Brownian dynamics simulations of hard rods in external fields and with contact interactions

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We propose a simulation method for Brownian dynamics of hard rods in one dimension for arbitrary continuous external force fields. It is an event-driven procedure based on the fragmentation and mergers of clusters formed by particles in contact. It allows one to treat particle interactions in addition to the hard-sphere exclusion as long as the corresponding interaction forces are continuous functions of the particle coordinates. We furthermore develop a treatment of sticky hard spheres as described by Baxter’s contact interaction potential.

I. INTRODUCTION

Motion of micron-sized particles in soft matter and biological systems can in general be well described by an overdamped Brownian motion. In these dynamics, the excluded volume part of the particle-particle interactions is often taken into account by considering the particles as hard spheres. The hard-sphere interaction represents a core part of any interaction if particles cannot penetrate each other and has a decisive impact on collective phenomena. Even if a softcore potential is more accurately accounting for the repulsive interactions when particles approach each other, the hard-sphere model is often used as an effective description. However, in Brownian dynamics simulations the hard-sphere interaction requires special care due to its singular nature.

The problem of simulating hard spheres has been around as long as modern computational physics [1, 2]. In these early studies, equilibrium properties were simulated based on Monte Carlo methods. The first dynamic algorithm was similar to standard Monte Carlo simulations and based on rejecting movements when particle positions violate the hardcore constraint [3]. A modification to this approach was made by placing particles in contact instead of rejecting the movements [4]. However, these first proposed methods are not suitable for high particle densities. Recent developments have shown that Monte-Carlo schemes can give quantitative agreement with Brownian dynamics simulations even at high densities [5–7].

An alternative means of simulating hardcore particle interaction are event-driven algorithms, where particle overlappings are avoided, similarly as it is often done in molecular dynamics simulations [8]. Most of these approaches are based on the idea that when an attempted displacement of the particles in the numerical procedure leads to an overlap of two particles, the corresponding encounter of the two particles is treated like a binary elastic collision. For treating a case where more than two particles overlap, a sequence of binary collision can be carried out. For zero external forces, a corresponding procedure was developed in Refs. [9, 10], and for non-zero external forces a refined method was proposed in Ref. [11]. One may also consider binary collisions which are not perfectly elastic [12–14].

In real systems, it is often found that more than two particles come into contact [15–17]. Then, particles can be grouped into clusters. The presence of clusters can be justified by kinetic theory, thereby establishing the limits of applicability of the binary collision theory (inelastic or elastic) [18]. Aggregation of such clusters in Brownian dynamics simulation was described in [19].

Here we present a method for simulating overdamped Brownian dynamics of hard rods in one dimension based on particle clusters. It relies on an exact method to solve the Langevin equation for this many-body system in arbitrary continuous external force fields. Particle configurations are evolved such that particle clusters are moved coherently. The clusters are formed by particles that are in contact. They can fragment into smaller or merge into larger clusters, and the fragmentation and merging processes are determined by conditions on the mean external forces acting on the clusters.

In several applications, in addition to excluded volume interactions, attractive interaction forces need to be taken into account. To this end, a simple generalization of the hard-sphere model was introduced by Baxter [20]. In this model of sticky hard spheres, adhesive interactions are present when particles are in contact. Equilibrium properties of this model can be studied analytically in one dimension [21]. Beyond one dimension, structural properties were studied extensively by Monte Carlo simulations in the past [22–31]. Approaches based on rates for contact breaking and formation between spheres were developed to describe nonequilibrium kinetics of such systems [32].

For including the contact interaction into Brownian dynamics simulations, we present a way of representing...
the $\delta$-singularity in the sticky hard-sphere interaction potential. This allows one to tackle the contact interaction in our cluster-based algorithm. One can use this method also in other simulation procedures for the Brownian dynamics of hard spheres.

II. LANGEVIN EQUATIONS FOR BROWNIAN MOTION OF HARD RODS

The one-dimensional overdamped Brownian motion of $N$ hard rods of length $\sigma$ with center positions at $x_i$, $i = 1, \ldots, N$, in an external force field $f(x)$ is described by the Langevin equations

$$\frac{dx_i}{dt} = \mu f(x_i) + \sqrt{2D} \xi_i(t),$$

where $\mu$ is the particle mobility, $D$ is the diffusion coefficient, and $\xi_i(t)$ are Gaussian white noise processes with zero mean and correlation functions $\langle \xi_i(t) \xi_j(t') \rangle = \delta_{ij} \delta(t - t')$.

If the particles would exert continuously varying interaction forces on each other, one could add these forces to the right hand side of Eqs. (1) and simulate the particles’ motion by standard methods [33, 34], e.g. by the Euler algorithm in the simplest approach. For hard rods, however, the interaction cannot be described by a continuously varying force. It enters the dynamics as a condition imposed on probability currents in the many-particle Fokker-Planck equation (see supplementary material of Ref. 35). The accurate and efficient treatment of this interaction in the Langevin equations (1) needs special care.

