

# Population dynamics method for rare events: systematic errors & feedback control

Esteban Guevara<sup>(1)</sup>, Takahiro Nemoto<sup>(2)</sup>,

Freddy Bouchet<sup>(3)</sup>, Rob L Jack<sup>(4)</sup>, Vivien Lecomte<sup>(5)</sup>

<sup>(1)</sup>IJM, Paris

<sup>(2)</sup>ENS, Paris

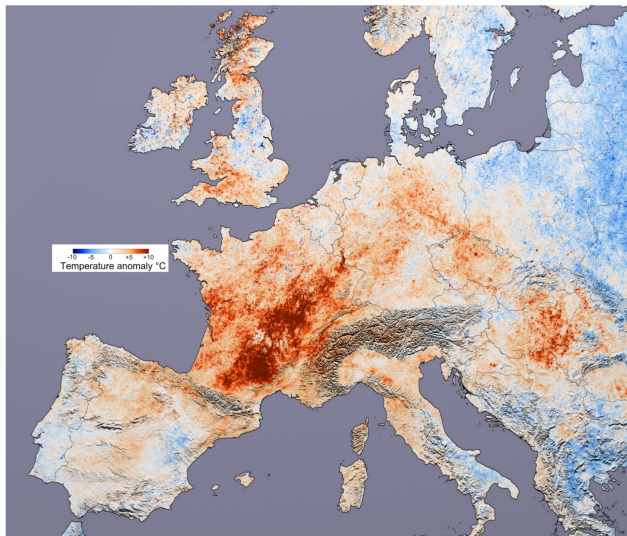
<sup>(3)</sup>ENS, Lyon

<sup>(4)</sup>Bath University & Cambridge University

<sup>(5)</sup>LPMA, Paris & LIPhy, Grenoble

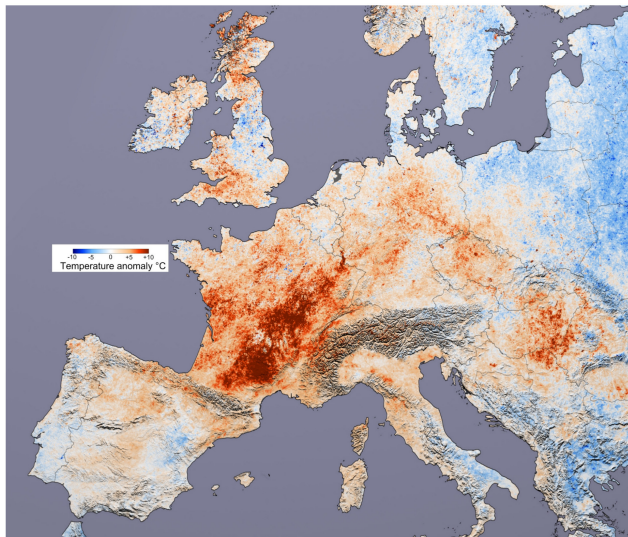
ICTS Bangalore — 5 September 2017

# Why studying rare events?



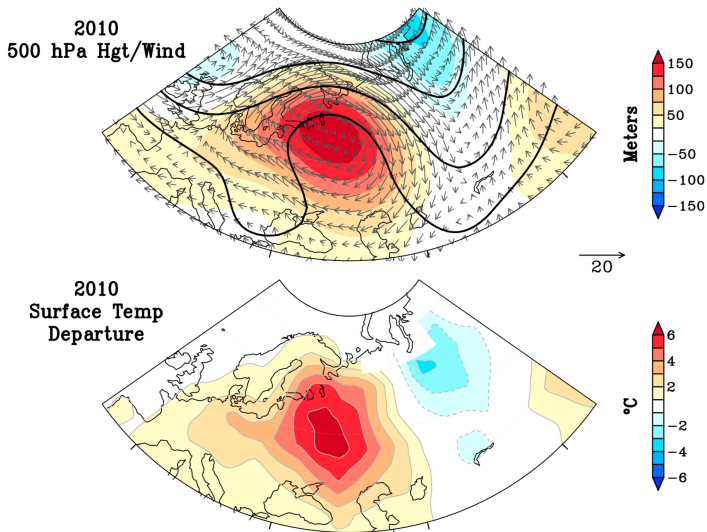
2003 heat wave, Europe [Terra MODIS]

# Why studying rare events?



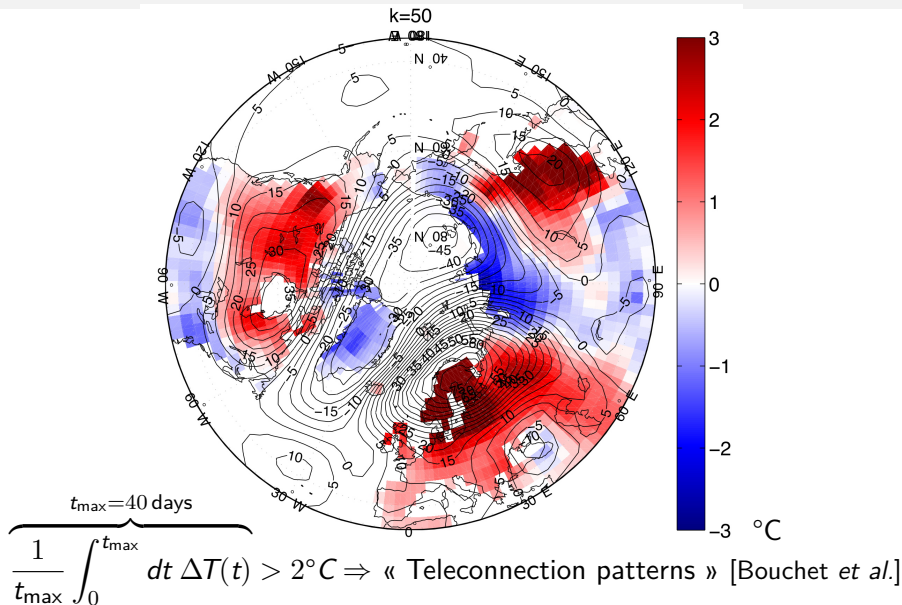
[Anomaly for **1-month** average] 2003 heat wave, Europe [Terra MODIS]

# Why studying rare events?



2010 heat wave in Western Russia [Dole *et al.*, 2011]

# Why studying rare events?



# Why studying rare events?

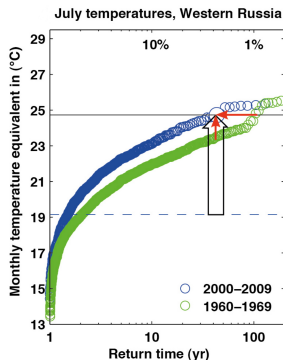
Questions for physicists and mathematicians:

- Probability and **dynamics** of rare events?
- How to **sample** these in numerical modelisation?
- Numerical **tools and methods** to understand their formation?

# Why studying rare events?

Questions for physicists and mathematicians:

- Probability and **dynamics** of rare events?
- How to **sample** these in numerical modelisation?
- Numerical **tools and methods** to understand their formation?



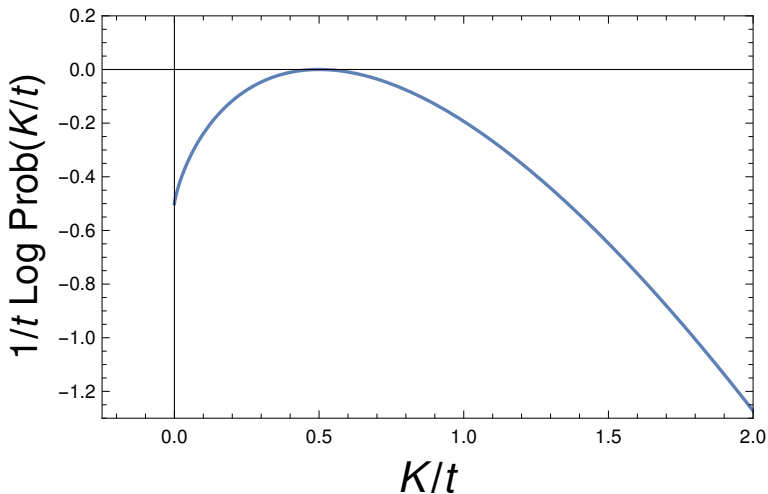
Evolution of the return time  
of the monthly averaged temperature

$$\frac{1}{t_{\max}} \int_0^{t_{\max}} dt T(t)$$

↔ anthropogenic impact on climate?

