Statistical physics of long dynamical trajectories for a system in contact with several thermal reservoirs

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Abstract

For a system in contact with several reservoirs \( r \) at different inverse-temperatures \( \beta_r \), we describe how the Markov jump dynamics with the generalized detailed balance condition can be analyzed via a statistical physics approach of dynamical trajectories \([C(t)]_{0 \leq t \leq T}\) over a long time interval \( T \to +\infty \). The relevant intensive variables are the time-empirical density \( \rho(C) \), that measures the fractions of time spent in the various configurations \( C \), and the time-empirical jump densities \( k_r(C',C) \), that measure the frequencies of jumps from configuration \( C \) to configuration \( C' \) when it is the reservoir \( r \) that furnishes or absorbs the corresponding energy difference \( (E(C') - E(C)) \).

Keywords: non-equilibrium, thermal reservoirs, large deviations, long dynamical trajectories

1. Introduction

The statistical physics of equilibrium is based on the ergodic principle, that states the equivalence between the ‘time average’ of any observable \( O \) over a sufficiently long time dynamical trajectory \([C(t)]_{0 \leq t \leq T}\) and an appropriate ‘ensemble average’ over some probability \( P_{\text{ensemble}}(C) \) over the configurations \( C \)

\[
\int_0^T dt \langle O(C(t)) \rangle \underset{T \to \infty}{\approx} \sum_C O(C) P_{\text{eq}}(C). \tag{1}
\]

The various equilibrium ensembles are adapted to the physical conditions one is interested in: for instance, the microcanonical ensemble is appropriate for isolated systems conserving the total energy, while the canonical ensemble is tailored for systems in contact with a single
thermal reservoir, with relations between the two in the thermodynamic limit (see the short reminder in the appendix).

For non-equilibrium systems, it is thus natural to also base the analysis on time-averages over long dynamical trajectories. However, it turns out that the time-empirical density measures the fractions of time spent in the various configurations $C$ (direct analog of equation (1))

$$\rho(C) \equiv \frac{1}{T} \int_0^T dt \delta_{C(t),C}$$

is not sufficient, and has to supplemented by the empirical flows between configurations, i.e. for instance the jump densities from configurations $C$ to configuration $C'$ for the case of jump processes

$$k(C',C) \equiv \frac{1}{T} \sum_{t \in [0,T]; \delta C(t^+) \neq \delta C(t^-)} \delta_{C(t^+),C'} \delta_{C(t^-),C}.$$  

For Markovian dynamics, the joint probability of the empirical density and of the empirical flows is known to follow a very general large deviation form with respect to the large time $T$ (the general theory of large deviations is summarized in the reviews [1–3] with applications to various fields). In addition, the corresponding rate function (known as ‘Level 2.5’ in the classification of Large Deviations) is explicit within various frameworks, namely for discrete-time/discrete-space Markov chains [3–6], for continuous-time/discrete-space Markov jump processes [4, 7–13] as well as for continuous-time/continuous-space diffusion processes [10, 13–15]. Moreover, this ‘Level 2.5’ formulation allows to reconstruct any time-additive observable of the dynamical trajectory via its decomposition in terms of the empirical densities and of the empirical flows, and is thus closely related to the studies focusing on the generating functions of time-additive observables via deformed Markov operators, that have attracted a lot of interest recently in various models [16–29]. More generally, the idea that the theory of large deviations is the appropriate language to analyze non-equilibrium dynamics has been emphasized from various perspectives (see the reviews [16, 25, 26, 30–33] and the PhD Theses [4, 13, 34, 35] or HDR Thesis [36]).

In the present paper, we focus on the Markov jump dynamics of a system in contact with $R > 1$ thermal reservoirs $r = 1, \ldots, R$ as described by the generalized-detailed-balance condition. We wish to keep track in the jump densities $k(C',C)$ from $C$ to $C'$ of equation (3) of the reservoir $r$ that is responsible for the jump, i.e. of the reservoir $r$ that furnishes or absorbs the energy difference between the initial and final configuration of the system

$$k(C',C) = \sum_{r=1}^R k_r(C',C).$$

The goal is to extend the standard ‘Level 2.5’ large deviation framework mentioned above [4, 7–13] to this finer analysis of the dynamical trajectories that contains all the information on the heats exchanges with the various reservoirs.

The paper is organized as follows. In section 2, we recall the Markov jump dynamics framework to describe a system in contact with several reservoirs, and we describe the relevant time-empirical observables of dynamical trajectories. In section 3, the probability of individual trajectories is analyzed in order to stress the important notion of intensive action of a function of the relevant empirical observables. Section 4 is devoted to the Boltzmann entropy of the set of dynamical trajectories with given empirical observables. In section 5, the large deviations properties of the joint probability of the time-empirical density and jump densities
are discussed. Finally the link with the statistics of other time-empirical additive observables is described in section 6. Section 7 summarizes our conclusions. The appendix contains a short summary of the statistical physics of equilibrium in order to make more explicit the comparison with the equations of the text concerning non-equilibrium.

2. Markov jump dynamics for a system in contact with several reservoirs

2.1. Markov jump dynamics in configuration space

Since our goal is to remain at the same level of generality as the statistical physics theory of equilibrium (see the short reminder in the appendix), we will consider that the system can be in a certain number of configurations $C$ of energies $E(C)$. The Markov jump dynamics between the configurations of the system is defined by the transition rates $w(C', C)$ from $C$ to $C' \neq C$ for the dynamical elementary moves $C' \leftarrow C$ that are allowed in the model (for instance single-spin-flip dynamics in spin models). It is convenient to introduce the following notation for the total rate out of configuration $C$

$$w^{\text{out}}(C) \equiv \sum_{C' \neq C} w(C', C). \quad (5)$$

Then the probability $P_t(C)$ to be in configuration $C$ at time $t$ then evolves with the master equation

$$\frac{\partial}{\partial t} P_t(C) = \sum_{C' \neq C} w(C, C') P_t(C') - w^{\text{out}}(C) P_t(C). \quad (6)$$

As recalled in the Introduction, many studies have been devoted to the analysis of dynamical trajectories within this general framework.

2.2. Generalized detailed balance

Here we wish to focus on the case where the system is in contact with $R$ reservoirs $r = 1, 2, ..., R$ characterized by different inverse temperature $\beta_r$. We will consider the generalized-detailed-balance condition in the following form (see the review [37] and references therein):

(i) the transition rate $w(C', C)$ is given by the sum over the contributions $w_r(C', C)$ of the reservoirs that allow this transition (i.e. when the reservoir $r$ is able to furnish or absorb the energy difference between the initial and final configuration of the system, otherwise the rate is zero)

$$w(C', C) = \sum_{r=1}^{R} w_r(C', C)$$

so that equation (5) becomes

$$w^{\text{out}}(C) \equiv \sum_{C' \neq C} \sum_{r=1}^{R} w_r(C', C). \quad (8)$$

(ii) the transitions rates $w_r(C', C)$ associated to the reservoir $r$ satisfy the usual detailed-balance condition at the inverse temperature $\beta_r$ of the reservoir $r$
\[ w_r(C, C') e^{-\beta E(C')} = w_r(C', C) e^{-\beta E(C)}. \] (9)

In particular, if the reservoir \( r \) were alone, the stationary state of the system would be the canonical distribution at inverse temperature \( \beta_r \).

