

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

between (2.1)
Stoch. process

NOTIFICATIONS:

- 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

GENERIC SETUP: are difficult to catch

e.g.: currents
events

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

$$\frac{\partial}{\partial t} P(\mathcal{C}, t) = \sum_{\mathcal{C}'} W(\mathcal{C}' \rightarrow \mathcal{C}) P(\mathcal{C}', t) - n(\mathcal{C}) P(\mathcal{C}, t)$$

with $n(\mathcal{C}) = \sum_{\mathcal{C}'} W(\mathcal{C} \rightarrow \mathcal{C}')$ (escape rate from conf. \mathcal{C})

one assumes $W(\mathcal{C} \rightarrow \mathcal{C}) = 0$

- Where does this come from? Come back to discrete time

$$P(\mathcal{C}, t+dt) = \sum_{\mathcal{C}'} \underbrace{dt W(\mathcal{C}' \rightarrow \mathcal{C})}_{\substack{\text{probability to have} \\ \text{jumped from } \mathcal{C}' \text{ to } \mathcal{C} \\ \text{between } t \text{ and } t+dt}} P(\mathcal{C}', t) + \underbrace{\left(1 - dt \sum_{\mathcal{C}'} W(\mathcal{C} \rightarrow \mathcal{C}')\right)}_{\substack{\text{prob. to have} \\ \text{remained in } \mathcal{C} \text{ b/w} \\ t \text{ and } t+dt}} P(\mathcal{C}, t)$$

This is also what we wrote $P(\mathcal{C}, t+dt | \mathcal{C}', t)$

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

$$\sum_{\mathcal{C}} P(\mathcal{C}, t+dt) = \sum_{\mathcal{C}} P(\mathcal{C}, t)$$

PROPERTIES:

Lecture on
Stoch. processes (2.2)

- Conservation of probability: one has

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

indeed $\frac{d}{dt} \sum_{\epsilon} \{ W(\epsilon|t) P(\epsilon, t) - W(t|\epsilon) P(t, \epsilon) \} = 0$

This property remains valid at all times
[This is the case if one can reach any state from any other]
through $\{W(t|\epsilon)\}$, and if there are no recurrent states

- Steady state:

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

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

"also named: "reversible dynamics"

- Equilibrium dynamics: detailed balance condition

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

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

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

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

$$P_{eq}(\epsilon_0) W(\epsilon_0 \rightarrow \epsilon_1) \dots W(\epsilon_{K-1} \rightarrow \epsilon_K) = P_{eq}(\epsilon_K) W(\epsilon_K \rightarrow \epsilon_{K-1}) \dots W(\epsilon_1 \rightarrow \epsilon_0)$$

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

probability density of the
history $\epsilon_0 \rightarrow \dots \rightarrow \epsilon_K$ of
the system

probability density of the
time-reversed history $\epsilon_K \rightarrow \dots \rightarrow \epsilon_0$ of
the system

Those probabilities are the same: reversibility \Leftrightarrow equilibrium.

OPERATOR NOTATION:

(2.3)
lecture on
Stoch. processes

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

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

- Traduction of the Markov evolution: (master equation)

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

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

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

- Traduction of the conservation of probability:

$$\partial_t \sum_e P(e,t) = 0 \Leftrightarrow \sum_e W_{ee'} \Leftrightarrow \langle -1 | W = 0 \quad \text{with } \langle -1 = \sum_e \langle e |$$

in other words, the vector $\langle -1$ is a left eigen vector of W of eigenvalue 0.

- Traduction of the steady state properties:

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

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 e.v. of W^T)

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

- Remark: All other eigenvalues of W are of real part < 0.

