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I. INTRODUCTION

There has been considerable progress in the research of Brownian motors during the past decade (see, e.g., [1–3]). Starting with the simple pointlike Brownian particles with time-dependent driving forces, research has expanded toward more complex objects such as interacting Brownian particles (e.g., [4–9]) and polymers [10–12]. In this paper, we study polymer motion with discrete lattice models, which allows us to consider different kinds of microscopic polymer dynamics in detail. Aside from being a purely theoretical branch of study, analysis of simplified discrete nonequilibrium particle models has become an important tool for studying biologically inspired Brownian motor systems (e.g., [13]).

Discrete models have been applied widely to single-particle ratchet problems (e.g., [14–17]). We expand this picture by considering a generalized Rubinstein-Duke model (RD model) [18,19] for polymer motion in discrete time-dependent potentials. An interesting question is, what kind of dynamics lies beyond simple pointlike particles and how can one calculate its properties, such as the effective diffusion coefficient and the drift. Although there are plenty of studies concerning the behavior of the RD polymer in zero (or uniform) field (e.g., [20,21]), only recently has a ratchet mechanism (tilting ratchet) been considered in this context [22].

Especially because of the high complexity of Brownian motors with internal structure, most studies of these systems have applied the Monte Carlo method. However, since the ratchet systems are quite sensitive to the values of parameters, and drifts generated by pure ratchet mechanism are usually very small, Monte Carlo simulations tend to be very time-consuming and inaccurate. In this paper, we study these systems with the master equation approach. The results obtained in this way are accurate enough to reveal the details of the dynamics.

The purpose of this paper is to give a hands-on example of how one applies the master equation method to systems involving time-dependent periodic potentials and complex molecules by using a modified RD model polymer as a prototype of such molecules. We perform calculations for short linear polymers in a nonequilibrium environment generated by flashing and traveling ratchets. To test the significance of the polymer type, reptating or not, we compare the motion of the RD polymer with the dynamics of a modified version of the RD polymer with less constrained microscopic movement.

The paper is organized as follows. In Sec. II, we expand the modified RD polymer model to periodic time-dependent potentials and give equations for the calculation of the drift and diffusion coefficients. In Sec. III, we present results of the calculations for short polymers. Finally in Sec. IV, we give our conclusions and discuss the implications to applications.

II. THE MODEL AND METHODS

The RD model was originally developed to model the random motion of a flexible polymer in a confined medium with static obstacles (e.g., pores in gel) that the polymer must bypass, therefore causing the polymer reptation. By assuming that the network of obstacles can be modeled (on average) by a latticelike structure, that the correlation length between the polymer segments is smaller than the distance between the obstacles, and that only the polymer heads are able to move into previously unoccupied cells (lattice sites), the problem can be discretized to a simple particle hopping model [18]. Soon after the original model was expanded [19] to be suitable for external potentials (e.g., static field), pure theoretical research of the model started to flourish such as in Refs. [20,23].

Technically the RD model is a spin-1 chain with special kinds of nearest-neighbor interactions between the particles (reptons). By assuming that the reptons experience random...
“pushes” by the environment modeled with a continuoustime Markov process with exponentially distributed waiting times, we can construct the stochastic generator of the system.

In order to study the effect of the intrinsic transition rules of the polymer in time-dependent periodic potentials on long-time dynamics, we will compare the results of the RD model to the results of a nonreptating polymer, which allow the breaking of the reptation tube. In this paper, we call this extended model the free-motion model (FM model). In Fig. 1 there is an illustration of an example configuration of a six-repton polymer with arrows indicating all allowed moves for both RD and FM models (see also Ref. [24]). All repton transitions are between nearest-neighbor lattice sites only. Similar extensions have been studied previously in a different context in Refs. [24–26].

As an environment for the polymers, we assume a discrete periodic potential $V(x)$ such that

$$V(x + L) = V(x).$$

To make contact with Kramers rate theory (see, e.g., [27] and the previous work related to discrete ratchets [16], we define the transition rate from state $i$ to $j$ by

$$\gamma_{i \rightarrow j} = \alpha \exp\{\beta[V(i) - V(j)]\}$$

and choose $\alpha = \beta = 1$ to define the time and energy scales along with the lattice constant 1 to define the spatial length scale. We shall next define the time-evolution operators for the RD and FM models (readers not interested in the formal development may skip the rest of this section).

The mathematical model for the polymer, which contains the RD model as a special case but also allows breaking of the reptation tube if wanted, is constructed as follows (see, e.g., [28]). Within the most compact, the inner coordinate representation, every bond between reptons can be in three states; up (state A), down (state B), or flat (state 0). In Fig. 1, reptons 1, 2, and 3 are in state 0, repton 4 in state A, and reptons 5 and 6 in state B. An $N$-repton polymer has $N-1$ bonds. The state corresponding to polymer configuration $y$ is thus given by a $3^{N-1}$-dimensional state vector $|\Psi_y\rangle$.