In the following we present a method to solve Eqs. (1), which we refer to as Brownian cluster dynamics (BCD) simulations. We will also consider Baxter’s sticky hard-sphere interaction [20], which corresponds to an additional attraction between two hard rods when they get in contact.

III. BCD SIMULATIONS FOR HARD RODS IN EXTERNAL FORCE FIELD

Our new algorithm gives an approximate solution of the Langevin equations (1) by evolving the system in fixed time steps $\Delta t$. It is based on conditions for cluster movements. Such movements are highly relevant, for example, to understand unexpected particle currents in driven Brownian motion through highly populated periodic potentials [36]. A cluster of size $n$ is a local arrangement of $n$ rods that are mutually in contact, but not in contact with other rods. We call such an arrangement an $n$-cluster. Single rods in this description are 1-clusters.

In dense systems, there will be clusters of all sizes that can fragment into smaller ones and/or collide with neighboring clusters during their motion. The challenge is to identify these fragmentation and collision events accurately.

We do this by first identifying at a starting time $t$ how the clusters move. This “cluster analysis” takes into account the fragmentation events. It amounts to assigning a velocity to each particle at time $t$, where particles moving together as a cluster have the same velocity. An attempt is then made to propagate the particle positions with these velocities in a small time step $\Delta t$. If there is no collision of clusters in the interval $[t, t + \Delta t]$, the particles are displaced correspondingly, and the cluster analysis is carried out for the new particle configuration at time $t + \Delta t$.

In the case when there is a collision of clusters during the interval $[t, t + \Delta t]$, the first collision event at time $t_c$ is determined. For the clusters involved in that collision, the cluster analysis is carried out, yielding updated velocities for the particles being part of the colliding clusters. Then again it is attempted to propagate the particle positions with the fixed velocities for all particles until time $t + \Delta t$. If there is a further collision, this is taken into account in the same manner as the first collision.

The process is repeated until the time $t + \Delta t$ is reached. Then all particle velocities are updated by a cluster analysis.

A. Cluster analysis

We consider $n$ hard rods of length $\sigma$ forming an $n$-cluster at time $t$. The particle positions in the $n$-cluster are ordered from left to right,

$$x_2 = x_1 + \sigma, \ldots, x_n = x_1 + (n - 1)\sigma,$$

where $x_1$ is the position of the leftmost particle. In the cluster, the particles can exert forces on each other that must obey the following conditions:

- The interaction force $F^\text{int}_{i,i+1}$ of particle $i$ on particle $(i+1)$ must be non-negative, $F^\text{int}_{i,i+1} \geq 0$.

- The force $F^\text{int}_{i,i+1}$ exerted by particle $(i+1)$ on particle $i$ is $F^\text{int}_{i+1,i} = -F^\text{int}_{i,i+1}$.

We furthermore define by

$$F^\text{ext}_i = f(x_i) + \frac{\sqrt{2D}}{\mu} \xi_i$$

the total external force, including the stochastic force mediated by the surrounding fluid.

The particles in the $n$-cluster can move in different ways. For example, in case of a 3-cluster, all particles can move as single particles (1-clusters). Or only the left particle moves as a single particle (1-cluster), while the middle and right particle move together as a 2-cluster. Or the left and middle particle move as a 2-cluster, and the right one as a 1-cluster. Or all 3 particle keep in touch and the 3-cluster moves as a whole. We can distinguish
between these possibilities by considering the 4 compositions \{1, 1, 1\}, \{1, 2\}, \{2, 1\}, and \{3\} of the 3-cluster into the respective subclusters.

For an \(n\)-cluster, there exist \(2^n - 1\) possible compositions (fragmentations) \(\{m_1, \ldots, m_s\}\) into subclusters of sizes \(m_1, \ldots, m_s\), with \(1 \leq m_j \leq n\), \(\sum_{j=1}^{s} m_j = n\). A possible fragmentation of a 10-cluster into 4 subclusters is illustrated in Fig. 1 together with the conditions for this fragmentation to occur. The conditions for a general composition (fragmentation) \(\{m_1, \ldots, m_s\}\) to occur, are explained next.

Let us first consider the condition for the \(j\)th subcluster of size \(m_j\) to become separated from the \((j+1)\)th subcluster of size \(m_{j+1}\). All particles in a moving subcluster must have the same velocity, and this velocity must be the same as that of the center (of mass) of the subcluster. The velocity of the \(j\)th subcluster is equal to \(\mu \bar{F}^\text{ext}_j\), where

\[
\bar{F}^\text{ext}_j = \frac{1}{m_j} \sum_{k=1}^{m_j} F^\text{ext}_{j,k}
\]

is the mean external force exerted on the particles in this subcluster. Here, \(F^\text{ext}_{j,k}\) is the external force on particle \(k\) in the subcluster \(j\), i.e. it is equal to the force \(F^\text{ext}_l\) on particle \(l\) in the \(n\)-cluster, where \(l = k + \sum_{i=1}^{j-1} m_i\).

