Spatial Distributions of Observables in Systems under Thermal Gradients

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Departures of observables from their thermal equilibrium expectation values are studied under heat flow in steady-state non-equilibrium environments. The relation between the spatial and temperature dependence of these non-equilibrium behaviors and the underlying statistical properties are clarified from general considerations. The predictions are then confirmed in direct numerical simulations within the FPU-β model. Non-equilibrium momentum distribution functions are also examined and characterized through their cumulants and the properties of higher order cumulants are discussed.

I. INTRODUCTION

In studies of non-equilibrium physics, especially those of steady states, local equilibrium is most often invoked and this assumption simplifies calculations through the use of equilibrium statistical mechanics and thermodynamics. The local equilibrium assumption allows the use of the equilibrium distribution function to compute observables. If local equilibrium conditions are not assumed, very little can be computed analytically and even the definition of temperature is no longer unique. Efforts have been made to quantify the goodness of local equilibrium assumptions or how transport coefficients differ from their linear response values, though only few quantitative studies exist. Without the knowledge of the non-equilibrium steady-state distribution, theoretical development becomes quite restrictive. We explore how observables depart from their equilibrium expectation values within a given non-equilibrium steady-state, specifically focusing on the spatial dependence of the non-equilibrium expectation values within a given system and their local temperature dependence. To make this concrete, heat flow in the FPU-β model is simulated to test the predictions. We further quantitatively examine the relationship between the momentum cumulants and the distribution and find that the lower order cumulants characterize the distribution quite well.

For systems in thermal gradients, it is natural to consider how an observable \( O \) in the non-equilibrium steady state departs from its equilibrium value, denoted \( O_{eq} \). The normalized deviation from equilibrium, when \( O_{eq} \neq 0 \), can be expanded as

\[
\delta_O = \frac{\delta O}{O} = \frac{O - O_{eq}}{O_{eq}} = C_O \left( \frac{\nabla T}{T} \right)^2 + C'_O \left( \frac{\nabla T}{T} \right)^4 + \cdots
\]  

(1)

When \( O_{eq} = 0 \), as is the case for higher order momentum cumulants, one can normalize by an observable which has the same dimensions. When local equilibrium is no longer valid, in general, no unique definition of temperature exists and a choice needs to be made. This definition of non-equilibrium temperature can be thought of as a choice of a coordinate system, on which the physics behavior of the system will not depend. If we assume analyticity in \( \nabla T \), the deviations \( \delta_O \) can be expanded in even powers as above. We shall see below that this expansion is adequate for describing the properties of the system.

The heat flow, \( J \), is the flow of energy and can be unambiguously defined in Hamiltonian systems. Near equilibrium, it satisfies Fourier’s law locally as \( J = -\kappa \nabla T(x) \), where \( \kappa \) is the thermal conductivity, \( T(x) \) is the temperature profile inside, and \( x \) is the position inside the system. Fourier’s law can be used to re-express the local departures from equilibrium in terms of the temperature profile \( T(x) \), or equivalently the position \( x \) once the coefficients \( C, C' \) are known since \( J \) does not depend on \( x \):

\[
\delta_O = C_O \left( \frac{J}{\kappa(T)T} \right)^2 + D'_O \left( \frac{J}{\kappa(T)T} \right)^4 + \cdots
\]  

(2)

We note that Fourier’s law itself receives non-equilibrium corrections, which is why the coefficient of \( O(J^4) \) term in the expansion differs from that of \( \delta_O \). In the following, the objectives will be to make the formula more explicit.
and understand its physical properties under rather general assumptions. This relation, together with \( \kappa(T) \) (and consequently \( T(x) \)) provides the basis for defining how non-equilibrium observables vary inside a finite system both near and far from global thermal equilibrium.