[Otto *et al.*, 2012]

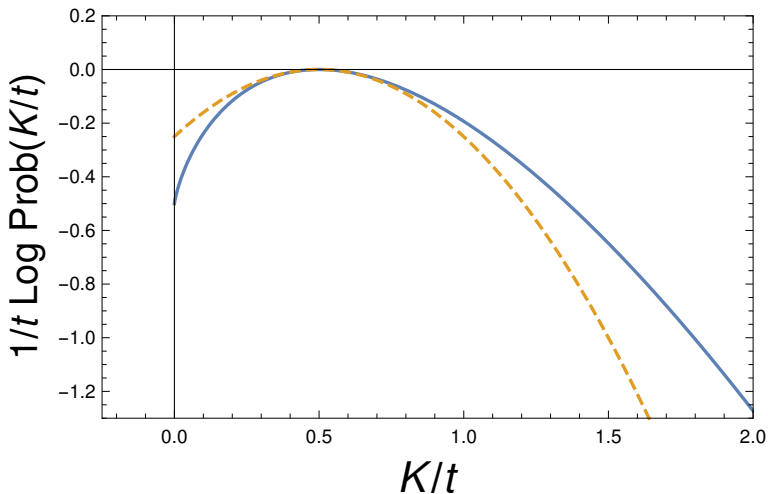
# Distribution of a time-extensive observable $K$ on $[0, t]$



$$\text{Prob}[K, t] \sim e^{t\varphi(K/t)}$$



# Distribution of a time-extensive observable $K$ on $[0, t]$



$$\text{Prob}[K, t] \sim e^{t\varphi(K/t)}$$

# s-modified dynamics

- Markov processes:

Configs.  $\mathcal{C}$ , jump rates  $W(\mathcal{C} \rightarrow \mathcal{C}')$

$$\partial_t P(\mathcal{C}, t) = \sum_{\mathcal{C}'} \left\{ \underbrace{W(\mathcal{C}' \rightarrow \mathcal{C}) P(\mathcal{C}', t)}_{\text{gain term}} - \underbrace{W(\mathcal{C} \rightarrow \mathcal{C}') P(\mathcal{C}, t)}_{\text{loss term}} \right\}$$

## s-modified dynamics

$$K = \text{activity} = \# \text{events}$$

- Markov processes:

Configs.  $\mathcal{C}$ , jump rates  $W(\mathcal{C} \rightarrow \mathcal{C}')$

$$\partial_t P(\mathcal{C}, t) = \sum_{\mathcal{C}'} \left\{ \underbrace{W(\mathcal{C}' \rightarrow \mathcal{C}) P(\mathcal{C}', t)}_{\text{gain term}} - \underbrace{W(\mathcal{C} \rightarrow \mathcal{C}') P(\mathcal{C}, t)}_{\text{loss term}} \right\}$$

- More detailed dynamics for  $P(\mathcal{C}, K, t)$ :

$$\partial_t P(\mathcal{C}, K, t) = \sum_{\mathcal{C}'} \left\{ W(\mathcal{C}' \rightarrow \mathcal{C}) P(\mathcal{C}', K-1, t) - W(\mathcal{C} \rightarrow \mathcal{C}') P(\mathcal{C}, K, t) \right\}$$

# $s$ -modified dynamics

$$K = \text{activity} = \# \text{events}$$

- Markov processes:

Configs.  $\mathcal{C}$ , jump rates  $W(\mathcal{C} \rightarrow \mathcal{C}')$

$$\partial_t P(\mathcal{C}, t) = \sum_{\mathcal{C}'} \left\{ \underbrace{W(\mathcal{C}' \rightarrow \mathcal{C}) P(\mathcal{C}', t)}_{\text{gain term}} - \underbrace{W(\mathcal{C} \rightarrow \mathcal{C}') P(\mathcal{C}, t)}_{\text{loss term}} \right\}$$

- More detailed dynamics for  $P(\mathcal{C}, K, t)$ :

$$\partial_t P(\mathcal{C}, K, t) = \sum_{\mathcal{C}'} \left\{ W(\mathcal{C}' \rightarrow \mathcal{C}) P(\mathcal{C}', K-1, t) - W(\mathcal{C} \rightarrow \mathcal{C}') P(\mathcal{C}, K, t) \right\}$$

- Canonical description:  $s$  conjugated to  $K$

$$\hat{P}(\mathcal{C}, s, t) = \sum_K e^{-sK} P(\mathcal{C}, K, t)$$

# s-modified dynamics

$$K = \text{activity} = \# \text{events}$$

- Markov processes:

Configs.  $\mathcal{C}$ , jump rates  $W(\mathcal{C} \rightarrow \mathcal{C}')$

$$\partial_t P(\mathcal{C}, t) = \sum_{\mathcal{C}'} \left\{ \underbrace{W(\mathcal{C}' \rightarrow \mathcal{C}) P(\mathcal{C}', t)}_{\text{gain term}} - \underbrace{W(\mathcal{C} \rightarrow \mathcal{C}') P(\mathcal{C}, t)}_{\text{loss term}} \right\}$$

- More detailed dynamics for  $P(\mathcal{C}, K, t)$ :

$$\partial_t P(\mathcal{C}, K, t) = \sum_{\mathcal{C}'} \left\{ W(\mathcal{C}' \rightarrow \mathcal{C}) P(\mathcal{C}', K-1, t) - W(\mathcal{C} \rightarrow \mathcal{C}') P(\mathcal{C}, K, t) \right\}$$

- Canonical description:  $s$  conjugated to  $K$

$$\hat{P}(\mathcal{C}, s, t) = \sum_K e^{-sK} P(\mathcal{C}, K, t)$$

- $s$ -modified dynamics [probability non-conserving]

$$\partial_t \hat{P}(\mathcal{C}, s, t) = \sum_{\mathcal{C}'} \left\{ e^{-s} W(\mathcal{C}' \rightarrow \mathcal{C}) \hat{P}(\mathcal{C}', s, t) - W(\mathcal{C} \rightarrow \mathcal{C}') \hat{P}(\mathcal{C}, s, t) \right\}$$

# s-modified dynamics

$$K = k_{c_0 c_1} + k_{c_1 c_2} + \dots$$

- Markov processes:

Configs.  $\mathcal{C}$ , jump rates  $W(\mathcal{C} \rightarrow \mathcal{C}')$

$$\partial_t P(\mathcal{C}, t) = \sum_{\mathcal{C}'} \left\{ \underbrace{W(\mathcal{C}' \rightarrow \mathcal{C}) P(\mathcal{C}', t)}_{\text{gain term}} - \underbrace{W(\mathcal{C} \rightarrow \mathcal{C}') P(\mathcal{C}, t)}_{\text{loss term}} \right\}$$

- More detailed dynamics for  $P(\mathcal{C}, K, t)$ :

$$\partial_t P(\mathcal{C}, K, t) = \sum_{\mathcal{C}'} \left\{ W(\mathcal{C}' \rightarrow \mathcal{C}) P(\mathcal{C}', K - k_{\mathcal{C}' \mathcal{C}}, t) - W(\mathcal{C} \rightarrow \mathcal{C}') P(\mathcal{C}, K, t) \right\}$$

- Canonical description:  $s$  conjugated to  $K$

$$\hat{P}(\mathcal{C}, s, t) = \sum_K e^{-sK} P(\mathcal{C}, K, t)$$

- $s$ -modified dynamics [probability non-conserving]

$$\partial_t \hat{P}(\mathcal{C}, s, t) = \sum_{\mathcal{C}'} \left\{ e^{-s k_{\mathcal{C}' \mathcal{C}}} W(\mathcal{C}' \rightarrow \mathcal{C}) \hat{P}(\mathcal{C}', s, t) - W(\mathcal{C} \rightarrow \mathcal{C}') \hat{P}(\mathcal{C}, s, t) \right\}$$

# Numerical method [JB Anderson; D Aldous; P Grassberger; P Del Moral; ...]

## Evaluation of large deviation functions

[à la “Diffusion Monte-Carlo”]

$$\sum_{\mathcal{C}} \hat{P}(\mathcal{C}, \mathbf{s}, t) = \langle e^{-\mathbf{s} \cdot \mathbf{K}} \rangle \sim e^{t \psi(\mathbf{s})}$$

- discrete time: Giardinà, Kurchan, Peliti [PRL **96**, 120603 (2006)]
- continuous time: VL, Tailleur [JSTAT P03004 (2007)]

