Maximum in density heterogeneities of active swimmers

This item was submitted to Loughborough University's Institutional Repository by the/an author.


Additional Information:

- This is an Open Access Article. It is published by the Royal society of Chemistry under the Creative Commons Attribution 3.0 Unported licence (CC BY 3.0). Full details of this licence are available at:https://creativecommons.org/licenses/by/3.0/

Metadata Record: https://dspace.lboro.ac.uk/2134/35123

Version: Published

Publisher: © The Royal Society of Chemistry

Rights: This work is made available according to the conditions of the Creative Commons Attribution 3.0 Unported (CC BY 3.0) licence. Full details of this licence are available at: http://creativecommons.org/licenses/by/3.0/

Please cite the published version.
Fabian Jan Schwarzendahl and Marco G. Mazza

Maximum in density heterogeneities of active swimmers
I. Introduction

Physical interactions in suspensions of microswimmers consisting of bacteria or algae have been recognized to play an important role in the swimmers’ collective behavior.1–3 The nonequilibrium character of active suspensions, where the energy injection takes place at the scale of the microorganisms, produces myriad instabilities. We find a maximum degree of clustering at intermediate filling fractions and at large Péclet numbers resulting from a competition of hydrodynamic and steric interactions between the swimmers. We develop an analytical theory that supports these results. This maximum might represent an optimum for the microorganisms’ colonization of their environment.

Thus, the state of motion is only determined by the forces acting at that very moment, and inertia is negligible.

Due to the microswimmers’ low Reynolds numbers, the sum of viscous drag and thrust balances out to zero, in most situations. A direct consequence of force-free motion is that the leading term of the solution of the Stokes equation for a microswimmer is a symmetric force dipole (or stresslet).

Biological microswimmers are complex systems because of the combination of biological, biochemical and physical processes all taking place at the same time. It is thus of great scientific value to develop theoretical models that isolate the relevant degrees of freedom and interactions. Considerable work has been done in recent years, and various models have been introduced, like the squirmer model,13–31 the shape anisotropic raspberry swimmer,32–34 the force–counterforce model,35–38 the catalytic dimers,39 or other hydrodynamic models.40–43 Experiments have confirmed that the flow field of flagellated bacteria like *E. coli* is to very good approximation modeled by a simple force dipole,44 whereas *Chlamydomonas reinhardtii* are modeled by three Stokeslets.45 Furthermore, as cell shapes vary greatly in the natural world, and realistic steric interactions are important in dense suspensions, a model that allows for flexibility in the shape of a microswimmer is a highly desirable feature. In this article we fill this lacuna. We derive a model for a flexible-shape microswimmer that produces self-propulsion by means of a force dipole for pusher-like microswimmers, or three Stokeslets for puller-like microswimmers.

An efficient method to simulate fluids at mesoscopic scales, and their hydrodynamics is the multiparticle collision dynamics (MPCD) technique.46 MPCD is a particle-based simulation method that allows for a wide range of fluid behaviors and can be used to model a variety of biological systems.
that correctly produces hydrodynamic modes. Due to its particle nature MPCD naturally includes thermal fluctuations, and can be easily coupled to molecular dynamics methods of solutes, colloids, and active swimmers. The MPCD technique in fact proves to be ideal for our purposes.

The nonequilibrium phase diagram of microswimmers has been subject to considerable interest, especially with regard to the emergence of density heterogeneities in the swimmers’ distribution. We explore the phase diagram of active swimmers and describe the presence of heterogeneities in the spatial distribution of both pushers and pullers. These heterogeneities arise due to the hydrodynamic interactions between the swimmers and relate to existing hydrodynamic theories. Interestingly, we find a maximum in the heterogeneities as filling fraction and Péclet number are varied. By using both computer simulations and analytical theory, we demonstrate that this maximum results from a competition between hydrodynamic and steric interactions, where the latter temper the hydrodynamic instability at higher filling fractions. This optimum might have important biological implications on the ability of motile bacteria and algae to form colonies or biofilms.

The remainder of this article is organized as follows. In Section II we introduce the model for the microswimmer. Section III describes the physical properties of the fluid and the microswimmer’s flow field. In Section IV we present the nonequilibrium phase diagram of our microswimmers, and specifically we characterize the density heterogeneities emerging from their hydrodynamic interactions and show that these are tempered by steric interactions at higher filling fractions. In Section V we present an analytical theory and show that we also find a maximum in heterogeneity, which is mediated by the interplay of hydrodynamic and steric interactions. Finally, in Section VI we discuss our main results and summarize our conclusions.

II. Model

We employ a stroke-averaged model of biological microswimmers, similarly to, taking into account the asymmetric shape of biological microswimmers due to the cell’s body and the flagella. The swimmer is thus modeled as an asymmetric dumbbell, as depicted in Fig. 1, that mimics a C. reinhardtii or an E. coli cell. The smaller sphere models the swimmer’s body and the larger sphere is a stroke average of the region spanned by flagellar motion. The rigid body dynamics of the dumbbells is simulated using Newton’s equations and quaternion dynamics (for details see Appendix A1 and A2).

In addition to the rigid body dynamics, we simulate the fluid surrounding the swimmers using MPCD (see Appendix A3), which is a mesoscopic, particle based method, that reproduces hydrodynamics at the Navier–Stokes level. Precise measurements show that the flow field of pusher-type microswimmers is well modeled by a force dipole. The pullers flow field, on the other hand, is well represented by a three-Stokeslet solution of the Stokes equation. These two flow-field models are included in our simulations by adding force regions to the fluid, as depicted by the red regions (3) in Fig. 1. Furthermore, we couple the fluid and swimmers’ dynamics by imposing a no-slip boundary condition on the swimmers’ surface (for details see Appendix A4).

It should be noted that the dumbbell model introduced here has an anisotropic rigid shape that can be easily modified to more complex shapes. Squirmers have so far been described for spherical or ellipsoidal shapes. Our hydrodynamic flow field is very similar to the three-bead-spring model. However, the rigid body dynamics differ in the fact that in the three-bead-spring model the beads are connected by springs, whereas ours is a rigid model.

In the following, we express all physical quantities in terms of the MPCD particle mass \( m \), the size of an MPCD grid cell \( a \), and the temperature \( T \) of the fluid. We simulate (see Section IV) \( N = 300–1560 \) active swimmers in a cubic domain with side length of \( 100a \), which is approximately 20 times the size of an individual swimmer, and with periodic boundary conditions. The resulting filling fraction ranges between \( \nu = 0.05 \) and \( \phi = 0.29 \). The average number of MPCD particles per cell is \( \langle N_C \rangle = 20 \) such that the total number of MPCD particles in a simulation is \( 2 \times 10^7 \). The Reynolds number considered here ranges from \( \mathcal{R} \sim 0.01 \) to \( \mathcal{R} \sim 0.1 \) and the Péclet number reaches from \( 2.2 \times 10^2 \) to \( 2.6 \times 10^4 \).

The full details of our numerical implementation are given in Appendix A.
III. Characterization of the fluid and active hydrodynamics

In the following we describe calculations aimed at characterizing the thermal (equilibrium) properties of our model in the passive case, and also the flow field generated by the active motion.