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

- Traduction of the detailed balance condition:

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

forall

$$\Leftrightarrow W_{ee'} P_{eq}(e') = W_{e'e} P_{eq}(e)$$

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

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

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

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

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

- Formal solution through matrix exponentiation:

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

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

one sets $Q(e,t) = e^{-t n(e)} P(e,t)$ which verifies: $\begin{cases} \text{Remark: if rates depend on time,} \\ \text{set } Q(e,t) = e^{-\int_0^t n(e(\tau)) d\tau} P(e,t) \end{cases}$

$$(*) \quad \partial_t |Q(t)\rangle = W(t) |Q(t)\rangle \text{ with } W(t)_{ee'} = W(e' \rightarrow e)_e$$

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

The solution is $|Q(t)\rangle = T_{exp}(\int_0^t W) |Q(0)\rangle$ ($t \in \mathbb{R}$)

where the time-ordered exponential T_{exp} writes

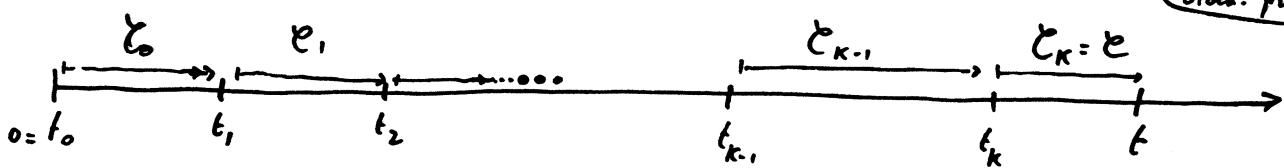
$$\begin{aligned} T_{exp}(\int_0^t W) &= \sum_{K \geq 0} T(\int_0^t W)^K = \sum_{K \geq 0} \int_0^t dt_K \int_0^{t_K} dt_{K-1} \dots \int_0^{t_2} dt_1 W(t_K) \dots W(t_1) \\ &= \sum_{K \geq 0} \int_0^t dt_1 \int_{t_1}^t dt_2 \dots \int_{t_{K-1}}^t dt_K W(t_K) \dots W(t_1) \end{aligned}$$

 $K \in \mathbb{N}$

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

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

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sum over the
number of jumps
↓
sum over the
histories of configurations
↓
integrals over the time jumps
 t_k 's between ϵ_{k-1} and ϵ_k

$$P(\epsilon, t) = \sum_{K \geq 0} \sum_{\epsilon_0, \dots, \epsilon_K} \int_{t_0}^t dt_1 \int_{t_1}^t dt_2 \dots \int_{t_{K-1}}^t dt_K$$

(A) →
 $n(\epsilon_0) e^{-(t_1 - t_0)n(\epsilon_0)} \times \dots \times n(\epsilon_{K-1}) e^{-(t_K - t_{K-1})n(\epsilon_{K-1})} e^{-(t - t_K)n(\epsilon_K)}$

(B) →
 $\times \frac{w(\epsilon_0 \rightarrow \epsilon_1)}{n(\epsilon_0)} \times \dots \times \frac{w(\epsilon_{K-1} \rightarrow \epsilon_K)}{n(\epsilon_{K-1})}$

sampling of
the initial
condition →

- in (A) : all the $n(\epsilon_{k-1}) e^{-(t_k - t_{k-1})n(\epsilon_{k-1})}$ ($1 \leq k \leq K$) represent the probability that the time t_k of jump between ϵ_{k-1} and ϵ_k is t_k .
- $e^{-(t - t_K)n(\epsilon_K)}$ represents the probability not to jump between t_K and t . hence (A) is the probability distribution of the times of jumps $\{t_k\}$.
- in (B) : $\frac{w(\epsilon_{k-1} \rightarrow \epsilon_k)}{n(\epsilon_{k-1})}$ represents the (normalised) probability to jump to ϵ_k starting from ϵ_{k-1} .

hence (B) represents the probability of the history of configurations $(\epsilon_0, \dots, \epsilon_K)$

