/t = j) \sim e^{-\hat{e}_w(j)t}$  ?)

# Temporal additivity principle

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- Conjecture:

$$\text{Prob}(\mathcal{Q}_t/t = j) \sim \exp \left[ - \min_{q(\tau)} \int_{t_0}^t \hat{e}_{w(q)}(q + \tau q') d\tau \right]$$

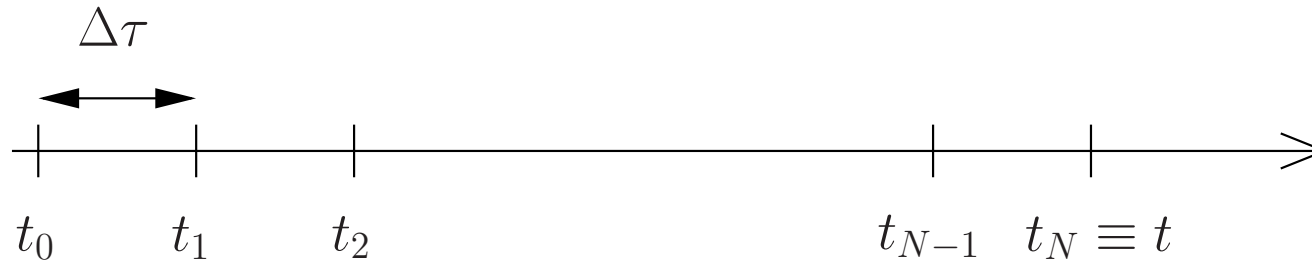
where integral is minimized over all  $q(\tau)$  with  $q(t_0) = j_0$  and  $q(t) = j$

- General idea: Look for most probable path  $q(\tau)$  satisfying boundary conditions
- Temporal analogue of additivity principle of [Bodineau and Derrida '04]

## Sketch of argument for temporal additivity principle

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1. Divide interval  $[t_0, t]$  into  $N$  subintervals of length  $\Delta\tau$ .



2. Chapman-Kolmogorov equation for joint probabilities of being found in configuration  $\sigma_i$  with average current  $q_i$ :

$$\begin{aligned} & p(q_N, \sigma_N, t | q_0, \sigma_0, t_0) \\ &= \sum_{\substack{q_1, \dots, q_{N-1} \\ \sigma_1, \dots, \sigma_{N-1}}} p(q_N, \sigma_N, t | q_{N-1}, \sigma_{N-1}, t_{N-1}) \cdots p(q_2, \sigma_2, t_2 | q_1, \sigma_1, t_1) p(q_1, \sigma_1, t_1 | q_0, \sigma_0, t_0) \end{aligned}$$

3. If  $\Delta\tau \gg 0$ , then assume  $p(q_{n+1}, \sigma_{n+1}, t_{n+1} | q_n, \sigma_n, t_n)$  independent of  $\sigma_n$  (true for an ergodic system with finite state space)

$$p(q_N, t | q_0, t_0) = \sum_{q_1, \dots, q_{N-1}} p(q_N, t | q_{N-1}, t_{N-1}) \cdots p(q_2, t_2 | q_1, t_1) p(q_1, t_1 | q_0, t_0)$$



## Sketch of argument for temporal additivity principle

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4. Now take  $t$  and  $N$  large whilst preserving their ratio (so  $t \gg \Delta\tau \gg 0$ );  
 $q(\tau)$  almost constant in each timeslice (adiabatic approx.)

