

Part II - Markov dynamics and large deviation functions: first steps

lecture (2.1)
Stoch. process
2012

MOTIVATIONS:

- Classical thermodynamics provides a frame to understand the mean-value of macroscopic observables (energy, density, susceptibility, ...)
- Physical systems are composed of numerous but finite # of particles → one wants to understand fluctuations of those observables.
- In particular: even at equilibrium, fluctuation of time-integrated observables are difficult to catch

GENERIC SETUP:

e.g: currents
events

- Set of configurations $\{c\}$ (finite or discrete)
e.g: occupation numbers $\{n_i\}$ on sites i of a network / lattice.
- Transition rates $W(c \rightarrow c')$ between configurations (continuous-time dynamics) ($c \neq c'$)
- Evolution of the probability $P(c, t)$ of being in c at time t :

Master Equation

$$\partial_t P(c, t) = \sum_{c'} W(c' \rightarrow c) P(c', t) - r(c) P(c, t)$$

with $r(c) = \sum_{c'} W(c \rightarrow c')$ (escape rate from conf. c)
one assumes $W(c \rightarrow c) = 0$

• Where does this come from? **come back** to discrete time

$$P(c, t+dt) = \sum_{c'} \underbrace{dt W(c' \rightarrow c) P(c', t)}_{\text{probability to have jumped from } c' \text{ to } c \text{ between } t \text{ and } t+dt} - \underbrace{\left(1 - dt \sum_{c'} W(c \rightarrow c')\right) P(c, t)}_{\text{probability to have remained in } c \text{ btw } t \text{ and } t+dt}$$

This is also what we wrote $P(c, t+dt | c', t)$

To understand / remember / recover the second term, one remarks that it ensures probability conservation:

$$\sum_c P(c, t+dt) = \sum_c P(c, t)$$

PROPERTIES:

Lecture on (2.2)
Stoch. processes
2012

• Conservation of probability: one has

$$\partial_t \sum_{\mathcal{E}} P(\mathcal{E}, t) = 0 \quad \text{as directly checked. Since } \sum_{\mathcal{E}} P(\mathcal{E}, 0) = 1$$

indeed

$$= \sum_{\mathcal{E}, \mathcal{E}'} \{ W(\mathcal{E}' \rightarrow \mathcal{E}) P(\mathcal{E}', t) - W(\mathcal{E} \rightarrow \mathcal{E}') P(\mathcal{E}, t) \}$$

this property remains valid at all times
[this is the case if one can reach any state from any other through $\{W(\mathcal{E} \rightarrow \mathcal{E}')\}$, and if there are no recurrent states]

• Steady state:

One assumes there exist a unique steady state $P_{st}(\mathcal{E})$ solution of $\partial_t P(\mathcal{E}, t) = 0$
i.e. verifying the detailed global balance condition

$$\sum_{\mathcal{E}'} W(\mathcal{E} \rightarrow \mathcal{E}') P_{st}(\mathcal{E}) = \sum_{\mathcal{E}'} W(\mathcal{E}' \rightarrow \mathcal{E}) P_{st}(\mathcal{E}') \quad \forall \mathcal{E}$$

• "Equilibrium dynamics": detailed balance condition

$$W(\mathcal{E} \rightarrow \mathcal{E}') P_{eq}(\mathcal{E}) = W(\mathcal{E}' \rightarrow \mathcal{E}) P_{eq}(\mathcal{E}') \quad \forall \mathcal{E}, \mathcal{E}'$$

- the steady state is then called "equilibrium state" $P_{st} = P_{eq}$
- this condition is much more restrictive

• interpretation 1: there is no current of probability in the steady state

$$0 = \sum_{\mathcal{E}'} \left(\underbrace{W(\mathcal{E}' \rightarrow \mathcal{E}) P_{eq}(\mathcal{E}') - W(\mathcal{E} \rightarrow \mathcal{E}') P_{eq}(\mathcal{E})}_{\text{current of probability in } \mathcal{E} (=0)} \right)$$

• interpretation 2: the dynamics starting from P_{eq} is reversible using detailed balance condition

$$P_{eq}(\mathcal{E}_0) W(\mathcal{E}_0 \rightarrow \mathcal{E}_1) \dots W(\mathcal{E}_{k-1} \rightarrow \mathcal{E}_k) = P_{eq}(\mathcal{E}_k) W(\mathcal{E}_k \rightarrow \mathcal{E}_{k-1}) \dots W(\mathcal{E}_1 \rightarrow \mathcal{E}_0)$$

probability density of the history $\mathcal{E}_0 \rightarrow \dots \rightarrow \mathcal{E}_k$ of the system

probability density of the time-reversed history $\mathcal{E}_k \rightarrow \dots \rightarrow \mathcal{E}_0$ of the system

those probabilities are the same: reversibility \Leftrightarrow equilibrium.

Very similar to the case of Langevin dynamics (part I)

OPERATOR NOTATION:

Lecture on
Stoch. processes

(2.3)

2012

- One considers a vector space of basis $|e\rangle$, e in the set of all configs.
- The space is orthonormal, the scalar product denoted $\langle \cdot | \cdot \rangle$:

$$\langle e' | e \rangle = \delta_{ee'} \quad (\text{Kronecker delta})$$

- Translation of the Markov evolution: (master equation)

Defining the vector $|P(t)\rangle = \sum_e P(e,t) |e\rangle$ of components $P(e,t)$:

$$\partial_t |P(t)\rangle = W |P(t)\rangle \quad \text{with} \quad \boxed{W_{ee'} = W(e \rightarrow e') - n(e) \delta_{ee'}}$$

$W_{ee'}$ is the element $e \rightarrow e'$ of the matrix (or 'operator') W .
named 'evolution operator'

- Translation of the conservation of probability:

This vector is termed "the total vector", or "Attila's vector"
(car toutes ses composantes sont des uns)

$$\partial_t \sum_e P(e,t) = 0 \Leftrightarrow \sum_{e'} W_{ee'} = 0 \Leftrightarrow \boxed{\langle - | W = 0 \quad \text{with} \quad \langle - | = \sum_e \langle e |}$$

in other words, the vector $\langle - |$ is a left eigen vector of W of eigenvalue 0.

