

# *Fluctuation relations and some applications*

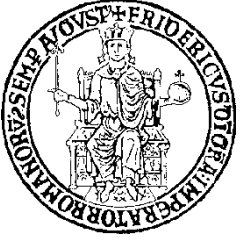
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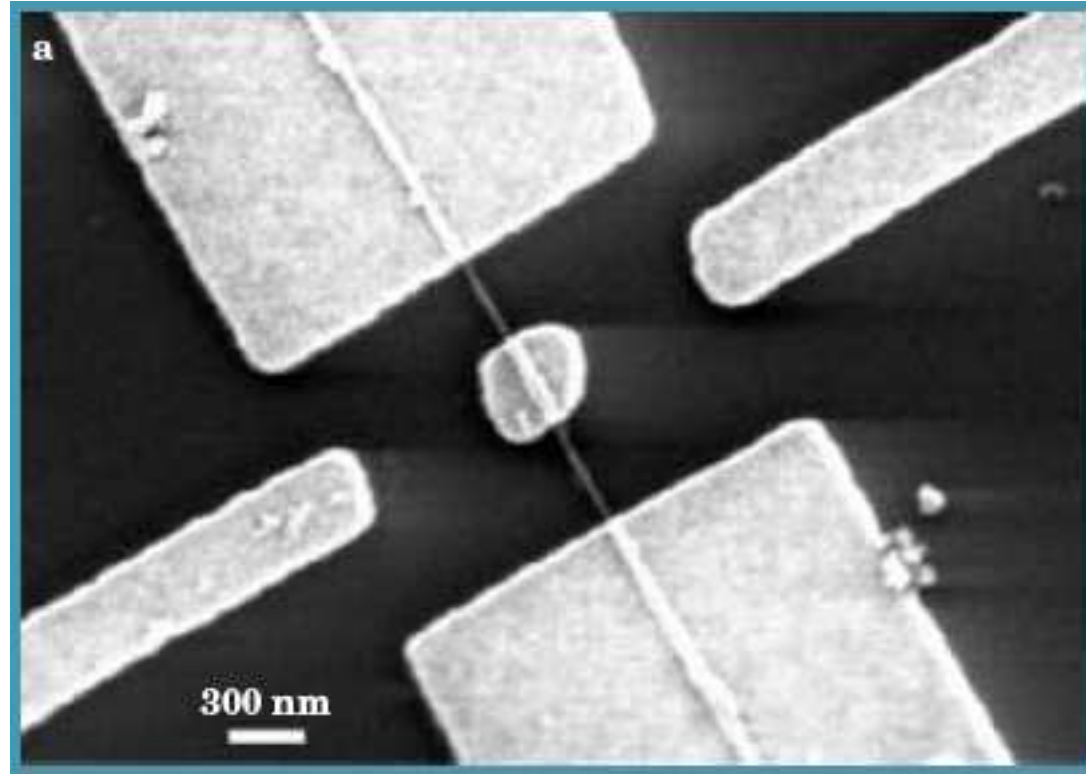
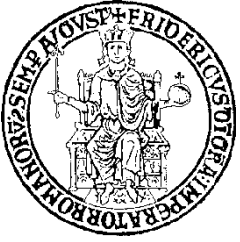
Associato INFN, Sezione di Napoli



# Summary

- ⑥ Small systems
- ⑥ Generalized second law: The Hasano-Sasa relation
- ⑥ Microscopic reversibility and Loschmidt's paradox
- ⑥ The “time-reversal” identities and the fluctuation theorems
- ⑥ Exploiting the Jarzynski relation: The histogram method
- ⑥ Exploring driven systems: The cloning technique
- ⑥ Perspectives

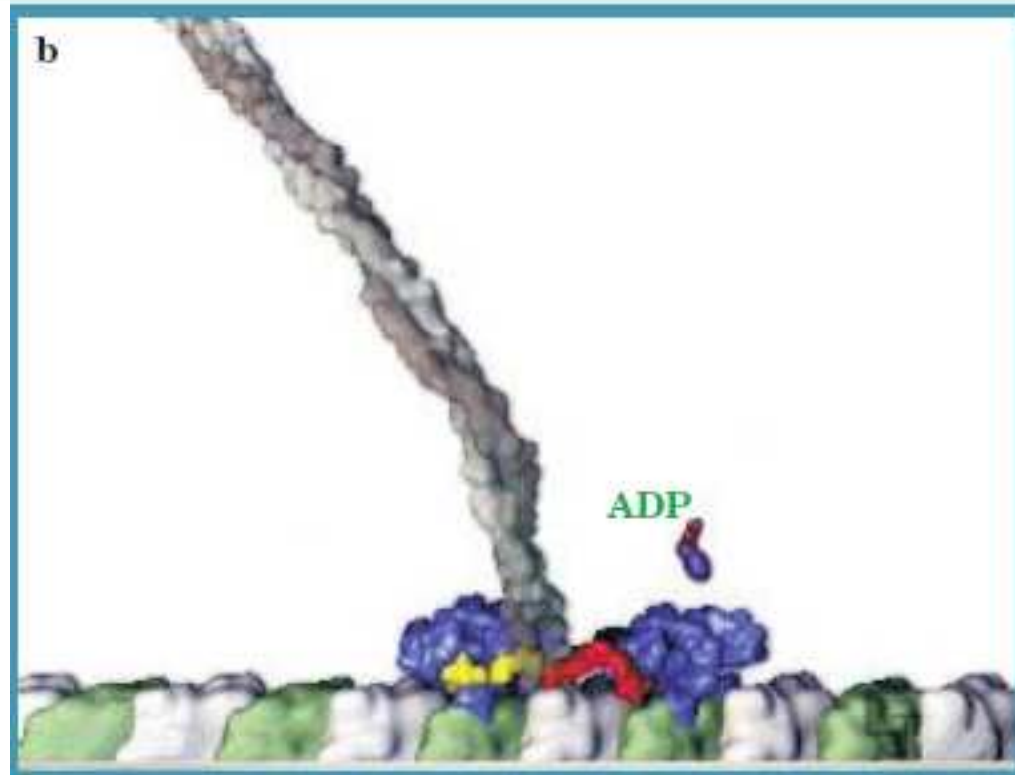
# Small systems: Artificial nanomachines



A metal-plate rotor attached to a multiwalled nanotube

*Fennimore et al., 2003*

# Small systems: Natural nanomachines



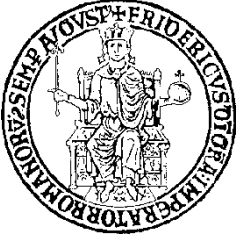
The kinesin-microtubule system: one 8-nm step every 10–15 ms

$ATP \longrightarrow ADP + P + \sim 20 T$  ( $k_B = 1$ )

Typically  $\Delta W \sim 12 T$ , efficiency  $\sim 60\%$

Dissipated power:  $\sim 20 T$  per second

*Milligan Laboratory, Scripps Research Institute*



# Evolution equation

$P(x, t)$ : pdf of microstate  $x$

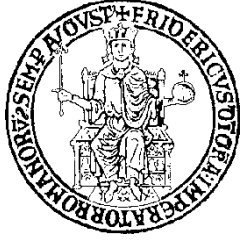
$$\frac{\partial P}{\partial t} = \hat{L}_{\mu(t)} P$$

$\hat{L}_{\mu}$ : Liouville operator depending on parameter  $\mu$

Manipulation:  $t \longrightarrow \mu(t)$ ,  $0 \leq t \leq t_f$ ,  $\mu(0) = 0$

Steady state: for each  $\mu$ ,

$$\hat{L}_{\mu} P_{\mu}^{\text{SS}} = 0, \quad \forall \mu$$



# The Hatano-Sasa relation

$\phi(x, \mu)$ : “Steady state hamiltonian”

$$\phi(x, \mu) = -\ln P_{\mu}^{\text{SS}}(x)$$

Define

$$A(t) = \int_0^t dt' \dot{\mu}(t') \left. \frac{\partial \phi}{\partial \mu} \right|_{\mu(t'), x(t')}$$

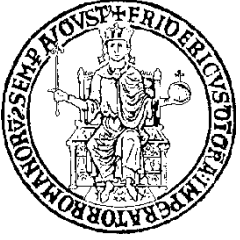
Then, if the pdf is  $P_0^{\text{SS}}$  at  $t = 0$ ,

$$P_{\mu(t)}^{\text{SS}}(x) = \langle \delta(x - x(t)) e^{-A(t)} \rangle$$

Average over initial condition and noise

*Hatano and Sasa, 2001*

# Conservative forces: The Jarzynski relation



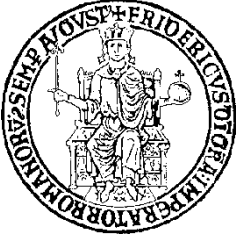
For conservative forces

$$\phi(x, \mu) = \frac{E(x, \mu) - F_\mu}{T}$$

$$\begin{aligned} dA(t) &= \frac{1}{T} \dot{\mu}(t) \left. \frac{\partial(E - F)}{\partial \mu} \right|_{\mu(t), x(t)} \\ &= \frac{1}{T} \dot{\mu}(t) \left. \frac{\partial E}{\partial \mu} \right|_{\mu(t), x(t)} - \frac{1}{T} dF_{\mu(t)} = \frac{1}{T} (dW - dF) \end{aligned}$$

$$\langle \delta(x - x(t)) e^{-W/T} \rangle = e^{-(E(x, \mu(t)) - F_0)/T} = \boxed{P_{\mu(t)}^{\text{eq}}(x) \frac{Z_{\mu(t)}}{Z_0}}$$

Jarzynski, 1997



Consider the joint pdf  $\Phi(x, A, t)$  of  $x$  and  $A$   
Evolution equation for  $\Phi$ :

$$\frac{\partial \Phi}{\partial t} = \hat{L}_\mu \Phi + \dot{\mu} \frac{\partial \phi}{\partial \mu} \frac{\partial \Phi}{\partial A}$$

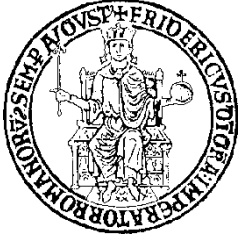
Define

$$\Psi(x, t) = \int dA e^{-A} \Phi(x, A, t)$$

Then

$$\frac{\partial \Psi}{\partial t} = \hat{L}_\mu \Psi - \dot{\mu} \frac{\partial \phi}{\partial \mu} \Psi = \frac{\partial}{\partial t} e^{-\phi(x, \mu(t))}$$





# Driven Brownian particle

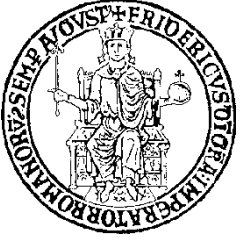
Langevin equation:

$$m\ddot{r}_i = -\gamma\dot{r}_i - \frac{\partial U}{\partial r_i} + f_i + \eta_i(t)$$

$$\langle \eta_i(t) \rangle = 0; \quad \langle \eta_i(t) \eta_j(t') \rangle = 2\gamma T \delta_{ij} \delta(t - t'), \quad \forall t, t'$$

Kramers equation ( $\dot{r}_i = p_i/m$ )

$$\frac{\partial P}{\partial t} = \sum_i \left\{ \left[ \frac{\partial}{\partial r_i} \left( -\frac{p_i}{m} \right) P \right] + \frac{\partial}{\partial p_i} \left[ \left( \gamma \frac{p_i}{m} + \frac{\partial U}{\partial r_i} - \mathbf{f}_i \right) P + \gamma T \frac{\partial}{\partial p_i} P \right] \right\}$$



