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Inequalities Generalizing the Second Law of Thermodynamics for Transitions between Nonstationary States

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We discuss the consequences of a variant of the Hatano-Sasa relation in which a nonstationary distribution is used in place of the usual stationary one. We first show that this nonstationary distribution is related to a difference of traffic between the direct and dual dynamics. With this formalism, we extend the definition of the adiabatic and nonadiabatic entropies introduced by M. Esposito and C. Van den Broeck in Phys. Rev. Lett. 104, 090601 (2010) for the stationary case. We also obtain interesting second-law-like inequalities for transitions between nonstationary states.

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Introduction.—The second law of thermodynamics provides fundamental limitations on the way transitions between equilibrium states can occur. For many years, this principle could only be expressed as an inequality. A broad number of works summarized under the name of fluctuations theorems [1,2] have changed that fundamentally by providing equalities valid for systems arbitrarily far from equilibrium. In particular, one equality defines the entropy production as the amount of time-symmetry breaking [3], a statement which not only encompasses the second law but also sharpens it by providing additional implications at the trajectory level. This notion of trajectories is also meaningful for stochastic optimization problems [4] and more generally for application to information theory. In this respect, the recent generalization of the Jarzynski relation for systems operating under feedback control [5] appears particularly significant. With all these exciting developments, the second law of thermodynamics, a rather old idea, appears nowadays more alive than ever.

In these generalizations, an essential step was made by Hatano and Sasa [2], who introduced the functional

$$Y[c] = \int_0^T dt \delta h_i \delta_s \phi(c, h_i),$$

(1)

where \(\phi(c, h) = -\ln p_a(c, h)\), and \(p_a(c, h)\) is the stationary probability distribution to be in a microstate \(c\) with a constant value of the control parameter \(h\). They have shown that \(\langle \exp(-Y[c]) \rangle = 1\), a relation which has been confirmed experimentally with small colloidal systems [6]. Their relation implies \(Y \geq 0\), which translates into a modified second law for transitions between nonequilibrium steady states (NESS) [7], and through an expansion it also leads to a modified fluctuation-dissipation theorem for systems near a NESS [8]. In this framework, in the limit of very slow driving, the probability distribution assumes at all times its stationary form evaluated at the value of the control parameter at this time. In contrast to this, we develop in this paper a framework for situations in which even in the limit of slow driving the probability distribution stays time dependent. This extension is important to treat the following situations: (i) the system never reaches a stationary state on any reasonable time scale, as in aging systems, (ii) the system is driven by at least two control parameters, so even when the protocol \(h\) is constant the dynamics remains nonstationary, or (iii) the system is prepared in a nonstationary distribution by the choice of initial conditions and is then further driven. In [9], it was shown that, remarkably, when a stationary distribution is used as reference, the second law can be split into two components, the so called adiabatic part (corresponding to the contribution of the entropy production in the limit of slow driving) and the remaining nonadiabatic part. In this Letter, we ask whether in the particular situations mentioned above, the second law can still be split into two components similar to the adiabatic and nonadiabatic entropy productions.

In order to investigate this question, we consider a system which evolves according to a continuous-time Markovian dynamics of a pure jump type. We denote a trajectory by \([c] = (c_0, c_1, \ldots, c_N; \tau_1, \ldots, \tau_N)\) where the \(c_i\) are the states which are visited by the system and \(\tau_i\) are the jumping times to go from \(c_{i-1}\) to \(c_i\). This trajectory is recorded from time 0 to time \(T\). The transition rate to jump from a configuration \(c\) to a configuration \(c'\) is denoted \(w_i^{h_j}(c, c')\). In this notation, the superscript \(h_i\) refers to the dependence on the driving protocol, while the subscript \(i\) refers to an additional time dependence not related to the protocol \(h\). The path probability for this trajectory is

$$P[c] = p_0(c_0) \prod_{j=1}^N \exp \left( -\int_{\tau_{j-1}}^{\tau_j} dt \lambda_i^{h_i}(c_{j-1}) w_i^{h_j}(c_{j-1}, c_j) \right) \times \exp \left( -\int_{\tau_N}^{T} dt \lambda_N^{h_N}(c_N) \right).$$