II. THE FPU MODEL AND TEMPERATURE PROFILES

The results we present here are derived from general considerations and we develop them in conjunction with a model in which they can be explicitly analyzed. We study the FPU \( \beta \) Hamiltonian, defined generally in the form

\[
\hat{H} = \sum_{k=0}^{L} \left[ \frac{\hat{p}_{k}^2}{2m} + \frac{1}{2}m\omega^2(\hat{q}_{k+1} - \hat{q}_{k})^2 + \frac{\beta}{4}(\hat{q}_{k+1} - \hat{q}_{k})^4 \right].
\]  

We use the FPU model since its physical properties are of wide interest ([12, 13, 14, 17] and references therein). Also as the model is well studied, we can understand the physical properties we find within a larger physics context. Under the rescaling \( \hat{p}_{k} = \hat{p}_{k}/\sqrt{\beta}, \hat{q}_{k} = \hat{q}_{k}/\sqrt{\beta} \), we obtain the conventional form of the FPU \( \beta \) model,

\[
H_{\beta} = \frac{1}{2} \sum_{k=0}^{L} \left[ \hat{p}_{k}^2 + (\hat{q}_{k+1} - \hat{q}_{k})^2 + \frac{\beta}{2}(\hat{q}_{k+1} - \hat{q}_{k})^4 \right],
\]  

where \( H_{\beta} = \hat{H}/(m^2\omega^4) \). We note that in finite temperature simulations, changing the temperature is equivalent to changing the coupling \( \beta \). Under the additional rescaling \( \hat{p}_{k} = p_{k}/\sqrt{\beta}, \hat{q}_{k} = q_{k}/\sqrt{\beta} \), one obtains a unique, dimensionless, Hamiltonian \( H = H_{\beta=1} = \beta H_{\beta} \), which we shall use without any loss of generality. Since \( p_{k}^2 = \beta p_{k}^2 \), the temperatures in the two formulations \( H \) and \( H_{\beta} \) are related by \( T = \beta T' \).

In this work, we study the non-equilibrium steady state physics of the theory under thermal gradients, making use of non-equilibrium states constructed numerically. (For general discussion, see, for instance, [14, 17].) The model is thermostatted at the boundaries \( k = 0, L \) at various temperatures \( T_{1}^{0}, T_{2}^{0} \), using the generalized versions of Nosé–Hoover thermostats as detailed in [18]. These additional thermostat degrees of freedom are added only at the boundaries and the degrees of freedom inside the system \( (0 < k < L) \) are exclusively those of the Hamiltonian [14]. By numerically integrating the equations of motion of the whole system (including those of the thermostats), we obtain the behavior of physical observables in the non-equilibrium steady state by averaging over time, in the standard manner [17]. The local temperature at site \( k \) is defined as \( T_{k} = \langle \hat{p}_{k}^2 \rangle \). In this work, we study the physics inside the system, away from the boundaries by much more than the mean free path of the system [12]. The sensitivity of the results to the manner in which we apply the boundary conditions — including both the number of thermostats and the strength of the couplings — have been examined to ensure that physics results below remain independent of their implementation. (The only exceptions are the boundary jumps in temperature which we discuss below.)

The numerical integrations were performed using the fourth order Runge-Kutta routines with time steps of \( 0.005 \sim 0.02 \) for \( 10^7 \sim 10^{10} \) time steps. The equilibrium properties have been readily verified with this method [14, 18].

In Fig. 1 some examples of temperature profiles for the FPU theory are shown. Generically, there are temperature jumps just inside the boundaries with smooth temperature variations within. The boundary jumps become larger as one moves away from global equilibrium. The jumps are dynamical in the sense that they depend on the model, the transport coefficient, heat flow, as well as the type of boundary conditions employed. The temperatures at the boundaries are at the thermostat temperatures to high degree of precision. For instance, in the examples of Fig. 1 the boundary temperatures are equal to the prescribed thermostat temperatures to within few in \( 10^5 \) relatively.

From temperature profiles and heat flow calculations, Fourier’s law can be verified to hold up to corrections of the form [11, 18], and the thermal conductivity, \( \kappa \), can be obtained for a given temperature and system size. In the 1-d FPU model, \( \kappa \) depends on the system size \( L \) and does not display bulk behavior [13]. \( \kappa \) is also dependent on the temperature in a known manner [17]. Generally, in cases where we have a one dimensional temperature gradient, the temperature profiles can be obtained by integrating Fourier’s law as long as we are not too far from equilibrium [11, 12, 16]:

\[
\int_{T_{1}}^{T(x)} \kappa(T) \, dT = -Jx, \quad J = -p_{k} \left[ (q_{k+1} - q_{k}) + (q_{k+1} - q_{k})^3 \right]
\]

\( x \) is the continuum extrapolation of the discrete lattice index \( k \). We note here that \( J \) is a constant within the system for a given set of temperature boundary conditions since there are no heat sinks or sources inside. \( T_{1} \) in the integral is the temperature extrapolated to the boundary and is explained below.
FIG. 1: Some examples of temperature profiles for the FPU model with $L = 128$. The thermostat temperatures at the boundaries are $(T^0_1, T^0_2) = (0.88, 16.72), (2.4, 15.2), (4.4, 13.2), (6.6, 11.0)$ for the four thermal profiles. The profiles predicted from Eq. (7) are indicated by × and agree well with the results from the numerical simulations.