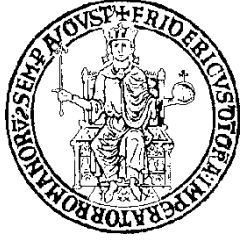
# Energy balance

$$E(x) = E(\vec{r}, \vec{p}) = \sum_i \frac{p_i^2}{2m} + U(\vec{r})$$

$$dE = \underbrace{\left( -\gamma \frac{\vec{p}}{m} + \vec{\eta}(t) \right) \cdot d\vec{r}}_{-dQ_{\text{tot}}} + \underbrace{\vec{f} \cdot d\vec{r} + \frac{\partial E}{\partial \mu} d\mu}_{dW_{\text{ext}}}$$

$$dQ_{\text{tot}} = dQ_{\text{ex}} + \underbrace{\left( \vec{f} - \frac{\partial U}{\partial \vec{r}} + T \frac{\partial \phi}{\partial \vec{r}} \right) \cdot d\vec{r} + \left( -\frac{p}{m} + T \frac{\partial \phi}{\partial \vec{p}} \right) \cdot d\vec{p}}_{dQ_{\text{hk}}}$$

$$dA = \frac{\partial \phi}{\partial \mu} \dot{\mu} dt = \frac{dQ_{\text{ex}}}{T} + d\phi$$

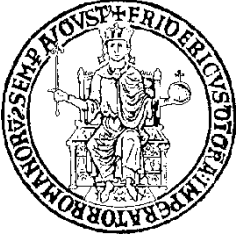


# Generalized Second Law

$$\begin{aligned} 0 &= \ln \text{Tr} P_{\mu(t)}^{\text{SS}} \\ &= \ln \langle \exp(-A) \rangle = \ln \left\langle \exp \left( -\frac{Q_{\text{ex}}}{T} - \Delta\phi \right) \right\rangle \\ &\leq -\frac{1}{T} \langle Q_{\text{ex}} \rangle - \Delta \langle \phi \rangle \end{aligned}$$

$$\langle \phi \rangle = -\text{Tr} \ln P^{\text{SS}} P^{\text{SS}} = S [P^{\text{SS}}]$$

$$\boxed{T \Delta S \geq -\langle Q_{\text{ex}} \rangle}$$



# Entropy balance

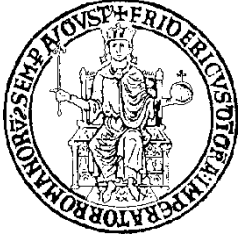
Local entropy:

$$s(x, t) = -\ln P(x, t)$$

*Crooks, 1999; Qian, 2002*

$$dQ_{\text{tot}} = dQ_1 + \underbrace{\left( \vec{f} - \frac{\partial U}{\partial \vec{r}} + T \frac{\partial s}{\partial r} \right) \cdot d\vec{r} + \left( -\frac{p}{m} + T \frac{\partial s}{\partial p} \right) \cdot d\vec{p}}_{dQ_2}$$

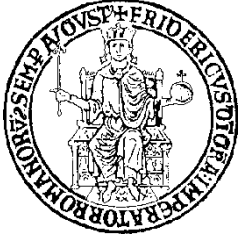
$$\left\langle \frac{dQ_1}{T} \right\rangle = ds$$



# *Loschmidt's paradox*

Thomson (1874) and Loschmidt (1876):

To every initial state  $x_0$  of a mechanical system leading to a decrease in Boltzmann's  $H$  function, corresponds an initial state  $I x_0$  leading to its increase



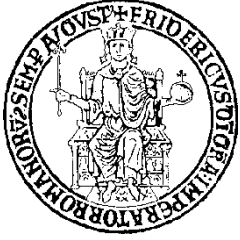
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There are infinitely many more states *in a large system* leading to a decrease in  $H$  than those leading to its increase



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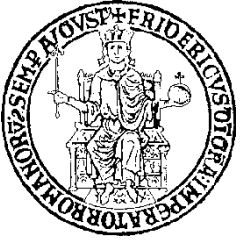
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In small systems, transient increases of Boltzmann's  $H$  are to be expected

# Microscopic reversibility I



Mechanical system described by  $x = (p, r)$

Time reversal operator:  $I x = (-p, r)$

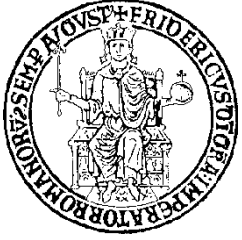
Time-reversal invariance of the hamiltonian:  $E(I x) = E(x)$

Solution of the equations of motion:

$$x(t, x_0) : \quad (\dot{p}, \dot{r}) = \left( -\frac{\partial E}{\partial r}, \frac{\partial E}{\partial p} \right)$$







# Microscopic reversibility II

Stochastic evolution equation for the pdf  $P(x, t)$ :

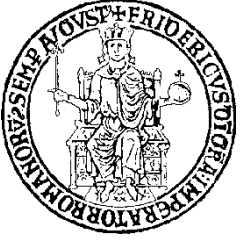
$$\frac{\partial P}{\partial t} = \hat{L}_\mu P$$

Microscopic reversibility:

$$Q \psi(x) = e^{-E(x)/T} \psi(Ix) : \quad Q^{-1} \hat{L} Q = \hat{L}^\dagger$$

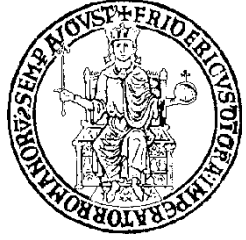
For the Kramers equation with non-conservative force  $\vec{f}$ :

$$Q^{-1} \hat{L} Q = \hat{L}^\dagger - \frac{\vec{f}}{T} \cdot \left( \frac{\vec{p}}{m} \right)$$



# Observations

- ⑥ I speak of Boltzmann's  $H$  and not of the entropy:  $H$  is a dynamic observable and the entropy is not
- ⑥ Typically the probability of a fluctuation is  $\propto e^{-\mathcal{F}/T}$
- ⑥ Thus a system is “small” if free energy differences are  $O(T)$



# The time-reversal relation

Evolution operator:

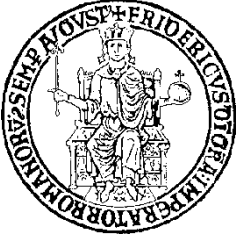
$$\mathcal{U}(t, t_0) = \text{T exp} \left( \int_{t_0}^t dt' \hat{L}_{\mu}(t') \right)$$

$$\tilde{\mathcal{U}}(t, t_0) := \mathcal{Q}_t^{-1} \mathcal{U}(t, t_0) \mathcal{Q}_{t_0}$$

satisfies

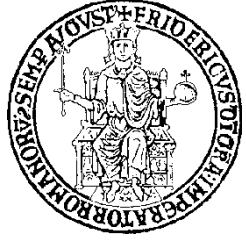
$$\begin{aligned} \frac{\partial}{\partial t} \tilde{\mathcal{U}}(t, t_0) &= \dot{\mathcal{Q}}_t^{-1} \mathcal{U}(t, t_0) \mathcal{Q}_{t_0} + \mathcal{Q}_t^{-1} \hat{L}_{\mu}(t) \mathcal{Q}_t \tilde{\mathcal{U}}(t, t_0) \\ &= \left[ \frac{\partial E/T}{\partial t} + \hat{L}_{\mu}^{\dagger}(t) - \frac{\vec{f}}{T} \cdot \frac{\vec{p}}{m} \right] \tilde{\mathcal{U}}(t, t_0) \end{aligned}$$

# Transition probabilities for short time intervals



$$\begin{aligned} & Q_{t+\Delta t}^{-1} \mathcal{U}(x', t + \Delta t; x, t) Q_t \\ &= e^{((E(x', t+\Delta t) - E(x, t))/T)} \mathcal{U}(I x', t + \Delta t; I x, t) \\ &= \mathcal{U}(x, t + \Delta t; x', t) \exp \left[ - \left( \frac{\vec{f}}{T} \cdot \frac{\vec{p}}{m} - \frac{\partial E(x')}{\partial t} \right) \Delta t \right] \end{aligned}$$

$$\begin{aligned} & \frac{\mathcal{U}(x', t + \Delta t; x, t)}{\mathcal{U}(I x, t + \Delta t; I x', t)} \\ &= \exp \left[ \left( \vec{f} \cdot \frac{\vec{p}}{m} \Delta t - \frac{\partial E}{\partial x} (x' - x)/T \right) \right] \\ &= \exp (dQ_2/T) \end{aligned}$$



# Crooks's reversal relation

A path is coarsely defined by the “gates”

$$\omega = (x_0, 0) \longrightarrow (x_1, t_1) \longrightarrow \dots \\ \dots \longrightarrow (x_{k-1}, t_{k-1}) \longrightarrow (x_k = x, t_k = t)$$

Then

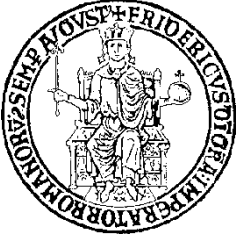
$$\frac{P(\omega, \mu | x_0, 0)}{P(\tilde{\omega}, \tilde{\mu} | I x, t)} = \left\langle \exp \left( \int_0^t \frac{dQ_2}{T} \right) \right\rangle$$

where

$$\tilde{\omega} = (I x, t_0) \longrightarrow (I x_{k-1}, \tilde{t}_{k-1}) \longrightarrow \dots \longrightarrow (I x_1, \tilde{t}_1) \longrightarrow (I x_0, t)$$

$$\tilde{t}_k = t - t_k, \tilde{\mu}(t) = \mu(\tilde{t})$$

Average over all paths  $x(t)$  conditioned by the gates



# Seifert's relation

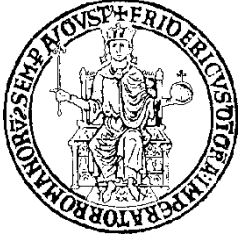
Continuous limit: paths  $\omega = x(t), \forall t$

Arbitrary initial pdf's  $p_0(x_0)$  for  $\omega$  and  $p_1(Ix)$  for  $\tilde{\omega}$  :

$$\begin{aligned} R[\omega, p_0, p_1] &:= \ln \frac{P(\omega, \mu | x_0, 0) p_0(x_0)}{P(\tilde{\omega}, \tilde{\mu} | Ix, t) p_1(Ix)} \\ &= \int_0^t \frac{dQ_2}{T} + \ln \frac{p_0(x_0)}{p_1(x_1)} \end{aligned}$$

Averaging over the paths

$$\begin{aligned} \langle e^{-R} \rangle &= \text{Tr} P(\omega, \mu | x_0, 0) p_0(x_0) e^{-R} \\ &= \text{Tr} P(\tilde{\omega}, \tilde{\mu} | Ix, t) p_1(Ix) = 1 \end{aligned}$$



# A first fluctuation theorem

1. Entropy production must sometimes be negative!
2. Take  $p_0(x, 0)$  arbitrary,  $p_1(x) = P(x, t)$  (starting from this initial condition)

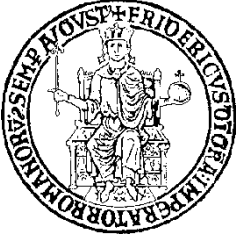
$$R = \int \frac{dQ_2}{T} - \Delta s = \int \frac{dQ_{\text{tot}}}{T} = \Delta s_{\text{tot}}$$

Thus

$$\frac{P(\Delta s_{\text{tot}})}{P(-\Delta s_{\text{tot}})} = e^{\Delta s_{\text{tot}}}$$