(2)

where \(\lambda_i^{h_i}(c') = \sum_{c \neq c'} w_i^{h_i}(c', c)\) represents the escape rate to leave the state \(c'\) and \(p_0(c_0)\) is the probability distribution...
of the initial condition \(c_0\). Let us consider a logratio of path probabilities of the form:

\[
\Delta \mathcal{A}[c] = \ln \frac{\mathcal{P}[c]}{\mathcal{P}[c']},
\]

where \(\mathcal{P}\) results from the application of an involution \(\sim\) on the path probability \(\mathcal{P}\), while \([c'] = (c_0, c_1, \ldots, c_N; \tau_1, \ldots, \tau_N)\) results from a different involution acting on the trajectory \([c]\). In the following, we assume that the involution \(\sim\) is either the identity \([c'] = [c]\) or the time-reversal symmetry \([c'] = [\bar{c}] = (c_N, c_{N-1}, \ldots, c_0; T - \tau_N, \ldots, T - \tau_1)\). By computing the logratio of these probabilities, one finds that

\[
\Delta \mathcal{A}[c] = \ln \frac{p_0(c_0)}{p_0(c_0^r)} - \int_0^T dt \left[ \lambda_t^{h_t}(c_t) - \lambda_t(c_t) \right] + \sum_{j=1}^N \ln \frac{w_{t_j}^{h_t}(c_{j-1}, c_j)}{w_{t_j}(c_{j-1}, c_j)},
\]

with \(c_t = c_j\) if \(t \in [\tau_j, \tau_{j+1}]\) and \(\lambda_t^{h_t} = \lambda_t^{h_{t-j}}\) if the involution \(\sim\) is the time reversal. Note that the second term in \(\Delta \mathcal{A}[c]\) corresponds to a difference of traffic (i.e., time integrated escape rates) between the dynamics generated by \(\mathcal{P}\) and that generated by \(\mathcal{P}\), and has similarities with the dynamical activity introduced in Ref. [10].

Now introducing \(\tilde{\mathcal{P}}(\Delta \mathcal{A}) = \sum_{c} \delta(\Delta \mathcal{A} - \Delta \mathcal{A}[c]) \times \mathcal{P}[c]\), and using the relation \(\Delta \mathcal{A}[c] = -\Delta \mathcal{A}[c']\), which follows from Eq. (3), one obtains a detailed fluctuation theorem (DFT) for \(\Delta \mathcal{A}\), namely

\[
P(\Delta \mathcal{A}) = \exp(\Delta \mathcal{A}) \tilde{\mathcal{P}}(-\Delta \mathcal{A}).
\]

As a first application of Eq. (4), we choose the involution star to be the time-reversal symmetry and the involution tilde to be the time reversal for the rates that we denote with a bar, such that \(\tilde{w}_{t_j}^{h_t}(c, c') = \tilde{w}_{t_j}^{h_{t-j}}(c', c)\). In this case, the second term in Eq. (4) is zero, and \(\tilde{\mathcal{A}}\) represents the total entropy production, \(\Delta S_{\text{tot}}\), which satisfies a DFT of the form above [11].

First main result.—We now introduce a new involution, namely, the duality transformation, that we denote by a hat (\(\Lambda\)). In analogy with the stationary case [9], we define the dual dynamics by the following transformation of the rates:

\[
\hat{w}_{t_j}^{h_t}(c, c') = \frac{w_{t_j}^{h_t}(c', c) \pi_{t}(c', h)}{\pi_t(c, h)},
\]

where the distribution \(\pi_t(c, h)\) represents the probability to observe the system in the state \(c\) at a time \(t > 0\) in the presence of a constant (time-independent) driving \(h\). This distribution, which plays a key role here satisfies

\[
\left(\frac{\partial}{\partial t}\right) \pi_t(c, h) = \sum_{c'} \pi_t(c', h) L_t^h(c', c),
\]