- Translation of the steady state property:

the global balance condition rewrites $\boxed{W |P_{st}\rangle = 0}$

in other words the vector $|P_{st}\rangle$ is a right eigenvalue of W of e.v. 0.

It must exist if probability is conserved, since W and W^T have the same spectrum (and a left eigen vector of W corresponds to a right \vec{e} of W^T)

→ We thus see how, using Algebra, conservation of probability ensures the existence of a steady state

- Remark: All other eigen values of W are of ~~real part~~ < 0 .

This is the Perron-Frobenius theorem. It ensures that $e^{tW} |P(0)\rangle = |P(t)\rangle \xrightarrow{t \rightarrow \infty} |P_{st}\rangle$.

Transduction of the detailed balance condition:

$$W(\mathcal{E}' \rightarrow \mathcal{E}) P_{eq}(\mathcal{E}') = W(\mathcal{E} \rightarrow \mathcal{E}') P_{eq}(\mathcal{E})$$

$\forall \mathcal{E}, \mathcal{E}'$

$$\Leftrightarrow W_{\mathcal{E}\mathcal{E}'} P_{eq}(\mathcal{E}') = W_{\mathcal{E}'\mathcal{E}} P_{eq}(\mathcal{E})$$

$$\Leftrightarrow P_{eq}^{-1/2}(\mathcal{E}) W_{\mathcal{E}\mathcal{E}'} P_{eq}^{1/2}(\mathcal{E}') = P_{eq}^{1/2}(\mathcal{E}') W_{\mathcal{E}'\mathcal{E}} P_{eq}^{-1/2}(\mathcal{E})$$

Hence W^{sym} can be interpreted as a quantum Hamiltonian (see also Langevin case)

$$\Leftrightarrow \left\{ \text{the 'symmetrised operator' } W^{sym} = \hat{P}_{eq}^{-1/2} W \hat{P}_{eq}^{1/2} \text{ is symmetric} \right\}$$

where \hat{P}_{eq} is the diagonal operator of elements $P_{eq}(\mathcal{E})$.

In this case, W^{sym} (and hence W) can be diagonalized in an orthonormal basis.

Formal solution through matrix exponentiation:

* form 1 : $\begin{cases} \partial_t |P(t)\rangle = W |P(t)\rangle \\ |P(0)\rangle = |P_0\rangle = \sum_{\mathcal{E}} P_0(\mathcal{E}) |\mathcal{E}\rangle \end{cases}$ has a solution $|P(\mathcal{E}, t)\rangle = e^{tW} |P_0\rangle = \sum_{n \geq 0} \frac{t^n}{n!} W^n |P_0\rangle$
but this form is not very useful.

* form 2 : description in terms of a jump process. To eliminate the diagonal term in $\partial_t |P(t)\rangle_{\mathcal{E}} = \sum_{\mathcal{E}'} W(\mathcal{E}' \rightarrow \mathcal{E}) |P(t)\rangle_{\mathcal{E}'} - n(\mathcal{E}) |P(t)\rangle_{\mathcal{E}}$

one sets $Q(\mathcal{E}, t) = e^{-t\lambda(\mathcal{E})} P(\mathcal{E}, t)$ which verifies: $\partial_t |Q(t)\rangle = W(t) |Q(t)\rangle$ with $W(t)_{\mathcal{E}\mathcal{E}'} = W(\mathcal{E}' \rightarrow \mathcal{E}) e^{t(\lambda(\mathcal{E}') - \lambda(\mathcal{E}))}$

Remark: if rates depend on time, set $Q(\mathcal{E}, t) = e^{-\int_0^t \lambda(\mathcal{E}(s)) ds} P(\mathcal{E}, t)$

this is a linear evolution with time-dependent operator $W(t)$

The solution is $|Q(t)\rangle = T_{exp} \left(\int_0^t W \right) |Q(0)\rangle$ (kk)

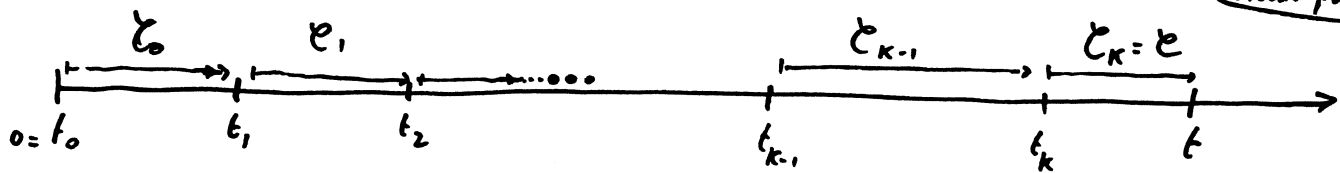
where the time-ordered exponential T_{exp} writes $K \in \mathbb{N}$

$$T_{exp} \left(\int_0^t W \right) = \sum_{K \geq 0} \mathcal{T} \left(\int_0^t W \right)^K = \sum_{K \geq 0} \int_0^t dt_K \int_0^{t_K} dt_{K-1} \dots \int_0^{t_2} dt_1 W(t_K) \dots W(t_1)$$

check that (kk) solves (*) using this expression

Final form of the result: coming back to $P(\mathcal{C}, t)$ from $Q(\mathcal{C}, t)$:

(2.5)
Lecture on
Stoch. processes
2012



sum over the number of jumps \downarrow sum over the histories of configurations \downarrow integrals over the time jump t_k 's between \mathcal{C}_{k-1} and \mathcal{C}_k \downarrow

$$P(\mathcal{C}, t) = \sum_{k \geq 0} \sum_{\mathcal{C}_0 \dots \mathcal{C}_k} \int_{t_0}^t dt_1 \int_{t_1}^t dt_2 \dots \int_{t_{k-1}}^t dt_k$$

(A) \rightarrow $n(\mathcal{C}_0) e^{-(t_1 - t_0)n(\mathcal{C}_0)} \times \dots \times n(\mathcal{C}_{k-1}) e^{-(t_k - t_{k-1})n(\mathcal{C}_{k-1})} e^{-(t - t_k)n(\mathcal{C}_k)}$

(B) \rightarrow $\times \frac{W(\mathcal{C}_0 \rightarrow \mathcal{C}_1)}{n(\mathcal{C}_0)} \times \dots \times \frac{W(\mathcal{C}_{k-1} \rightarrow \mathcal{C}_k)}{n(\mathcal{C}_{k-1})}$

sampling of the initial condition \rightarrow $\times P_0(\mathcal{C}_0)$

• in (A): all the $n(\mathcal{C}_{k-1}) e^{-(t_k - t_{k-1})n(\mathcal{C}_{k-1})}$ ($1 \leq k \leq K$) represent the probability that the time t_k of jump between \mathcal{C}_{k-1} and \mathcal{C}_k is t_k .

