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Gyrotactic Suppression and Emergence of chaotic trajectories of swimming particles in three-dimensional flows

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We study the effects of imposed three-dimensional flows on the trajectories and mixing of gyrotactic swimming micro-organisms, and identify new phenomena not seen in flows restricted to two dimensions. Through numerical simulation of Taylor–Green and ABC flows, we explore the role that the flow and the cell shape play in determining the long-term configuration of the cells’ trajectories, which often take the form of multiple sinuous and helical ‘plume-like’ structures, even in the chaotic ABC flow. This gyrotactic suppression of Lagrangian chaos persists even in the presence of random noise. Analytical solutions for a number of cases reveal the how plumes form and the nature of the competition between torques acting on individual cells. Furthermore, studies of Lyapunov exponents reveal that as the ratio of cell swimming speed relative to the flow speed increases from zero, the initial chaotic trajectories are first suppressed and then give way to a second unexpected window of chaotic trajectories at speeds greater than unity, before suppression of chaos at high relative swimming speeds.

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I. INTRODUCTION

The behaviour of swimming micro-organisms in active suspensions [1, 2] is crucial in the survival of species, since it allows them to attain favourable locations in relation to resources, predators and each other. Modelling the spatio-temporal population dynamics has a rich history due to the importance of unicellular organisms such as phytoplankton in ocean currents [3]. Phytoplankton are responsible for the production of half of the world’s oxygen from biomass, and form the base of the aquatic food chain. Identifying optimal distributions may also have applications in biofuel production [6] which is becoming increasingly important for our planet’s long term sustainability. Orientation due to bottom heaviness (gyrotaxis), or equivalently negative-buoyancy plus body-shape (rheotaxis), is a feature of many free-swimming micro-organisms such as the ubiquitous freshwater single-cell green alga Chlamydomonas reinhardtii and the single-cell, green-alga genus Dunaliella found in salt water.

Being bottom-heavy, gyrotactic cells experience a gravitational torque which acts to re-orient them if they move away from their preferred vertical orientation. Consequently cells focus into downwelling regions resulting in the formation of plumes. Irrespective of shape, orientation will also be affected by local vorticity in the fluid, while non-spherical cells may experience an additional torque due to the rate-of-strain. The balance of these torques and their effect on trajectories is examined in this paper.

Torney & Neufeld [7] and Khurana et al. [8, 9] computed the dynamics of freely-rotating prolate swimming particles in both steady and unsteady 2D imposed laminar test flows consisting of periodic sinusoidal vortices given by a stream function of the form

\[ \psi(x, y, t) = U \sin(x + B \sin(\Omega t)) \sin y, \]  

where \( x \) and \( y \) are Cartesian coordinates, \( t \) is time, \( U \) is the maximum flow speed, \( \Omega \) is the angular frequency of the lateral oscillation of the flow and \( B \) is its amplitude. For steady flows (\( B = 0 \)), cell trajectories divide into two regions at slow swimming speeds. In one region, cells aggregate around the boundaries of the vortices with chaotic trajectories that move between vortices. In the other region, cells are trapped within chaotic domains in centres of the vortices, but above a threshold speed, the particles escape from this regular elliptic region. For thin rod-like particles, the threshold velocity decreases to zero so that there are no barriers to transport across cells. Furthermore, as the swimming velocity increases, the rate of particle transport can decrease due to the formation of traps near elliptic islands that hold swimming cells for long times. This effect is enhanced for more elongated particles and when stochastic terms are added to the swimming model. At higher swimming speeds, rod-like cells are attracted to stable manifolds of hyperbolic fixed points.

In a recent numerical study, Durham et al. [10] examined the behaviour of gyrotactic, spherical cells in a two-dimensional (2D) imposed flow and discovered both regimes in which the cells were concentrated into patterns and those in which they were randomly distributed.

When in addition to a 2D flow, the orientation of gyrotactic swimming cells is restricted to two dimensions as in [10], cells tumble when vorticity is high, but with realistic unrestricted 3D orientation (even in a 2D flow), cells reorientate in the third (cross-flow) dimension and spin about their axis, so that the cell swimming direction has an additional degree of freedom, as shown by
Pedley & Kessler [11] (Section 3.1). This means that even the components of the mean cell swimming speed projected on to the 2D plane are different when the cells are allowed unrestricted 3D orientation.

Thorn & Bearon [5] investigated the transport and dispersion of spherical gyro tactic organisms with 3D orientation in steady 2D homogeneous linear shear flows, with and without additive stochastic reorientation. Gyrotaxis in fully developed turbulent flows was also studied by them and more recently in [12] and [13].