Since equation (9) only fixes the ratios
\[ \frac{w_r(C', C)}{w_r(C, C')} = e^{-\beta_r [E(C') - E(C)]} \] (10)
but not the products \( [w_r(C', C) w_r(C, C')] \), various choices for the transition rates exist in the literature. The simplest choice from a technical point of view corresponds to the case
\[ w_r(C', C) w_r(C, C') = 1 \] [38–40]
while a more physical choice (in order to have transition rates that remain finite in the zero-temperature limit \( \beta_r = \infty \)) is given by the Glauber transition rates [41]
\[ w^\text{Glauber}_r(C', C) = \frac{e^{-\frac{\beta_r}{2} [E(C') - E(C)]}}{e^{\frac{\beta_r}{2} [E(C') - E(C)]} + e^{-\frac{\beta_r}{2} [E(C') - E(C)]}} = \frac{e^{-\frac{\beta_r}{2} [E(C') - E(C)]}}{2 \cosh \left( \frac{\beta_r}{2} [E(C') - E(C)] \right)}. \] (12)

2.3. Relevant time-empirical observables for the dynamical trajectories and their constraints

As recalled in the introduction, the fundamental time-empirical observables for a Markov jump dynamical trajectory \( \mathcal{C}(t) \) over the large-time interval \( T \) are [4, 7–13].

(a) the time-empirical density \( \rho(C) \) of equation (2) with the normalization
\[ \sum_C \rho(C) = 1. \] (13)
(b) the time-empirical jump densities \( k(C', C) \) of equation (3), with the following constraint for each \( C \)
\[ \sum_{C'} k(C', C) = \sum_{C'} k(C, C'). \] (14)

This quantity represents the number of sojourns in configuration \( C \) over the time \( T \), as computed from the number of departures from \( C \) or from the number of arrivals into \( C \) (for large \( T \), one may neglect the boundary terms where \( C \) happens to be the initial configuration at \( t = 0 \) or the final configuration at \( t = T \)).

In the present paper, we keep track of the reservoir \( r \) in the jump densities (equation (4)) so that it is convenient to introduce the following notation to summarize the constraints of equations (13) and (14) with the decomposition of equation (4),
\[ \mathcal{D} [\rho(\cdot), k(\cdot \cdot \cdot)] = \left[ \delta(1 - \sum_C \rho(C)) \right] \prod_C \delta \left( \sum_{C' \neq C} \sum_{r=1}^R k_r(C', C) - \sum_{C' \neq C} \sum_{r=1}^R k_r(C, C') \right) \] (15)
(the notation \( \delta(Y) \) represents the discrete Kronecker symbol \( \delta_{0,Y} \) but has been chosen here for better readability of the argument \( Y \)).
2.4. Replacing the empirical jump densities \( k_i(C', C) \) by the empirical activities \( a_i(C', C) \) and currents \( j_i(C', C) \)

For each given pair of configurations \( C \) and \( C' \), one can choose some order \( C < C' \), and decide to replace the two time-empirical jump densities \( k_i(C', C) \) and \( k_i(C, C') \) by their antisymmetric and symmetric parts called respectively the time-empirical current \( j_i(C', C) \) and the time-empirical activity \( a_i(C', C) \) [9]

\[
\begin{align*}
  j_i(C', C) &= k_i(C', C) - k_i(C, C') = -j_i(C, C') \\
  a_i(C', C) &= k_i(C', C) + k_i(C, C') = a_i(C, C').
\end{align*}
\]

(16)

This decomposition is more appropriate when one wishes to analyze the properties upon time reversal [9]: if the dynamical trajectory \( [C(t)]_{0 \leq t \leq T} \) is characterized by the time empirical observables \( (\rho(C), a_i(C', C), j_i(C', C)) \), then the time-reversed trajectory \( [C_{\text{Backward}}(t)]_{0 \leq t \leq T} \) defined by

\[
C_{\text{Backward}}(t) \equiv C(T - t)
\]

is characterized by the same empirical density and empirical activities, but by opposite empirical currents

\[
\begin{align*}
  \rho_{\text{Backward}}(C) &= \rho(C) \\
  a_{r,\text{Backward}}(C', C) &= a_i(C', C) = a_i(C', C) \\
  j_{r,\text{Backward}}(C', C) &= j_i(C', C) = -j_i(C', C).
\end{align*}
\]

(18)

Via this change of variables, the constraints of equation (15) do not contain the activities \( a_i(C', C) \), i.e. they involve only the empirical density \( \rho(C) \) and the empirical currents \( j_i(C', C) \)

\[
\mathcal{D} [\rho(\cdot), j(\cdot, \cdot)] = \left[ \delta(1 - \sum_C \rho(C)) \right] \prod_C \delta \left( \sum_{C' \neq C} \sum_{i=1}^R j_i(C', C) \right).
\]

(19)

2.5. Link with thermodynamics: intensive time-empirical heats \( q_r \) furnished by the various reservoirs

During the time interval \( [0, T] \), each reservoir \( r \) will furnish some extensive empirical heat \( (Tq_r) \), either positive or negative, where the intensive time-empirical heat \( q_r \) reads in terms of the empirical jump densities \( k_i(C', C) \), or in terms of the empirical currents \( j_i(C', C) \) introduced above

\[
q_r = \sum_C \sum_{C' \neq C} \left[ E(C') - E(C) \right] k_i(C', C)
\]

\[
= \sum_C \sum_{C' \neq C} \left[ E(C') - E(C) \right] \frac{j_i(C', C)}{2} = \sum_C \sum_{C' > C} \left[ E(C') - E(C) \right] j_i(C', C).
\]

(20)

Besides this decomposition into pairs of configurations, it is useful to write the following alternative expression with respect to the energy of a single configuration

\[
q_r = - \sum_C E(C) \left( \sum_{C' \neq C} j_i(C', C) \right)
\]

(21)
in order to make obvious that the sum over the reservoirs vanishes as a consequence as the constraint of equation (19)

\[
\sum_{r=1}^{R} q_r = - \sum_{C} E(C) \left( \sum_{C' \neq C}^{\ast} \sum_{r=1}^{R} j_r(C', C) \right) = 0
\]

(22)
as it should since the energy cannot accumulate in the system. This is the empirical version of the first principle of thermodynamics at the level of a very long dynamical trajectory (see the review [37] and references therein).