The nonzero elements of the local creation and annihilation operators defining the dynamics of the bonds are

$$[a_i],_{1,1} = [a_i],_{2,2} = [b_i],_{3,3} = 1, \quad [a_i],_{1,2} = [a_i'],_{1,2} = [b_i],_{2,3} = [b_i'],_{3,2} = 1.$$ 

The operators $a$ and $b$ produce changes in the local bond configuration as indicated in Fig. 1. To extend the model to include a periodic potential $V$, we must add an additional state. One repton is chosen as a marker repton that keeps track of the polymer position within the potential. The transition rates of a single repton now depend on the position of the marker repton and all other bonds separating it from the marker. Either of the head reptons is the most convenient choice for the marker repton, hence we choose here the repton labeled 1 (see Fig. 1). The dimension of the marker state is $L$, so the dimension of the total system of equations becomes $L \times 3^{N-1}$.

By denoting

$$L(i) = \exp[-V(i + 1) + V(i)], \quad R(i) = \exp[-V(i - 1) + V(i)],$$

indicating transitions to the left and right (corresponding down and up in Fig. 1), the nonzero matrix elements for the marker state and transition operators are

$$[c_{1,l}],_{-1,l} = 1 \quad \text{for} \quad l \neq 1, \quad [c_{1,l}],_{l+1,l} = 1 \quad \text{for} \quad l \neq L,$$

$$[c_{1,l}],_{l,l} = [c_{1,l}],_{l+1,l} = [n_{1,l}],_{l} = 1,$$

where $1 \leq l \leq L$. The state of the polymer now has the form $|\text{marker repton}\rangle \otimes |\text{polymer configuration}\rangle = |\Psi\rangle \otimes |\Psi_\gamma\rangle$, where $|\Psi\rangle$ is the marker repton state vector with dimension $L$. The stochastic generator of the polymer model in the $L$-periodic potential thus becomes

$$H = \sum_{i=1}^{L} A_i + \sum_{j} \left( B_{j} + \sum_{i=1}^{N-2} M_{i,j},j \right),$$

where the operator $A$ applies to bond 1 and the marker repton, $M$ applies to bulk reptons, and $B$ applies to bond $N-1$. The explicit forms of these operators are given in Appendix A.

### A. Time-dependent potentials

The time dependence of the environment can break the detailed balance and may result in a directed drift. We assume that the switching between the distinct environments is independent of the polymer state in the potential, i.e., there is no feedback from the polymer. The switching mechanism between the potentials can be either stochastic or deterministic. The stochastic Markovian switching allows us to evaluate the stationary state directly by solving an eigenvalue problem. This is the most widely used way of studying Brownian motors and similar systems. With deterministic switching, we must numerically integrate to get the periodically stationary state.

Due to the time- and position-dependent transition rates, extra care must be taken to numerically study the process.
accurately. For example, one should not use the standard
discrete-time Monte Carlo simulation method that has been
widely used in various RD model studies. It does not pro-
duce correct results for our models. Instead one should
handle the master equation directly by means of numerical
integration or use the continuous-time Monte Carlo method.
In general, transport of particles in time-dependent potentials
is a hard problem to solve exactly. Even for a single particle
in a periodic potential, the general solution is not known.
The solution for stationary potentials, however, is available (pioneered by Derrida [29]).

First assume Markovian switching between the potentials.
We must include an additional state that keeps track of the
current potential,

$$|\text{potential state}\rangle \otimes |\text{marker repton}\rangle$$

$$\otimes |\text{polymer configuration}\rangle = |\Psi_s\rangle \otimes |\Psi_h\rangle \otimes |\Psi_y\rangle,$$

where $|\Psi_s\rangle$ is the state vector of the potential with dimension $S$, i.e., the number of different potentials. Since there is no feedback mechanism, adding this new state is straightforward. The nonzero state and transition operator elements for the potential state are

$$[\hat{h}_i]_{i+1,i} = 1 \text{ for } 1 \leq i \leq S, \quad [\hat{h}_S]_{1,S} = [n_s]_{x,i} = 1,$$

where $1 \leq s \leq S$. By defining the operator $\hat{h}$ like this, we consider only cyclic transitions between the potentials (i.e., $1 \rightarrow 2 \rightarrow \cdots \rightarrow S \rightarrow 1 \rightarrow \cdots$) to preserve the analogy with the deterministically switching potentials. The stochastic generator becomes

$$H = \sum_{s=1}^{S} [\hat{h}_s + T_s^{-1} (n_s - \hat{h}_s)],$$

where $\hat{h}_s$’s are formed by extending all elements of the operator in Eq. (1) with their corresponding potential state $s$ (e.g., $a_{i,j} \rightarrow \{n_a_{i,j}\}$ and $T_1, T_2, \ldots, T_S$ are the mean lifetimes of the potentials.