## Cloning dynamics

$$\partial_t \hat{P}(\mathcal{C}, \mathbf{s}) = \underbrace{\sum_{\mathcal{C}'} W_{\mathbf{s}}(\mathcal{C}' \rightarrow \mathcal{C}) \hat{P}(\mathcal{C}', \mathbf{s}) - r_{\mathbf{s}}(\mathcal{C}) \hat{P}(\mathcal{C}, \mathbf{s})}_{\text{modified dynamics}} + \underbrace{\delta r_{\mathbf{s}}(\mathcal{C}) \hat{P}(\mathcal{C}, \mathbf{s})}_{\text{cloning term}}$$

- $W_{\mathbf{s}}(\mathcal{C}' \rightarrow \mathcal{C}) = e^{-\mathbf{s} \cdot \mathbf{W}} W(\mathcal{C}' \rightarrow \mathcal{C})$
- $r_{\mathbf{s}}(\mathcal{C}) = \sum_{\mathcal{C}'} W_{\mathbf{s}}(\mathcal{C} \rightarrow \mathcal{C}')$   $r(\mathcal{C}) = \sum_{\mathcal{C}'} W(\mathcal{C} \rightarrow \mathcal{C}')$
- $\delta r_{\mathbf{s}}(\mathcal{C}) = r_{\mathbf{s}}(\mathcal{C}) - r(\mathcal{C})$

# Numerical method [JB Anderson; D Aldous; P Grassberger; P Del Moral; ...]

## Evaluation of large deviation functions

[à la “Diffusion Monte-Carlo”]

$$\sum_{\mathcal{C}} \hat{P}(\mathcal{C}, \mathbf{s}, t) = \langle e^{-\mathbf{s} \cdot \mathbf{K}} \rangle \sim e^{t \psi(\mathbf{s})}$$

- discrete time: Giardinà, Kurchan, Peliti [PRL **96**, 120603 (2006)]
- continuous time: VL, Tailleur [JSTAT P03004 (2007)]

## Cloning dynamics

$$\partial_t \hat{P}(\mathcal{C}, \mathbf{s}) = \underbrace{\sum_{\mathcal{C}'} W_{\mathbf{s}}(\mathcal{C}' \rightarrow \mathcal{C}) \hat{P}(\mathcal{C}', \mathbf{s}) - r_{\mathbf{s}}(\mathcal{C}) \hat{P}(\mathcal{C}, \mathbf{s})}_{\text{modified dynamics}} + \underbrace{\delta r_{\mathbf{s}}(\mathcal{C}) \hat{P}(\mathcal{C}, \mathbf{s})}_{\text{cloning term}}$$

- $W_{\mathbf{s}}(\mathcal{C}' \rightarrow \mathcal{C}) = e^{-\mathbf{s} \cdot \mathbf{W}}(\mathcal{C}' \rightarrow \mathcal{C})$
- $r_{\mathbf{s}}(\mathcal{C}) = \sum_{\mathcal{C}'} W_{\mathbf{s}}(\mathcal{C} \rightarrow \mathcal{C}')$   $r(\mathcal{C}) = \sum_{\mathcal{C}'} W(\mathcal{C} \rightarrow \mathcal{C}')$
- $\delta r_{\mathbf{s}}(\mathcal{C}) = r_{\mathbf{s}}(\mathcal{C}) - r(\mathcal{C})$



# Explicit construction

$$\partial_t \hat{P}(\mathcal{C}, s) = \underbrace{\sum_{\mathcal{C}'} W_s(\mathcal{C}' \rightarrow \mathcal{C}) \hat{P}(\mathcal{C}', s) - r_s(\mathcal{C}) \hat{P}(\mathcal{C}, s)}_{\text{modified dynamics}} + \underbrace{\delta r_s(\mathcal{C}) \hat{P}(\mathcal{C}, s)}_{\text{cloning term}}$$

How to take into account loss/gain of probability?

- handle a large number of copies of the system
- implement a **selection** rule: on a time interval  $\Delta t$  a copy in config  $\mathcal{C}$  is replaced by  $e^{\Delta t \delta r_s(\mathcal{C})}$  copies
- $\psi(s)$  = the rate of exponential growth/decay of the total population
- optionally: keep population constant by non-biased pruning/cloning

# Explicit construction

$$\partial_t \hat{P}(\mathcal{C}, s) = \underbrace{\sum_{\mathcal{C}'} W_s(\mathcal{C}' \rightarrow \mathcal{C}) \hat{P}(\mathcal{C}', s) - r_s(\mathcal{C}) \hat{P}(\mathcal{C}, s)}_{\text{modified dynamics}} + \underbrace{\delta r_s(\mathcal{C}) \hat{P}(\mathcal{C}, s)}_{\text{cloning term}}$$

How to take into account loss/gain of probability?

- handle a large number of copies of the system
- implement a **selection** rule: on a time interval  $\Delta t$  a copy in config  $\mathcal{C}$  is replaced by  $\lfloor e^{\Delta t \delta r_s(\mathcal{C})} + \epsilon \rfloor$  copies,  $\epsilon \sim [0, 1]$
- $\psi(s)$  = the rate of exponential growth/decay of the total population
- optionally: keep population constant by non-biased pruning/cloning

# Explicit construction

$$\partial_t \hat{P}(\mathcal{C}, s) = \underbrace{\sum_{\mathcal{C}'} W_s(\mathcal{C}' \rightarrow \mathcal{C}) \hat{P}(\mathcal{C}', s) - r_s(\mathcal{C}) \hat{P}(\mathcal{C}, s)}_{\text{modified dynamics}} + \underbrace{\delta r_s(\mathcal{C}) \hat{P}(\mathcal{C}, s)}_{\text{cloning term}}$$

How to take into account loss/gain of probability?

- handle a large number of copies of the system
- implement a **selection** rule: on a time interval  $\Delta t$  a copy in config  $\mathcal{C}$  is replaced by  $\lfloor e^{\Delta t \delta r_s(\mathcal{C})} + \epsilon \rfloor$  copies,  $\epsilon \sim [0, 1]$
- $\psi(s)$  = the rate of exponential growth/decay of the total population
- optionally: keep population constant by non-biased pruning/cloning

# Explicit construction

$$\partial_t \hat{P}(\mathcal{C}, s) = \underbrace{\sum_{\mathcal{C}'} W_s(\mathcal{C}' \rightarrow \mathcal{C}) \hat{P}(\mathcal{C}', s) - r_s(\mathcal{C}) \hat{P}(\mathcal{C}, s)}_{\text{modified dynamics}} + \underbrace{\delta r_s(\mathcal{C}) \hat{P}(\mathcal{C}, s)}_{\text{cloning term}}$$

How to take into account loss/gain of probability?

- handle a large number of copies of the system
- implement a **selection** rule: on a time interval  $\Delta t$  a copy in config  $\mathcal{C}$  is replaced by  $\lfloor e^{\Delta t \delta r_s(\mathcal{C})} + \epsilon \rfloor$  copies,  $\epsilon \sim [0, 1]$
- $\psi(s)$  = the rate of exponential growth/decay of the total population
- optionally: keep population constant by non-biased pruning/cloning

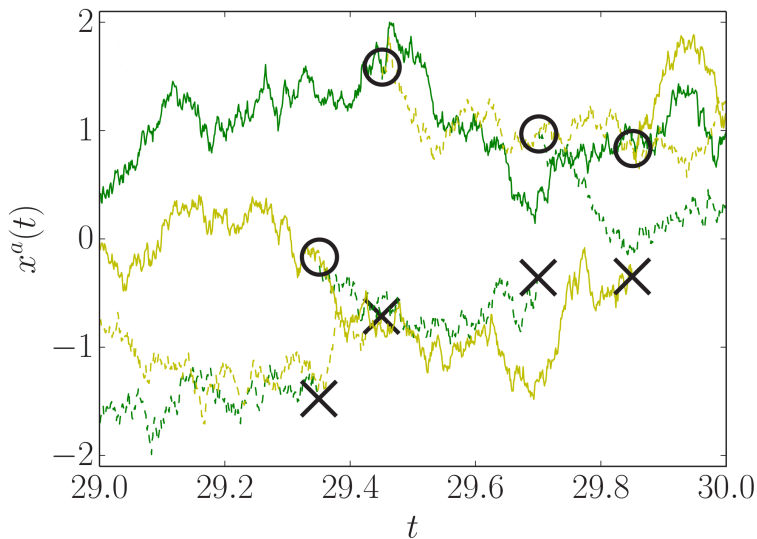
# Explicit construction

$$\partial_t \hat{P}(\mathcal{C}, s) = \underbrace{\sum_{\mathcal{C}'} W_s(\mathcal{C}' \rightarrow \mathcal{C}) \hat{P}(\mathcal{C}', s) - r_s(\mathcal{C}) \hat{P}(\mathcal{C}, s)}_{\text{modified dynamics}} + \underbrace{\delta r_s(\mathcal{C}) \hat{P}(\mathcal{C}, s)}_{\text{cloning term}}$$