However, there are few studies in steady inhomogeneous three-dimensional flows, which increase the complexity by introducing mixing and Lagrangian chaos [14, 15], but remain amenable to analytical progress. In this paper, we fill this gap in knowledge and relax the unphysical restrictions in [10] by allowing the cells to orientate in three dimensions and investigating the role of cell shape in inhomogeneous three-dimensional test flows for ellipsoidal cells subject to gyrotaxis. We identify new phenomena not seen in flows restricted to two dimensions.

II. MODEL FORMULATION

We consider two incompressible velocity fields. The first is the Taylor–Green Vortex (TGV) flow, where the fluid velocity is [10, 17]

$$
\mathbf{u} = \begin{pmatrix}
-2 \cos x \sin y \sin z \\
\sin x \cos y \sin z \\
\sin x \cos y \cos z
\end{pmatrix}.
$$

Here $x = (x, y, z)^T$ is the laboratory-frame position vector in Cartesian coordinates where $x$ and $y$ are the horizontal components and $z$ is vertically upwards. This is a very commonly used test flow, both due to its tractability and as it represents an exact closed form solution of the incompressible Euler equations. The streamlines are steady closed counter-rotating vortices, and the vorticity is

$$
\omega \equiv \nabla \times \mathbf{u} = 3 \cos x \begin{pmatrix}
0 \\
- \sin y \cos z \\
\cos y \sin z
\end{pmatrix}.
$$

(Note that $\omega$ is a 2D TGV flow when $B = 0$.) Secondly, we impose the Arnold–Beltrami–Childress (ABC) flow [18]

$$
\mathbf{u} = \begin{pmatrix}
\cos y + \sin z \\
\cos z + \sin x \\
\cos x + \sin y
\end{pmatrix}.
$$

This is a more complex but consequently biologically-relevant flow as it exhibits Lagrangian chaos. It also possesses the Beltrami property that $\mathbf{u} = \omega$ and has open streamlines [19].

Each velocity field requires a separate non-dimensionalisation. The same scaling as in [10] has been used in equation (1) for the TGV flow i.e. lengths have been scaled by $L/2\pi$, velocities by $L\omega_0/2\pi$, vorticity by $\omega_0$ and time by $1/\omega_0$, where $L$ is the spacing between the cube of adjacent vortices in the flow and $\omega_0$ is the maximum vorticity at the centre of these vortices. For the ABC flow in (2), lengths are non-dimensionalised by $1/2\pi$, velocities by $u_0$, vorticity by $2\pi u_0$ and time by $1/2\pi u_0$, where $u_0$ is the maximum fluid velocity.

The swimming cells are modelled as gyro tactic prolate ellipsoids and are so small ($\approx 10 - 20\mu$m) that the local Reynolds number is effectively zero. Using the well-established model of Pedley and Kessler [20], the non-dimensionalised equations determining cell orientation $\mathbf{p}$ and position $x$ are the gyro tactic equation,

$$
p = \frac{d\mathbf{p}}{dt} = \frac{1}{2G} [\mathbf{k} - (\mathbf{k} \cdot \mathbf{p})\mathbf{p}] + \frac{1}{2} (\omega(x) \times \mathbf{p}) + \alpha \mathbf{f},
$$

and cell velocity equation,

$$
x = \frac{dx}{dt} = V\mathbf{p} + \mathbf{u}(x).
$$

Here $\mathbf{f} = (I - \mathbf{p}\mathbf{p}) \cdot \mathbf{E}(x) \cdot \mathbf{p}$ and $\mathbf{E}(x)$ is the rate-of-strain tensor, which has components

$$
E_{ij} = (\partial u_i/\partial x_j + \partial u_j/\partial x_i)/2 \quad \forall \quad i, j = 1, 2, 3.
$$

In [5] and [10], the unit vector $\mathbf{k}$ defines the preferred swimming direction of cells which is vertically upwards, $I$ is the identity tensor, $t$ is time, $V$ defines swimming speed of cells relative to the flow, $G$ describes the rate of gyro tactic reorientation. When $G \ll 1$ cells tend to swim vertically upwards regardless of the ambient flow, and as $G \to \infty$ the cells rotate freely in the flow. The cell eccentricity is $\alpha = (a^2 - b^2)/(a^2 + b^2)$, where $a$ and $b$ are the lengths of a cell’s semi-major and minor axis respectively. Note that $\alpha = 0$ describes a sphere (e.g. a spherical squirmer [21]) and $\alpha = 1$ a thin rod. Here we assume that the suspension is sufficiently dilute that cell-cell interactions can be neglected and the cells have no influence on the bulk flow. Equations (5) and (6) are expanded below for later reference:

$$
p = \begin{pmatrix}
\frac{p_x p_x + \omega_y p_z - \omega_z p_y + \alpha f_x}{2G} \\
\frac{p_y p_z + \omega_z p_x - \omega_x p_z + \alpha f_y}{2G} \\
\frac{1 - p_z^2 + \omega_x p_y - \omega_y p_x + \alpha f_z}{2G}
\end{pmatrix}
$$

and

$$
x = \begin{pmatrix}
V p_x + u_x \\
V p_y + u_y \\
V p_z + u_z
\end{pmatrix}.
$$
where
\[
\begin{align*}
    f_x &= E_{11}p_x(1 - p_x^2) + E_{12}p_y(1 - 2p_x^2) + E_{13}p_z(1 - 2p_x^2) \\
     & \quad - 2E_{22}p_y^2p_x - 2E_{23}p_xp_yp_z - E_{33}p_z^2p_x^2, \\
    f_y &= -E_{11}p_x^2p_y + E_{12}p_x(1 - 2p_y^2) - 2E_{13}p_xp_yp_z \\
     & \quad + E_{22}p_y(1 - p_y^2) + E_{33}p_y^2p_z, \\
    f_z &= -E_{11}p_x^2p_z - 2E_{12}p_zp_y + E_{13}p_x(1 - 2p_z^2) \\
     & \quad - E_{22}p_y^2p_z + E_{33}p_y^2p_z + E_{33}p_z(1 - p_z^2).
\end{align*}
\]

Cartesian coordinates are retained in (8) instead of converting to spherical polars (11) to avoid potential computational difficulties at the poles.

For direct numerical simulation of the cells’ trajectories in the TGV and ABC flows, a fourth order Runge–Kutta scheme with a time step of \( \delta t = 0.01 \) was implemented; simulations with \( \delta t = 0.005 \) were also carried out as a convergence check. Each simulation was initialised with 1000 cells with random orientations and positions in a \((2\pi)^3\) box with periodic boundary conditions on all sides.

For both of the three-dimensional TGV and ABC flows, we explore the parameter ranges
\[
(G, V) \in (0, 10) \times (0, 10) \quad \text{and} \quad 0 \leq \alpha \leq 1.0.
\]

No qualitatively new features were found outside this range, and moreover we note that in their example of 2D flow with 2D trajectories, Durham et al. [10] did cover a wider parameter range \((G, V) \in (10^{-2}, 10^2) \times (10^{-2}, 10^2)\) but the extreme ranges show no additional patterns.

III. TAYLOR–GREEN VORTEX FLOW

For all values of \( G \) and \( \alpha \), simulating passive tracers with \( V = 0 \) results in no discernible pattern, with cells instead remaining randomly spread as \( t \to \infty \), with each cell remaining within its initial vortex. When \( V > 0 \), cells can swim across streamlines and are no longer confined to a single vortex. Setting \( G = 1 \) for a slow swimming speed \( V \approx 0.025 \) results in slight clustering around \( y = \pi/2 \) and \( 3\pi/2 \). Increasing the swimming speed to \( V = 0.1 \) results in the cells forming the organised structures shown in Figure 1. When \( \alpha = 0.1/2 \), the system collapses onto two sheets at \( y = \pi/2 \) and \( 3\pi/2 \) so that cell movement is exclusively in the \( xz \)-plane. For both of these eccentricities cells form structures that almost exclusively follow the streamlines of the flow due to their relatively low swimming speed. When \( V \approx 0.15 \) any structure degenerates and as \( t \to \infty \) cell trajectories again appear to be random. In general, organised structures do not always develop, but are instead only found for specific parameter ranges.

Simulating spherical cells with \( V = 1 \) and \( G = 0.1 \) results in all trajectories tending to one of four distinct points,
\[
(3\pi/2, 3\pi/2, \pi), (\pi/2, 3\pi/2, \pi), (3\pi/2, \pi, 0), (\pi/2, \pi, 0),
\]
where cells subsequently settle and orient upwards, i.e. \( p = (0, 0, 1)^T \). Cells are now positioned so that up-swimming exactly cancels the fluid velocity and

FIG. 1: Simulations of cells in the TGV flow with \( V = 0.1 \) and \( G = 1 \): (a) \( \alpha = 0 \), (b) \( \alpha = 0.5 \) and (c) \( \alpha = 1 \).
a fixed steady state has been reached. Substituting these values of \( x \) and \( p \) into (5) and (6) confirms that \( \dot{p} = \dot{x} = 0 \). These single-point solutions are also seen in Table II when \( \alpha = 0, V = 1 \) and \( G = 1 \), just above ‘x-shape’ closed-loop trajectories reminiscent of Bowditch–Lissajous figures \([22]\) that lie in the planes \( y = \pi/2 \) and \( y = 3\pi/2 \).

