Rules for transition rates in nonequilibrium steady states

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Just as transition rates in a canonical ensemble must respect the principle of detailed balance, constraints exist on transition rates in driven steady states. I derive those constraints, by maximum information-entropy inference, and apply them to the steady states of driven diffusion and a sheared lattice fluid. The resulting ensemble can potentially explain nonequilibrium phase behaviour and, for steady shear, gives rise to stress-mediated long-range interactions.

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If we wish to design a driven stochastic model that exhibits a nonequilibrium steady state with a given flux, how should we choose its transition rates? I answer the question by applying Jaynes’ principle of maximum entropy inference (MaxEnt). It might seem perverse to specify a macroscopic result, and then infer an equation of motion, since theoretical modelling usually involves the reverse procedure. However, for equilibrium systems, the principle of detailed balance (DB) is derived in just this way. The mean energy is fixed, and the equilibrium ensemble defined as the distribution of states respecting that constraint, unbiased by any other information. One then infers the properties of the appropriate reversible dynamics, obtaining a set of rules, DB, which demands that the ratio of rates for a transition and its time-reverse is given by the Boltzmann factor of the energy cost $\Gamma$. It is commonly assumed that the same conditions of DB should be used in nonequilibrium models subjected to a finite through-put of flux, so that the dynamics of local transitions is governed by the same physics as at equilibrium. However, I shall show that those are not the transition rates predicted by MaxEnt when a mean flux, as well as a mean energy, is specified. The hypothesis that the phase-space paths adopted by nonequilibrium systems are distributed according to MaxEnt has been supported by some notable successes, including the recovery of linear transport theory and, more recently, the fluctuation theorem and self-organised criticality. It is also a cornerstone of the GENERIC approach to nonequilibrium kinetics.

We define a particular nonequilibrium ensemble to be the set of phase-space paths available to a system, minimally constrained by fixing only the mean energy and flux on those paths. The unbiased distribution of paths appropriate to those constraints is given by MaxEnt. Having defined this ensemble, we may investigate its properties without controversy. A leap of faith is required only to hypothesise that the ensemble is a good description of some physical systems under realistic forcing conditions. If so, then the method will have important implications for nonequilibrium phase transitions such as shear-banding and jamming. I derive physically convincing results for two applications: driven diffusion, and a model of interacting particles under shear.

Jaynes showed that Shannon’s information entropy $S_I = -\sum_{\Gamma} p(\Gamma) \ln p(\Gamma)$ is maximized by the distribution $p(\Gamma)$ of states in the equilibrium canonical ensemble, recovering the Boltzmann distribution. But Jaynes also applied the method to nonequilibrium problems, for which $\Gamma$ represents an entire path through phase space, spanning the duration $\tau$ of the nonequilibrium experiment in question. The maximization is constrained by whatever information is known about the paths. At equilibrium, a constant mean energy $\mathcal{E}$ is specified. To define a driven steady-state ensemble, we shall additionally stipulate a (possibly multi-component) mean flux $\mathcal{J}$. We shall write $p_r(X|Y)$ as the normalized probability for any quantity $X$ exhibited by the system during the time interval $\tau$, subject to conditions $Y$. Thus, MaxEnt yields the conditional probability $p_r(\Gamma|\mathcal{J},\mathcal{E})$ that, over the duration $\tau$, the system takes a path $\Gamma$, given that the mean energy and flux have the values specified. If ergodicity is assumed then, for $\tau \to \infty$, $\mathcal{E}$ and $\mathcal{J}$ can be interpreted as time-averages. So the system’s energy and flux are allowed to fluctuate, but their averages over the duration $\tau$ must have exactly the specified values on all the paths considered.