For the \(j\)th and \((j+1)\)th subclusters to separate, the velocity of the \((j+1)\)th subcluster must be larger than that of the \(j\)th subcluster, i.e. the condition \(\bar{F}^\text{ext}_j < \bar{F}^\text{ext}_{j+1}\) must be fulfilled. The same holds true for the other subcluster separations, i.e. we obtain

\[
\bar{F}^\text{ext}_j < \bar{F}^\text{ext}_{j+1}, \quad j = 1, \ldots, s-1.
\]

as a first set of conditions for the composition \(\{m_1, \ldots, m_s\}\) to occur.

In addition, the particles in the \(j\)th subclusters need to keep in contact. This means that the condition (5) for subcluster separation must be violated for all possible fragmentations of this subcluster. We thus obtain

\[
\frac{1}{i} \sum_{k=1}^{i} F^\text{ext}_{j,k} \geq \frac{1}{m_j - i} \sum_{k=i+1}^{m_j} F^\text{ext}_{j,k}, \quad i = 1, \ldots, m_j - 1.
\]

The same holds true for all other subclusters, i.e. inequalities (6) must be obeyed for all \(j = 1, \ldots, s\). This gives the second set of conditions for the composition \(\{m_1, \ldots, m_s\}\) to occur.

For completeness, let us derive these conditions on the forces in a more formal manner by considering the equations of motions of particles in a subcluster of size \(m\). For simplicity, we here label the particle coordinates in the subcluster as \(x_1, \ldots, x_m\). The equations are:

\[
\begin{align*}
\dot{x}_1 &= \mu (F^\text{ext}_1 - F^\text{int}_{1,2}), \\
\dot{x}_i &= \mu (F^\text{ext}_i + F^\text{int}_{i-1,i} - F^\text{int}_{i,i+1}), \quad i = 2, \ldots, m-1, \\
\dot{x}_m &= \mu (F^\text{ext}_m + F^\text{int}_{m-1,m}).
\end{align*}
\]

All particles in the subcluster have the same velocity,

\[
\dot{x}_1 = \ldots = \dot{x}_m = \mu \bar{F}^\text{ext},
\]

where \(F^\text{ext} = \sum_{j=1}^{m} F^\text{ext}_j / m\) is the mean external force. Unless the external forces \(F^\text{ext}\) acting on the particles in the subcluster are all equal, the interaction forces \(F^\text{int}_{i,i-1}\) between the particles must enforce these conditions.

In fact, Eqs. (8) constitute \(m\) linear equations for determining the \((m-1)\) interaction forces \(F^\text{int}_{1,2}, F^\text{int}_{2,3}, \ldots, F^\text{int}_{m-1,m}\). Only \((m-1)\) of these equations are independent, since \(\sum_{i=1}^{m} \dot{x}_i\) does not depend on the interaction forces. The solution of the system of linear equations gives

\[
F^\text{int}_{i,i+1} = \frac{m-i}{m} \sum_{j=1}^{i} F^\text{ext}_j - \frac{i}{m} \sum_{j=i+1}^{m} F^\text{ext}_j.
\]

for \(i = 1, \ldots, m-1\). These interaction forces have to be non-negative, yielding the \((m-1)\) conditions in Eq. (6) for the subcluster to move as a whole (with \(m_j = m\)).

Moreover, each subcluster must become separated from its neighboring subclusters. This gives the set of conditions (5).

### B. Algorithm

We consider the positions \(x_i\) of the \(N\) rods to be ordered as

\[
x_1 < x_2 < \ldots < x_N,
\]

where \((x_{i+1} - x_i) \geq \sigma\). In the absence of the hardcore interaction, the particle positions could be propagated as

\[
x_i(t + \Delta t) = x_i(t) + \mu f(x_i) \Delta t + \sqrt{2D \Delta t} N_i,
\]

when applying the Euler scheme with time step \(\Delta t\) for solving Eqs. (1). Here \(N_i\) are independent random numbers drawn from a Gaussian distribution with zero mean.
and unit variance (standard normal distribution). Accordingly, the total external forces \( F^\text{ext} \) at time \( t \) are approximated by

\[
F^\text{ext}_i(t) = \frac{x_i(t + \Delta t) - x_i(t)}{\mu \Delta t} = f(x_i(t)) + \frac{1}{\mu} \sqrt{2D/t} N_i.
\]

Knowing these forces, the cluster analysis is carried out. For each cluster of touching particles, we identify the decomposition into subclusters according to the condition (5). Specifically, for a \( k \)-cluster with \( k > 1 \), we consider the fragmentation into each possible pair of subclusters \( (j, k - j), \ j = 1, \ldots, k - 1 \). For these subclusters, we calculate the respective mean external forces \( F^\text{ext}_j \) and \( F^\text{ext}_{k-j} \) using Eq. (12) for the force exerted on each individual particle of the subcluster. We split the \( k \)-cluster into that pair of subclusters, for which the difference \( (F^\text{ext}_{k-j} - F^\text{ext}_j) > 0 \) is the largest. Then we repeat this procedure for the two subclusters, and again for the pair of subclusters resulting from the two subclusters and so on. The procedure stops, if no further splitting into subclusters occurs.