With deterministic switching, the stochastic generator is given by

$$H(i) = \begin{cases} H_1, & t \in [0, T_1) \\ H_2, & t \in [T_1, T_1 + T_2) \\ \vdots \\ H_S, & t \in \left[ \sum_{i=1}^{S-1} T_i, T \right), \end{cases}$$

where $H_s$ is the operator of the type (1) in the potential $s$ and $T = \sum_{i=1}^{S} T_i$ is the time period and $H(t+T) = H(t)$. In this case, there exists a $T$-periodic stationary solution. Once $H$ is given, the time evolution of the system is governed by the master equation $dq(t)/dt = H(t)q(t)$, where $q(t)$ is the probability vector. Since $H_s$’s are generally nonsymmetric, $q(t)$ usually has an oscillating behavior.

B. Drift and diffusion

We are interested in the drift and diffusion of the center of mass (CM) of the polymer. The velocity and the diffusion coefficient can be defined as

$$\mathbb{v} = \lim_{t \to \infty} \frac{d}{dt} \langle x_{CM}(t) \rangle,$$

$$D_{\text{eff}} = \frac{1}{2} \lim_{t \to \infty} \frac{d}{dt} \left[ \langle x_{CM}(t)^2 \rangle - \langle x_{CM}(t) \rangle^2 \right],$$

where $x_{CM}$ is the center of mass of the polymer. Here $\mathbb{v}$ and $D_{\text{eff}}$ could also be defined for single reptons instead of the center of mass and this local approach naturally leads to the same long-time values.

From the previous equations we define the Peclet number

$$\text{Pe} = \frac{|\mathbb{v}|}{D_{\text{eff}}},$$

where we choose the length scale $\ell = 1$. Since our polymer is simply composed of several neighbor-hopping random walkers, we can generalize the formalism of Ref. [17] (which generalizes the ideas of Ref. [29]). First define

$$q_y(t) = \sum_{n=0}^{\infty} p_{n,y}(t),$$

$$s_y(t) = \sum_{n=0}^{\infty} (l + nL)p_{n,y}(t) - \langle x_{CM}(t) \rangle q_y(t),$$

where $p_{n,y}$ is the probability to find the marker-repton in the position $l+nL$ with the polymer inner configuration $y$ and the redefined state $y$ includes both the marker-repton position $(l)$ and the inner configuration $(y)$ within the $L$-periodic potential. Assume that the stochastic generator $H$ of the total system is defined by the rates $\Gamma_{i,j}$ from state $i$ to $j$. It can be shown by using the definitions above, by taking the time derivatives and using the master equation (see, e.g., [30] for a similar calculation), that

$$\mathbb{v}(t) = h \sum_{y} (\bar{R}_y - \bar{L}_y)q_y(t),$$

where

$$\bar{R}_y = \sum_{i} \Gamma_{y,i}, \quad \bar{L}_y = \sum_{i} \Gamma_{i,y},$$

with arrows indicating the direction (right or left) of those repton transitions that lead from the state $y$ to states $i$, neglecting all the rest. Since this expression is for the center of mass, $h=1/N$ is chosen as the new lattice constant.

Similarly, we get

$$D_{\text{eff}}(t) = \frac{h^2}{2} \sum_{y} (\bar{R}_y + \bar{L}_y)q_y(t) + h \sum_{y} (\bar{R}_y - \bar{L}_y)s_y(t).$$

The evolution equations for $q_y(t)$ and $s_y(t)$ can be found by differentiating Eqs. (2) and (3) in time and using the master equation once more. We arrive at...
\[ \frac{dq_y(t)}{dt} = -\left( R_{\text{out}}^y + E_{\text{out}}^y \right) q_y(t) + \tilde{R}_{\text{in}}^y[q(t)] + \tilde{E}_{\text{in}}^y[q(t)], \]

\[ \frac{ds_y(t)}{dt} = -\left( R_{\text{out}}^y + L_{\text{out}}^y \right) s_y(t) + \tilde{R}_{\text{in}}^y[s(t)] + \tilde{L}_{\text{in}}^y[s(t)] - h[L_{\text{in}}^y q(t)] - \tilde{R}_{\text{in}}^y[q(t)] - v(t) q_y(t), \]

where

\[ \tilde{R}_{\text{in}}^y[q(t)] = \sum_i \Gamma_{i,y} q_i(t), \quad \tilde{R}_{\text{in}}^y[s(t)] = \sum_i \Gamma_{i,y} s_i(t), \]

\[ \tilde{L}_{\text{in}}^y[q(t)] = \sum_i \Gamma_{i,y} q_i(t), \quad \tilde{L}_{\text{in}}^y[s(t)] = \sum_i \Gamma_{i,y} s_i(t). \]