# The Gallavotti-Cohen fluctuation theorem

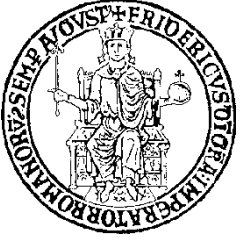


3. In particular for  $\mu(t) = \text{const.}$ ,  $p_0(x) = p_1(x) = P_\mu^{\text{SS}}(x)$ :

$$R \simeq \frac{\dot{Q}_2}{T} t = \sigma t$$

$$\boxed{\frac{P(\sigma)}{P(-\sigma)} = e^{\sigma t}}$$

*Gallavotti and Cohen, 1995*



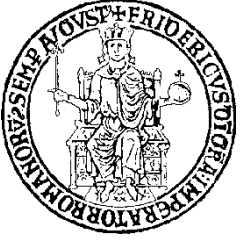
## Back to Jarzynski

4. Take  $p_0 = \exp(-(E_{\mu(0)} - F_{\mu(0)})/T)$  and  $p_1 = \exp(-(E_{\mu(t)} - F_{\mu(t)})/T)$ . Then

$$\begin{aligned} R &= \Delta S_m + [(E(x, t) - F_{\mu(t)}) - (E(x, 0) - F_{\mu(0)})] / T \\ &= W_d / T \end{aligned}$$

Thus

$$1 = \langle e^{-W_d/T} \rangle = \langle e^{-W/T} \rangle e^{\Delta F/T}$$



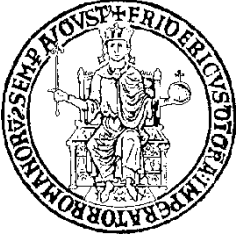
# Evaluation of free-energy landscapes

$$\mathcal{F}_0(M) = -T \ln \text{Tr} \delta(M - M(x)) e^{-E_0(x)/T}$$

$$Z_0 = \int dM e^{-\mathcal{F}_0(M)/T} = \text{Tr} e^{-E_0(x)/T}$$

Manipulation:  $t \longrightarrow E_{\mu(t)}(x)$ ,  $E_{\mu(0)}(x) = E_0(x)$ ,

$$E_{\mu}(x) = E_0(x) + U_{\mu}(M(x))$$

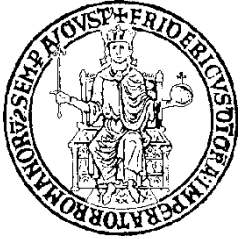


# The basic identity

*A. Imparato and L. Peliti, 2005*

$$\begin{aligned}\langle \delta(M - M(x)) e^{-\beta W} \rangle_t &= \int dx \delta(M - M(x)) \frac{e^{-\beta E_{\mu(t)}(x)}}{Z_0} \\ &= e^{-(\mathcal{F}_0(M) - F_0)/T} e^{-U_{\mu(t)}(M)/T}\end{aligned}$$

Generalization of Hummer and Szabo, 2001



# The histogram method

Thus

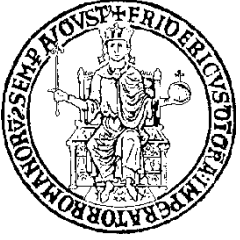
$$e^{U_{\mu(t)}(M)/T} \langle \delta(M - M(x)) e^{-W/T} \rangle_t = e^{-(\mathcal{F}_0(M) - F_0)/T}$$

$\mathcal{N}$  trajectories  $(M_t^k, W_t^k)$ , sampled at discrete times  $t_j$

Discrete bins  $M_\ell \leq M \leq M_\ell + \Delta M_\ell$

$$\begin{aligned} r(M_\ell, t_j) &= Z_0 e^{U_{\mu(t_j)}(M_\ell)/T} \overline{\theta_\ell(M(t_j)) e^{-W/T}} \\ &= Z_0 e^{U_{\mu(t)}(M_\ell)/T} \frac{1}{\mathcal{N}} \sum_{k=1}^{\mathcal{N}} \theta_\ell(M_{t_j}^k) e^{-W_{t_j}^k/T} \end{aligned}$$

$$\Delta R(M_\ell) = e^{-(\mathcal{F}_0(M_\ell) - F_0)/T} \delta M_\ell = \langle r(M_\ell, t_j) \rangle, \quad \forall \ell, j$$



# The best estimate

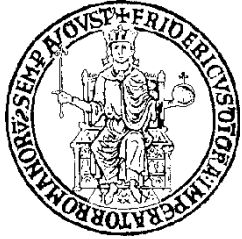
$$\Delta R^*(M_\ell) = \sum_j r(M_\ell, t_j) p_j$$

$$0 \leq p_j \leq 1, \quad \sum_j p_j = 1$$

Best estimate:

$$p_j = \frac{\lambda}{\text{Var } r(M_\ell, t_j)} \propto \frac{e^{U_{\mu(t_j)}(M_\ell)/T}}{e^{-W_{t_j}}}$$

*Braun et al., 2004*



# A mean-field Ising model

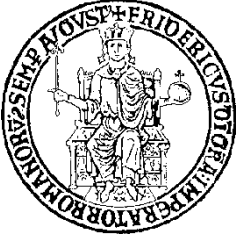
$$\mathcal{F}_0(M) = -\frac{J}{2N}M^2 - TS(M)$$

$$S(M) = - \left[ \left( \frac{N+M}{2} \right) \log \left( \frac{N+M}{2} \right) + \left( \frac{N-M}{2} \right) \log \left( \frac{N-M}{2} \right) \right]$$

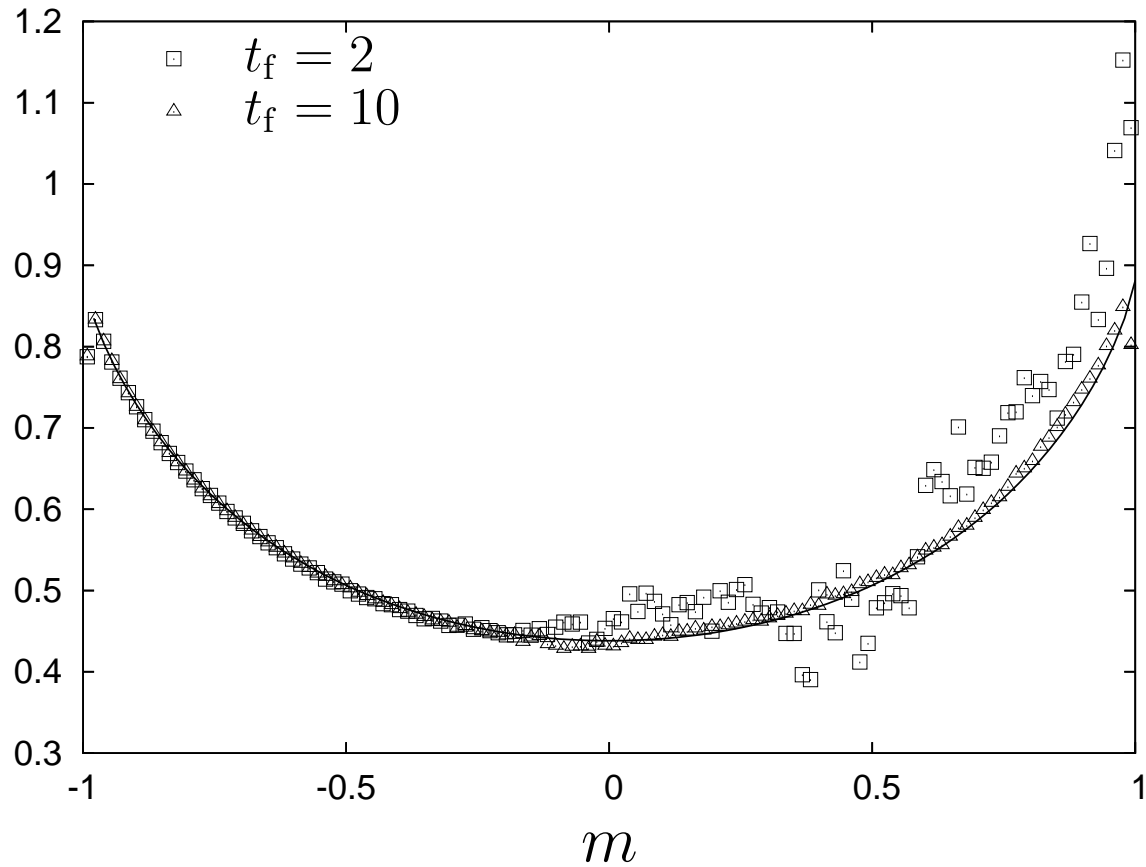
$$U_h(M) = -hM$$

$$m = \frac{M}{N}$$

$$f_0^*(m) = -\frac{T}{N} \ln R^*(M)$$



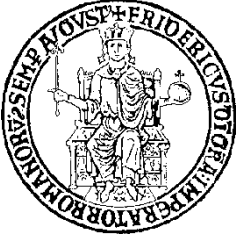
# Linear protocol



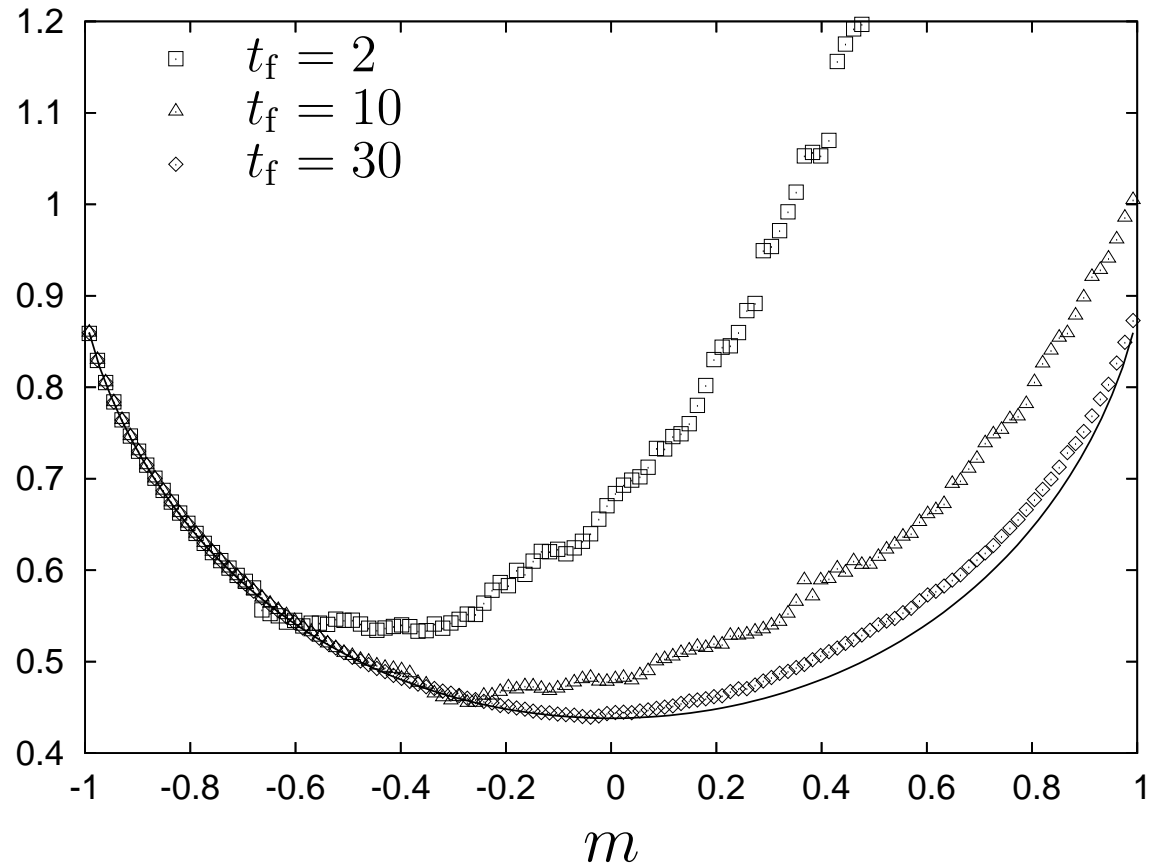
$$h(t) = h_0 + \frac{h_1 - h_0}{t_f} t$$