We first consider a passive colloid (with the same geometry described above, and in Fig. 1) immersed in the MPCD fluid, that is, we carried out equilibrium simulations without activity. The equipartition theorem applied to the passive colloid for the translational motion predicts that the average of each velocity component squared is \( \langle U^2 \rangle = k_B T M \), where \( M \) is the colloid’s mass, \( k_B \) is the Boltzmann constant, and \( x = x, y, z \). For our system, we find a theoretical value of the translational motion \( \langle U_{\text{theory}}^2 \rangle = 2.7 \times 10^{-4} k_B T m \) and the simulations give \( \langle U_x^2 \rangle = \langle U_y^2 \rangle = \langle U_z^2 \rangle = 2.3 \times 10^{-4} k_B T m \). In case of the rotational motion, the equipartition theorem predicts for the angular velocity \( \langle \Omega^2 \rangle = k_B T I_{\text{mz}} \), where \( I_{\text{mz}} \) is the moment of inertia tensor. Considering our swimmer, whose long axis is aligned with the \( z \) axis of a Cartesian reference frame, the theoretical predictions for both the puller [Fig. 2(b)] and the pusher [Fig. 2(a)] show a good quantitative agreement with the simulations yield \( \langle \Omega_x^2 \rangle = 1.1 \times 10^{-3} k_B T m a^2 \), \( \langle \Omega_y^2 \rangle = 1.4 \times 10^{-3} k_B T m a^2 \) and the simulations give \( \langle \Omega_z^2 \rangle = 1.2 \times 10^{-3} k_B T m a^2 \).

We now consider the active motion at \( \Re \ll 1 \). Hydrodynamics at low Reynolds numbers (relevant for micron-sized objects) allow a great simplification of the Navier–Stokes equations: the nonlinear, inertial effects can be neglected, and the governing equations are the Stokes equations

\[
\eta \nabla^2 \mathbf{u} = \nabla p - \mathbf{f}^\text{ext}, \quad \nabla \cdot \mathbf{u} = 0, \tag{1}
\]

where \( \mathbf{u}(r) \) is the fluid velocity, \( p(r) \) the pressure, \( \mathbf{f}^\text{ext}(r) \) is a body force acting on the fluid at position \( r \), and \( \eta \) is the viscosity of the fluid. Solving the Stokes equations means obtaining expressions for \( \mathbf{u} \) and \( p \) that satisfy eqn (1) and the boundary conditions. From this knowledge, the stress tensor \( \sigma \) can be calculated. For a Newtonian fluid, \( \sigma \) depends linearly on the instantaneous values of the velocity gradient, so that one can write

\[
\sigma = -p \mathbf{I} + \eta [\nabla \otimes \mathbf{u} + (\nabla \otimes \mathbf{u})^T],
\]

where \( \otimes \) indicates the tensor product, and \( \mathbf{I} \) the identity tensor. Because the Stokes equations are linear, their solution can be formally written in terms of the convolution of a Green’s function with the inhomogeneous term \( \mathbf{f}^\text{ext} \):

\[
\mathbf{u}(r, t) = \int \mathbf{O}(r - r') \cdot \mathbf{f}^\text{ext}(r', t) \, dr'. \tag{2}
\]

In free, three-dimensional space, the Green’s function is found by considering a point source \( \mathbf{f}^\text{ext} = f e \delta(r) \) acting on a infinite, quiescent fluid, where \( e \) is the unit vector representing the direction of the force, and \( \delta(r) \) the Dirac distribution. A straightforward calculation given the Oseen tensor \( \mathbf{O}(r) \equiv \frac{1}{8 \pi \eta r^4} (\mathbf{I} + \hat{r} \otimes \hat{r}) \), where \( \hat{r} \equiv r / |r| \), and the resulting flow field

\[
\mathbf{u}(r) = \frac{1}{8 \pi \eta r^4} (e + (\hat{r} \cdot e) \hat{r}),
\]

is termed a ‘Stokeslet’ and decays with distance as \( r^{-4} \).

A theoretical prediction for the puller flow field is constructed from three Stokeslets, and for the pusher we use two Stokeslets. The Stokeslets positions are placed at the midpoints of the respective force regions in the swimmer model (see Fig. 1).

We now consider the flow field generated by the active motion of our model microswimmer. We switch on the active motion with a force \( f_0 = 50 k_B T / a \) and carry out the full dynamics as described in Section II (see also Appendix A for a full definition of \( f_0 \)). Fig. 2 shows the flow fields of a pusher and a puller in the lab frame. As expected, the flow field of the puller is contractile, as fluid is drawn in from the front and the back, while fluid is pushed away normal to the swimming direction [Fig. 2(a)]. The situation is reversed in the pusher case [Fig. 2(c)] where fluid is pushed out at the front and back of the swimmer.

The theoretical predictions for both the puller [Fig. 2(b)] and the pusher [Fig. 2(d)] show a good qualitative agreement with the simulated flow fields. For the Stokesian algebraic decay of \( \mathbf{u}(r) \), the reader is referred to Section I in the ESL.

The effective velocity \( \mathbf{v}_\text{eff} \equiv |(e \cdot \mathbf{U})| \) of an isolated swimmer in the steady state depends linearly on the active force \( f_0 \). Here, \( e \) is the swimmer’s orientation and \( \mathbf{U} \) is the swimmer’s velocity. From our simulations, we calculate \( \mathbf{v}_\text{eff} \) for a pusher [see Fig. 3].

---

Fig. 2 Time-averaged flow field generated by (a) our model puller, (b) theoretical puller, (c) our model pusher, and (d) theoretical pusher. We show cross-sections on the \( x-y \) plane at \( z = 0 \). The force strength is \( f_0 = 50 k_B T / a \). The large central white regions show the hard cores of the active swimmers. The thin lines with arrows mark the streamlines, while the color code shows the magnitude of the flow velocity normalized to the thermal velocity. The large black arrows indicate the direction of motion.
The linear fit has a slope of $s = (1.45 \pm 5 \times 10^{-2}) \times 10^{-3}$.

The analogous results for pullers are also shown in Fig. 3, where the slope of the linear fit is $s = (4.4 \pm 7 \times 10^{-2}) \times 10^{-3}$.

The linear dependence of $v_{\text{eff}}$ on $f_0$ shows that our simulations correspond to the Stokes flow regime.

Further simulations on the two particle interaction statistics and two particle flow fields can be found in Section I (ESI).

IV. Density heterogeneities

In the following we will study the nonequilibrium phase behavior of our active swimmer model. The filling fraction we employ here is computed using the volume of the swimmer’s body $V_b$, as well as the volume that is spanned by the flagella sphere $V_F$, while taking into account their overlap volume $V_{\text{ol}}$.

$$\phi = \frac{(V_b + V_F - V_{\text{ol}})N}{V},$$

where $V$ is the volume of the simulated system. Note, that in an experiment only the body of the cell would be taken into account, thus the filling fraction should then be rescaled by $V_b/(V_b + V_F - V_{\text{ol}}) = 7.5 \times 10^{-2}$. Furthermore, we vary the strength of the active force $f_0$, which changes the propulsion speed $v_{\text{eff}}$ as well as the strength of the hydrodynamic interactions between the swimmers. The Péclét number captures the ratio of advection to diffusion, and can be computed using

$$\mathcal{P} = \frac{v_{\text{eff}} \sigma}{D},$$

where we used the linear relation (fitted slope) between active velocity and force dipole strength from Section III. Furthermore, $\sigma = 5a$ is the typical length of the swimmer and for the diffusion constant we assume $D = \frac{k_B T}{6 \pi \eta a}$.

We analyze the system’s density using a Voronoi tessellation and compute the local volume for each swimmer.