Examination of intermediate speeds reveals a number of diverse formations, three of which are seen in Figures 3a, 3b and 3c. Simulating with the parameters \( V = 3 \) and \( G = 1 \) produces four curved plumes in the \( yz \)-plane. From Figure 3a, we observe that \( p_x = 0 \) and the trajectories lie in the planes \( x = \pi/2 \) and \( x = 3\pi/2 \), on which the vorticity vanishes (see (3)). Substituting these values into (8) and (9) yields

\[
\dot{p} = \begin{pmatrix}
0 \\
\frac{p_{0}p_{z}}{2G} \\
\frac{1}{1 - p_{z}^{2}} \\
\frac{2G}{2G}
\end{pmatrix}
\]

and

\[
\dot{x} = \begin{pmatrix}
0 \\
Vp_{y} \pm \cos y \sin z \\
Vp_{z} \pm \sin y \cos z
\end{pmatrix}.
\] (13)

The general solution of (13) is

\[
p = \begin{pmatrix}
0 \\
\frac{D e^{-t/2G}}{C e^{-t/G} - 1} \\
\frac{1}{1 + C e^{-t/G}}
\end{pmatrix},
\]

(15)

where \( C \) and \( D \) are arbitrary constants. Taking the limit as \( t \to \infty \) and inserting into (14) yields

\[
\dot{x} = \begin{pmatrix}
0 \\
\pm \cos y \sin z \\
V \pm \sin y \cos z
\end{pmatrix}.
\]

(16)

These equations now define a simple two-dimensional flow with closed streamlines and plotting particle trajectories reveals a similar pattern to that in Figure 3a.

For rod-like cells (\( \alpha = 1 \)), clustering again occurs leading to the formation of four plumes, now in the \( xz \)-plane. From Figure 3b, it is clear that

\[
p_{y} = 0 \quad \text{and} \quad y = \pi/2 \quad \text{or} \quad 3\pi/2,
\]

which are inserted into (5) and (6) to obtain

\[
\dot{p} = \begin{pmatrix}
\frac{p_{0}p_{z}}{2} \pm g_{x} \\
0 \\
\frac{1}{1 - p_{z}^{2}} \pm g_{x}
\end{pmatrix}
\]

and

\[
\dot{x} = \begin{pmatrix}
3p_{x} \pm 2 \cos z \sin z \\
0 \\
3p_{z} \pm \sin x \cos z
\end{pmatrix}.
\]

(17)

(18)

where

\[
g_{x} = p_{z}(-2 + p_{z}^{2}) \cos x \sin z
\]

\[
+ p_{x}(2 - 2p_{z}^{2} + p_{z}^{2}) \sin x \sin z,
\]

(19)

\[
g_{z} = (p_{z}/2)(5/2 + p_{z}^{2}) \cos x \cos z
\]

\[
+ p_{z}(-2p_{z}^{2} + 1 + p_{z}^{2}) \sin x \sin z.
\]

(20)

As \( \alpha > 0 \), cells are now affected by the rate-of-strain in the flow, so that the complexity of (14) is increased and an analytical solution cannot be obtained. However tracking the orientation of cells numerically for large values of \( t \), shows that

\[
-0.6 < p_{x} < 0.6, \quad p_{y} = 0 \quad \text{and} \quad 0.8 < p_{z} < 1,
\]

which is consistent with the \( xz \)-view in Figure 3b. Particle trajectories can again be plotted using (18). Without an analytical solution for (17) we must now select a few of the extreme values of \( p_{x} \) and \( p_{z} \) to gain a structure comparable to Figure 3b. This can be seen in the \( xz \)-view of Figure 3b in which red arrows show the trajectories when \( p_{x} = -0.6 \) and the blue dashed arrows correspond to \( p_{x} = 0.6 \).

Lastly, trajectories for which \( \alpha = 1/2 \) are examined. Figure 3c shows that cells now form an intricate structure involving eight curved plumes on one-dimensional manifolds. Since \( \alpha \neq 0 \), an analytical solution again cannot be found for (3) but moreover as cells now continue to move in all three planes (9) cannot be simplified. However this does demonstrate the importance of cell shape, with \( \alpha \) acting as a bifurcation parameter.

Recalling Figure 2, a pattern involving four plumes in the \( yz \)-plane was seen for spherical cells (\( \alpha = 0 \)), simulating rods (\( \alpha = 1 \)) resulted in four plumes in the \( xz \)-plane while \( \alpha = 1/2 \) gives eight plumes on one dimensional manifolds. Thus varying eccentricity affects both the number and location of plumes, while we conclude that increasing \( \alpha \) leads to more complex structures, or cells follow three-dimensional flow and any clustering breaks down leading to full mixing \([23, 24]\).