## Biological interpretation

- copy in configuration  $\mathcal{C} \equiv$  organism of **genome**  $\mathcal{C}$
- dynamics of rates  $W_s \equiv$  **mutations**
- cloning at rates  $\delta r_s \equiv$  **selection** rendering **typical** the rare histories

An example: 4 copies, 1 degree of freedom  $\mathcal{C} = x \in \mathbb{R}$



# How to perform averages? (i)

[spectral analysis]

- ★ Final-time distribution: *proportion* of copies in  $\mathcal{C}$  at  $t$

$$\langle N_{\text{nc}}(t) \rangle_s$$

$$\langle N_{\text{nc}}(\mathcal{C}, t) \rangle_s$$

$$p_{\text{end}}(\mathcal{C}, t) = \frac{\langle N_{\text{nc}}(\mathcal{C}, t) \rangle_s}{\langle N_{\text{nc}}(t) \rangle_s}$$

[ $N_{\text{nc}}$  = number in non-constant population dynamics]

## How to perform averages? (i)

[spectral analysis]

$$\partial_t |\hat{P}\rangle = \mathbb{W}_s |\hat{P}\rangle$$

★ Final-time distribution: *proportion* of copies in  $\mathcal{C}$  at  $t$

$$\langle N_{\text{nc}}(t) \rangle_s$$

$$\langle N_{\text{nc}}(\mathcal{C}, t) \rangle_s$$

$$p_{\text{end}}(\mathcal{C}, t) = \frac{\langle N_{\text{nc}}(\mathcal{C}, t) \rangle_s}{\langle N_{\text{nc}}(t) \rangle_s}$$

[ $N_{\text{nc}}$  = number in non-constant population dynamics]



## How to perform averages? (i)

[spectral analysis]

$$\partial_t |\hat{P}\rangle = \mathbb{W}_s |\hat{P}\rangle$$

$$\mathbb{W}_s |R\rangle = \psi(s) |R\rangle$$

$$\langle L | \mathbb{W}_s = \psi(s) \langle L |$$

$$[ \quad \langle L | = \langle - | \quad @ \textcolor{red}{s} = 0 \quad ]$$

★ Final-time distribution: *proportion* of copies in  $\mathcal{C}$  at  $t$

$$\langle N_{\text{nc}}(t) \rangle_s$$

$$\langle N_{\text{nc}}(\mathcal{C}, t) \rangle_s$$

$$p_{\text{end}}(\mathcal{C}, t) = \frac{\langle N_{\text{nc}}(\mathcal{C}, t) \rangle_s}{\langle N_{\text{nc}}(t) \rangle_s}$$

[ $N_{\text{nc}}$  = number in non-constant population dynamics]

## How to perform averages? (i)

## [spectral analysis]

$$\partial_t |\hat{P}\rangle = \mathbb{W}_s |\hat{P}\rangle$$

$$\mathbb{W}_s |R\rangle = \psi(s) |R\rangle$$

$$e^{t\mathbb{W}_s} \underset{t \rightarrow \infty}{\sim} e^{t\psi(s)} |R\rangle \langle L|$$

$$\langle L | \mathbb{W}_s = \psi(s) \langle L |$$

$$[ \quad \langle L | = \langle - | \quad @ \textcolor{red}{s} = 0 \quad ]$$

★ Final-time distribution: *proportion* of copies in  $\mathcal{C}$  at  $t$

$$\langle N_{\text{nc}}(t) \rangle_s$$

$$\langle N_{\text{nc}}(\mathcal{C}, t) \rangle_s$$

$$p_{\text{end}}(\mathcal{C}, t) = \frac{\langle N_{\text{nc}}(\mathcal{C}, t) \rangle_s}{\langle N_{\text{nc}}(t) \rangle_s}$$

[ $N_{\text{nc}}$  = number in non-constant population dynamics]

## How to perform averages? (i)

## [spectral analysis]

$$\partial_t |\hat{P}\rangle = \mathbb{W}_s |\hat{P}\rangle$$

$$\mathbb{W}_s |R\rangle = \psi(s) |R\rangle$$

$$e^{t\mathbb{W}_s} \underset{t \rightarrow \infty}{\sim} e^{t\psi(s)} |R\rangle \langle L|$$

$$\langle L | \mathbb{W}_s = \psi(s) \langle L |$$

$$[ \quad \langle L | = \langle - | \quad @ \textcolor{red}{s} = 0 \quad ]$$

★ Final-time distribution: *proportion* of copies in  $\mathcal{C}$  at  $t$

$$\langle N_{\text{nc}}(t) \rangle_s = \langle - | e^{t\mathbb{W}_s} | P_i \rangle N_0 \underset{t \rightarrow \infty}{\sim} e^{t\psi(s)} \langle L | P_i \rangle N_0$$

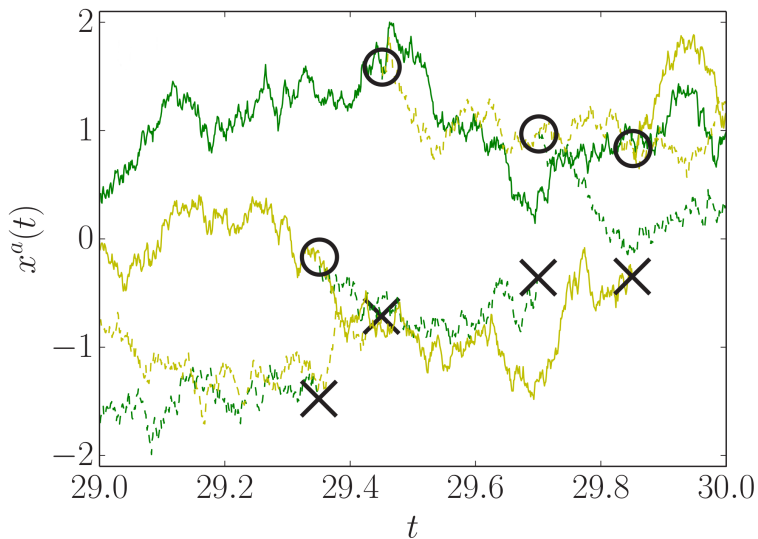
$$\langle N_{\text{nc}}(\mathcal{C}, t) \rangle_s = \langle \mathcal{C} | e^{t\mathbb{W}_s} | P_i \rangle N_0 \underset{t \rightarrow \infty}{\sim} e^{t\psi(s)} \langle \mathcal{C} | R \rangle \langle L | P_i \rangle N_0$$

$$p_{\text{end}}(\mathcal{C}, t) = \frac{\langle N_{\text{nc}}(\mathcal{C}, t) \rangle_s}{\langle N_{\text{nc}}(t) \rangle_s} \underset{t \rightarrow \infty}{\sim} \langle \mathcal{C} | R \rangle \equiv p_{\text{end}}(\mathcal{C})$$

[ $N_{\text{nc}}$  = number in non-constant population dynamics]

Final-time distribution governed by **right** eigenvector.