A global measure for the heterogeneity of a configuration of swimmers is given by the standard deviation of the distribution of local Voronoi volumes $\sigma_{\text{loc}}$. In order to remove trivial scaling factors, we compare $\sigma_{\text{loc}}$ to the standard deviation $\sigma_{\text{rnd}}$ of local Voronoi volumes for random homogeneous configurations of nonoverlapping, passive dumbbells (as in Fig. 1) with the same filling fraction. Fig. 4 shows the resulting phase diagram, with the dependence of $\sigma_{\text{loc}}/\sigma_{\text{rnd}}$ on Péclét number and filling fraction. Here, positive values of $\mathcal{P}$ correspond to pusher-type swimmers, whereas negative values are puller-type swimmers.

The phase diagram shows that for both pullers and pushers, initially $\sigma_{\text{loc}}/\sigma_{\text{rnd}}$ grows with $\mathcal{P}$ and filling fraction, then it reaches a maximum and drops to lower values. The initial increase is related to an instability that is mediated by the hydrodynamic interactions of the microswimmers which has also been found in ref. 1, 2 and 48–50. Intuitively this can be understood in the following way: the hydrodynamic flow field from the swimmers creates heterogeneities in the fluid’s velocity, which couple back to the swimmers and produce heterogeneities in the density. It should be noted that the Péclét number of the system has to be rather large, such that the hydrodynamic interaction between the swimmers is strong enough to produce heterogeneous structures. As the filling fraction increases, steric interactions grow in importance and compete with the hydrodynamic instability. Thus, we ascribe the presence of the maximum in $\sigma_{\text{loc}}/\sigma_{\text{rnd}}$ to a tapering effect of the steric interactions on the hydrodynamic instability. This tapering effect becomes visible only when steric interactions are fully accounted for.

To test the hypothesis that steric interactions stabilize the hydrodynamic instability, we carry out two more types of simulations: first, we exclude the steric effects by setting the hard-core repulsion [eqn (A6)] to zero, thus only hydrodynamic effects are included. Second, we carry out Brownian dynamics simulations, which completely neglect hydrodynamic interactions. More details about the Brownian dynamics simulations are given in Section III of the ESI.†

In Fig. 5 we show our original simulations that fully account for hydrodynamic and steric interactions, the simulations...
without steric interactions, and the active Brownian simulations. The simulations with hydrodynamics alone give rise to a strong increase of $\sigma_{loc}/\sigma_{rnd}$, but no maximum occurs, while the active Brownian simulations show a monotonic increase of $\sigma_{loc}/\sigma_{rnd}$, which is much less pronounced. Our original simulations (with Brownian simulations) confirm our hypothesis. The simulations with both hydrodynamic and steric interactions, and the active Brownian simulations. The circles are the results for simulations including only hydrodynamic interactions; and the triangles are simulations including only steric interactions.

Furthermore, fluctuations in the swimmers' motion are added to the fluid via

$$f^{\text{noise}} = \sum_i \zeta_i^L(t) \delta(r-r_{Li}) - \zeta_i^S(t) \delta(r-r_{Si}),$$

(9)

where $\zeta_i^L(t)$ and $\zeta_i^S(t)$ are noise terms with $r_{ij}^{L,S}(t) r_{ij}^{L,S}(t') = 2 \Gamma_{ij}^{L,S} \delta(t-t')$, and $\Gamma_{ij}^{L,S} = 6 \pi \eta a_i a_j$ are the friction coefficients of the front and back sphere. Considering only the active term eqn (7) can be solved

$$u = f \sum_i [O(r-r_{Li}) - O(r-r_{Si})] \cdot e_i,$$

(10)

with the Oseen tensor $O(r) \equiv \frac{1}{8 \pi \eta r} (I + \hat{r} \otimes \hat{r})$, which is regularized by using $O[r] \leq a_{ij} \equiv U^* r_{ij}^{L,S}$. Given the solution (10), the flow velocity can be eliminated from eqn (5) and (6). Using the hydrodynamic center $r^C = \frac{\Gamma_{Li} + \Gamma_{Si} r_{Si}}{\Gamma_{Li} \Gamma_{Si}}$, we can simplify eqn (5) and (6) to

$$\frac{dr^C}{dt} = v_0 e_i + \frac{1}{\eta a_i} \sum_{i 
eq j} F_{ij} + \zeta_i,$$

(11)

$$\frac{de_i}{dt} = \left[ \frac{1}{\eta a_i} \sum_{i \neq j} \tau_{ij} + \zeta^R_i \right] \times e_i,$$

(12)

where $v_0$ is the propulsion speed and $\tau_{ij} = 3m(a_i + a_j)$ is the hydrodynamic friction coefficient. The leading order of the multipole expansion of the hydrodynamic force $F_{ij}$ and torque $\tau_{ij}$ between swimmers $i$ and $j$ are given in Section IV (ESI†). The random forces $\zeta_i$ and $\zeta^R_i$ are Gaussian white noises with

$$\langle \zeta_i(t) \zeta_{ij}(t') \rangle = 2D \delta(t-t'),$$

(13)

$$\langle \zeta^R_i(t) \zeta^R_{ij}(t') \rangle = 2D_R \delta(t-t'),$$

(14)

where $D$ is the translational and $D_R$ the rotational diffusion coefficient.