In many situations, the temperature dependence of the thermal conductivity, within some temperature range, can be well described by

$$\kappa(T) = c T^{-\gamma}. \quad (6)$$

While this power law may not hold globally in $T$, it is often the case that it is sufficient for the region of interest, which is the case here. In such a situation, the temperature profile can be explicitly computed from (5) to be

$$T(x) = \begin{cases} T_1 \left[ 1 - \left( \frac{T_2}{T_1} \right)^{1-\gamma} \right]^{1/\gamma}, & \gamma \neq 1 \\ T_1 \left( \frac{T_2}{T_1} \right)^{\gamma/L}, & \gamma = 1 \end{cases} \quad (7)$$

Here, $T_{1,2}$ denote the boundary temperatures obtained by extrapolating the temperature profile inside the system and differs from the thermostat temperatures $T^0_{1,2}$ by the boundary temperature jumps. From (5) and (6), the temperatures $T_{1,2}$ are found to obey a relation

$$\frac{JL}{c} = \frac{T_{2}^{1-\gamma} - T_{1}^{1-\gamma}}{1-\gamma} \quad (8)$$

To understand the temperature profile of the whole system, we further need an understanding of the temperature jumps at the boundaries[20]. Similar boundary slips have been seen in sheared systems and these effects have been known for a long time in real systems. To leading order, the temperature jumps can be described by (with $n$ being the normal to the boundary)

$$|T_i - T^0_i| \simeq \frac{\alpha c}{L(1-\gamma)} \left[ T_{2}^{1-\gamma} - T_{1}^{1-\gamma} \right] \sim \lambda \frac{\partial T}{\partial n}, \quad (i = 1, 2) \quad (9)$$

Here $\lambda$ is the mean free path of the excitations, which for the FPU lattice model, is essentially the $\kappa(T)$ (up to a constant factor of order one) due to kinetic theory arguments[14]. $\alpha$ reflects the efficacy of the boundary conditions. The last relation is obtained by using Fourier’s Law and (8). The jumps on the hot and cold side are the same provided the system is reasonably close to equilibrium. The jumps at the boundaries and the temperature profile within describe the temperature profile of the complete system. The predicted values for the temperature profiles are plotted in Fig. 1 at a number points inside the systems (× symbols) away from the boundaries and are seen to be consistent with the simulation results. The thermal conductivity is roughly constant with respect to the temperature in this region so that $\gamma = 0$ was used in the profile calculations. This demonstrates that all aspects of the non-equilibrium temperature profile can be quantitatively captured through (7) and (9), irrespective of whether $\kappa(T)$ is a power law in temperature for all $T$ or not. With this understanding of $T(x)$ we can now turn to the question of general observables.
distribution since the temperature profile is known and can be understood as in (7). Theory\[18\] for comparison.

In this case in thermal equilibrium, the behavior of the observables have been seen to be well described by (1)\[2, 3\]. The cumulants of the observables are well defined local variables and their values in local equilibrium are known precisely. The low order cumulants are defined as

\[ \langle\langle p^2 \rangle\rangle = \langle p^2 \rangle, \quad \langle\langle p^4 \rangle\rangle = \langle p^4 \rangle - 3\langle p^2 \rangle^2, \quad \langle\langle p^6 \rangle\rangle = \langle p^6 \rangle - 15\langle p^2 \rangle^3 + 30\langle p^2 \rangle^3, \ldots \]  

where, in equilibrium,

\[ \langle\langle p^2 \rangle\rangle_{eq} = T, \quad \langle\langle p^4 \rangle\rangle_{eq} = 0 \quad (n \neq 2) \]  

This property is also of practical importance. Since the deviations we compute can be small, it is desirable to use observables whose local equilibrium values are known exactly. In this case in thermal equilibrium, \( \mathcal{O}_{eq} = 0 \), so we use \( \delta_{\mathcal{O}} = \langle\langle p^{2n} \rangle\rangle/T^n \). We list the coefficient for the case \( \mathcal{O} = \langle p^4 \rangle \) in Table I for the FPU \( \beta \)-model as well as \( \phi^4 \) theory for comparison.