$$h_1 = -h_0 = 1, t_f = 2, 10, N = 10, \mathcal{N} = 10^4 \text{ samples}$$



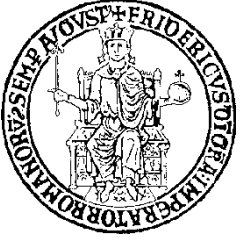


# A larger system

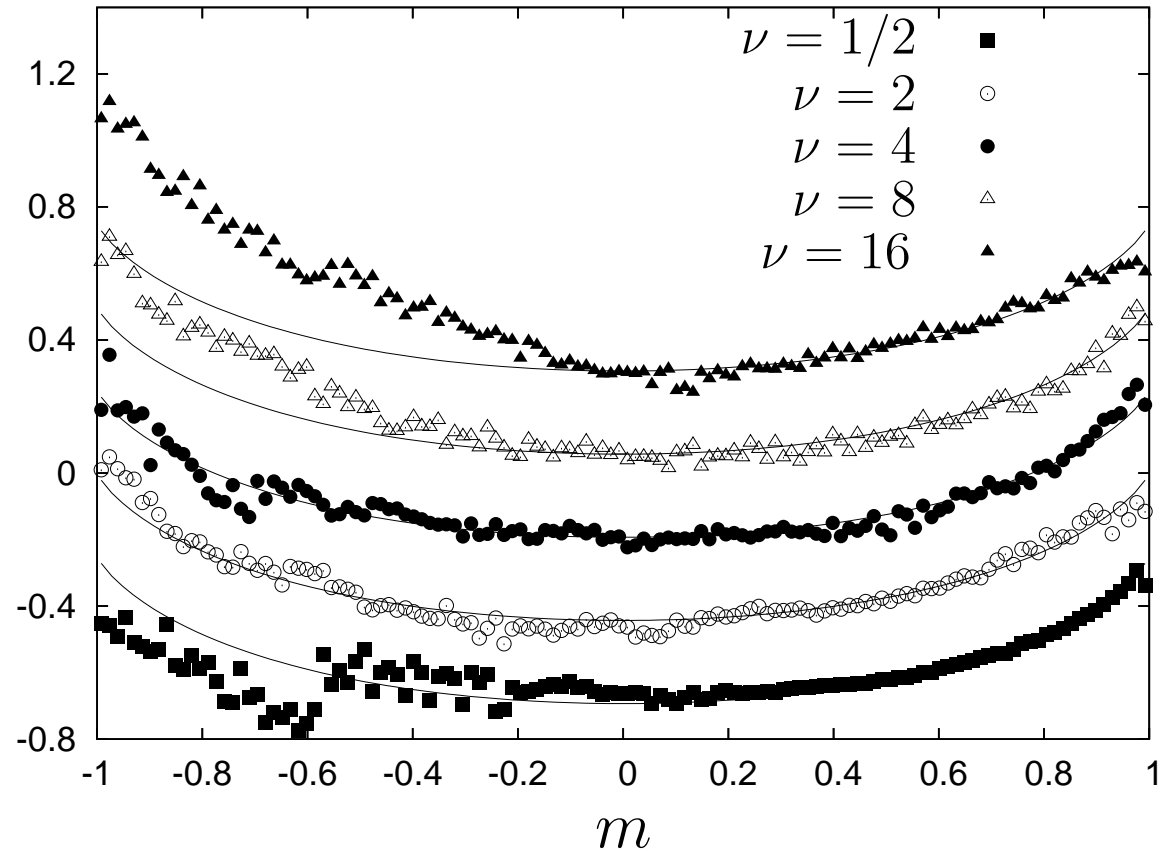


$$h(t) = h_0 + \frac{h_1 - h_0}{t_f} t$$

$$h_1 = -h_0 = 1, t_f = 2, 10, N = 100, \mathcal{N} = 10^4 \text{ samples}$$

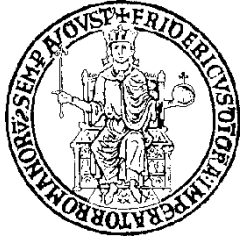


# Oscillatory protocol

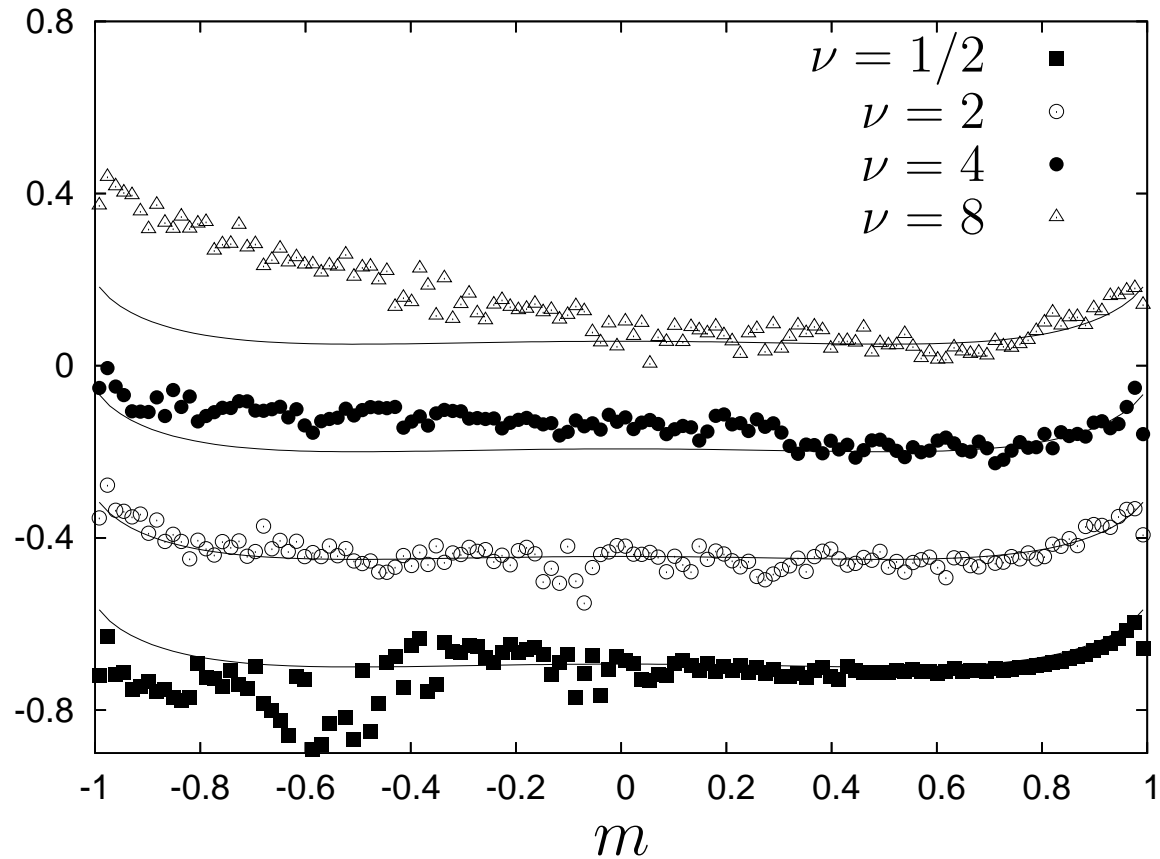


$$h(t) = h_0 \sin(2\pi\nu t), \quad 0 \leq t \leq t_f$$

$$h_0 = 1, N = 10, J_0 = 0.5, t_f = 2, \mathcal{N} = 10^4 \text{ samples}$$

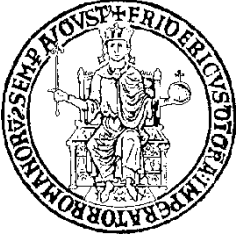


# At lower temperatures

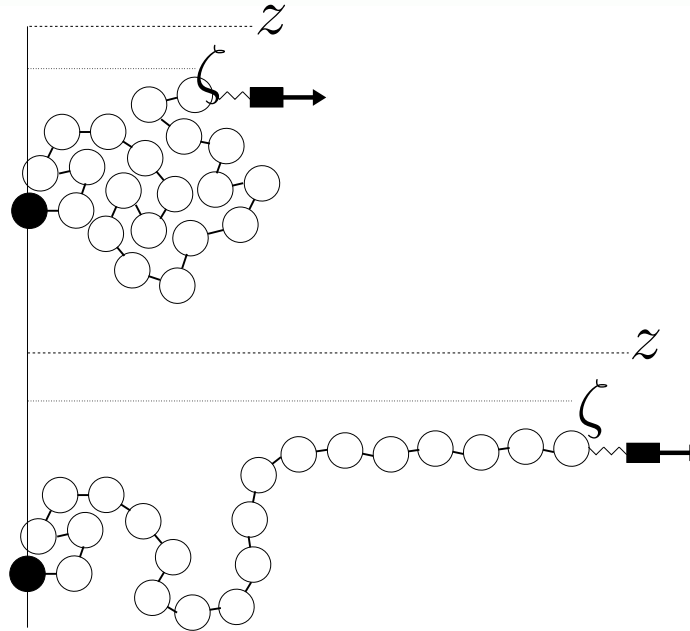


$$h(t) = h_0 \sin(2\pi\nu t), \quad 0 \leq t \leq t_f$$

$$h_0 = 1, N = 10, J_0 = 1.1, t_f = 2, \mathcal{N} = 10^4 \text{ samples}$$



