

1 **Loopy Lévy flights enhance tracer diffusion in active suspensions**

2 Kiyoshi Kanazawa*

3 *Faculty of Engineering, Information and Systems,*
4 *University of Tsukuba, Tennodai, Tsukuba, Ibaraki 305-8577, Japan*

5 Tomohiko G. Sano

6 *Flexible Structures Laboratory, Institute of Mechanical Engineering,*
7 *École polytechnique fédérale de Lausanne, Lausanne, CH-1015, Switzerland.*

8 Andrea Cairoli

9 *Department of Bioengineering, Imperial College London, London SW7 2AZ, UK*

10 Adrian Baule

11 *School of Mathematical Sciences, Queen Mary University of London, London E1 4NS, UK*

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* kiyoshi@sk.tsukuba.ac.jp

13 Brownian motion is widely used as a paradigmatic model of diffusion in equi-
14 librium media throughout the physical, chemical, and biological sciences. How-
15 ever, many real world systems are intrinsically out-of-equilibrium due to the
16 energy-dissipating active processes underlying their mechanical and dynamical
17 features [1]. The diffusion process followed by a passive tracer in prototypical
18 active media such as suspensions of active colloids or swimming microorgan-
19 isms [2] indeed differs significantly from Brownian motion, manifest in a greatly
20 enhanced diffusion coefficient [3–10] and non-Gaussian statistics of the tracer dis-
21 placements [6, 9, 10]. While such characteristic features have been extensively
22 observed in experiments, there is so far no comprehensive theory explaining
23 how they emerge from the microscopic dynamics. Here we develop a theoretical
24 framework to coarse-grain the hydrodynamic interactions between the tracer
25 and the active swimmers, which shows that the tracer follows a non-Markovian
26 coloured Poisson process accounting for all empirical observations. The theory
27 predicts in particular a long-lived Lévy flight regime [11] of the tracer motion
28 with a non-monotonic crossover between two different power-law exponents. The
29 duration of this regime can be tuned by the swimmer density, thus suggesting
30 that the optimal foraging strategy of swimming microorganisms might crucially
31 depend on the density in order to exploit the Lévy flights of nutrients [12]. Our
32 framework can be applied to address important conceptual questions, such as
33 the thermodynamics of active systems [13], and practical ones, such as the in-
34 teraction of swimming microorganisms with nutrients and other small particles
35 like degraded plastic [14] and the design of artificial nanoscale machines [15].

36 A passive tracer immersed in a fluid at equilibrium moves randomly due to its collisions
37 with the surrounding fluid molecules. Understanding how the stochastic process of the tracer
38 relates to the statistical mechanics of the surrounding fluid, as accomplished in the seminal
39 works by Einstein, Smoluchowski, and Langevin [16], has provided deep insight into the
40 connection between molecular transport and equilibrium thermodynamics, which has been
41 widely exploited to describe soft matter and other complex physical systems [17]. However,
42 when either artificial self-propelled colloids or biological swimming micro-organisms, such as
43 bacteria like *Escherichia coli* or algae like *Volvox* and *Chlamydomonas reinhardtii* [2], are also
44 suspended, the diffusion of the tracer changes dramatically due to the active stirring of the

45 fluid exerted by the swimmers. Indeed, the tracer exhibits empirically the following unique
 46 features that can no longer be explained as a Brownian motion: (i) the tracer exhibits loopy
 47 trajectories [4, 8, 9]; (ii) its mean square displacement (MSD) exhibits a crossover between
 48 superdiffusion with characteristic scaling t^α ($1 < \alpha \leq 2$) for short times and normal diffusion
 49 ($\alpha = 1$) for long times, where the effective diffusion coefficient D_e is greatly enhanced
 50 compared with the equilibrium coefficient D_0 : $D_e = D_0 + B\rho v_A$, with the density of swimmers
 51 ρ , characteristic swimming velocity v_A , and a system-dependent parameter B [3–5, 7–9];
 52 (iii) the probability density function (PDF) $P_{\Delta t}$ of tracer displacements in a given time
 53 interval Δt exhibits strong non-Gaussian features manifest as power-law tails [6, 10]; (iv) $P_{\Delta t}$
 54 eventually reverts to a Gaussian shape for large Δt [9, 10]; (v) the associated non-Gaussian
 55 parameter (NGP) exhibits a characteristic scaling Δt^{-1} for large times [10].