Let us investigate how well \( \phi^4 \) describes the spatial distribution of \( \langle\langle p^4 \rangle\rangle/T^2 \). We find

\[ \frac{\langle\langle p^4 \rangle\rangle}{T^2} = a_4 T^{2(\gamma-1)+s_4} = a_4 \left( \frac{T_1}{T} \right)^{1-\gamma} \left( 1 - \frac{T_2}{T_1} \right)^{-1} \left( \frac{x}{L} \right)^{1-\gamma} \]  

\[ \frac{\langle\langle p^4 \rangle\rangle}{T^2} \]

| \( \phi^4 \) Theory | \( \langle\langle p^4 \rangle\rangle/T^2 \) |
|---------------------|----------------------------------|
| \( d = 1 \) | \( T = 1 \) | 3.3(24) | 0.96(15) |
| | \( T = 5 \) | 1.6(6) | 1.18(9) |
| \( d = 2 \) | \( T = 1 \) | 1.9(4) | 1.09(5) |
| | \( T = 5 \) | 0.4(2) | 1.6(2) |
| \( d = 3 \) | \( T = 1 \) | 4 (1) | 0.96(10) |
| | \( T = 5 \) | 0.2(5) | 1.6(6) |

TABLE I: Non-Equilibrium coefficients \( C_4 = \langle\langle p^4 \rangle\rangle L^4 \) for \( \langle\langle p^4 \rangle\rangle/T^2 \) (cf. Eq. (10), (14)). The results are shown for the FPU \( \beta \) model and the \( \phi^4 \) theory in \( d = 1 \sim 3 \)-dimensions. The value of \( s \) is extracted from fitting to several temperatures.

III. SPATIAL DEPENDENCE OF CUMULANTS IN THE NON-EQUILIBRIUM STEADY STATE

In non-equilibrium steady states, physical observables show deviations from their equilibrium values reflecting the lack of local equilibrium in the system. The behavior of the observables have been seen to be well described by \( \phi^4 \) on average, at least in some cases\[11\]. Here, we now would like to investigate a more detailed issue — whether these properties can be used to understand the nature of the spatial profiles of these observables in a given non-equilibrium situation. We will assume that within some range of \( T \) and \( L \) that we can represent the expansion coefficients in \( \phi^4 \) as

\[ C_\mathcal{O} = \mu_\mathcal{O} T^{s_\mathcal{O}} L^{n_\mathcal{O}} \]  

The behavior of \( C_\mathcal{O} \) with respect to \( T, L \) clearly must depend on the dynamics of the theory and is not expected to be generic.

To study the spatial distribution of physical observables in non-equilibrium, we make use of (2) which describes how the observables should behave in non-equilibrium locally in space, given the thermal conductivity. Using this property and (4), we obtain to leading order that observables will deviate from their local equilibrium values as

\[ \delta_\mathcal{O} = C_\mathcal{O} \left( \frac{JT(x)^{\gamma-1}}{c} \right)^2 = a_\mathcal{O} T(x)^{2(\gamma-1)+s} \]  

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Here \( a_\mathcal{O} \) is defined through this equation and should be proportional to \( J^2 \). This implicitly contains the spatial distribution since the temperature profile is known and can be understood as in (7).

While these arguments apply to any physical observable in the system, we choose to study cumulants of momenta, \( p \), mainly for the following reasons; conceptual and practical. There seems to be no universal rigorous definition of local equilibrium, yet the concept in the least seems to include a unique meaning for temperature, which in this case would lead to the Maxwellian distribution for \( p \). To put another way, when the momentum distribution is not Maxwellian, we can choose different definitions of the temperature based on the various moments of \( p \). The cumulants of the momentum distribution provide insight into how the physical properties of a non-equilibrium system deviates from those of local equilibrium. The cumulants are well defined local variables and their values in local equilibrium are known precisely. The low order cumulants are defined as

