Generalized Entropy approach to far-from-equilibrium statistical mechanics.

Alexei V. Tkachenko
Bell Labs, Lucent Technologies, 600-700 Mountain Ave., Murray Hill, NJ 07974
E–mail: alexei@bell-labs.com

We present a new approach to far–from–equilibrium statistical mechanics, based on the concept of generalized entropy, which is a microscopically-defined generalization of Onsager-Machlup functional. In the case when a set of slow (adiabatic) variables can be chosen, our formalism yields a general form of the macroscopic evolution law (Generalized Langevin Equation) and extends Fluctuation-Dissipative Theorem. It also provides for a simple understanding of recently–discovered Fluctuation Theorem.

PACS numbers: 05.70.Ln, 05.20.-y

Foundations of non-equilibrium statistical physics remain in focus of intensive research for nearly a century. Over past several decades, there has been a dramatic progress in application of stochastic equations to wide variety of complex systems, as well as in understanding some of their generic features. However, the methods of microscopic derivation of such coarse-grained descriptions (e.g. starting with a Hamiltonian), remain essentially limited to traditional kinetic theory and linear response scheme [1, 2]. In this work, we emphasize the reductionistic mission of non-equilibrium statistical mechanics, by building a constructive formalism whose conceptual framework resembles that of equilibrium theory. Our approach is not limited to the vicinity of thermal equilibrium, and becomes equivalent to the classical linear response theory in this limit.

The central concept in our discussion is Generalized Entropy (GE), which goes back to works by Onsager and Machlup [3, 4]. Further development of this paradigm includes its generalization for far-from-equilibrium case by Graham [5] and action-principle approach to Markovian stochastic dynamics by Eyink [6, 7], whose technique and conclusions have many common points with ours. The distinct feature of the present approach is an explicit microscopic interpretation of GE, which enables us to derive the macroscopic evolution equations starting from a microscopic level. Normally, entropy is assigned a macroscopic state of a system as the logarithm of its statistical weight. Similarly, we introduce GE, as a logarithm of the statistical weight of a given macroscopic evolution (trajectory). We limit ourselves to the class of microscopic dynamics (parameterized with microvariables \( q_j, j = 1...N \)) which are deterministic and phase-volume-preserving (as e.g. in Hamiltonian systems). In this case, there is a natural measure in the space \( q_j \), i.e. the statistical weight of any subset of the space is its phase volume. Let \( \mathbf{A} = \mathbf{A}(q) \) be a macroscopic (possibly, multicomponent) variable. The formal definition of the GE associated with a given evolution \( \mathbf{A}(t) \) between two moments of time, \( t_i \) and \( t_f \), is the following:

\[
S[\mathbf{A}(t)]_{t_i}^{t_f} \equiv \log \left[ \int D[q_f] \prod_{t=t_i}^{t_f} \delta \left( \frac{d}{dt} \mathbf{A}(q(t)) - \dot{\mathbf{A}}(t) \right) \right], \quad (1)
\]

Here, integration is performed over all actual microscopic trajectories \( q_f(t) \), i.e. those satisfying the microscopic equations of motion. Since we assume the deterministic and phase–volume–preserving evolution on that level, all the actual trajectories are equally weighted.

Our next step is to distinguish between two contributions to GE: \( S[\mathbf{A}(t)]_{t_i}^{t_f} = S^{(0)}[\mathbf{A}] + S^{(kin)}[\dot{\mathbf{A}}] \) \((1)\). The first term, \( S^{(0)} \), is the logarithm of the total number of trajectories of length \( \Delta t \equiv t_f - t_i \) with a fixed "midpoint" \( \mathbf{A} \equiv \left( \mathbf{A}_{t_i} + \mathbf{A}_{t_f} \right)/2 \). The other contribution, \( S^{(kin)} \), to which we will refer as kinetic entropy, is the logarithm of the probability of a given evolution \( \dot{\mathbf{A}}(t) \) for the fixed \( t_i, t_f \) and \( \mathbf{A} \):

\[
S^{(kin)}[\dot{\mathbf{A}}(t)]_{t_i}^{t_f} \equiv \log \left( \prod_{t=t_i}^{t_f} \delta \left( \frac{d}{dt} \mathbf{A}(q(t)) - \dot{\mathbf{A}}(t) \right) \right)_{\mathbf{A}} = \log \left( \int D[\mathbf{X}(t)] \exp \left( \Sigma[\mathbf{X}(t)] - \int_{t_i}^{t_f} \mathbf{X}(t) \cdot \dot{\mathbf{A}}(t) dt \right) \right). \quad (2)
\]