56 Developing a single theory that captures all features (i)–(v) has been a major challenge.
 57 While the loop-like motion (i) results from an individual scattering event of the tracer in
 58 the dipolar flow field of a single swimmer [7, 18, 19], and the linear form of D_e (ii) has been
 59 explained phenomenologically based on the active flux ρv_A of the swimmers [5, 7, 8, 19, 20],
 60 the statistical observations (iii)–(v) could so far not be explained consistently. The power-
 61 law tails in $P_{\Delta t}$ and their convergence to Gaussian scaling for large Δt , which is expected
 62 based on central limit theorem (CLT) arguments [21, 22], have been reproduced in [10, 23, 24]
 63 assuming a static force distribution akin to the Holtsmark theory of gravitating particles [25]
 64 (see Supplementary Information (SI) Sec. VI C for a review). However, this approach
 65 neglects any dynamics of the swimmers and is thus not sufficient to capture the enhanced
 66 diffusion observed in experiments [3–10]. Here we present a derivation of the stochastic
 67 process underlying the tracer diffusion from the microscopic hydrodynamic interactions. The
 68 resulting process is valid at all timescales and captures all characteristic features (i)–(v).

69 We consider a three-dimensional system composed of m active particles (swimmers) and
 70 a passive tracer suspended in a viscous fluid inside a cubic box (Fig. 1a). The swimmers are
 71 assumed self-propelled moving unidirectionally with constant velocity v_A [26]. In general,
 72 the dynamics of such a multi-particle system with long-range hydrodynamic interactions
 73 is analytically intractable. However, suspensions of micro-organisms often considered in
 74 experiments are characterized by (a) low Reynolds number swimming and (b) a low density
 75 of swimmers (dilute condition, see below). In particular the dilute condition (b) allows
 76 us to neglect the mutual hydrodynamic interactions of the swimmers [27], leading to the

77 overdamped equations of motion:

$$\frac{d\mathbf{x}_i}{dt} = v_A \mathbf{n}_i, \quad \Gamma \frac{d\mathbf{X}}{dt} = \sum_{i=1}^m \mathbf{F}(\mathbf{x}_i - \mathbf{X}, \mathbf{n}_i), \quad (1)$$

78 where \mathbf{F} is the force on the tracer generated by a single swimmer and Γ is a viscous coefficient
 79 for the passive particle. In Eq. (1), $\mathbf{x}_i(t)$ and $\mathbf{X}(t)$ denote the positions of the i -th active
 80 particle and the passive particle, respectively, and the unit vector \mathbf{n}_i specifies the swimming
 81 direction.

82 The low Reynolds number condition (a) further yields a closed form expression for \mathbf{F} as
 83 the solution of the Stokes equation. For force- and torque-free swimmers, the leading-order
 84 term in a far-field expansion is a dipole force [26]

$$\mathbf{F}(\mathbf{r}_i, \mathbf{n}_i) \approx \frac{p}{r_i^2} \left[3 \frac{(\mathbf{n}_i \cdot \mathbf{r}_i)^2}{r_i^2} - 1 \right] \frac{\mathbf{r}_i}{r_i}, \quad r_i > d \quad (2)$$