5. Observed average current in timeslice  $(t_n, t_{n+1}]$  is

$$q_{\Delta\tau}^{(n)} = \frac{q_{n+1}t_{n+1} - q_n t_n}{\Delta\tau}$$

6. So using *Markovian* large deviation principle have

$$p(q_{n+1}, t_{n+1} | q_n, t_n) \approx A_n e^{-\Delta\tau \hat{e}_w(q_n)(q_{\Delta\tau}^{(n)})}$$

7. Putting all the slices together gives

$$p(q_N, t | q_0, t_0) \approx A \sum_{q_1, \dots, q_{N-1}} e^{-\sum_{n=0}^{N-1} \Delta\tau I_w(q_n)(q_{\Delta\tau}^{(n)})}.$$

8. Then pass to continuum limit  $N, t, \Delta\tau \rightarrow \infty$ ,  $q_n \rightarrow q(\tau)$

$$p(j, t | j_0, t_0) \sim \int_{q(t_0)=j_0}^{q(t)=j} \mathcal{D}[q] e^{-\int_{t_0}^t \hat{e}_w(q)(q+\tau q') d\tau}$$

## Sketch of argument for temporal additivity principle

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9. In  $t \rightarrow \infty$  limit, path integral dominated by most probable path in  $q$ -space, so

$$\text{Prob}(\mathcal{Q}_t/t = j) \sim \exp \left[ - \min_{q(\tau)} \int_{t_0}^t \hat{e}_{w(q)}(q + \tau q') d\tau \right]$$

where integral is minimized over all  $q(\tau)$  with  $q(t_0) = j_0$  and  $q(t) = j$

10. To make  $t$ -dependence more explicit write

$$\text{Prob}(\mathcal{Q}_t/t = q) \sim e^{-t^\alpha F(j)},$$

If  $F(j)$  exists and is not everywhere zero then have large deviation principle.

$$F(j) = \lim_{t \rightarrow \infty} \min_{q(\tau)} \frac{1}{t^\alpha} \int_{t_0}^t \hat{e}_{w(q)}(q + \tau q') d\tau.$$

*If Markovian rate function is known, can find large deviation principle for system with current-dependent rates by minimizing relevant integral...*

- Analytically (Euler-Lagrange, Gaussian approximation)
- Numerically

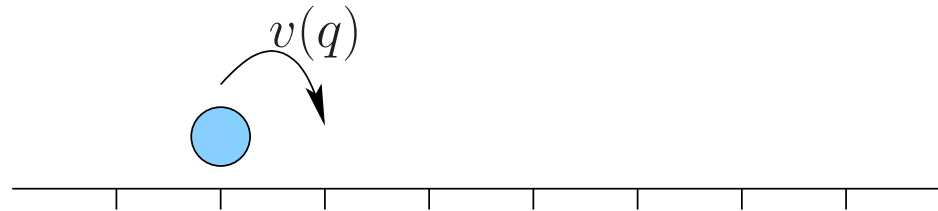
## Example 1: Uni-directional random walk

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- Recall Markovian case of single particle hopping rightwards on an infinite lattice

$$\hat{e}_v(j) = v - j + j \ln \frac{j}{v}$$

- Now modify picture so that rate for hopping at time  $t$  depends on average current  $q(t)$  up to  $t$



- Predict that distribution of number of jumps  $Q_t$  has asymptotic form

$$\text{Prob}(Q_t/t = j) \sim \exp \left[ - \min_{q(\tau)} \int_{t_0}^t \hat{e}_{v(q)}(q + \tau q') d\tau \right]$$

- Minimizing integral gives Euler-Lagrange equation

$$\frac{dv}{dq} - q \frac{dv/dq}{v_R} - \frac{2\tau q'}{q + \tau q'} - \frac{\tau^2 q''}{q + \tau q'} = 0$$

## Example 1: Uni-directional random walk

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- Exactly solvable cases include  $v(q) = aq$ , i.e., rate for particle to move at given time is directly proportional to average velocity up to that time
- In this case, solving E-L equation and carrying out integration gives

$$\min_{q(\tau)} \int_{t_0}^t \hat{e}_{v(q)}(q + \tau q') d\tau \sim \begin{cases} jt_0^a t^{1-a} & \text{for } a < 1 \\ (a-1)j_0 t_0 \ln t & \text{for } a > 1 \end{cases}$$