As \( V \) increases, the diversity in behaviour subsides and when \( V \gtrsim 5 \) cells only form the upwelling structures seen in Figure 4. Spherical cells collect in two sheets in the \( yz \)-plane, while clustering results in four upwelling plumes when \( \alpha = 1/2 \) and \( \alpha = 1 \). In Figure 4 irrespective of shape,

\[
p = (0, 0, 1)^{T} \quad \text{and} \quad x = \pi/2 \quad \text{or} \quad 3\pi/2,
\]

so (16) holds again. When \( \alpha = 1/2 \) and \( \alpha = 1 \), cells cluster at \( y = \pi/2 \) or \( 3\pi/2 \), and (16) simplifies to

\[
\dot{x} = \begin{pmatrix}
0 \\
0 \\
V \pm \cos z
\end{pmatrix}.
\] (22)

Consequently, trajectories are now restricted to the vertical plumes seen in Figures 3(b) and 4(c). This demonstrates that increasing swimming speed suppresses the effects of the surrounding flow \([25]\). The full range of
parameter values that have been used in simulations of the TGV flow and the steady state patterns that are found are given in Tables I and II.

FIG. 2: Simulations of cells in the TGV flow with $V = 3$ and $G = 1$: (a) $\alpha = 0$, (b) $\alpha = 0.5$ and (c) $\alpha = 1$. 
(b) $\alpha = 1$. The red and blue arrows show trajectories formed using (18) when $p_x = \pm 0.6$ respectively.

(c) $\alpha = 1/2$.

FIG. 3: Projections of cell trajectories in the TGV flow for $V = 3$ and $G = 1$.

FIG. 4: Simulations of cells in the TGV flow with $V = 9$ and $G = 1$: (a) $\alpha = 0$, (b) $\alpha = 0.5$ and (c) $\alpha = 1$. 
IV. ABC FLOW

We now turn our attention to the more complex of the two test flows which we shall see does not prevent clustering despite the presence of Lagrangian chaos. Following [14], we calculate the time- and ensemble-averaged principal Lyapunov exponent for the trajectories of cells in the given flow. To be specific, the cells’ initial positions are chosen randomly from a uniform distribution across the spatial domain, and their initial orientations are chosen independently from a uniform distribution on the unit sphere, the simulations are stepped forward in time until all transient effects have decayed, and then the Lyapunov exponents are evaluated for each cell, from which the mean value over all cells at each time step is calculated. This mean value is then averaged over 500 time steps. Positive values of the Lyapunov exponent indicate exponential divergence of cell trajectories and hence a chaotic flow since the domain is bounded; negative values correspond to suppression of Lagrangian chaos.

Setting \( V = 0 \) corresponds to the cells behaving as passive tracers and, as expected, \( \lambda > 0 \) because the ABC flow is chaotic, and the cells become randomly distributed throughout the computational box (see Figs. 5 and 6). We now consider what happens for increasing values of \( V \).

For all values of \( \alpha \), as \( V \) increases to \( V \approx 1 \), \( \lambda \) decreases and becomes negative (Figs. 5 and 6), showing that gyrotaxis suppresses Lagrangian chaos as plume formation begins to dominate. The patterns formed by the plumes are illustrated in Tables III and IV. A helical plume forms along the central axis of the box, with a secondary plume at the periodic side boundary for small non-zero values of \( V \) when \( \alpha = 1 \) (see e.g. Fig. 7), together with other secondary features.

The numerical results show that the trajectories in

\[
\begin{align*}
\alpha = 0, G = 1 & \quad \alpha = 0.5, G = 1 & \quad \alpha = 1, G = 1 \\
\end{align*}
\]

FIG. 7: Simulations of cells in the ABC flow with \( V = 0.1 \) and \( G = 1 \): (a) \( \alpha = 0 \), (b) \( \alpha = 0.5 \) and (c) \( \alpha = 1 \).

Figure 7(a), when \( \alpha = 0 \), lie on the helix

\[
x = \pi + a \cos z, \quad y = 3\pi/2 - a \sin z
\]

of radius \( a \approx 0.35 \). As a validation of the numerical results, it is possible to derive an approximate solution for this manifold as follows. Defining new coordinates \( x' = x - \pi \) and \( y' = y - 3\pi/2 \) then, from (11), the fluid velocity and vorticity on the helix are

\[
\mathbf{u} = \mathbf{\omega} = \begin{pmatrix}
\sin z - \sin(a \sin z) \\
\cos z - \sin(a \cos z) \\
\cos(a \sin z) - \cos(a \cos z)
\end{pmatrix}
\]

\[
= (1 - a)e_\theta + (-2 + a^2/2)k + O(a^3),
\]

using cylindrical polar coordinates \( (r, \theta, z) \) with unit base vectors \( e_r, e_\theta \) and \( k \). Dropping the primes, the position of a cell on the helix is

\[
x = a e_r(\theta(t)) + z(t)k \implies \dot{x} = -a \sigma e_\theta + \sigma k,
\]

where \( z = -\theta = \sigma t \) and \( \sigma < 0 \) is a constant speed.
From (6),
\[ Vp = p_\theta e_\theta + p_z k + O(a^3), \quad \text{where} \]
\[ p_\theta = [a(1 - \sigma) - 1], \quad p_z = [2 + \sigma - a^2/2] \]
\[ \Rightarrow V\dot{p} = \sigma p_\theta e_r + O(a^3). \]  
(24)

