Diffusion coefficient and power spectrum of active particles with a microscopically reversible mechanism of self-propelling

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(Dated: August 05, 2022)

Catalytically active macromolecules are envisioned as key building blocks in development of artificial nanomotors. However, theory and experiments report conflicting findings regarding their dynamics. The lack of consensus is mostly caused by a limited understanding of specifics of self-propulsion mechanisms at the nanoscale. Here, we study a generic model of a self-propelled nanoparticle that does not rely on a particular mechanism. Instead, its main assumption is the fundamental symmetry of microscopic dynamics of chemical reactions: the principle of microscopic reversibility. Significant consequences of this assumption arise if we subject the particle to the action of an external time-periodic force. The particle diffusion coefficient then becomes enhanced compared to the unbiased dynamics. The enhancement can be controlled by the force amplitude and frequency. We also derive the power spectrum of particle trajectories. Among new effects stemming from the microscopic reversibility are the enhancement of the spectrum at all frequencies and sigmoid-shaped transitions and a peak at characteristic frequencies of rotational diffusion and external forcing. The microscopic reversibility is a generic property of a broad class of chemical reactions, therefore we expect that the presented results will motivate new experimental studies aimed at testing of our predictions. This could provide new insights into dynamics of catalytic macromolecules.

I. INTRODUCTION

Active microparticles capable of self-propelled motion are at the forefront of current research in physics, chemistry, and biology. The vivid interest in their properties is driven by potential applications, e.g., in microscopic robotics for targeted transport of drugs and microcargoes, for cleaning polluted habitats and working as sensors in biological environments to name a few. Inspired by molecular motors, there is an ongoing miniaturization efforts to develop nanosized artificial machines capable to operate, e.g., in the intracellular environment.

However, tracking individual physical processes that can induce the self-propulsion of nanoparticles is a rather challenging task. Instead, several experimental works have focused on average transport characteristics like particle’s diffusion constant. Conclusions regarding the nature and existence of self-propulsion are then inferred indirectly based on values of such ensemble-averaged characteristics and confronting them with results for basic theoretical models. Conversely, theoretical works often conjecture a specific mechanisms of self-propelling and discuss the behaviour of the diffusion constant. The resulting link between experimental data and theoretical models is thus indirect only. As a consequence, there are often conflicting views on the nature of active motion at these tiny scales and even on whether the catalytic activity of macromolecules can lead to their active self-propelled dynamics.

Despite a number of theoretical and experimental studies of dynamics of active macromolecules, the effects of a fundamental premise of nonequilibrium statistical mechanics – the principle of microscopic reversibility remains largely unexplored in this domain. In our recent work, we have analyzed a generic model of the active motion at the nanoscale. The developed model did not rely on a specific self-propulsion mechanisms, but assumed that both the chemical reactions powering the self-propulsion and the translational Brownian motion of the particle comply with the MR principle. We predicted an increased diffusivity and mobility of active nanoparticles compared to a passive particle and to models where MR is not included. As a result of assuming MR, the both parameters become dependent on an amplitude of constant external force applied on the nanoparticle.

Here, we extend the model of Ref. by incorporating a time-dependent external force of magnitude $F(t)$ acting upon the active nanoparticle. Motivated by spectroscopic measurements, we assume the force being a superposition of a constant and harmonically oscillating forces, i.e.,

$$F(t) = F_{dc} + F_{ac} \cos(\Omega t + \alpha).$$

Our main objective is to describe new qualitative effects stemming from MR in this model.
First, we will show how parameters of the external force (1), i.e., $F_{\text{dc}}$, $F_{\text{ac}}$ and $\Omega$, can modify the particle’s diffusion coefficient. Second, we will derive the power spectrum of the particle stochastic trajectories. The spectrum can be measured in single-particle tracking experiments. It contains more detailed information on the underlying dynamic mechanisms than the diffusion coefficient alone.

All discussed quantities will be compared with corresponding ones calculated for a reference model, where MR is neglected. Experimental tests of our predictions can decide on the relevance of the MR principle for the self-propulsion at the nanoscale. This in turn, can exclude from consideration (or confirm) a broad range of mechanisms obeying the assumed symmetry, narrowing down possibilities for theoretical modeling. Such tests can also guide an experimental development toward a resolution of existing controversies.

In the following Section II, we introduce our model of a nanoparticle with microscopically reversible propulsion (Sec. II A), discuss its coarse-grained continuous-space dynamics and estimate all model parameters in accord with relevant experimental data (Sec. II B). In addition, we solve the Langevin equations for the time-periodic driving force (Sec. II C), introduce reference models (Sec. II D and Sec. II E), and compare our approach with previous theoretical works (Sec. II F). In Sec. III, the derivation of two-time correlation functions is explained. These are then used to derive and discuss the particle’s diffusion coefficient (Sec. IV) and the power spectrum (Sec. V).

II. MICROSCOPICALLY REVERSIBLE ACTIVE PROPULSION

A. Microscopic model

Consider a particle driven by active propulsion and undergoing rotational and translational overdamped Brownian motion. Dynamics of the center of mass position $r$ of such a particle is governed by the Langevin equation \(^{37}\)

$$\frac{dr}{dt} = u_a(t)\mathbf{n}(t) + \mu F(t) + \sqrt{2D}\xi(t). \quad (2)$$

The last two terms on the right-hand side of (2) represent overdamped Brownian motion in the external force field $F$, $\mu$ is the mobility, $D = \mu k_B T$ the diffusion coefficient, where $k_B$ is the Boltzmann constant, and $T$ the temperature of ambient environment. Components of zero-mean Gaussian white noise vector $\mathbf{\xi}(t) = (\xi_x(t), \xi_y(t))$ satisfy $(\xi_i(t_1)\xi_j(t_2)) = \delta_{ij}\delta(t_1 - t_2)$.