• $e^{-(t - t_k)n(\mathcal{C}_k)}$ represents the probability not to jump between t_k and t

hence (A) is the probability distribution of the times of jump $\{t_k\}$.

• in (B): $\frac{W(\mathcal{C}_{k-1} \rightarrow \mathcal{C}_k)}{n(\mathcal{C}_{k-1})}$ represents the (normalised) probability to jump to \mathcal{C}_k starting from \mathcal{C}_{k-1} .

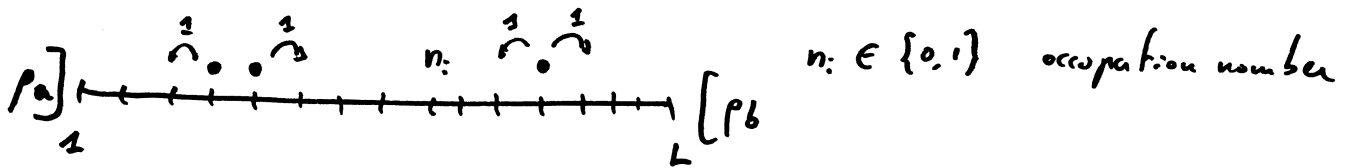
hence (B) represents the probability of the history of configurations $\{\mathcal{C}_0 \dots \mathcal{C}_k\}$

Remark: one can rewrite (A) = $\prod_{k=0}^{k-1} n(\mathcal{C}_k) \cdot e^{-\int_0^t dt' n(\mathcal{C}_{t'})}$

check that this is the correct solution when rates depend on time and hence $\nabla n(\mathcal{C}) = n(\mathcal{C}, t)$

LARGE DEVIATION FUNCTIONS

• Examples and motivation:



• total current on a time interval $[0, t]$: $Q = \#\{\text{jumps to the right}\} - \#\{\text{jumps to the left}\}$
 in time: $Q \mapsto Q+1$ each time a particle jumps to the right
 $Q \mapsto Q-1$ — — — — — left.

• total time and space integrated density:

$$\rho = \frac{1}{L} \int_0^t d\tau \sum_{i=1}^L n_i(\tau)$$

evolution in time: continuous (no jumps)

on each interval $[t_{k-1}, t_k]$ where the system does not change configuration:

$$\rho \mapsto \rho + \frac{1}{L} \int_{t_{k-1}}^{t_k} d\tau \sum_{i=1}^L \overbrace{n_i(t_{k-1})}^{\text{constant}} = \rho + (t_k - t_{k-1}) \frac{1}{L} \sum_{i=1}^L n_i(t_{k-1})$$

• 'dynamical activity' K : on a time interval $[0, t]$, $K = \#\{\text{change of configuration}\}$
 in time: $K \mapsto K+1$ each time a configuration changes

• 'integrated escape rate' R : configuration at time τ , constant $\mathcal{E}_\tau = \mathcal{E}_{k-1}$ for $\tau \in [t_{k-1}, t_k]$

$$R = \int_0^t d\tau r(\mathcal{E}_\tau)$$

$$R = \sum_{k=1}^K (t_k - t_{k-1}) r(\mathcal{E}_{k-1}) + (t - t_K) r(\mathcal{E}_K)$$

→ What is the distribution of those history-dependent observables?
 What can we learn from those distribution in terms of dynamical phase transitions?

• Generic cases: A_1, A_2 observables depending on the history of the system on $(0, t]$

$$A_1 \text{ defined as: } \begin{cases} A_1|_{t=0} = 0, & A_1 \mapsto A_1 + a_1(\mathcal{E} \rightarrow \mathcal{E}') \\ & \text{at each jump } \mathcal{E} \rightarrow \mathcal{E}' \end{cases}$$

(A_1 only changes at jump times)_t

$$A_2 \text{ defined as } A_2 = \int_0^t dt a_2(\mathcal{E}_t) = \sum_{k=1}^K (t_k - t_{k-1}) a_2(\mathcal{E}_{k-1}) + (t - t_K) a_2(\mathcal{E}_K)$$

(A_2 evolves continuously in time)

• Large deviation function:

x in direct space: for A an observable of type A_1 or A_2 :

• the probability density of being in \mathcal{E} at time t , having observed a value A of the observable, is denoted $P(\mathcal{E}, A, t)$.

• the probability distribution of A at time t is

$$P(A, t) = \int_{\mathcal{E}} P(\mathcal{E}, A, t) \quad \text{and scales as}$$

$$\boxed{P(A, t) \sim \exp(+t\pi(A/t))} \quad \text{as } t \rightarrow \infty \quad \left\{ \begin{array}{l} \text{Other scalings,} \\ \text{are possible, but for} \\ \text{finite systems, this} \\ \text{is the most generic.} \end{array} \right.$$

π is a dynamical equivalent of the entropy
 $\pi(a)$ describes more than the mean and the variance of a "large deviation function"

• π is difficult to determine in general ("microcanonical problem")
 one prefers to go to the "canonical dynamical ensemble"

x in Laplace space: one has the $t \rightarrow \infty$ scaling

$$\boxed{\langle e^{-sA} \rangle \sim \exp(t\psi(s))} \quad \text{as } t \rightarrow \infty$$

average taken on histories on time interval $(0, t)$

$\psi(s)$ is the cumulant generating function: (ψ is a "dynamical free energy")

$$\boxed{\partial_s^k \psi|_{s=0} = (-1)^k \frac{1}{t} \langle A^k \rangle_{\mathcal{E}_t}} \quad \langle A^k \rangle_{\mathcal{E}_t} \text{ is the } k^{\text{th}} \text{ cumulant of } A.$$

Interpretation of s; link between $\Psi(s)$ and $\Pi(a)$:

* Starting from $\langle e^{-sA} \rangle = \int dA P(A,t) e^{-sA}$ A = at

$e^{t\Psi(s)} \sim \int da e^{t[\Pi(a) - sa]}$
in the $t \rightarrow \infty$ limit, one may use the saddle point theorem

One has $\Psi(s) = \max_a (\Pi(a) - sa)$ Ψ and Π are Legendre transformed.