An example: 4 copies, 1 degree of freedom  $\mathcal{C} = x \in \mathbb{R}$



# How to perform averages? (ii) Intermediate times

$$\partial_t |\hat{P}\rangle = \mathbb{W}_s |\hat{P}\rangle$$

$$\mathbb{W}_s |R\rangle = \psi(s) |R\rangle$$

$$e^{t\mathbb{W}_s} \underset{t \rightarrow \infty}{\sim} e^{t\psi(s)} |R\rangle \langle L|$$

$$\langle L | \mathbb{W}_s = \psi(s) \langle L |$$

$$[ \quad \langle L | = \langle - | \quad @ \textcolor{red}{s} = 0 \quad ]$$

★ Mid-time distribution: *proportion* of copies in  $\mathcal{C}$  at  $t_1 \ll t$

$$\langle N_{\text{nc}}(t) \rangle_s$$

$$\langle N_{\text{nc}}(t | \mathcal{C}, t_1) \rangle_s$$

$$p(t | \mathcal{C}, t_1) = \frac{\langle N_{\text{nc}}(t | \mathcal{C}, t_1) \rangle_s}{\langle N_{\text{nc}}(t) \rangle_s}$$

## How to perform averages? (ii) Intermediate times

$$\partial_t |\hat{P}\rangle = \mathbb{W}_s |\hat{P}\rangle$$

$$\mathbb{W}_s |R\rangle = \psi(s) |R\rangle$$

$$e^{t\mathbb{W}_s} \underset{t \rightarrow \infty}{\sim} e^{t\psi(s)} |R\rangle \langle L|$$

$$\langle L | \mathbb{W}_s = \psi(s) \langle L |$$

$$[ \quad \langle L | = \langle - | \quad @ \textcolor{red}{s} = 0 \quad ]$$

★ Mid-time distribution: *proportion* of copies in  $\mathcal{C}$  at  $t_1 \ll t$

$$\langle N_{\text{nc}}(t) \rangle_s = \langle - | e^{t\mathbb{W}_s} | P_i \rangle N_0 \underset{t \rightarrow \infty}{\sim} e^{t\psi(s)} \langle L | P_i \rangle N_0$$

$$\langle N_{\text{nc}}(t | \mathcal{C}, t_1) \rangle_s = \langle - | e^{(t-t_1)\mathbb{W}_s} | \textcolor{red}{C} \rangle \langle \textcolor{red}{C} | e^{t_1 \mathbb{W}_s} | P_i \rangle N_0 \sim e^{t\psi(s)} \langle L | \mathcal{C} \rangle \langle \mathcal{C} | R \rangle \langle L | P_i \rangle N_0$$

$$p(t | \mathcal{C}, t_1) = \frac{\langle N_{\text{nc}}(t | \mathcal{C}, t_1) \rangle_s}{\langle N_{\text{nc}}(t) \rangle_s} \underset{t \rightarrow \infty}{\sim} \langle L | \mathcal{C} \rangle \langle \mathcal{C} | R \rangle \equiv p_{\text{ave}}(\mathcal{C})$$

## How to perform averages? (ii) Intermediate times

$$\partial_t |\hat{P}\rangle = \mathbb{W}_s |\hat{P}\rangle$$

$$\mathbb{W}_s |R\rangle = \psi(s) |R\rangle$$

$$e^{t\mathbb{W}_s} \underset{t \rightarrow \infty}{\sim} e^{t\psi(s)} |R\rangle \langle L|$$

$$\langle L | \mathbb{W}_s = \psi(s) \langle L |$$

$$[ \quad \langle L | = \langle - | \quad @ \textcolor{red}{s} = 0 \quad ]$$

★ Mid-time distribution: *proportion* of copies in  $\mathcal{C}$  at  $t_1 \ll t$

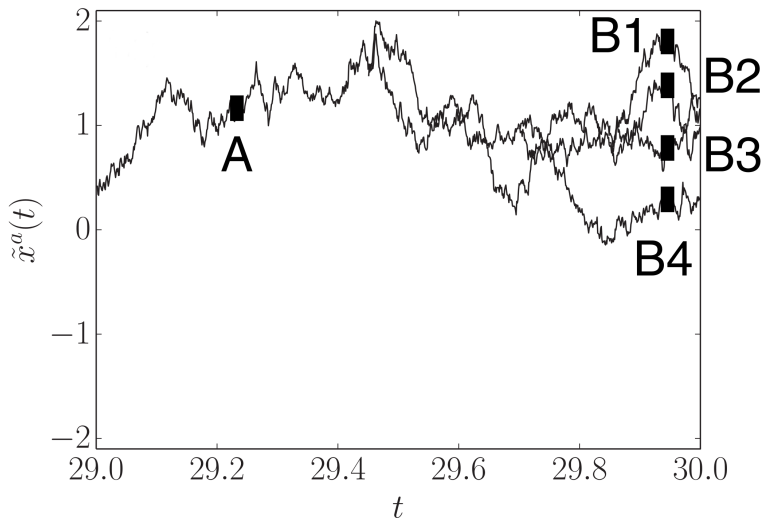
$$\langle N_{\text{nc}}(t) \rangle_s = \langle - | e^{t\mathbb{W}_s} | P_i \rangle N_0 \underset{t \rightarrow \infty}{\sim} e^{t\psi(s)} \langle L | P_i \rangle N_0$$

$$\langle N_{\text{nc}}(t | \mathcal{C}, t_1) \rangle_s = \langle - | e^{(t-t_1)\mathbb{W}_s} | \textcolor{violet}{\mathcal{C}} \rangle \langle \textcolor{violet}{\mathcal{C}} | e^{t_1 \mathbb{W}_s} | P_i \rangle N_0 \sim e^{t\psi(s)} \langle L | \mathcal{C} \rangle \langle \mathcal{C} | R \rangle \langle L | P_i \rangle N_0$$

$$p(t | \mathcal{C}, t_1) = \frac{\langle N_{\text{nc}}(t | \mathcal{C}, t_1) \rangle_s}{\langle N_{\text{nc}}(t) \rangle_s} \underset{t \rightarrow \infty}{\sim} \langle L | \mathcal{C} \rangle \langle \mathcal{C} | R \rangle \equiv p_{\text{ave}}(\mathcal{C})$$

Mid-time distribution governed by **left** and **right** eigenvectors.

An example: 4 copies, 1 degree of freedom  $\mathcal{C} = x \in \mathbb{R}$



Huge sampling issue



# How to perform averages?

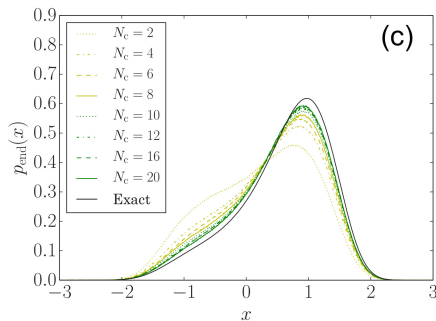
## ★ Mid-time ancestor distribution:

fraction of copies (at time  $t_1$ ) which were in configuration  $\mathcal{C}$ , knowing that there are in configuration  $\mathcal{C}_f$  at final time  $t_f$ :

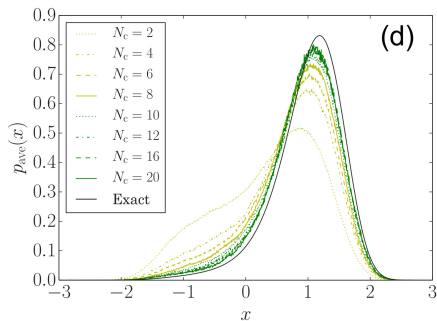
$$p_{\text{anc}}(\mathcal{C}, t_1; \mathcal{C}_f, t_f) = \frac{\langle N_{\text{nc}}(\mathcal{C}_f, t_f | \mathcal{C}, t_1) \rangle_s}{\sum_{\mathcal{C}'} \langle N_{\text{nc}}(\mathcal{C}_f, t_f | \mathcal{C}', t_1) \rangle_s} \underset{t_{f,1} \rightarrow \infty}{\sim} \langle L | \mathcal{C} \rangle \langle \mathcal{C} | R \rangle = p_{\text{ave}}(\mathcal{C})$$

**The “ancestor statistics” of a configuration  $\mathcal{C}_f$  is thus independent (far enough in the past) of the configuration  $\mathcal{C}_f$ .**

# Example distributions for a simple Langevin dynamics

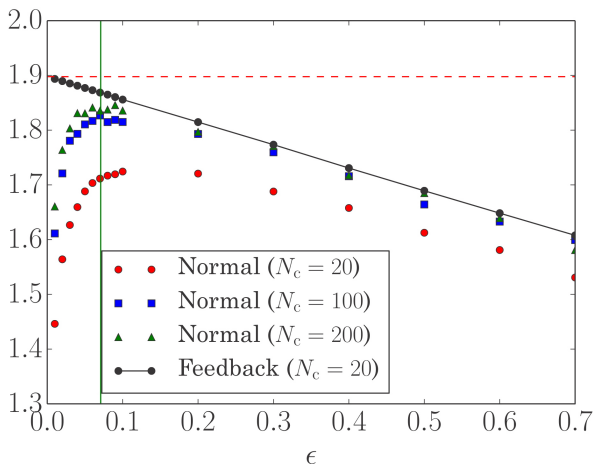


final-time:  $p_{\text{end}}(x)$



intermediate-time:  $p_{\text{ave}}(x)$

# The small-noise crisis: systematic errors grow as $\epsilon \rightarrow 0$



Cause: as  $\epsilon \rightarrow 0$ ,  $p_{\text{ave}}(x)$  &  $p_{\text{end}}(x) \rightarrow$  sharply peaked at *different points*  
*i.e.* the clones do not ~~attack~~ sample correctly the phase space