Note that all transitions are assumed to be between nearest-neighbor lattice sites only. Otherwise transitions of a certain length should be collected in their own sets according to their hopping distances (\(\leq L\)), which would appear as coefficients of additional sum terms in the equations. In the matrix form

\[ \frac{dq(t)}{dt} = H(t) q(t), \quad (4) \]

\[ \frac{ds(t)}{dt} = H(t) s(t) - hH_{\text{sign}}(t) q(t) - v(t) q(t), \quad (5) \]

where \(H_{\text{sign}}\) has the structure

\[ [H_{\text{sign}}]_{i,j} = [H]_{i,j} \text{ for all right transitions}, \]

\[ [H_{\text{sign}}]_{i,j} = -[H]_{i,j} \text{ for all left transitions}, \]

\[ [H_{\text{sign}}]_{i,j} = 0 \text{ for all other transitions } i,j. \]

This operator is easily built while building the stochastic generator itself. Since \(v(t)\) in Eq. (5) is governed by Eq. (4), these systems must be solved simultaneously. See also Refs. [31,32], where a similar approach has been applied to find the drift and the effective diffusion coefficient for complex molecules.

### 1. Time-independent stationary states

When \(H\) is time-independent, we can take the limit \(t \to \infty\) and define the steady-state parameters as

\[ Q_y = \lim_{t \to \infty} q_y(t), \quad S_y = \lim_{t \to \infty} s_y(t). \]

By using these, we get well-defined stationary values

\[ v = h \sum_y (\tilde{R}_{\text{out}}^y - \tilde{L}_{\text{out}}^y) Q_y, \]

\[ D_{\text{eff}} = \frac{\hbar^2}{2} \sum_y (\tilde{R}_{\text{out}}^y + \tilde{E}_{\text{out}}^y) Q_y + h \sum_y (\tilde{R}_{\text{in}}^y - \tilde{L}_{\text{in}}^y) S_y, \quad (6) \]

for the velocity and the effective diffusion coefficient. Now \(Q_y\)'s and \(S_y\)'s are found by solving the equations

\[ HQ = 0, \quad HS = hH_{\text{sign}} Q + v Q. \quad (7) \]

So far equations like these have been solved exactly only for a single particle on a periodic lattice. The first solution was given in Ref. [29] for the nearest-neighbor hopping particle with arbitrary transition rates. This was later extended, e.g., for parallel one-dimensional lattices in Ref. [30]. However, for more complex systems (like RD polymers), solutions cannot be found by exact methods and numerics must be applied. The structure of \(H\) also raises some issues. Since the determinant of \(H\) is always zero, mathematically there is no unique solution for the nonhomogeneous linear set of equations in Eqs. (7). This can be easily seen by using the fact that an ergodic stochastic system always has a nontrivial stationary state, therefore by the rank-nullity theorem we have \(\text{rank}(H) = \text{dim}(H) - 1\), meaning that we have one free parameter and all solutions are separated by a constant [i.e., \(S_y\) is a solution \(\iff (S_y + \text{const})\) is a solution]. However, since we also have conditions

\[ \sum_y Q_y = 1, \quad \sum_y S_y = 0, \]

which can be derived from the definitions of \(q_y(t)\) and \(s_y(t)\), there indeed exist unique solutions for \(S\) and \(Q\) (which is of course also required on physical grounds).

Equation (6) is a generalization of the result derived in Ref. [20]. This can be seen by considering the case \(L = 1\) without external potentials (i.e., \(y\)'s are simply inner configurations, \(v = 0\) and \(Q = Q_y = 3^{N-1}\)), so that with \(a_y := -2S_y/\hbar Q\) we have (for lattice constant 1)

\[ D_{\text{eff}} = \frac{\hbar^2}{2} Q \sum_y \left( \tilde{R}_{\text{out}}^y + \tilde{L}_{\text{out}}^y \right) + \frac{2}{\hbar Q} (\tilde{R}_{\text{out}}^y - \tilde{L}_{\text{out}}^y) S_y \]

\[ = \frac{1}{2N^2 3^{N-1}} \sum_y \left[ (r_y + l_y) - (r_y - l_y) a_y \right]. \]

Here we used the fact that, in this case, every state \(y\) has a weight \(1/3^{N-1}\) and \(r_y/l_y\)'s can be interpreted as the “number of arrows” for right-left transitions out from the state \(y\). In Ref. [20], symmetry properties (reflections) of polymer configurations were used to find a unique solution for \(a_y\)'s, but this is not possible when external potentials are present and the problem is nonsymmetric. However, numerical linear algebra tools can be used to find the solution.