\[ \langle\langle p^2 \rangle\rangle = \langle p^2 \rangle, \quad \langle\langle p^4 \rangle\rangle = \langle p^4 \rangle - 3\langle p^2 \rangle^2, \quad \langle\langle p^6 \rangle\rangle = \langle p^6 \rangle - 15\langle p^2 \rangle^3 + 30\langle p^2 \rangle^3, \ldots \]  

where, in equilibrium,

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This property is also of practical importance. Since the deviations we compute can be small, it is desirable to use observables whose local equilibrium values are known exactly. In this case in thermal equilibrium, \( \mathcal{O}_{eq} = 0 \), so we use \( \delta_\mathcal{O} = \langle\langle p^{2n} \rangle\rangle/T^n \). We list the coefficient for the case \( \mathcal{O} = \langle p^4 \rangle \) in Table I for the FPU \( \beta \)-model as well as \( \phi^4 \) theory for comparison.

Let us investigate how well \( \phi^4 \) describes the spatial distribution of \( \langle\langle p^4 \rangle\rangle/T^2 \). We find

\[ \frac{\langle\langle p^4 \rangle\rangle}{T^2} = a_4 T^{2(\gamma-1)+s_4} = a_4 \left( \frac{T_1}{T} \right)^{1-\gamma} \left( 1 - \frac{T_2}{T_1} \right)^{-1} \left( \frac{x}{L} \right)^{1-\gamma} \]  

\[ \frac{\langle\langle p^4 \rangle\rangle}{T^2} \]
FIG. 2: (left) Spatial dependence of the rescaled 4-th momentum cumulant, $⟨⟨p^4⟩⟩/T^2$ for the four systems in Fig. 1. Larger cumulant values are seen for larger boundary temperature differences. (right) Temperature dependence of $⟨⟨p^4⟩⟩/T^2$ for the same systems. In both panels, the predictions $11$ are indicated by the dashes.

FIG. 3: $J$ dependence of the non-equilibrium expansion coefficient, $a_4$, for various boundary conditions, $(T_0^1, T_0^2)$ and system sizes $L$. The dashed line is $3.72J^2$ and the $\sim J^2$ behavior of the coefficient can be clearly seen, as predicted from theory. Each data point represents a particular temperature boundary condition for $L = 32$ (×), $L = 64$ (□) and $L = 128$ (○) systems.

$s_4$ is the temperature dependence of the coefficient $C_4$ which is reflected in Table IV. To understand the validity of the prediction Eq. (1), fits were made with just one parameter $a_4$ for the whole profile. We find that this describes the situation quite well, as seen in the examples of Fig. 2 where the predictions are denoted by dashes. In these figures, we have compared the fits with the spatial as well temperature dependence of $⟨⟨p^4⟩⟩$ for the four systems shown in Fig. 1. In this temperature range, temperature dependence of the thermal conductivity is weak so we used $\gamma = 0$ and $s_4 = -0.14$ extracted from the data in Table IV. Similar results were found for different temperature boundary conditions and for different $L$. To further verify the underlying physics, we study the $J$ dependence of the coefficient $a_4$. The behavior for various systems, including the four systems in Fig. 1, are shown in Fig. 3. Each data point represents a system with a particular size and temperature boundary conditions. The central temperature is around $T = 8.8$ and is kept fixed. The observed behavior is clearly well described by $a_4 \sim$ const. $\times J^2$. The coefficient $a_4$ seems $L$-independent and this can roughly be understood since $c^2$ grows in $L$ in a manner similar to $C_4$. We have in addition systematically studied the results to see if we can discern the contribution of higher order terms in the expansions $1, 2$ (of order $J^4$ and higher) but have found no consistent evidence for them. In other physical situations, non-analytic behavior seems to have been seen in some cases $21, 22$.

While the logic seems to work for the lowest non-trivial order cumulant, $⟨⟨p^4⟩⟩$, we find it instructive to analyze if it works at higher orders. In this direction, we have analyzed the next non-trivial order $⟨⟨p^6⟩⟩$ and have found that its behavior is quite consistent with physics of $11$, as was the case of $⟨⟨p^4⟩⟩$, in all the systems we have studied. In practice, higher order cumulants are more prone to errors and the computations are more difficult. The results for
FIG. 4: Spatial dependence (left) and temperature dependence (right) of \( \langle p^6 \rangle / T^3 \) for the four systems in Fig. 1. Larger cumulant values are seen for larger boundary temperature differences. Predictions are shown with dashes.