85 with the difference vector $\mathbf{r}_i = \mathbf{x}_i - \mathbf{X}$ and the system specific cut-off d , which separates
 86 the far-flow field from any near-flow field hydrodynamic contributions and hard-core in-
 87 teractions. The dipole strength parameter p specifies the universal features of the far-flow
 88 hydrodynamic field [26]: $p < 0$ denotes *pusher* swimmers whose flow lines are oriented out-
 89 ward along the direction of its velocity vector and inward laterally (e.g., *E. coli* [26]); $p > 0$
 90 denotes instead *puller* swimmers whose flow lines are oriented in the opposite directions (e.g.,
 91 *C. reinhardtii* [26]). For $r_i \leq d$ the interaction force \mathbf{F} is not universal but system-specific.
 92 Nevertheless, all swimmer-tracer interactions in this regime can be accurately captured us-
 93 ing arguments based on the CLT, which do not require a detailed form of \mathbf{F} . Therefore, we
 94 set for simplicity $\mathbf{F} = 0$ in this regime [10]. The model has one further length parameter
 95 $b^* \equiv \sqrt{|p|/(\Gamma v_A)}$, which can be related to the typical length scale of the swimmers [26]. All
 96 these parameters can be determined experimentally: for *E. coli* $v_A \simeq 30 \mu\text{m/s}$, $d \simeq 6 \mu\text{m}$ and
 97 $b^* \simeq 2 \mu\text{m}$ [5, 27]; for *C. reinhardtii* $v_A \simeq 100 \mu\text{m/s}$, $d \simeq 35 \mu\text{m}$ and $b^* \simeq 8 \mu\text{m}$ [10, 23, 24, 28];
 98 both are satisfying $d \geq b^*$.

99 Simulations of Eqs. (1,2) under the dilute condition reproduce all features (i)–(v) (see
 100 Fig. 1b, Fig. 3, and Supplementary Video). Crucially, the time-series of the force exerted on
 101 the tracer highlights that its dynamics can be resolved as a sequence of individual scattering
 102 events, where only two-body tracer-swimmer interactions are relevant (Fig. 2a). In the
 103 kinetic theory of gases, a similar description in terms of binary scattering events is ensured
 104 by requiring $\rho r_c^3 \ll 1$, where r_c denotes the range of the interparticle interaction force. Even

105 though for long-range hydrodynamic forces an interaction range cannot be well defined,
 106 we identify in our system r_c as the maximum of the characteristic length scales available:
 107 $r_c \equiv \max\{d, b^*\} = d$. The motivation behind this definition of r_c is two-fold. Firstly, the
 108 dilute condition $\rho d^3 \ll 1$ is indeed realized in experiments that exhibit the features (i)–(v)
 109 (e.g., $\rho d^3 \approx 0.15$ in [10] and $\rho d^3 \approx 0.10$ in [5]). Secondly, this parameter regime allows for
 110 a self-consistent description of the tracer dynamics, in which the dipole interaction governs
 111 the displacement statistics on short and intermediate time scales, while the statistics for
 112 longer times reverts to a Gaussian due to the CLT, as follows.

113 In a dilute system defined this way, at every instant in time, swimmers on average have
 114 $b \gg d$, where b is the impact parameter of a binary swimmer–tracer interaction (see Fig. 1c).
 115 The tracer statistics will then be governed by three distinct dynamical regimes: (1.) For
 116 short times $\Delta t \ll \tau_H \equiv b^*/v_A$, with τ_H the timescale below which the motion of the swimmers
 117 is effectively negligible, the tracer experiences static long-range forces as in the Holtmark
 118 theory [25] (“Holtmark regime”). (2.) For times $\tau_H \ll \Delta t \ll \tau_C$, the tracer is displaced by
 119 the moving swimmers in a sequence of binary “scatterings” governed by the far-flow field
 120 interaction force (2) (“scattering regime”). The time scale $\tau_C \equiv 1/(\rho v_A \pi d^2)$ (see Fig. 1c)
 121 thus estimates the time necessary for a swimmer to come close enough to the tracer to
 122 interact via hard-core and near-field hydrodynamic interactions. (3.) For $\Delta t \gg \tau_C$, the
 123 tracer is displaced by an accumulation of these “collisions” with the swimmers such that the
 124 CLT applies (“CLT regime”).