- Crossover at  $a = 1$ :
  - $a > 1$ , escape regime: no large deviation principle
  - $a < 1$ , localized regime:
    - \* System approaches state where particle has zero velocity
    - \* Large deviation principle with “speed”  $t^{1-a}$ :

$$\text{Prob}(Q_t/t = j) \sim e^{-jt_0^a t^{1-a}}, \quad \text{for } j > 0$$

- \* Can show

$$\text{Var}[Q_t] \sim (t/t_0)^{2a}$$

so transition from subdiffusive regime to superdiffusive regime at  $a = 1/2$

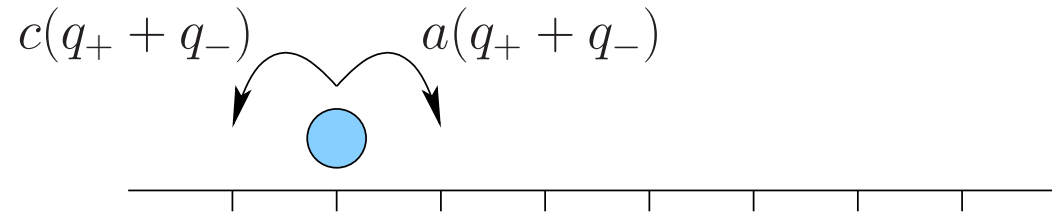
## Example 2: Bi-directional random walk with activity dependent rates

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- Bi-directional random walk, count separately jumps to right and left so that

$$Q_t = Q_{+,t} - Q_{-,t}$$

- Consider rates proportional to “activity”



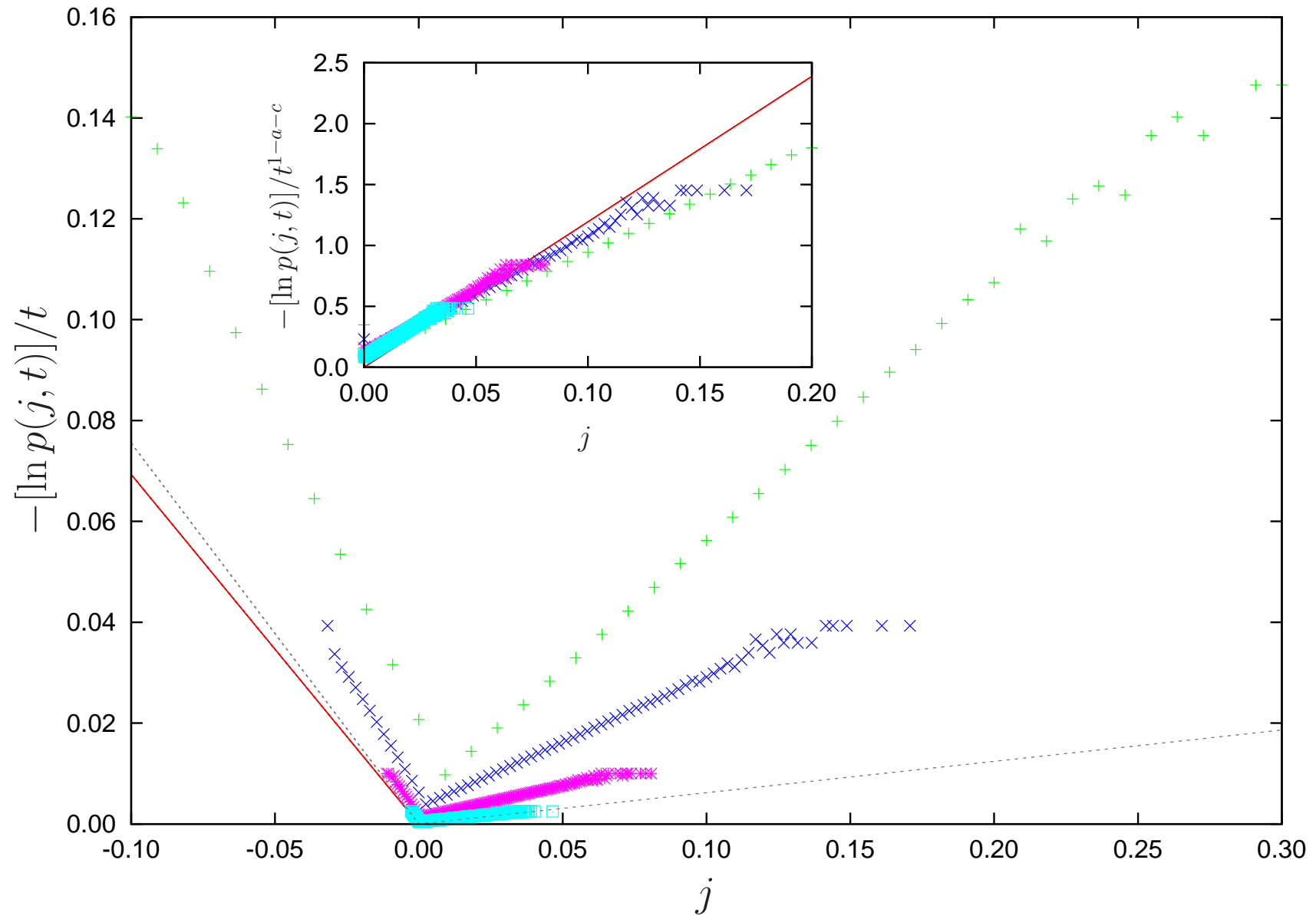
- Without loss of generality take  $a > c$ , i.e., drive to right
- For  $a + c < 1$ , we find

