Flux boundary conditions in particle simulations

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Flux boundary conditions are interesting in a number of contexts ranging from multiscale simulations to simulations of molecular hydrodynamics in nanoscale systems. Here we introduce, analyze, and test a general scheme to impose boundary conditions that simultaneously control the momentum and energy flux into open particle systems. The scheme is shown to handle far from equilibrium simulations. It acquires its main characteristics from the requirement that it fulfills the second law of thermodynamics and thus minimizes the entropy production, when it is applied to reversible processes. It is shown both theoretically and through simulations that the scheme emulates the effect of an extended particle system as far as particle number fluctuations, temperature, and density profiles are concerned. The numerical scheme is further shown to be accurate and stable in both equilibrium and far from equilibrium contexts.

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I. INTRODUCTION

Both equilibrium and nonequilibrium particle simulations always require boundary conditions in addition to the prescription of the bulk dynamics. These boundary conditions may act to simulate an extended particle bath, as in standard equilibrium ensembles or a condition that maintains the system away from equilibrium as in a shear flow simulation. This is true whether the particle interactions are conservative as in a classical molecular dynamics simulation [1], or dissipative as in dissipative particle dynamics [2] or granular systems.

While well-established techniques exist to introduce a pressure [3] or a temperature [4] in a nonlocal way, the present paper introduces local boundary conditions that allow the fluxes of momentum and energy to be simultaneously specified either as free parameters or in order to simulate an external reservoir. Such boundary conditions may be required in a wide variety of contexts where the hydrodynamic flow needs to be simulated on the molecular level, whether one studies imbibition in carbon nanotubes [5], flows around proteins, or through layered liquid films [6]. Examples of such systems and processes also include temperature controlled, strong shears in nanolayers, or cross-phenoomena such as the thermomolecular effect in nanotubes [7], and the osmotic flow of water through membranes [8,9].

Particle flux boundary conditions are interesting in particular for the design of multiscale simulation schemes where particle and continuum simulations are coupled by flux exchange algorithms [10–14]. For these schemes to be conservative the flux leaving one system must enter the other. In existing schemes the particle fluxes have been imposed by techniques that only produce the desired average fluxes, and are unstable or burdened with certain undesirable features [15,16]. The present scheme introduces the exact fluxes into the particle system. Moreover, it is analyzed and derived by a statistical mechanical argument.

Our model is derived from the requirement that it be possible to impose reversible changes through the flux boundary conditions, i.e., that local equilibrium distributions be preserved and the entropy production minimized. From this it is shown how our algorithm may be used to perform molecular dynamics, or other particle simulations of systems which are governed by the grand canonical ensemble. Our approach differs from previous simulations using this ensemble [17] by the use of open boundaries that connect to a reservoir of particles from which the flow of momentum and energy is controlled. Processes that are most effectively studied in the grand canonical ensemble include the equilibration of thin Black Newton films [18], the swelling of clays in water [19,20] or water between lamellar lipid structures [21].

We note that the scheme may also be applied to some nonstandard equilibrium ensembles. For instance, by setting the energy flux to zero and the momentum flux to a prescribed pressure , an ensemble of fixed energy and pressure, but variable particle number , may be simulated.

In the present paper some simple 2D test simulations are carried out using soft sphere potentials, of both equilibrium and nonequilibrium processes. The scheme is shown to exhibit the predicted equilibrium behavior and remain both stable and accurate in far from equilibrium processes. The equilibrium behavior coincides with the theoretical predictions as far as density, temperature, and chemical potential profiles are concerned. Also, the energies are shown to satisfy the Boltzmann distribution and the particle number fluctuations reproduce the standard predictions of the grand canonical ensemble. The nonequilibrium processes examined include a freely expanding gas, in which case the model is shown to be both stable and numerically accurate, and shear flow simulations where the flow measurements agree with simple hydrodynamic predictions.