# How to make mid- and final-time distribution closer?

Driven/auxiliary dynamics: [Maes, Jack&Sollich, Touchette&Chetrite]

- Probability preserving
- No mismatch between  $p_{\text{ave}}$  and  $p_{\text{end}}$
- Constructed as

$$\mathbb{W}_s^{\text{aux}} = L\mathbb{W}_sL^{-1} - \psi(s)\mathbf{1}$$

# How to make mid- and final-time distribution closer?

Driven/auxiliary dynamics: [Maes, Jack&Sollich, Touchette&Chetrite]

- Probability preserving
- No mismatch between  $p_{\text{ave}}$  and  $p_{\text{end}}$
- Constructed as

$$\mathbb{W}_s^{\text{aux}} = L\mathbb{W}_sL^{-1} - \psi(s)\mathbf{1}$$

# How to make mid- and final-time distribution closer?

Driven/auxiliary dynamics: [Maes, Jack&Sollich, Touchette&Chetrite]

- Probability preserving
- No mismatch between  $p_{\text{ave}}$  and  $p_{\text{end}}$
- Constructed as

$$\mathbb{W}_s^{\text{aux}} = L\mathbb{W}_sL^{-1} - \psi(s)\mathbf{1}$$

- Issue: determining  $L$  is difficult
- Solution: evaluate  $L$  as  $L_{\text{test}}$  on the fly and simulate

$$\mathbb{W}_s^{\text{test}} = L_{\text{test}}\mathbb{W}_sL_{\text{test}}^{-1}$$

- **Iterate.** [For any  $L_{\text{test}}$ , the simulation is in principle correct.]

# How to make mid- and final-time distribution closer?

Driven/auxiliary dynamics: [Maes, Jack&Sollich, Touchette&Chetrite]

- Probability preserving
- No mismatch between  $p_{\text{ave}}$  and  $p_{\text{end}}$
- Constructed as

$$\mathbb{W}_s^{\text{aux}} = L\mathbb{W}_sL^{-1} - \psi(s)\mathbf{1}$$

- Issue: determining  $L$  is difficult
- Solution: evaluate  $L$  as  $L_{\text{test}}$  on the fly and simulate

$$\mathbb{W}_s^{\text{test}} = L_{\text{test}}\mathbb{W}_sL_{\text{test}}^{-1}$$

- **Iterate.** [For any  $L_{\text{test}}$ , the simulation is in principle correct.]

# How to make mid- and final-time distribution closer?

Driven/auxiliary dynamics: [Maes, Jack&Sollich, Touchette&Chetrite]

- Probability preserving
- No mismatch between  $p_{\text{ave}}$  and  $p_{\text{end}}$
- Constructed as

$$\mathbb{W}_s^{\text{aux}} = L\mathbb{W}_sL^{-1} - \psi(s)\mathbf{1}$$

- Issue: determining  $L$  is difficult
- Solution: evaluate  $L$  as  $L_{\text{test}}$  on the fly and simulate

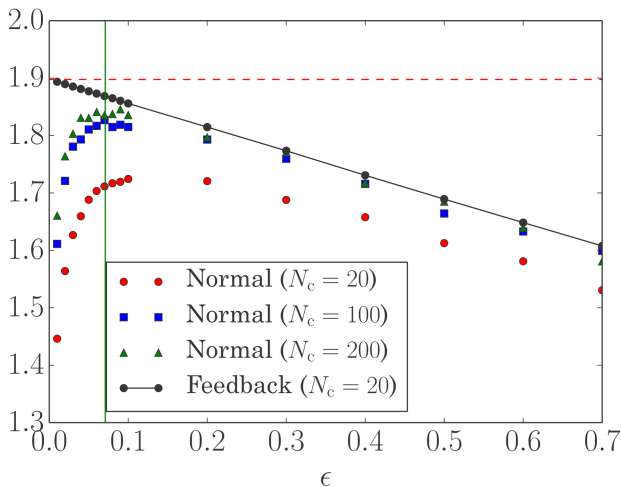
$$\mathbb{W}_s^{\text{test}} = L_{\text{test}}\mathbb{W}_sL_{\text{test}}^{-1}$$

- **Iterate.** [For any  $L_{\text{test}}$ , the simulation is in principle correct.]

Similar in spirit to **multi-canonical** (e.g. Wang-Landau) approaches in static thermodynamics. [Here, one flattens the left-eigenvector of  $\mathbb{W}_s^{\text{test}}$ .]

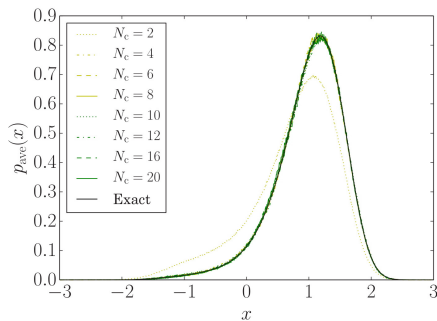
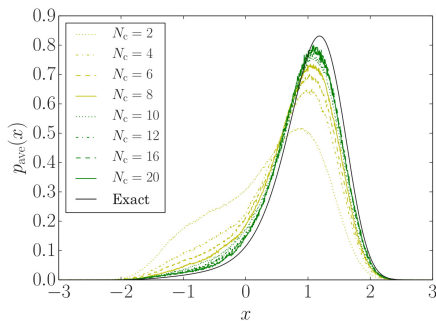


# Improvement of the small-noise crisis (i.i)



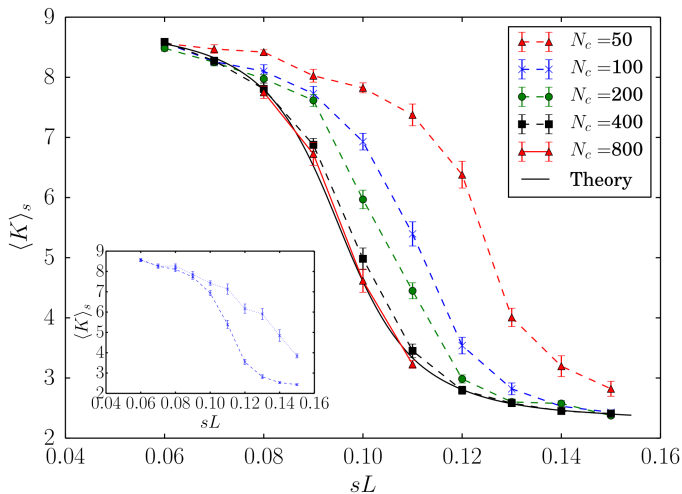
Physical insight: probability loss transformed into *effective forces*.

## Improvement of the small-noise crisis (i.ii)



Much more efficient evaluation of the biased distribution.  
Even for a very crude (polynomial) approximation of the effective force.

# Improvement of the small-noise crisis (ii)



Interacting system in 1D.

Effective force: 1-, 2-, 3- body interactions only [also crude approx.].

# Summary and open questions (1)

## Multicanonical approach

[with F Bouchet, R Jack, T Nemoto]

- Sampling problem (depletion of ancestors)
- On-the-fly evaluated auxiliary dynamics
- Solution to the small-noise crisis
- Systems with large number of degrees of freedom

# Summary and open questions (1)

## Multicanonical approach

[with F Bouchet, R Jack, T Nemoto]

- Sampling problem (depletion of ancestors)
- On-the-fly evaluated auxiliary dynamics
- Solution to the small-noise crisis
- Systems with large number of degrees of freedom

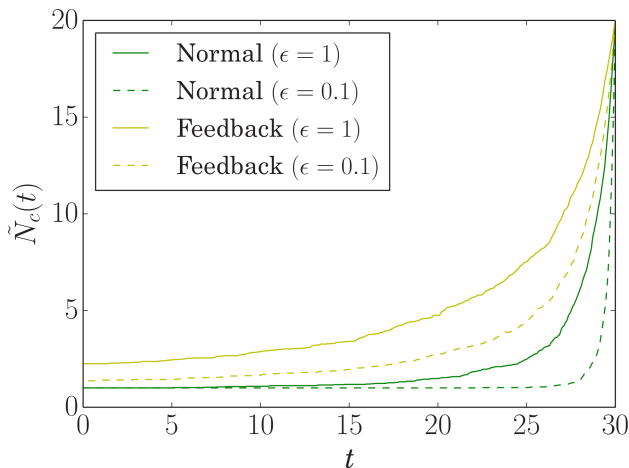
## Finite-population effects

[with E Guevara, T Nemoto]

- Quantitative finite- $N_{\text{clones}}$  scaling  $\rightarrow$  interpolation method
- Initial transient regime due to small population
- Analogy with biology: many small islands vs. few large islands?
- Question: effective forces  $\leftarrow$  selection?