Here we have transformed the expression by using the exponential representation of \( \delta \)-function, which resulted in introduction of a new variable \( \mathbf{X} \) conjugated to \( \dot{\mathbf{A}} \). \( \Sigma \) is the generating functional for the variable \( \dot{\mathbf{A}} \), which in particularly means that its variation of any order in \( \mathbf{X} \) coincides with the corresponding irreducible correlator of the conjugated field (in our case, \( \dot{\mathbf{A}} \)):

\[
\Sigma[\mathbf{X}(t)] \equiv \log \left( \exp \left( \int_{t_i}^{t_f} \mathbf{X}(t) \cdot \frac{d}{dt} \dot{\mathbf{A}}(q(t)) dt \right) \right)_{\mathbf{X}} = \log \left( \exp \left( \int_{t_i}^{t_f} \mathbf{X}(t) \cdot \frac{d}{dt} \dot{\mathbf{A}}(q(t)) dt \right) \right)_{\mathbf{X}} = \left( \int dt_1...dt_n \langle \dot{\mathbf{A}}(t_1)...\dot{\mathbf{A}}(t_n) \rangle \mathbf{X}(t_1)...\mathbf{X}(t_n) \right). \quad (3)
\]

Here \( \langle \rangle \) is the irreducible part of the correlator, i.e. the n-point average with subtracted contribution reducible to the lower–order correlation functions. Here and in the future \( \dot{\mathbf{A}}(t_1)\dot{\mathbf{A}}(t_2) \) denotes direct tensor product, which is to be distinguished from scalar one, e.g.
$X(t) \cdot \dot{A}(t)$. The averaging is performed over all trajectories with the given initial and final times $(t_i, t_f)$ and fixed midpoint, $\overline{A}$.

Now, after we have related kinetic entropy $S^{\text{kin}}$, to the statistics of $\dot{A}$, it becomes possible to express another contribution to GE, $S^{(0)}(\overline{X})$, in terms of regular thermodynamic entropy $S(A)$. For doing so we notice that integration of the statistical weights of all trajectories originating from a given point $A$ is, by definition, the weight of the initial state, $\exp S(A)$. After making a simple calculation based on this observation, one gets $S(A) = S^{(0)}(A) + \Sigma \left[\frac{1}{2} \delta S^{(0)}/\delta A\right]$. The time interval $\Delta t$, and the associated change of $A$ is assumed to be sufficiently small so that the linear expansion of $S^{(0)}$ in $\Delta A$ be valid.

In a general case, $\Sigma$–functional is an awkward mathematical object, because of its non–local structure. However, until this point we have not restricted the choice of the macroparameters, $A$. From the practical point of view, it is clear that the coarse–grained description of a system may be reasonable if one can choose a set of relatively slow variables as the macroparameters. Below we specify this choice in a more formal way.

By definition, the form of $\Sigma$–functional (and the correlators of $\dot{A}$) depends on the midpoint position, $\overline{A}$. However, there naturally exists an interval $\delta A$ within which such dependence can be neglected. This allows one to introduce a concept of drift time, $\tau_d$, over which most of the trajectories remain within this interval of constant statistics of $\dot{A}$. Suppose there exists a shorter time scale, $\tau_0 \ll \tau_d$, such that any correlator $\langle \dot{A}(t_1) \ldots \dot{A}(t_n) \rangle$ becomes negligible when $|t_1 - t_n| > \tau_0$. In this case, one can choose the initial and final times such that $\tau_0 \ll \Delta t \equiv t_f - t_i \ll \tau_d$. This considerably simplifies the expression for $\Sigma$–functional: if we are only interested in the behavior of the system on times larger than $\tau_0$, $\Sigma$ becomes local in $X$:

$$\Sigma[X(t)] = \int_{t_i}^{t_f} \Xi(X(t)) dt,$$

here

$$\Xi(X) = \sum_{n=1}^{\infty} \frac{X^n}{n!} \int dt_1 \ldots dt_{n-1} \langle \dot{X}(0) \ldots \dot{X}(t_{n-1}) \rangle. \quad (5)$$

We will refer to time scale $\tau_0$ as ergodicity time. It can be shown that from the methodological point of view, the assumption of the existence of such time scale does play a role similar to that of the ergodic principle in equilibrium theory.