If Π is convex, one may perform the inverse Legendre transform:

$\Pi(a) = \min_s (\Psi(s) + sa)$

* If $a^*(s)$ is the a where the max is reached or if $s^*(a) = s = \min_s$ one says that $a^*(s)$ and s are 'conjugated' and $s^*(a)$ and a are 'conjugated'

* s plays a role similar to the inverse temperature β : it fixes the average value of A (in the same way as β fixes the average value of the energy E)

* Indeed, in the large time limit: the mean value of an observable O in the 's-state' writes

$\langle O(t) \rangle_s \equiv \frac{\langle O(t) e^{-sA} \rangle}{\langle e^{-sA} \rangle} = \frac{1}{\langle e^{-sA} \rangle} \sum_{\mathcal{E}, A} \tilde{P}(\mathcal{E}, A, t) O(\mathcal{E}) e^{-sA}$ A = at
 $\sim \frac{\int da \sum_{\mathcal{E}} \tilde{P}(\mathcal{E}, a) O(\mathcal{E}) e^{t(\Pi(a) - sa)}}{\int da e^{t(\Pi(a) - sa)}}$
the saddle is reached at the same value $a = a^*(s)$

$\langle O(t) \rangle_s \sim \sum_{\mathcal{E}} \tilde{P}(\mathcal{E}, a^*(s)) O(\mathcal{E})$: mean value in the s-state, of $O(\mathcal{E})$ at final time
 $\langle O(t) \rangle_s \xrightarrow{t \rightarrow \infty} \langle O(t) \rangle_{A = a^*(s)t}$
Mean value of $O(\mathcal{E})$ at final time for histories with $A = a^*(s)t$

In the same way, if $\pi(a)$ is convex: by Legendre transform

$$\langle O(\epsilon) \rangle_{A=at} \stackrel{t \rightarrow \infty}{=} \langle O(\epsilon) \rangle_{s=s^*(a)}$$

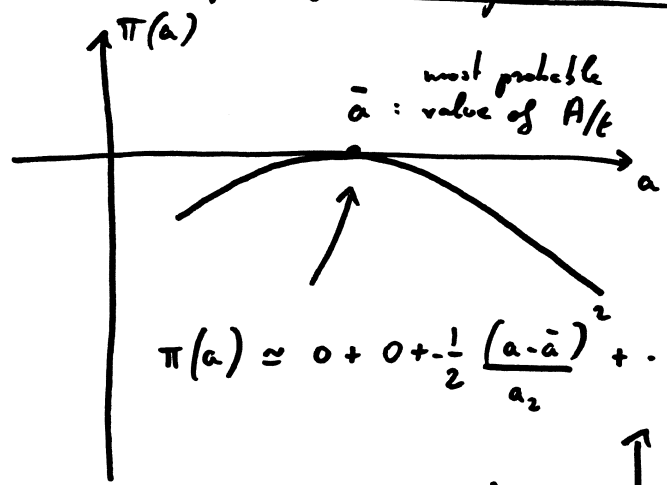
Mean value of $O(\epsilon)$ at final time for histories with a value $A=at$ of the observable A

Mean value in the s -state of $O(\epsilon)$ at final time:

$$\langle O(\epsilon) \rangle_s = \frac{\langle O(\epsilon(t)) e^{-sA} \rangle}{\langle e^{-sA} \rangle}$$

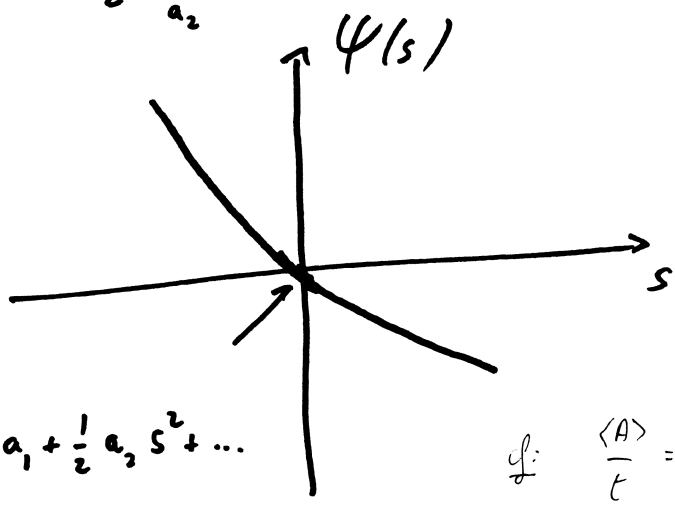
In other words, to characterize the value of the observable O in histories with a value $A=at$ of the observable A , one has to compute $\langle O(\epsilon) \rangle_s$ at $s=s^*(a)$.