In Eq. (2), $u_a(t)\mathbf{n}(t)$ denotes the active propulsion velocity with magnitude $u_a(t)$ and direction $\mathbf{n}(t)$ being a unit vector

$$\mathbf{n}(t) = (\cos \phi(t), \sin \phi(t)), \quad (3)$$

which we also call as the orientation of the particle. Over time, the orientation undergoes rotational diffusion\(^{38}\) characterized by the diffusion constant $D_\phi$, meaning that the angle $\phi(t)$ itself performs a Brownian motion

$$\phi(t) = \phi(0) + \sqrt{2D_\phi} \int_0^t \xi(t') dt', \quad (4)$$

driven by the delta-correlated zero-mean Gaussian white noise $\xi(t)$. The noises $\xi_x(t)$ and $\xi_y(t)$ are statistically independent.

It can be rather challenging to precisely control the particle orientation in experiments with nanoswimmers. Therefore, in this work, the initial value $\phi(0)$ in (4) is assumed to be a random variable homogeneously distributed within the interval $[0, 2\pi)$, yielding random initial orientation of the particle.

In the microscopic model, the magnitude $u_a(t)$ represents stochastic jumps of the particle driven by chemical reactions,\(^{37}\) see Fig. 1. We assume that a forward chemical reaction and the corresponding jump from $r$ to $[r + \mathbf{n}(t)\delta r]$ occur with rate $k_+$. The microscopic reversibility\(^{36,39-41}\) then ensures that there exists also a backward reaction accompanied by the jump from $[r + \mathbf{n}(t)\delta r]$ to $r$ and its rate $k_-$ is related to $k_+$ by the local detailed balance
condition \( k_+ / k_- = \exp \left( \frac{\Delta G - \delta W}{k_B T} \right) \),

where \( \Delta G \) stands for the reaction free energy and \( \delta W \) for the work done on the particle by all external (mechanical, electromagnetic) forces acting on the particle during its jump \( r \to [r + n(t)] \delta r \).

The condition (5) follows from reversibility with respect to time-reversal of a more detailed microscopic dynamics (classical Hamiltonian or quantum) upon which our effective stochastic time-evolution (2) is constructed via elimination of fast degrees of freedom.\(^{37}\)

B. Thermodynamically consistent propulsion at macroscale: Estimating parameters of the model

Miscellaneous single-particle tracking methods,\(^{20,44,45}\) pulsed field gradient nuclear magnetic resonance (NMR),\(^{30,31,46,47}\) neutron scattering,\(^{48}\) and other experimental techniques capable of measure diffusivity of nanoparticles have a common feature: They typically probe the particle dynamics on length scales much larger than \( \delta r \) and time-intervals between measurements are longer than the duration of individual reactions.

When observed on such macroscopic time and length scales, the magnitude of active velocity can be approximated by\(^{37}\)

\[ u_a(t) \approx u + \mu_c F_n(t) + \sqrt{2D_c} \xi_a(t), \tag{6} \]

with

\[ F_n(t) = n(t) \cdot F(t), \tag{7} \]

being the projection of external force \( F(t) \) onto the particle orientation \( n(t) \). During the coarse-graining procedure, the projection (7) arises from the work \( \delta W \approx F_n(t) \delta r \) present in Eq. (5). The constant term \( u \) in (6), on the other hand, is proportional to the reaction free energy per elementary displacement: \( (u/\mu_c) \delta r = \Delta G \), where the mobility \( \mu_c \) is related to \( D_c \) by the fluctuation-dissipation theorem \( \mu_c k_B T = D_c \) and can be estimated based on the reaction rate constants \( k_{\pm} \) measured in absence of the external force \( \{F(t) = 0\} \)

\[ D_c \approx \frac{\langle \delta r \rangle^2}{2 (k_+ + k_-).} \tag{8} \]

This “diffusion constant” sets magnitude of fluctuations in number of chemical reactions powering the particle active dynamics. At the macroscale, these fluctuations are represented by zero-mean Gaussian white noise \( \xi_a(t) \) in (6).

The derivation of Eq. (6) from the microscopic model can be found in our previous work in Ref. 37. The essence of this derivation is as follows: In the microscopic model, the active motion is represented by a random walk in the direction of the particle orientation. The forward (backward) steps of length \( \delta r \) happen with the rate \( k_+ \) (\( k_- \)). The continuum (diffusive) approximation of the dynamics of this random walk process, subjected to the condition that reaction rates \( k_\pm \) obey the detailed balance condition (5), leads to the Langevin equation with velocity (6).

Relation (8), \( \mu_c k_B T = D_c, \mu k_B T = D \), and Stokes’ law \( \mu = 1/6 \pi \eta R \) with \( \eta \) being the dynamic viscosity of ambient environment and \( R \) the hydrodynamic radius of nanoparticle, allow us to estimate magnitudes of all the model parameters, which we will use for illustrating our results in Figs. 2, 3, and partly in Fig. 4. For this purpose, we adopt numerical values inspired by ones reported for catalytically active urease.\(^{20-22}\) We take \( k_+ \approx 10^5 \, \text{s}^{-1} \), \( D_c \approx 10^5 \, \text{s}^{-1} \), \( \delta r \approx 5 \, \text{nm} \). Furthermore, we neglect \( k_- \), set \( T \approx 300 \, \text{K} \) for the room temperature, and \( \eta \approx 8.5 \times 10^{-4} \, \text{Ns/m}^2 \) for the dynamic viscosity of water at this temperature. As for \( u \), surprisingly, its impact on the results turns out to be insignificant even for relatively large velocities like \( u = 1 \, \mu \text{m/s} \).