125 Remarkably, these three regimes are captured by a coarse-grained description of the tracer
 126 dynamics in terms of the Langevin equation

$$\Gamma \frac{d\mathbf{X}}{dt} = \sum_{i=1}^{N(t)} \mathbf{f}_{\mathbf{b}_i}(t - \tau_i), \quad (3)$$

127 where $N(t)$ counts the number of scattering events up to t and $\mathbf{f}_{\mathbf{b}}(t)$ is the force shape func-
 128 tion (FSF) describing the force exerted on the tracer during each scattering. The FSF is
 129 centred at the scattering time point τ_i , which is set by the condition $\mathbf{n}_i \cdot \mathbf{r}_i = 0$ for the i -th
 130 swimmer. The transition from the fully deterministic dynamics of Eqs. (1, 2) to a stochastic
 131 description by Eq. (3) is realised by assuming $N(t)$ to be a Poisson process with intensity
 132 $\lambda(\mathbf{b})$, whose functional form can be determined from the microscopic dynamics assuming
 133 a uniform distribution of swimmers in the box and an isotropic distribution of swimming
 134 directions (see SI Sec. II A). The FSF $\mathbf{f}_{\mathbf{b}}(t)$ is characterized by the impact parameter and

135 injection angles (combined as \mathbf{b} ; see Fig. 2b) and is obtained in analytical form by solv-
 136 ing the binary swimmer–tracer scattering problem. Remarkably, for $b \gg b^*$ an analytical
 137 approximation of the FSF can be obtained using a Picard iteration up to 2nd order and
 138 subsequent Taylor expansion (see SI Sec. II B–D), which is in very good agreement with
 139 numerics (Fig. 2c). The total intensity of the process diverges in the large system size limit
 140 because $\lambda(\mathbf{b})d\mathbf{b} \propto b$ such that $\lim_{L \rightarrow \infty} \int d\mathbf{b} \lambda(\mathbf{b}) = \infty$. Consequently, $\mathbf{X}(t)$ is a coloured
 141 Poisson process with infinite intensity, a generalization of the Campbell process [29]. Rather
 142 than being unphysical, this infinite intensity is a physical consequence of the long-range hy-
 143 drodynamic interactions, which cause an infinite number of small scatterings at arbitrarily
 144 large distances.

145 From Eq. (3) we calculate the statistics of the tracer displacement in the x -direction
 146 $\Delta X \equiv \mathbf{e}_x \cdot [\mathbf{X}(t + \Delta t) - \mathbf{X}(t)]$, which is typically measured in experiments [4, 6, 9, 10].
 147 Using functional techniques (see SI Sec. III), we derive the displacement PDF $P_{\Delta t}(|\Delta X|)$
 148 and obtain its scaling behaviours in the regimes (1.)–(3.) as

$$P_{\Delta t}(|\Delta X|) \propto \begin{cases} |\Delta X|^{-\alpha_H} & (\Delta t \ll \tau_H) \\ |\Delta X|^{-\alpha_S} & (\tau_H \ll \Delta t \ll \tau_C) \\ e^{-\Delta X^2/(2\sigma^2)} & (\tau_C \ll \Delta t) \end{cases} \quad (4)$$

149 with power-law exponents $\alpha_H = 5/2$, $\alpha_S = 5/3$ and a positive constant σ that depends on
 150 d . For finite d a truncation appears in the power-law tails, which is realistic because any
 151 physical system must accommodate finite cutoffs [30].