### 2. Time-dependent stationary states

When \(H\) is time-dependent, we must integrate Eqs. (4) and (5) in time until the system arrives at the periodically stationary state (with period \(T\)). The mean velocity and the diffusion coefficient are determined by

\[ v = \lim_{t \to \infty} \frac{1}{T} \int_{t-T}^{t} v(s) ds, \]

\[ D_{\text{eff}} = \lim_{t \to \infty} \frac{1}{T} \int_{t-T}^{t} D_{\text{eff}}(s) ds. \]

In practice, these are calculated by integrating in time long enough so that results have converged.
C. Fast and slow switching regimes

When the switching times of the potential are close to the characteristic time scales of the system (i.e., relaxation times), the behavior depends heavily on the switching type and lifetimes of the states. However, when the potential changes very rarely or extremely fast, the system becomes independent of the switching type and even of the relative lifetimes of the states.

First assume that the total mean switching period $T \rightarrow 0$ such that $T_i \gg \tau_i$ for all mean lifetimes of the potentials (1 $\leq i \leq 5$). In this case, particles experience an effective average potential (“mean-field” [36]) and the transition rates become

$$\Gamma_{i,j}^{MF} = \sum_k x_k \Gamma_{i,j}^k,$$  \hspace{1cm} (8)

where $\Gamma_{i,j}^k$ are the transition rates of the stochastic generator of type (1) in the potential $k$ and $x_k=T_k/T$’s are weight factors determined by the mean lifetimes of the potentials. This leads to a mean-field stochastic generator with dimension $L \times 3^L$. This approach was used in Ref. [33] to solve exactly the single-particle dynamics in two arbitrary alternating periodic potentials. Although this mean-field limit is mathematically well defined, from the physical point of view it is artificial since real-world systems have inertia, and changing the potential state takes some finite time (e.g., charge redistribution to build up an electric field). So the velocity always goes to zero in the fast switching limit.

Now assume that $T_i \gg \tau_i$ for all $1 \leq i \leq S$, where $\tau_i$ is the longest relaxation time of the system in the potential $i$. This means that the system always converges close to the stationary state in the current potential before the potential is switched to the next one. By the model assumptions, drift is always zero at the stationary state in all potentials. Let $d_{j|i}$ denote the mean travel distance of the molecule center of mass within the potential $j$ using the stationary state of the potential $i$ as an initial state and then letting the system fully relax [37]. Summing over all $d_{j|i}$’s gives the total expected distance within one time period $T$, and by assuming cyclic switching of the states, we define

$$d = \sum_{i=1}^{S} d_{j|i+1}.$$  \hspace{1cm} (9)

The sign of $d$ determines the drift direction in the large $T$ limit and the asymptotic drift thus becomes $v = d/T$.

That internal molecular states may have a strong influence on the dynamics can be already seen in the slow switching regime. Letting the molecule first find its equilibrium in some nonflat potential and then turning the potential off may indeed result in directed motion of the molecule after the switching, due to internal relaxation, whereas a single particle would be immobile in the mean. These rebounds might be dominating and define the sign of $d$. 

**III. RESULTS**

A. Choice of the potentials

We have numerically analyzed RD and FM models with the polymer length of $N=1, \ldots, 11$ reptons and with two potentials ($S=2$) and stochastic switching. All calculations were done with MATLAB. We used the standard Runge-Kutta 4 method to integrate Eq. (4) in time and a trapezoid method to calculate the resulting integral in Eq. (9). The Arnoldi and BiConjugate gradient stabilized methods were used to solve homogeneous and nonhomogeneous systems in Eqs. (7).

We consider two basic potential types: flashing and traveling ratchets. The first type is the most general nonsymmetric potential that has been extensively used in studies of Brownian motors and the latter one is a generic example of asymmetrically placed symmetric potentials and has been recently used with single-particle models [17,34]. We consider the simplest case $L=3$, which is the smallest possible length that can form both of these potentials with the ratchet effect. See Fig. 2 for sketches of these potentials. A positive drift sign indicates motion in the increasing lattice-site index direction. Because of symmetries, the next choice would be $L=5$, but this choice would also need longer polymer lengths ($N \geq 11$) than we can efficiently handle. We require that the polymer must be able to cover several potential periods when fully extended. We also set $V_{max}=\frac{1}{2}$, which we found to give interesting results while also being computationally feasible [38]. The results concerning the general behavior and drift inversion do not significantly depend on the choice of $V_{max}$. With these parameter choices, the relaxation times $\tau$ for $N = 3, 11$ fall between $\ln(\tau)=0.8, \ldots, 4.8$ for RD polymers and $\ln(\tau)=0.3, \ldots, 3.4$ for FM polymers in all potentials studied here. Below we let $N$ and $T$ vary. By the limit $T \rightarrow 0$, we mean going to the mean-field stochastic generator with the rates given by Eq. (8). Overviews of the dynamics of polymers of lengths $N=1, 3, 5, 7, 9$ are given in Figs. 3

![FIG. 2. Sketch of the flashing nonsymmetric ratchet (left columns) and traveling symmetric ratchet (right columns) for $L=3$ (two period lengths shown).](image-url)
and 5, while Figs. 4 and 6 provide more detail for \( N = 1, \ldots, 11 \). We are especially interested in the current inversions and the general effects of the polymer size.