Following,¹ we derive the one-body Smoluchowski equation from eqn (11) and (12). As we have seen in Section IV steric interactions among swimmers may play a crucial role in their dynamics.
Thus, we explicitly include the effect of steric interactions by means of the Ansatz\textsuperscript{27,28}
\[ v(c) = v_0 - c \zeta. \] (15)
Here, \( c \) is the concentration and the constant \( \zeta \) quantifies how much the swimmers are slowed down by the steric interactions (for more details see ref. 57 and 58). The Smoluchowski equation then reads
\[ \partial_t p = -\nabla \cdot [v(c) ep] - \frac{1}{\xi_{by}} \nabla \cdot (F_{by} p) \]
\[ - \frac{1}{\xi_{by}^2} \left( e \times \frac{\partial}{\partial e} \right) \tau_{ny} p \]
\[ + D \Delta p + D_R \left( e \times \frac{\partial}{\partial e} \right) p, \]
where \( p(r, e, t) \) is the one-particle probability distribution function of finding a swimmer at position \( r \), with orientation \( e \) at time \( t \). The first term on the right hand side of eqn (16) takes into account the active motion with a density dependent velocity, due to steric interactions; the second term accounts for the hydrodynamic forces and the third term for the hydrodynamic torques; the last two terms are responsible for translational and rotational diffusivity respectively. Here, \( F_{by} \) are the hydrodynamic two-body forces, and \( \tau_{ny} \) is the hydrodynamic torque between two particles (see also ESI\textsuperscript{+} Section V). To make progress with this equation, we follow the standard path: we consider a multipole expansion and compute moment equations for the concentration \( c \), the polarization \( P \) and the nematic order tensor \( Q \)
\[ c(r, t) = \int \text{dep}(r, e, t), \]
\[ P(r, t) = \frac{1}{c(r, t)} \int \text{dep}(r, e, t), \]
\[ Q(r, t) = \frac{1}{c(r, t)} \int \left( e \otimes e - \frac{1}{3} I \right) p(r, e, t). \]
The full equations for \( c, P \) and \( Q \) are given in Section V (ESI\textsuperscript{+}). We linearize these moment equations around the isotropic state, described by \( c = c_0 + \delta c, P = \delta P \) and \( Q = \delta Q \) and turn to Fourier space, with wave vector \( k \), where the fields are denoted by \( \delta \bar{c}(k), \delta \bar{P}(k), \) and \( \delta \bar{Q}(k) \). First order in the fluctuations the equations governing the temporal evolution read
\[ \partial_t \delta \bar{c} = - \left[ ik \left( v_0 c_0 - \zeta c_0^2 + c_0 \frac{\Delta alf^2}{30 \eta d} \right) \delta \bar{P} + D k \delta \bar{c} \right], \]
\[ \partial_t \delta \bar{P} = - \left[ ik (v_0 - \zeta c_0) \delta \bar{Q} + ik \frac{v_0}{3 c_0} - 2 \zeta \right] \delta \bar{c} \]
\[ + (D k \delta \bar{k} + D_R) \delta \bar{P}, \]
\[ \partial_t \delta \bar{Q}_{ij} = - \left[ \frac{7}{5} (v_0 - \zeta c_0) |k| \delta \bar{P}_{ij} \right]^{\text{ST}} + \frac{3 M f}{8 \pi \eta} c_0 \delta \bar{c} \delta \bar{Q}_{ij} \]
\[ + (4 D_R + k_n k_a D) \delta \bar{Q}_{ij}, \]
where \( k = |k| \) is the absolute value of the wavevector,
\[ (-\bar{M} g) = \frac{4}{5} \left( \begin{array}{ccc} -\frac{2}{3} \delta \bar{Q}_{2,2} + \delta \bar{Q}_{3,3} & \frac{2}{3} \delta \bar{Q}_{2,3} & -\delta \bar{Q}_{3,1} \\ \frac{2}{3} \delta \bar{Q}_{2,1} & \frac{2}{3} \delta \bar{Q}_{3,2} & -\delta \bar{Q}_{1,3} \\ -\delta \bar{Q}_{1,2} & -\delta \bar{Q}_{1,3} & \frac{2}{3} \delta \bar{Q}_{3,3} \end{array} \right), \] (23)
\[ [A]_{ij}^{\text{ST}} \] is the symmetric traceless form of the tensor \( A_{ij} \) and we now denote \( i, j = x, y, z \). The terms involving the force dipole strength \( f \) stem from the hydrodynamic interactions and can cause instabilities in the system. To analyze the stability of the system we will first consider pullers \( (f < 0) \) and in the second step pushers \( (f > 0) \), where we systematically keep terms up to order \( k^2 \).
For pullers \( (f < 0) \) the concentration fluctuations \( \delta \bar{c} \) are dominant. To analyze the fluctuations in the concentration we use a large length-scale and long time-scale \( (D_{ct} \gg 1) \) approximation for the longitudinal polarization \( \bar{P}_1 = \bar{k} \delta \bar{P}_1, \bar{k} = k / k \), which reads
\[ \delta \bar{P}_1 \approx - \frac{ik}{3 D_R} \left( \frac{v_0}{c_0} - 2 \zeta \right) \delta \bar{c}, \]
whereas the fluctuations \( \delta \bar{Q}_{ij} \) are of higher order in \( k \), when terms of order \( k^2 \) in eqn (20) are kept. Inserting the quasistationary solution eqn (24) into the eqn (20) yields
\[ \partial_t \delta \bar{c} = - \frac{k^2}{3 D_R} \left[ 3 D_R v_0 + c_0 \left( -3 v_0 \zeta + \frac{\Delta alf^2}{30 \eta d} \right) \right] \]
\[ + c_0^2 2 \zeta \left( \frac{\zeta - \Delta alf}{30 \eta d} \right) \delta \bar{c}. \]
For pullers the term \( c_0 \left( -3 v_0 \zeta + \frac{\Delta alf^2}{30 \eta d} \right) \) introduces an instability at low concentrations \( c_0 \), which are counteracted by the term \( c_0^2 2 \zeta \left( \zeta - \Delta alf \right) > 0 \), that stabilizes the system at higher concentrations. Since the first term is dominated by the hydrodynamic interactions and the second term by the steric interactions, we can draw the same conclusion as from the simulations: the hydrodynamic interactions cause heterogeneities in the system which are suppressed by the steric effects at larger \( c_0 \). Moreover, inspection of the fastest-growing eigenvalue reveals a maximum of instability as a function of \( c_0 \) and, hence, a maximum degree of heterogeneity. The position of the maximum of the filling fraction can be estimated from microscopic information extracted from scattering events between two swimmers; we find \( \phi_{\text{max}} = 0.12 \), which is consistent with the simulation in Fig. 5(b) (for details see Appendix B).
We also find an instability in the splay fluctuations of the nematic tensor \( \delta \bar{Q}_{11} = \bar{k} \bar{Q}_{11} \bar{k} \). Here, the approximation of the polarization fluctuations [eqn (21)] is different, since we need to consider eqn (22) for the counting of powers of \( k \). In the large length and time scale limit we arrive at
\[ \delta \bar{P}_1 = - \frac{1}{D_R} \left( \frac{v_0}{c_0} - \zeta c_0 \right) \delta \bar{Q}_{11}. \]
(26)
Inserting eqn (26) into eqn (22) and projecting on the splay part yields
\[
[k_i \delta \hat{P}_j]_{ST} \frac{k_i}{k^2} = -k^2 \frac{1}{2D_R} (v_0 - \zeta c_0) \delta \hat{Q}_{||ll}. \tag{27}
\]
Combining eqn (27) with eqn (22) gives
\[
\partial_t \delta \hat{Q}_{||ll} = \left[-k^2 \left(D + \frac{4v_0^2}{15D_R}\right) + 4D_R - c_0 \left(8\zeta v_0 k^2 \left(\frac{2}{15D_R} + \frac{2f/25}{\eta}\right)\right) + c_0^2 \frac{4k^2 + 2}{15D_R} \right] \delta \hat{Q}_{||ll}. \tag{28}
\]
From this term we also find a maximum instability, but the resulting instability is sub-dominant compared to the instability in the concentration fluctuations.

For pushers ($f > 0$) the bend component of the nematic tensor fluctuations $\delta \hat{Q}_{ij} = \hat{k} \delta \hat{Q} (1 - \hat{k} \otimes \hat{k})$ become unstable, since here $k_i \delta \hat{Q}_j (1 - \hat{k} \otimes \hat{k}) < 0$, $\forall l = x, y, z$. For the polarization we use the large length and time scale limit in eqn (26). Inserting eqn (26) into the term $[k_i \delta \hat{P}_j]_{ST}$ of eqn (22), and projecting onto the bend part gives
\[
[k_i \delta \hat{P}_j]_{ST} \frac{k_i}{k^2} (\delta \hat{Q}_j - k_j \delta \hat{Q}) = -k^2 \frac{1}{2D_R} (v_0 - \zeta c_0) \delta \hat{Q}_{||ll}. \tag{29}
\]
Therefore, we have a single equation for the bend nematic fluctuations $\delta \hat{Q}_{||ll}$, which is given by
\[
\partial_t \delta \hat{Q}_{||ll} = \left[-k^2 \left(D + \frac{v_0^2}{5D_R}\right) + 4D_R - c_0 \left(\frac{4\zeta v_0 k^2}{5D_R} + \frac{3f/25}{\eta}\right) + c_0^2 \frac{2k^2 + 2}{5D_R} \right] \delta \hat{Q}_{||ll}. \tag{30}
\]
Here, the term $-c_0 \left(\frac{4\zeta v_0 k^2}{5D_R} + \frac{3f/25}{\eta}\right)$ destabilizes the system at low concentrations $c_0$, through nematic fluctuations, which are counteracted by the terms $c_0^2 \frac{2k^2 + 2}{5D_R}$ that stabilizes the system for higher concentrations. Again, the first term, which destabilizes the system, is dominated by the hydrodynamic interactions, whereas the second, stabilizing term comes from the steric interactions. Additionally from microscopic information extracted from scattering events between two swimmers we can estimate the filling fraction of the maximum heterogeneities, with the result $\phi_{\text{max}} = 0.23$, which is in accordance with the simulation from Fig. 5(a) (for details see Appendix B).