The paper is organized as follows: In Sec. II the basics of the model are defined. In Sec. III the details are worked out and statistical mechanical arguments applied to derive the particle insertion and removal procedure. In Sec. IV the model is applied to simulate equilibrium ensembles, in Sec. V a numerical technique to improve accuracy is introduced, and in Sec. VI the simulation results are presented and dis-
II. MODEL

It follows from Newton’s laws that it is impossible to prescribe the force on a body and the resulting velocity independently. For this reason we cannot simultaneously specify mass, momentum, and energy fluxes, and we choose to let the mass flux be the dependent quantity. For example, if our boundary conditions were applied at the ends of a pipe in which particles bounced back upon hitting the walls, the average mass flux would be determined by the pressure drop in which particles bounced back upon hitting the walls, the interface area. The momentum change \( \Delta m v_i \) of the added/removed particle is the momentum flux due to \( j_p \) if the particle is removed. The energy change \( \Delta e_j \) of the added/removing particles uses the sum convention. The sums \( \sum_i F_i \cdot v_i dt \) and \( \sum_i F_i \cdot v_i dt + \sum_i \Delta e_i \) are the momentum and energy inputs due to \( F_i \) during the time \( dt \). In order to simplify Eqs. (1) and (2) we define \( \bar{j}_p \) and \( \bar{j}_e \) through the relations

\[
Adt \bar{j}_p = Adt j_p - \sum_{i'} \Delta (mv_{i'}) = \sum_i F_i \cdot v_i dt,
\]

\[
Adt \bar{j}_e = Adt j_e - \sum_{i'} \Delta e_{i'} = \sum_i F_i \cdot v_i dt.
\]

Provided that the force \( F_i \) satisfies these conditions the correct energy and momentum fluxes into the particle system will result.

III. STATISTICAL MECHANICS OF FLUX BOUNDARY CONDITIONS

In order to specify \( F_i \) and the particle addition/removal procedure we need a criterion in addition to Eqs. (3) and (4). This criterion is that it shall be possible to impose reversible changes to the system through the boundary conditions. This implies that the scheme must evolve the buffer via equilibrium states, or arbitrarily close to such states. In order for the

heat capacity the flux across the bulk boundary will both fluctuate and be somewhat delayed if \( j_p \) and \( j_e \) depend on time. Correspondingly, the momentum and energy contained in the buffer will fluctuate. However, in real applications typically \( N \sim 10^5 \) or larger while \( M \sim 10^2 \) or smaller. The time scale of momentum and energy flow through the buffer will be correspondingly smaller than these time scales in the bulk. Hence the relative effect of the buffer mass and heat capacity will normally be negligible.

The particle dynamics itself conserves mass, momentum, and energy. In the buffer zone there is an additional nonconservative force \( F_i \). The average of this force will act to taper the particle density profile continuously to zero. The particle creation procedure is thus easily implemented as particles may be added in the dilute region away from other particles. While efficient techniques to add particles with the correct potential energy in dense particle systems do exist [22], the present model circumvents any need for them.

The flux boundary conditions introduce energy and momentum to the particle system both through the force \( F_i \) and particle addition/removal. We require that the combined effect is to create the prescribed momentum and energy fluxes, i.e.,

\[
j_p Adt = \sum_i F_i dt + \sum_i \Delta (mv_{i'}), \tag{1}
\]

\[
j_e Adt = \sum_i F_i \cdot v_i dt + \sum_i \Delta e_{i'}, \tag{2}
\]

FIG. 1. Illustration of the model. The buffer is to the left of \( x = -x_B \) and the nonconservative buffer force \( F_i \) acts only on the \( M = 16 \) particles there. At \( x = x_B \) the particle velocities are reversed. In the vertical direction periodic boundary conditions are used.

cussed. Our conclusions are contained in Sec. VII.
the same factor, i.e.,

$$ \sum_{i=1}^{M} F_i \cdot v_i = MF \cdot \langle v \rangle + \sum_{i=1}^{M} F'_i \cdot v'_i, $$

which gives \( f \) and the particle force

$$ F'_i = \frac{A v'_i}{\sum_{i=1}^{M} v'_i^2} [\bar{\mu}_i - \bar{F}_p \cdot \langle v \rangle]. $$

This concludes the derivation of \( F_i \). External particle forces that are proportional to the particle velocity have been widely applied and studied as heating mechanisms in molecular dynamics simulations [24,25]. However, to our knowledge such forcing schemes have not previously been generalized to control the momentum flux.