152 The Holtmark regime (1.) yields the same scaling behaviour $|\Delta X|^{-\alpha_H}$ found in the
 153 approaches [10, 23, 24] considering static swimmers. For $\Delta t \gtrsim \tau_H$, these theories are not
 154 applicable because the rearrangements of the active swimmers are no longer negligible. Tak-
 155 ing into account the scatterings from moving swimmers, the process (3) is non-Markovian.
 156 Nevertheless, in the scattering regime (2.), the description becomes effectively Markovian
 157 as the FSF can be approximated as a Dirac δ -function (see SI Sec. II). In this regime the
 158 coloured Poisson model (3) is equivalent to a compound Poisson process with jump-length
 159 distribution prescribed as a power law $\propto |\Delta X|^{-\alpha_S}$ with a finite cutoff. Our results thus
 160 suggest that the celebrated Lévy flight model [11] holds as an approximate description of
 161 the tracer dynamics at this timescale. The Lévy flight is a generalization of the standard
 162 random walk model to power-law distributed jump sizes, which can capture long jumps in a

163 single time-step (see SI Sec. IV). If confirmed experimentally, our framework would provide
 164 the first validation of this model from a physical microscopic dynamics. In the collision
 165 regime (3.), collisions become dominant over scatterings. Since collisional impact has a fi-
 166 nite cutoff, the accumulation of a sufficient number of collisions leads to the Gaussian tail
 167 as a consequence of the CLT. We note that the detailed form of \mathbf{F} in Eq. (1) for $r_i \leq d$ is
 168 renormalized into the variance σ^2 .

169 Using the exact expression for $P_{\Delta t}$ (see SI, Eq. (S40)), the tracer MSD for different
 170 swimmer densities is predicted to collapse onto a universal curve upon rescaling by $1/(\rho b^{*5})$,
 171 which reveals a crossover from ballistic motion $\propto \rho v_A^2 \Delta t^2$ at short timescales to normal
 172 diffusion $\propto \rho v_A \Delta t$ at longer ones (see SI Sec. III A). Likewise, the NGP is predicted to
 173 exhibit the scaling $\text{NGP}(\Delta t) \propto (\rho v_A)^{-1} \Delta t^{-1}$ upon rescaling by ρb^{*3} (see SI Sec. III B).
 174 Eq. (3) thus not only predicts the enhanced diffusion at all time scales but also captures
 175 quantitatively the linear dependence of the diffusion coefficient D_e on the active flux ρv_A
 176 as well as the scaling behaviour of the NGP, consistent with experimental observations [3–
 177 5, 7–10]. The striking non-monotonic behaviour of the scaling exponents of $P_{\Delta t}$ as predicted
 178 by Eq. (4), as well as the data collapse and the scalings for both the MSD and NGP are
 179 confirmed in simulations (see Fig. 3).

180 Our theory can be extended to incorporate thermal fluctuations and more general hy-
 181 drodynamic forces in arbitrary dimensions. Under thermal noise, the scaling predictions
 182 Eq. (4) and that of the NGP are preserved, while the MSD is shifted by a term $2D_0\Delta t$, with
 183 $D_0 = k_B T/\Gamma$ and T the temperature of the bath (see SI Secs. III A and B). Interestingly,
 184 the data collapse for both the MSD and the NGP no longer holds in this case. Therefore,
 185 the presence of data collapse can be probed as a footprint of the relevance of thermal fluc-
 186 tuations on the tracer dynamics. Moreover, our analysis shows that while α_H is universally
 187 determined by the leading power-law exponent of the hydrodynamic interaction force, n_H ,
 188 as $\alpha_H = 1 + D/n_H$ in D dimensions, the exponent α_S depends crucially on the details of
 189 the force law (see SI Sec. III C). We find $\alpha_S = 1 + (D - 1)/n_S$, where the exponent n_S is
 190 related to the net tracer displacement during a scattering event and is thus determined by
 191 the geometry of particle loops.