Noting that
\[ \omega \wedge Vp = \sigma(1 - 3a)e_r + O(a^3) \]
and
\[ k - (k \cdot p)p = -p_z p_r e_r - p_\theta p_\theta e_\theta + (1 - p_z^2)k, \]
equating coefficients in (5) and assuming that
\[ V = \tilde{V}a^2, \quad \text{where} \quad \tilde{V} = O(1), \]
yields
\[ a = 1/3 \quad \text{and} \quad \sigma = -2 + (1 + \tilde{V})a^2 + O(a^3), \]
which is consistent with the numerical estimates from the simulations. The value of \( a \) ensures that \( \omega \wedge p = 0 \) while the value of \( \sigma \) renders \( p_z \approx 1 \). To this order, we see that the shape of the manifold is independent of \( G \), and \( p_r = p_\theta = O(a^3) \), so that the cells are swimming vertically upwards while being carried downwards in the flow on the helical trajectory.

\[ \begin{align*}
(a) & \\
(b) & \\
(c) &
\end{align*} \]

FIG. 8: Simulations of cells in the ABC flow with \( V = 9 \) and \( G = 1 \): (a) \( \alpha = 0 \), (b) \( \alpha = 0.5 \) and (c) \( \alpha = 1 \).

When \( V \) is increased further and the swimming cells move more rapidly through the flow field, \( \lambda \) increases and becomes positive as a second window of chaotic trajectories is encountered, provided \( \alpha < 1 \). Figure 9 illustrates that the window exists for a wide range of values of \( G \geq 1 \). Finally as \( V \to \infty \), the cells move so quickly that plumes form once again and \( \lambda \) decreases through zero as chaotic trajectories are suppressed. Patterns transition from the second chaotic window through complicated braided patterns that are spatially periodic in the vertical direction with wavelengths that are multiples of \( 2\pi \), the height of the computational box. Examples are described in more detail below. At very large values of \( V \), single helical plumes are found in the same region of the box as when \( 0 < V < 1 \), but now they have the opposite sense and cells are no longer carried downwards in the local flow (see e.g. Fig. 8).

The case \( \alpha = 1 \) (rod-like cells) appears to be singular, in that there is no second window of chaos and plumes form in different quadrants of the computational domain. In the Discussion below, we give evidence that in this limit, orientation by gyrotaxis and by the local rate-of-strain in the flow exactly cancel along plume axes.

For parameter ranges which produce organised structures, cell trajectories are periodic; this is illustrated in Figure 10 where the evolution as swimming speed increases is examined. For cells in a downwelling plume,
where the downward motion is entirely suppressed and only an upwelling plume is formed. The region between the dashed black lines in Figure 10 indicate swimming speeds where both down and/or upwelling plumes form in different regions of the same flow, as the parameter $V$ is varied.

The outcome of simulations for $V = 4$ and $G = 4$ are seen in Figure 11. Views projected onto different planes when $\alpha = 2/3$ are seen in Figure 12. This gives one of the most intricate structures observed, and again features a plume on a one-dimensional manifold. Tracking orientation, it was observed that

$$-0.5 < p_x < 0.5, \quad -0.5 < p_y < 0.5, \quad 0.88 < p_z < 1.$$  

Solutions were found to be periodic, with period equal to 19.03, during which time each cell travels along all of the nine trajectories in the periodic box. Alternatively, one may think of this as a single trajectory extending through nine vertically connected boxes. Figure 13 shows how $p_x$ varies during this periodic circuit. Decomposition into Fourier series reveals that the dominant frequency is related to the number of times a cell crosses the box in the $z$-direction.

To study the movement of cells relative to each other, the nearest neighbour distance was calculated after aggregation had occurred and then tracked over a long time period. The results for several different structures are seen in Figure 14. Comparing the results for random arrangements to those which have formed plumes provides clear evidence that the formation of these structures has led to a suppression of the Lagrangian chaos.
Figure 11 also illustrates the importance of cell shape on trajectories. When \( \alpha = 0 \), a simple plume forms whereas taking \( \alpha = 2/3 \) results in a more complex braided structure. Closer investigation revealed that braiding only occurs for \( 0.66 \lesssim \alpha \lesssim 0.72 \), showing how small the window for this behaviour to occur is. When \( \alpha = 1 \) a single plume, at the periphery of the box is seen.