Collecting the above results gives the following expression for GE:

$$S[A(t)], X(t)] = \int_{t_i}^{t_f} dt \left[ A \cdot \left( \frac{1}{2} \frac{\delta S}{\delta A} - X \right) + \Xi(X) - \Xi \left( \frac{1}{2} \frac{\delta S}{\delta A} \right) \right]. \quad (7)$$

The first contribution here is the conventional entropy of the initial state, and the other one is the logarithm of the probability of a given evolution starting at that point. The latter is expressed in terms of the functional $S'$, which has a meaning of GE in $(A, X)$ space:

$$S'[A(t), X(t)] = \int_{t_i}^{t_f} dt \left[ A \cdot \left( \frac{1}{2} \frac{\delta S}{\delta A} - X \right) + \Xi(X) - \Xi \left( \frac{1}{2} \frac{\delta S}{\delta A} \right) \right]. \quad (7)$$

This quantity becomes additive on the time scales exceeding $\tau_0$, which means that the dynamics becomes Markovian. This fact allows us to extend the applicability of the above expression to the case when $(t_f - t_i) > \tau_d$, i.e. when $\Xi(X)$ is no longer independent of $A$.

In a particular case of reversible microscopic dynamics, $\Xi$ does not change if the sign of $X$ reversed. This implies that the ratio of probabilities of direct and reversed evolutions along the same path $A(t)$ is independent of the form of $\Sigma$–functional and is given by $\exp(A(t_f) - A(t_i))$ (see (4)-(5)). This property is known as Fluctuation Theorem [4], which has been recently established for a wide variety of non-equilibrium systems.

One can eliminate the “fictitious” variable $X$ from Eqs. (4)-(5):

$$S[A(t)], X(t)] = \int_{t_i}^{t_f} dt \left[ A \cdot \left( \frac{1}{2} \frac{\delta S}{\delta A} - X \right) + \Xi(X) - \Xi \left( \frac{1}{2} \frac{\delta S}{\delta A} \right) \right],$$

here the kinetic entropy, which is now a local functional, has been expressed as a time integral of pseudo–Lagrangian $A$:

$$S^{(\text{kin})}[A(t)] = \int_{t_i}^{t_f} dt \Delta A(t) dt = \int_{t_i}^{t_f} dt \left[ \Delta A(t) - \Xi \left( \frac{1}{2} \frac{\delta S}{\delta A} \right) \right], \quad (9)$$

Because of the obvious analogy with classical mechanics, we will refer the above $(A, X)$ and $(A, \dot{A})$ forms for GE as pseudo–Hamiltonian and pseudo–Lagrangian ones, respectively. Though they are completely equivalent, the pseudo–Hamiltonian formalism requires introduction of additional variables $X$ (which plays the role of momentum conjugated to measurable $A$), while the pseudo–Lagrangian form obscures the relationship between the conjugated functions $A$ and $\Xi$, given by Eq. (4).

In the case of a distributed system, when both $A$ and $X$ are fields, $A$ and $\Xi$ would typically become local functionals. If there is a global conservation law for one or several components of $A$, the $\Sigma$–functional, Eq. (3) is invariant with respect to the global transformation $X(r) \to X(r) + \sum_{\alpha} \delta^{(\alpha)} n^{(\alpha)}$ (here index $\alpha$ counts all.
the conserved components of field $\mathbf{A}$, $\mathbf{n}^{(\alpha)} \cdot \mathbf{A}$). Existence of this transformation means that $\Sigma$ and $\Xi$ should depend only on gradients (and higher spatial derivatives) of the corresponding components of the field $\mathbf{X}$: $\Sigma = \int \xi (\nabla \mathbf{X}^{(\alpha)}) dtdr^\tau$. If the microscopic fluxes $\mathbf{j}^{(\alpha)}$ of the conserved parameters can be introduced, the expansion of $\xi$ in powers of $\nabla \mathbf{X}^{(\alpha)}$ is given by the form similar to Eq. [5], with all the correlators of $\dot{\mathbf{A}}$ replaced with those of microfluxes $\mathbf{j}^{(\alpha)}$.