C. Harmonic driving force

In general, the total external driving force \( F(t) \) in Eq. (7) can represent any mechanical, electromagnetic, van der Waals, and other forces exerted on the active particle during its dynamics. Here, we assume the \( r \)-independent force \( F(t) = (F(t), 0) \) acting along the \( x \) coordinate axis with harmonically oscillating amplitude given in Eq. (1), where \( \Omega \) is the angular frequency of oscillations, \( F_{dx} \) and \( F_{dx} \) are positive constants, and \( \alpha \) is the initial phase. To eliminate transient effects caused by a particular value of \( \alpha \), we shall average all following results over \( \alpha \in [0, 2\pi] \).\(^{49}\)

We are interested in this particular form of \( F(t) \) with an outlook on experimental verification of our predictions, since a qualitatively similar external driving appears in many spectroscopic methods.

For such driving, the projection \( F_n(t) \) defined in Eq. (7) reads \( F_n(t) = F(t) \cos \phi(t) \). It is multiplied by \( \mu_c \cos \phi(t) \) [\( \mu_c \sin \phi(t) \)] in the Langevin equation for \( x(t) \) [\( y(t) \)]. After integrating the Langevin equations subjected to initial conditions \( x(0) = y(0) = 0 \), we get

\[ x(t) = \int_0^t \left\{ \left[ \mu + \mu_c \cos^2 \phi(t') \right] F(t') + u \cos \phi(t') \right. \]

\[ + \sqrt{2D} \xi_x(t') + \cos \phi(t') \sqrt{2D} \xi_n(t') \right\} dt', \tag{9} \]

\[ y(t) = \int_0^t \left\{ \mu_c F(t') \sin \phi(t') + u \sin \phi(t') \right. \]

\[ + \sqrt{2D} \xi_y(t') + \sin \phi(t') \sqrt{2D} \xi_n(t') \right\} dt'. \tag{10} \]

These integral expressions constitute the starting point for all following derivations.
D. Reference cases

Assuming the detailed balance condition (5) has remarkable consequences for the velocity magnitude \( u_a(t) \) in Eq. (6): \( u_a(t) \) becomes dependent on the external force and the noise strength \( D_c \) in Eq. (6) satisfies the fluctuation-dissipation theorem \( D_c = \mu c k_B T \). Setting heuristically

\[
  u_a(t) = u = \text{const}
\]

would eliminate the force-dependence and the noise. Also, for nano-sized particles, this would obliterate the connection between the coarse-grained model based on the Langevin equation (2) with \( u_a(t) = u \) and a microscopic model, where chemical kinetics responsible for the self-propelling is consistent with the principle of microscopic reversibility.\(^{37}\)

Formally, neglecting the consequences of MR in our model, in particular the force-dependence in (6), can be realized by setting \( \mu c = D_c/k_B T = 0 \) in Eq. (6). Therefore, in the following, we shall refer to the Langevin model of the nanoparticle dynamics with \( u_a(t) = u \) as to the one where MR of the active propulsion mechanism is neglected.

Accordingly, any term occurring in the resulting formulas that would explicitly depend on \( \mu c \) or \( D_c \), can be regarded as stemming from MR of the active propulsion mechanism.

The dynamics in the limit \( \mu c \to 0 \) shall serve as a reference case in all discussions of our results. Moreover, we will normalize all plotted results by corresponding quantities calculated for \( u = 0 \) and \( \mu c = D_c/k_B T = 0 \), i.e., by the results for the passive two-dimensional overdamped Brownian motion (BM).

E. Relation to dynamics of micron-sized particles

Let us note that the model with dynamics obeying the Langevin equation (2) with the constant active velocity (11) is known in the literature simply as the active Brownian particle (ABP) model.\(^{50-65}\) ABP model describes well the dynamics of micron-sized Janus colloidal particles driven by a large number of chemical reactions per second. Due to this large scale, the active motion of micron-sized particles is nearly deterministic and the aforementioned effects of MR are expected to be negligible. That is, for micron-sized particles, the terms in (6) containing \( \mu c \) are expected to be much smaller than \( u \).

F. Preliminary work and related models with MR

In Ref. 37, we have analyzed the microscopic model of Sec. II A for the case of a constant (time-independent) external force. In that simpler situation, we could derive exact analytical expressions for first two moments of the particle position. The expressions are valid for the fully microscopic dynamics and all \( t \). We have then explained in detail the derivation of the macroscopic limit for active velocity presented in the current Sec. II B and found the corresponding moments in this limit. Comparing the results, it turned out that the long-time behavior is qualitatively similar for both levels of description.

In the current, technically more demanding, time-dependent case, having in mind this equivalency revealed in the previous study,\(^{37}\) we analyze the macroscopic model only. Another advantage of focusing on the macroscopic limit is that it allows to access a more advanced quantity compared to that discussed in the previous microscopic analysis – the full power spectrum of an ensemble of stochastic trajectories.

Let us now sum up some other situations, where certain aspects of MR of the active propulsion were used. In fact, stochastic models of active particles with MR have emerged relatively recently.\(^{56,67}\) Assuming MR, included into Markovian models by means of the local detailed-balance condition, \(^{68,69}\) is crucial for defining the entropy production of individual stochastic trajectories\(^{66}\) within the formalism of stochastic thermodynamics.\(^{67}\) Following works focused on particle’s phoretic velocity,\(^{70}\) motility-induced phase separations,\(^{71}\) and performance of active heat engines.\(^{72,73}\)

Another remarkable class of theoretical studies relying on MR focuses on linearized dynamics of both chemical concentrations and mechanical degrees of freedom.\(^{74-80}\) MR is then enforced via coupling time-evolutions of the two types of variables by a symmetric matrix of Onsager coefficients, similarly to the formalism of classical linear irreversible thermodynamics.\(^{81}\) In contrast to these works, in our case, there is no explicit modeling of time-evolution of chemical degrees of freedom. They are treated by means of so called chemiostats, i.e., thermodynamic reservoirs of chemical free energy.

III. TWO-TIME CORRELATION FUNCTIONS

In this section, we derive two-time correlation functions, which form mathematical foundations for the physical discussion of diffusion coefficients (Sec. IV) and the power spectra (Sec. V).