$$\text{Prob}(Q_t/t = j) \sim \begin{cases} \exp[-jt_0^{a+c} (\frac{a+c}{a-c}) t^{1-a-c}] & \text{for } j \geq 0 \\ \exp[j(\ln \frac{a}{c})t + jt_0^{a+c} (\frac{a+c}{a-c}) t^{1-a-c}] & \text{for } j < 0. \end{cases}$$

- *Leading term in exponent is different for currents in forward and backward directions*

## Example 2: Bi-directional random walk with activity dependent rates

Comparison with simulation:



## Example 2: Bi-directional random walk with activity dependent rates

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- What about fluctuation symmetry?
- Since

$$\text{Prob}(Q_t/t = j) \sim \begin{cases} \exp[-jt_0^{a+c} \left(\frac{a+c}{a-c}\right) t^{1-a-c}] & \text{for } j \geq 0 \\ \exp[j(\ln \frac{a}{c})t + jt_0^{a+c} \left(\frac{a+c}{a-c}\right) t^{1-a-c}] & \text{for } j < 0. \end{cases}$$

then

$$\frac{\text{Prob}(Q_t/t = -j)}{\text{Prob}(Q_t/t = +j)} \sim \exp \left[ -j \left( \ln \frac{a}{c} \right) t \right]$$

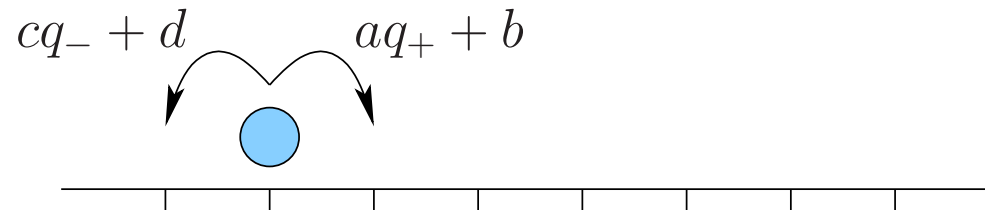
i.e., fluctuation theorem still holds

- Expected here since relative bias is constant  $v_R/v_L = a/c$   
(also holds for  $a + c > 1$  when there is obviously no stationary state)