Our aim is to find the rate, in the driven ensemble,

$$\omega_{a \to b}^{\text{driv}} = \lim_{\Delta t \to 0} \frac{p_{\Delta t}(a \to b|a,\mathcal{J},\mathcal{E})}{\Delta t} \quad (1)$$

for a system to undergo some transition $a \to b$. This is the probability (per unit time) that the transition will occur within an interval $\Delta t$ of the current time ($t = 0$), given that the current state is $a$, and that the energy and flux will eventually (over duration $\tau$) have the specified time-averages. The constraints $\mathcal{E}$ and $\mathcal{J}$, which define the nonequilibrium ensemble, apply only to the duration $\tau$ as a whole; we do not constrain the energy and flux at each interval $\Delta t$ separately. A conditional probability of the kind in Eq. (1) can be manipulated by Bayes’ theorem, which gives two equivalent expressions for the joint probability of performing the transition $a \to b$ within time $\Delta t$, and acquiring a mean flux $\mathcal{J}$ in time $\tau$, given the mean energy and initial state, thus:

$$p_{\Delta t}(a \to b|a,\mathcal{J},\mathcal{E})p_r(\mathcal{J}|a,\mathcal{E}) = p_{\Delta t}(a \to b|a,\mathcal{E})p_r(\mathcal{J}|a \to b,\mathcal{E}).$$

Normalization implies that $p_r(\mathcal{J}|\ldots)$ has dimensions of reciprocal flux. Substituting into Eq. (1) and recognising
that quantities for which only the mean energy is constrained belong to the equilibrium ensemble, yields
\[ \omega^{\text{driv}}_{a \rightarrow b} = \omega^{\text{eq}}_{a \rightarrow b} \eta_{ab}. \] (2)

Equation (2) confirms, as asserted above, that the rate of a transition \( a \rightarrow b \) in the driven ensemble is not equal to the rate at equilibrium, but is enhanced by a factor
\[ \eta_{ab} \equiv \lim_{\tau \rightarrow \infty} \frac{p^{\text{eq}}(\mathcal{J}(a \rightarrow b))}{p^{\text{eq}}(\mathcal{J}(a))}. \] (3)

This factor is known in principle, as it is a property of the corresponding system at equilibrium, not in the driven ensemble. The denominator in Eq. (3) is the probability that a system, starting in the current state \( a \), will exhibit an average flux \( \mathcal{J} \) over the duration \( \tau \) if it is governed by the equilibrium transition rates. Of course it is exceedingly unlikely for an equilibrium system to spontaneously perform sustained flow, so the denominator is infinitesimal. The numerator in Eq. (3) measures the infinitesimal probability of that same flux at equilibrium, given that the dynamics begins with a transition to state \( b \). We shall see that Eqs. (2, 3) make intuitive sense in some examples below.

The above derivation exploits the fact that the driven ensemble is a sub-set of the equilibrium ensemble (albeit in the extreme tail of the flux distribution), since it is defined by one extra constraint. But the ‘sub-ensemble dynamics’ (SED) [Eqs. (2, 3)] should not be mistaken for a near-equilibrium approximation since the sub-set of paths has properties very different from the equilibrium set. Nevertheless, an equilibrium Markov process remains Markovian under SED. For many transitions, that contribute no flux and do not alter the future likelihood of flux, \( \eta_{ab} = 1 \) so Eq. (2) says the rate is equal in the driven and equilibrium ensembles, as often assumed. Two types of transition are boosted in the driven ensemble: (A) a transition that carries a positive flux in the direction of \( \mathcal{J} \), (B) a transition to a state that is more amenable to subsequent flux-carrying transitions.

Many choices of prior rates \( \omega^{\text{eq}}_{a \rightarrow b} \) are possible, either fully implementing Newton’s laws, or embodying approximate (e.g. Brownian) dynamics. Consider the exact Newtonian evolution of the particles of a fluid element surrounded by a reservoir of more fluid. Unphysical transitions, e.g. violating momentum conservation for internal degrees of freedom, have zero prior rate, so Eq. (2) also forbids such transitions in the driven case, e.g. under shear flow. Thus, the scheme respects Newton’s laws and is consistent with the Liouville equation. The reservoir introduces randomness into the dynamics by coupling to particles at the surface of the fluid element. SED provides the unbiased description of the reservoir’s influence.