To proceed with the calculations, we first must evaluate various averages over the particle orientation as given by the angle \( \phi(t) \), which performs a Brownian motion, see Eqs. (3) and (4). In particular, mean values of \( \sin \phi(t) \) and \( \cos \phi(t) \) are zero,

\[
  \langle \sin \phi(t) \rangle = \langle \cos \phi(t) \rangle = 0,
\]

for all \( t \geq 0 \) meaning that there is no preferable orientation of the particle, i.e., \( \langle n(t) \rangle = 0 \).
Furthermore, we have
\[
\langle \cos(\phi(t_1)) \cos(\phi(t_2)) \rangle = \langle \sin(\phi(t_1)) \sin(\phi(t_2)) \rangle = \frac{1}{2} e^{-D_r|t_1-t_2|},
\] (13)
\[
\langle \cos^2(\phi(t_1)) \cos(\phi(t_2)) \rangle = \frac{1}{4} + \frac{1}{8} e^{-4D_r|t_1-t_2|},
\] (14)
\[
\langle \sin(\phi(t_1)) \cos(\phi(t_1)) \sin(\phi(t_2)) \cos(\phi(t_2)) \rangle = \frac{1}{8} e^{-4D_r|t_1-t_2|},
\] (15)
and
\[
\langle \sin(\phi(t_1)) \cos(\phi(t_2)) \rangle = 0,
\] (16)
\[
\langle \cos^2(\phi(t_1)) \cos(\phi(t_2)) \rangle = \langle \sin^2(\phi(t_1)) \cos(\phi(t_2)) \rangle = 0,
\] (17)
\[
\langle \sin(\phi(t_1)) \cos(\phi(t_1)) \cos(\phi(t_2)) \rangle = 0,
\] (18)
\[
\langle \sin(\phi(t_1)) \cos(\phi(t_1)) \sin^2(\phi(t_2)) \rangle = 0.
\] (19)

These identities may be derived by rewriting the goniometric functions in terms of complex exponentials and taking mean values of resulting expressions with respect to the Gaussian distribution of process \( \phi(t) \).

We shall also use the following average over the initial phase of the external force
\[
\int_0^{2\pi} \cos(\Omega t + \alpha) \cos(\Omega t_2 + \alpha) \frac{d\alpha}{2\pi} = \frac{1}{2} \cos[\Omega(t_1-t_2)],
\] (20)
and the fact that the mean value of \( \cos(\Omega t + \alpha) \) when \( \alpha \in [0,2\pi] \) is zero.

Averaging Eqs. (9) and (10) over all noises and \( \alpha \), specifically using Eqs. (12), (13), and (16), we arrive at expressions
\[
\langle x(t) \rangle = \left( \mu + \frac{\mu_c}{2} \right) F_{dc} t,
\] (21)
\[
\langle y(t) \rangle = 0,
\] (22)
giving mean values of particle coordinates at time \( t \). Thus, the mean position \( \langle r(t) \rangle = \langle x(t), y(t) \rangle \) drifts along the direction of \( F(t) \) with the enhanced mobility \( \mu + \mu_c/2 \) as compared to both the passive Brownian particle and ABP case where \( \mu_c = 0 \).

To examine diffusive dynamics in the presence of such mean drift, it is instructive to focus on the displacement of the particle measured relative to its mean position,
\[
\Delta r(t) = r(t) - \langle r(t) \rangle,
\] (23)
with individual coordinates being \( \Delta x(t) = x(t) - \langle x(t) \rangle \) and \( \Delta y(t) = y(t) - \langle y(t) \rangle = 0 \).

In experiments, one most often measures quantities that can be derived from two-time correlation functions
\[
C_{xx}(t_1,t_2) = \langle (\Delta x(t_1) \Delta x(t_2)) \rangle,
\] (24a)
\[
C_{xy}(t_1,t_2) = \langle (\Delta x(t_1) \Delta y(t_2)) \rangle,
\] (24b)
\[
C_{yx}(t_1,t_2) = \langle (\Delta y(t_1) \Delta x(t_2)) \rangle,
\] (24c)
\[
C_{yy}(t_1,t_2) = \langle (\Delta y(t_1) \Delta y(t_2)) \rangle.
\] (24d)

Starting with the calculation of \( C_{xx}(t_1,t_2) \), we subtract \( \langle x(t) \rangle \) given in (21) from \( x(t) \) in Eq. (9) and average the product \( \Delta x(t_1) \Delta x(t_2) \) over all noises and the initial phase \( \alpha \) using Eqs. (13), (14), (16), (17), and (20). This yields
\[
C_{xx}(t_1,t_2) = (2D + D_c) \min(t_1,t_2)
\] (25)
\[
+ \int_0^{t_1} \int_0^{t_2} \left\{ \frac{u^2}{2} e^{-D_r|t'|-t_2|} + \left( \mu + \mu_c \right) \frac{F^2}{2} \cos[\Omega(t_1-t_2)] \right\} e^{-4D_r|t_1-t_2|} \mathrm{d}t'_2 \mathrm{d}t_1.
\]

Here, terms on the first line result from \( \delta \)-correlated noises \( \xi_x \) and \( \xi_\alpha \) in (9), the both terms on the second line would be present also in ABP model (with \( \mu_c = 0 \)), and, the second term also in the harmonically driven passive Brownian motion (\( \mu = 0, \mu_c = 0 \)). All the terms on the third line are intrinsic to the present microscopically reversible model and would occur neither in ABP nor in BM models.

Calculation of the two-time correlation function \( C_{yy}(t_1,t_2) \) proceeds along similar lines. The result reads
\[
C_{yy}(t_1,t_2) = C_{xx}(t_1,t_2)
\] (26)
\[
- \int_0^{t_1} \int_0^{t_2} \left\{ \mu + \mu_c \right\} \frac{F^2}{2} \cos[\Omega(t_1-t_2)] \mathrm{d}t'_2 \mathrm{d}t_1.
\]

That is, \( C_{yy} \) differs from \( C_{xx} \) merely by the second term on the second line of Eq. (25).