The only thing that remains in order to define the full algorithm is to specify the particle addition/removal procedure. We start from the general observation that a factorized equilibrium distribution is easily preserved under the addition of a particle. The \( N \)-particle equilibrium distribution \( \Pi_i^{N+1} P_i^{eq} \) simply becomes the \( N+1 \)-particle distribution \( \Pi_i^{N+1} P_i^{eq} \) if the new particle is picked from the single particle distribution \( P_i^{eq} \).

However, \( P_i^{eq} \) is in general not factorizable. It may be written

$$ P_i^{eq} = \Lambda(\{x_i\}) \prod_i P_{MB}(v_i) $$

where the Maxwell-Boltzmann distribution

$$ P_{MB}(v_i) \propto \exp\left(-\frac{(m/2)v_i^2}{T}\right) $$

when \( \beta = 1/T \). As long as particle interactions cannot be neglected \( \Lambda(\{x_i\}) \) will contain information on spatial correlations and will not factorize. Only in the low density limit will it do so. So, for the dilute end of the buffer we may write

$$ \Lambda(\{x_i\}) \approx \prod_i \rho^{eq}(x_i), $$

where \( \rho^{eq}(x_i) \) is the dilute limit of the equilibrium density profile and \( i \) now only runs over particles in the corresponding region.

By adding particles with positions and velocities picked from the single particle distribution

$$ p_i^{eq}(x,v) \approx \rho^{eq}(x)P_{MB}(v), $$

local equilibrium is preserved in the dilute part of the buffer. For global equilibrium to be preserved as well, we must again allow the system to remain sufficiently close to overall equilibrium states, so that the entropy production of the hydrodynamic modes following from the external fluxes remains negligible. In particular the heat conduction, which is likely to be the slower of the hydrodynamic modes, must be allowed to relax.

Note that if we remove particles by randomly selecting them in the region where particles are added, the distribution of the removed particles is automatically the same as that of the added particles, i.e., Eq. (15).
The buffer density profile is easily predicted from the assumption of equilibrium. Since the buffer particle number \( M \) is fixed we may under equilibrium conditions write the average force on the buffer particles (\( \mathbf{F} \)) as a potential force

\[
\langle \mathbf{F} \rangle = \frac{pA}{M} \mathbf{n} = -\nabla \Phi,
\]

where \( \Phi = -pAx/M \) and \( p \) is a constant pressure. The fluctuations in \( \mathbf{F} \) around the average \( \langle \mathbf{F} \rangle \) are caused by the particle insertion/removal and temperature fluctuations. In equilibrium the contributions from both these effects vanish. The potential \( \Phi \) should then be considered as part of \( U \) and will thus act to define \( P^\theta \). In equilibrium \( \mu \) is constant throughout the system. We may thus write \( \mu = \mu_0 (p(x), T) + \Phi(x) \) \( = \text{const} \) where \( \mu_0 \) is the chemical potential at \( p(x) \) in the absence of any external potential. In the dilute part of the buffer we may use the ideal gas expression \( \mu_0 = T_0 \log(p(x)/\rho_0) \), where \( \rho_0 \) is the quantum density. This gives the barometric density profile

\[
\rho^\theta(x) \approx e^{pAx/M T_0}.
\]

The above result could also have been predicted from force balance arguments.