## Example 3: Generalized elephant

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- Again consider bi-directional random walk but with rates



- For  $a, c < 1$  have mean currents

$$\bar{q}_+ = \frac{b}{1-a}, \quad \bar{q}_- = \frac{d}{1-c} \quad \text{and} \quad \bar{q} = \bar{q}_+ - \bar{q}_-$$

- Gaussian expansion (about means) and minimization of integral gives, for  $a = c$ :

–  $0 < a < 1/2$ , diffusive behaviour:

$$\text{Prob}(\mathcal{Q}_t/t = j) \sim \exp \left\{ - \left[ \frac{1}{2} \frac{\left( j - \frac{b-d}{1-a} \right)^2}{\frac{b+d}{(1-a)(1-2a)}} \right] t \right\}$$

–  $1/2 < a < 1$ , superdiffusive behaviour:

$$\text{Prob}(\mathcal{Q}_t/t = j) \sim \exp \left\{ - \left[ \frac{1}{2} \frac{\left( j - \frac{b-d}{1-a} \right)^2}{\frac{b+d}{(1-a)(2a-1)}} \right] t_0^{2a-1} t^{2-2a} \right\}$$

(generalization of results for original symmetric discrete-time elephant)



## Example 3: Generalized elephant

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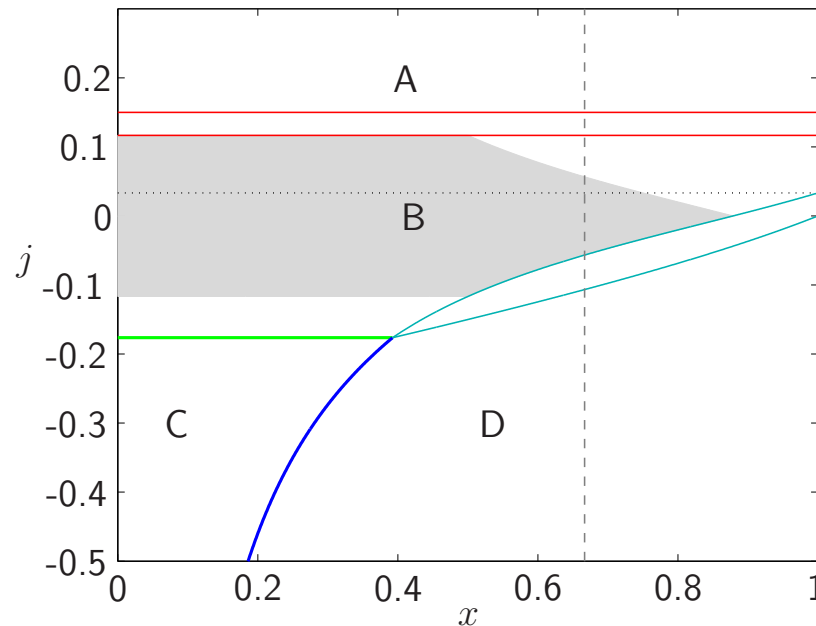
- Within this Gaussian approximation

$$\frac{\text{Prob}(Q_t/t = -j)}{\text{Prob}(Q_t/t = +j)} \sim \begin{cases} \exp \left[ -j \frac{2(b-d)(1-2a)}{1-a} t \right] & \text{for } 0 < a < 1/2 \\ \exp \left[ -j \frac{2(b-d)(1-2a)}{1-a} t_0^{2a-1} t^{2-2a} \right] & \text{for } 1/2 < a < 1. \end{cases}$$

- Both cases have well-defined mean stationary current...
- ...but only have usual fluctuation symmetry for  $0 < a < 1/2$
- For  $1/2 < a < 1$  symmetry is apparently modified by superdiffusive spreading about the mean
  - Logarithm of probabilities for forward and backward currents still asymptotically proportional to  $j$  but sublinear in  $t$
- Scenario merits closer investigation