# Unzipping of a model homopolymer

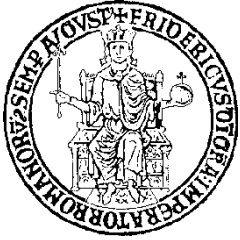


$$U_{z(t)}(\zeta) = \frac{1}{2}k (\zeta - z(t))^2$$

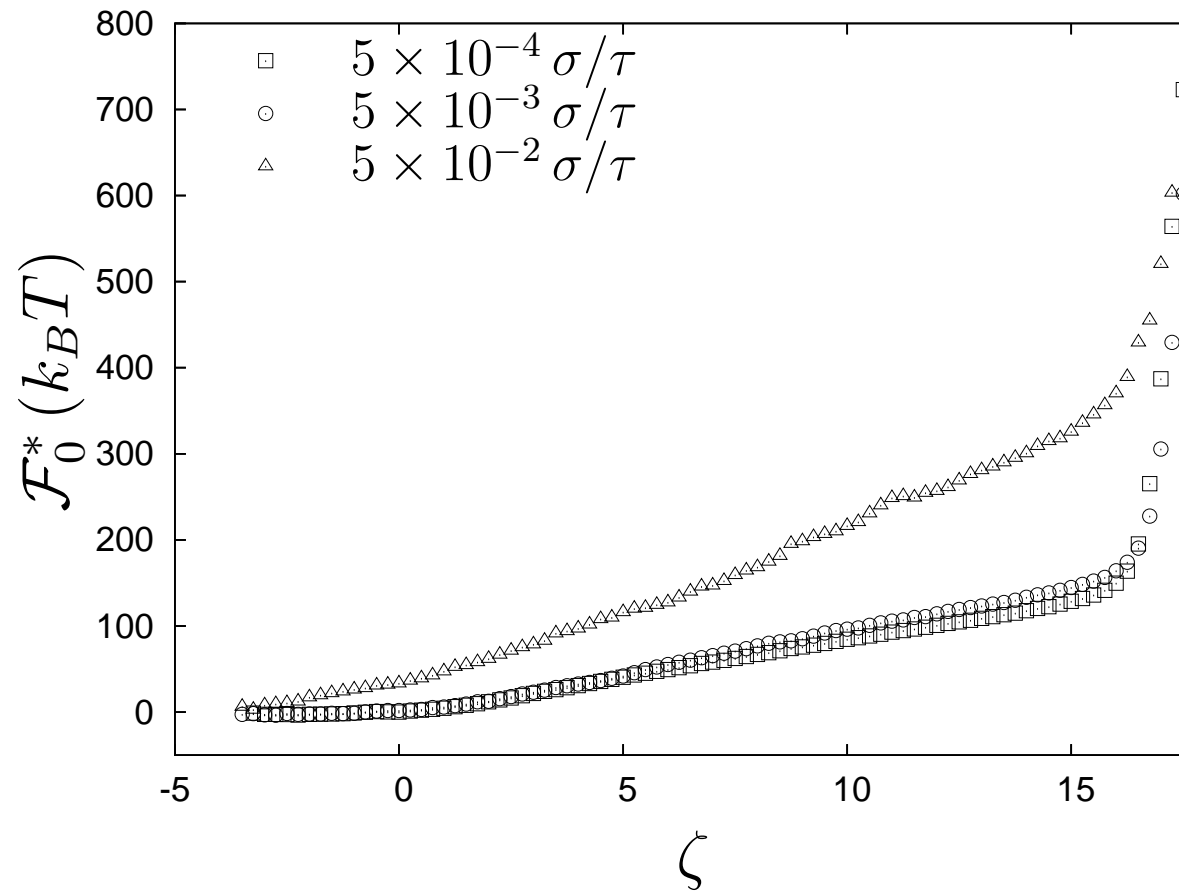
L-J potential ( $\epsilon, \sigma$ ) + harmonic potential for successive beads

$$N = 20, \sigma = 0.5 \text{ nm}, \epsilon = 1 \text{ kcal/mol}, m = 3 \cdot 10^{-25} \text{ kg } \tau = \sqrt{m\sigma^2/\epsilon} \simeq 3.3 \text{ ps},$$

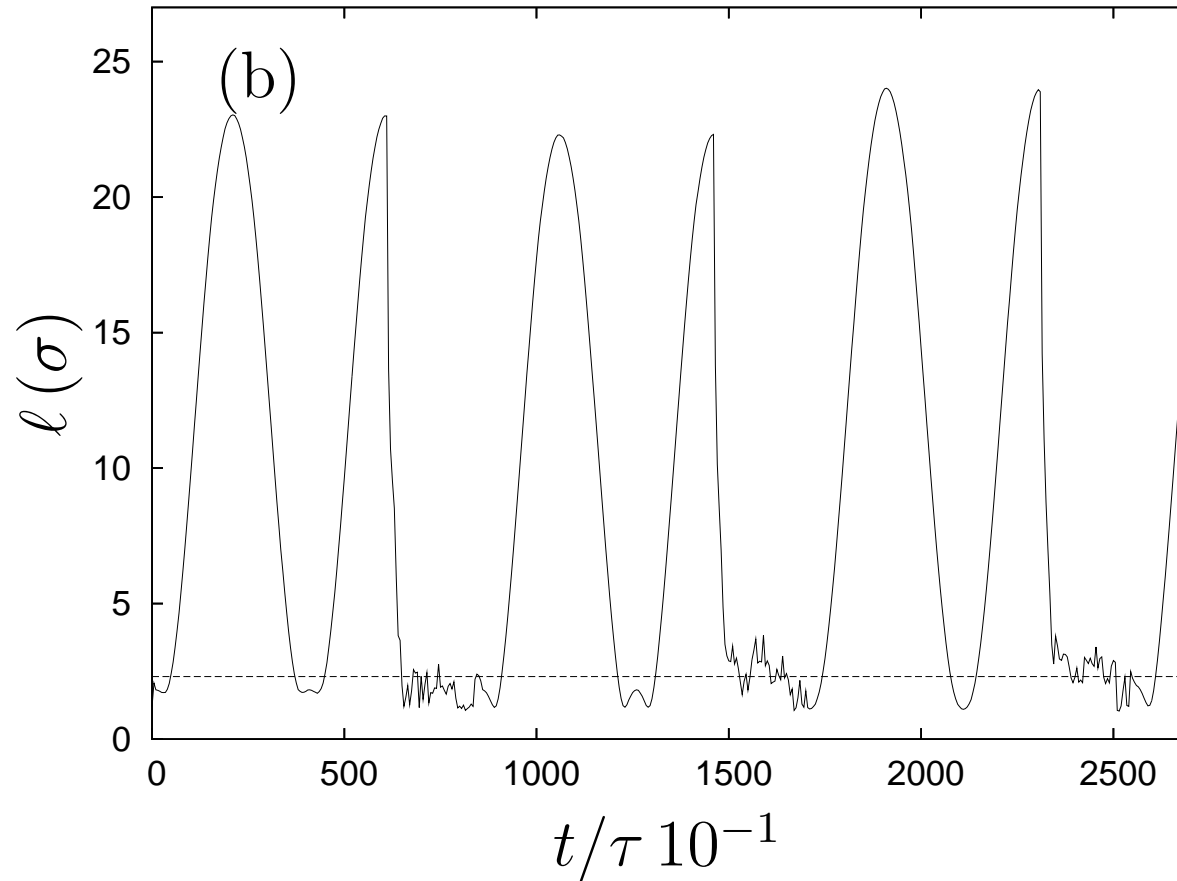
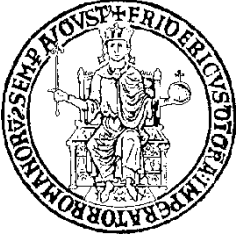
$$\gamma = 15m/\tau, k = 5000\epsilon/\sigma^2, T = 300 \text{ K}$$

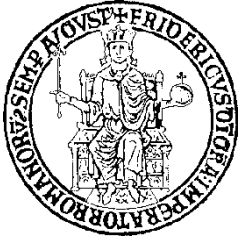


# Linear protocol

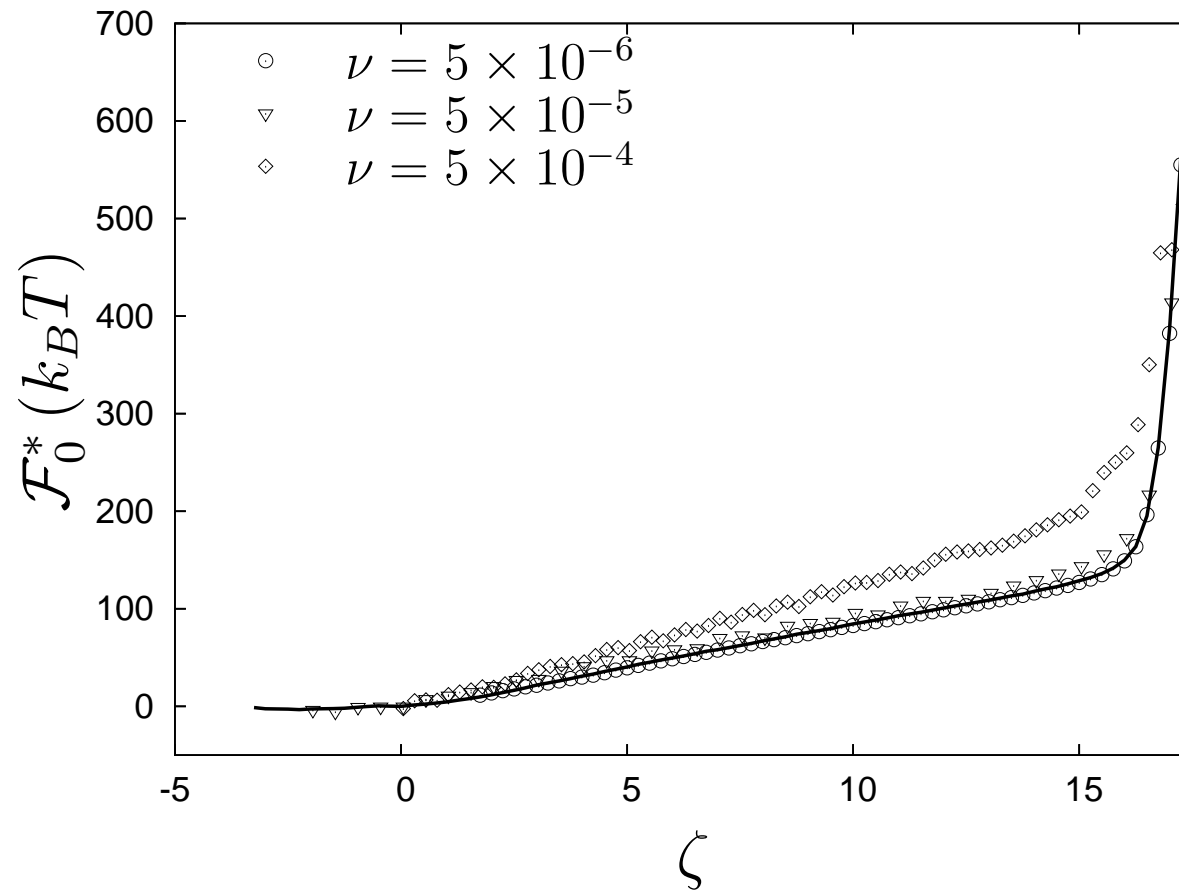


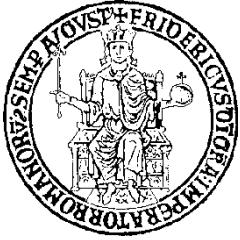
# Oscillatory protocol: "Pulsed" protocol