In practice, it is an easy matter to observe where the profile acquires an exponential decay. In the simulations particle insertion/removal is carried out at \( x < -3x_B \), which is well within this domain, and the spatial probability distribution for particle insertion is normalized to unity over this region.

Since both particle addition and removal are reversible in the sense that they minimize the entropy production we may write the entropy change associated with the addition of \( \Delta N \) particles as

\[
T \Delta S' = \sum_{i'} \Delta \epsilon_{i'} - \mu \Delta N.
\]

This is the standard form of the second law when the volume available to the system is not changed. But the total entropy change to the system during the time \( \Delta t \), during which the \( \Delta N \) particles are added, also includes the heating due to \( \mathbf{F}' \). For isolated systems (\( \Delta N = 0 \)) it has been shown both analytically and through simulations [24] that \( T \Delta S = \sum_{i=1}^{M} \mathbf{F}'_i \cdot \mathbf{v}'_i \Delta t \). Hence since particle addition and the application of \( \mathbf{F}' \) happen sequentially we may assume the following additive form for the entropy change:

\[
T \Delta S = T \Delta S' + \sum_{i=1}^{M} \mathbf{F}'_i \cdot \mathbf{v}'_i \Delta t.
\]

Combining this result with the expression for the corresponding energy change

\[
\Delta E = \sum_{i=1}^{M} \mathbf{F}_i \cdot \mathbf{v}_i + \sum_{i'} \Delta \epsilon_{i'} = M \mathbf{F} \cdot \langle \mathbf{v} \rangle \Delta t + \sum_{i=1}^{M} \mathbf{F}'_i \cdot \mathbf{v}'_i \Delta t
\]

\[
+ \sum_{i'} \Delta \epsilon_{i'}
\]

we obtain

\[
T \Delta S = \Delta E - \mu \Delta N - M \mathbf{F} \cdot \langle \mathbf{v} \rangle \Delta t.
\]

We define the volume change \( \Delta V \) of the system as \( A \) times the buffer center of mass displacement in the \( x \) direction, i.e.,

\[
\Delta V = -A \langle \mathbf{v} \rangle \cdot \mathbf{n} \Delta t.
\]

By using Eq. (16) to eliminate \( A \mathbf{n} \) we get

\[
p \Delta V = -M \mathbf{F} \cdot \langle \mathbf{v} \rangle \Delta t,
\]

and we obtain the standard form of the second law for reversible processes

\[
T \Delta S = \Delta E + p \Delta V - \mu \Delta N.
\]

Finally, we note that the existence of the barometric density profile Eq. (17) requires a positive pressure. Indeed, the existence of a dilute region relies on this. With a positive external pressure \( p \) the existence of a finite equilibrium vapor pressure will guarantee some ideal gas region where \( x < -x_B \) and the local pressure drops to zero [27].

However, with a negative pressure, as may, for instance, occur in water in narrow conduits, the present algorithm will not work and must be replaced by schemes that add particles in dense regions [22,26]. Theoretically, it is not known how the imposition of the exact energy flux or preservation of local equilibrium may be achieved in these cases, although as shown in Ref. [22], the system relaxes to the expected equilibrium state within a few collision times if the correct amount of potential energy is released upon each particle insertion.

**IV. APPLICATIONS AND EQUILIBRIUM ENSEMBLES**

The model may now be applied in various ways. As mentioned already it solves the problem of flux imposition in hybrid schemes where \( j_x \) and \( j_p \) are provided as time dependent boundary conditions by some continuum solver.

As a means to simulate equilibrium ensembles the boundary condition most naturally realizes the grand canonical ensemble. By taking \( j_x = -\lambda (T - T_0) \) where \( \lambda \) is a relaxation parameter that plays the role of a heat conductivity, and the temperature is defined by the relation \( T = (a m^2/2) \), we simulate a heat bath of temperature \( T_0 \). By setting \( j_p = n \mathbf{p} \) we fix the pressure \( p \). The freely adjustable particle number \( N \) will then equilibrate around an average that is governed by the chemical potential \( \mu(p, T_0) \). The bulk part of the system which has fixed volume also has fixed external \( T_0 \) and \( \mu \), and is thus described by the grand canonical ensemble.