We now discuss the dynamics of the system in the deterministic limit, which corresponds to the settle point of the GE functional. One has to emphasize that although the structure of the functional is similar to regular action, the result of the variation procedure is dramatically different from that in mechanics. In our case, only initial point $\mathbf{A}(t_i)$ should be kept fixed, while the initial rate $\dot{\mathbf{A}}(t_i)$ (or, equivalently, the final point $\mathbf{A}(t_f)$), is subject to optimization. As a result, the current value of $\mathbf{A}$, rather than the pair $(\mathbf{A}, \dot{\mathbf{A}})$ determine the future dynamics of the system, and the equation of motion is of the first, rather than the second order. Namely, maximization of the pseudo-Lagrangian form of GE, Eq. (8), with respect to $\dot{\mathbf{A}}(t_f)$, yields the following dynamic law:

$$\frac{\delta \Lambda (\dot{\mathbf{A}})}{\delta \mathbf{A}} = \frac{1}{2} \frac{\delta S}{\delta \dot{\mathbf{A}}}. \tag{10}$$

This equation can be interpreted as a balance between thermodynamic driving force (the right hand side) and the dissipative force (the apparent physical meaning of the left hand side). An alternative description can be obtained by variation of the functional in Eq. (8) with respect to $\dot{\mathbf{A}}$ and $\mathbf{X}$:

$$\dot{\mathbf{A}} = \frac{\delta \Xi}{\delta \mathbf{X}} \bigg|_{\mathbf{X}=\frac{\delta S}{\delta \mathbf{A}}}. \tag{11}$$

The equation shows how the system moves in response to the thermodynamic driving force, which in the deterministic limit appears to be identical to variable $\mathbf{X}$.

In the case of conserved components of $\mathbf{A}$-field, Eq. (11), will be replaced with regular continuity equation, $A^{(\alpha)} = -\nabla \cdot \mathbf{j}^{(\alpha)}$, in which the macroscopic flux is given by constitutive equation, $\mathbf{j}^{(\alpha)} = \xi (\nabla \mathbf{X}^{(\alpha)})$. It is interesting to note that the macroscopic flux can be introduced even if there is no well-defined fluxes on microscopic scale: this concept follows from the spatial locality of $\Sigma$-functional and conservation of quantity $\int A^{(\alpha)} dtdr^\tau$.

“Kinetic potentials” $\Lambda$ and $\Xi$ in the deterministic limit are related through Lagender transform, i.e. $\Xi (\mathbf{X}) = \Lambda (\dot{\mathbf{A}}) + \mathbf{X} \cdot \dot{\mathbf{A}}$, where $\mathbf{X} = -\delta \Lambda (\dot{\mathbf{A}})/\delta \dot{\mathbf{A}}$. In the vicinity of equilibrium the driving forces are small, and therefore only leading terms in expansion of $\dot{\mathbf{A}}$ remain relevant: $\Xi = \langle \dot{\mathbf{A}} \rangle \cdot \mathbf{X} + \frac{1}{2} \Gamma^{(2)} \cdot \mathbf{X}^2$. After substituting this expression for $\Xi$ into equation of motion (11), we recover the classical linear response result: $\dot{\mathbf{A}} = \langle \dot{\mathbf{A}} \rangle + \frac{1}{2} \Gamma^{(2)} \cdot \delta S/\delta \mathbf{A}$, i.e. the dissipative contribution to $\dot{\mathbf{A}}$ is proportional to the thermodynamic driving force, and our result for the corresponding kinetic coefficient (see (5)) coincides with the one given by Fluctuation–Dissipative Theorem (FDT). It is a general practice to assume the same linear rate-force relationship even in the regime in which entropy (free energy) is no longer a harmonic function of deviations from equilibrium. Langevin equation (8) is one of the most noticeable examples of such an approach. Furthermore, Öttinger et al. [6] have recently proposed an elegant unified way of representing most of the known stochastic models in a single generic form, which again assumes a linear rate-force relationship for dissipative dynamics. Our Eq. (11) (or (10)) in a general case would result in a nonlinear relationship between them and, in this sense, can be referred to as Generalized Langevin Equation (GLE) (an additional noise term will be discussed below). The fact that a particular form of the evolution equation depends on the correlators of macroparameters, suggests a possibility for a synergy between our scheme and earlier field–theoretical approaches to non-equilibrium statistical mechanics [1].