For the sake of further analysis, we have solved all integrals in Eqs. (25) and (26). The resulting (somewhat extensive) expressions are presented in Appendix, see Eqs. (A.1) and (A.2).

Finally, after averaging the products \( \Delta x(t_1) \Delta y(t_2) \) and \( \Delta y(t_1) \Delta x(t_2) \), for the cross-correlations, we get
\[
C_{xy}(t_1,t_2) = C_{yx}(t_1,t_2) = 0.
\] (27)

IV. DIFFUSION COEFFICIENTS

Sum of results (25) and (26) for \( C_{xx} \) and \( C_{yy} \) evaluated at \( t_1 = t_2 = t \) gives us the mean squared displacement (MSD) of the particle
\[
\langle [\Delta r(t)]^2 \rangle = C_{xx}(t,t) + C_{yy}(t,t).
\] (28)

When studied in experiments, MSD can provide a valuable insight into the type of microscopic propulsion mechanism used by the nanoparticle. Its exact analytical expression contains a superposition of linear, bounded oscillating and exponentially decaying terms that follow from Eqs. (A.1) and (A.2) at \( t_1 = t_2 = t \). Let us inspect two experimentally relevant regimes where this result considerably simplifies.

Rotational diffusion is the fastest diffusive process in our model. It happens on the characteristic time scale \( \sim 1/D_r \). If the time-resolution in an experiment is high enough to
The long-time diffusion coefficient $D_{\text{ABP}}$ would observe the linear growth of MSD superimposed with the ballistic term,

$$\langle (\Delta \mathbf{r}(t))^2 \rangle \approx 4 \left( D + \frac{D_c}{2} \right) t + \left[ u^2 + \frac{\mu_c^2}{4} \left( F_{\text{dc}}^2 + F_{\text{ac}}^2 \right) / 2 \right] t^2,$$

for $t \ll 1/D_r$.

Equation (29) gives MSD of very short persistent trajectories. While moving along such a trajectory, the particle essentially does not rotate. Therefore, the diffusive growth of MSD described by the linear term in (29) is caused just by the translational Brownian motion ($D$) and fluctuations in the active propulsion velocity $u_a(t)$ ($D_c/2$). The magnitude of the ballistic term ($\sim t^2$) is enhanced by force amplitudes $F_{\text{ac}}^2$ and $F_{\text{dc}}^2$ as compared to the corresponding result for ABP model.\(^{82-84}\)

On the other hand, experimental techniques such as the pulsed field gradient NMR,\(^{30,31,46,47}\) neutron scattering,\(^{48}\) and various single-particle tracking methods\(^{20,44,45}\) can measure the long-time effective diffusion coefficient

$$D = \lim_{t \to \infty} \frac{\langle (\Delta \mathbf{r}(t))^2 \rangle}{4t}. \quad (30)$$

In this limit, we have

$$\lim_{t \to \infty} \frac{\langle (\Delta x(t))^2 \rangle}{t} = \lim_{t \to \infty} \frac{\langle (\Delta y(t))^2 \rangle}{t} = \lim_{t \to \infty} \frac{\langle (\Delta r(t))^2 \rangle}{2t}, \quad (31)$$

i.e., the diffusive spreading of the particle probability density function around its mean position is isotropic. This holds despite the fact that the driving force $F(t)$ breaks spatial isotropy of the problem by dragging the particle along the $x$ axis.

By evaluating the limit in (30) we get the effective diffusion coefficient

$$D = D + \frac{D_c}{2} + \frac{u^2}{2D_r} + \frac{(\mu_c F_{\text{dc}})^2}{32D_r}$$

$$+ \frac{(\mu_c F_{\text{ac}})^2}{4D_r} \frac{D_r t^2}{(4D_r + \Omega^2)}.$$ \quad (32)

The long-time diffusion coefficient $D$ depends on the constant part of active velocity $u$, on the magnitude and frequency of the external force $F(t)$, and on the rotational diffusion constant $D_r$ in a rather nontrivial manner. Physical origins of individual contributions to $D$ are as follows.

The contribution $(D + D_c/2)$ is identical to the short-time diffusivity in Eq. (29). Its first part reflects the passive translational Brownian motion, for which we have

$$D_{\text{nm}} = D$$ \quad (33)

even in the presence of the external harmonic driving. The second part, $D_c/2$, arises due to fluctuations in magnitude of the active velocity $u_a(t)$, Eq. (6).

The $u$-dependent term in (32) quantifies uncertainty of the particle position stemming from the rotational diffusion of the constant part $u$ of the active velocity $u_a(t)$. This contribution vanishes with increasing $D_r$ (at fixed $u$) since the rapidly rotating direction of the velocity has hardly any effect on the translational motion. This term appears also in the effective diffusion coefficient for the harmonically driven ABP model:

$$D_{\text{ABP}} = D + \frac{u^2}{2D_r}. \quad (34)$$

Remarkably, the value of $u^2/2D_r$ is rather low as compared to $D$ of a nanoparticle. Using parameter estimates from Sec. II B, we get $D \approx 1.3 \times 10^{-11}$ m$^2$/s$^2$ and $u^2/2D_r \approx 5 \times 10^{-18}$ m$^2$/s$^2$ at $u = 1$ mm/s, which is a relatively large value of $u$ when related to particle's hydrodynamic radius $R_H \approx 20$ nm. The both terms become comparable at $u \approx 1.6$ mm/s that would correspond to a nanoparticle traveling at staggering $80000R_H$ per second.

The fundamental reason for $u^2/2D_r$ being vanishingly small as compared to $D$ is that rotational diffusion of small particles since the rotational diffusion coefficient as $D_r \sim 1/R_H^2$.