VI. Conclusions

We have presented a new model for biological microswimmers that is based on Stokeslets and the stroke averaged motion of their flagella. The Stokeslets were distributed to model the flow fields of *C. reinhardtii* or *E. coli* cells. Furthermore, our model takes into account the anisotropic shape of a microswimmer. Typical for this is the shape of a *C. reinhardtii* cell, which is well modeled by an asymmetric dumbbell. Self-propulsion is generated through a symmetry breaking due to the asymmetric shape and force-free motion. The fluid and the hydrodynamic interactions are explicitly included with MPCD.

We show that the flow fields produced in our simulations can be predicted using simple formulae from the literature. These formulae also correspond to the experimentally measured flow fields. Additionally, we test the effective velocity of the microswimmer model, and find that it depends linearly on the applied force, in agreement with the Stokes flow regime.

We study the phase diagram in terms of filling fraction and Pécel number. We find that both pullers and pushers exhibit density heterogeneities. The density heterogeneities show a maximum at intermediate filling fractions and high Pécel number. To determine the mechanism underpinning this phenomenon, we perform additional simulations, in which either the steric interactions or the hydrodynamic interactions were switched off. Simulations with active Brownian particles showed a small linear increase in the density heterogeneities, while simulations without the steric interactions show strong density heterogeneities. This is an instability caused only by the hydrodynamic interactions, which is known from the literature. However, no maximum arises in the simulations with only hydrodynamic or only steric interactions, which shows that the maximum in the density heterogeneities is mediated by an interplay of the hydrodynamic and the steric interactions. The hydrodynamic interactions destabilize the system, whereas the steric interactions stabilize the system as the filling fraction grows and thus a maximum in density heterogeneity arises.

We have also developed an analytical theory based on a Smoluchowski equation which includes steric as well as hydrodynamic interactions. We computed the hydrodynamic moments of this equation and performed a linear stability analysis of the moment equations around the homogeneous state. For both puller and pusher-type swimmers, we found that at low concentration the system is destabilized by hydrodynamic interactions. At higher concentrations, however, the instabilities are counteracted by the steric interaction. This interplay gives rise to a maximum in the instability of the homogeneous state, and thus a maximum heterogeneity in the concentration of swimmers. The position of the maxima calculated from the analytics is in accordance with the simulations. Our continuum theory does not explicitly account for steric effects at the microscopic level, which induce short range correlations, and lubrication forces, and influence the short range hydrodynamics. These effects, however, are captured by the numerical simulations. The agreement between our theory and simulations gives us confidence that our assumptions effectively includes the dominant physical effects.

The physical pictures from both simulations and analytical theory fit together: both show that the homogeneous state is not stable and there is a maximum of instability. Also, both analyses show that the instability arises from hydrodynamic interactions and is suppressed by the steric interactions.

The maximum in the density heterogeneities might have important biological implications: it points to a possible, optimal filling fraction and Pécel number for the formation of heterogeneous distributions of motile microorganisms. Bacteria or microalgae exhibiting these optimal parameters are more likely to form colonies or biofilms.
Conflicts of interest

There are no conflicts of interest to declare.

Appendix A: model details

The microswimmer is characterized by its mass $M$, center of mass position $\mathbf{R}$ and orientation $\mathbf{q}$. In the following we describe the equations governing the motion of the microswimmer, the fluid dynamics implemented through the MPCD, and their coupling.

1. Rigid body dynamics of swimmers

As we do not consider shape deformable swimmers, we are only concerned with rigid-body dynamics. The most general motion of a rigid body is the combination of a translation along an axis (the Mozzi axis) and a rotation around the same axis, as per the Mozzi–Chasles theorem. Any orientation in space can be described using three numbers, that are commonly represented with the Euler angles, which correspond to three elementary rotations. However, there are a number of issues with the choice of the Euler angles. For instance, the composition of rotations with Euler angles or rotation matrices is rather complex, and involves trigonometric functions which lead to an accumulation of rounding-off errors. Eventually the matrices representing the rotations may become not orthogonal. Importantly, for some values of the Euler angles there are discontinuous jumps in the representation. More fundamentally, the Euler angles do not generate a covering map of the rotation group $SO(3)$, that is, the map from Euler angles to $SO(3)$ is not always a local homeomorphism. Fortunately, the topology of $SO(3)$ is diffeomorphic to the real projective space $\mathbb{P}^3(\mathbb{R})$ which admits a universal cover represented by the group of unit quaternions $\mathbf{q} = (q_0, q_1, q_2, q_3)^T$, where the superscript $T$ indicates the matrix transposition.

The equations of motion for the rigid body dynamics in three dimensions and in terms of quaternions read:

$$m\ddot{\mathbf{R}} = \mathbf{F}, \quad (A1)$$

$$\dot{\mathbf{q}} = \frac{1}{2} \left[ \mathbf{W}(\mathbf{q}) \left( \begin{array}{c} 0 \\ \Omega^b \end{array} \right) + \mathbf{W}(\mathbf{q}) \left( \begin{array}{c} 0 \\ -\Omega^b \end{array} \right) \right], \quad (A2)$$

$$\dot{\Omega}^b = (\mathbf{I}_m)^{-1} \left( \Omega^b \times (\mathbf{I}_m)^{-1} \Omega^b - (\mathbf{I}_m)^{-1} \right), \quad (A3)$$

$$\dot{\mathbf{q}}_a^b = (\mathbf{I}_m)^{-1} \left( \mathbf{I}_b^a \right), \quad (A4)$$

where $\Omega^b$ is the angular velocity of the swimmer, $\mathbf{I}_m$ the moment of inertia tensor of the swimmer in the body frame, and the indices $(x,y,z)$ take on as values the cyclic permutations of $(x,y,z)$. In eqn (A1), $\mathbf{F} = -\nabla \Phi$ and $\mathbf{T} = \mathbf{R}_b \times \mathbf{F}$ are the force and torque, respectively, acting on the swimmer due to steric interactions with the neighbor, where $\mathbf{R}_b$ is the vector connecting the center of mass of the swimmer to the point of contact with the neighbor, and the matrix $\mathbf{W}$ is (see also ref. 61)

$$\mathbf{W}(\mathbf{q}) = \begin{pmatrix} q_0 & -q_1 & -q_2 & -q_3 \\ q_1 & q_0 & -q_3 & q_2 \\ q_2 & q_3 & q_0 & -q_1 \\ q_3 & -q_2 & q_1 & q_0 \end{pmatrix}. \quad (A5)$$

The repulsive, steric interactions among swimmers are modeled using a Weeks–Chandler–Andersen potential:

$$\Phi(r_{ij,ab}) = 4\varepsilon \left[ (\sigma_{ab}/r_{ij,ab})^{12} - (\sigma_{ab}/r_{ij,ab})^6 \right] + \varepsilon \quad (A6)$$

if $r_{ij,ab} < 2^{1/6} \sigma_{ab}$, and $\Phi(r_{ij,ab}) = 0$ otherwise, where $r_{ij,ab} = |r_{ia} - r_{jb}|$ is the distance between sphere $a$ of swimmer $i$ and sphere $b$ of swimmer $j$. $\varepsilon$ is the energy scale and $\sigma_{ab}$ is the sum of the radii of sphere $a$ and sphere $b$. For the numerical integration we use the Verlet algorithm proposed in ref. 60, which was also used and discussed in detail in ref. 25.