**B. Flashing ratchet potential**

Let us first define the time period \( T = T_{on} + T_{off} \) and the symmetry parameter \( x = T_{off} / T \), where \( T_{on} / T_{off} \) are the corre-

**FIG. 3.** (Color online) Drift and diffusion of the RD and FM polymers in the flashing ratchet \( (L = 3, V_{\text{max}} = \frac{1}{2}) \). (a)–(c) Drift as a function of the symmetry parameter \( x = T_{off} / T \) with the total flashing period \( T \rightarrow 0 \) (a), \( T \rightarrow \exp(3) \) (b), and \( T \rightarrow \exp(5) \) (c) with \( N = 1 \) (solid black), \( N = 3 \) (dash), \( N = 5 \) (dot), \( N = 7 \) (dash-dot), \( N = 9 \) [solid gray (blue)]. (d) Effective diffusion coefficient of \( N = 9 \) with \( T \rightarrow 0 \) (solid), \( T \rightarrow \exp(3) \) (dash), and \( T \rightarrow \exp(5) \) (dot).

**FIG. 4.** (Color online) Drift and Peclet numbers of the RD polymers (left column) and FM polymers (right column) in the flashing ratchet \( (L = 3, V_{\text{max}} = \frac{1}{2}) \) with \( N = 1 \) (solid black), \( N = 3 \) (dash), \( N = 5 \) (dot), \( N = 7 \) (dash-dot), and \( N = 9 \) [solid gray (blue)]. Symmetry parameters \( x = T_{off} / T \) are \( x = \frac{1}{5} \) (a), \( x = \frac{1}{3} \) (b), and \( x = \frac{3}{4} \) (c).

**FIG. 5.** Mean travel distances of the RD polymers (circles) and FM polymers (squares) in the flashing ratchet \( (L = 3, V_{\text{max}} = \frac{1}{2}) \) as a function of the polymer length in one time period at the asymptotic limit (i.e., the stationary state is reached before the switching).

**FIG. 6.** (Color online) Drift and diffusion of the RD and FM polymers in the traveling potentials \( (L = 3, V_{\text{max}} = \frac{1}{2}) \). (a)–(c) Drift as a function of the symmetry parameter \( x = T_{l} / T \) with \( T \rightarrow 0 \) (a), \( T \rightarrow \exp(3) \) (b), and \( T \rightarrow \exp(7) \) (c) with \( N = 1 \) (solid black), \( N = 3 \) (dash), \( N = 5 \) (dot), \( N = 7 \) (dash-dot), \( N = 9 \) [solid gray (blue)]. (d) The effective diffusion coefficient for \( N = 9 \) with \( T \rightarrow 0 \) (solid), \( T \rightarrow \exp(3) \) (dash), and \( T \rightarrow \exp(7) \) (dot).
in different scales of $T$. The drift in the positive direction (generated by the short slope) arises when the ratchet is switched on for such a short time that the larger rate of the short slope wins the smaller rate of the longer slope (see Fig. 2). Therefore, for increasing $T$, the ratio $x$ must get smaller to retain the dominance of the shorter potential slope, and finally $x$ goes to zero at $T \rightarrow \infty$.

When we add more reptons, the overall shape of the $\nu(T)$ curves remains very similar with small $T$'s. However, a clear effect of the polymer length and internal mechanisms can be seen with the long-time period $T = \exp(5)$, where the drift curve of the $N=9$ RD polymer turns positive for $x \approx \frac{1}{2}$. The velocities of the FM polymers remain on the negative side and no change in their general drift behavior can be seen as the parameters $N$ and $T$ are varied. In Fig. 3(d), we have plotted an example of $D_{\text{eff}}$ behavior of the $N=9$ polymers with corresponding $T$’s of Figs. 3(a)–3(c). Here $x=1$ simply gives the diffusion coefficient of the free polymers, and the diffusion constant in the static potential (at $x=0$) is always smaller. As can be seen, the effect of $T$ and $x$ on the diffusion is quite small in general.