FIG. 5: \( J \) dependence of the coefficient \( a_6 \) for various boundary conditions, \((T_0^1, T_0^2)\) and system sizes \( L \). The dashed line denotes \( 156 J^2 \). \( \sim J^2 \) dependence of \( a_6 \) is evident, in agreement with the predictions. Each data point represents a particular temperature boundary condition for system sizes \( L = 32 \) (\( \times \)), \( L = 64 \) (\( \square \)) and \( L = 128 \) (\( \bigcirc \)), as in Fig. 3.

the same four systems in Fig. 1 are shown in Fig. 4. As in the \( \langle p^4 \rangle \) case, the coefficient \( a_6 \) shows \( J^2 \) behavior within error, as it should. \( a_6 \) shows a weak \( L \) dependence, as we would generically expect. A common value of \( s_6 = -1.6 \) was adopted for all the data in Fig. 4 and Fig. 5. What is evident is that the spatial behavior of non-equilibrium observables can be explicitly related to transport and other physical properties of the system using rather general considerations. From the cumulants we now consider what can be said about the full momentum distribution function.

IV. CUMULANTS AND THE DISTRIBUTION

The cumulants are quantitative indicators of the non-Maxwellian nature of the momentum distribution or the violations of local equilibrium. All the cumulants are non-zero unless the system is in local equilibrium, in which case only the linear and quadratic cumulants are non-zero. There are very few problems where cumulants can all be computed analytically and it becomes numerically intractable to compute them as we go to higher orders. It is then of interest to see how well the lower order cumulants characterize the distribution. The cumulants are properties of the distribution function, which has an infinite number of degrees of freedom. A priori, there is no reason to assume that the lower order cumulants characterize the distribution. In order to clarify this issue, first note that the distribution...
function $f(p)$ and the cumulants are related explicitly through the generating function as

$$
\int dp e^{ipf(p)} = \langle e^{ip} \rangle = \exp \left( \sum_{n=0}^{\infty} \frac{i^n p^n}{n!} \langle p^n \rangle \right) = \exp \left( \sum_{n=0}^{\infty} \frac{(-p^2)^n}{(2n)!} \langle p^{2n} \rangle \right)
$$

(15)

Here, in the last equality, the symmetry under $p \leftrightarrow -p$ was used, which leads to $\langle p^{2n+1} \rangle = 0$. We see from this equation that given all the cumulants (or equivalently, moments), we may recover the distribution function by performing an inverse Fourier transform. However, in practice, not all the cumulants are available.

Intuitively, we expect the lower order cumulants to be the leading order results with higher order cumulants becoming more important as we move further away from equilibrium. In Fig. 8, we plot the relative difference of the measured distribution $f(p)$ to the thermal distribution, $f_0(p)$, for the distribution directly measured in the simulations and the distribution computed from the low order cumulants, $\langle p^{2.4.6} \rangle$. The comparisons are performed for the four systems in Fig. 1 at a point in the middle of the system. From these graphs, we observe the following: (a) The agreement between the distribution computed from lower order cumulants and the distribution is quite good in all cases; (b) the relative deviation from the thermal distribution is larger as we move away from equilibrium (larger $\Delta T/T$), as expected; (c) the small discrepancy between the computed distribution and the measured one seems to be larger for larger $\Delta T/T$; (d) the deviation from the thermal distribution becomes more noisy for smaller $\Delta T/T$, since the deviation itself is smaller and the relative error is larger. We mention here that strictly speaking, the distributions can have different behavior, such as long tails, beyond the region we have investigated. However, these tails would have to be quite small since the distributions decay as $\exp(-p^2/(2T))$ and the agreement is good up to reasonably large $p$, as seen in Fig. 8. We have examined numerous systems for different $T$ and $L$ and found similar good agreement. Therefore, we see that the lower order cumulants provide good physical observables that quantitatively describe the deviations of the systems from local equilibrium, at least in the FPU model.

It is possible to examine the characteristics of the higher order cumulants. It should be noted that unlike the even moments $\langle p^{2n} \rangle$, even cumulants, $\langle p^{2m+1} \rangle$, need not be positive and in general will not be. So to study the general trend of the cumulants for higher order, we examine the magnitude of the cumulants. In Fig. 7 (left), we show the behavior of the cumulants up to 20-th order for the same four systems in Fig. 1 specifically for the point at which the momentum distributions in Fig. 6 were computed. Only data points with reasonable error are shown and an explanation of the relevant errors is given below. We see an increase in the magnitude with the order is roughly exponential. This growth is far milder than the (2n)! seen in (18).