• Generic shape of the functions $\pi(a)$ and $\psi(s)$:



$\bar{a} = a_1 = \frac{\langle A \rangle}{t}$ mean value
 $a_2 = \frac{\langle A^2 \rangle_c}{t}$ second cumulant

$\pi(a) \approx 0 + 0 + \frac{1}{2} \frac{(a-\bar{a})^2}{a_2} + \dots \rightarrow P(A=at, t) \approx e^{-\frac{1}{2} \frac{(a-\bar{a})^2}{a_2} t}$ [Gaussian distribution]



For a reference on large deviations, Laplace & Legendre transforms, see Hugo Touchette
 The large deviation approach to statistical mechanics
 Physics Reports 478 1 (2009)

$$\psi(s) \approx -s a_1 + \frac{1}{2} a_2 s^2 + \dots$$

$$\frac{\langle A \rangle}{t} = - \frac{\partial \psi}{\partial s} \Big|_{s=0}$$

$$\frac{\langle A^2 \rangle_c}{t} = \frac{\partial^2 \psi}{\partial s^2} \Big|_{s=0}$$

• Large deviation functions as maximal eigenvalues

Lectures on (2.10)
Stoch. processes
2012

1. Cases of observables of type A_1

* Time evolution for $P(e, A_1, t)$:

$$\partial_t P(e, A_1, t) = \sum_{e'} W(e' \rightarrow e) P(e', A_1 - a_1(e' \rightarrow e), t) - r(e) P(e, A_1, t)$$

this evolution is non-diagonal in the direction A_1 ,
the operator of evolution is very difficult to diagonalize.

* Going to the s -state: Laplace transform

This quantity is more detailed than $\langle e^{-sA_1} \rangle$, but it obeys closed equation of evolution, as detailed below.

One introduces $\hat{P}(e, s, t) = \sum_{A_1} e^{-sA_1} P(e, A_1, t)$

It verifies $\langle e^{-sA_1} \rangle = \sum_{e, A_1} e^{-sA_1} P(e, A_1, t) = \sum_e \hat{P}(e, s, t)$

* Time evolution: from the time evolution of $P(e, A_1, t)$ one finds

$$\partial_t \hat{P}(e, s, t) = \sum_{e'} e^{-s a_1(e' \rightarrow e)} W(e' \rightarrow e) \hat{P}(e', s, t) - r(e) \hat{P}(e, s, t)$$

This time, the operator is diagonal in direction s . Putting $|\hat{P}(t)\rangle = \sum_e \hat{P}(e, s, t) |e\rangle$

one has

$$\partial_t |\hat{P}(s, t)\rangle = W(s) |\hat{P}(s, t)\rangle$$

$$(W(s))_{ee'} = e^{-s a_1(e' \rightarrow e)} W(e' \rightarrow e) - r(e) \delta_{ee'}$$

here, only the non-diagonal part of the evolution operator is modified by s : it is the part describing jumps

* Largest eigen value of $W(s)$: one has $\langle e^{-sA_1} \rangle = \sum_e \hat{P}(e, s, t)$
 but $|\hat{P}(s, t)\rangle = e^{t W(s)} |\hat{P}_0\rangle \sim e^{t \max Sp W(s)} \rightarrow \nu e^{t \max Sp W(s)}$
 maximal eigenvalue of $W(s)$

Hence $\langle e^{-sA_1} \rangle \sim e^{t \max Sp W(s)}$ and by definit°

$$\psi(s) = \max Sp W(s)$$

it is the largest eigenvalue of $W(s)$

2. Case of observable of type A_2

$\partial_t P(x, A_2, t)$ is not as directly obtained as previously.

* first approach: time discretization ok up to order dt

$$P(x, A_2, t+dt) = \int_{x'} dt W(x' \rightarrow x) P(x', A_2, t) + (1 - dt r(x)) P(x, A_2 - \int_t^{t+dt} a_2(x(\tau)) d\tau) \approx dt a_2(x)$$

Hence, in the $dt \rightarrow 0$ limit: $P(x, A_2, t) - dt \partial_{A_2} P(x, A_2, t) a_2(x)$

$$\partial_t P(x, A_2, t) = \int_{x'} W(x' \rightarrow x) P(x', A_2, t) - (r(x) P(x, A_2, t) + a_2(x) \partial_{A_2} P(x, A_2, t))$$

Or, introducing $\hat{P}(x, s, t) = \int dA_2 e^{-s A_2} P(x, A_2, t)$
one again has $\langle e^{-s A_2} \rangle = \int dA_2 \langle P(x, A_2, t) \rangle = \int \hat{P}(x, s, t)$

Thanks to Laplace transform, the derivative ∂_{A_2} becomes a simple factor

The equation of evolution is thus (by integration by part)

$$\partial_t \hat{P}(x, s, t) = \int_{x'} W(x' \rightarrow x) \hat{P}(x', s, t) - (r(x) + s a_2(x)) \hat{P}(x, s, t)$$

Or, in vector notation:

$$\partial_t |\hat{P}(s, t)\rangle = W(s) |\hat{P}(s, t)\rangle$$

$$(W(s))_{xx'} = W(x' \rightarrow x) - (r(x) + s a_2(x)) \delta_{xx'}$$

this time, it is the diagonal part of the operator of evolution which is modified by s

This corresponds to the fact that $A_2 = \int_0^t a_2(x(\tau)) d\tau$ is an observable which does not evolve at jumps.

Remark: This formula, written as: $\partial_t |\hat{P}(s, t)\rangle = W |\hat{P}(s, t)\rangle - s \hat{a}_2 |\hat{P}(s, t)\rangle$ diagonal operators of elements $a_2(x)$

jumps are described by the non-diagonal part of W

is an example of Feynman-Kac formula

* second derivation of the result:

(2.12)
Lecture on
Stoch. processes
2012

One has $A_2 = \sum_{k=1}^K (t_k - t_{k-1}) a_2(\xi_{k-1}) + (t - t_K) a_2(\xi_K)$

Thus, using the expression previously obtained for the probability of an history.

$$\langle e^{-sA_2} \rangle = \sum_{K \geq 0} \sum_{\xi_0, \dots, \xi_K} \int_{\xi_0}^t dt_1 \dots \int_{\xi_{K-1}}^t dt_K e^{-\sum_{k=1}^K (t_k - t_{k-1}) (r(\xi_{k-1}) + s a_2(\xi_{k-1})) - (t - t_K) (r(\xi_K) + s a_2(\xi_K))} \times \prod_{k=1}^K W(\xi_{k-1} \rightarrow \xi_k) P_0(\xi_K)$$