## Open question (2): why is it working?

Improvement of the depletion-of-ancestors problem:



Dashed line: lower noise

Continuous line: higher noise

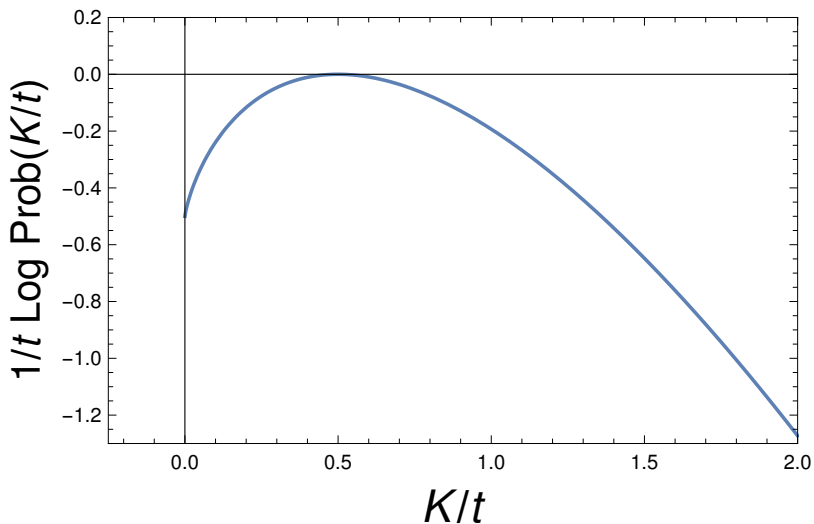
# Thanks for your attention!

## References:

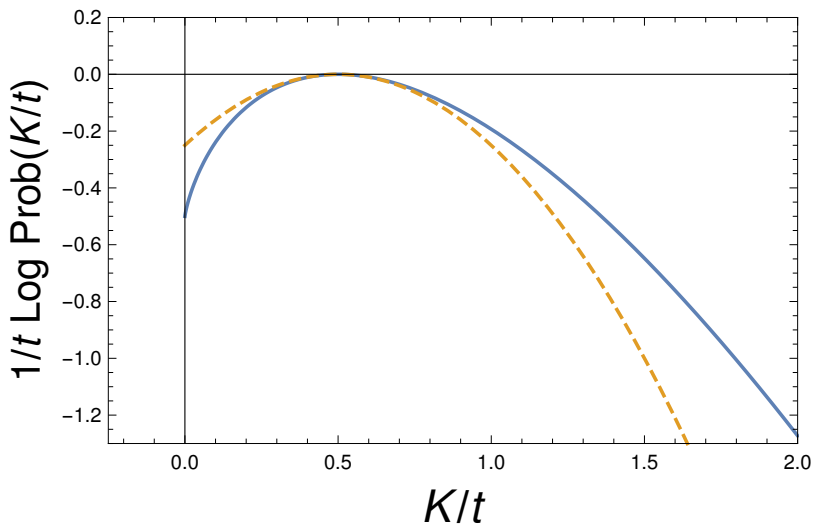
- ★ *Population dynamics method with a multi-canonical feedback control*  
Takahiro Nemoto, Freddy Bouchet, Robert L. Jack and Vivien Lecomte  
PRE **93** 062123 (2016)
- ★ *Finite-time and finite-size scalings in the evaluation of large deviation functions: Analytical study using a birth-death process*  
Takahiro Nemoto, Esteban Guevara Hidalgo and Vivien Lecomte  
PRE **95** 012102 (2017)
- ★ *Finite-size scaling of a first-order dynamical phase transition: adaptive population dynamics and effective model*  
Takahiro Nemoto, Robert L. Jack and Vivien Lecomte  
PRL **118** 115702 (2017)

# Supplementary material

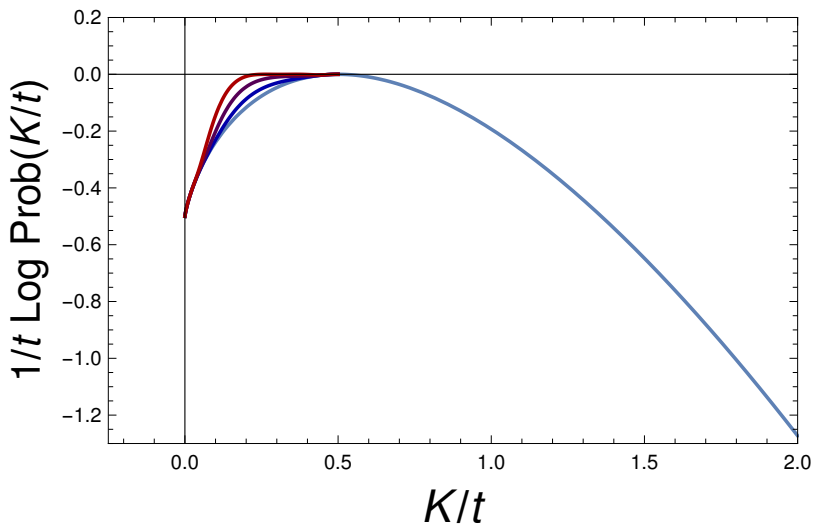




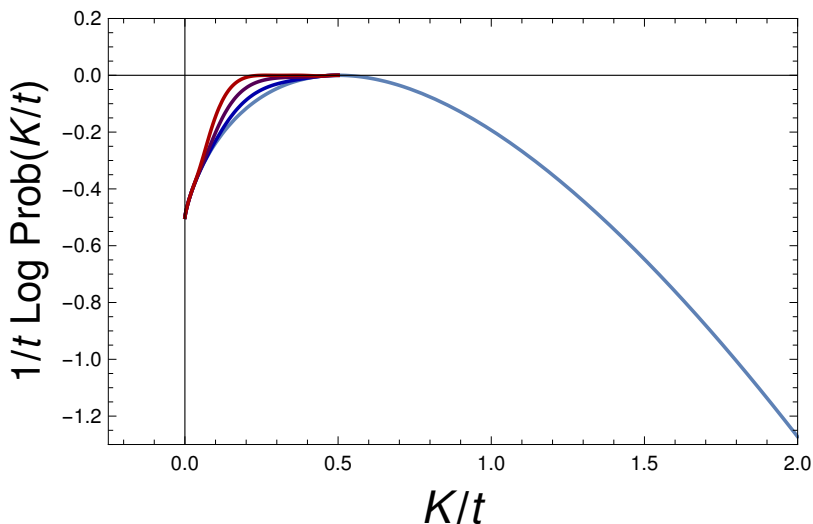
$$\text{Prob}[K] \sim e^{t\varphi(K/t)}$$



$$\text{Prob}[K] \sim e^{t\varphi(K/t)}$$

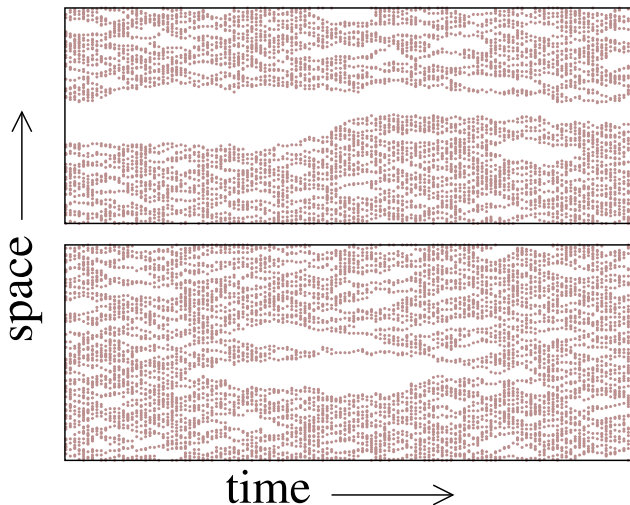


$$\text{Prob}[K] \sim e^{t\varphi(K/t)}$$

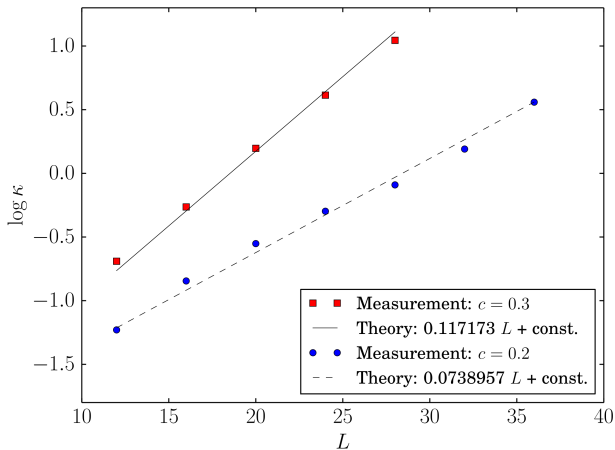


$$\text{Prob}[K] \sim e^{t\varphi(K/t)}$$

Finite-time & -size scalings matter.