Other choices are of course possible too. By setting \( j_x = 0 \), \( p \) fixed and specifying the Maxwell-Boltzmann distribution for the added particles by the internal temperature \( T \), one defines an ensemble where energy and pressure are fixed but where \( N \) and \( T \) are fluctuating. Likewise, one could fix the enthalpy \( H \) of the total system by setting \( j_x = M \langle \mathbf{v} \rangle \cdot \mathbf{F} \), or equivalently \( \Delta E + p \Delta V = \Delta H = 0 \). However, the system is still open, so this choice does not correspond to an insulated system. Hence these two latter ensembles may be of theoretical interest only as it is unclear how one would realize them experimentally.
V. ACCURACY ENHANCEMENT AND STABILITY OF THE FLUX BOUNDARY SCHEME

Particle addition and removal cannot be done in a simple way without causing discontinuous jumps in $F_i$. While this is not a problem in principle it reduces the numerical accuracy of the scheme. In order to smooth these jumps and improve the numerical accuracy we introduce the reservoir variables, $\Delta P_r$ and $\Delta E_r$ for momentum and energy. The idea is that when a particle is removed or added the corresponding momentum and energy correction is returned to the system gradually, rather than in a single punch.

For this purpose the following fluxes are defined to enter in Eqs. (8) and (11)

$$\tilde{j}_p = j_p - \frac{\Delta P_r}{\tau_r A} \quad \text{and} \quad \tilde{j}_e = j_e - \frac{\Delta E_r}{\tau_r A},$$

(25)

where $\tau_r$ is a time scale larger than $dt$ but smaller than any relevant hydrodynamic time scale. The momentum and energy added to the particles from the reservoir variables over a time step $dt$ are $-\Delta P_r dt / \tau_r$ and $-\Delta E_r dt / \tau_r$, respectively. Correspondingly, these variables are updated according to the rules

$$\Delta P_r \rightarrow \Delta P_r - \Delta P_r dt / \tau_r + \sum_{i'} \Delta(mv_{i'}),$$

$$\Delta E_r \rightarrow \Delta E_r - \Delta E_r dt / \tau_r + \sum_{i'} \Delta e_{i'}. \quad (26)$$

Initially, $\Delta P_r$ and $\Delta E_r$ are both 0. While the scheme of Eqs. (3) and (4) leads to fixed momentum and energy fluxes into the system, the present scheme compensates the delay of the momentum and energy input associated with particle addition/removal by a characteristic time $\tau_r$. If no more particles are added or removed $\Delta P_r$ and $\Delta E_r$ will decrease exponentially on the time scale $\tau_r$. Hence the variables $\Delta P_r$ and $\Delta E_r$ act to smooth $F_i$ as a function of time while the total momentum and energy input instead become (slightly) discontinuous with particle addition/removal. If we set $\tau_r = dt$ we recover the original scheme and the actual fluxes will again exactly equal the prescribed ones.

What are the stability criteria of the above algorithm? Two types of instabilities have been observed, one linked to overheating at the dilute end of the system, and one associated with the effect of cooling the system below the available thermal energy.

Since the mean free time $\tau$ that controls the relaxation rate to local equilibrium increases with decreasing density, it will become large at the far end of the buffer region where the density is low. Since $F'_i \propto v'_i$ hot particles here may become increasingly hot and escape due to lack of equilibration. If this happens the distribution of Eq. (15) breaks down and the velocity distribution becomes position dependent. This problem may be dealt with by increasing $M$. When the magnitude of $j_e$ may be chosen at will, say by changing $\lambda$ in our case, one may also decrease $j_e$ thus lowering the heating per particle.