# Many-particle systems

- In general would need to minimize integral numerically to find large deviations for memory-dependent case
- For example, Markovian 1-site open-boundary ZRP [RJH, Rákos & Schütz '06]



A:  $\hat{e}_w(j) = f_j(\alpha, \gamma)$

B:  $\hat{e}_w(j) = f_j\left(\frac{\alpha\beta}{\beta+\gamma}, \frac{\gamma\delta}{\beta+\gamma}\right)$

C:  $\hat{e}_w(j) = f_j(\alpha, \gamma) + f_j(\beta, \delta)$

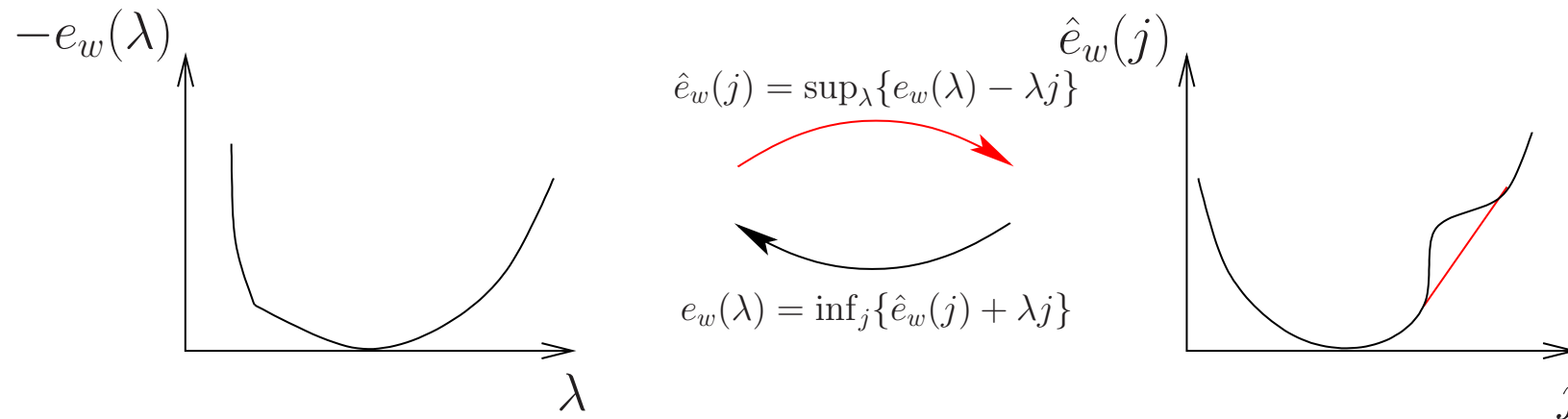
D:  $\hat{e}_w(j) = f_j(\alpha, \gamma) + \beta(1-x) + \delta(1-x^{-1}) + j \ln x$

with  $f_j(a, b) = a + b - \sqrt{j^2 + 4ab} + j \ln \frac{j + \sqrt{j^2 + 4ab}}{2a}$

- Particularly interested in effect of memory on dynamical phase transitions...

## Non-convex rate functions

- For  $e_w(\lambda)$  non-differentiable, Legendre transform *only* yields convex envelope of  $\hat{e}_w(j)$



- For short-range temporal correlations then system can phase separate in time...
  - Gives straight-line section of rate function
- ...But not necessarily so for systems with memory/long-range temporal correlations
  - Non-convex rate functions are possible
- Analogy: long-range spatial correlations in equilibrium give non-concave entropies
- Can we demonstrate this explicitly in ZRP with appropriate current-dependent rates?