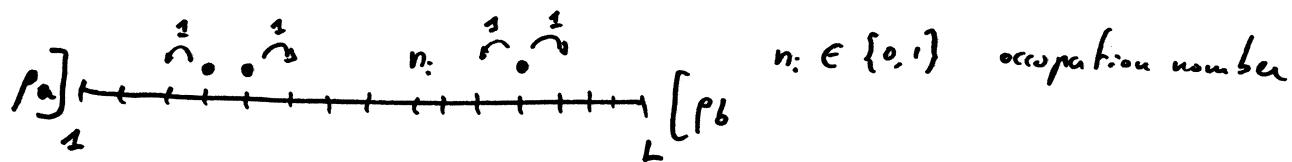
Remark: one can rewrite (A) = $\prod_{k=0}^{K-1} n(\epsilon_k) \cdot e^{-\int_0^t dt' n(\epsilon_{t'})}$

check that this is the correct solution when rates depend on time and hence $d n(\epsilon) = n(\epsilon, t)$

LARGE DEVIATION FUNCTIONS

(2.6)
Lecture
Stoch. Process

- Examples and motivation:



$n_i \in \{0, 1\}$ occupation number

- total current on a time interval $[0, t]$: $Q = \#\{\text{jumps to the right}\} - \#\{\text{jumps to the left}\}$
in time: $Q \rightarrow Q+1$ each time a particle jumps to the right
 $Q \rightarrow Q-1$ — — — 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 $[t_{k..}, t_k]$ [where the system does not change configuration]:

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

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

- 'integrated escape rate' R : configuration at time τ , constant $C_\tau = C_{t_{k..}}$ for $\tau \in [t_{k..}, t_k]$

$$R = \int_0^t d\tau n(C_\tau)$$

$$R = \sum_{k=1}^K (t_k - t_{k..}) n(C_{t_{k..}}) + (t - t_K) n(C_K)$$

→ 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 defined as: $\{A_1|_{t=0}, A_1 \mapsto A_1 + a_1(\epsilon \rightarrow \epsilon')\}$
 at each jump $\epsilon \rightarrow \epsilon'$
 $(A_1 \text{ only changes at jump times})$

A_2 defined as $A_2 = \int_0^t d\tau a_2(\epsilon_\tau) = \sum_{k=1}^K (t_k - t_{k-1}) a_2(\epsilon_{k-1}) + (t - t_K) a_2(\epsilon_K)$
 $(A_2 \text{ evolves continuously in time})$

- Large deviation function:

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

. the probability density of being in ϵ at time t , having observed a value A of the observable, is denoted $P(\epsilon, A, t)$.

. the probability distribution of A at time t is

$$P(A, t) = \sum_{\epsilon} P(\epsilon, A, t) \quad \text{and scales as}$$

$$P(A, t) \sim \exp\left(t\pi(A/t)\right)$$

as $t \rightarrow \infty$ Other scalings
are possible, but for
finite systems, this
is the most generic.

π is a dynamical equivalent of the entropy
 $\pi(a)$ describes more than the mean and the variance of a \rightarrow "large deviation function"
. it is difficult to determine in general ("microcanonical problem")
one prefers to go to the "canonical dynamical ensemble"

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

$$\langle e^{-sA} \rangle \sim \exp\left(-t\psi(s)\right)$$

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")

$$\frac{\partial_s^k \psi}{s=0} = (-1)^k \frac{1}{t} \langle A^k \rangle_c$$

$\langle A^k \rangle_c$ is the k^{th} cumulant of A .

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

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* Starting from $\langle e^{-sA} \rangle = \int dA P(A,t) e^{-sA}$ $A = at$
 $e^{st\Psi(s)} \sim \int da e^{\underbrace{t(\Pi(a)-sa)}_{\text{in the } t \rightarrow \infty \text{ limit, one may}}}$
 use the saddle point theorem

One has

$$\boxed{\Psi(s) = \max_a (\Pi(a) - sa)}$$

Ψ and Π are
Legendre transformed.