However, this is not a major problem, and even far from equilibrium simulations are stable and converge to equilibrium, as is described below. The parameter $\lambda$ sets the time needed to relax the buffer to any desired value of $T_0$. The simulations described below were all stable and the choice of $\lambda$ corresponds to a relaxation time smaller than a few $\tau$'s, and thus corresponds to relatively strong heating/cooling.

Finally, if $j_e$ and $j_p$ are set to maintain a state for which $j_e < \langle v \rangle \cdot j_p$, i.e., where the overall energy input is less than the work done by $j_p$, the system will be unstable. The physical reason is that more energy is extracted than added, and the thermal velocities will decay to zero. Physical considerations would normally exclude such a boundary condition.

VI. SIMULATIONS

In these simulations Newton's second law is integrated using the velocity-Verlet algorithm and an interparticle potential $U(r < d) = (k/2)(d - r)^2$ where $d$ is the particle diameter, $r$ the center separation, and $k$ a spring constant.

The simulation domain includes a bulk particle region at $-x_B < x < x_B$ and $-y_B < y < y_B$. At $y = \pm y_B$ periodic boundary conditions are applied and at $x = x_B$ the particles bounce back with reversed velocities, thus conserving energy.

In the simulations the length unit is the particle diameter $d$ and the unit of time the inverse frequency $\sqrt{m/k}$ associated with the potential. This implies the unit $kd^3$ for energy.

The velocity Verlet scheme is as follows:

$$r_i(t + \Delta t) = r_i(t) + v_i(t) \Delta t + \frac{F_i(t)}{2m} \Delta t^2 + o(\Delta t^3),$$

$$v_i\left(t + \frac{\Delta t}{2}\right) = v_i(t) + \frac{F_i(t)}{2m} \Delta t,$$

$$v_i(t + \Delta t) = v_i\left(t + \frac{\Delta t}{2}\right) + \frac{F_i(t + \Delta t)}{2m} \Delta t + o(\Delta t^2). \quad (27)$$

This is the standard scheme [4]. Note, however, that since $F_i$ depends not only on position but also on velocities we have to use

$$F_i(t + \Delta t) = F_i\{r_i(t + \Delta t), v_i(t + \Delta t)\}.$$

(28)

in order to get third order accuracy in $v_i(t + \Delta t)$.

In summary the model is implemented by the repeated performance of the following three steps.

1. Propagate the particle positions and velocities using Eq. (28) with the current values of $F$ and $F'$ given by Eqs. (8) and (11).

2. Update the momentum and energy reservoirs, Eqs. (26), and the fluxes $\tilde{j}_p$ and $\tilde{j}_e$, Eq. (25).

3. Update the forces Eqs. (8) and (11) and return to 1.

First, the stability and accuracy of the scheme is tested. Every simulation was started with particles in the buffer zone only, and then the system was left to relax. Initially, a fixed $j_e = 0.5$ was applied and then replaced by $j_e = -\lambda(T - T_0)$ with $\lambda = 0.5$. Figure 2 shows the relaxation of such a free expand-
sion simulation where initially there were only (randomly placed) particles in the buffer. The stability of such a freely expanding system is of interest for the study of shock fronts. The error in energy $\delta E$ is defined as the difference between the measured energy, which includes $\Delta E_{\text{e}}$, and the prediction $E_0 + \int dt A_j$, where $E_0$ is the initial energy.

Figures 3 and 4 show the resulting error after $t=400$ as functions of $dt$ and $M$, respectively. When $dt \geq 0.01$ the error scales roughly as $dt^3$. When $dt < 0.01$ the round-off errors appear to dominate. The variations with $M$ are small down to the $M=64$ value below which the heat capacity of the buffer starts to become too small to accommodate the variations in the flux $j_e$.