Indeed the derivation of the time-ordered exp. formula does not involve the conservation of probability presented by W .
And one recognizes precisely the expression of:
It thus also works for $W(s)$

$$\langle - | T \exp \underline{W}_s(t) | P_0 \rangle$$

time dependent operator of evolution
corresponding precisely to $W(s)$ for s associated to A_2

which is compatible with $|\hat{P}(s, t)\rangle = T \exp W_s(t) | P_0 \rangle$

• Remark: mixed observables:

The l.d.f. $\Psi(s_1, s_2) = \lim_{t \rightarrow \infty} \frac{1}{t} \log \langle e^{-s_1 A_1 - s_2 A_2} \rangle$ is the max e.v. of

$$\left(W(s_1, s_2) \right)_{\xi \xi'} = e^{-s_1 a_1(\xi' \rightarrow \xi)} W(\xi' \rightarrow \xi) - (r(\xi) + s_2 a_2(\xi)) \delta_{\xi \xi'}$$

both diagonal and non-diagonal parts of the operator of evolution are modified, respectively by s_2 and s_1

Example:

Time reversal symmetry and fluctuation theorem:

(J. Stat. Phys. 95 333 (1999), Lebowitz & Spohn)

Lecture on Stoch. processes 2.13 2012

Consider the observable $Q_{\epsilon_s} = \text{Log} \frac{W(\epsilon_0 \rightarrow \epsilon_1) \dots W(\epsilon_{K-1} \rightarrow \epsilon_K)}{W(\epsilon_K \rightarrow \epsilon_{K-1}) \dots W(\epsilon_1 \rightarrow \epsilon_0)}$ ("entropy current")

Upon a jump $\epsilon \rightarrow \epsilon'$ one has $Q_{\epsilon_s} \mapsto Q_{\epsilon_s} + \text{Log} \frac{W(\epsilon \rightarrow \epsilon')}{W(\epsilon' \rightarrow \epsilon)}$

The modified operator of evolution for s associated to Q_{ϵ_s} is

$$\begin{aligned} (W(s))_{\epsilon\epsilon'} &= e^{-s \text{Log} \frac{W(\epsilon' \rightarrow \epsilon)}{W(\epsilon \rightarrow \epsilon')}} W(\epsilon' \rightarrow \epsilon) - n(\epsilon) \delta_{\epsilon\epsilon'} \\ &= W(\epsilon' \rightarrow \epsilon)^{1-s} W(\epsilon \rightarrow \epsilon')^s - n(\epsilon) \delta_{\epsilon\epsilon'} \end{aligned}$$

Results of this class are among the few results valid in non-equilibrium steady states.

One thus has $(W(s))_{\epsilon\epsilon'} = (W(1-s))_{\epsilon'\epsilon}$
 $W(s)^T = W(1-s)$

Relates rare events ($s \neq 1$) to not rare ones ($s=0$). Implies Fluctuation-Dissipation Relation, and Onsager Relations.

Since these two operators have the same spectrum, one has $\Psi(s) = \Psi(1-s)$

This symmetry is an instance of 'Gallavotti-Cohen relation' (or 'Fluctuation Theorem') [J. Stat. Phys 80: 931 (1995)]

Numerical algorithm for large deviation function: s conjugated to $A=A$.

By writing $W_s(\epsilon \rightarrow \epsilon') = e^{-sA(\epsilon \rightarrow \epsilon')} W(\epsilon \rightarrow \epsilon')$; $r_s(\epsilon) = \sum_{\epsilon'} W_s(\epsilon \rightarrow \epsilon')$

$$(W(s))_{\epsilon\epsilon'} = \underbrace{W_s(\epsilon' \rightarrow \epsilon) - r_s(\epsilon) \delta_{\epsilon\epsilon'}}_{\text{probability-conserving evolution with modified rates } W_s(\epsilon \rightarrow \epsilon')} + \underbrace{(r_s(\epsilon) - n(\epsilon)) \delta_{\epsilon\epsilon'}}_{\text{cloning with rate } r_s(\epsilon) - n(\epsilon)}$$

(probability-conserving) evolution with modified rates $W_s(\epsilon \rightarrow \epsilon')$ for a population of copies of the system.

One can devise an algorithm to measure $\Psi(\epsilon)$ numerically:

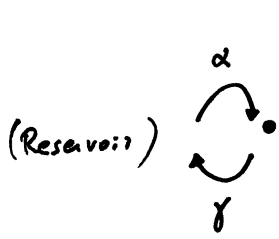
- evolution with rate $W_s(\epsilon \rightarrow \epsilon')$ of each copy
- cloning (pruning) of copies with rate $r_s(\epsilon) - n(\epsilon)$

} favoring configurations with an atypical value of A .

then the global increase/decrease rate of the population is $\Psi(s)$

[There are tricks to keep the population size constant - see Refs.]

• An example: death and birth process on one site



$n=0$ site empty $W(0 \rightarrow 1) = \alpha$ (birth)
 $n=1$ site occupied $W(1 \rightarrow 0) = \gamma$ (death)

$$\partial_t P(0,t) = \gamma P(1,t) - \alpha P(0,t)$$

$$\partial_t P(1,t) = \alpha P(0,t) - \gamma P(1,t)$$

$\alpha > 0$
 $\gamma > 0$

$$\partial_t \begin{pmatrix} P(0,t) \\ P(1,t) \end{pmatrix} = \underbrace{\begin{pmatrix} -\alpha & \gamma \\ \alpha & -\gamma \end{pmatrix}}_W \underbrace{\begin{pmatrix} P(0,t) \\ P(1,t) \end{pmatrix}}_{|P(t)\rangle}$$

indeed

$$W_{01} = W(1 \rightarrow 0) = \gamma$$

$$W_{10} = W(0 \rightarrow 1) = \alpha$$

$$W_{00} = -\alpha(0) = -W(0 \rightarrow 1) = -\alpha$$

$$W_{11} = -\alpha(1) = -W(1 \rightarrow 0) = -\gamma$$

* search for a steady-state: one even has an equilibrium state (detailed balance ok)

"Grand-canonical" form $P(n) = \frac{1}{Z} e^{\mu n}$

St. st. eq:

$$\begin{cases} 0 = \gamma e^{\mu} - \alpha \\ 0 = \alpha - \gamma e^{\mu} \end{cases}$$

$$\left. \begin{aligned} e^{\mu} &= \frac{\alpha}{\gamma} \\ Z &= e^{0\mu} + e^{1\mu} = 1 + \frac{\alpha}{\gamma} \end{aligned} \right\}$$

μ = "chemical potential"

$$P(n)_{eq} = \frac{\left(\frac{\alpha}{\gamma}\right)^n}{1 + \frac{\alpha}{\gamma}} \quad (\text{normalized})$$

In vector form: $|P_{eq}\rangle = \frac{1}{\alpha + \gamma} \begin{pmatrix} \gamma \\ \alpha \end{pmatrix}$ is indeed a right eigen-vector of W of ev. 0.