After this step, it is known which particles form clusters that will move with the same velocity at the beginning of the time interval \( [t, t + \Delta t] \). For particles \( i = j, \ldots, j + n \) forming such cluster with \( n \) particles, the velocities according to Eq. (8) are

\[
v_i(t) = \frac{\mu}{n} \sum_{k=j}^{j+n} F^\text{ext}_k(t),
\]

with the \( F^\text{ext}_k(t) \) from Eq. (12).

Knowing the velocities, all particle positions are propagated within the time interval \( [t, t + \Delta t] \) in an event-driven procedure from a collision at time \( t_c \) to a next collision at time \( t'_c \).

Initially we set \( t_c = t \). A possible collision can occur between neighboring rods, if their velocities satisfy

\[
v_i(t_c) > v_{i+1}(t_c).
\]

The time of this possible collision is

\[
t_{i,i+1} = t_c + \frac{x_{i+1}(t_c) - x_i(t_c) - \sigma}{v_i(t_c) - v_{i+1}(t_c)}. \quad (15)
\]

The next collision realized is the one occurring at the smallest time in the set \( \{t_{i,i+1}\} \), i.e.,

\[
t'_c = \min_i \{t_{i,i+1}\}. \quad (16)
\]

If \( t'_c < (t + \Delta t) \), we determine the particle \( i_c \) taking part in this collision:

\[
i_c = \arg \min_i \{t_{i,i+1}\}. \quad (17)
\]

This collision leads to the merging of two clusters, where in the first cluster the rod \( i_c \) is the rightmost one and in the second cluster the rod \( (i_c + 1) \) is the leftmost one. As a consequence, after time \( t'_c \), the particles in the merged cluster move together and must have the same velocity, which is determined by the average external force, see Eq. (8). This implies that the velocities of the particles in the merging clusters become equal and are given by the weighted average of the velocities \( v_{i_c}(t_c) \) and \( v_{i_c+1}(t_c) \). If the first and second cluster contain \( m \) and \( n \) rods, respectively (merger of an \( m \)- with an \( n \)-cluster), the velocities \( v_j \) of the rods \( j = i_c - m + 1, i_c - m + 2, \ldots, i_c + n \) at time \( t'_c \) become

\[
v_j(t'_c) = \frac{m v_{i_c}(t_c) + n v_{i_c+1}(t_c)}{m + n}. \quad (18)
\]

This change of velocities means that the two clusters undergo a perfectly inelastic collision. The velocities of all other rods are kept, i.e. \( v_i(t'_c) = v_i(t_c) \) (no updating).

To keep the implementation simple, the positions \( x_i \) of all rods are updated at time \( t'_c \),

\[
x_i(t'_c) = x_i(t_c) + v_i(t_c)(t'_c - t_c), \quad i = 1, \ldots, N. \quad (19)
\]

This event-driven procedure given by Eqs. (14)-(19) is repeated until \( t'_c \geq t + \Delta t \). Then the positions of all rods \( i = 1, \ldots, N \) at time \( t + \Delta t \) are calculated using

\[
x_i(t + \Delta t) = x_i(t_c) + v_i(t_c)(t + \Delta t - t_c). \quad (20)
\]

Thereafter, updated external forces \( F^\text{ext}(t + \Delta t) \) at time \( t + \Delta t \) are calculated, the cluster analysis is per-

\[\text{Calculation of external forces [Eq. (11)]; } t_c \leftarrow t\]

\[\text{Cluster analysis [Eqs. (4), (5)] based on known cluster identification in previous time step}\]

\[\text{Calculation of velocities [Eq. (12)]}\]

\[\text{Calculation of set of times of possible collisions [Eqs. (13), (14)]}\]

\[\text{Determination of next collision time } t'_c \text{ and clusters taking part in collision [Eqs. (15), (16)]}\]

\[\begin{cases} t'_c \geq t + \Delta t \rightarrow \text{yes} \\ t'_c < t + \Delta t \rightarrow \text{no} \end{cases}\]

\[\text{Update of velocities of merging clusters [Eq. (17)]}\]

\[\text{Update of rod positions [Eq. (18)]}\]

\[\text{Update of rod positions [Eq. (19)]}\]

\[\begin{cases} t \leftarrow t + \Delta t \\ t_c \leftarrow t'_c \end{cases}\]

FIG. 2. Flow diagram of the BCD simulation algorithm [37].
formed again, and the event-driven procedure is carried out for the next time step.