In order to demonstrate how GLE works outside the linear regime, we consider a trivial kinetic problem: an ensemble of independent two–state systems, each of which has the same transition rate $\kappa$ in either direction. The relaxation dynamics for the population difference between the two states, $N_- \equiv N_1 - N_2$ is given by equation $\dot{N}_- = -2\kappa N_-$. Although linear, it is not a linear response result. The thermodynamic driving force conjugated to $N_-$ is the chemical potential difference between the two states, i.e. in our notations $2X_- = \partial S/\partial N_- = \log N_2/N_1$. This expression can be linearized in $N_-$ (for constant $N_+ \equiv N_1 + N_2$) only sufficiently close to equilibrium, i.e. when $N_- \ll N_+$. This means that the simple linear kinetic equation is a result of a non-linear dependence of the response $\dot{N}_-$ on the driving force $X_-$. In the considered case, one can calculate $\Xi (X_-)$ exactly. In the limit of large $N_+$, the number of switches happening over small time $\delta t$ is $N_+ \delta t$. Since all the switches are completely uncorrelated, the original formula for $\Sigma$–potential, Eq. (8), results in the following expression for $\Xi (X)$: $\Xi = N_+ \kappa \log[cosh(2X_-)]$. Here we have taken into account the fact that the change of population difference, $\delta N_-$ is either 2 or −2 for any individual switch and the both possibilities are equally weighted. Indeed, any microscopic trajectory (sequence of individual switches) can be reversed, and this does not change the position of its midpoint, $\overline{N}_- \equiv (N_- (t_i) + N_- (t_f))/2$. After using Eq. (11), we recover the expected linear equation for $N_-$. We now proceed with the discussion of fluctuations around the deterministic dynamics. Let $\mathbf{A}^{(0)} (t)$ be a solution to equation of motion (11) and $\mathbf{a} (t) \equiv \mathbf{A} (t) - \mathbf{A}^{(0)} (t)$ is the deviation of an actual trajectory from it ($\mathbf{a} (t_i) = 0$). The corresponding deviation of the gener-
alized entropy from its local maximum is given by the following quadratic expression:

\[
\delta S[a(t)] = -\int_{t_i}^{t_f} \frac{\dot{\lambda} - 1}{2} \left( \dot{a} + \dot{\lambda} \cdot \hat{\kappa} \cdot a \right)^2 dt = -\frac{1}{2} \int a_{\omega} \cdot \left( \omega^2 \dot{\lambda}^{-1} + \hat{\kappa} \cdot \dot{\lambda} \cdot \hat{\kappa} \right) \cdot a_{\omega} \frac{d\omega}{2\pi},
\]

(12)

here \(\dot{\lambda} \equiv \delta^2 \Xi / \delta X^2\) and \(\hat{\kappa} \equiv \frac{1}{2} \delta^2 S / \delta A^2\). Note that the deterministic trajectory is stable only if the response matrix \(\dot{\lambda} \cdot \hat{\kappa}\) is positive-definite. Otherwise, any two trajectories originating from the same point diverge exponentially fast. By definition, \(\exp(\delta S[a(t)])\) is the statistical weight of a given trajectory. Therefore, the above quadratic functional, Eq. (12), corresponds to Gaussian statistics of \(a\), with \(\langle a_{\omega} a_{-\omega} \rangle = \left( \omega^2 \dot{\lambda}^{-1} + \hat{\kappa} \cdot \dot{\lambda} \cdot \hat{\kappa} \right)^{-1}\). Equivalently, this result can be represented by introduction of a random Gaussian noise \(\eta(t)\) to the second equation of (11), with \(\langle \eta(t) \eta(t') \rangle = \dot{\lambda} \delta(t - t')\).

The fact that the same matrix \(\delta^2 \Xi / \delta X^2\) controls both the strength of the fluctuations and the response to a small variation in driving force (see (12) or (11)), is a signature of FDT [1] [2], which is conventionally applied only in linear-response regime. In order to extend FDT to our far-from-equilibrium case, we probe the system with a random Gaussian noise \(\eta(t)\) to the second equation of (11), with \(\langle \eta(t) \eta(t') \rangle = \dot{\lambda} \delta(t - t')\).