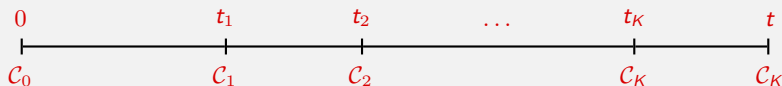


[Merolle, Garrahan and Chandler, 2005]



Exponential divergence of the susceptibility

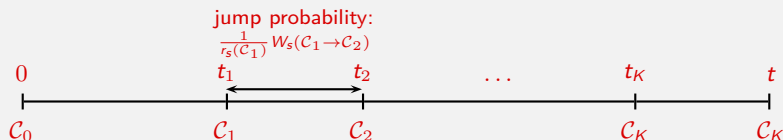
# Explicit construction (1/3)



## Probability-preserving contribution

$$\partial_t \hat{P}(C, t) = \sum_{C'} \left\{ \underbrace{W_s(C' \rightarrow C) \hat{P}(C', t)}_{\text{gain term}} - \underbrace{W_s(C \rightarrow C') \hat{P}(C, t)}_{\text{loss term}} \right\}$$

# Explicit construction (1/3)



Which configurations will be visited?

Configurational part of the trajectory:  $\mathcal{C}_0 \rightarrow \dots \rightarrow \mathcal{C}_K$

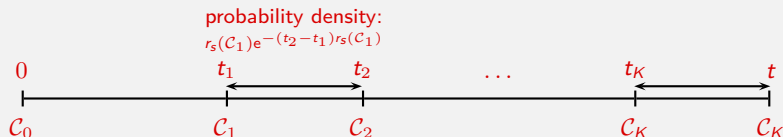
$$\text{Prob}\{\text{hist}\} = \prod_{n=0}^{K-1} \frac{W_s(\mathcal{C}_n \rightarrow \mathcal{C}_{n+1})}{r_s(\mathcal{C}_n)}$$

where

$$r_s(\mathcal{C}) = \sum_{\mathcal{C}'} W_s(\mathcal{C} \rightarrow \mathcal{C}')$$



# Explicit construction (2/3)

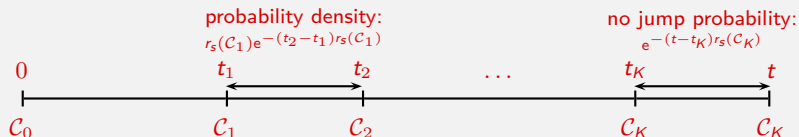


When shall the system jump from one configuration to the next one?

- probability density for the time interval  $t_n - t_{n-1}$

$$r_s(C_{n-1})e^{-(t_n-t_{n-1})r_s(C_{n-1})}$$

# Explicit construction (2/3)



When shall the system jump from one configuration to the next one?

- probability density for the time interval  $t_n - t_{n-1}$

$$r_s(C_{n-1})e^{-(t_n-t_{n-1})r_s(C_{n-1})}$$

- probability not to leave  $C_K$  during the time interval  $t - t_K$

$$e^{-(t-t_K)r_s(C_K)}$$

# Explicit construction (3/3)

$$\partial_t \hat{P}(\mathcal{C}, s) = \underbrace{\sum_{\mathcal{C}'} W_s(\mathcal{C}' \rightarrow \mathcal{C}) \hat{P}(\mathcal{C}', s) - r_s(\mathcal{C}) \hat{P}(\mathcal{C}, s)}_{\text{modified dynamics}} + \underbrace{\delta r_s(\mathcal{C}) \hat{P}(\mathcal{C}, s)}_{\text{cloning term}}$$

How to take into account loss/gain of probability?

- handle a large number of copies of the system
- implement a **selection** rule: on a time interval  $\Delta t$  a copy in config  $\mathcal{C}$  is replaced by  $e^{\Delta t \delta r_s(\mathcal{C})}$  copies
- $\psi(s)$  = the rate of exponential growth/decay of the total population
- optionally: keep population constant by non-biased pruning/cloning

# Explicit construction (3/3)

$$\partial_t \hat{P}(\mathcal{C}, s) = \underbrace{\sum_{\mathcal{C}'} W_s(\mathcal{C}' \rightarrow \mathcal{C}) \hat{P}(\mathcal{C}', s) - r_s(\mathcal{C}) \hat{P}(\mathcal{C}, s)}_{\text{modified dynamics}} + \underbrace{\delta r_s(\mathcal{C}) \hat{P}(\mathcal{C}, s)}_{\text{cloning term}}$$

How to take into account loss/gain of probability?

- handle a large number of copies of the system
- implement a **selection** rule: on a time interval  $\Delta t$   
a copy in config  $\mathcal{C}$  is replaced by  $\lfloor e^{\Delta t \delta r_s(\mathcal{C})} + \epsilon \rfloor$  copies,  $\epsilon \sim [0, 1]$
- $\psi(s)$  = the rate of exponential growth/decay of the total population
- optionally: keep population constant by non-biased pruning/cloning

# Explicit construction (3/3)

$$\partial_t \hat{P}(\mathcal{C}, s) = \underbrace{\sum_{\mathcal{C}'} W_s(\mathcal{C}' \rightarrow \mathcal{C}) \hat{P}(\mathcal{C}', s) - r_s(\mathcal{C}) \hat{P}(\mathcal{C}, s)}_{\text{modified dynamics}} + \underbrace{\delta r_s(\mathcal{C}) \hat{P}(\mathcal{C}, s)}_{\text{cloning term}}$$

How to take into account loss/gain of probability?

- handle a large number of copies of the system
- implement a **selection** rule: on a time interval  $\Delta t$   
a copy in config  $\mathcal{C}$  is replaced by  $\lfloor e^{\Delta t \delta r_s(\mathcal{C})} + \epsilon \rfloor$  copies,  $\epsilon \sim [0, 1]$
- $\psi(s) =$  the rate of exponential growth/decay of the total population
- optionally: keep population constant by non-biased pruning/cloning

# Explicit construction (3/3)

$$\partial_t \hat{P}(\mathcal{C}, s) = \underbrace{\sum_{\mathcal{C}'} W_s(\mathcal{C}' \rightarrow \mathcal{C}) \hat{P}(\mathcal{C}', s) - r_s(\mathcal{C}) \hat{P}(\mathcal{C}, s)}_{\text{modified dynamics}} + \underbrace{\delta r_s(\mathcal{C}) \hat{P}(\mathcal{C}, s)}_{\text{cloning term}}$$

How to take into account loss/gain of probability?

- handle a large number of copies of the system
- implement a **selection** rule: on a time interval  $\Delta t$   
a copy in config  $\mathcal{C}$  is replaced by  $\lfloor e^{\Delta t \delta r_s(\mathcal{C})} + \epsilon \rfloor$  copies,  $\epsilon \sim [0, 1]$
- $\psi(s)$  = the rate of exponential growth/decay of the total population
- optionally: keep population constant by non-biased pruning/cloning

# Explicit construction (3/3)

$$\partial_t \hat{P}(\mathcal{C}, s) = \underbrace{\sum_{\mathcal{C}'} W_s(\mathcal{C}' \rightarrow \mathcal{C}) \hat{P}(\mathcal{C}', s) - r_s(\mathcal{C}) \hat{P}(\mathcal{C}, s)}_{\text{modified dynamics}} + \underbrace{\delta r_s(\mathcal{C}) \hat{P}(\mathcal{C}, s)}_{\text{cloning term}}$$

## Biological interpretation

- copy in configuration  $\mathcal{C} \equiv$  organism of **genome**  $\mathcal{C}$
- dynamics of rates  $W_s \equiv$  **mutations**
- cloning at rates  $\delta r_s \equiv$  **selection** rendering atypical histories typical