We note that one could also update the external forces and perform the cluster analysis after each collision. This would lead to an implementation of the BCD algorithm with a variable time step. However, the use of a fixed time step becomes more efficient in particular in dense systems with high collision rate because the external forces do not need to be updated during a time interval $\Delta t$.

A flow diagram illustrating the algorithm is given in Fig. 2. A C++ code implementing this BCD algorithm, which uses vector manipulation procedures from Refs. [38, 39], is available on [37].

C. Validation: Comparison with exact analytical results

For testing our algorithm, we compare simulation results for density profiles $\varrho(x)$ and two-particle densities $\varrho_2(x, x + \sigma)$ at contact with analytical results derived from the exact density functional [40]

$$\Omega[\varrho(x)] = \int dx \varrho(x) \left\{ U(x) - \mu_{\chi} - k_B T \left[ 1 - \ln \left( \frac{\varrho(x)}{1 - \eta(x)} \right) \right] \right\}$$

of hard rods in equilibrium, where $U(x)$ is an external potential, $\mu_{\chi}$ is the chemical potential, and

$$\eta(x) = \int_{x-\sigma}^{x} dy \varrho(y).$$

We choose the periodic potential

$$U(x) = \frac{U_0}{2} \cos \left( \frac{2 \pi x}{\lambda} \right),$$

for which the integration limit in Eq. (21a) can be taken from zero to the wavelength $\lambda$.

The equilibrium density profiles follow by minimization of the density functional $\Omega[\varrho(x)]$. Numerically, we perform this minimization by evolving an initial profile using the dynamical density functional theory (DDFT) [41, 42] until the stationary equilibrium profile is obtained. As initial profile, we used the Boltzmann one $\varrho(x) \propto \exp[-U(x)/k_B T]$ of noninteracting particles with a normalization

$$\bar{\varrho} = \frac{1}{\lambda} \int_{0}^{\lambda} dx \varrho(x) = \frac{N}{L},$$

where the system length $L$ is a multiple of $\lambda$. The coverage (“volume fraction”) is equal to $\bar{\varrho} = N \sigma / L$.

The two-particle density at contact is [40, 43]:

$$\varrho_2(x, x + \sigma) = \frac{\varrho(x) \varrho(x + \sigma)}{1 - \eta(x + \sigma)},$$

To determine the density profile from our simulations, we discretize our periodicity interval into $M_{\text{bin}}$ bins $[x_\alpha, x_\alpha + \Delta x]$ with

$$x_\alpha = \alpha \Delta x, \quad \Delta x = 1 / M_{\text{bin}}.$$ (25)

After each time step $\Delta t$ in the simulation, we increase a counter $p_\alpha$ by the number of particles in bin $\alpha$. We thus obtain

$$\varrho(x_\alpha + \Delta x / 2) = \frac{p_\alpha}{M_{\text{samp}} \Delta x} + \mathcal{O}(\Delta x, M_{\text{samp}}^{-1}),$$ (26)

where $M_{\text{samp}}$ is the number of time intervals used in the sampling. Equation (26) allows one to calculate the density at position $x = x_\alpha + \Delta x / 2$ with controlled accuracy.

For determining the two-particle density at contact, we check after each time step whether one particle with coordinate $x$ falls into the bin $\alpha$ and another particle coordinate falls into the interval $[x + \sigma, x + \sigma + \Delta']$, where $\Delta' \ll 1$ is a small length. If this is the case, we increase a counter $q_\alpha$ for the bin $\alpha$ by one. We thus obtain

$$\varrho_2(x_\alpha + \Delta x / 2, x_\alpha + \Delta x / 2 + \sigma) = \frac{q_\alpha}{M_{\text{samp}} \Delta x \Delta'} + \mathcal{O}(\Delta x, \Delta', M_{\text{samp}}^{-1})$$

for calculating the two-particle density at contact if the left particle is at position $x = x_\alpha + \Delta x / 2$.

In the following we take $\lambda$ as length unit, $\lambda^2 / D$ as time unit, and $k_B T$ as energy unit ($\lambda = 1, D = 1, k_B T = 1$).

![Equilibrium density profiles of hard rods in the cosine potential (22) obtained from BCD simulations (symbols) and from calculations based on the exact density functional [Eq. (21a)]. The panels show results for various filling factors $\sigma = 0.2$, $\bar{\varrho} = 0.2$, $\sigma = 0.5$, $\bar{\varrho} = 0.8$.](image)
Furthermore we must set \( D = k_B T \mu \) for simulating equilibrium properties, i.e. \( \mu = 1 \) in our units. We consider a high amplitude \( U_0 = 6 \) of the external potential. In the simulations, we used the system length \( L = 100 \) with periodic boundary conditions, and chose \( \Delta x = \Delta' = 10^{-3} \) for the discretization lengths. The time step is reduced with increasing particle density \( \bar{\varrho} \) to properly resolve collisions between clusters. For the largest simulated \( \bar{\varrho} = 0.8 \) here, we chose \( \Delta t = 10^{-7} \).