Next we fix values $x=\frac{1}{3}, \frac{1}{2}, \frac{1}{3}$ and examine the $T$ dependence of the drift and the Peclet number in detail. The results in Fig. 4 reveal a complex behavior of the drift. The overall form of the curves is as expected: the drift and the Peclet number have some (local) maxima around $\ln(T)=0$. For small $T$, the single particle remains the fastest in all cases excluding $x=\frac{1}{2}$ for FM polymers, where it is the slowest one. However, as $T$ gets larger, longer polymers eventually become faster, which is caused by their longer relaxation time (short polymers have already reached their stationary state). This can be clearly seen from Figs. 4(b) and 4(c), but it also takes place in Fig. 4(a) to some extent. Similar behavior of coupled particles being faster than single ones and also having drift inversions were also reported in Ref. [5]. Although the relaxation times are quite different (see Sec. III A), the maxima of the drift fall close to $\ln(T) \approx 1$ for all polymer lengths and the position of the maximum Peclet number is almost constant. The drift sign change, already seen in Fig. 3, is present in Fig. 4(c).

The behavior of the Peclet number is very clear and similar in every case in Fig. 4: the larger the polymer, the larger the Peclet number. Thus the transport of longer polymers is more coherent than of shorter ones. Similar behavior was found in a continuum model consisting of elastically coupled Brownian particles [4]. By comparing the values of the Peclet number between polymer types, we see no significant differences between the curves. There is a slight difference for large values of $T$, where the Peclet number remains larger for FM polymers. This holds with every choice of parameters, excluding the possible current inversion points [e.g., the interval $\ln(T) \approx 1, \ldots, 0$ in Fig. 4(c)].

Next we take a closer look at the asymptotic behavior at $T \rightarrow \infty$. In Fig. 5, we have plotted the mean travel distance $d$ defined in Eq. (9). For $N=1,2$ there are no bulk reptons so the mean travel distances of RD and FM polymers may differ for $N \geq 3$ only. The calculation reveals that for long RD polymers ($N > 5$, a “critical length”) the rebound effect wins (i.e., $d > 0$) and the polymer starts traveling backwards while the single particle and FM polymers are traveling to the expected negative direction. The rebound effect is also present in FM polymers, but it is not strong enough to reverse the drift direction. For RD polymers with $L > 3$ with feasible polymer lengths, our Monte Carlo test simulations do not display this kind of an anomalous current inversion, suggesting that it may be related to spatial discretization and that longer-range interactions (e.g., stiffness) between reptons need to be introduced to see such inversions for $L > 3$.

We also note that a similar effect of multiple current inversions with tightly connected Brownian particles (rods) was reported in Ref. [12]. In that work, however, current inversions were not found for objects able to vary their length (rotating rods) in the ratchet direction, whereas the polymers in our work are able to vary their length between 1 and $N$ and still have drift inversion.

The reason for the stronger rebound effect of the RD polymer is caused by the strong tendency to enter (possibly deformed) U-shaped configurations because of the strict repulsion rule. After the potential is turned off, this shape unwinds and causes the drift. This also happens with time-dependent fields [22]. Since FM polymers lack the reptation rule, there is not as much variation in their shape as RD polymers have, thus resulting in a weaker rebound effect.

### C. Traveling ratchet potential

Let now $T = T_1 + T_2$ for the mean lifetimes $T_1$ and $T_2$ of the potentials depicted in the right column of Fig. 2 and define the asymmetry parameter $x = T_1 / T$. A Similar drift and diffusion behavior as previously reported in Refs. [17,34] for a single particle is expected. In Fig. 6, we show $\nu$ as a function of $x$ with three different $T$’s [Figs. 6(a)–6(c)]; $T \rightarrow 0$, $\ln(T) = 3$, and $\ln(T) = 7$. The behavior for the single particle is as expected; the drift is antisymmetric with respect to $x = \frac{1}{2}$ and goes to zero at $x = -\frac{1}{2}, \frac{1}{2}$. With longer polymers, the drift changes sign not trivially for large $T$’s [Fig. 6(c)] for both polymer types. This result is unexpected. An example of the behavior of the diffusion coefficient is shown in Fig. 6(d) for $N=9$ and different $T$’s. $D_{\text{eff}}$ always reaches its maximum at $x = \frac{1}{2}$ and decreases as the system goes to a static potential state at $x = 0$ and 1. The similarity of Figs. 3(a) and 6(a) is caused by the fact that, as can be easily seen from Eq. (8) for $T \rightarrow 0$, the traveling potential creates an effective rate structure similar to the nonsymmetric flashing ratchet.