Given a vector in the laboratory frame $\mathbf{f}$ the transformation to the body frame vector $\mathbf{f}^b$ is given by

$$\mathbf{f}^b = \mathbf{D} \mathbf{f}, \quad (A7)$$

where the matrix $\mathbf{D}(\mathbf{q})$ is constructed from the quaternions.

Quaternions are represented as $\mathbf{q} = q_0 + q_1 \mathbf{i} + q_2 \mathbf{j} + q_3 \mathbf{k}$, with $q_0, \ldots, q_3 \in \mathbb{R}$, and $\mathbf{i}^2 = \mathbf{j}^2 = \mathbf{k}^2 = \mathbf{i} \mathbf{j} \mathbf{k} = -1$.

The unitary matrix $\mathbf{D}$ that transforms vectors from the lab to the body frame is (see also ref. 61)

$$\mathbf{D} = \begin{pmatrix} q_0^2 + q_1^2 - q_2^2 - q_3^2 & 2(q_1 q_2 + q_0 q_3) & 2(q_1 q_2 - q_0 q_3) \\ 2(q_2 q_1 - q_0 q_3) & q_0^2 - q_1^2 + q_2^2 - q_3^2 & 2(q_2 q_3 + q_0 q_1) \\ 2(q_3 q_1 + q_0 q_2) & 2(q_3 q_2 - q_0 q_1) & q_0^2 - q_1^2 - q_2^2 + q_3^2 \end{pmatrix} \quad (A8)$$

Thus, the orientation of the swimmer at any given time is found from $\mathbf{D}^{-1}(\mathbf{q}(t))(0,0,1)^T$.

Note that all quantities that do not carry an index $b$ are calculated in the laboratory frame.

2. Center of mass and moment of inertia

Fig. 6 shows the detailed geometry and dimensions of our model microswimmers (both pullers and pushers). In the body frame, the swimmer is aligned with the $z$ direction and the coordinates of the centers of the $B$ and $F$ spheres are $z_1$ and $z_2$, respectively. Given a homogeneous mass distribution the center of mass of the swimmer is given by

$$z_{CoM} = \begin{pmatrix} V_1 z_1 + V_2 z_2 - V_{Sc1} \left[ z_1 + \frac{3}{4} \frac{(d_f - h_1)^2}{3d_f/2 - h_1} \right] \\ -V_{Sc1} \left[ z_2 + \frac{3}{4} \frac{(d_f - h_1)^2}{3d_f/2 - h_2} \right] \end{pmatrix} (V_1 + V_2 - V_{Sc1} - V_{Sc2})^{-1}, \quad (A9)$$

where $V_i$ are the volumes of the spheres and $V_{Sc}$ are the volumes of their spherical caps, which are cut by the other sphere. $V_{Sc1} = \frac{1}{3} \pi h_1^2 (3d_f/2 - h_1)$, (A10)
and $h_i$ are the heights of the spherical caps

$$h_1 = \frac{(d_f/2 - d_f/2 + l)(d_f/2 + d_f/2 - l)}{2l}$$

$$h_2 = \frac{(d_f/2 - d_f/2 + l)(d_f/2 + d_f/2 - l)}{2l}$$

The moment of inertia for a spherical cap about the $x$- axis as well as $y$- direction is

$$I_{Sc,(x,y)} = \rho \int_V (x^2 + y^2) dV$$

$$= \rho \int_0^{2\pi} d\phi \int_0^{\cos^{-1}(d_f/2 - h_i) \cos \theta} \frac{d^2r^2}{d\theta^2} \left[ (\cos \phi \sin \theta)^2 + (\cos \theta)^2 \right]$$

$$\times \frac{d^2r^2}{d\theta^2}$$

$$= \rho \frac{\pi}{6} h_i^3 \left( -9h_i^3 + 45h_i^2 d_f/2 - 80h_i (d_f/2)^2 + 60(d_f/2)^3 \right)$$

and about the $z$-direction is

$$I_{Sc,z} = \rho \int_V (x^2 + y^2) dV$$

$$= \rho \int_0^{2\pi} d\phi \int_0^{\cos^{-1}(d_f/2 - h_i) \cos \theta} \frac{d^2r^2}{d\theta^2} \left[ (\cos \phi \sin \theta)^2 + (\sin \phi \sin \theta)^2 \right]$$

$$\times \frac{d^2r^2}{d\theta^2}$$

$$= \rho \frac{\pi}{30} h_i^3 \left( 3h_i^2 - 15h_i d_f/2 + 20(d_f/2)^2 \right).$$

By using the moment of inertia of a sphere $I_{Sp,0} = \frac{8}{15} \pi (d_f/2)^5$ and with the use of the parallel axis theorem, we compute the moments of inertia of the swimmer as

$$I_{(x,y)} = I_{Sp,1(x,y)} - I_{Sc,(x,y)} + \rho(V_f - V_{Sc}) x_1^2,$$

$$I_z = I_{Sp,z} + I_{Sc,z} - I_{Sc,z}.$$ 

3. Multiparticle collision dynamics

To simulate a fluid at fixed density $\rho$ and temperature $T$ surrounding the swimmers, we use the MPCD algorithm, which is a mesoscopic, particle based method to simulate a fluid at the Navier–Stokes level of description. We include the Andersen thermostat and the conservation of angular momentum into the MPCD dynamics; the resulting algorithm is usually denoted as MPC-AT+. The fluid is modeled using $N_f$ point-like particles of mass $m$, whose dynamics are executed through two steps: the streaming step and the collision step. In the streaming step the fluid particles' positions $r_i, i \in \{1,N_f\}$ are updated according to

$$\frac{d}{dt} r_i = v_i,$$

where $v_i$ is their velocity and $\Delta t$ is the MPCD timestep.

The collision step mediates the interactions between the particles. Here, the system is divided into $N_c$ collision cells with a regular grid of lattice constant $a$. The center of mass velocity in each cell $C(i)$ is calculated and remains constant during the collision step, whereas the fluctuating part of the velocity of every fluid particle $i$ is randomized, which mimics the collision between particles. Hence, the velocity of particle $i$ is updated as follows:

$$v_i = \frac{1}{N_{C(i)}} \sum_{j \in C(i)} v_j + v_i^{ran} - \frac{1}{N_{C(i)}} \sum_{j \in C(i)} v_j^{ran} + m \left\{ \Pi^{-1} \sum_{j \in C(i)} \left[ r_{j,c} \times (r_j - r_i^{ran}) \right] \right\} \times r_i,$$

where the random velocity $v_i^{ran}$ has components sampled from a Gaussian distribution with zero mean and variance $\sqrt{k_B T/m}$, $N_{C(i)}$ is the number of fluid and ghost particles (see Section A4) in cell $C(i)$. The vector $r_{j,c}$ is the position of the neighboring particle $j$ relative to the center of mass of the cell $C(i)$. In eqn (A17), $\Pi^{-1}$ is the inverse of the moment of inertia tensor $\Pi \equiv \sum_{j \in C(i)} m_j [\langle r_j \cdot r_j \rangle I - r_j \otimes r_j]$ for the fluid particles in cell $C(i)$.

Note that $\Pi^{-1}$ is a dynamical quantity that has to be updated at every timestep, and it also includes the ghost particles within the swimmer (see Section A4).

To ensure Galilean invariance and avoid the build-up of spurious correlations in the velocities, the usual grid shift is performed at each timestep, that is, the grid is shifted by a random vector, whose components are uniformly distributed in the interval $[-a/2,a/2]$. 