In Fig. 5 we have plotted the density and temperature profiles throughout the buffer region. The solid line represents a best fit to an exponential in the buffer region. Due to the presence of interactions between particles, this fitted curve decays somewhat more slowly than the low density profile of Eq. (17). However, in the dilute tail of the curve the measured decay rate of $\rho(x)$ agrees well with the prediction of Eq. (17). In the $\rho < 0.1$ region the discrepancies between the predicted decay coefficient in the exponential, $\rho \lambda/(MT_0)$, and the measured one were smaller than 5%.

The result of Fig. 5 is less obvious than it may appear. The fact that the temperature is roughly constant throughout the buffer and bulk particle region, and that the density falls off exponentially in the buffer region is drastically changed if the particle addition/removal scheme is changed. For in-

FIG. 2. Snapshots of a free expansion simulation at $t=0$, 10, 30, 50, and 70 where initially only the buffer zone contained particles. The error in energy is given as a fraction of the total energy. Here $x_B=-12$, $M=200$, $dt=0.02$, $p=1.0$, $\lambda=1.0$, $T_0=0.5$, and $\tau=1$.

FIG. 3. The relative error in energy after $t=400$ as a function of $dt$ when $\lambda=0.5$, $\tau=1$, and $M=200$. The dashed line has slope 3. FIG. 4. The relative error in energy after $t=400$ as a function of $M$ when $\lambda=0.5$, $\tau=1$.

FIG. 5. The particle density $\rho(x)$ ($\bigcirc$) and temperature $T(x)$ ($\triangle$) as a function of $-x$. The full line shows a best fit to an exponential in the buffer region. The profile was averaged over time $500 < t < 4000$. Except for the value $x_B=-6$ parameters are as in Fig. 2.
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FIG. 6. The time-averaged pressure as a function of time at the rightmost wall, \( x=x_B \) divided by the imposed \( P=j_p \cdot n \), corresponding to the simulation in Fig. 2. The dashed line shows the instantaneous pressure, averaged only over the time \( dt \) while the full line shows the corresponding running average over a time window \( \Delta t = 50 \). The straight line is the expected value.

stance, by inserting and removing particles at different average energies, or at different average locations, the profiles are significantly changed.

The results of Figs. 4 and 5, which show the numerical accuracy and the proximity of equilibrium in the simulations, dictate the choice of \( M \). These results reflect what is happening in the buffer, so this choice does not depend on the number of bulk particles. In practice one needs \( M > 60 \).

Figure 6 shows the results of measuring the pressure as a function of time. The pressure is measured as the momentum change per unit area per unit time at the right hand wall where particles are subject to the bounce back condition. When averaged the pressure is indeed seen to conform to the imposed value. The initial overshoot is associated with the free-expansion phase before the system reaches equilibrium. Figure 7 shows the local distributions of kinetic energy. The curves in Fig. 7 are averaged over sections of the buffer region. The lowest curve corresponds to an average over particles located at \(-50 < x+x_B < -40 \), the next curve to particles in \(-40 < x+x_B < -30 \) and so forth to the upper curve which is obtained from particles in \(-10 < x+x_B < 0 \). The probabilities are not normalized by the local density, hence the shift between the curves directly corresponds to the decrease in \( \rho(x) \), which is almost two orders of magnitude less

FIG. 7. The probability \( P_x(E_K) \) of finding the kinetic energy \( E_K \) as a function of \( E_K \) for different values of \( x \) (\( \circ \)). The straight lines are linear fits. Here \( x_B = -6 \) and the other parameters are as in Fig. 2.

FIG. 8. The temperature as a function of density. \( T_0=0.5 \) and \( T/T_0 \) are the slopes in Fig. 7 in the buffer region.

for the lower than for the upper curve. It is seen from Fig. 7 that

\[
T_0 \log(P_x(E_K)) = \text{const} - E_K
\]

to a good approximation, i.e., that the Boltzmann distribution \( P_x(E_K) \propto \exp(-E_K/T_0) \) is satisfied locally. To quantify this statement the slopes of the curves in Fig. 7 are plotted in Fig. 8. The expected slopes are \(-1 \), while the actual slopes measure \(-T_0/T \) where \( T \) is the local temperature. The local value of \( T \) does not fluctuate significantly beyond expected equilibrium fluctuations. It is reassuring that the system is well equilibrated even within the low density tail as the consistency of the model relies on this property.