where \(L_t^h\) is the generator defined by \(L_t^h(c', c) = \frac{d}{dt} \delta(c, c') \sum_{c} \hat{w}_{t_j}^{h_t}(c', c)\). We emphasize that \(\pi_t(c, h)\) depends only on the driving at time \(t\) unlike \(p_t(c, [h_t])\), the solution of the master equation with the same initial condition but with the generator \(L_t^h\), which depends functionally on the driving history \([h_t]\) up to time \(t\). Using the duality transformation introduced above, we consider the following two cases for the involutions entering in Eq. (3): Case (A) where the involution \(\sim\) is the combination of duality plus time reversal \((\Lambda)\) and \(\Lambda\) is the time reversal, with \(\Delta \mathcal{A}[c] = \Delta S_{\text{na}}[c]\), and case (B) where the involution \(\sim\) is the duality \((\Lambda)\) and \(\Lambda\) is the identity, and \(\Delta \mathcal{A}[c] = \Delta B_{\text{na}}[c]\). In both cases, the integral in the right-hand side of Eq. (4) is the same and, using Eq. (6) and (7), it can be written as

\[
\int_0^T dt \left[ \lambda_t^{h_t}(c_t) - \hat{\lambda}_t^{h_t}(c_t) \right] = -\int_0^T dt \left( \partial_t \ln \pi_t(c_t, h_t) \right).
\]

We call that quantity \(\Delta T^\ast[c]\) in the following. Note that in Eq. (8), the time derivative acts only on \(\ln \pi_t\), but not on the arguments of that function. This relation is our first result. It establishes an important link between the difference of traffic associated with the direct and dual dynamics and the accompanying distribution \(\pi_t(c, h_t)\).

Second main result.—We now show that using the quantity \(\Delta T^\ast\) we can generalize the notions of adiabatic and nonadiabatic contribution to the total entropy production, denoted respectively \(\Delta S_{\text{a}}\) and \(\Delta S_{\text{na}}\). We define them to be

\[
\Delta S_{\text{a}}[c] = \ln \frac{p_0(c_0)}{\tilde{p}_0(c_0^r)} + \sum_{j=1}^N \ln \frac{\pi_{t_j}(c_{j-1}, h_{t_j})}{\tilde{\pi}_{t_j}(c_{j-1}, h_{t_j})},
\]

\[
\Delta S_{\text{na}}[c] = \sum_{j=1}^N \ln \frac{w_{t_j}(c_{j-1}, c_j) \pi_{t_j}(c_{j-1}, h_{t_j})}{\tilde{w}_{t_j}(c_{j-1}, c_j) \tilde{\pi}_{t_j}(c_{j-1}, h_{t_j})},
\]

where \(\pi_{t_j}(c, h)\) replaces again the stationary distribution in the usual definition [9]. These two quantities are such that \(\Delta S_{\text{tot}} = \Delta S_{\text{a}} + \Delta S_{\text{na}}\). Since \(\Delta S_{\text{tot}}\) can be further split into reservoir entropy \(\Delta S_{\text{r}}\) and system entropy with \(\Delta S = \ln p_0(c_0) - \ln \tilde{p}_0(c_0^r)\), one can introduce an excess entropy \(\Delta S_{\text{ex}}\) such that \(\Delta S_{\text{r}} = \Delta S_{\text{a}} + \Delta S_{\text{ex}}\) and \(\Delta S_{\text{na}} = \Delta S - \Delta S_{\text{ex}}\). Unfortunately, the splitting into adiabatic and nonadiabatic contributions does not have, in general, the property that each term satisfies a DFT, although the joint distribution of \(\Delta S_{\text{a}}\) and \(\Delta S_{\text{na}}\) satisfies such a relation [12]. Nonetheless, being of the form of Eq. (3), \(\Delta A_{\text{na}} = \Delta S_{\text{na}} - \Delta T^\ast\) and \(\Delta B_{\text{a}} = \Delta S_{\text{a}} - \Delta T^\ast\) do verify separately a DFT

\[
\ln \frac{P(\Delta A_{\text{na}})}{\tilde{P}(-\Delta A_{\text{na}})} = \Delta A_{\text{na}}, \quad \ln \frac{P(\Delta B_{\text{a}})}{\tilde{P}(-\Delta B_{\text{a}})} = \Delta B_{\text{a}}.
\]

These relations represent the second main result of this Letter, which we now discuss in more detail:

Let us assume that the driving starts at time \(t_{\text{di}} > 0\) and ends at time \(t_{\text{df}} < T\) for a total duration \(t_d = t_{\text{df}} - t_{\text{di}}\). When \(\pi_t(c, h)\) relaxes quickly to the stationary
distribution (on a time scale $\tau_{st}$ such that $\tau_{st} \ll T$ and $\tau_{st} \ll \tau_d$), one recovers from Eqs. (9) and (10) the usual definitions of the nonadiabatic and adiabatic parts of the entropy production. In this case, $\Delta T = 0$, and Eqs. (11) become the usual detailed fluctuation theorems satisfied by the adiabatic and nonadiabatic entropies [9].