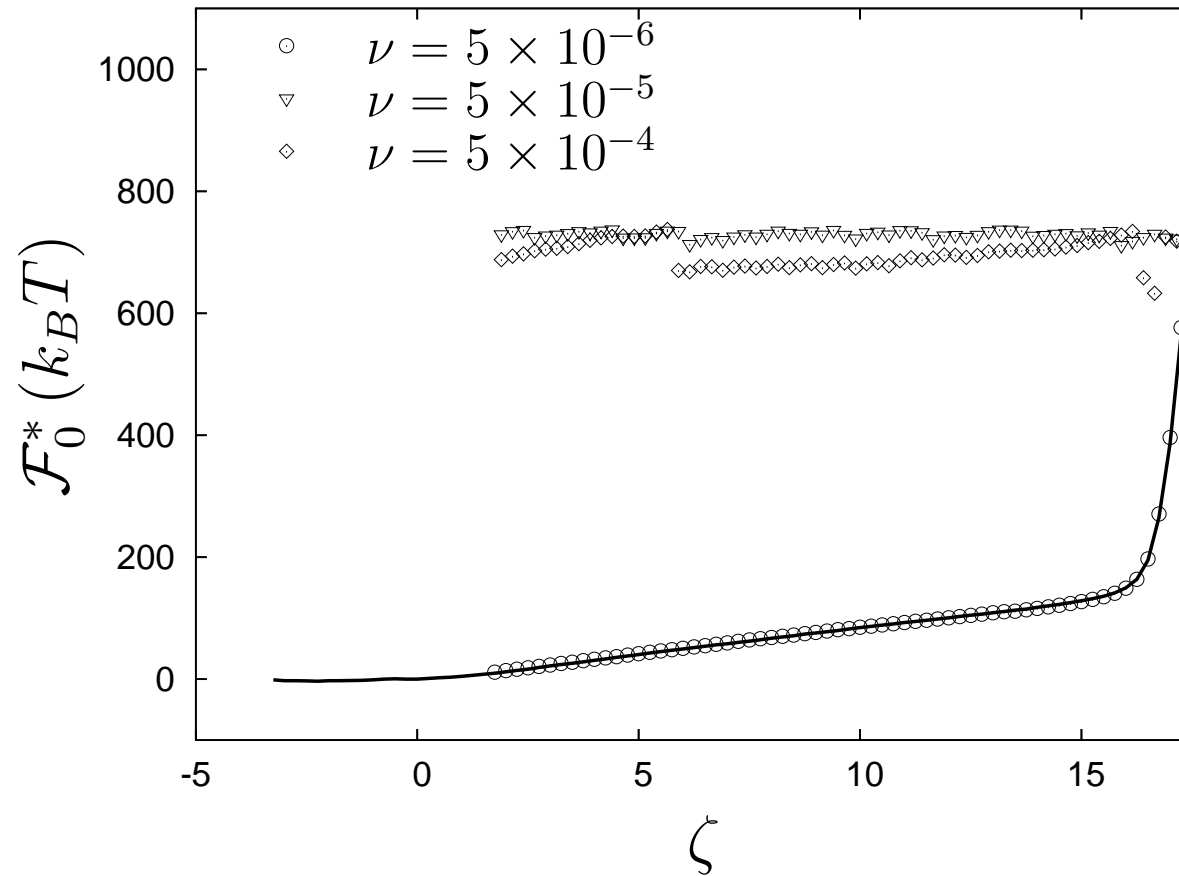


# The free energy

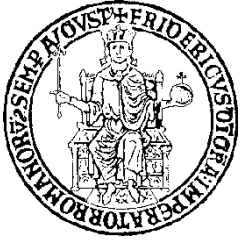




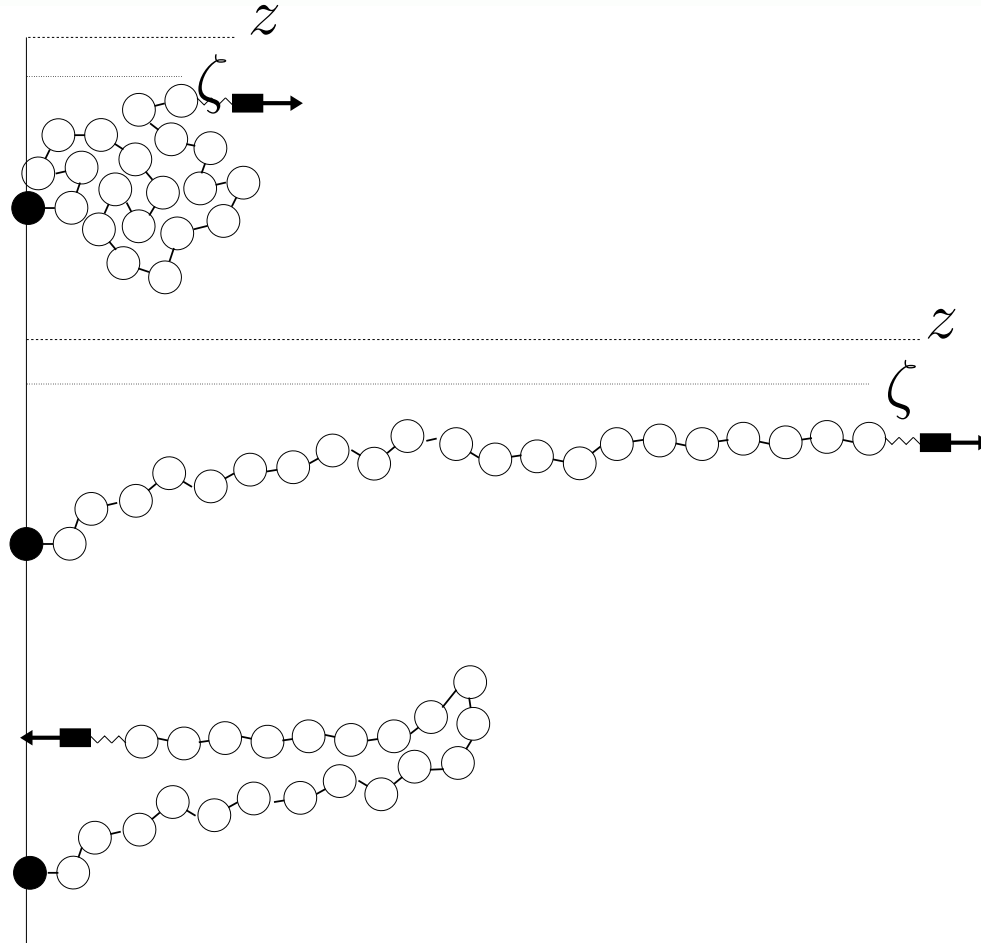
# The “always attached” protocol

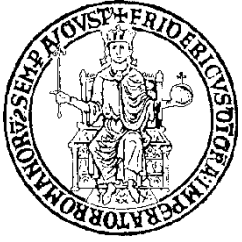




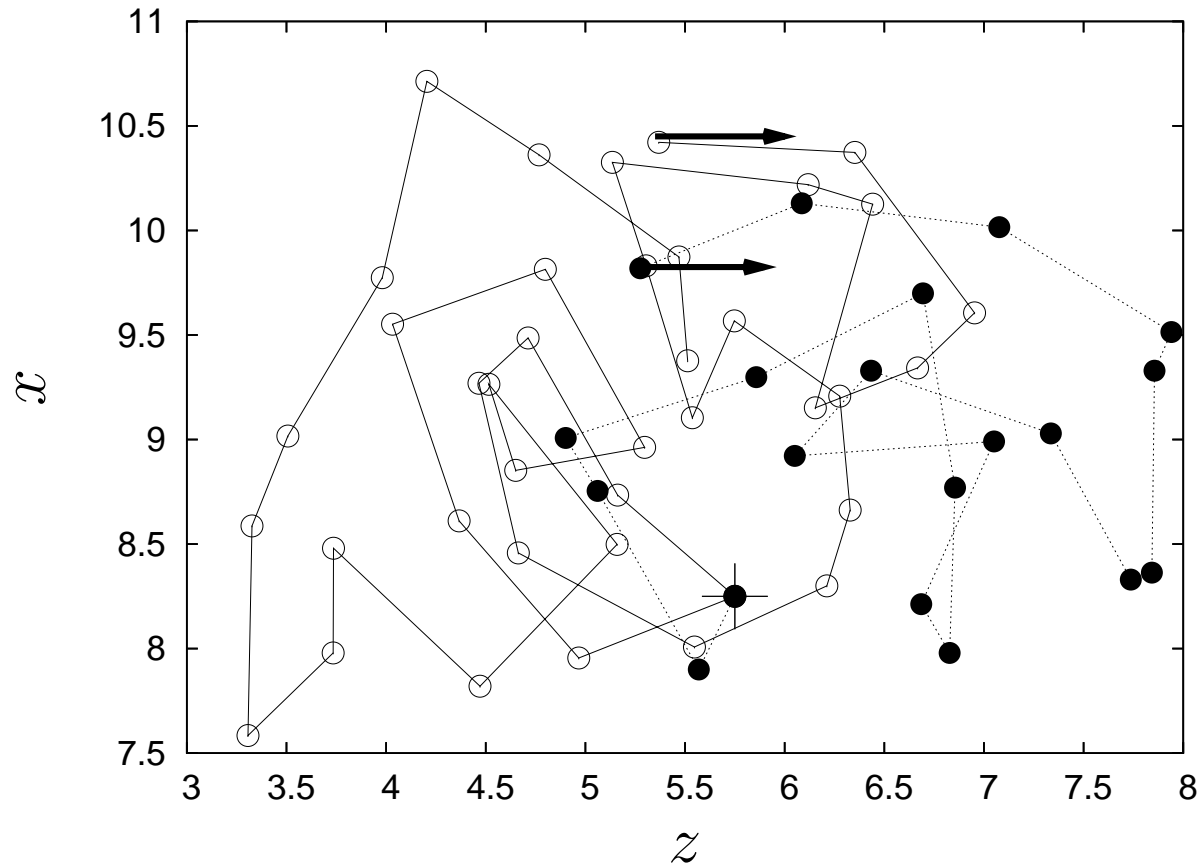


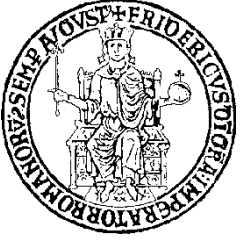
# What is happening?





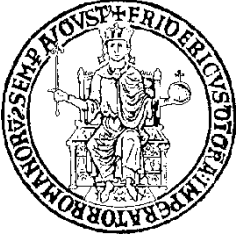
# The configurations





## Discussion

- ⑥ The JE is effective (via the histogram method) to reconstruct free-energy landscapes for systems small enough (small energy barriers)
- ⑥ Care must be taken that the monitored collective coordinate is “good”, i.e., that the distribution of the *transverse* degrees of freedom is sufficiently sampled during manipulation
- ⑥ The choice of the manipulation protocol affects the reliability of the results