Figure 9 shows the particle number as a function of time. In the very first stages \( t \leq 50 \), there is a rapid increase in \( N(t) \) as the system fills up with particles. After the free expansion phase a density wave and hydrodynamic oscillations in \( N(t) \) are seen. After \( t=500 \) the wave that is caused by the filling process decays and we are left with the equilibrium fluctuations.

To investigate whether the flux boundary conditions really play the role of an extended particle system, we measure the fluctuations in \( N \) at constant \( T_0 \). According to the standard of statistical mechanics [23]

\[
\langle \Delta N^2 \rangle = (N-M)T \left. \frac{\partial \rho}{\partial \rho} \right|_T,
\]

where \( N-M \) is the bulk particle number. In the ideal gas limit this reduces to \( \langle \Delta N^2 \rangle = N-M \), which is the result we

FIG. 9. The particle number \( N \) as a function of time. Except for the value \( x_B = -6 \) parameters are as in Fig. 2.

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measurements and the measurements of $T/H_20849$ are based on equation of state measurements, and the full line shows a polynomial fit to these.

we expect for low pressures. For higher pressures the repulsive interparticle potential is felt and $\langle \Delta N^2 \rangle \ll N-M$.

In an independent set of measurements using periodic boundary conditions in all directions the constant temperature equation of state $p=p(\rho)$ was measured. From these measurements $\partial p/\partial \rho|_T$ was obtained, both from finite differencing the $p(\rho)$ data and from the derivative of a polynomial fit to $p(\rho)$. In Fig. 10 $\langle \Delta N^2 \rangle/(N-M)$ is compared to $T\partial p/\partial \rho|_T$. It is seen that the fluctuations conform to the prediction within the noise which is intrinsic both to the equation of state measurements and the measurements of $\langle \Delta N^2 \rangle$. In other words, as far as particle number fluctuations go the effect of our flux boundary conditions is the same as that of an extended particle system.

Finally, we briefly examine the case where there is a shear imposed on the system. Figure 11 shows a shear flow where the average velocity reaches the thermal velocity over a few mean free paths. The velocity profile is quite linear within the bulk particle region as one would expect from hydrodynamics. Also the buffer density profile is still close to exponential, and the temperature profile roughly constant as in equilibrium. Interestingly the particle number fluctuations $\langle \Delta N^2 \rangle$ (not shown) have decayed to half the equilibrium value in these nonequilibrium simulations.

VII. CONCLUSIONS

We have proposed a scheme that has been used to impose arbitrary energy and momentum fluxes across the boundaries of an open particle system. This scheme relies on the combination of a buffer force and particle addition/removal, and it may be used to impose any thermodynamic state on the particle system, both at and away from equilibrium. This algorithm and the fact that it closely approximates a particle reservoir when applied to reversible processes is the main outcome of the paper. The algorithm may be used in conjunction with continuum solvers to carry out hybrid computations of far from equilibrium processes where no adequate continuum description exists, for instance, transitions to boundary shear localization, the moving contact line, or the molecular hydrodynamics around nanostructures.

The present simulation scheme has been shown to be stable and accurate to less than 1% in errors of energy in all 2D simulations studied here, both near and far from equilibrium. The scheme can easily be generalized to 3D by the simple replacement $T \rightarrow (3/2)T$ for the average kinetic energy.

Work is in progress to implement the present scheme in hybrid simulations that hitherto have lacked an accurate scheme for flux imposition which maintains the particle domain in a close to equilibrium state.

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[27] We note that with strong interparticle attraction the number of molecules in the dilute vapor region may also become impractically small in parts of the $pT$ phase space of the Lennard-Jones liquid.