In Figs. 3 and 4 we compare simulated equilibrium density profiles \( \varrho(x) \) with the theoretical profiles derived from the Percus functional. These figures show an excellent agreement between theory and simulation for various representative values of rod lengths \( \sigma \) and mean particle densities \( \bar{\varrho} \).

### IV. ADDITIONAL INTERACTIONS

If there are additional interaction forces between the hard rods, it is straightforward to include them in the numerical treatment as long as the interactions are continuous functions of the particle coordinates. We focus here on pair interactions and denote by \( f^{\text{int}}(x_i, x_j) \) the force of rod \( j \) exerted on rod \( i \). Such interactions can include van der Waals, electrostatic, magnetic, steric and/or depletion forces [44]. The Langevin equations for overdamped Brownian motion of the rods then are

\[
\frac{dx_i}{dt} = \mu f(x_i) + \mu \sum_{j=1}^{N} f^{\text{int}}(x_i, x_j) + \sqrt{2D} \xi_i(t),
\]

where we set \( f^{\text{int}}(x_i, x_j) = 0 \).

When applying the algorithm discussed in Sec. III B, the only change is that the external forces in Eq. (12) need to be replaced by

\[
F^\text{ext}_i(t) = f(x_i(t)) + \mu \sum_{j=1}^{N} f^{\text{int}}(x_i(t), x_j(t)) + \frac{1}{\mu} \sqrt{\frac{2D}{\Delta t}} N_i.
\]

The prominent potential for sticky hard spheres contains a \( \delta \)-singularity and accordingly does not give a continuous interaction force between the rods. It therefore requires a special treatment.

#### A. Treatment of Baxter’s sticky hard-sphere potential

Baxter introduced his model of sticky hard spheres to capture major characteristics of real interactions, which exhibit a repulsive core and attractive part. If the attractive part is a short-range surface adhesion between particles, the pair interaction potential \( V^{\text{int}}(r) \) may be modeled by including a \( \delta \)-function in the corresponding Boltzmann factor,

\[
\exp\left[ -\frac{V^{\text{int}}(r)}{k_B T} \right] = \Theta(r - \sigma) + \gamma \delta(r - \sigma).
\]

Here, \( \Theta(.) \) is the Heaviside step function, \( k_B T \) is the thermal energy, and \( r \) is the distance between particle positions. The parameter \( \gamma \) is a measure for the strength of the adhesive interaction. This yields the potential \([20, 21] \)

\[
V^{\text{int}}(r) = \begin{cases} \infty, & r < \sigma, \\ -k_B T \ln[1 + \gamma \delta(r - \sigma)], & r \geq \sigma. \end{cases}
\]

In the recent literature, particles with this type of interaction are often referred to as adhesive hard spheres.

Baxter’s model has been applied as an approximation to understand properties of Lennard-Jones fluids [45, 46]. Interestingly, when increasing the exponents in the Lennard-Jones potential, possible cluster formations approach the ones of Baxter’s model [47]. It was considered also to uncover effects of adhesive interaction on percolation behavior [22, 48–51]. Further applications include analysis of colloid sedimentation on liquid-solid [52] and fluid-fluid interfaces [53], different types of sequential adsorption [54], gelation [24, 26, 27, 55] and gel elasticity [28], aggregation of biomolecules [56, 57], crystallization of macromolecules [58], self-assembly of inactive [59] and active [60, 61] particles, detachment dynamics of colloidal spheres [62], viscosity of adhesive hard sphere dispersions [63, 64], and charge regulation...
of colloids inside electrolyte solutions [65]. Another
important application pertains to the analysis of scattering
experiments on colloids [66, 67] and nanoparticle mix-
tures [68–73]. The model has been used also to describe
dynamics in glassy states and liquid-glass phase transi-
tions [74–77]. The model has been extended to binary
mixtures of particles with different strength of Baxter’s
interactions [78, 79] and binary mixtures of different par-
ticles sizes [80]. Under certain conditions (size disparity
and high concentrations), such mixtures can undergo a
fluid-fluid phase separation [81]. Furthermore, models
based on Baxter’s model have been suggested for mix-
tures of colloids and telechelic polymers [82, 83] and for
protein-protein interactions between proteins of different
size [84].

In order to deal with the δ-singularity in Eq. (31), we
use a power-law representation δε(r) of the δ-function:

$$\delta_\epsilon(r) = \frac{p+1}{\epsilon^{p+1}} (\epsilon - r)^p \left[ \Theta(r) - \Theta(r - \epsilon) \right], \quad (32)$$

where $p > 1$. For $p \leq 1$, the force would no longer be
continuous at $r = \sigma + \epsilon$. The virtual interaction range
$\epsilon > 0$ needs to be much smaller than the average distance
between neighboring particles. For $\epsilon \to 0$, $\delta_\epsilon(r) \to \delta(r)$.