3. Probabilities of dynamical trajectories: notion of intensive action

3.1. Extensive action \( A (|C(t)|_{0 \leq t \leq T}) \) of a dynamical trajectory \(|C(t)|_{0 \leq t \leq T}\)

A dynamical trajectory \( C(t) \) on the time interval \( 0 \leq t \leq T \) corresponds to a certain number \( M \geq 0 \) of jumps \( m = 1, \ldots, M \) occurring at times \( 0 < t_1 < \ldots < t_M < T \) between the successive configurations \( \{C_0 \rightarrow C_1 \rightarrow C_2 \rightarrow \ldots \rightarrow C_M\} \) that are visited between these jumps. Since we wish to keep track of the reservoirs labels \( r_m \) that are responsible of the jumps at times \( t_m \), it will be convenient to denote this trajectory as

\[
[C(t)]_{0 \leq t \leq T} = \left[ C_0; (t_1, r_1); C_1; (t_2, r_2); C_2; (t_3, r_3); \ldots; C_{M-1}; (t_M, r_M); C_M \right].
\]

(23)
The probability of this trajectory conditioned to start in configuration \( C_0 \) reads in terms of the transitions rates

\[
P \left( [C(t)]_{0 \leq t \leq T} \right) = \sum_{C_0} \prod_{m=1}^{M} \prod_{r_m} \prod_{C_{m+1} \neq C_m} \prod_{C_{m-1} \neq C_m} \prod_{C_{m+2} \neq C_m} \prod_{C_{m-2} \neq C_m} e^{-\tau w_{m}(C)} w_{m}(C_{m+1}) w_{m}(C_{m-1}) w_{m}(C_{m+2}) w_{m}(C_{m-2}) \ldots
\]

(24)
The normalization over all possibilities trajectories on \([0, T]\) for a fixed initial configuration \( C_0 \) involves the sum over the number \( M \) of jumps, the sum over the \( M \) configurations \( \{C_1, \ldots, C_M\} \) where \( C_m \) has to be different from \( C_{m-1} \), the sum over the possible reservoirs \( r_m \) for each jump, the integration over the jump times with the measure \( dt_1 \ldots dt_M \) and the constraint \( 0 < t_1 < \ldots < t_M < T \)

\[
1 = \sum_{M=0}^{\infty} \int_{0}^{\infty} dt_1 \int_{0}^{t_1} \int_{0}^{t_2} \int_{0}^{t_2} \ldots \int_{0}^{t_M} dt_M = \int_{0}^{t_1} \int_{0}^{t_2} \int_{0}^{t_2} \ldots \int_{0}^{t_M} dt_M
\]

(25)
The trajectory probability of equation (24) can be rewritten

\[
P \left( [C(t)]_{0 \leq t \leq T} \right) = e^{-A([C(t)]_{0 \leq t \leq T})}
\]

(26)
in terms of the action

\[
A ([C(t)]_{0 \leq t \leq T}) = \int_{0}^{T} dt w(C(t)) - \sum_{m=1}^{M} \left[ w_{m}(C(t_m^{+}), C(t_m^{+})) - w_{m}(C(t_m^{-}), C(t_m^{-})) \right]
\]

(27)

while the normalization of equation (25) can be formally rewritten as
where the sum over all the possible trajectories \([\mathcal{C}(t)]_{0 \leq t \leq T}\) is meant to replace the more explicit first line of equation (25).

### 3.2. Intensive action \(A[\rho(\cdot), k(\cdot, \cdot)]\) of a dynamical trajectory in terms of its intensive empirical observables

The action of the dynamical trajectory of equation (27) is extensive with respect to the large time-interval \(T\)

\[
A ([\mathcal{C}(t)]_{0 \leq t \leq T}) = TA[\rho(\cdot), k(\cdot, \cdot)]
\]  

(29)
and the corresponding intensive action \(A[\rho(\cdot), k(\cdot, \cdot)]\) reads in terms of the empirical density \(\rho(C)\) and of the empirical jump densities \(k_i(C', C)\) of the trajectory

\[
A[\rho(\cdot), k(\cdot, \cdot)] = \sum_{C} \rho(C)w^{\text{out}}(C) - \sum_{C'} \sum_{C' \neq C} \sum_{r=1}^{R} k_r(C', C) \ln (w_r(C', C))
\]

\[
= \sum_{C} \sum_{C' \neq C} \sum_{r=1}^{R} \left[ w_r(C', C) \rho(C) - k_r(C', C) \ln (w_r(C', C)) \right]
\]  

(30)
where we have used equation (8) in the second line in order to stress the decomposition of the action onto the contributions of the reservoir dependent transition rates \(w_r(C', C)\).

In summary, the probability of an individual trajectory \([\mathcal{C}(t)]_{0 \leq t \leq T}\) of equation (24) decays exponentially in \(T\)

\[
P ([\mathcal{C}(t)]_{0 \leq t \leq T}) \xrightarrow{T \rightarrow +\infty} e^{-TA[\rho(\cdot), k(\cdot, \cdot)]}
\]  

(31)
and the intensive action \(A[\rho(\cdot), k(\cdot, \cdot)]\) only depends on the time-empirical density \(\rho(C)\) and the time-empirical jump densities \(k_i(C', C)\) of the trajectory, that are thus the only relevant intensive observables of the trajectory.

This statement can be considered as the qualitative analog of equation (A.9) concerning the canonical ensemble, where the probability of a configuration involves the extensive energy \(E(C)\) in the exponential, so that the appropriate intensive variable in the large \(N\) limit is the intensive energy \(\varepsilon = \frac{E(C)}{N}\) of equation (A.5).

### 3.3. Intensive action \(A[\rho(\cdot), j(\cdot, \cdot), a(\cdot, \cdot)]\) in terms of the empirical currents and activities

Via the change of variables of equation (16), the intensive action of equation (30) becomes in terms of the empirical currents \(j_i(C', C)\) and empirical activities \(a_i(C', C)\)

\[
A[\rho(\cdot), j(\cdot, \cdot), a(\cdot, \cdot)] = \sum_{C} \rho(C)w^{\text{out}}(C)
\]

\[
- \sum_{C} \sum_{C' \neq C} \sum_{r=1}^{R} \left[ \frac{\alpha_r(C', C)}{2} \ln [w_r(C', C)w_r(C, C')] + \frac{\beta_r(C', C)}{2} \ln \left( \frac{w_r(C', C)}{w_r(C, C')} \right) \right]
\]  

\[
= \sum_{C} \sum_{C' \neq C} \sum_{r=1}^{R} \left[ \frac{w_r(C', C) \rho(C) - k_r(C', C) \ln (w_r(C', C))}{w_r(C, C')} \right]
\]  

(32)
In particular, one sees why the simplest choice of equation (11) for the transitions rates is also the simplest choice from the point of view of this intensive action, since the activities
3.4. Entropy production in terms of the time-empirical heats $q_r$ of the trajectory

The probability of the backward trajectory $C^{\text{Backward}}(t) \equiv C(T - t)$ conditioned to start in configuration $C_T$ involves the backward empirical observables of equation (18)

$$P\left(\{C^{\text{Backward}}(t)\}_{0 \leq t \leq T} \right) \underset{T \to +\infty}{=} e^{-T\sum_{r=1}^{R} \beta_r \sum_{C' > C} j_r(C', C) \ln \left( \frac{w_r(C', C)}{w_r(C, C')} \right)}.$$  

(33)