We now study two examples with Brownian prior dynamics. The first, a driven Brownian ideal gas, is simple enough for exact calculation, but exhibits only ‘type A’ transitions. More of the physical richness of SED will appear in the second example, a complex system under shear that demonstrates both types of enhancement.

Let us find the equation of motion for the particles of a Brownian ideal gas with a drift velocity \( v \). The problem de-couples for each component of each particle’s displacement. The non-trivial part is the component in the flux direction. We must assume prior knowledge of the motion at equilibrium, for which each coordinate \( x(t) \) performs an unbiased random walk according to the Langevin equation, \( \xi \) (where \( \xi \) is a function constant, and \( \xi \) a delta-correlated noise function). We introduce a time step \( \Delta t \) that will eventually be taken to zero, so that the thermal noise \( \xi \) is drawn from a well-behaved Gaussian distribution and the Langevin equation relates this stochastic variable \( \xi \) to a step \( x \rightarrow x + \Delta x \) such that \( \Delta x = \xi \Delta t/\zeta \). This equilibrium dynamics dictates that each Brownian particle follows a path with steps drawn from the distribution
\[ p^{\text{eq}}(\Delta x) = G(\Delta x, \Delta t) = \frac{1}{\sqrt{4\pi D\Delta t}} \exp \left( -\frac{\Delta x^2}{4D\Delta t} \right) \] (4)

which corresponds (per unit time) to a transition rate (per unit distance) \( \omega^{\text{eq}}_{x \rightarrow x+\Delta x} \) in the notation of Eq. (2).

Here, \( G(x, t) \) is the Green function for free diffusion, and the diffusion coefficient is given by the Einstein relation \( D = k_B T/\zeta \). Now, from the set of all equilibrium paths, we extract the sub-set exhibiting the required flux \( v \), by introducing \textit{a posteriori} the constraint \( x(\tau) = \tau v + x_0 \).

On those paths, the rate of a step \( \Delta x \) from position \( x \) at time \( t \) is enhanced by
\[ \eta_{x, x + \Delta x} = \frac{G(x_0 - x - \Delta x, \tau - t - \Delta t)}{G(x_0 - x, \tau - t)} \] (5)

[from Eq. (3)], since the probability of an equilibrium particle achieving the required displacement in the remaining time is given by the Green function for free diffusion. Substituting Eq. (5) into Eq. (2) in the limits \( \tau \gg t \) and \( |x_0| \gg |x| \) yields
\[ p^{\text{driv}}(\Delta x) \rightarrow p^{\text{eq}}(\Delta x) \left[ 1 + \Delta x \frac{v}{2D} - \Delta t \frac{v^2}{4D} \right] \rightarrow p^{\text{eq}}(\Delta x - v \Delta t) \] (6)

where the second line follows by inspection of Eq. (3). Drawing \( \Delta x \) from \( p^{\text{driv}}_{\Delta t} \) in Eq. (1) yields the equilibrium dynamics. So Eq. (1) specifies that substituting \( \Delta x - v \Delta t \) for \( \Delta x \) in the equilibrium equation of motion will yield the dynamics of the driven Brownian ideal gas. Making that substitution, with \( \Delta t \rightarrow 0 \), gives
\[ \frac{dx}{dt} = v + \xi(t)/\zeta \] (7)

where \( \xi \) is the usual Gaussian white noise. This is the equation of motion for driven diffusion that one could easily have written down. But it was not conjectured; it was generated from the Langevin equation for free diffusion, by the sub-ensemble method. In fact, one could write any number of stochastic equations that yield the net drift \( v \tau \) over the fixed duration \( \tau \), e.g. with some
temporal correlations or an additional oscillatory forcing term that integrates to zero. But Eq. (2) specifies a unique solution. Any equation of motion other than the simple one specified by these dynamics would violate the MaxEnt hypothesis, indicating that it introduces unwarranted new information about the paths, additional to the prior dynamics and posterior constraint.