192 Over several magnitudes in time, the tracer is characterized by a linear-in-time MSD with
 193 non-Gaussian displacement statistics, an example of “Brownian yet non-Gaussian” diffusion
 194 widely observed in non-equilibrium systems [31]. While this behaviour poses considerable

195 challenges already for purely phenomenological modelling approaches [32], it is here fully
196 captured by the process (3) that is derived from the microscopic dynamics. Moreover,
197 our results pose interesting biological questions on the possible foraging behaviour of real
198 swimming micro-organisms like *C. reinhardtii*. The power-law tail in Eq. (4) persists for
199 longer timescales when the swimmer density ρ decreases, because $\tau_C \propto 1/\rho$. Since Lévy
200 flights increase encounter probabilities in stochastic search [12, 33, 34], our results suggest
201 that the foraging strategy of swimming micro-organisms might depend on ρ : for large ρ
202 the Gaussian (spatially localized) displacements of nutrients make an active search like
203 intermittent Lévy-Brownian strategies [35] suitable. Conversely, at low ρ the nutrients
204 exhibit large power-law distributed displacements, such that it might be advantageous for
205 the forager to simply wait for a nutrient to come close. Finally, we remark that superimposing
206 additional force fields such as electric or magnetic forces can be treated within our framework
207 in full analogy to hydrodynamic interactions, which would allow the theoretical investigation
208 of mechanisms to control and exploit enhanced diffusion.

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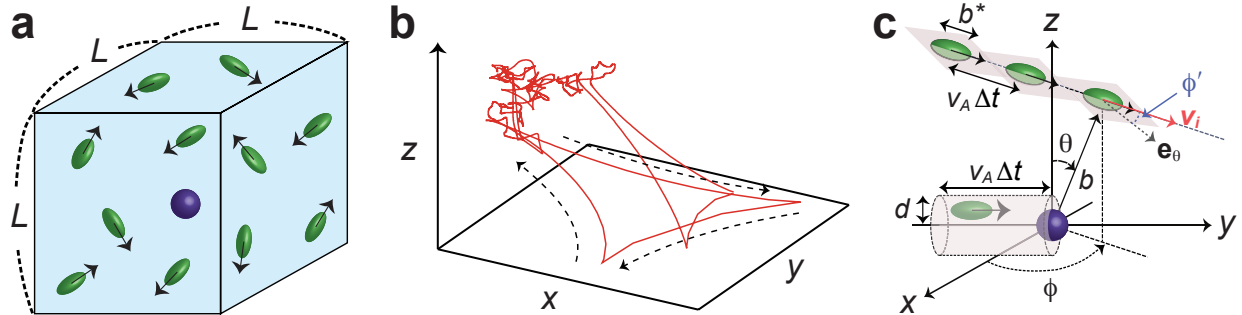


FIG. 1. **Illustration of the microscopic model.** (a) Active swimmers (green ellipsoids) and a passive tracer (violet sphere) are suspended in a cubic box of edge length L . The direction of the swimmers $\mathbf{n}_i \equiv (\sin \theta_i \cos \phi_i, \sin \theta_i \sin \phi_i, \cos \theta_i)$ is randomized upon hitting the box boundary. We consider a sufficiently large box $L \rightarrow \infty$ under the dilute condition $\rho d^3 \ll 1$, where $\rho \equiv m/L^3$, and set length and time units as $b^* = 1$ and $\tau_H \equiv b^*/v_A = 1$ (see SI Sec. I for protocol details). (b) Typical simulated tracer trajectory exhibiting characteristic non-closed loops for pusher swimmers ($p < 0$, $d = b^*$). (c) Two exemplary swimmer–tracer interaction events. Top: a two-body scattering event with impact parameter $b > d$ and injection angles θ , ϕ , and ϕ' , with the unit vector \mathbf{e}_θ . Bottom: a collision event with $b \leq d$. A swimmer travels the distance $v_A \Delta t$ in a time interval Δt . We define the time at which this distance is equal to the characteristic length scale b^* as τ_H . Since b^* is related to the typical length scale of a swimmer, for $\Delta t \ll \tau_H$ the motion of the swimmers can be effectively neglected (see SI, Fig. S3). Conversely, for a given set of injection angles swimmers can collide against the tracer in the time interval Δt if they are contained in a cylinder with cross section area πd^2 and length $v_A \Delta t$ surrounding the tracer. The mean free time of the tracer τ_C is then estimated as $\rho \pi d^2 v_A \tau_C = 1$, *i.e.*, $\tau_C \equiv 1/(\rho v_A \pi d^2)$. For $\Delta t \ll \tau_C$ such collision events do not contribute to the tracer dynamics.