The behavior of the higher order cumulants is of some import and we briefly explain semi-quantitatively why they are difficult to obtain. The difficulty lies mainly in the statistical error in the simulations. This can be estimated from the number of samples for computing the expectation values as

$$
\Delta \langle p^n \rangle \sim \frac{n}{\sqrt{N}}
$$

(16)

where $\Delta$ denotes the error and $N$ is the total number of samples or the number of time steps in the simulation. Note that $\langle p^{2n} \rangle = \langle p^n \rangle + \ldots$ so that an error estimate for the moment should suffice as the error estimate for the cumulant. An adequate value for the moment can be obtained in equilibrium,

$$
\langle p^n \rangle \sim (n-1)!!
$$

(17)

Combining these relations, we find the statistical error for the cumulants which increases rapidly for higher order cumulants.

$$
\Delta \left( \frac{\langle p^n \rangle}{T^{n/2}} \right) \sim \frac{(n-1)!! n}{\sqrt{N}}
$$

(18)

These estimates for the error also apply to the equilibrium situation. In contrast to the non-equilibrium cumulants, the equilibrium cumulants should vanish, with the exception of $\langle p^2 \rangle$. As the measured values will converge to zero, at any given time-step in the simulation, their values will be generically non-zero. In Fig. 7 (right), we compare the equilibrium cumulants, in the middle of the system to the above error estimates. It can be seen that the rough estimate (18) seems to be consistent with the results. As one samples more ($N$ increases), these will tend to zero. However, for a finite sample size, this is found to explain the order of the uncertainty.

With $N = 10^9$ time-steps — which we used for the values in Fig. 4 — for 8 and 10-th order cumulants, the errors are 0.03 and 0.3. As we can see from Fig. 7 (left), this means that we can obtain up to the 8 or 10-th cumulant with reasonable error for the four systems but the higher order cumulants are expected to be unreliable for systems closer
FIG. 6: The relative deviation of the distribution from the Maxwell distribution for the four systems in Fig. 1. Distribution obtained from the cumulants \( \langle p^4 \rangle, \langle p^6 \rangle \) (dashed) are compared with the measured distributions (solid). The agreement is excellent. \( \Delta T/T \) denotes the boundary temperature difference over the average temperature and is an indication of how far the system is from equilibrium.

to equilibrium. These error estimates are quite consistent with the estimates we obtain from the statistical properties of the simulations. These errors can be overcome with higher statistics which quickly becomes unrealistic for higher order. We have analyzed systems with various other temperature boundary conditions and \( L \) and have found the increasing behavior of the cumulants seen in Fig. 4 (left) to be quite generic.

V. SUMMARY AND DISCUSSIONS

The spatial distribution of cumulants in non-equilibrium steady states under thermal gradients were predicted from general considerations and tested in the the FPU model. The understanding of the temperature profile for a given non-equilibrium steady state, combined with the deviations of physical observables from their equilibrium values, can be used to develop a consistent description of the spatial distribution of observables. In principle, the behavior of observables probably have higher order corrections in the non-equilibrium nature of the system, which in this case is \( \nabla T \), but higher order effects could not be separated within the current numerical simulation results.

We quantitatively analyzed the relation between the momentum cumulants and the distribution in the non-equilibrium steady state. It was found that the lower order cumulants characterize the difference of the non-equilibrium distribution from the one in local equilibrium quite well. Understanding and characterizing the properties of the distribution is of manifest importance since the distribution function for physical variables allows us to compute any observable constructed from these variables. To understand the properties of any local variable in the non-equilibrium state, the physical properties of the coordinate variables also need to to be clarified.

A comment is perhaps in order: lack of local equilibrium behavior can in some cases be attributed to the lack of
So coarse graining will average out the violations of local equilibrium seen above. Also, the non-local equilibrium properties found in this paper pertain to systems in the non-equilibrium steady state and therefore are not transient.

We have also performed similar analyses of spatial distributions on the $\phi^4$ model. The physical properties of the model are different from those of FPU model and we intend to report on this in the near future.

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