---

**Fig. 6** Details on the geometry of our model puller-type swimmer (a) and pusher-type swimmer (b). Regions $B$ (green) and $F$ (empty circle) are the body with diameter $d_F$ and stroke-averaged flagella with diameter $d_f$, respectively, and they are separated by a distance $d$. The red arrow denotes the swimmers orientation and $C$ is the center of mass. Black circles are the force poles acting on the fluid. For pullers (a) the region $R_{pl,1}$ has the diameter $d_{pl,1}$ and the regions $R_{pl,2,R_{pl,2}'}$ have the diameter $d_{pl,2}$. For pushers (b) the regions $R_{ps,1}$ and $R_{ps,2}$ have the diameter $d_{ps,1}$. 

---

This journal is © The Royal Society of Chemistry 2018
4. Coupling of the swimmer's and fluid's dynamics

No velocity field is prescribed in our model of microswimmers. Locomotion is achieved by obeying the conservation of momentum in the collisions between the fluid particles and the swimmers; the shape asymmetry then induces self-propulsion. Two physical effects need to be included: we impose no-slip boundary conditions on the model swimmer's surface, and the force poles are explicitly included (see Fig. 1). Both effects induce modifications of the streaming and collision steps of the MPCD algorithm that we explain in the following.

a. Streaming step. To ensure the no-slip boundary condition the bounce-back rule is applied to the MPCD particles that hit the spheres during the streaming step. The velocity of the fluid particle is reversed and the change in momentum is given by

\[ \mathbf{f}_s = 2m(\mathbf{v}_s - \mathbf{U} \times (\mathbf{r}_s - \mathbf{R})) \] (A18)

where \( \mathbf{R} \) is the center-of-mass position of the swimmer colliding with the fluid particle, \( \mathbf{U} \) and \( \mathbf{\Omega} \) are the linear and angular velocity of the swimmer, \( \mathbf{r}_s \) is the position of the fluid particle upon collision with the sphere. The updated fluid particle velocity reads

\[ \mathbf{v}'_s = \mathbf{v}_s - \mathbf{f}_s / m. \] (A19)

In addition, the fluid particles are reflected back along the direction of their initial velocity. For this, we use an exact ray tracing method to detect the collision of the MPCD particle onto the swimmer’s surface. If a collision is detected the MPCD particle is propagated back onto the swimmer’s surface and then the bounce-back rule is applied. If the particle travels a fraction \( 0 < \lambda < 1 \) of the timestep towards the swimmer, then after the collision with the swimmer, it will travel for the time \((1 - \lambda)\delta t\) away from the swimmer’s surface. Furthermore, we allow for multiple collisions within the same timestep; this has been shown to prevent spurious depletion forces among colloids. The new linear and angular velocities of the swimmer after the collision with the fluid particles read

\[ \mathbf{U}' = \mathbf{U} + \sum_j \mathbf{J}_j / M, \] (A20)

\[ \mathbf{\Omega}' = \mathbf{\Omega} + \mathbf{I}_m^{-1} \sum_j (\mathbf{r}_j - \mathbf{R}) \times \mathbf{J}_j. \] (A21)

The force poles are added as external force regions in the streaming step for each swimmer. This is done by modifying the streaming step inside the force regions to

\[ \mathbf{r}_s(t + \delta t) = \mathbf{r}_s(t) + \mathbf{v}_s(t) \delta t + \mathbf{f}_s \delta t^2 / 2 \] (A22)

\[ \mathbf{v}_s(t + \delta t) = \mathbf{v}_s(t) + \mathbf{f}_s \delta t, \] (A23)

where the force in the lab frame reads

\[ \mathbf{f}_s = \mathbf{f}_s - (\mathbf{U} + \mathbf{\Omega} \times (\mathbf{r}_s - \mathbf{R})) \delta t, \] (A24)

and \( \mathbf{f}_s \) is the active force discussed in the following. The flow fields are modeled by force poles. While mathematically such force poles are point forces, any numerical implementation must mollify this requirement.

Pullers. The flow field is modeled by three Stokeslets, and the active force \( \mathbf{f}_s \) is applied to \( \mathbf{r}_{ps,1} \), \( \mathbf{r}_{ps,2} \) and \( \mathbf{r}_{pl,2} \) [see Fig. 6(a)]. The region \( \mathbf{r}_{ps,1} \) with diameter \( d_{ps,1} \) is located at the rear of the swimmer and its force points into the direction of the swimmer’s orientation. The other two regions \( \mathbf{r}_{pl,2} \) and \( \mathbf{r}_{ps,2} \) are placed on the side of the swimmer and have the opposite orientation. The angle \( \varphi_{pl} \) between the orientation of the puller and the line connecting the center of mass \( C \) and the midpoint of the region \( \mathbf{r}_{pl,2} \) (or \( \mathbf{r}_{ps,2} \)) defines their position on the boundary of the swimmer. The diameter of both \( \mathbf{r}_{pl,2} \) and \( \mathbf{r}_{ps,2} \) is \( d_{pl,2} = d_{pl,1} / (2)^{1/2} \), such that they have half the volume of \( \mathbf{r}_{ps,1} \), making the fluid force free. The total force in the region \( \mathbf{r}_{ps,1} \) is

\[ f = f_0 \frac{1}{6} \pi d_{pl,1}^3 \rho \]

and in the regions \( \mathbf{r}_{pl,2}, \mathbf{r}_{ps,2} \) is

\[ f = f_0 \frac{1}{6} \pi d_{pl,2}^3 \rho. \]

Pushers. The flow field is modeled by a force dipole. We apply the force \( \mathbf{f}_{ac} \) to all fluid particles located within spherical regions [see Fig. 6(b)]. The regions \( \mathbf{r}_{ps,1} \) and \( \mathbf{r}_{ps,2} \) where \( \mathbf{f}_{ac} \) is applied are equally sized spheres with diameter \( d_{ac} \) and the two forces \( \mathbf{f}_{ac} \) are equal and opposite, to ensure that the fluid is overall force free. To generate a smooth flow on the boundary of the swimmer the direction of the applied force in regions \( \mathbf{r}_{ps,1} \) and \( \mathbf{r}_{ps,2} \) is modeled as follows. For fluid particles \( \mathbf{r}_i \in \mathbf{r}_{ps,1} \) or \( \mathbf{r}_{ps,2} \) we apply the force

\[ f_{ac}^b = \begin{cases} (0, 0, f_0), & \text{if } s_b^2 > d_{ps}/2, \\ (0, 1, f_0), & \text{if } s_b^2 < d_{ps}/2, \end{cases} \] (A25)

Here, \( s_b = (s_{x,b}, s_{y,b}, s_{z,b})^T \) is the distance between the MPCD particle and the center of the region \( \mathbf{r}_{ps,1} \) or \( \mathbf{r}_{ps,2} \). The small, black arrows in Fig. 6(b) give a schematic representation of the flow field arising from eqn [A25]. As before the superscript \( b \) denotes the body frame, in which the swimmer’s orientation is aligned with the \( z \)-axis. The constant \( f_0 \) gives the strength of the force that is applied. The total force in one of the two regions can be estimated by integrating eqn (A25), which yields \( f = f_0 \frac{5}{48} \pi d_{ps}^3 \rho \).