![Figure 14: Distance between cells and their nearest neighbour, averaged over all cells.](image)

The full range of parameter values that have been used in simulations of the ABC flow and the steady state patterns that are given in Tables III, IV, V and VI and in the Supplementary Material [26].

To test the robustness of the most intricate structures, the effect of adding small perturbations (noise) to cell orientation was examined. To do this [8] was altered to

\[
\dot{p} = \begin{pmatrix}
\frac{\omega y p_z - \omega z p_y + \alpha f_z + \eta \sqrt{1 - \xi^2} \cos \theta}{2G} \\
\frac{\omega x p_z - \omega z p_x + \alpha f_y + \eta \sqrt{1 - \xi^2} \sin \theta}{2G} \\
-\frac{1 - p_x^2}{2G} + \frac{\alpha f_z}{2} + \frac{\eta \xi}{2}
\end{pmatrix}
\]

Taking \( \theta \) and \( \xi \) to be uniformly distributed such that \( \theta \in [0, 2\pi] \) and \( \xi \in [-1, 1] \) adds an additional noise term, uniformly distributed on \( S^2 \) with amplitude \( \eta \). Figure 15 shows four Poincaré sections from the plane, \( z = 0 \), with different values of \( \eta \) when \( V = G = 4 \) and \( \alpha = 2/3 \). When \( \eta = 0 \) there exist nine distinct crossing points which correspond to the nine braids. However as the level of noise increases so does the number of crossing points. This results in trajectories crossing each other with regularity meaning the new arrangement is lacking the previous intricate braiding. For parameter ranges which produce single plumes, adding noise revealed that cell aggregation is less focused than before. This causes patterns to be less defined although this does not impede the inherent similarities to structures without noise. The same behaviour was seen whether noise was added at \( t = 0 \) or when the structures had formed.

![Figure 15: Poincaré sections in the \( z = 0 \) plane with \( V = G = 4 \) and \( \alpha = 2/3 \), showing the effect of noise on cell trajectories when (a) \( \eta = 0 \), (b) \( \eta = 0.0001 \), (c) \( \eta = 0.001 \) and (d) \( \eta = 0.01 \).](image)

V. DISCUSSION

Some insight into the role of the shape of the microorganisms and the competition between the vorticity and rate-of-strain terms in the gyrotactic orientation equation [5] can be gained by considering how gyrotaxis varies with \( \alpha \) in an axisymmetric orientation equation [8] becomes

\[
p = \frac{1}{2G} \begin{pmatrix}
-p_z p_x \\
-p_0 p_z \\
1 - p_z^2
\end{pmatrix} - \frac{w'(r)}{2} \left( p_z - \alpha p_z (1 - 2p_z^2) \right) \begin{pmatrix}
-2p_r p_0 p_z \\
-p_r - \alpha p_r (1 - 2p_r^2)
\end{pmatrix}.
\]

(27)
where \( w'(r) = 2r \). The only stable steady solution when \( 0 < \alpha \leq 1 \) is

\[
p_r = \frac{\sqrt{1 - 8\Gamma^2(r)\alpha(1 - \alpha) - 1}}{4\Gamma(r)\alpha}, \quad p_\theta = 0 \quad \text{and} \quad p_z = \sqrt{1 - p_r^2}, \tag{28}\]

where \( \Gamma(r) = Gw'(r) \). In this solution \( p_r \leq 0 \) always and is subject to the additional constraint that

\[
p_r \geq -1 \iff \begin{cases} \Gamma(1 + \alpha) \leq 1, & \text{for } 0 < \alpha \leq 1/3, \\ \forall \Gamma, & \text{for } 1/3 < \alpha \leq 1, \end{cases} \tag{29}\]

since \( \mathbf{p} \) is a unit vector by definition. Moreover, the solution is linearly stable only when \( 8\Gamma^2\alpha(1 - \alpha) < 1 \). Combining the two constraints shows that the linearly stable solution exists

\[
\iff \begin{cases} \Gamma(1 + \alpha) \leq 1, & \text{for } 0 < \alpha \leq 1/3, \\ 8\Gamma^2\alpha(1 - \alpha) < 1, & \text{for } 1/3 < \alpha \leq 1. \end{cases} \tag{30}\]

For the special case \( \alpha = 0 \),

\[
p_r = -\Gamma(r), \quad p_\theta = 0, \quad p_z = \sqrt{1 - \Gamma^2(r)}, \tag{31}\]

which exists and is linearly stable when \( 0 \leq \Gamma < 1 \). Thus we see that \( p_r \to 0 \) as \( \alpha \to 1 \), i.e. the torques due to local vorticity and rate-of-strain cancel each other out for rod-like cells, so that the cells swim vertically upwards, even though they are gyrotactic. For details of the other steady but unstable solutions, see [27].