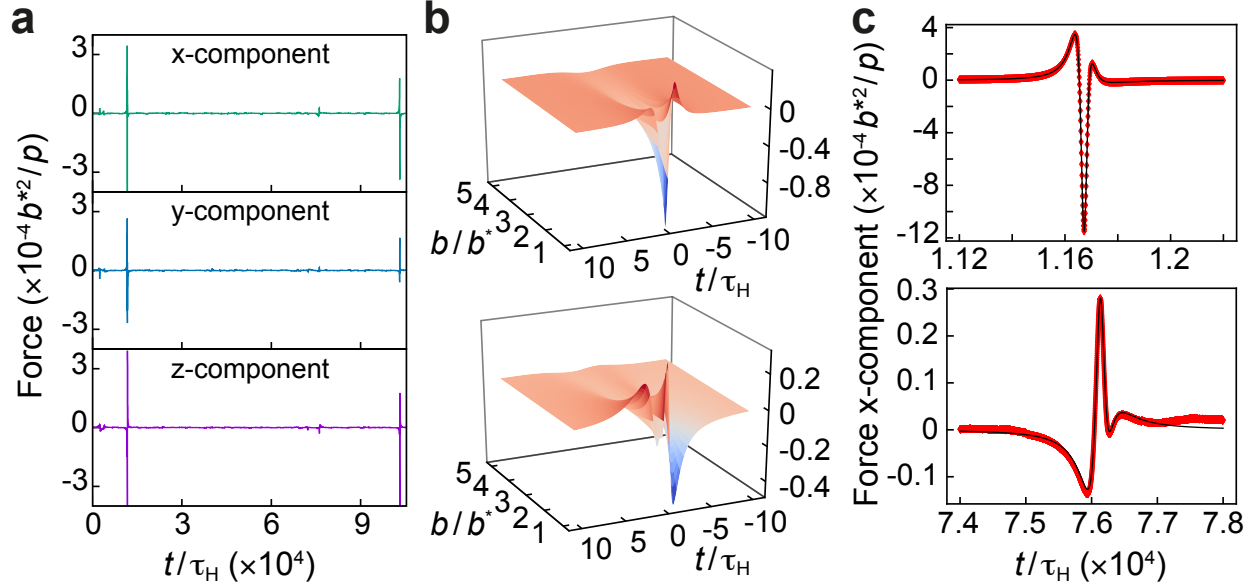


FIG. 2. **The force exerted on the tracer by the swimmers.** (a) Typical time series in the simulation ($p < 0$, $d = b^*$). (b) 3D plots of the projected FSF $f_{\mathbf{b}}(t) \equiv \mathbf{e}_x \cdot \mathbf{f}_{\mathbf{b}}(t)$, which is characterized by the parameter set $\mathbf{b} \equiv (b, \theta, \phi, \phi')$. We consider $\{\theta, \phi, \phi'\} = \{\pi/2, 0, \pi/2\}$ (top panel) and $\{\theta, \phi, \phi'\} = \{0, 0, 0\}$ (bottom panel). These two special cases form a base representation of the general solution for $f_{\mathbf{b}}(t)$ up to 2nd order (see SI Sec. II D). (c) Exemplary fits (black solid line) of $f_{\mathbf{b}}(t)$ to simulation data of two scattering events arbitrarily extracted from the force time-series (a). Fit parameters are \mathbf{b} and τ , which are obtained using nonlinear least squares (see SI Sec. II D).