This takes into account the redirection of the force on the boundary of the swimmer and the density \( \rho \) of the fluid.

b. Collision step. To guarantee the no-slip boundary conditions on the surface of the swimmers, it is necessary to fill each swimmer with ghost particles, such that the collision step can be properly executed. \(^{65}\) The positions of the ghost particles \( \mathbf{r}_g^b \) are uniformly distributed within the swimmer, \( \mathbf{z} \) and are advected with the swimmer in each timestep. The ghost

\[^{66}\text{We recommend to fill the swimmer with multiple ghost particles, rather than using a single ghost particle of large mass, as this would result in a wrong value of the torque.}\]
particles density is matched to the fluid’s density so as to make the swimmer neutrally buoyant. Before every collision step the ghost velocities $v_i^g$ are updated according to

$$v_i^g = U + \Omega \times (r_i^g - R) + v_i^{an}, \quad (A26)$$

where the components of $v_i^{an}$ are sampled from a Gaussian distribution. The ghost particles then (together with the fluid particles) take part in the collision step [see eqn (A17)], and their velocities are updated to $v_i^g$. The resulting change in linear momentum due to the ghost particles is $J_i^g = m (v_i^g - r_i^g)$ and the change in angular momentum is $L_i^g = (r_i^g - R) \times J_i^g$. These changes are then transferred to the swimmer$^{17}$

$$U' = U + \sum_j J_j^g / M, \quad (A27)$$

$$\Omega' = \Omega + I_{an}^{-1} \sum_j L_j^g. \quad (A28)$$

5. Algorithm implementation

In this section we explain how the present algorithm is implemented. First, note that the MPCD algorithm scales as $O(N)$, and thus is particularly prone to an efficient implementation with parallel programming. We therefore implemented the entire dynamics (both swimmer and MPCD) on graphics processing unit (GPU) cards using the Nvidia CUDA language. Parallelization of the MPCD algorithm is rather straightforward. The streaming step is performed for each fluid particle independently in a separate CUDA kernel, whereas ghost particles are simply translated together with the corresponding swimmer. Then, a kernel for each particle is started to carry out the bounce-back rule and afterwards a kernel for each particle is started to apply the periodic boundary conditions. The collision step eqn (A17) is implemented with the following kernels:

1. the cell of each particle is found and the center of mass for each cell is found by starting a kernel for each particle;
2. the two sums over the velocities in eqn (A17) are computed, where a kernel for each velocity component of each particle is started;
3. a kernel for each particle is started to compute the number of particles in each cell;
4. a kernel for each cell is started to normalize the velocities and compute the center of mass;
5. a kernel for each particle is started to compute its position with respect to the center of mass;
6. a kernel for each particle is started to compute the cross product in eqn (A17) and the contributions in each cell are summed;
7. six kernels for each particle are started to compute the contribution to the six components of the inertia tensor in the corresponding cell;
8. a kernel for each cell is started to compute the inverse of the respective inertia tensors;
9. for each cell a kernel is started to compute the curly bracket in eqn (A17);
10. a kernel for each particle is started to add all contributions of eqn (A17) and finish the collision step, while considering the rule for ghost particles eqn (A26);
11. a kernel for each ghost particle is started, which computes the momentum and angular momentum transfer [eqn (A27) and (A28)] to the swimmers.

Some of these computations could be combined into single kernels, but it is computationally more efficient to start many kernels with small computations, which we have opted for in this algorithm. After the MPCD steps the rigid body dynamics part (see also Section A1) of the code is executed. It consists of the following kernels, one for each swimmer:

1. the velocity and angular velocity stemming from the fluid interaction are added to the swimmers;
2. the positions and quaternions are updated;
3. periodic boundary conditions are applied;
4. the neighbor list between swimmers is updated (here we use a linked list);
5. the steric forces and torques are computed;
6. the velocities and angular velocities are updated.

This concludes the algorithm, which now goes back to the streaming step of the MPCD part.

6. Computational details

We carried out three-dimensional simulations with an average of $\langle N_c \rangle = 20$ fluid particles per cell. The timestep of the MPCD algorithm is fixed to $\delta t = 10^{-2}\sqrt{ma^2/(k_B T)}$, whereas the MD timestep is $\delta t_{MD} = 5 \times 10^{-4}\sqrt{ma^2/(k_B T)}$. The resulting kinematic viscosity $\nu = \eta / \rho$ of the fluid for the MPC-AT+α algorithm (including both kinetic and collisional contribution) can be calculated exactly as $\nu = 3.88a\sqrt{k_B T/m}$. Simulations using a forced flow (for details see ref. 69) produced a viscosity of $\nu = 3.69a\sqrt{k_B T/m}$.

The large sphere $F$ associated to the stroke-averaged flagella of the swimmer has a diameter of $d_l = 7a$, while the small sphere $B$ associated to the body of the swimmer has $d_B = 3a$, and the distance between the spheres centers is $l = 7a$. The choice of the geometrical parameters is dictated by a combination of factors. First, it is computationally convenient to make the swimmers’ linear size a few times the grid spacing $a$. Second, inspired by the geometric properties of *C. reinhardtii* a ratio $d_l/d_B \geq 2$ is advisable.$^{10}$ For the sake of clarity in the comparison of our results, we maintain the same geometry also for pushers. The energy scale of the steric interactions is set to $\epsilon = 10k_B T$. For pushers we fix the diameter of the force dipole regions $R_{p,1}$ and $R_{p,2}$ to $d_p = 3a$. The region $R_{p,1}$ of the pullers has the same diameter $d_{p,1} = 3a$ and accordingly the regions $R_{p,2}$ and $R_{p,2}$ have the diameter $d_{p,2} = 3a/(2)^{1/3}$. The angle between the swimmers orientation and the line connecting the center of mass of the pullers $C$ to the midpoint of the regions $R_{p,1}$ and $R_{p,2}$ is $\theta_{p} = 107^\circ$.

To initialize the simulations, we distribute the swimmers homogeneously (and without overlaps) across a cubic box with periodic boundary conditions.
Appendix B: estimate of maximum from theoretical prediction

To estimate the maximum predicted from the analytical treatment, we computed the maximum of the eigenvalues in eqn (30) for pushers and eqn (25) for pullers. As a first step, we have to find estimates for both the wavenumber \( k \) and the constant quantifying the steric interactions \( \zeta \). Since we expect the hydrodynamic interactions to be relevant on the size of the swimmer, we chose the distance between the force poles \( l \) to determine the characteristic wavenumber \( k = 2\pi /l \). The constant \( \zeta \) quantifying the steric interactions can be estimated from ref. 58

\[
\zeta = v_0^2 \sigma_s \tau_c \tag{A29}
\]

where \( v_0 \) is the propulsion speed, \( \sigma_s \) the geometrical cross section, and \( \tau_c \) is the collision time. The collision time can be estimated from the center-of-mass distance of two colliding swimmers, where we considered collisions as shown in Fig. S5 and S6 (ESI†). The resulting collision time for pullers is \( \tau_c = 9.5\mu s / \sqrt{kbT/m} \) and for pushers \( \tau_c = 18.0\mu s / \sqrt{kbT/m} \). The propulsion speed is extracted from Fig. 3 and the geometrical cross section is \( \sigma_s = 4\pi \sigma^2 \). Here, \( \sigma = l/2 = 7.5a \) is the effective steric radius of the swimmer also estimated from the distance of the force poles. To finally obtain a filling fraction we use \( \phi = c_0(\zeta + V_F - V_0) \) [see eqn (3)].

Acknowledgements

We gratefully acknowledge insightful conversations with Johannes Blaschke, Jens Elgeti, Stephan Herminghaus, Sebastian Mair, and Kuang-Wu Lee. We also gratefully acknowledge support from the Deutsche Forschungsgemeinschaft (SFB 937, project A20). Open Access funding provided by the Max Planck Society.

References