Although the $F_{\text{dc}}$-dependent term in (32) might seem similar to the $u$-dependent one [both $(\mu_c F_{\text{dc}})$ and $u$ have dimensions of velocity], there is a significant difference in physical origins between the two. While the velocity $u(t)$ rotates with the particle, the line of action of the external force $F(t)$ is fixed in space (parallel with the $x$ axis). The force, however, influences the magnitude of active velocity $u_a(t)$, as described by the projection $F_n(r,t)$ in Eq. (6). When $n(t)$ and $F(t)$ are parallel, the projection attains its maximal magnitude and it vanishes for perpendicular orientation of the two vectors. Because $n(t)$ rotates erratically, the stochastic changes of $F_n(r,t)$ enhance uncertainty of the particle position as quantified by $D$.

The last, $F_{\text{ac}}$-dependent term in (32) appears for analogous physical reasons. Notably, its magnitude can be controlled by the frequency $\Omega$ of the external force. This dependence is reflected in Fig. 2 showing $D$ from Eq. (32) (solid line) and $D_{\text{ABP}}$ from (34) (dashed), both being normalized by the diffusion coefficient of the overdamped Brownian motion (33).

Interestingly, the force-dependent contributions to $D$, whose fundamental origins can be traced back to the detailed balance condition (5), provide a way to control the diffusivity by applying external forcing upon the particle. This effect is a direct consequence of the microscopic reversibility of active propulsion mechanism. Therefore, it is missing in the ABP model and in the passive Brownian motion.

Testing such dependence in an experiment may provide a hint regarding microscopic reversibility of the underlying active propulsion mechanism. The zero-mean harmonic force seems to be well-suited for such purposes because it does not induce a net mean displacement of the particle.
The power-law dependence on frequency (inverse time scale) \( D \) with exponent 2 reflects the typical Brownian scaling of position with time. At low frequencies \( (\omega \to 0) \), corresponding to \( t \to \infty \) limit in the time domain, the whole expression (38) reduces to just such a power law,

\[
S_{rr}(\omega) \approx \frac{8D}{\omega^2}, \quad \omega \to 0, \tag{40}
\]

where the effective diffusion coefficient \( D \) is that from Eq. (32).

The \( \omega \)-dependent term in (38) consists of a combination of the Brownian power-law dependence on \( \omega \) and the Lorentzian function describing the rotational diffusion of the particle orientation. As the damping rate in the Lorentzian part of this term, there is the characteristic “frequency” (inverse time scale) \( D_r \) for this process to happen.

V. POWER SPECTRUM

The power spectrum of an ensemble of long trajectories, \(^85,86\)

\[
S_{rr}(\omega) = \lim_{\tau \to \infty} \frac{1}{\tau} \left\langle \left| \int_0^\tau \Delta r(t) e^{i\omega t} dt \right|^2 \right\rangle, \tag{35}
\]

bare a more-detailed information on the particle dynamics than MSD (28) does. Moreover, the power spectrum frequently attains a simple form, where individual terms characterize various diffusive mechanisms involved.

We will derive \( S_{rr}(\omega) \) by breaking it down into two parts: \( S_{rr}(\omega) = S_{xx}(\omega) + S_{yy}(\omega) \), with the marginal power spectrum of \( \Delta x(t) \) given by

\[
S_{xx}(\omega) = \lim_{\tau \to \infty} \frac{1}{\tau} \left\langle \left| \int_0^\tau \Delta x(t) e^{i\omega t} dt \right|^2 \right\rangle = \lim_{\tau \to \infty} \frac{1}{\tau} \int_0^\tau C_{xx}(t_1, t_2) e^{i\omega(t_1-t_2)} dt_1 dt_2,
\]

and \( S_{yy}(\omega) \) being defined similarly for \( \Delta y(t) \). To evaluate \( S_{xx}(\omega) \), we perform the double integration in the second line of Eq. (36) inserting there the exact expression (A.1) for \( C_{xx}(t_1, t_2) \). Carrying out a similar calculation for \( S_{yy}(\omega) \), we find the isotropy relation

\[
S_{xx}(\omega) = S_{yy}(\omega), \tag{37}
\]

analogous to Eqs. (31) for MSDs in the long-time limit. Eventually, we arrive at the final result

\[
S_{rr}(\omega) = \left( D + \frac{D_c}{2} \right) \frac{8}{\omega^2} + \frac{u^2}{D_r} \left( \frac{4}{\omega^2} - \frac{2}{D_r^2 + \omega^2} \right) + \frac{(\mu_c F_{dc})^2}{16D_r} \frac{4}{\omega^2} - \frac{2}{(4D_r)^2 + 2},
\]

in the time domain, the whole expression (38) reduces to just such a power law,

\[
S_{rr}(\omega) \approx \frac{8D}{\omega^2}, \quad \omega \to 0, \tag{40}
\]

where the effective diffusion coefficient \( D \) is that from Eq. (32).

The \( \omega \)-dependent term in (38) consists of a combination of the Brownian power-law dependence on \( \omega \) and the Lorentzian function describing the rotational diffusion of the particle orientation. As the damping rate in the Lorentzian part of this term, there is the characteristic “frequency” (inverse time scale) \( D_r \) for this process to happen.
At the nanoscale, its magnitude is rather negligible when the isotropy relation (37). Regardless the orientation of the force is transferred evenly to power spectra of both coordinates orientation changes as well. Interestingly, this correlation changes due to the rotational diffusion, the magnitude since $F_n$ of the product (37) can be roughly understood as the decay time of auto-correlations occurring in the previous term. The time $1/D_r$ shorter than that of the rotational diffusion (1/D_t) time scale (1) analogous form as the $S_{rr}^{(BM)}(\omega)$ for ABP model where the microscopic reversibility of active propulsion is neglected, see Eq. (41). The $F_{ac}$-dependent term in (38) causes the sigmoid-shaped changes of MRABP curves near $\omega \approx \Omega$ and $\omega \approx D_t$ observed at $F_{dc} = 0$. The $F_{dc}$-dependent term enhances the variation of the solid line near $\omega \approx D_t$, as compared to the $F_{dc} = 0$ case.