We now notice that $\Delta A_{na} = \Delta S_b + \mathcal{Y}_f$, where $\Delta S_b = \Delta S - \Delta \psi$ is a boundary term, with $\Delta \psi = \ln \pi_0(c_0, h_0) - \ln \pi_T(c_T, h_T)$ and

$$\mathcal{Y}_f[c] = \int_0^T dt h_c \partial_c \psi_f(c, h).$$

This quantity is the exact analog of Eq. (1), when the stationary distribution $p_0(c, h)$ is replaced by $\pi_f(c, h) = \exp(-\psi_f(c, h))$. Then, a consequence of the first equation in Eq. (11) is that the functional $\mathcal{Y}_f$ satisfies a generalized Hatano-Sasa relation

$$\langle \exp(-\mathcal{Y}_f[c]) \rangle = 1,$$

when we consider transitions between nonstationary states, that is to say, when the time $T$ is such that $T - \tau_d \gg \tau$ where $\tau$ is the relaxation time of $p_0(c_0, h_0)$ towards $\pi_f(c_0, h_0)$. Since $p_0(c_0, h_0) = \pi_0(c_0, h_0)$, in this case the boundary term $\Delta S_b$ vanishes. We note that by expanding Eq. (13) we have a modified fluctuation-dissipation theorem valid for systems near a general nonequilibrium state [13,14].

Note also that the remaining parts in the entropy production $\Delta A_a = \Delta S_{tot} - \Delta A_{na} = \Delta S_a + \Delta T$ and $\Delta B_a = \Delta S_{tot} - \Delta B_{na} = \Delta S_{na} + \Delta T$ do not satisfy a DFT of the form of Eq. (5) except in the stationary case due to the fact that in this case $\Delta T = 0$.

Third main result.—Using the Jensen inequality on integrated fluctuation theorem associated to Eq. (11), we get $\langle \Delta A_{na} \rangle \geq 0$ and $\langle \Delta B_{na} \rangle \geq 0$. We can equivalently write that $\langle \Delta S_a \rangle \geq \langle \Delta T \rangle$ and $\langle \Delta S_{na} \rangle \geq \langle \Delta T \rangle$, which taken together imply $\langle \Delta S_{tot} \rangle \geq \max(2\langle \Delta T \rangle, 0)$. These inequalities are very general, they hold for Markov processes in finite time $T$, arbitrary initial probability distribution $p_0$, arbitrary driving, and arbitrary dynamics of the system at constant time-independent driving. Note that there are no lower bounds for $\Delta A_a$ or $\Delta B_{na}$, which need not be positive on average. That this should be the case can be understood by considering a system at equilibrium on which two protocols that exactly compensate each other are applied and only the second protocol is considered as driving. In this case, the system is in equilibrium at all times, the rates satisfy a detailed balance condition, and $\langle \Delta S_{tot} \rangle = 0$. Therefore, $\langle \Delta A_a \rangle = 0$ and $\langle \Delta B_{na} \rangle \leq 0$.