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

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

- * If $a^*(s)$ is the a where the \max_a is reached
 or if $s^*(a) = s - \frac{\min_s}{a}$ one says that
 $a^*(s)$ and s are conjugated
- * $\boxed{s \text{ plays a role similar to the temperature } \beta:}$
 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 Θ in the 's-state' writes

$$\langle \Theta(e) \rangle_s \equiv \frac{\langle \Theta(e(t)) e^{-sA} \rangle}{\langle e^{-sA} \rangle} = \frac{1}{\langle e^{-sA} \rangle} \sum_{e,A} \tilde{P}(e,A,t) \Theta(e) e^{-sA} \quad A = at$$

$$\sim \frac{\int da \sum_e \tilde{P}(e,a) \Theta(e) e^{t(\Pi(a)-sa)}}{\int da e^{t(\Pi(a)-sa)}} \quad \begin{array}{l} \text{the saddle is reached} \\ \text{at the same value} \\ a = a^*(s) \end{array}$$

$$\langle \Theta(e) \rangle_s \sim \left\{ \begin{array}{l} \tilde{P}(e, a^*(s)) \Theta(e) : e \\ \text{mean value in the} \\ \text{s-state, of } \Theta(e) \\ \text{at final time} \end{array} \right\}$$

$$\boxed{\langle \Theta(e) \rangle_s \stackrel{t \rightarrow \infty}{=} \langle \Theta(e) \rangle \Big|_{A=a^*(s)t}}$$

Mean value of $\Theta(e)$ at final time
for histories with $A = a^*(s)t$

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

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$$\left\langle \Theta(c) \right\rangle \Big|_{A=a(t)} \stackrel{t \rightarrow \infty}{=} \left\langle \Theta(c) \right\rangle_{s=s^*(a)}$$

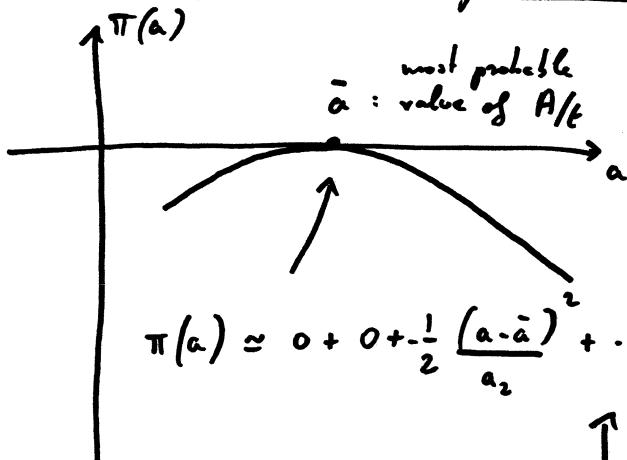
Mean value of $\Theta(c)$ at final time
for histories with a value $A=a(t)$
of the observable A

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

$$\left\langle \Theta(c) \right\rangle_s = \frac{\left\langle \Theta(c(t)) e^{-sA} \right\rangle}{\left\langle e^{-sA} \right\rangle}$$

In other words, to characterize the value of the observable Θ in histories with a value $A=a(t)$ of the observable A , one has to compute $\left\langle \Theta(c) \right\rangle_s$ at $s=s^*(a)$.

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



$$\bar{a} = a_1 = \frac{\langle A \rangle}{t} \quad \text{mean value}$$

$$a_2 = \frac{\langle A^2 \rangle_c}{t} - \bar{a}^2 \quad \text{second cumulant}$$

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

$\psi(s)$

For a reference on large deviation,
Laplace & Legendre transforms,
see Hugo Touchette

The large deviation approach to
statistical mechanics

Physics Reports 478 1 (2009)

$$\psi(s) \approx -s\bar{a}_1 + \frac{1}{2} \bar{a}_2 s^2 + \dots$$

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

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