Such $u$-dependent term is also the only additional one that appears in the power spectrum of a harmonically driven ABP model:

$$S_{rr}^{(ABP)}(\omega) = \frac{8D}{\omega^2} + \frac{u^2}{D_c} \left( \frac{4}{\omega^2} - \frac{2}{D_t^2 + \omega^2} \right).$$

(41)

At the nanoscale, its magnitude is rather negligible when compared to the Brownian part $8D/\omega^2$, i.e., $S_{rr}^{(ABP)}(\omega) \approx S_{rr}^{(BM)}(\omega)$ holds. The reason for this is the large $D_r$ of small particles as discussed in details in the paragraph after Eq. (34).

The $F_{dc}$-dependent contribution in (38) has an analogous form as the $u$-dependent one. Yet, the characteristic time scale (1/4D_t) for the process it represents is 4 times shorter than that of the rotational diffusion (1/D_t) occurring in the previous term. The time 1/4D_t can be roughly understood as the decay time of auto-correlations of the product $n(t)F_n(r, t)$ in Eq. (6) for $u_n(t)$. The individual constituents of this product are already correlated since $F_n(r, t)$ depends on $n(t)$: As the particle orientation changes due to the rotational diffusion, the magnitude of the force projection onto the instantaneous particle orientation changes as well. Interestingly, this correlation is transferred evenly to power spectra of both coordinates regardless the orientation of the force $F(t)$ in space, viz the isotropy relation (37).

Behavior of the $F_{ac}$-dependent term in (38) with $\omega$ can be rather rich as compared to other terms. Figure 3 illustrates the power spectrum (38) of the microscopically reversible model (MRABP) and the one of a standard periodically driven ABP without the microscopic reversibility of the active propulsion, Eq. (41), for parameters estimated in Sec. II B and two values of $F_{dc}$. In Fig. 3, curves marking $S_{rr}^{(ABP)}(\omega)$ (solid and dashed-dotted line) indicate that the $F_{ac}$-dependent term is responsible for two sigmoid-shaped transitions: one close to the driving frequency $\omega \approx \Omega$ and one near $\omega \approx D_t$. The transitions are visible at $F_{dc} = 0$ and $F_{dc} = 10$ pN. As a result of nonzero $F_{dc}$, we observe an enhancement in the low-frequency part of the spectrum. For very high $\omega$, spectra (38) at $F_{dc} = 0$ and $F_{dc} \neq 0$ coincide. Both functions remain significantly larger than $S_{rr}^{(BM)}(\omega)$, Eq. (41), which is nearly identical with $S_{rr}^{(MRABP)}(\omega)$, Eq. (39), for all plotted $\omega$.

Moreover, if parameters $F_{ac}$, $D_r$, and $\Omega$ meet certain conditions, the sigmoid-shaped transition around the driving frequency $\omega \approx \Omega$ transforms into a peak, which we demonstrate in Fig. 4. To observe the peak, the $F_{ac}$-dependent term should attain significant values compared to other terms. This can be achieved either by increasing $F_{ac}$ or by a proper choice of $D_t$, see solid and dashed-dotted curves in Fig. 4 demonstrating the latter option. Also, $\Omega$ should be far enough from low frequencies, where $S_{rr}(\omega)$ is dominated by the Brownian power-law behavior (20). The width and the height of the peak can be controlled by the “damping” $4D_t$. If $4D_t$ is smaller than $\Omega$, a rather pronounced peak occurs (solid line). The

![Figure 3](https://example.com/fig3.png)

**FIG. 3.** Spectral power densities of active nanoparticles driven by the external time-periodic force with amplitude (1). Plotted spectra are normalized by $S_{rr}^{(BM)}(\omega)$, the spectrum of overdamped Brownian motion given in Eq. (39). Parameters used are those estimated in Sec. II B, $F_{ac} = 10$ pN, and $\Omega = 10^3$ s$^{-1}$. For the active particle with microscopically reversible propulsion (MRABP), we plot $S_{rr}(\omega)$ from Eq. (38) for $F_{dc} = 10$ pN (solid line) and $F_{dc} = 0$ (dashed-dotted line). The nearly constant dashed line represents the power spectrum $S_{rr}^{(ABP)}(\omega)$ for ABP model where the microscopic reversibility of active propulsion is neglected, see Eq. (41). The $F_{ac}$-dependent term in (38) causes the sigmoid-shaped changes of MRABP curves near $\omega \approx \Omega$ and $\omega \approx D_t$ observed at $F_{dc} = 0$. The $F_{dc}$-dependent term enhances the variation of the solid line near $\omega \approx D_t$, as compared to the $F_{dc} = 0$ case.

![Figure 4](https://example.com/fig4.png)

**FIG. 4.** Spectral power density (38) of an active particle with microscopically reversible propulsion subjected to the external time-periodic driving. The spectrum is plotted for three different values of the rotational diffusion coefficient $D_r$ and normalized by the power spectrum $S_{rr}^{(BM)}(\omega)$ of the overdamped Brownian motion given in Eq. (39). Other model parameters are chosen as in Sec. II B, $F_{dc} = 0$, $F_{ac} = 10$ pN, and $\Omega = 10^3$ s$^{-1}$. The peak in $S_{rr}(\omega)$ around $\omega \approx \Omega$ develops when $D_r < \Omega$. Its height can be controlled by $D_r$, which also changes the background spectral power at other frequencies $\omega$. The
condition \( \Omega > D_t \) means that the external force should oscillate faster than the rotational diffusion happens. Contrary, if the rotational diffusion is faster, the peak vanishes (dashed line in Fig. 4).