As a consequence, the ratio between the probability of the initial trajectory $[C(t)]_{0 \leq t \leq T}$ and the probability of the backward trajectory $[C^{\text{Backward}}(t)]_{0 \leq t \leq T}$ only involves the contributions related to the empirical currents $j_r(C', C)$

$$\frac{P\left(\{C(t)\}_{0 \leq t \leq T} \right)}{P\left(\{C^{\text{Backward}}(t)\}_{0 \leq t \leq T} \right)} = e^{-T \sum_{r=1}^{R} \beta_r \sum_{C'} \sum_{C > C'} j_r(C', C) \ln \left( \frac{w_r(C', C)}{w_r(C, C')} \right)}.$$  

(34)

Since the generalized-detailed condition of equation (9) fixes the ratios of equations (10) and (34) reduces to

$$\frac{P\left(\{C(t)\}_{0 \leq t \leq T} \right)}{P\left(\{C^{\text{Backward}}(t)\}_{0 \leq t \leq T} \right)} = e^{-T \sum_{r=1}^{R} \beta_r \sum_{C'} \sum_{C > C'} j_r(C', C) [E(C') - E(C) \right] - T \sum_{r=1}^{R} \beta_r q_r}.$$  

(35)

and thus only involves the intensive time-empirical heats $q_r$ introduced in equation (20) and the inverse-temperatures $\beta_r$ of the reservoirs. The intensive entropy production $\sigma_{\text{prod}}^{\text{int}}([C(t)]_{0 \leq t \leq T})$ associated to the long dynamical trajectory $[C(t)]_{0 \leq t \leq T}$, that measures its irreversibility by the comparison with the probability of the backwards trajectory $[C^{\text{Backward}}(t)]_{0 \leq t \leq T}$

$$\sigma_{\text{prod}}^{\text{int}}([C(t)]_{0 \leq t \leq T}) \equiv \frac{1}{T} \ln \left( \frac{P\left(\{C(t)\}_{0 \leq t \leq T} \right)}{P\left(\{C^{\text{Backward}}(t)\}_{0 \leq t \leq T} \right)} \right).$$  

(36)

thus only involves the weighted sum of the empirical heats $q_r$ of the dynamical trajectory

$$\sigma_{\text{prod}}^{\text{int}}([C(t)]_{0 \leq t \leq T}) = - (A[\rho(.), j(\ldots), a(\ldots)] - A[\rho(\ldots), -j(\ldots), a(\ldots)]) = - \sum_{r=1}^{R} \beta_r q_r.$$  

(37)

with coefficients given by the inverse-temperatures $\beta_r$ of the reservoirs (while the non-weighted sum over the $q_r$ vanishes (equation (22)): this is the empirical version of the second principle of thermodynamics at the level of a very long dynamical trajectory (see the review [37] and references therein).

3.5. Relation between the action $A[\rho(.), k(\ldots)]$ and the Kolmogorov-Sinai entropy $h^{KS}$

In information theory and in stochastic thermodynamics, the ‘surprise’ $S([C(t)]_{0 \leq t \leq T})$ to observe as outcome the trajectory $[C(t)]_{0 \leq t \leq T}$ when its probability is $P\left(\{C(t)\}_{0 \leq t \leq T} \right)$ is defined as minus the logarithm of this probability, so that it coincides with the extensive action $A([C(t)]_{0 \leq t \leq T})$ of equation (26)
\( \mathcal{S} ([C(t)]_{0 \leq t \leq T}) \equiv - \ln [\mathcal{P} ([C(t)]_{0 \leq t \leq T})] = A ([C(t)]_{0 \leq t \leq T}) \xrightarrow{T \to +\infty} TA[\rho(\cdot), k(\cdot) ]. \) \hfill (38)

As a consequence, the intensive action \( A[\rho(\cdot), k(\cdot)] \) of a dynamical trajectory represents the ‘intensive surprise’ of the trajectory, and depends only of the empirical observables \([\rho(\cdot), k(\cdot)] \) of this trajectory.

The Shannon entropy \( S^\text{dyn}_T \) associated to the trajectories probabilities \( \mathcal{P} ([C(t)]_{0 \leq t \leq T}) \) then corresponds to the average of the surprise \( \mathcal{S} ([C(t)]_{0 \leq t \leq T}) \) over the trajectories, or to the average of the extensive action over the trajectories

\[
S^\text{dyn}_T \equiv - \sum_{[C(t)]_{0 \leq t \leq T}} \mathcal{P} ([C(t)]_{0 \leq t \leq T}) \ln [\mathcal{P} ([C(t)]_{0 \leq t \leq T})] = \langle A ([C(t)]_{0 \leq t \leq T}) \rangle \xrightarrow{T \to +\infty} T A[\rho(\cdot), k(\cdot) ].
\] \hfill (39)

The Kolmogorov-Sinai entropy \( h^{KS} \) defined as the coefficient of growth with respect to the time \( T \) of equation (39) thus corresponds to the average of the intensive action \( A[\rho(\cdot), k(\cdot)] \)

\[
h^{KS} \equiv \lim_{T \to +\infty} \left( \frac{S^\text{dyn}_T}{T} \right) = \langle A[\rho(\cdot), k(\cdot)] \rangle \] \hfill (40)

over the dynamical trajectories, that will be translated into an average over the empirical variables in the next sections.

4. Entropy of the set of trajectories with given empirical observables

4.1. Measure of the set of dynamical trajectories with given empirical observables

Since all the individual dynamical trajectories \([C(t)]_{0 \leq t \leq T} \) that have the same empirical observables \( \rho(C) \) and \( k(C, C') \) have the same probability given by equation (31), one can rewrite the normalization of equation (28) as a sum over these empirical observables

\[1 = \sum_{\rho(\cdot)} \sum_{k(\cdot)} \Omega_T[\rho(\cdot), k(\cdot)] e^{-TA[\rho(\cdot), k(\cdot)]} \] \hfill (41)

where \( \Omega_T[\rho(\cdot), k(\cdot)] \) represents the measure of the set of the dynamical trajectories associated to given values of these empirical observables. So besides the empirical constraints \( \mathcal{D} [\rho(\cdot), k(\cdot)] \) of equation (15), one expects an exponential growth with respect to \( T \)

\[
\Omega_T[\rho(\cdot), k(\cdot)] \xrightarrow{T \to +\infty} \mathcal{D} [\rho(\cdot), k(\cdot)] e^{TS^{\text{Boltzmann}}[\rho(\cdot), k(\cdot)]} \] \hfill (42)

where \( S^{\text{Boltzmann}}[\rho(\cdot), k(\cdot)] \) represents the Boltzmann intensive entropy of the set of trajectories with given intensive empirical observables.

This statement can be considered as the qualitative analog of equation (A.7) concerning the microcanonical ensemble, where the number of configurations with a given intensive energy \( e \) grows exponentially in \( N \) with a coefficient given by the Boltzmann intensive entropy \( s(e) \).