* mean occupation number:

$$\langle n \rangle = \sum_n n P(n)_{eq} = \frac{\alpha/\gamma}{1 + \alpha/\gamma} = \frac{\alpha}{\alpha + \gamma}$$

In details: one rewrites $\partial_t |P(t)\rangle = (\alpha + \gamma) \begin{pmatrix} -\frac{\alpha}{\alpha + \gamma} & \frac{\gamma}{\alpha + \gamma} \\ \frac{\alpha}{\alpha + \gamma} & -\frac{\gamma}{\alpha + \gamma} \end{pmatrix} |P(t)\rangle$
and one changes time t into $t' = (\alpha + \gamma)t$

* reparametrization: one sets $\alpha = c \quad \gamma = 1 - c$ which fixes the unit of time $0 < c < 1$

$$\begin{cases} P_{eq}(n) = c^n (1-c)^{1-n} \\ \langle n \rangle = c \end{cases}$$

Binomial law

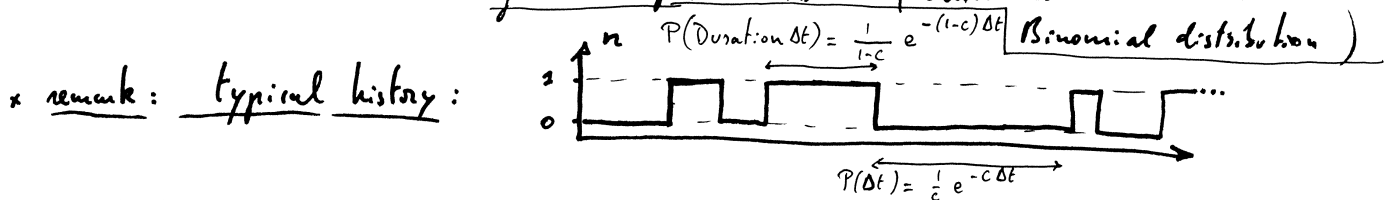
$$W = \begin{pmatrix} c & 1-c \\ c & -(1-c) \end{pmatrix}$$

In vector form $|P_{eq}\rangle = \begin{pmatrix} 1-c \\ c \end{pmatrix} = (1-c)|0\rangle + c|1\rangle$

* generating function of the cumulant of the occupation number: $\langle e^{-sn} \rangle_{eq} = e^{\psi_{occ.}(s)}$

$$\psi_{occ.}(s) \equiv \log \langle e^{-sn} \rangle_{eq} = \log \{ 1 - c + c e^{-s} \}$$

Describes all the statistics of the occupation number (which is trivial here:



* Large deviations of the time-averaged density $\rho(t) = \int_0^t d\tau n(\tau)$:

This is an observable of type 2 (time-integrated)

Mean value: one expects $\langle \rho \rangle = \int_0^t d\tau \langle n(\tau) \rangle \sim ct$
 $\approx c$ at large times

Variance: $\langle \rho(t)^2 \rangle = \langle (\int_0^t d\tau n(\tau))^2 \rangle = \int_0^t d\tau \int_0^t d\tau_2 \langle n(\tau) n(\tau_2) \rangle$
 difficult to determine since we don't know the 2-time correlation function

We thus go to the generating functional of the cumulants of ρ

$\langle e^{-s\rho(t)} \rangle \underset{t \rightarrow \infty}{\sim} e^{t \Psi_\rho(s)}$
The index ρ is there to remind us of which observable $\Psi_\rho(s)$ is the c.g.f.

We have seen that

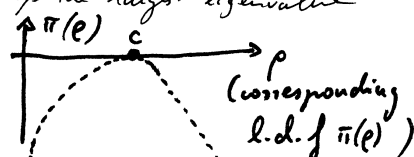
$\Psi_\rho(s) = \text{Max}_{S_p} W_p(s)$, $(W_p(s))_{ee'} = W(e \rightarrow e') - (n(e) + s n(e)) \delta_{ee'}$
spectrum

In our case:

$W_p(s) = \begin{pmatrix} c & 1-c \\ c & -[(1-c)+s] \end{pmatrix}$

Eigenvalues are solutions in λ of $\det \begin{pmatrix} c-\lambda & 1-c \\ c & -\lambda-[(1-c)+s] \end{pmatrix} = 0$

One finds $S_p W_p(s) = \left\{ -\frac{1+s}{2} \pm \frac{1}{2} \sqrt{(1+s)^2 - 4cs} \right\}$

$\Psi_\rho(s) = -\frac{1+s}{2} + \frac{1}{2} \sqrt{(1+s)^2 - 4cs}$ \rightarrow 
always the largest eigenvalue (corresponding l.d.f. of $\pi(e)$)

Which is very different from $\Psi_{occ.}(s) = \log \{ 1-c + c e^{-s} \}$

One can compare the successive cumulants from (i) $\langle n^k \rangle_c = \frac{\partial^k}{\partial s^k} \Psi_{occ.}(s)$; (ii) $\langle \rho^k \rangle_c = \frac{\partial^k}{\partial s^k} \Psi_\rho(s)$