Let us note that \( F_{\text{dc}} \) - and \( F_{\text{ac}} \)-dependent terms in the power spectrum \( S_{rr}(\omega) \) given in Eq. (38) emerge due to the assumption of the microscopic reversibility of the active propulsion mechanism. Therefore, these terms are naturally missing in the ABP model, where MR is not taken into account. Accordingly, the derived spectrum \( S_{rr}^{(\text{ABP})}(\omega) \) in Eq. (41) for the periodically driven ABP model is identical with the one in case without the external driving force. \(^{86}\)

VI. SUMMARY AND PERSPECTIVES

The principle of microscopic reversibility is inherently related to the time reversal symmetry of microscopic dynamics. It must be enforced whenever the consistency of a studied nonequilibrium stochastic dynamics with the second law of thermodynamics is required. In case of Markov jump processes, the principle is incorporated by means of the local detailed balance condition obeyed by transition rates, while in continuous-space driven diffusions it is implemented via the fluctuation-dissipation relation. Both these ways occur in the modeling of the active propulsion mechanism in the present work.

As our main results, we have discussed prominent effects stemming from this fundamental principle in the case of a chemically driven active nanoparticle whose propulsion mechanism is made consistent with MR. When an external time-periodic force is acting upon such a particle, its effective diffusion coefficient and the spectral power density are significantly enhanced in comparison to the force-free dynamics and to the corresponding reference model not obeying MR.

The diffusion coefficient of such particle contains new contributions that increase its value beyond the one obtained for the reference active Brownian particle model without MR. The magnitude of the enhancement can be externally controlled by varying the amplitude of a constant part of the applied force, the amplitude of a time-periodic part of the force, and the frequency of the force oscillations.

All the new contributions to the diffusion coefficient have their counterparts in the power spectrum of stochastic trajectories of the nanoparticle. The spectrum also reflects time scales of underlying dynamic processes responsible for these contributions. As new qualitative effects caused by MR in the spectrum, we report sigmoid-shaped transitions and a sharp peak that can occur at the frequency of the external driving and at the characteristic time scale associated with rotational diffusion of the particle.

Overall, we have selected the external time-periodic driving force having in mind possible spectroscopic verifications of the reported hallmarks of the microscopic reversibility. To this end, we have also chosen values of the model parameters close to the ones of recently studied catalytic nanoparticles. We expect that the results presented here will motivate new experimental studies in line with our predictions, which in turn can shed a new light on the nature of chemically driven self-propulsion at the nanoscale. Moreover, our findings can provide new insights into high values of experimentally measured diffusion coefficients of small active particles, view of a potentially significant impact of local (constant and time-dependent) forces on the diffusivity and the power spectrum.

ACKNOWLEDGEMENTS

We acknowledge financial support from the Portuguese Foundation for Science and Technology (FCT) under Contracts nos. PTDC/FIS-MAC/5689/2020, UIDB/00618/2020, and UIDP/00618/2020. AR gratefully acknowledges financial support from the Czech Science Foundation (Project No. 20-02955J) and from the Department of Physics and Mathematics at Nottingham Trent University (grant no. 01/PHY/-/X1175). Computational resources were supplied by the project “e-Infrastruktura CZ” (e-INFRA CZ LM2018140) supported by the Ministry of Education, Youth and Sports of the Czech Republic.
Appendix: Exact expressions for correlation functions

The analytical expression for $C_{xx}(t_1, t_2)$, which has been given in Eq. (25) in terms of double integrals, reads

\[
C_{xx}(t_1, t_2) = 2 \left[ D + \frac{D_c}{2} + \frac{u^2}{2D_r} + \frac{(\mu_c F_{dc})^2}{32D_r} + \frac{(\mu_c F_{ac})^2}{4} \frac{D_r}{(4D_r)^2 + \Omega^2} \right] \min(t_1, t_2)
\]

\[
+ \frac{u^2}{2D_r^2} \left( e^{-D_r t_1} + e^{-D_r t_2} - e^{-D_r |t_2 - t_1|} - 1 \right) + \frac{(\mu_c F_{dc})^2}{128D_r^2} \left( e^{-4D_r t_1} + e^{-4D_r t_2} - e^{-4D_r |t_2 - t_1|} - 1 \right)
\]

\[
+ \left( \mu + \frac{\mu_c}{2} \right)^2 \frac{F_{ac}^2}{2\Omega^2} \left\{ \cos(\Omega(t_2 - t_1)) - \cos(\Omega t_1) - \cos(\Omega t_2) + 1 \right\}
\]

\[
+ \frac{(\mu_c F_{ac})^2}{16} \frac{\Omega^2 - (4D_r)^2}{[(4D_r)^2 + \Omega^2]^2} \left\{ 1 - e^{-4D_r |t_2 - t_1|} \left[ \cos(\Omega(t_2 - t_1)) - \frac{8D_r \Omega}{(4D_r)^2 - \Omega^2} \sin(\Omega t_2 - t_1) \right] \right\}
\]

\[
+ e^{-4D_r t_1} \left[ \cos(\Omega t_1) - \frac{8D_r \Omega}{(4D_r)^2 - \Omega^2} \sin(\Omega t_1) \right] + e^{-4D_r t_2} \left[ \cos(\Omega t_2) - \frac{8D_r \Omega}{(4D_r)^2 - \Omega^2} \sin(\Omega t_2) \right] \right\},
\] (A.1)

where, on the first line, the expression enclosed in squared brackets is nothing but the effective diffusion coefficient $D$, Eq. (32). Similarly, we evaluate $C_{yy}(t_1, t_2)$, related to $C_{xx}(t_1, t_2)$ in Eq. (26) and get

\[
C_{yy}(t_1, t_2) = C_{xx}(t_1, t_2) - \left( \mu + \frac{\mu_c}{2} \right)^2 \frac{F_{ac}^2}{2\Omega^2} \left\{ \cos(\Omega(t_2 - t_1)) - \cos(\Omega t_1) - \cos(\Omega t_2) + 1 \right\}.
\] (A.2)

That is, $C_{yy}(t_1, t_2)$ is given by Eq. (A.1) after removing all terms displayed on the third line of (A.1).
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