In Baxter’s original work [20], the δ-singularity in Eq. (31) was represented by a rectangular well, corre-
sponding to $p = 0$ in Eq. (32). This was particularly
suitable for the solution of the Percus-Yevick equations.
However, a rectangular potential well would yield δ-
singularities in the forces and is thus not a useful rep-
eresentation for simulation of Brownian dynamics based
on Langevin equations.

By inserting this representation in Eq. (31), we obtain
the interaction force

$$f_{\text{int}}^p(r) = \begin{cases} -k_B T \frac{\gamma p (\epsilon + \sigma - r)^{p-1}}{\sigma^{p+1}} + \gamma (\epsilon + \sigma - r)^p, & \sigma \leq r < \sigma + \epsilon, \\ 0, & r \geq \sigma + \epsilon. \end{cases} \quad (33)$$

Because of the constant one introduced in the argument
of the logarithmic function in Eq. (31), this interaction
force is continuous at $r = \sigma + \epsilon$.

While $\gamma$ is a physical parameter characterizing the
strength of the adhesive interaction, $p$ and $\epsilon$ are auxiliary
parameters defining the representation of the δ-function
in Eq. (31). What is a good choice of these auxiliary
parameters?

The parameter $\epsilon$ gives a virtual range of the attrac-
tive contact interaction and should be much smaller than
all relevant length scales in the system like the particle
size, mean distance between particles, and characteristic
lengths of the external force field, as, e.g., the wavelength
in a periodic field. At the same time, $\epsilon$ should not be
taken too small in order to resolve the virtual interaction
range, when particles come close to each other and are
driven by $f_{\text{int}}^p$ towards each other. This can be accounted
for by imposing the condition

$$\mu \max_r \{|f_p^\text{int}(r)|\} \Delta t \ll \epsilon, \quad (34)$$

which means that the maximum displacement caused by
the adhesive interaction in a time step $\Delta t$ is much smaller
than $\epsilon$.

In Fig. 5(a), we plot $f_{\text{int}}^p(r)$ for $\epsilon = 0.05$ and various
$p$. For this $\epsilon$, Fig. 5(b) shows that $\max_r \{|f_p^\text{int}(r)|\}$ is
smallest for $p \approx 2.95$. As $\epsilon$ should be as small as possible
in Eq. (34), this value for $p$ is preferable. For this rea-
soning, we used that $\max_r \{|f_p^\text{int}(r)|\}$ is a weakly varying function of $\epsilon$.

B. Validation: Comparison with exact
analytical results

Similarly as in Sec. III C, we test our treatment of Bax-
ter’s potential in connection with the algorithm described
in Sec. III B by comparison with exact analytical results
for equilibrium density profiles. These profiles are gen-
erated for the periodic potential in Eq. (22) by applying
the dynamic density functional theory, where now the
exact functional derivative of the density functional for
hard rods with contact interaction [21] is used:

$$\frac{\delta \Omega[\rho(x)]}{\delta \rho(x)} = \ln\left( \frac{K(x) \hat{K}(x - \sigma)}{\rho(x)} \right) + \frac{U(x) - \mu_{ch}}{k_B T} + \frac{1}{2} \ln(1 - \eta(x))\left(1 - \eta(x + \sigma)\right) - \frac{\sigma}{2\gamma} \frac{1}{1 - \eta(x + \sigma)} \nonumber$$

$$+ \int_{x-\sigma}^x \frac{dy}{2\gamma} \left[ \gamma \rho_y(y + \sigma) - \gamma \rho_y(y + \sigma) \right]^{\gamma} \left[ \rho(x + \sigma) - \rho_x(x + \sigma) \right]^{2} \nonumber$$

with

$$\hat{K}(x) = \frac{1}{2\gamma} \left( 1 - \gamma \frac{\rho(x + \sigma) - \rho(x)}{1 - \eta(x + \sigma)} \right) \nonumber$$

$$(35b)$$

$$+ \sqrt{ \left(1 - \gamma \frac{\rho(x + \sigma) - \rho(x)}{1 - \eta(x + \sigma)} \right)^2 + 4\gamma \frac{\rho(x + \sigma) - \rho(x)}{1 - \eta(x + \sigma)} }$$

and

$$K(x) = \hat{K}(x - \sigma) - \frac{\rho(x) - \rho(x - \sigma)}{1 - \eta(x)} \nonumber$$

$$(35c)$$

The solution of the structure equation \(\delta \Omega[\rho(x)]/\delta \rho(x) = 0\) gives the equilibrium density profile. As in the case of hard rods without contact interaction, we generated this equilibrium profile by evolving the corresponding DDFT equations into the stationary long-time limit. The computational time for reaching this long-time limit is much longer than for the hard rods without contact interaction, because of the more complex mathematical structure of the functional derivative \(\delta \Omega[\rho(x)]/\delta \rho(x)\) in Eqs. (35a)-(35c). For obtaining results in reasonable computing time, we were limited to use a spatial discretization length \(\Delta x = 10^{-3}\) [in units of \(\lambda\)] [85]. As we discuss below, this limitation causes small systematic deviations at high coverages \(\bar{\rho} \sigma\) in spatial regions of low particle densities.