Let us consider a second example, demonstrating that rates respecting DB are not generally correct for driven systems. In a simple 2D model of Brownian particles under continuous shear, a triangular lattice has some fraction of its sites occupied by monomers with nearest-neighbour interactions. An average velocity difference fraction of its sites occupied by monomers with nearest-neighbor interactions. An average velocity difference between the top and bottom boundaries is established by stochastically selecting a horizontal layer $l$ between two rows of the lattice, and shifting all of the system above this layer to the right by one lattice spacing. If the layers are selected with equal probability, these discrete shear transitions will result in a uniform shear rate when observed on large length- and time-scales. In addition to the shear steps, local dynamics consists of choosing a pair $s$ of neighbouring sites at random and swapping them, with a rate $\omega_s$. If one site is occupied by a monomer and the other empty, the transition causes the monomer to hop, and might result in an energy change by making or breaking bonds. Repeating the same swap $s$ recovers the original configuration. In the absence of shear, transition rates respecting DB, e.g. ‘heat-bath dynamics’[1], will correctly generate all static correlations. With shear applied to our model, let us initially violate Eq. (2), and assume that the same equilibrium dynamics is chosen for the local swap transitions.

The model as defined would settle into a driven steady state in the long-time limit and, by tuning the interactions, could be made to exhibit nonequilibrium phase transitions and amusing mesophase structures. But it is unphysical, as becomes apparent with a particular choice of interactions. Let the monomers be coupled in pairs by bonds of infinite strength, to form dimers. A shear transition rotates some dimers (Fig. 1). Any dimer in a north-east south-west orientation prevents shear on the layer that it straddles, e.g. in Fig. 1 further shear on the same layer is disallowed by a dimer that would be broken by the transition. In the thermodynamic limit, the chance of finding a lattice layer that is not blocked by at least one such adversely oriented dimer vanishes. The model cannot be driven to shear.

It is unreasonable for a dimer to wander through configurations under equilibrium-like dynamics, unaware that it is blocking a macroscopic flux. Clearly such dynamics will not work for rigid Brownian dimers. Instead, DB should be violated, even at the local level, so that adversely oriented dimers are pushed out of the way of the applied flux. Even a dimer of finite strength should prefer re-orientation to shear-induced dissociation. The rate of re-orientation is prescribed by SED.

Consider first how SED treats shear steps. A shear step is a ‘type A’ transition as it contributes a quantum of shear flux. In an equilibrium model with both forward and reverse shear steps, the probability of accumulating a large net shear by time $t = \tau$ is tiny. If a forward shear transition $a \rightarrow b$ takes place before $t = \Delta t$, then one step fewer is subsequently required to attain the desired shear, so the numerator of Eq. (3) is larger than the denominator. Thus Eq. (4) correctly prescribes a higher rate for forward shear steps in the driven ensemble than at equilibrium. Similarly, reverse steps are suppressed. As shear steps on different layers $l$ (see Fig. 1) contribute equal flux, they have approximately equal enhancement factors. But their rates $\omega_l^{\text{triv}}$ are not necessarily equal (yielding affine shear) since, by DB, $\omega_l^{\text{triv}}$ in Eq. (2) depends on the energy cost. Hence the driven shear steps are concentrated on the softest layers, e.g. where fewest attractive neighbours will be separated. Rather than imposing affine shear, the driven ensemble, with its weak constraint fixing only the total mean flux, allows authentic inhomogeneous flow.