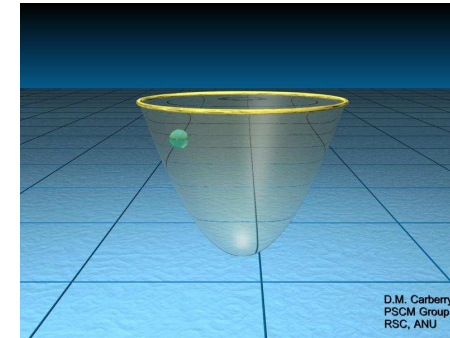
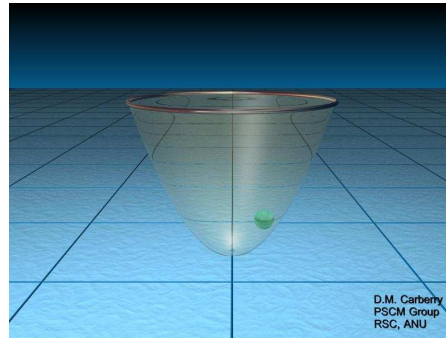
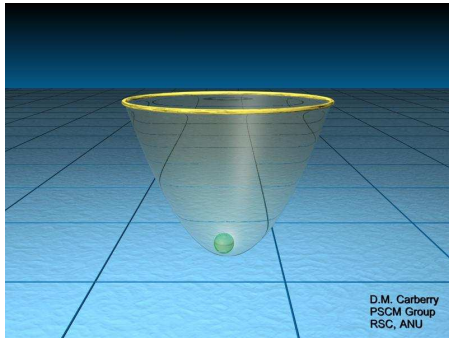
## Harder problem

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- Suppose rates at time  $t$  depend not on  $q(t)$  but on full history, i.e.,  $q(\tau)$  for  $0 \leq \tau \leq t$ .
- Now have an intrinsically non-Markovian problem
- For example, take rates at time  $t$  which depend on  $q(t/2)$ 
  - cf. “Alzheimer random walk” [Cressoni *et al.* '07, Kenkre '07]
- In principle, can still use additivity-type approach but have to minimize non-local integral...

# Experiment: colloidal particle in optical trap

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“Experimental Demonstration of Violations of the Second Law of Thermodynamics for Small Systems and Short Time Scales”

G. Wang *et al.* Phys. Rev. Lett. **89** 050601 (2002)

# Non-equilibrium fluctuation theorems

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*“Relate the probability of observing a given entropy increase to the probability of observing the same magnitude of entropy decrease”*