# Exploring nonequilibrium systems

*C. Giardinà, J. Kurchan, L. Peliti, 2005*

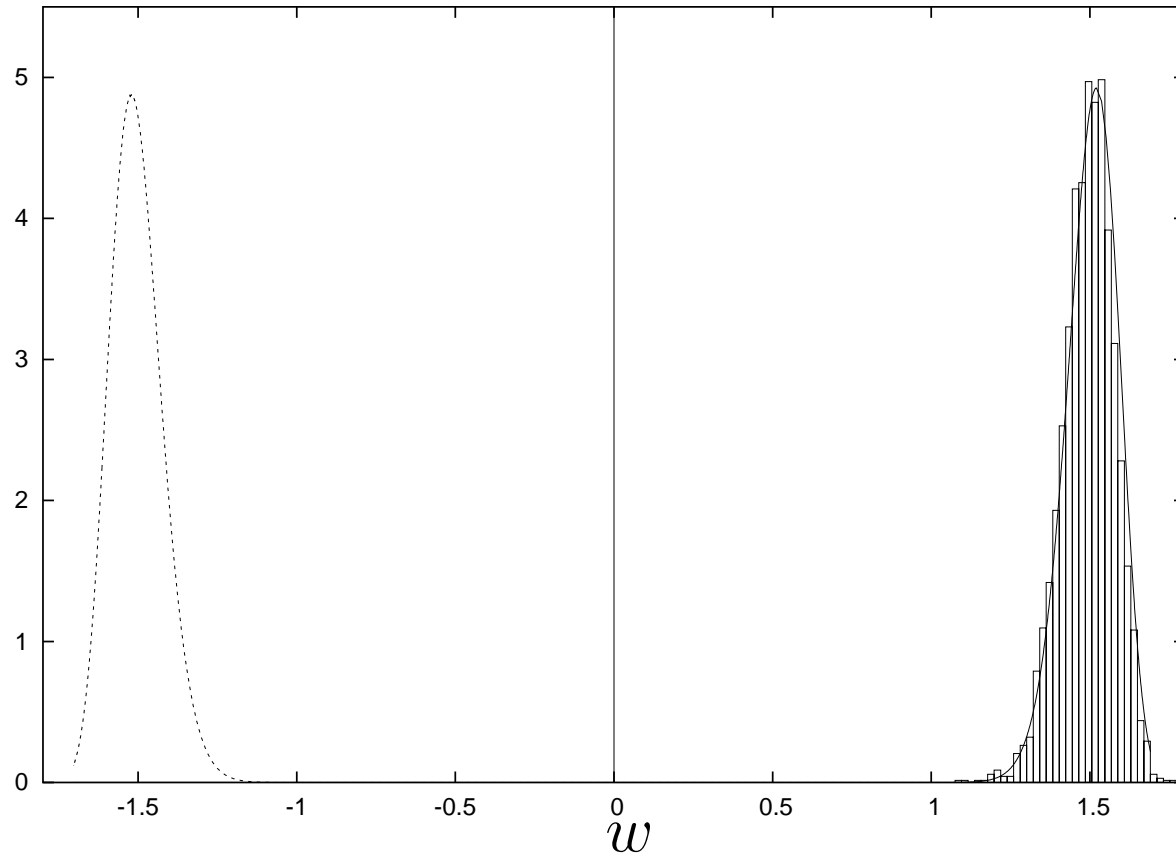
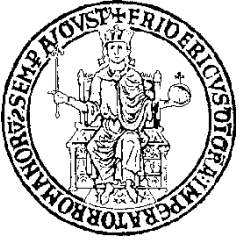
Evolution equations:

$$\frac{\partial \Psi}{\partial t} = \hat{L} \Psi + A \Psi$$

Then  $\Psi$  is given by a weighted average:

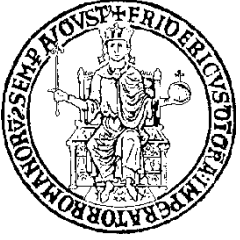
$$\Psi(x, t) = \left\langle \delta(x - x(t)) \exp \left( \int_0^t dt' A(t') \right) \right\rangle$$

# The weight can be wild



Work distribution  $P(W/N)$  for the Ising model

Dashed line: Weighted work distribution  $P(W/N)e^{-W/T}$

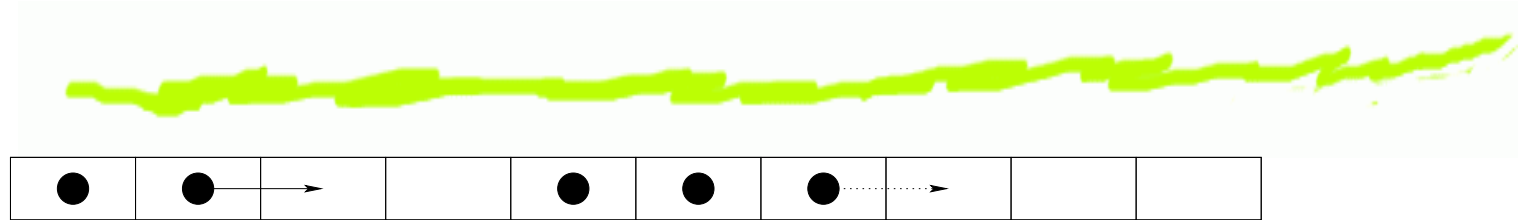
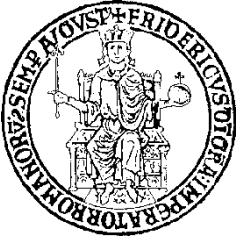


# *Can we improve our statistics?*

- ⑥ Interpret  $\Psi(x, t)$  as a density of walkers
- ⑥ Walkers move according to the Langevin equation
- ⑥ Walkers reproduce or die depending on the local value of  $A$
- ⑥ Thus  $\Psi$  samples the weighted probability, not the original one!

# The Totally Asymmetric Exclusion

## Process (TASEP)



At any given time step  $t$ , a given particle moves to the right with probability  $\alpha$  if the target site is empty

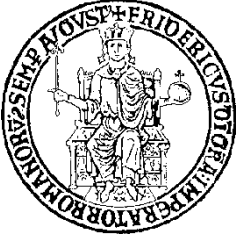
Configuration  $\mathcal{C} = (n_i)$ ,  $n_i \in \{0, 1\}$ ,  $i = 1, L$ , periodic b.c.

Current  $J$ :

$$J_{\mathcal{C}'\mathcal{C}} = \begin{cases} 1, & \text{if one particle jumps to the right;} \\ 0, & \text{if nothing happens.} \end{cases}$$

We wish to evaluate

$$e^{\Lambda(\lambda)} = \left\langle \exp \left( \lambda \sum_t J_{\mathcal{C}_{t+1}\mathcal{C}_t} \right) \right\rangle$$



# The large-deviation function

$$\text{Prob}[\mathcal{C}_0, \mathcal{C}_1, \dots, \mathcal{C}_T] = U_{\mathcal{C}_T \mathcal{C}_{T-1}} \cdots U_{\mathcal{C}_2 \mathcal{C}_1} \cdot U_{\mathcal{C}_1 \mathcal{C}_0}$$

$$e^{\Lambda(\lambda)} = \sum_{\mathcal{C}_1, \dots, \mathcal{C}_T} \tilde{U}_{\mathcal{C}_T \mathcal{C}_{T-1}} \cdots \tilde{U}_{\mathcal{C}_1 \mathcal{C}_0} = \sum_{\mathcal{C}_T} \left[ \tilde{U}^T \right]_{\mathcal{C}_T \mathcal{C}_0}$$

where

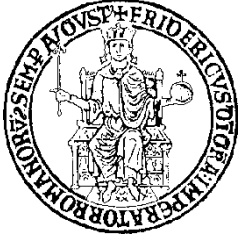
$$\tilde{U}_{\mathcal{C}'\mathcal{C}} := e^{\lambda J_{\mathcal{C}'\mathcal{C}}} U_{\mathcal{C}'\mathcal{C}}$$

Define

$$K_{\mathcal{C}} := \sum_{\mathcal{C}'} \tilde{U}_{\mathcal{C}'\mathcal{C}}, \quad U'_{\mathcal{C}'\mathcal{C}} \equiv \tilde{U}_{\mathcal{C}'\mathcal{C}} K_{\mathcal{C}}^{-1}$$

$$e^{\Lambda(\lambda)} = \sum_{\mathcal{C}_2, \dots, \mathcal{C}_T} U'_{\mathcal{C}_T \mathcal{C}_{T-1}} K_{\mathcal{C}_{T-1}} \cdots U'_{\mathcal{C}_1 \mathcal{C}_0} K_{\mathcal{C}_0}$$



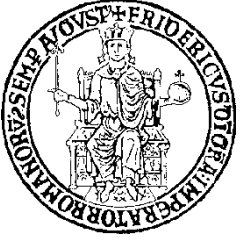


# The simulation steps

- 6 A cloning step:

$$P_c(t + 1/2) = K_c P_c(t)$$

$$G \text{ clones of } \mathcal{C} : G = \begin{cases} [K_c] + 1, & \text{with probability } K_c - [K_c] \\ [K_c], & \text{otherwise} \end{cases}$$



# The simulation steps

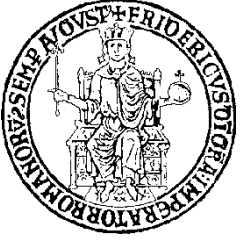
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- 6 A shift step:

$$P_{c'}(t + 1) = \sum_c U'_{c'c} P_c(t + 1/2)$$



# The simulation steps

- ⑥ A cloning step:

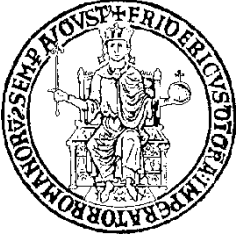
$$P_c(t + 1/2) = K_c P_c(t)$$

$$G \text{ clones of } \mathcal{C} : G = \begin{cases} [K_c] + 1, & \text{with probability } K_c - [K_c] \\ [K_c], & \text{otherwise} \end{cases}$$

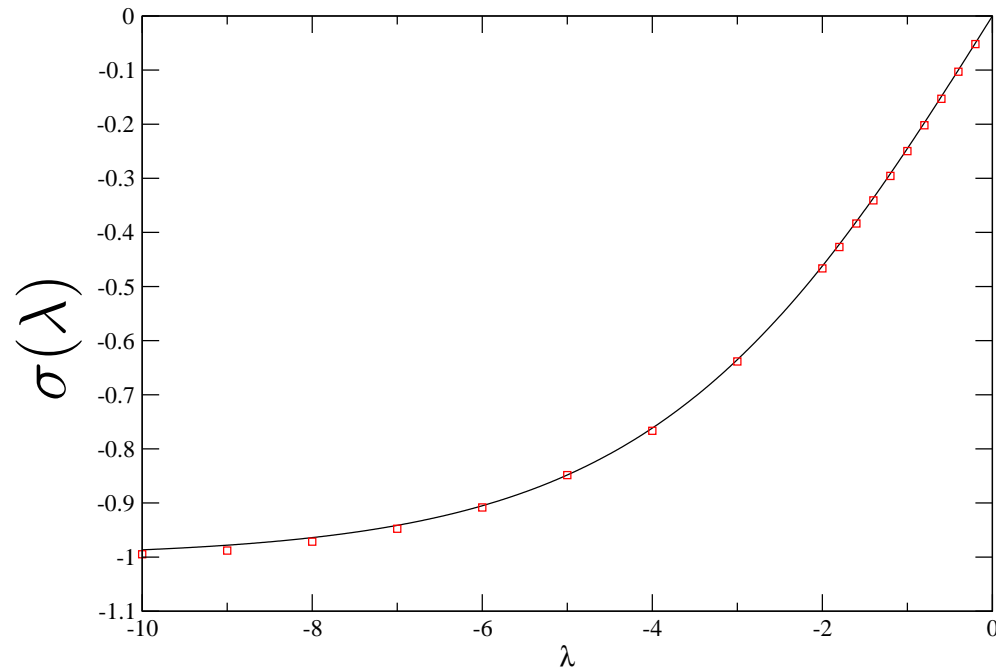
- ⑥ A shift step:

$$P_{c'}(t + 1) = \sum_c U'_{c'c} P_c(t + 1/2)$$

- ⑥ Overall cloning step with an adjustable rate  
 $M_t = N/(N + G)$  (the same for all configurations)