Then equation (41) becomes

\[1 \xrightarrow{T \to +\infty} \sum_{\rho(\cdot)} \sum_{k(\cdot)} \mathcal{D} (\rho(\cdot), k(\cdot)) e^{T[S^{\text{Boltzmann}}[\rho(\cdot), k(\cdot)] - A[\rho(\cdot), k(\cdot)]]}, \] \hfill (43)
4.2. Analysis for typical values of the empirical observables

The typical values of the empirical observables read
\[ \rho^{\text{typ}}(C) = \rho_C(C) \]
\[ k^{\text{typ}}_r(C', C) = w_r(C', C)\rho_C(C) \]
\[ (44) \]
in terms of the normalized stationary solution \( \rho_C(C) \) of the master equation (6), that should satisfy for all \( C \)
\[ 0 = \sum_{C' \neq C} \sum_{r=1}^{R} w_r(C, C')\rho_C(C') - w^{\text{out}}_r(C)\rho_C(C) = \sum_{C' \neq C} \sum_{r=1}^{R} w_r(C, C')\rho_C(C') - \sum_{r=1}^{R} w_r(C', C)\rho_C(C) \]
\[ (45) \]

For the typical values of equation (44), the exponential behavior in \( T \) of equation (43) should exactly vanish, i.e. the entropy \( S^{\text{Boltzmann}}[\rho^{\text{typ}}(\cdot), k^{\text{typ}}(\cdot, \cdot)] \) should exactly compensate the action \( A[\rho^{\text{typ}}(\cdot), k^{\text{typ}}(\cdot, \cdot)] \) of equation (30)
\[ S^{\text{Boltzmann}}[\rho^{\text{typ}}(\cdot), k^{\text{typ}}(\cdot, \cdot)] = A[\rho^{\text{typ}}(\cdot), k^{\text{typ}}(\cdot, \cdot)] = \sum_{C} \sum_{C' \neq C} \sum_{r=1}^{R} \left[ w_r(C', C)\rho^{\text{typ}}(C') - k^{\text{typ}}_r(C', C) \ln \left( w_r(C', C) \right) \right]. \]
\[ (46) \]

4.3. Intensive entropy for generic values of the intensive empirical observables

To obtain the intensive entropy \( S^{\text{Boltzmann}}[\rho(\cdot), k(\cdot, \cdot)] \) for generic values \( \rho(C) \) and \( k_r(C', C) \) of the intensive empirical observables, one just needs to introduce the effective transitions rates \( w^{\text{eff}}_r(C', C) \) that would make the empirical values typical for this effective modified dynamics, namely (equation (44))
\[ w^{\text{eff}}_r(C', C) = \frac{k_r(C', C)}{\rho(C)}. \]
\[ (47) \]

Then one may use equation (46) with these effective rates to obtain
\[ S^{\text{Boltzmann}}[\rho(\cdot), k(\cdot, \cdot)] = \sum_{C} \sum_{C' \neq C} \sum_{r=1}^{R} \left[ w^{\text{eff}}_r(C', C)\rho(C) - k_r(C', C) \ln \left( w^{\text{eff}}_r(C', C) \right) \right] = \sum_{C} \sum_{C' \neq C} \sum_{r=1}^{R} \left[ k_r(C', C) - k_r(C', C) \ln \left( \frac{k_r(C', C)}{\rho(C)} \right) \right]. \]
\[ (48) \]

As explained after equation (42), this intensive Boltzmann entropy \( S^{\text{Boltzmann}}[\rho(\cdot), k(\cdot, \cdot)] \) of the set of dynamical trajectories that are characterized by the intensive empirical observables \( [\rho(\cdot), k(\cdot, \cdot)] \) is the analog of the intensive microcanonical entropy \( s(e) \) as a function of the intensive energy \( e \) in the microcanonical ensembles (equation (A.7)). Here the explicit expression of equation (48) was derived by the method of ‘change of measure’ which is very common in the whole field of large deviations, since it allows to obtain directly the result without any actual computation (i.e. one does not need to use combinatorial methods to count the appropriate configurations).
4.4. Alternative expression of the intensive entropy in terms of empirical activities and currents

Via the change of variables of equation (16), the intensive entropy of equation (48) becomes in terms of the empirical currents and activities

$$S_{\text{Boltzmann}}[\rho(\cdot), j(\cdot), a(\cdot)] = \sum_c \sum_{C' > C} R \left[ a_r(C', C) - \frac{a_r(C', C)}{2} \ln \left( \frac{a_r^2(C', C) - j_r(C', C)}{4 \rho(C') \rho(C)} \right) \right] - \frac{j_r(C', C)}{2} \ln \left( \frac{a_r(C', C) + j_r(C', C)}{a_r(C', C) - j_r(C', C)} \times \frac{\rho(C')}{\rho(C)} \right).$$

(49)

For the next section, it is interesting to note here the asymmetry property with respect to the empirical currents $j_r(C', C)$, for any values of the empirical density $\rho(C)$ and activities $a_r(C', C)$ (equation (18))

$$S_{\text{Boltzmann}}[\rho(\cdot), j(\cdot), a(\cdot)] - S_{\text{Boltzmann}}[\rho(\cdot), -j(\cdot), a(\cdot)] = \sum_c \sum_{C' > C} R \left[ \frac{j_r(C', C)}{2} \ln \left( \frac{\rho(C')}{\rho(C)} \right) \right].$$

(50)

5. Large deviations for the probability of intensive empirical densities

5.1. Joint probability of the empirical density $\rho(C)$ and empirical jump densities $k_r(C', C)$

Equation (41) means that the probability $P_T[\rho(\cdot), k(\cdot)]$ of the empirical observables $\rho(\cdot)$ and $k(\cdot)$ over the set of dynamical trajectories, with the normalization

$$1 = \sum_{\rho(\cdot)} \sum_{k(\cdot)} P_T[\rho(\cdot), k(\cdot)]$$

(51)

is given by

$$P_T[\rho(\cdot), k(\cdot)] = \Omega_T[\rho(\cdot), k(\cdot)] e^{-TA[\rho(\cdot), k(\cdot)]}.$$  

(52)

Plugging equations (42) into (52) yields that the probability $P_T[\rho(\cdot), k(\cdot)]$ displays the large deviation form

$$P_T[\rho(\cdot), k(\cdot)] \underset{r \to +\infty}{\sim} D[\rho(\cdot), k(\cdot)] e^{-T[I[\rho(\cdot), k(\cdot)]]}$$

(53)

where the rate function $I[\rho(\cdot), k(\cdot)]$ is simply given by the difference between the action $A[\rho(\cdot), k(\cdot)]$ of equation (30) and the entropy $S_{\text{Boltzmann}}[\rho(\cdot), k(\cdot)]$ of equation (48)

$$I[\rho(\cdot), k(\cdot)] = A[\rho(\cdot), k(\cdot)] - S_{\text{Boltzmann}}[\rho(\cdot), k(\cdot)]$$

(54)

This formula is thus a very direct generalization in the presence of several reservoirs $r$ of the standard large deviation rate at level 2.5 for Markov jump processes. [4, 7–13]. It vanishes only for the typical values of equation (44)

$$I[\rho^{\text{yp}}(\cdot), k^{\text{yp}}(\cdot)] = 0$$

(55)
while the positive values \( I[\rho(\cdot), k(\cdot)] > 0 \) for non-typical empirical observables \([\rho(\cdot), k(\cdot)] \neq [\rho^p(\cdot), k^p(\cdot)]\) measure how rare it is to observe them for large time \( T \).

Here the qualitative analogy with equilibrium is that the intensive free-energy of the canonical ensemble contains an energetic contribution and an entropic contribution (equation (A.15)).

### 5.2. Joint probability of the empirical density, empirical activities and currents

Via the change of variables of equation (16), the probability of equation (53) can be translated into the joint probability of the empirical density, the empirical activities and the empirical currents

\[
P_T[\rho(\cdot), j(\cdot), a(\cdot)] \sim \mathcal{D}[\rho(\cdot), j(\cdot), a(\cdot)] e^{-T I[\rho(\cdot), j(\cdot), a(\cdot)]} \tag{56}
\]

where the constraints \( \mathcal{D}[\rho(\cdot), j(\cdot), a(\cdot)] \) are given in terms of the currents in equation (19), while the rate function reads (equation (54))

\[
I[\rho(\cdot), j(\cdot), a(\cdot)] = \sum_{C} \sum_{C'} \sum_{r=1}^{R} \left[ \sum_{c} a_r(C', C) \rho(C) + w_r(C, C') \rho(C') - a_r(C', C) \right] 
+ \sum_{C} \sum_{C'} \sum_{r=1}^{R} \left[ \frac{a_r(C', C) - j_r(C', C)}{2} \ln \left( \frac{a_r(C', C)}{j_r(C', C)} \right) \right] 
+ \frac{j_r(C', C)}{2} \ln \left( \frac{a_r(C', C)}{j_r(C', C)} \times \frac{w_r(C, C') \rho(C')}{w_r(C, C') \rho(C)} \right). \tag{57}
\]

An important property of this form of the rate function is that the ratio of the probabilities to observe the empirical currents \( j_r(C', C) \) and to observe the opposite empirical currents \( (-j_r(C', C)) \) that would characterize the Backwards trajectories (equation (18)) simplifies into a linear term with respect to the currents in the exponential for any empirical density \( \rho(C) \) and activities \( a_r(C', C) \)

\[
\frac{P_T[\rho(\cdot), j(\cdot), a(\cdot)]}{P_T[\rho(\cdot), -j(\cdot), a(\cdot)]] = e^{-T [I[\rho(\cdot), j(\cdot), a(\cdot)] - I[\rho(\cdot), -j(\cdot), a(\cdot)]]} \tag{58}
\]

Again this is a very direct generalization in the presence of several reservoirs \( r \) of the asymmetry with respect to the empirical currents \([8, 43]\): one recognizes the contribution of the action asymmetry (equation (37)) and the contribution of the entropy asymmetry (equation (50)) discussed previously.

### 5.3. Joint probability of the empirical density and the empirical currents (without the empirical activities)

From equation (56), one can obtain the joint probability of the empirical density and the empirical currents by integrating over the empirical activities that do not appear in the constraints

\[
P_T[\rho(\cdot), j(\cdot)] \sim \sum_{a(\cdot)} P_T[\rho(\cdot), j(\cdot), a(\cdot)] e^{-T I[\rho(\cdot), j(\cdot), a(\cdot)]} \sum_{a(\cdot)} e^{-T I[\rho(\cdot), j(\cdot), a(\cdot)]} \tag{59}
\]
As a consequence, one just needs to optimize the rate function \( I[\rho(\cdot), j(\cdot), a(\cdot)] \) over the activities \( a_r(C', C) \)

\[
0 = \frac{\partial I[\rho(\cdot), j(\cdot), a(\cdot)]}{\partial a_r(C', C)} = \frac{1}{2} \ln \left( \frac{a_r^2(C', C) - \hat{\rho}_r(C', C)}{4w_r(C, C')\rho(C')w_r(C', C)\rho(C)} \right).
\]

(60)

Plugging the optimal solution

\[
a_r^{opt}(C', C) = \sqrt{f_r(C', C) + 4w_r(C, C')\rho(C')w_r(C', C)\rho(C)}
\]

(61)

into the rate function yields the optimal value

\[
I^{opt}[\rho(\cdot), j(\cdot), a^{opt}(\cdot)] = \sum C \sum C' \sum r \rho(C')w_r(C', C)\rho(C)\left( \sqrt{f_r(C', C) + 4w_r(C, C')\rho(C')w_r(C', C)\rho(C)} \right)
\]

(62)

that governs the large deviation form of equation (59)

\[
P_T[\rho(\cdot), j(\cdot)] \propto D[\rho(\cdot), j(\cdot)] e^{-T I^{opt}[\rho(\cdot), j(\cdot)]}.
\]

(63)

Again equation (62) is a very direct generalization in the presence of several reservoirs \( r \) of the known formula for Markov jump processes [8, 42, 43].

Equation (62) yields the property directly inherited from equation (58)

\[
\frac{P_T[\rho(\cdot), j(\cdot)]}{P_T[\rho(\cdot), -j(\cdot)]} \xrightarrow{T \to +\infty} e^{-T \int R \left[ I^{opt}[\rho(\cdot), j(\cdot)] - I^{opt}[\rho(\cdot), -j(\cdot)] \right]}
\]

\[
= e^{T \sum C \sum C' \sum r j_r(C', C) \ln \left( \frac{w_r(C', C)\rho(C)}{w_r(C, C')\rho(C')} \right)}.
\]

(64)

6. Statistics of the other intensive time-empirical observables

In this section, we explain for our present framework with several thermal reservoirs how the statistics of any intensive empirical time-additive observable of the dynamical trajectory can be reconstructed from its decomposition in terms of the empirical densities and of the empirical jump densities studied in the previous sections. We also mention the link with the generating function method based on deformed Markov operators, that have attracted a lot of interest recently in various models [16–29].

6.1. Decomposition onto the time empirical density and time-empirical jump densities

For the trajectory \( [C(t)]_{0 \leq t \leq T} \) of equation (23), any intensive time-empirical observable

\[
b( [C(t)]_{0 \leq t \leq T} ) \equiv \frac{1}{T} \int_0^T df(C(t)) + \frac{1}{T} \sum_{k=1}^M g_r(\rho(C(t_k^+), C(t_k^-)))
\]

(65)
only depends on the time-empirical density \( \rho(C) \) and the time-empirical jump densities \( k_r(C', C) \) introduced in section 2.3

\[
b ([C(t)]_{0 \leq t \leq T}) = \sum_C \rho(C)F(C) + \sum_C \sum_{r=1}^\infty k_r(C', C)g_r(C', C) \equiv b[\rho(\cdot), k(\cdot)]. \tag{66}
\]

### 6.2. Large deviation form of the probability distribution \( p_T(b) \) of \( b \)

Its probability distribution over trajectories \([C(t)]_{0 \leq t \leq T}\) with the measure of equation (28)

\[
p_T(b) = \sum_{[C(t)]_{0 \leq t \leq T}} \mathcal{P}([C(t)]_{0 \leq t \leq T}) \delta(b - b([C(t)]_{0 \leq t \leq T})) \tag{67}
\]
can be thus rewritten in terms of the probability distribution of the intensive empirical observables with the measure of equation (51) with the large deviation form of equation (53)

\[
p_T(b) \approx \sum_{\rho(.)} \mathcal{P}_T[\rho(\cdot), k(\cdot)] \delta(b - b[\rho(\cdot), k(\cdot)]) e^{-TL[\rho(\cdot), k(\cdot)]} \tag{68}
\]

where the constraint \( \delta(b - b[\rho(\cdot), k(\cdot)]) \) appears as a supplementary constraint with respect to the constitutive ones \( \mathcal{D}[\rho(\cdot), k(\cdot)] \) of equation (15). As a consequence, the probability distribution of \( b \) will follow the large deviation form

\[
p_T(b) \approx e^{-TL(b)} \tag{69}
\]

where the rate function \( L(b) \) corresponds to the minimum of the rate functional \( I[\rho(\cdot), k(\cdot)] \) over \( \rho(\cdot) \) and \( k(\cdot) \) in the presence of the constraints \( \delta(b - b[\rho(\cdot), k(\cdot)]) \) and \( \mathcal{D}[\rho(\cdot), k(\cdot)] \); this operation is generically called ‘contraction’ in the large deviation language.

### 6.3. Scaled cumulant generating function of \( b \) via the appropriate deformed Markov operator

The large deviation form of equation (69) translates into the large deviation form for generating function of \( b \)

\[
\int dbp_T(b) e^{Tb} \approx \int db e^{T(b-L(b))} e^{T\phi(\nu)} \tag{70}
\]

where the saddle-point evaluation of the integral over \( b \) yields that the scaled cumulant generating function \( \phi(\nu) \) is the Legendre transform of the rate function \( L(b) \)

\[
\phi(\nu) = \sup_b (\nu b - L(b)). \tag{71}
\]

The generating function reads in terms of the probability \( \mathcal{P}_T[\rho(\cdot), k(\cdot)] \) of the empirical densities

\[
\int dbp_T(b) e^{Tb} \approx \sum_{\rho(.)} \sum_{k(.)} \mathcal{P}_T[\rho(\cdot), k(\cdot)] e^{Tb[\rho(\cdot), k(\cdot)]} \approx \sum_{\rho(.)} \sum_{k(.)} \mathcal{D}[\rho(\cdot), k(\cdot)] e^{Tb[\rho(\cdot), k(\cdot)]-I[\rho(\cdot), k(\cdot)]}. \tag{72}
\]
Using the explicit expressions of equations (66) and (54), the factor of \( T \) in the exponential reads
\[

\nu b [\rho(t), k (\ldots)] - I [\rho(t), k (\ldots)] = \nu \sum_C \rho(C) f(C) + \nu \sum_C \sum_{C' \neq C} \sum_{r=1}^R k_r(C', C) g_r(C', C)
\]

\[ - \sum_C \sum_{C' \neq C} \sum_{r=1}^R \left[ w_r(C', C) \rho(C) - k_r(C', C) + k_r(C', C) \ln \left( \frac{k_r(C', C)}{w_r(C', C) \rho(C)} \right) \right] \]

\[ = - \sum_C \left[ \sum_{C' \neq C} \sum_{r=1}^R w_r(C', C) - \nu f(C) \right] \rho(C) - \sum_C \sum_{C' \neq C} \sum_{r=1}^R k_r(C', C) \left[ -1 + \ln \left( \frac{k_r(C', C)}{w_r(C', C) \rho(C)} \right) \right] .
\]

(73)

This type of result for Markov jump processes can be re-interpreted as the effect of some \( \nu \)-dependent deformed Markov operator for a vector \( \Psi_{t}^{(\nu)}(C) \) (see the detailed proof in the section 4.2 of [10])
\[

\partial_t \Psi_{t}^{(\nu)}(C') = \sum_{C' \neq C} \sum_{r=1}^R w_r(C', C) \Psi_{t}^{(\nu)}(C') - w_{(\nu)}^{\text{out}}(C) \Psi_{t}^{(\nu)}(C')
\]

where the reservoir-dependent transition rates for \( C' \neq C \) are tilted according to
\[

w_{r}(C', C) \equiv e^{\nu \rho(C', C)} w_{r}(C', C)
\]

(75)

while the diagonal terms are changed into
\[

w_{(\nu)}^{\text{out}}(C) \equiv \sum_{C' \neq C} \sum_{r=1}^R w_{r}(C', C) - \nu f(C).
\]

(76)

Again, this is a very direct generalization in the presence of several reservoirs \( r \) of the known result for Markov jump dynamics \([10, 24, 42]\). Then the scaled cumulant generating function \( \phi(\nu) \) of equation (70) corresponds to the highest eigenvalue of thee deformed Markov operator \( w_{(\nu)} \) of equations (75) and (76), while the corresponding left and right eigenvectors are essential to understand the properties of the ‘conditioned’ dynamics (see \([23, 24, 28, 29, 36]\) and references therein).

7. Conclusion

For a system in contact with several reservoirs \( r \) at different inverse-temperatures \( \beta_r \), we have described how the Markov jump dynamics with the generalized detailed balance condition could be analyzed via a statistical physics approach of dynamical trajectories \([C(t)]_{0 \leq t \leq T}\) over a long time interval \( T \to +\infty \). At each step, we have stressed the similarities and differences with the statistical physics theory of equilibrium. The main points can be summarized as follows:

(i) the relevant intensive variables are the time-empirical density \( \rho(C) \) and the time-empirical jump densities \( k_r(C', C) \) from configuration \( C \) to configuration \( C' \) produced by the reservoir \( r \) (that furnishes or absorbs the energy difference). The probability of an individual trajectory \( P \{ C(t) \}_{0 \leq t \leq T} \sim e^{-TA[\rho(\cdot), k (\ldots)]} \) involves the intensive action \( A[\rho(\cdot), k (\ldots)] \) that only depends on these empirical observables. The analogy with equilibrium is that, in the canonical ensemble, the intensive energy \( e = \frac{E(C)}{N} \) of a configuration is the only
relevant intensive variable that determines the probability of a configuration in the thermodynamic limit \( N \to +\infty \).

(ii) the measure \( \Omega_T[\rho(\cdot), k(\ldots)] \) of the set of dynamical trajectories \( [\mathcal{C}(t)]_{t \in \mathbb{T}} \) that have the same empirical variables grows exponentially \( \Omega_T[\rho(\cdot), k(\ldots)] \sim e^{TS_{\text{Boltzmann}}[\rho(\cdot), k(\ldots)]} \) with respect to \( T \), where \( S_{\text{Boltzmann}}[\rho(\cdot), k(\ldots)] \) represents the intensive Boltzmann entropy.

The analogy with equilibrium is that, in the microcanonical ensemble, the number of configurations with a given intensive energy \( e \) grows exponentially in \( N \) with a coefficient given by the Boltzmann intensive entropy \( s(e) \).

(iii) the measure over dynamical trajectories \( P_t ([\mathcal{C}(t)]_{t \in \mathbb{T}}) \) can be thus replaced by a measure \( P_T[\rho(\cdot), k(\ldots)] \) over the empirical observables with their constraints: the large deviation form for large \( T \) involves the rate function \( I[\rho(\cdot), k(\ldots)] = A[\rho(\cdot), k(\ldots)] - S_{\text{Boltzmann}}[\rho(\cdot), k(\ldots)] \) that involves the action of (i) and the entropy of (ii). The analogy with equilibrium is that the intensive free-energy contains an energetic contribution and an entropic contribution.

(iv) This measure \( P_T[\rho(\cdot), k(\ldots)] \) over the relevant empirical variables allows to study the statistics of any intensive time-additive observable \( b \) of the dynamical trajectory, in particular those that contain information on the exchanges with the various reservoirs.

(v) The behavior of various observables upon time-reversal has been analyzed via the replacement of the time-empirical jump densities \( k_{\tau}(\mathcal{C}', \mathcal{C}) \) by the time-empirical activities \( a_{\tau}(\mathcal{C}', \mathcal{C}) \) and currents \( j_{\tau}(\mathcal{C}', \mathcal{C}) \).

This non-equilibrium framework for a system in contact with several reservoirs \( r \) has been formulated at the same level of generality of the statistical physics theory of equilibrium, where the system can be in a certain number of configurations \( \mathcal{C} \) of energies \( E(\mathcal{C}) \). For applications to specific models, besides this space of configurations with the energy function \( E(\mathcal{C}) \), one needs to choose the transition rates \( w_{\tau}(\mathcal{C}', \mathcal{C}) \), i.e. what dynamical elementary moves \( \mathcal{C}' \leftarrow \mathcal{C} \) are allowed between configurations (for instance single-spin-flip dynamics in spin models) and how the various reservoirs \( r \) are involved in these possible transitions, depending on the real-space structure of the model under study.

As a final remark, let us note that here we have focused on a system that could only exchange energy with the various thermal reservoirs, but a closely-related framework for systems exchanging particles with reservoirs in terms of empirical densities and empirical flows can be found in [44] for independent particles and in [43] for interacting particles, with possibly time-dependent rates, in order to take into account the possibility of external driving with some protocol \( \lambda(t) \), in particular for the case of periodic driving, which has attracted a lot of interest recently from the point of view of large deviations [44–48].

Appendix. Short reminder on the statistical physics theory of equilibrium

In this appendix, we recall some basic ideas of the statistical physics of equilibrium in order to make more explicit the analogy with some equations of the text concerning non-equilibrium.

A.1. Enumeration of the configurations \( \mathcal{C} \) and of their energies \( E(\mathcal{C}) \)

The statistical physics theory of equilibrium is formulated at a very general level where a system is described by a set of configurations \( \mathcal{C} \) characterized by their energies \( E(\mathcal{C}) \). To simplify the discussion, let us focus on the standard example of the the Ising model of \( N \) classical spins \( S_j = \pm 1 \) in dimension \( d \) with nearest-neighbor coupling \( J \). The \( N = 2^N \) configurations \( \mathcal{C} = (S_1, \ldots, S_N) \) can be decomposed with respect to the discrete extensive energy \( E \).
\[ N = 2^N = \sum_E N(E) \]  
\[ \text{(A.1)} \]

where

\[ N(E) \equiv \sum_C \delta_{E,E(C)} \]  
\[ \text{(A.2)} \]

counts the number of configurations \( C \) that have the same energy \( E(C) = E \).

**A.2. Microcanonical ensemble at fixed energy \( E \) (for an isolated system)**

In the microcanonical ensemble associated to the fixed energy \( E \), the \( (N - N(E)) \) configurations having a different energy \( E(C) \neq E \) are not possible, while the \( N(E) \) configurations having this energy \( E(C) = E \) are equiprobable

\[ P^\text{micro}_E(C) \equiv \frac{\delta_{E,E(C)}}{N(E)}. \]  
\[ \text{(A.3)} \]

The corresponding Shannon entropy of this microcanonical probability coincides with the Boltzmann entropy

\[ S^\text{micro}_E \equiv -\sum_C P^\text{micro}_E(C) \ln (P^\text{micro}_E(C)) = \ln N(E) \equiv S^\text{Boltzmann}_E. \]  
\[ \text{(A.4)} \]

For a large number \( N \) of spins, both the energy \( E \) and the entropy \( S^\text{micro}_E \) are extensive in \( N \). It is then useful to introduce the intensive energy

\[ e \equiv \frac{E}{N} \]  
\[ \text{(A.5)} \]
as well as the intensive entropy \( s^\text{micro}(e) \) as a function of the intensive energy \( e \)

\[ s^\text{micro}(e) \equiv \frac{S^\text{micro}_E(E=Ne)}{N}. \]  
\[ \text{(A.6)} \]

that governs the exponential growth in \( N \) of the number of configurations with a given intensive energy \( e \)

\[ N(e) \simeq e^{N s^\text{micro}(e)}. \]  
\[ \text{(A.7)} \]

The inverse temperature associated to this microcanonical ensemble reads

\[ \beta^\text{micro} = \frac{ds^\text{micro}(e)}{de}. \]  
\[ \text{(A.8)} \]

**A.3. Canonical ensemble at inverse temperature \( \beta \) (for a system in contact with a thermal reservoir)**

The canonical ensemble at inverse temperature \( \beta \) corresponds to the probability distribution

\[ P^\text{cano}_\beta(C) \equiv \frac{e^{-\beta E(C)}}{Z_\beta} \]  
\[ \text{(A.9)} \]
where the canonical partition function $Z_\beta$ that ensures the normalization allows to define the free-energy $F_\beta$ via

$$e^{-\beta F_\beta} \equiv Z_\beta = \sum_C e^{-\beta E(C)} = \sum_E N(E) e^{-\beta E} = \sum_E e^{S_{\text{micro}}(E) - \beta E}. \quad \text{(A.10)}$$

The corresponding Shannon entropy reads

$$S_{\beta_{\text{cano}}} = -\sum_C P_{\beta_{\text{cano}}}(C) \ln \left( P_{\beta_{\text{cano}}}(C) \right) = \beta (F_\beta - U_\beta) \quad \text{(A.11)}$$

in terms of the free-energy $F(\beta)$ of equation (A.10) and of the internal energy

$$U_\beta = \sum_C E(C) P_{\beta_{\text{cano}}}(C). \quad \text{(A.12)}$$

For small $N$, the canonical ensemble is of course completely different from the microcanonical ensemble. However for large $N$, the extensivity of the energy (equation (A.5)) and of the microcanonical entropy (equation (A.6)) leads to the saddle-point evaluation of the partition function of equation (A.10)

$$e^{-\beta F_\beta} \equiv Z_\beta \simeq N \int \text{d}e e^{N s_{\text{micro}}(e) - \beta e} \simeq e^{-\beta N f(\beta)} \quad \text{(A.13)}$$

where the intensive free-energy

$$f(\beta) \equiv \frac{F(\beta)}{N} \quad \text{(A.14)}$$

is given by the saddle-point value of equation (A.13)

$$0 = \partial_e \left( s_{\text{micro}}(e) - \beta e \right) = \beta s_{\text{micro}}(e) - \beta$$

and

$$-\beta f(\beta) = s_{\text{micro}}(e) - \beta e \quad \text{(A.15)}$$

so that the canonical ensemble at inverse temperature $\beta$ is actually dominated in the thermodynamic limit by the configurations of the microcanonical ensemble at intensive energy $e$ satisfying $\beta s_{\text{micro}}(e) = \beta$.

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