In the simulations, all parameters are chosen as for the hard rods without contact interaction considered in Sec. III C. To represent the contact interaction we use the integer exponent \(p = 3\) in Eq. (32) and set \(\epsilon = 0.05\).

In Fig. 6, simulated equilibrium density profiles \(\rho(x)\) are compared with theoretical profiles calculated from Eqs. (35) for three strengths \(\gamma\) of the contact interaction and various representative rod lengths \(\sigma\) and mean particle densities \(\bar{\rho}\). For \(\bar{\rho} = 0.2\) and 0.5, the simulated data show an excellent agreement with the analytical predictions.

For \(\bar{\rho} = 0.8\) (graphs in the right column in Fig. 6), the agreement between simulations and the numerical solution of the structure equations is also good, but close to \(x = 0\) (or \(x = 1\)) small deviations are seen. Note that the density profiles in Fig. 6 are plotted on a logarithmic scale, i.e. the absolute deviation between the data obtained from the simulations and the numerical solution is very small.

In order to see whether the deviations are caused by our choice of \(\epsilon\) or the limited accuracy of the numerical solution of the structure equations, we in addition carried out simulations for a smaller value \(\epsilon = 0.025\) (blue circles in the right column of Fig. 6). For \(\gamma = 0.1\) and \(\gamma = 1\), the simulated results for \(\epsilon = 0.05\) and \(\epsilon = 0.025\) overlap, and for \(\gamma = 10\) the results are almost converged with respect to a decrease of \(\epsilon\). This indicates that the simulated data represent the true profiles and that the numerical solution of the structure equation is not perfectly accurate close to \(x = 0\). As discussed above, the accuracy of solving the structure equations is limited by the spatial resolution \(\Delta x\).

C. Time-dependent mean-square displacement

As an application of our method to a dynamical quantity, we show in Fig. 7 the time-dependent mean-square displacement (MSD) of a tagged rod for a system with \(U_0 = 0\) (no periodic potential) in the absence of a contact interaction \((\gamma = 0)\), and for sticky rods with \(\gamma = 1\). At short times, where the root of the MSD is smaller than
the mean interparticle spacing, the dynamics must reflect that of independent rods, i.e. the MSD increases linearly with time. The short-time diffusion coefficient is smaller in the presence of contact interaction due to enhanced cluster formation. When the MSD becomes comparable to the squared mean interparticle distance \((L/N - \sigma)^2\), as indicated by the dashed horizontal line in Fig. 7, collisions become relevant. The behavior then crosses over to the well-known anomalous single-file diffusion, where the MSD grows as a square root of \(t\). Interestingly, the subdiffusion is faster in the presence of contact interaction. A more detailed investigation of this effect and the change of behavior for \(U_0 > 0\) will be presented elsewhere.

V. SUMMARY AND CONCLUSION

Overdamped Brownian motion of particles is ubiquitous in soft matter and biological systems and clustering of particles is widely observed in the dynamics observed at high densities. Here we have presented a new method to simulate Brownian dynamics of clusters formed by hardcore interacting particles in one dimension. This method can be applied to the motion in arbitrary external force fields and allows one to determine interaction forces between particles in contact. It relies on a fragmentation and merging of clusters based on the evaluation of mean external forces acting upon individual clusters. This includes the random forces exerted by the fluid environment. We proposed an algorithm to implement the fragmentation and merging processes in an event-driven scheme.

Particle interactions beyond hardcore can be straightforwardly included into the algorithm as long as the corresponding interaction forces are continuous functions of the particle coordinates. Other types of interactions need addition considerations. In particular, Baxter’s sticky hard-sphere interaction for describing adhesive forces is an important case. We developed a method to tackle the \(\delta\)-singularity in this model which occurs when particles get in contact.

Our event-driven numerical method for treating cluster dynamics in collective overdamped Brownian motion has been successfully tested by comparison with theoretical results derived from exact density functionals, both for hardcore interacting systems and in the presence of the additional attractive contact interaction in the sticky hard-sphere model.

In view of the existing methods for treating Brownian motion of hard spheres, we consider our new method to fill a gap when it becomes necessary to model collective dynamics of particle clusters, as they can manifest themselves, for example, as Brownian solitons [36]. As for further developments of the method, a challenging task its extension to higher dimensions. This requires a refined procedure of cluster analysis and the cluster dynamics may be treated similarly as in the one-dimensional case by decomposing forces in the directions normal to the hard spheres at contact points and perpendicular to it.

VI. ACKNOWLEDGEMENTS

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