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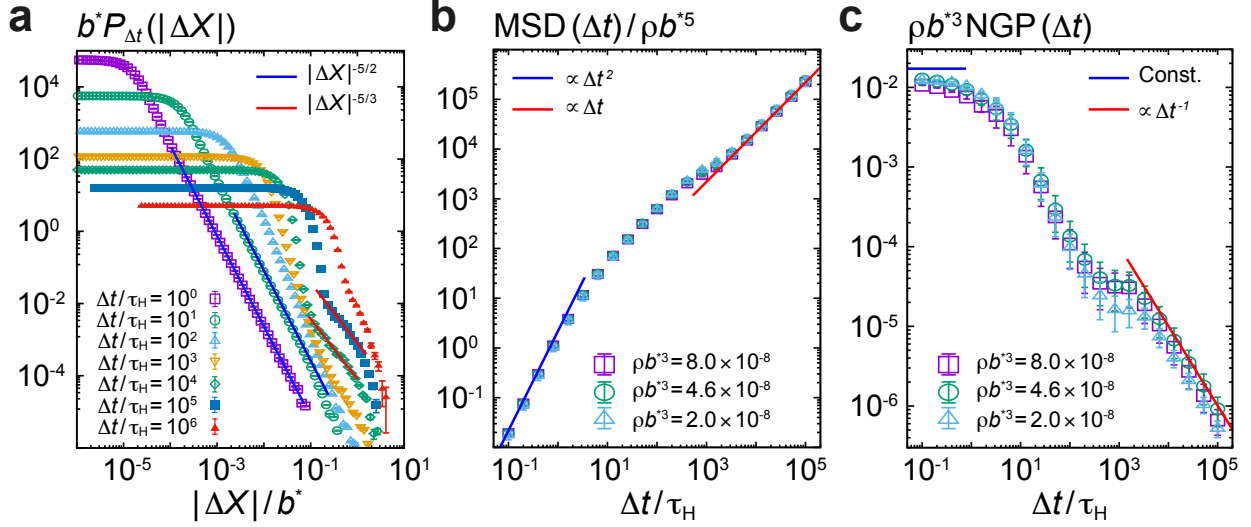


FIG. 3. **Simulation results confirming the theoretical predictions.** (a) The PDF $P_{\Delta t}(|\Delta X|)$ of the tracer displacements ΔX exhibiting power-law tails and the reversal to a Gaussian for large Δt . The crossover scaling from $\alpha_H = 5/2$ to $\alpha_S = 5/3$ should be experimentally testable (see SI Sec. V). (b) Data collapse of $\text{MSD}(\Delta t) \equiv \langle \Delta X^2 \rangle$ upon rescaling by $1/(\rho b^{*5})$ exhibiting a crossover between ballistic and normal diffusive motion [3, 4, 6, 7, 10]. (c) Data collapse of $\text{NGP}(\Delta t) \equiv \langle \Delta X^4 \rangle / (3\langle \Delta X^2 \rangle^2) - 1$ upon rescaling by ρb^{*3} exhibiting a power-law decay Δt^{-1} for large Δt [10]. Error bars denote ± 1 s.e.m..

292 AUTHOR CONTRIBUTIONS

293 KK, TGS, AC, and AB designed the research and interpreted results. KK and AC
 294 performed analytical calculations. TGS performed numerical simulations. AB managed the
 295 project. KK, AC, and AB wrote the paper.

296 COMPETING INTERESTS

297 The authors declare no competing financial interests.

298 **DATA AND CODE AVAILABILITY**

299 Simulation data were generated at XC40 at Yukawa Institute for Theoretical Physics in
300 Kyoto University. Derived data supporting the findings of this study are reproducible by the
301 simulation code, available at <https://doi.org/10.5281/zenodo.3550834>. The time series data
302 used in Fig. 1b and Fig. 2b are accessible as a text file at <https://doi.org/10.5281/zenodo.3550838>.
303 The further data will be available from Tomohiko G. Sano upon request.