$$\frac{p(-\mathcal{X}, t)}{p(\mathcal{X}, t)} \sim e^{-\chi t}$$

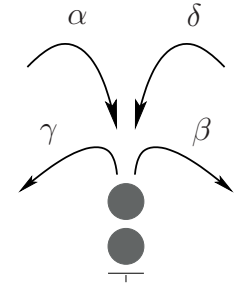
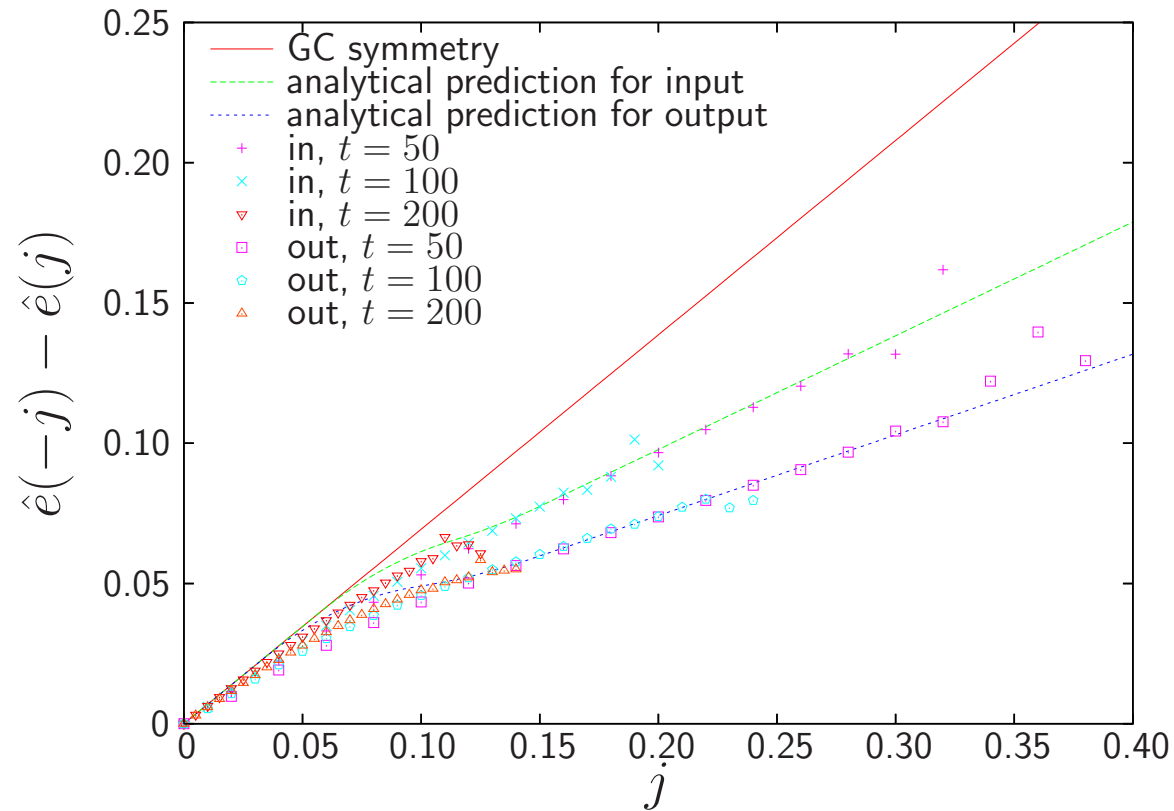
1. Computer simulations of sheared fluids [D.Evans *et al.* '93 ]
2. Steady state of deterministic systems [Gallavotti & Cohen '95]:
  - $\mathcal{X}$  is rate of phase space contraction
3. Stochastic systems (with bounded state space) [Lebowitz & Spohn '99]
  - $\mathcal{X}$  can often be identified with average particle current
  - Symmetry  $\tilde{H}(\lambda)^T = P_{\text{eq}}^{-1} \tilde{H}(E - \lambda) P_{\text{eq}} \Rightarrow e(\lambda) = e(E - \lambda)$
  - *But the zero-range process has unbounded state space!*

# Back to single-site ZRP

- Prediction:

$$\frac{p(-j, t)}{p(j, t)} \sim e^{-Ejt} \quad \text{with } E \text{ an effective field}$$

- e.g., 1 site ZRP with steady-state initial condition:



- *Breakdown of Gallavotti-Cohen symmetry*
  - Physically due to “instantaneous condensates”

# Fluctuation Theorems: General perspective

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- Consider “dissipation function”  $W(t)$  [for  $w_n = 1$ ]

$$W(t) = \sum_{l=0}^L E_l J_l(t) - \ln \frac{P_0(\sigma(t))}{P_0(\sigma(0))}$$

- Distribution of  $w(t) = W(t)/t$  obeys

$$\frac{p(-w, t)}{p(w, t)} = e^{-wt}$$

→ transient fluctuation theorem [D.Evans & Searles '94]

- *For bounded state space*, in the long-time limit one can replace  $W(t)$  by  $(\sum_{i=0}^L E_i) J_t$
- **For unbounded state space, boundary terms are non-vanishing and GC symmetry can be violated**
- Analogous effects due to unbounded potentials:
  - Deterministic forces, single-particle Langevin dynamics  
[Bonetto *et al.* '05, van Zon & Cohen '03, Farago '02, Baiesi *et al.* '06]

[Experimentally relevant, e.g., trapped colloids, granular media, electric circuits, ... ]