As well as prescribing the rate of shear steps, Eq. (2) also governs the local site-swapping dynamics. The site-swapping transition $s$ indicated in Fig. 2 is not of ‘type A’, as it contributes no shear flux, but we shall see that its rate is boosted by SED as it is a ‘type B’ transition. In Fig. 2, forward shear is blocked on every lattice layer by adversely-oriented dimers. So there is no chance of immediate shear and, of course, little hope of the desired net flux over the duration $\tau$ of an equilibrium experiment. Hence the denominator of Eq. (3) is very small. The numerator asks for the likelihood of that same shear flux if swap $s$ (represented by $a \rightarrow b$) is first performed. This would rotate an offending dimer, allowing forward shear on the layer shown in bold. A shear step is not guaranteed to follow swap $s$, but its likelihood is greatly increased by the swap. Further shear could follow, until eventually some dimers re-block the layer, returning to a configuration statistically similar to Fig. 2. Hence, if $s$ is performed, the probability of achieving the desired net flux at equilibrium, although small, is many times larger than the denominator of Eq. (3). So SED greatly boosts the rate for swap $s$ in Fig. 2 and the highlighted dimer is quickly ‘pushed out of the way’. Similarly, if transition $s$ were blocked by another dimer it too would be moved by SED, and so on, with correlated chains of events enabling the stipulated mean flux to be realised.

It is startling that SED has generated long-range cou-
plings governing the local swaps, whereas their prior equilibrium dynamics depends only on nearest-neighbour interactions. Because the dimer highlighted in Fig. 2 is alone in blocking its layer of the lattice, it is quickly rotated under applied shear. It feels that there are no other soft planes in the whole lattice that could yield. If there were, the dimer would not feel such an imperative to move, as the enhancement factor would be smaller. We interpret that stress is concentrated at this point. The physics of long-range stress-mediated interactions has arisen naturally.

Unlike the example of the Brownian gas, the Green function for the dimers is unknown, so an exact calculation is not presented. The problem is that evaluation of Eq. (3) appears to require clairvoyance of the full consequences of a proposed transition. However, those consequences need only be forecast for a finite time into the future, exceeding any correlation time, following which the steady state can be assumed. The problem is generally tractable in terms of a cluster expansion. We can speculate that the limit in which such an expansion breaks down (because many correlated particle movements are required for any finite flux probability) might be identified with a jamming transition, just as breakdown of the virial expansion accompanies a critical point.

The method outlined here (detailed elsewhere) puts the simulation of driven steady states on a closer footing to equilibrium numerics, for which any DB-respecting algorithm yields the Gibbs ensemble. By contrast, sheared fluids have hitherto required microscopically accurate simulations, with the associated processing overhead and thermostatting issues. This scheme is not the only route to nonequilibrium transition rates. Models may be defined that violate Eq. (2), but one should then be aware that extraneous information has been introduced, not present in the prior dynamics and macroscopic observables. In this respect the rules set out here have the same status as the principle of DB.

The driven ensemble has been defined here by fixing the time-averaged flux on each path, analogous to fixing energy in an equilibrium microcanonical ensemble. In the limit \( \tau \to \infty \) (analogous to the thermodynamic limit), identical results follow from an alternative ensemble, in which paths are weighted exponentially by their flux, analogous to the canonical ensemble. Details will appear in a longer paper, but note that simply boosting a transition exponentially by the immediate flux that it carries would neglect the subtle non-mean-field time correlations of SED.

I have studied the physical implications of MaxEnt for driven steady states and shown how to implement the resulting dynamics. The approach is unique in deriving dynamical rules for a driven system under the stochastic influence of a reservoir, without requiring any approximate coarse-graining or near-equilibrium assumption. The rules have yielded a rich variety of correct physics for driven diffusion and Brownian dimers under shear. Many quiescent systems are well approximated by the laws of canonical equilibrium, but exceptions include glasses, granular media, and some cellular automata. Similarly, not every nonequilibrium steady state will respect the conditions presented here, but those that do are expected to form a large and significant class.

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