- | | |
|---|--|
| $\langle n \rangle = c$ | $\frac{1}{t} \langle \rho \rangle = c$ |
| $\langle n^2 \rangle_c = c(1-c)$ | $\frac{1}{t} \langle \rho^2 \rangle_c = 2c(1-c)$ |
| $\langle n^3 \rangle_c = c(1-c)(1-2c)$ | $\frac{1}{t} \langle \rho^3 \rangle_c = 6c(1-c)(1-2c)$ |
| $\langle n^4 \rangle_c = c(1-c)(1-6c(1-c))$ | $\frac{1}{t} \langle \rho^4 \rangle_c = 24c(1-c)(1-5c(1-c))$ |
| ... | ... |

x Large deviations of the activity $K = \# \{ \text{configuration changes} \}$: (2.16)

lecture on
Stoch. processes
2012

. The mean value only depends on the eq. equilibrium state, in the large time limit:

Indeed:

$$\partial_t P(n, K, t) = c P(0, K-1, t) \delta_{n1} + (1-c) P(1, K-1, t) \delta_{n0} - [c P(0, K, t) + (1-c) P(1, K, t)]$$

multiply by K
and sum
over K and n

for these terms of the sum, one does $K \rightarrow K+1$

$$\partial_t \sum_{n, K} K P(n, K, t) = \sum_K \left\{ c K P(0, K-1, t) + K(1-c) P(1, K-1, t) - c K P(0, K, t) - (1-c) K P(1, K, t) \right\}$$

by definit^o: $\langle K \rangle$

averages over histories or $[0, t]$

$$\partial_t \langle K \rangle = \sum_K \left\{ c \underbrace{[K+1-K]}_{=1} P(0, K, t) + (1-c) \underbrace{[K+1-K]}_{=1} P(1, K, t) \right\}$$

$$\partial_t \langle K \rangle = c P(0, t) + (1-c) P(1, t) \quad \left\{ \begin{array}{l} \sum_K P(0, K, t) = P(0, t) \\ \sum_K P(1, K, t) = P(1, t) \end{array} \right.$$

for $t \rightarrow \infty$: $\partial_t \langle K \rangle = \langle c(1-n) + (1-c)n \rangle_{eq} = \langle r(n) \rangle_{eq}$ this relation is generic
check it!

average over histories

average in the equilibrium steady state

$r(n) = c(1-n) + (1-c)n$
is the escape rate from configuration n .

In our case: $\frac{1}{t} \langle K \rangle = 2c(1-c) \quad t \rightarrow \infty$

There is no generic way to extend this result for higher moments $\frac{1}{t} \langle K^k \rangle_c$ $k \geq 2$

. Generating function: $\Psi_K(s) = \frac{1}{t} \log \langle e^{-sK} \rangle = \text{Max}_p S_p W_K(s)$

$(W_K(s))_{pp'}$ = $e^{-s} W(p' \rightarrow p) - r(p) \delta_{pp'}$, hence

$$W_K(s) = \begin{pmatrix} -c & e^{-s}(1-c) \\ e^{-s}c & -(1-c) \end{pmatrix}$$

One finds $\Psi_K(s) = -\frac{1}{2} + \frac{1}{2} \sqrt{1 - 4c(1-c)(1 - e^{-2s})}$

No surprise here:
everything is analytic

from which one gets for instance

$$\frac{1}{t} \langle K \rangle = 2c(1-c)$$

$$(t \rightarrow \infty) \quad \frac{1}{t} \langle K^2 \rangle_c = 4c(1-c)(1 - 2c(1-c))$$

⋮

Population algorithm to measure $\Psi_K(s)$: for $s \leftrightarrow K$

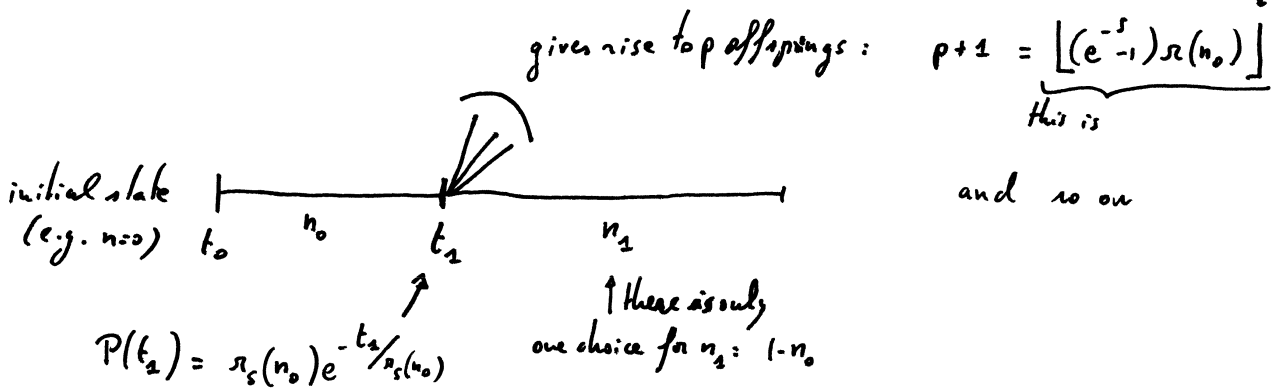
$$\partial_t \hat{P}(n, s, t) = \underbrace{e^{-s} [cn + (rc)(n)]}_{\text{death \& birth process with rates}} \hat{P}(n-1, s, t) - \underbrace{e^{-s} (c(n) + n)}_{\substack{\text{death rate } e^{-s}c \\ \text{birth rate } e^{-s}(rc)}} \hat{P}(n, s, t) + \underbrace{(e^{-s}-1) (c(n) + n(rc))}_{\substack{\text{cloning at rate } (e^{-s}-1)r(n)}} \hat{P}(n, s, t)$$

For this simple example, we keep total population $N(t)$ non-constant.

Start with N_0 copies of the system ($N_0 \gg 1$, typically $10^2 - 10^3$)

Take $s < 0$ (so that $e^{-s} > 1$).

Each copy of the system evolves as follows:



Remark: it is better to take $p+1 = \lfloor (e^{-s}-1)r(n_0) + \epsilon \rfloor$ where ϵ is uniformly distributed on $[0, 1]$

Each of the offsprings evolves independently afterwards.

In the large time limit: $N(t) \sim e^{t \Psi_K(s)}$