From the inequality $\langle \Delta A_{na} \rangle \geq 0$, one obtains

$$\langle \Delta S \rangle \geq -\langle \Delta S_{ex} \rangle + \langle \Delta T \rangle,$$

which contains the second law for transitions between equilibrium states and the modified second law for transitions between NESS [2,7] as particular cases. For this reason, we call Eq. (14) a modified second law for transitions between nonstationary states. Alternatively, one has also $\langle \mathcal{Y}_f \rangle \geq -\langle \Delta S_{ex} \rangle = D(p_f||\pi_f) \geq 0$. The equality in these relations holds in the adiabatic limit, corresponding to infinitively slow driving on $h_1$, in which case $p_f$ has relaxed towards $\pi_f$ and thus $\mathcal{Y}_f$ is zero. In this limit, $\Delta A_{na} = 0$ which justifies the adiabatic or nonadiabatic terminology: $\Delta A_a = \Delta B_{na} = \Delta S_{tot}$ and $\Delta B_{na} = 0$. Taken together, these relations imply that $\Delta T = 0$ and that the second equation in Eq. (11) becomes the DFT satisfied by the total entropy production. Clearly, the driving can be slow even if $\pi_f(c, h)$ has not relaxed to a stationary distribution. Note also that in this adiabatic limit, the Shannon entropy constructed from $\pi_f(c, h)$, namely $\langle \Delta \psi \rangle$, equals the opposite of the excess entropy $-\langle \Delta S_{ex} \rangle$ as in the case of NESS.

Example.—As an illustrative example, we consider a time-dependent two states model, which may be realized experimentally in quantum optics for instance [15]. We have chosen for simplicity rates of the form $w^{b|h}(a, b) = w(a, b) e^{-h/2}$ and $w^{h|b}(b, a) = w(b, a) e^{h/2}$, where the driving $h_1$ follows a time-symmetric half sinusoidal protocol of duration $\tau_d = T$. The initial probability to be in one of the two states is chosen according to an arbitrary value different from the stationary value [here we chose arbitrarily $p_0(b) = 0.9$]. Therefore, the reference dynamics is nonstationary by the choice of initial condition. Through extensive kinetic Monte Carlo simulations of the trajectories followed by this system, we determine the distributions $\pi_f(c, h)$, the corresponding dual dynamics, and the distribution of $\mathcal{Y}_f$ from this data. Figure 1 shows that the first DFT of Eq. (11) is well obeyed in this case irrespective of whether the relaxation of the $\pi_f$ distribution is fast or not with respect to the driving. Figure 2 illustrates transitions

![Figure 1](image-url)
between two nonstationary states. The nonstationary states at given \( h \) are created by sinusoidal reference protocols with rates \( w^h_i(a, b) = w(a, b)e^{-h \cdot \sin \omega t} \) and \( w^h_i(b, a) = w(b, a)e^{h \cdot \sin \omega t} \). The transition is produced by a piecewise linear driving protocol \( h_t \) as shown in the inset of the figure. The various quantities \( \langle Y_T \rangle, \langle \Delta S_{na} \rangle, \text{and} \langle \Delta T \rangle \) are shown as a function of the driving of the duration \( t_d \). As expected, in the quasistatic limit \( t_d \to \infty \), one has \( \langle \Delta T \rangle = \langle Y_T \rangle = \langle \Delta S_{na} \rangle = 0 \), whereas \( t_d \to 0 \) corresponds to a quenched limit which is consistent with \( \langle \Delta S_{na} \rangle - \langle \Delta T \rangle = \langle Y_T \rangle \). Note that the general evolution of \( \langle Y_T \rangle \) as a function of \( t_d \) is similar to that of the dissipated work in the equilibrium case [3].

In this particular example, we have chosen a simple dynamics for which the distribution \( \pi(c, h) \) is analytically solvable. More generally for applications in complex systems, this distribution will not be available analytically; however, if the system (or subsystem) of interest is of small size, the numerical determination of this distribution is possible through simulations [14]. Among the various strategies which can facilitate this numerical determination, one recent suggestion is to determine the distribution iteratively by starting from an approximate ansatz function [16].

**Conclusion.**—We have connected the accompanying distribution \( \pi(c, h) \) introduced in [14] to the difference of traffic between the direct and dual dynamics. Using a nonstationary probability as reference, we have extended the notion of adiabatic and nonadiabatic contribution to the total entropy production. Unfortunately, the two new parts of the total entropy production do not verify separately a DFT as in the stationary case. Despite this, we have obtained two detailed fluctuation theorems with interesting consequences: a generalization of the Hatano-Sasa relation and second-law-like inequalities for transitions between nonstationary states. These results could have important applications, in particular, in force measurements with biopolymers or proteins, for the characterization of small glassy systems, or for stochastic optimization problems.