The perturbation results in adding a coupling term \(\sum_{\omega} i \omega h_{\omega} a_{\omega} / 2\) to functional \(\delta S\), Eq. (12). The response now can be determined by variation of \(\delta S\) with respect to \(a_{\omega}\). An important aspect of this procedure is that the causality principle should be taken into account: in \(\hat{a}\)-representation, \(a(t)\) should be varied with the specified past and unknown future (as we did while deriving equations of motion, (10) and (11)). In \(\omega\) representation, the response is given by

\[
\hat{\Gamma}(\omega) \equiv \frac{\delta a_{\omega}}{\delta h_{\omega}} = \frac{i \omega}{2} \langle a_{\omega} a_{-\omega} \rangle_R.
\]

(13)

Here index \(R\) stands for the retarded part of the correlator, i.e. the one with all poles at \(\Im(\omega) > 0\) half-plane. The above relationship extends the classical FDT towards strongly non-equilibrium regime. It has to be stressed that \(\hat{\Gamma}(\omega)\) does determine the response to small perturbations but, in contrast to linear response regime, it does not relate the total driving force \(\delta S/\delta X\) to the evolution rate \(\dot{A}\) (the relationships is given by non-linear Eq. (10) or (11)).

Surprisingly enough, the applicability of FDT might be extended even further, to time scales comparable or shorter than ergodicity time \(\tau_0\). Namely, if the frequencies of interest are considerably higher than the characteristic relaxation rates (eigenvalues of relaxation matrix \(\dot{\lambda} \cdot \hat{\kappa}\)), we may neglect the dependency of the the driving force on \(a\). In this case, the generalized entropy in \((a, x)\)-representation can be written in the following form:

\[
S' = \sigma[x(t)] - \sigma[h(t)/2] + \int_{t_i}^{t_f} \dot{a} \cdot \left( \frac{h(t)}{2} - x(t) \right) dt.
\]

(14)

Here \(x(t)\) is the deviation of \(X\) from its deterministic value \(X_0\), and \(\sigma[x(t)] \equiv \Xi[X_0 + x(t)] - \Xi[X_0]\) is the corresponding deviation of \(\Sigma\)-functional, which is no longer assumed to be local, i.e. \(\omega \tau_0\) is not small. In this regime, to which one may refer as sub-ergodic, the statistics of \(a\) need not to be Gaussian.

It is easy to show that \(\sigma[x(t)]\) is the generating functional for \(\hat{a}\), i.e. its variations coincide with the corresponding correlators of \(\hat{a}(t)\). On the other hand, in accordance with (14), these variations of \(\sigma\) determine the response of the system in any order of \(h\), \(\hat{\Gamma}^{(n)}(\omega_1, ..., \omega_n) \equiv \delta^n a_{\sum \omega_k} / \delta h_{\omega_1} ... \delta h_{\omega_n}\). This results in the following extension of FDT to sub-ergodic time scales:

\[
\hat{\Gamma}^{(n)}(\omega_1, ..., \omega_n) = \langle (a_{-\sum \omega_k} a_{\omega_1} ... a_{\omega_n}) \rangle_R \prod_{k=1}^{n} \frac{i \omega_k}{2}.
\]

(15)

This version of FDT is remarkable: (i) it may be applicable to systems with considerable memory effects, e.g. glasses; (ii) it establishes the relationship between the non-linear response of a system and the deviation of its fluctuation statistics from Gaussian. The above relationship is very similar to recent results for Markovian stochastic processes, [6]. It should be noted that the direct experimental or numeric check of Eq. (13) may be difficult to perform without correct interpretation of field \(h\) (its physical meaning is straightforward only in the case of ideal (fast) thermal bath coupled to the system). In particular, such interpretation may involve frequency-dependent temperature [4].

In conclusion, we have proposed a framework for construction of non-equilibrium macroscopic theory of a complex system, starting with its fundamental non-dissipative dynamics. This approach assumes the possibility of choosing a set of relatively slow (adiabatic) variables. Our major results include the general form of equation of motion of the system under a given thermodynamic driving force (GLE) and extended FDT. Among the immediate possible applications of our scheme is the development of non-equilibrium statistical theories of various complex systems starting with their model Hamiltonians, such as Heisenberg and XY models [12] [2], or Gross-Pitaevsky model of Bose condensate. Another intriguing direction of the development of the generalized entropy approach is its use for Landau-type description of bifurcations. It also provides us with an apparatus to study the problem of kinetic tunneling between various steady states (attractors) of a non-
equilibrium system. A natural extension of our theory would be its quantum generalization.

Acknowledgement The author is grateful to T. Witten, P. Cvitanovic, A. Sengupta, B. Shraiman, Y. Rabin, C. Varma, E. Balkovski for useful discussions.

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