Next we fix $x = \frac{1}{2}$ and examine the $T$ dependence in detail. In Fig. 7, we have plotted $\nu$ and the Peclet number for $\ln(T) = -4, \ldots, 7.5$. As $N > 2$, drift inversions can be seen around $\ln(T) = 2$ for both polymer types. As before, the single particle remains the fastest for small $T$, but eventually the drift curves begin to intersect as $T$ gets larger and the single particle is not always the fastest [see, e.g., the $N=3$ FM polymer in Fig. 7(a), right column]. The behavior of the Peclet number is as before: Longer polymers have more coherent transport, excluding the possible drift inversion points and their neighborhood. With small values of $T$, the Peclet number is also the same for both polymer types, but because of unequal drifts for moderate and large values of $T$ [$\ln(T) > 0$], also differences exist.
switching rates, we found nontrivial inversions of the polymer drift direction, which cannot occur with simple pointlike noninteracting particles. We also found that the Peclet number grows as the polymer gets longer and is largely independent of the polymer type, thus allowing more coherent transport for longer polymers. The overall polymer dynamics in ratchet potentials was found to be very model-specific. The discretization of the problem in this work may be far from many real-world applications but, nevertheless, since our model captures the essential characteristics of the Brownian motor system, we expect that similar properties could be found in the nanoscale objects that can be described with discrete states instead, such as molecular motors with internal structure. Drift inversions are especially interesting since they facilitate more efficient separation methods of molecules. The next step would be to consider larger L and N and the differences between deterministic and stochastic switching [35].

**APPENDIX A: OPERATORS IN H**

The explicit definitions of the operators in Eq. (1) are

\[ A_j(d) = \left\{ R(l) + L(l) \right\} \tilde{a}_{0.i,j} - R(l) \tilde{a}_{1.i,j} - L(l) \tilde{b}_{1.i,j} + L(l) \tilde{a}_{A.i,j} \]

\[ - L(l) \tilde{a}_{1.j} + R(l) \tilde{b}_{A.i,j} - R(l) \tilde{b}_{1.j}, \]

\[ B_{j,j'} = \left\{ R[l + f(N - 1, y)] + L[l + f(N - 1, y)] \right\} n_{0,0,y,j}, \]

\[ - R[l + f(N - 1, y)] a_{0,y,j}^i - L[l + f(N - 1, y)] b_{0,y,j}^i + L[l + f(N - 1, y)] a_{N,y,j}^i + R[l + f(N - 1, y)] n_{N,y,j}^i \]

\[ M_{i,j} = \left\{ R[l + f(i, y)] \right\} n_{A.i,j}^i n_{B.i,j}^i n_{0,i,y,j} + n_{0,i,y,j} n_{B.i,j}^i \]

\[ - a_{i,y,j} a_{i,y,j}^i - b_{i,y,j} b_{i,y,j}^i + L[l + f(i, y)] \]

\[ \times \left\{ n_{0,i,y,j} a_{i,y,j}^i + n_{i,y,j} a_{0,i,y,j} - a_{i,y,j}^i a_{i,y,j} + b_{i,y,j} b_{i,y,j}^i \right\} \]

\[ + \Omega R[l + f(i, y)] n_{A.i,j} n_{B.i,j} \]

\[ + \Omega L[l + f(i, y)] n_{B.i,j} n_{A.i,j} + n_{0,i,y,j} n_{B.i,j} \]

\[ - b_{i,y,j} a_{i,y,j}^i - a_{i,y,j} b_{i,y,j}^i, \]

where \( \Omega = 0 \) for RD polymers and 1 for FM polymers, and

\[ \tilde{a}_{i,j} = c_i a_{i,j}, \quad \tilde{a}_{1,j} = c_i a_{i,j}^i, \]

\[ \tilde{b}_{i,j} = c_i b_{i,j}, \quad \tilde{b}_{1,j} = c_i b_{i,j}^i, \]

\[ n_{1,i,j} = n_{i,j} n_{1,i,j}, \]

\[ x_{i,j} = n_{i,j} \left( \prod_{j=0}^{l-1} n_{g(y,j),j} \right) x_{i,j}, \]

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\[ n_{x,i,j} = \prod_{j=0}^{i-1} n_{g(y,j),j} n_{x,i} \]

with \( x \in \{a, b, a', b'\} \), \( z \in \{A, 0, B\} \). The function \( g(y,i) \in \{A, 0, B\} \) gives the state of the \( i \)th bond in the configuration \( y \), and the function \( f \),

\[ f(i,y) = \sum_{j=0}^{i} \langle \Psi_{n_{A,j} - n_{B,i}} | \Psi_{y} \rangle, \quad 1 \leq i \leq N-1, \]

gives the position of the repton \( i+1 \) in marker-centered coordinates. The detailed forms of the functions \( g \) and \( f \) depend on the selection of the state basis.

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[36] Not to be confused with the mean field in the sense of an approximation tool for the stochastic generator \( H \) itself.
[37] Note that when \( V \) is relatively small, the integral \( \lim_{t \to \infty} \int |p(t,s)| ds \) dies out fast and in practice we can truncate the integral at some moderate value of \( t \) (i.e., few times of the relaxation time).
[38] Since the asymmetry of the stochastic generator and the relaxation time of the polymer grow exponentially as the potential becomes larger, the problem becomes increasingly hard to handle numerically.