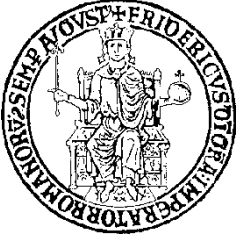


# Results

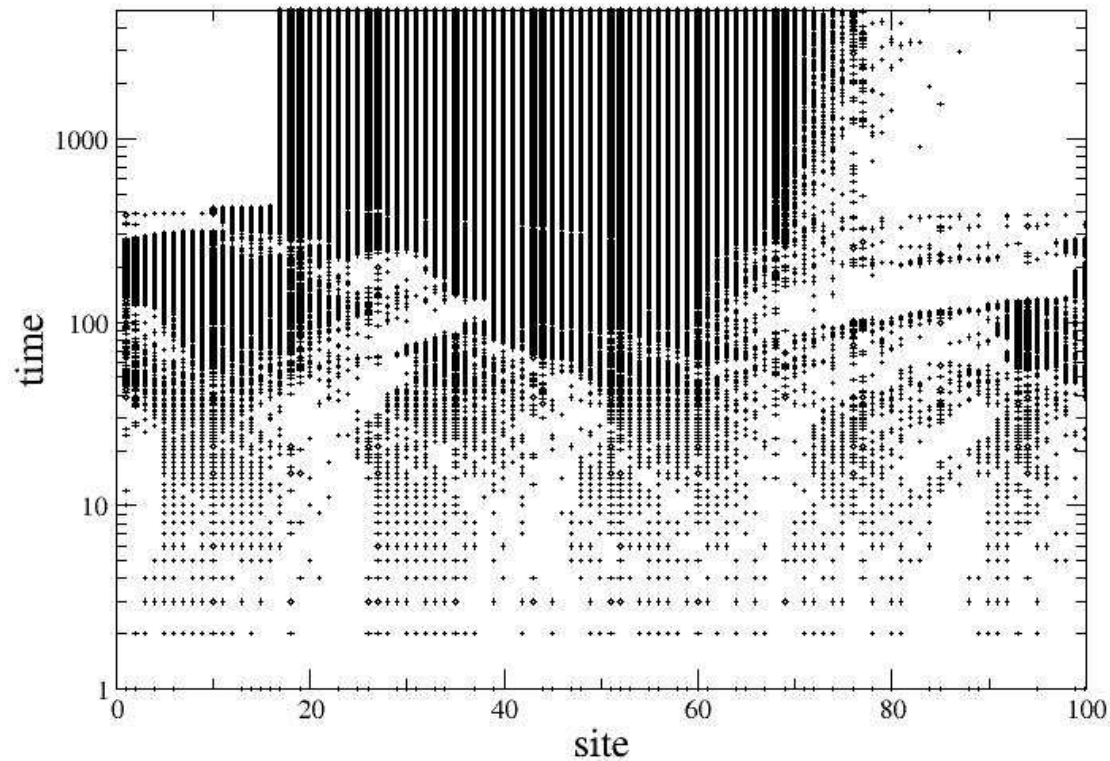


For long times

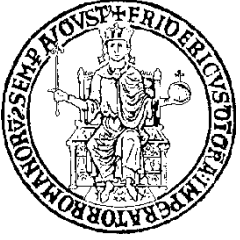
$$- \lim_{t \rightarrow \infty} \frac{1}{t} \ln[M_T \cdots M_2 \cdot M_1] = \lim_{t \rightarrow \infty} \frac{\Lambda(\lambda)}{t} = \sigma(\lambda)$$



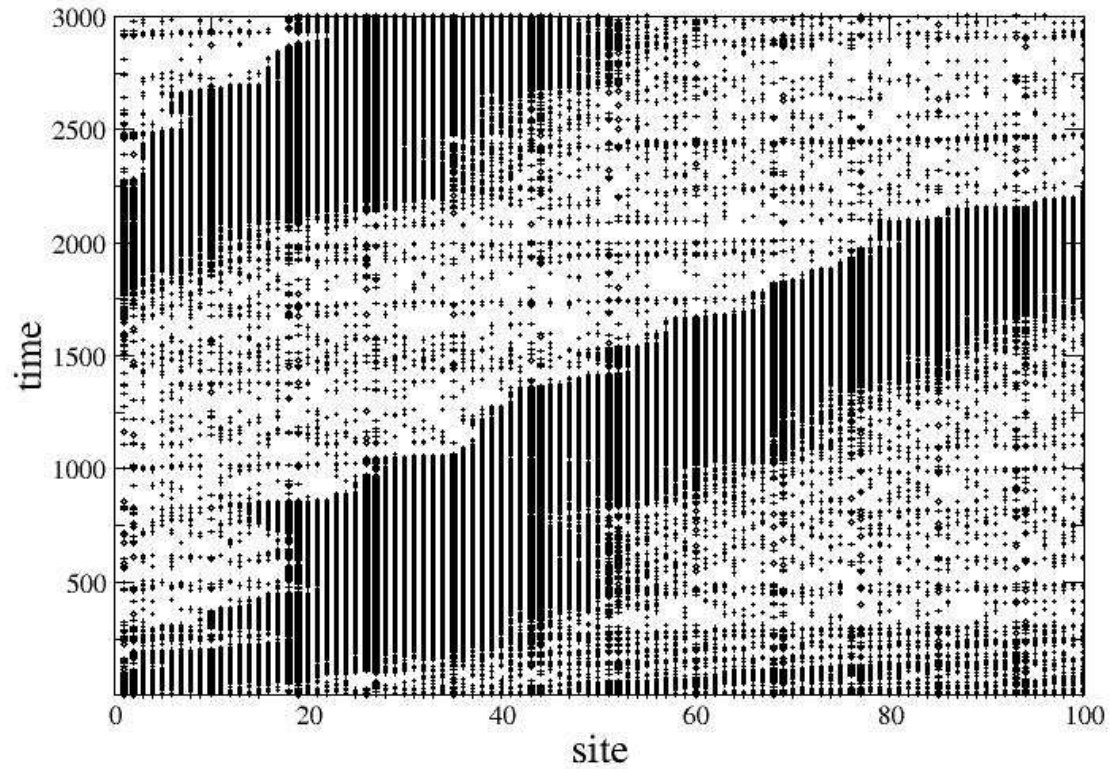
# The configurations



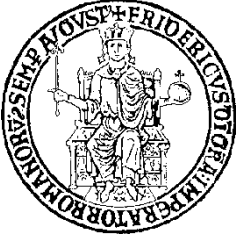
Space-time diagram for a ring of  $N = 100$  sites,  $\lambda = -50$  and density 0.5  
Note the logarithmic scale on the  $y$ -axis



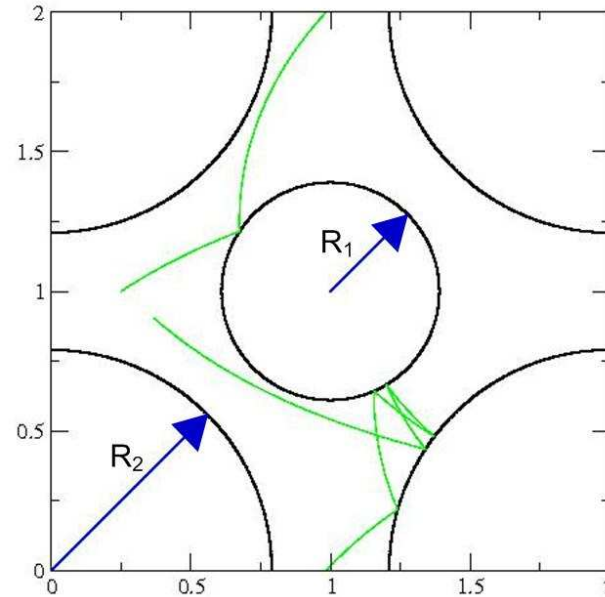
# *Moving shock waves*



Space-time diagram for a ring of  $N = 100$  sites,  $\lambda = -30$  and density 0.3



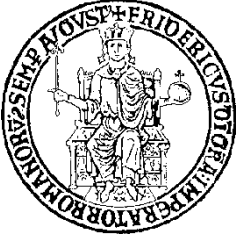
# The Lorentz gas



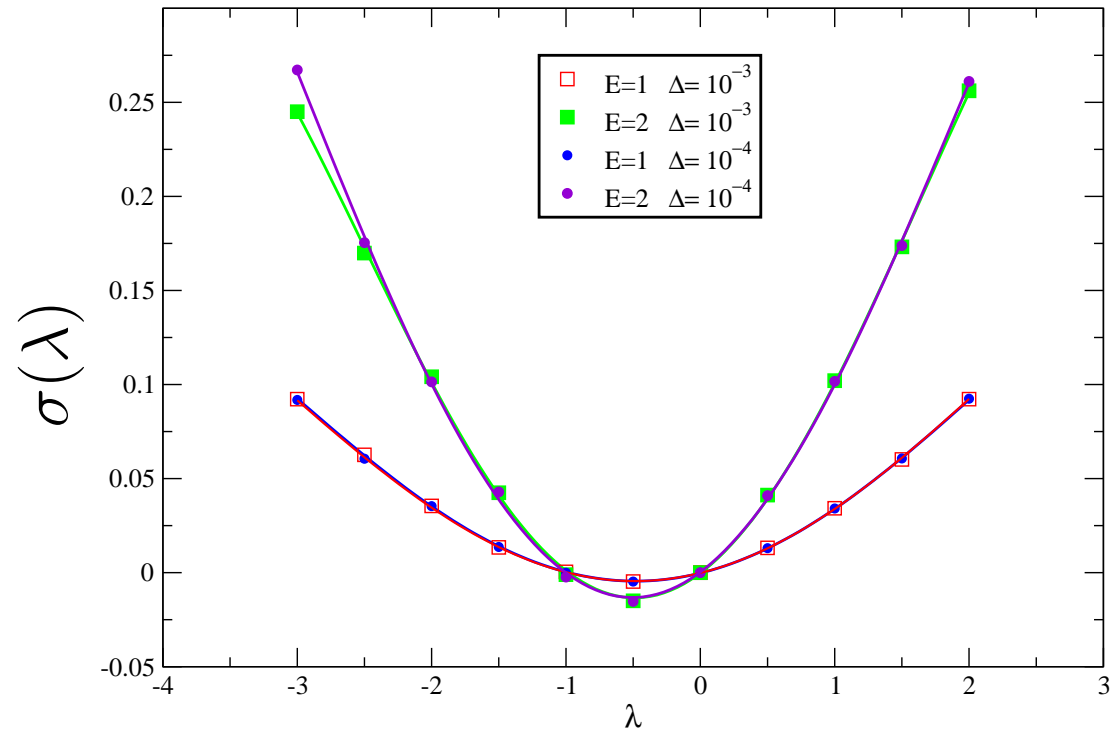
$$\ddot{x}_i = -E_i + \gamma(t)\dot{x}_i, \quad i = 1, 2$$

$$\gamma(t) = \sum_i E_i \dot{x}_i$$

$$\Lambda(\lambda) = \ln \left\langle \exp \left( \int_0^t dt' \gamma(t') \right) \right\rangle$$

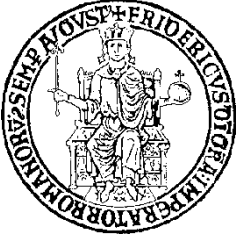


# The Gallavotti-Cohen relation



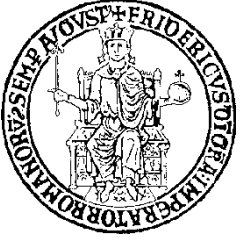
Data for  $\vec{E} = (E, 0)$ ,  $E = 1, 2$  and noise intensity  $\Delta = 10^{-3}, 10^{-4}$





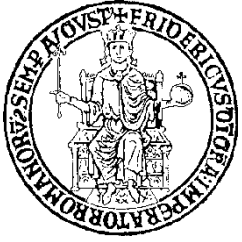
## *Discussion*

- ⑥ Efficient sampling technique (minutes on PC)
- ⑥ So far restricted to steady states
- ⑥ Beyond steady states: sampling of the initial condition (TBD)



# *Perspectives*

- ⑥ “Local” fluctuation relations
- ⑥ Experimental checks: Electrical circuits (Ciliberto et al.)
- ⑥ Energetics of molecular engines



**Thanks**

My collaborators:

