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Filling an Emulsion Drop with Motile Bacteria

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We have measured the spatial distribution of motile *Escherichia coli* inside spherical water droplets emulsified in oil. At low cell concentrations, the cell density peaks at the water-oil interface; at increasing concentration, the bulk of each droplet fills up uniformly while the surface peak remains. Simulations and theory show that the bulk density results from a "traffic" of cells leaving the surface layer, increasingly due to cell-cell scattering as the surface coverage rises above ~10%. Our findings show similarities with the physics of a rarefied gas in a spherical cavity with attractive walls.

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The study of self-propelled particles [1], natural or synthetic "swimmers," is a part of the emerging field of "active matter physics" dealing with collections of intrinsically out of equilibrium entities. While problem-specific continuum theories exist [2], there is as yet no general recipe comparable to equilibrium free energy minimization for predicting the behavior of a general active matter system from knowledge of its microscopic constituents alone.

Confinement is of significant interest in diverse areas of physics. In active matter, it is already known that self-propelled particles can confine themselves spontaneously. Motile *Escherichia coli* and other bacteria encountering a surface continue to swim along it, giving rise to self-organised confinement to a 2D layer. Experimentally, the number density of motile *E. coli* between two parallel glass slides peaks strongly at the slides [3,4]. However, in this geometry, the cell density between the walls remains low and there is little 3D confinement, because the "wall-hugging" swimmers can escape essentially to infinity along the wall. Even the 2D confinement is weak: surface swimmers spread out to minimize interaction, and single-body physics suffices to explain wall hugging [3–6].

The interesting question now arises: what would happen if there is confinement in all spatial dimensions? Biologically, motile bacteria in nature are sometimes confined in this way, e.g., in raindrops [7] or infected host cells [8], possibly leading to motility loss [9]. In physics, the collective behavior of confined swimmers has attracted recent interest. At high density, motile *Bacillus subtilis* in a cylindrical drop develop stable vortices [10,11], while simulations of swimmers in a 2D box suggest novel forms of phase separation near close packing in the absence of hydrodynamic interactions [12]. In this work, we perform experiments starting from the opposite limit, and probe the way in which interaction effects emerge amongst motile bacteria confined within spherical emulsion droplets as their average density, ρ_0 , increases from a small value.

As expected, at $\rho_0 \rightarrow 0$, we observe motile cells hugging the inner surface in a layer [3,4]. Instead, our focus here is on the unexpected way that the drop fills as ρ_0 increases: the bulk density increases uniformly while the surface peak remains. We present simulations and theory that reproduce essential features of our observations, and which suggest that the physics is reminiscent of a rarefied gas in a spherical cavity with attractive walls [13].

We studied spherical water-in-oil emulsion drops with a range of radii, R, containing a smooth-swimming mutant of *Escherichia coli* AB1157 bacteria in phosphate motility buffer at increasing average cell density, ρ_0 . Green fluorescent protein expressed by the bacteria and a dye that preferentially adsorbs to the emulsifier on the water-drop surfaces allowed us to take high-resolution fluorescent confocal image stacks and reconstruct the cell positions inside droplets, Fig. 1 (cf. Supplemental Material movies



FIG. 1 (color online). Snapshots of 2 μ m-thick cross sections of droplets, (ϕ , R) = 0.47%, 17.4 μ m; 2.7%, 16.1 μ m; 6.2%, 14.3 μ m (left to right). Top: experiment, confocal images, ~2 μ m below the equator, with red droplet edges and green bacteria. Bottom: simulated equatorial cross sections.

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1 and 2 [14]). The majority of droplets had $R = 10 \ \mu m$ 20 μm (Fig. S1), and are therefore significantly smaller than the persistence length of our swimmers (estimated to be $\lambda \gtrsim 100 \ \mu m$). In situ optical characterization using an oxygen-sensitive dye [25] and differential dynamic microscopy [26,27] confirmed the absence of oxygen gradients within these droplets and that cells swam during the duration of our experiments with essentially a constant speed distribution. Taking care to minimize aberration effects arising from working with spherical drops (Figs. S2, S3 [14]), we counted cells within concentric shells to obtain the cell density as a function of distance from the center, $\rho(r)$, which we assume to be isotropic. (See [14] for all preparative and imaging details.)

Typical density distributions, $\rho(r)$, for $R = (14 \pm 2) \mu m$ over a range of cell-body volume fraction, ϕ_0 [for $(2 \times 1) \mu m$ spherocylinderical cells] are shown in Fig. 2, with $\rho(r)$ normalized by the average number density ρ_0 and the radius by the droplet radius *R*. Each curve is the result of averaging over 10 stacks of analyzed images.

At low ϕ_0 , bacteria are localized in a shell beneath the water-oil interface. Visually, almost all of these were motile, although a few nonmotile cells were also localized at this interface. As ϕ_0 increases, the peak in ρ/ρ_0 drops and migrates inward, while the cell density throughout the rest of the drop increases uniformly. This scenario occurs at all *R* studied (Fig. S4 [14]), although there was not enough statistics to investigate *R* dependence systematically. It is easily explained why our findings are independent of droplet size. We work with $2R < \lambda$; a qualitative change is only expected when droplets become larger than the swimmers' persistence length, λ .

While the $\rho(r)/\rho_0$ peak decreases with ϕ_0 , Fig. 2, the absolute number of cells hugging the surface increases with ϕ_0 . Figure 3 (inset) shows the total number of cells, N_s , found within the peak [28] plotted against the total number



FIG. 2 (color online). Radial bacterial number density distributions, $\rho(r)$, normalized by the average number density of the whole droplet, ρ_0 , plotted against the radial distance from the center, *r*, normalized by the droplet radius, *R*, averaged over 10 data sets. Left: Experimental data for $R \approx 14 \ \mu\text{m}$. Right: Simulation data for $R = 16 \ \mu\text{m}$ over a similar range of volume fractions as in experiment obtained for the case of maximal scattering at cell-cell collision.

of cells, N_0 . We report N_s as the total area the surface cells would cover as a monolayer, N_sA_b (where A_b = area covered by one cell), normalized by the droplet surface area ($A = 4\pi R^2$), $\eta = N_sA_b/A$ (the "surface area fraction"); N_0 is also reported as the total cell area NA_b similarly normalized, $\eta_0 = N_0A_b/A$.

Note that the peaks in Fig. 2 are much larger than crowding-induced layering in confined hard particles. For hard spheres in a rigid spherical cavity, there is no surface peak until $\phi_0 \gtrsim 20\%$ [29]; we see well-developed peaks at $\phi_0 \ll 5\%$ due to wall hugging [3,4] and not crowding.

The bulk of the droplet fills in a surprising way as ϕ_0 increases. Recent simulations [12] may suggest the interpretation that the wall layer is a high-density "liquid" phase coexisting with a low-density "vapor" phase in the bulk of the droplet. In this scenario, we should expect the interface between these two phases to move continuously inward as ϕ_0 is raised, as was observed for motile *B. subtilis* in a cylindrical droplet [10]. Instead, the bulk of our spherical droplets fill up uniformly with *E. coli* as ϕ_0 increases; see Fig. 2.

A qualitative explanation is as follows. At $\phi_0 \rightarrow 0$, cells are found almost exclusively at the inner droplet surface due to wall hugging [3–6]. This lower-density surface layer is as yet noninteracting. As ϕ_0 becomes finite and more and more cells arrive at the interface, the surface coverage eventually reaches a point when there will only be room for another cell if an existing surface cell leaves, spontaneously due to reorientation, or by scattering off the arriving cell or with another surface cell. Since $\lambda > 2R$ and the bulk density remains relatively low, a "departing" cell most likely travels along an approximately straight trajectory to



FIG. 3 (color online). The surface area fraction of bacteria, η , expressed as a fraction of the total potential surface coverage, η_0 , as a function of η_0 , from experiment (circle) and simulations $(\theta_r^{(c)} = 0, \text{ downward triangle}; \theta_r^{(c)} \approx 0.077, \text{ upward triangle}; \theta_r^{(c)} = \pi, \text{ star})$. Inset: The same data plotted in terms of the un-normalized surface coverage η : the absolute number of bacteria at the droplet surface increases sublinearly with the average cell density. Both plots: solid line denotes fit to Eq. (1); dashed line denotes total surface accumulation.

another part of the interface, where the process repeats. This cross-droplet traffic is manifested as a uniform increase in bulk density.

Consider first a simple analytical model for this picture. Bacteria with speed v give rise to a uniform flux towards a flat surface. We measure the bulk and surface concentrations using the equivalent area fractions η_0 and η already introduced, so that the inward flux is $\propto (\eta_0 - \eta)v$, where the factor $\eta_0 - \eta$ measures the number of cells in the bulk. We assume that an arriving cell will be trapped at the surface if it is presented with an empty surface, the latter with probability $\propto (1 - \eta)$. Thus, the arriving flux $\propto (1 - \eta)(\eta_0 - \eta)v$. Surface cells also swim with speed v, and remain trapped until they either leave spontaneously with probability $\propto \eta$ due to orientational fluctuations, or are scattered by other cells with probability $\propto \eta^2$. In the steady state, fluxes to and from the surface balance read as

$$(1 - \eta)(\eta_0 - \eta) = c\eta + b\eta^2;$$
 (1)

i.e., the arriving flux (left-hand side) is balanced (right-hand side) by cells leaving the surface due to spontaneous reorientation (*c* term) and scattering (*b* term). For swimmers of size $\sim a$ confined to a region of size $\sim R$, simple kinematics give [14] $c \simeq R/v\tau$, where τ is a typical reorientation time after which a surface cell spontaneously returns to the bulk, and $b \simeq kR/a$, where *k* is the probability that a surface cell-cell collision scatters a cell into the bulk.

To compare this simple kinetic model with experiments, we use Eq. (1) with $c = R/v\tau$ and b = kR/a to fit the experimental data, Fig. 3 (•), taking into account that each point pertains to a droplet with a different R (Fig. S5 [14]) shows the R dependence explicitly). This gives best-fit values $\tau_{\rm fit}^{-1} = (0.18 \pm 0.04) \text{ s}^{-1}$ and $k_{\rm fit} = 0.04 \pm 0.03$, the significant uncertainties in the latter quantity being a direct reflection of the rather noisy data. In our kinematic model, $\tau_{\rm fit}^{-1} = 0.18 \, {\rm s}^{-1}$ is the rate at which trapped cells leave the surface due to spontaneous reorientation [14]; this is comparable to our estimate of the free rotational diffusion coefficient of a single flagellated cell, $D_r \simeq 0.2 \text{ s}^{-1}$ [14]. The other parameter, $k_{\rm fit} = 0.04$, represents the probability of cell-cell scattering leading to a cell leaving the surface. This parameter is difficult to determine directly from observations, but visual inspection of Supplemental Material movies 1 and 2 [14] suggests that a k of a few percent is not unreasonable.

To compare directly with experimental data, we use the average droplet radius, $R = 16 \,\mu\text{m}$, and average cell velocity, $v = 13.5 \,\mu\text{m s}^{-1}$ m, to find $c_{\text{fit}} = 0.21$ and $b_{\text{fit}} = 0.38$, and plot Eq. (1) as the full line in Fig. 3. The agreement is satisfactory, especially given the significant amount of noise in the data. In the Supplemental Material [14], we show that the predicted *R* dependence of

the coefficients in Eq. (1) is also verified to within experimental uncertainties.

There are apparent similarities between our kinetic model, Eq. (1), and one proposed recently [30] to explain activity-driven phase separation [12]. However, the absence of particle-particle interaction in [12,30] gives rise to coexistence been liquid and vapor phases each of fixed density. On the other hand, the density dependence of the fluxes in Eq. (1) means that our surface layer and bulk densities change with the average cell concentration, and what we observed is incompatible with the coexistence of two phases with invariant densities.

We next simulated N_0 spherocylinders (end-to-end $\ell = 2 \ \mu$ m, diameter 0.8 μ m) initially distributed uniformly inside a rigid sphere of radius *R*. Each cell self-propels at $v = 13.5 \ \mu$ m s⁻¹ along its long axis and diffuses translationally and rotationally with isotropic diffusivities $D = 0.25 \ \mu$ m² s⁻¹ and $D_r = 0.2 \ s^{-1}$ respectively, mimicking *E. coli*. Surface cells can reorient away from the surface with rotational diffusivity D_r .

What happens when two motile *E. coli* cells collide is likely very details dependent [6]. To access the essential physics, we simulated this process using a single phenomenological parameter, $\theta_r^{(c)}$, the "angular deflection at collision": two colliding cells (bulk or surface) change their propulsion directions by an angle chosen uniformly from [0, $\theta_r^{(c)}$], and rotate around the original direction by an angle chosen uniformly in $[-\pi, \pi]$. The time step was 1 ms. (See [14] for details.)

If $\theta_r^{(c)} = 0$, two colliding cells remain stuck until Brownian motion frees them. Since our data suggest that collision-induced movement of cells away from the surface layer is important, we explore first the opposite limit of maximal reorientation, viz., $\theta_r^{(c)} = \pi$, which reproduces the most prominent aspects of the observed phenomenology (see Fig. 2): a peak in $\rho(r)/\rho_0$ that decreases as ϕ_0 increases, with a uniformly increasing bulk density.

The peak width, Δ , is narrower in simulations than in experiments. In the former, $\Delta \simeq 0.05R \simeq 1 \ \mu\text{m}$, about half of the simulated cell length, $\ell = 2 \ \mu\text{m}$. Previous simulations of wall hugging at moderate propulsion forces [5] have also found $\Delta \approx \ell/2$, which is explained as a remnant of the depletion zone next to a wall in the case of passive hard rods. The wider experimental peaks, $\Delta \simeq 3 \ \mu\text{m}$, presumably result partly because real *E. coli* with flagella behave as rods considerably longer than $\ell = 2 \ \mu\text{m}$. However, already-noted [14] near-edge optical aberrations may also contribute to the apparent Δ .

We turn next to the number of cells in the surface layer. As expected, the case of no collision-induced reorientation, $\theta_r^{(c)} = 0$, gives a constant η/η_0 as η_0 increases, Fig. 3 (\checkmark). This does not reproduce our data. Complete randomization at collision, $\theta_r^{(c)} = \pi$, Fig. 3 (\star), is more realistic. The actual data (•), which lie between these two limits, are accounted for by $\theta_r^{(c)} \approx 0.077$, Fig. 3 (\blacktriangle). To make sense of this value,

note that the average reorientation at a cell-cell collision given by a particular value of $\theta_r^{(c)}$ can be recast as an effective collisional rotational diffusivity, $D_r^{(c)} = \theta_r^{(c)}/6\Delta t$, where $\Delta t = 1$ ms is our time step [14]. Thus, our data suggest $D_r^{(c)} \approx 1 \text{ s}^{-1}$, which is about five times the Brownian D_r . The plots of $\rho(r)/\rho_0$ at various cell densities at this value of $\theta_r^{(c)}$ (Fig. S6 [14]) display the same phenomenology as those shown in Fig. 2, although the uniform rise of the bulk density with ϕ_0 is not as rapid as observed.

It would be unrealistic to expect our simple simulated model to reproduce exactly the totality of the data shown in Figs. 2 and 3. Most importantly, the details of surface swimming depends sensitively on precise geometric parameters of the swimmers [31]. Furthermore, the distance between a wall-hugging cell and the surface can fluctuate by up to a cell width or more [32,33], partly due to the complicated wobble of the cell body, and the effect of surface curvature remains largely unexplored. A basic model in which cells arriving at a surface simply align perfectly with it cannot be expected to account for such complexities, and therefore of the shape of the surface peak, Fig. 2. Such complexities may have less effect on an "integral measure" such as the total number of trapped cells, Fig. 3, which is indeed what we have fitted to theory and simulations.

Our system shows certain similarities with a confined classical rarified gas in which the mean free path, λ , is larger than or comparable to the confinement length, L, i.e., the Knudsen number, $Kn = \lambda/L > 1$. In both cases the particles can traverse the confined space in a straight line. For hard spheres of radius *a* at volume fraction ϕ_0 confined to a sphere of radius R, $\text{Kn} = 1/(6\sqrt{2}\phi_0 R/a)$ [13], which for the droplets reported in Fig. 2 ranges from $\text{Kn} \simeq 1.9$ at $\phi_0 = 0.22\%$ through Kn $\simeq 0.5$ at $\phi_0 = 0.79\%$ to Kn \simeq 0.06 at $\phi_0 = 6.2\%$. Simulations [13] show that in a confined rarefied gas at Kn > 1 with attractive walls, the evolution of the density profile as a function of average gas density is closely similar to that shown in Fig. 2: a broadening surface peak and uniformly increasing bulk density. The attraction in our case comes from wall hugging [3,4].

This analogy is no longer appropriate either at large ϕ_0 or when the persistence length of the swimmers drops below the system size, $\lambda/2R < 1$. The latter can be probed using wild-type cells, which tumble every 1 s or so between straight runs. Given $v \leq 20 \ \mu m s^{-1}$, we now have $\lambda \leq 20 \ \mu m$. Figure 4 compares the density profile for a smooth swimmer in a drop with $2R \ll \lambda$, and a wild type in a drop with $2R > \lambda$. The density peak at the droplet edge has disappeared in the latter case, presumably because surface tumbles now remove cells from the trapped layer too rapidly for a peak to build up.

At cell densities 5–10 times higher than the maximum reported so far, we observed vortices with constantly



FIG. 4 (color online). Comparison of density profiles for smooth swimmers (red) and wild-type run-and-tumble cells (blue).

changing orientations (Supplemental Material movie 3 [14]). This may be compared to *B. subtilis* confined to cylindrical water drops at high densities, where a single vortex aligned to the cylinder axis is seen [10]. The difference may partly be due to the greater length of *B. subtilis* cells, and partly to differing spatial symmetry. Such collective motion is left to future work.

Finally, if internal flows from bacterial motility can set up exterior flows, then our droplets should display at least enhanced positional fluctuations. Tracking revealed no such activity. This is likely because the lecithin layer stabilizing each droplet is rigid enough to decouple internal and external flows.

To summarize, we have observed the emergence of many-body behavior in spherical water droplets filled with increasing density of motile *E. coli* bacteria. The singlebody physics of previously studied wall hugging [3,4] together with Brownian reorientation taking cells from the surface into the bulk suffice to explain observations up to a surface coverage of $\eta_0 \approx 0.1$, Fig. 3. Thereafter, the decrease of η/η_0 with η_0 evidences cell-cell scattering; fitting to a simple theory suggests that a few percent of collisions scatter cells from the surface layer to the bulk. Bulk traffic of cells from one part of the inner surface to another, initially due solely to Brownian reorientation, and then increasingly due to cell-cell scattering, explains the observed uniform increase in the bulk density as the average cell density increases.

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Filling an emulsion drop with motile bacteria: Supplementary Material

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Sample preparation

P1 phage transduction [1] was used to create a smooth swimming strain (AB1157 Δ cheY) using the appropriate *E. coli* K12 single knockout mutant from the KEIO collection [2]. Kanamycin (final concentration 30 µg ml⁻¹) was added to all growth media. The GFP encoding plasmid pHC60 was extracted using a QIAGEN Plasmid mini kit, transformed into AB1157 Δ cheY using a method based on CaCl₂ as detailed in [3], and maintained by tetracycline (final concentration 5 µg ml⁻¹).

Bacteria were grown overnight in Luria-Bertani broth at 30 °C shaken at 200 rpm; harvested in the exponential phase; washed three times by careful filtration with a 0.45 µm filter and resuspended in a phosphate motility buffer (6.2 mM K₂HPO₄; 3.8 mM KH₂PO₄; 67 mM NaCl; 0.1 mM EDTA at 7.0 pH) to optical densities of OD = 1 to 3 (at 600 nm), corresponding (from plate count) to cell densities of (1.55 to 4.65) × 10⁹ ml⁻¹.

Emulsions were obtained by mechanically dispersing bacterial suspension ($\sim 2 \% v/v$) in sunflower oil (Sigma, used as purchased). We could create droplets stable for days without adding surfactant, presumably due to native lecithins in the oil [4]. Samples of emulsion ($\sim 400 \,\mu$) were loaded into ($8 \times 8 \times 8$) mm open coverglass chambers for microscopic observation from below.

In situ characterisation

In situ differential dynamic microscopy (DDM) [5, 6] was performed to measure the swimming speed distribution, P(v), of the cells within individual droplets and the fraction of non-motile organisms. We found an average swimming speed inside a droplet to be $\bar{v} \simeq (13.5 \pm 0.7) \,\mu m \, s^{-1}$ at cell concentrations similar to those used in our experiment (OD=2, initial cell density used for emulsification). The fraction of non-motile organisms inside an emulsion drop was around 30%. Interestingly, at high cell densities (OD=24, initial cell density used for emulsification) although the average swimming speed was $\bar{v} \simeq 12 \,\mu m \, s^{-1}$, the fraction of non-motile organisms decreased to around 10%.

The oxygenation condition inside droplets was monitored using a ruthenium dye, RTDP [7], whose fluorescence is quenched by molecular oxygen. We found using DDM that the dye did not affect cell motility at concentrations $\lesssim 50 \,\mu$ M. Imaging the intensity of RTDP-containing baterial droplets showed that our cells experienced spatiotemporally constant oxygen conditions for many hours; in particular, there were no oxygen gradients near droplet edges. Thus, oxytaxis is absent. Presumably, as in [8], the high solubility of oxygen in oil [9] keeps each drop well oxygenated.

Imaging & image reconstruction

We studied the spatial distribution of cells in bacteriacontaining droplets using confocal microscopy. A green channel imaged GFP-labelled cells. The dye 1,1'dioctadecyl-3,3,3',3'-tetramethylindocarbo-cyanine perchlorate (Dil, Molecular Probes) was dissolved in the oil to highlight in a red channel the droplet-stabilising lecithins, with which it has an enhanced affinity. Each image, taken in an inverted Zeiss AXIO Observer.Z1 microscope with an LSM700 scanning module and a $60 \times$ oil immersion objective, covered a $(193 \times 193) \,\mu\text{m}$ field to an optical depth of $2\,\mu\text{m}$, and typically showed ≤ 10 droplets with various (R, ρ_0) . We acquired z-stacks of typically $30 \text{ images} (256 \text{ lines}, 0.8 \text{ s per slice}) \text{ spaced at } 2 \,\mu\text{m of all}$ the droplets in a field of view, starting several μm below the bottom of the sample compartment (Supplementary Movie 1). All images were acquired within 2 hours of sample preparation.

Local intensity maxima were identified in the green channel (GFP-labelled bacteria, excited at 488 nm) of Fiji-filtered [10] images to give cell coordinates, while red channel (Dil-stained oil, excited at 555 nm) images were analysed to yield droplet centres and radii, Fig. 1, main text. The droplet were approximately spherical. We selected typically 12 points at the w/o interface of each droplet within its stack of images and fit these points to an ellipsoid [11]. We found $0.03 \leq 1 - c/a \leq 0.17$, where a and c are the longest and shortest principal axes, with little systematic correlation between c/a and droplet size.

The measured distribution of average droplet radius, R, is shown in Fig. S1.

Supplemental Movie 1: Confocal *z*-stack of a typical w/o emulsion that encapsulates active bacteria. a) Processed stack. b) Overlay of the positions of bacteria with the unprocessed stack.

Supplemental Movie 2: Phase-contrast movie of a typical w/o emulsion that encapsulates active bacteria.



FIG. S1. Distribution of droplet radii in our emulsions.

Optical Distortions

In principle, counting cells within concentric shells ('bins') of width Δr gives the cell density as a function of distance from the center, $\rho(r)$, which we assume to be isotropic. However, it is well known that refractive index differences within a sample induce distortions in the image [12]. These distortions can range from simple local blurring of features (i.e. reduction in resolution) to severe 3-dimensional geometric distortions, producing images with little resemblence to the object. In the current context, bacteria are imaged through a smooth spherical interface between the sunflower oil (n_o = 1.46) and the aqueous buffer (n_w = 1.33).

In order to assess the nature and severity of the resultant distortions, we modelled confocal imaging of objects within water droplet using a commercial ray tracing package (ZEMAX 13 Professional, Radiant Zemax, LLC). The modelled microscope is based around a publicly available desription of a high numerical aperture oil immersion lens [13]. It assumes illumination using a blue (450 nm)point source and confocal detection, with signal from either green fluorescent objects or red fluorescent oil (with increased fluorescence in a thin shell surrounding the water droplet) being recorded. For simplicity, transverse scanning was simulated by moving the sample through a fixed laser focus rather than scanning the laser as in the experiments. Axial scans were performed by changing the thickness of the oil immersion layer by a prescribed amount dz.

Representative results showing an x - z cross-section passing through the centre of a droplet of radius 20 µm are presented in Fig. S2. Distortions within the upper hemisphere are evident, both in the red signal from the oil, which gives an upper hemisphere with an apparently smaller radius, and the location of the green objects. Therefore only experimental data from the lower hemisphere were included in our analysis. However, even



FIG. S2. Simulation of optical distortion due to imaging inside a spherical droplet of higher refractive index than the surrounding medium. The outline of the upper hemisphere is severely distorted. White points inside become blurred and shifted into the green patches. Positional errors increase as the angle from the central axis increases, to the extent that (inset) an object on the equator can become two. Thus, we only used points within a $\pi/3$ cone to acquire quantity data for calculating $\rho(r)$.

within the lower hemisphere distortions are noticeable.

Within a cylinder of radius $\sim 0.75R$ the shape and contrast of the green objects are hardly affected and the position shows a relatively small systematic underestimation of the axial distance (~ 10 %). This implies in particular that the shape of the radial bacteria density distribution near the droplet centre is little affected by imaging distortions. However, for large transverse distances the contrast and shape are clearly affected, leading to a low contrast region near the equator. The distortions of the particle shape are highlighted by a spherical object touching the droplet surface at the equator (see inset): it shows up as two distinct maxima. The onset of this region of low contrast and severe distortions can be roughly estimated by looking at a ray in z-direction hitting the spherical interface at a distance x from the central axis of the droplet. Its angle of incidence is $\sin(\theta) = x/R$, and within the droplet Snell's law gives $\sin(\theta') = n_o/n_w \sin(\theta) = (n_o/n_w)(x/R)$. However, beyond the critical angle $\theta_c = \sin^{-1}(n_w/n_o) \approx 66^\circ$ the ray can not actually enter the droplets any more, giving rise to a 'blind' region. It was therefore decided to restrict analysis of the experimental data to the bottom sector with $\theta_S = 60^\circ$ opening half angle.

Our observed shift and broadening of the peak could at least partially be due to these distortions, so the detailed



FIG. S3. Comparison of the bacterial radial number density distribution for one droplet, considering measured bacterial positions only within a spherical cone of half-angle θ_{\max} from the imaging axis. At large deviations from the imaging axis data points are distorted by spherical abberations. As θ_{\max} is decreased, the distribution converges on a robust shape. In our data analysis we chose $\theta_{\max} = \pi/3$, as a compromise between minimising random (statistical) and systematic (optical distortion) errors.

shape of the radial profile plots close to the edge should not be overinterpreted, Fig. S3. On the other hand, there appears no systematic radius dependence of the observed phenomenology, Fig. S4.

Diffusivities and persistence length

We worked mostly with smooth (non-tumbling) swimmers, whose persistence length, λ , is due to loss of orientation as a result of rotational Brownian motion. The mean-squared angular drift of each cell is given by $\langle \theta^2 \rangle \simeq D_r t$, where the rotational diffusivity D_r is controlled mainly by the length of the cell body plus flagellum. Modelling cell+flagella by an ellipsoid with semi-axes $a \simeq 5 \,\mu\text{m}$ and $b \simeq 0.5 \,\mu\text{m}$, then gives $D_r = k_B T (16\pi \eta a b^2/3)^{-1} \simeq 0.2 \,\text{s}^{-1}$ [14]. (This value is the same order of magnitude as but larger than that measured recently from a related strain of *E. coli* [8].) The rotational relaxation time is $D_r^{-1} \simeq 5$ s, and the persistence length for $\bar{v} \simeq 20 \,\mu\text{m} \,\text{s}^{-1}$ swimmers is $\lambda \simeq \bar{v}\tau_r \simeq 100 \,\mu\text{m}$. Thus, all of our emulsion droplets are considerably smaller than the persistence length of our smooth swimmers.

We next estimate upper and lower bounds of the translational (center of mass) diffusivity of the cells. If we model a cell as a sphere with volume appropriate to a $(1 \times 2) \,\mu\text{m}$ spherocylinder, then the Stokes-Einstein relation gives $D \simeq 0.3 \,\mu\text{m}^2 \,\text{s}^{-1}$. On the other hand, modelling (cell + flagella) as an ellipsoid with the dimensions given above gives $D \simeq 0.1 \,\mu\text{m}^2 \,\text{s}^{-1}$. We use $D = 0.25 \,\mu\text{m}^2 \,\text{s}^{-1}$ in our simulations.

Analytic model

We consider a slab of bacterial solution of size R with N_b uniformly distributed bacteria. Bacteria swim with a constant velocity v. The number of bacteria hitting the surface in time dt is then $N_b v dt/R$. The probability of staying at the surface after hitting it is taken to be equal to the fraction of the surface that is free from bacteria, $1 - N_s A_b/A$, where N_s is the number of bacteria at the surface, A_b is the surface area covered by one bacterium and A is the total surface area. Therefore, the total number of bacteria arriving and staying at the surface during time dt is

$$dN_s = \left(1 - \frac{N_s A_b}{A}\right) N_b \frac{v \, dt}{R} \,. \tag{S1}$$

We assume that there are two mechanisms for bacteria leaving the surface. First, each cell may leave by its own reorientation, with characteristic time τ , i.e. with rate $N_s(\gamma dt)$ where $\gamma = \tau^{-1}$. Secondly, a call may encounter another cell in a two-body 'scattering' event, which can be modelled by $\beta N_s^2 dt$. The change in surface cell number in time dt is therefore:

$$dN_s = \left(1 - \frac{N_s A_b}{A}\right) N_b \frac{v \, dt}{R} - \gamma N_s \, dt - \beta N_s^2 \, dt \,, \quad (S2)$$

where β is the scattering frequency, and γ the self-scattering frequency. These values are difficult to estimate for arbitrary surface coverages but can be calculated for low N_s .

With regards to inter-bacterial scattering, consider bacteria swimming at the surface and select one bacterium as a 'target'. The probability of another bacterium hitting the target from a distance λ is $2a/2\pi\lambda$, where *a* is the radius of a bacterium, which is assumed to have a circular projection on the surface. The number of bacteria hitting the target from a thin shell $(\lambda, \lambda + d\lambda)$ is

$$\frac{2a}{2\pi\lambda} \left[\pi \left(\lambda + \mathrm{d}\lambda \right)^2 - \pi\lambda^2 \right] \frac{N_s}{A} = 2a \frac{N_s}{A} \,\mathrm{d}\lambda \,. \tag{S3}$$

The target can be hit only from a circle of radius v dt, so that the total number of scattering events between all bacteria and a selected target is

$$\int_0^{v \, \mathrm{d}t} 2a \frac{N_s}{A} \, \mathrm{d}\lambda = 2av \, \mathrm{d}t \frac{N_s}{A} \,. \tag{S4}$$

Since the same argument is valid for every bacterium on the surface, the total number of scattering events is proportional to

$$\frac{1}{2}N_s 2av \,\mathrm{d}t \frac{N_s}{A} = av \,\mathrm{d}t \frac{N_s^2}{A} \,, \tag{S5}$$

where the factor 1/2 is introduced to account properly for the number of bacterial pairs. By comparing this



FIG. S4. Radial bacterial number density distributions, $\rho(r)$, normalised by the average number density of the whole droplet, ρ_0 and the droplet radius, R, for several datasets with different radii and cell densities. Upper: Experiment. Lower: Simulation.

FIG. S5. Experimentally determined normalised fraction of cells in the surface peak, η/η_0 , plotted against the droplet radius R (•) compared to best-fit theoretical predictions (•).

expression with Eq. (S2)), we identify

$$\beta = k \frac{av}{A} \,, \tag{S6}$$

where k is a probability that one of the bacteria participating in a scattering event would come off the surface. Once again, this argument only properly works for a low surface coverage. Using Eq. (S2) with the approximation Eq. (S6), we obtain in the steady-state

$$\left(1 - \frac{N_s A_b}{A}\right) N_b \frac{v}{R} - \gamma \frac{R}{v} N_s - k \frac{av}{A} N_s^2 = 0.$$
 (S7)

Finally, observing that $N_s + N_b = N$, where N is the total number of bacteria in the system, and introducing $\eta = N_s A_b/A$ and $\eta_0 = N A_b/A$, we obtain Eq. (1) in the main text, where $b = kaR/A_b \simeq kR/a$ and $c = \gamma R/v = R/v\tau$. Note that η_0 can be significantly larger than 1.

Testing R dependence

The coefficients of Eq. (1) in the main text, c and b, are R-dependent. Rewriting it to show the R-dependent coefficients explicitly, this equation predicts that the peak density η_i of a droplet of radius R_i is given by:

$$(1 - \eta_i)(\eta_0 - \eta_i) = \left(\frac{R_i}{v\tau}\right)\eta_i + \left(\frac{kR_i}{a}\right)\eta_i^2, \qquad (S8)$$

which is the expression that we fitted to the data to obtain $\tau_{\rm fit}^{-1} = (0.18 \pm 0.04) \, {\rm s}^{-1}$ and $k_{\rm fit} = 0.04 \pm 0.03$. In the main text we use the average droplet radius of $R = 16 \, {\rm \mu m}$ and an average cell velocity of $v = 13.5 \, {\rm \mu m \, s}^{-1}$ to find $c_{\rm fit} = 0.21$ and $b_{\rm fit} = 0.38$. The theoretical prediction resulting from substituting these values into Eq. (1) is given as the full line in Fig. 3 of the main text.

Here, we replot the data points in Fig. 3 of the main text but now showing the *R*-dependence of the normalised peak density, η/η_0 , Fig. S5.

For comparison, theoretical points $\{\eta_i\}$ are calculated from Eq. (S8) with $\tau = \tau_{\text{fit}}$ and $k = k_{\text{fit}}$. Thus, in Fig. S5, each experimental data point has a best-fit theory point directly above or below it (i.e. with the same R). Especially given the noisy data, the overall agreement is satisfactory.

Simulation algorithm

We simulated our system in continuous, 3-dimensional space and discrete time. Thus, the algorithm iterated for each spherocylinder i at position r_i self propelled with velocity v_i parallel to the long axis, at each time-step of size Δt is,

$$\begin{aligned} \mathbf{r}_{i}(t + \Delta t) &= \mathbf{r}_{i}(t) + \mathbf{v}_{i}(t)\Delta t + \sqrt{2}\mathrm{D}\Delta t \boldsymbol{\eta}_{i}(t) ,\\ \mathbf{v}_{i}(t + \Delta t) &= \mathbf{R}(\theta \eta_{i,r}(t)) \mathbf{v}_{i}(t) , \end{aligned} \tag{S9}$$

where $\boldsymbol{\eta}$ is a vector of unit gaussian noise representing translational Brownian motion with diffusion constant D and **R** is a rotation matrix representing random rotational noise of magnitude $\theta_{\rm r}$ ($\eta_{\rm i,r}(t)$ is a sample from a unit gaussian distribution).

The magnitude of rotational noise θ is determined with

$$\theta_{\rm r} = \sqrt{2D_{\rm r}\Delta t} + \begin{cases} \theta_{\rm r}^{\rm (c)} & \text{ collision} \\ 0 & \text{ no collision} \end{cases}$$
(S10)

where D_r is the bulk rotational diffusion constant, and $\theta_r^{(c)}$ is a maximum deflection angle in $[0, \pi]$.

If two cells overlap after propagation over a time step, then the two cells are 'back tracked' to their previous position. In subsequent time steps, Brownian motion (D and D_r) and collisional deflection, $\theta_r^{(c)}$, eventually free the two cells. If $\theta_r^{(c)} = \pi$, then the two cells take randomised orientations during a single time step.

If a cell overlaps with a surface after propagation over a time step, the cell is 'back tracked' to a position where the overlap vanishes, and the velocity is aligned parallel to the surface,

$$\hat{\boldsymbol{v}}_{i\parallel}(t) = \frac{\boldsymbol{v}_{i}(t) - (\boldsymbol{v}_{i}(t) \cdot \hat{\boldsymbol{u}}) \boldsymbol{v}_{i}(t)}{\left| \boldsymbol{v}_{i}(t) - (\boldsymbol{v}_{i}(t) \cdot \hat{\boldsymbol{u}}) \boldsymbol{v}_{i}(t) \right|}$$
(S11)
$$\boldsymbol{v}_{i}(t + \Delta t) = \left| \boldsymbol{v}_{i}(t) \right| \hat{\boldsymbol{v}}_{i\parallel}(t) ,$$

where \hat{u} is the outward normal to the droplet surface.

Finally, to understand the significance of the parameter $\theta_{\rm r}^{\rm (c)}$, we recast it in terms of an effective collision-induced diffusivity. The variance of the uniform distribution on $[-\theta_{\rm r}^{\rm (c)}, \theta_{\rm r}^{\rm (c)}]$ out of which we choose the reorientation angle at collision is $\sigma^2 = \theta_{\rm r}^{\rm (c)2}/3$. To map this to an effective

FIG. S6. Radial bacterial number density distributions, $\rho(r)$, normalised by the average number density of the whole droplet, ρ_0 , plotted against the radial distance from the center, r, normalised by the droplet radius, R, averaged over 10 data sets for $R = 16 \,\mu\text{m}$. Left: No collision-induced reorientation, $\theta_r^{(c)} =$ 0. Right: Intermediate level of collision-induced reorientation, $\theta_r^{(c)} = 0.077$. We have used a log vertical scale to highlight the fact that the bulk density does *not* rise in the case of $\theta_r^{(c)} = 0$, and the rise is less rapid than is seen in either plots in Fig. 2 of the main text in the case of $\theta_r^{(c)} = 0.077$.

rotational diffusion with diffusivity $D_r^{(c)}$, we recall that this process produces a Gaussian distribution with variance $\sigma^2 = 2D_r^{(c)}\Delta t$ that scales linearly with the time interval, which we take as our time step (1 ms). Equating variances gives the expression quoted in the main text: $D_r^{(c)} = \theta_r^{(c)2}/6\Delta t$.

Surface peak evolution with fitted value of $\theta_{r}^{(c)}$

Figure S6 (right) shows how the surface peak evolves with average cell volume fraction, ϕ_0 using the value of $\theta_r^{(c)} = 0.077$, which corresponds to a collision-induced rotational diffusivity that is about five times the Brownian rotational diffusivity. These simulations give a very similar phenomenology, but the quantitative comparison with experiments is less exact than the case of full randomisation at collisions, especially at the highest ϕ_0 shown. A corresponding plot for the case of no collision-induced orientation ($\theta_r^{(c)} = 0$) is also shown, Fig. S6 (left), where there is little evolution as ϕ_0 increases.

Collective motion

We show in a movie the emergence of collective vortices in droplets with cell densities an order of magnitude higher than the highest ϕ_0 considered in the main text, with an estimated $\phi_0 \sim 30 \%$.

Supplemental Movie 3: Phase-contrast movie of a an emulsion drop encapsulating swimming *E. coli* at a density of ϕ_0 estimated at 30 %. Note the existence of large-scale (~ droplet size) vortices.

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