In fact, it is readily shown (although to the best of our knowledge not previously reported) that, for any steady parallel vertical flow of the form \( \mathbf{u} = w(x, y) \mathbf{k}, \) \( \mathbf{p} = \mathbf{k} \) everywhere is an exact solution of the gyrotactic orientation equation (5) when \( \alpha = 1 \). This shows that rod-like gyrotactic cells focus less well into downwelling flows than more spherical cells, leading to a broadening of cell concentration profiles. In particular when \( \alpha = 1 \), cells are much less influenced by the ambient flow and tend to swim up at all gyrotaxis numbers which explains why Lagrangian chaos in the ABC flow is suppressed for all swimming speeds above a critical value as in Figs. 5 and 9. The limit \( G \to \infty \) i.e. no gyrotaxis (freely-rotating cells) is, of course, singular in that there is no preferred orientation. The results of simulations (see Supplementary Material [26]) for \( G = \infty \) in the ABC flow for \( \alpha = 0, 0.5 \) and 1.0, and \( 0.1 \leq V \leq 9.0 \) show that in all cases the Lyapunov exponents are positive, so shape alone cannot inhibit mixing in this flow.

We note that the computational results of Bearon et al. [25] for 2D particle orientation and trajectories in 2D Poiseuille flow in a channel are consistent with this result, in that the equilibrium cell concentration broadens as \( \alpha \) increases.

VI. CONCLUSIONS

Our numerical results, supported by analytical solutions, reveal that gyrotactic cells in incompressible three-dimensional flows can form organised one- or two-dimensional structures for swimming speeds both faster and slower than the ambient flow. These features are not found in the absence of gyrotaxis (see Section 4 in the Supplementary Material [26]). The two flows used here are spatially periodic solutions of the Euler equations. In the TGV flow, trajectories collapse (‘focus’) onto a one-dimensional manifold, except when the cells are spherical in which case two-dimensional manifolds are also found.

It is of particular interest to note that collapse to a one- or two-dimensional manifold suppresses mixing in the ABC flow, which shape alone cannot do, and complex braids that extend over many spatial periods are seen. Two windows of chaotic trajectories are found as the relative swimming speed \( V \) is increased: the first found when cells do not swim vanishes as \( V \) increases towards one, while the second occurs over a range of values of \( V > 1 \). Trapping in islands, which is a feature of 2D flows [7,9], is not seen in the 3D ABC flow. Furthermore, these results are robust, in that the structures persist in the presence of noise, suggesting that both the inhibition of Lagrangian chaos by gyrotactic focussing and the re-emergence of chaos may be biologically relevant in taking advantage of the local flow environment for aggregation or mixing, outwith carefully-controlled laboratory environments.

A further observation is that (as noted in the discussion of Figure 7) robust structures form in the ABC flow for values of \( V \approx 0.01 \) which is some five times smaller than required in the TGV flow. Whether this is specific to these two flows, or is a more general result that the presence of Lagrangian chaos enhances the formation of structures merits further study beyond the scope of this paper.

VII. ACKNOWLEDGEMENTS

This research was funded by an EPSRC PhD scholarship awarded to SIHR.
### TABLE I

**TGV flow for values of $V < 1$, when $G = 1$**

<table>
<thead>
<tr>
<th>$\alpha$ = 1</th>
<th>Similar to $V = 0.4$</th>
</tr>
</thead>
<tbody>
<tr>
<td>$\alpha$ = 0.5</td>
<td>Similar to $V = 0.9$</td>
</tr>
<tr>
<td>$\alpha$ = 0</td>
<td>Similar to $V = 0.2$</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>$V$ = 0.1</th>
<th>$V$ = 0.2</th>
<th>$V$ = 0.3</th>
<th>$V$ = 0.4</th>
<th>$V$ = 0.5</th>
<th>$V$ = 0.6</th>
<th>$V$ = 0.7</th>
<th>$V$ = 0.8</th>
<th>$V$ = 0.9</th>
</tr>
</thead>
</table>

### TABLE II

**TGV flow for values of $V \geq 1$, when $G = 1$.**

<table>
<thead>
<tr>
<th>$\alpha$ = 1</th>
<th>Similar to $V = 5$</th>
</tr>
</thead>
<tbody>
<tr>
<td>$\alpha$ = 0.5</td>
<td>Similar to $V = 5$</td>
</tr>
<tr>
<td>$\alpha$ = 0</td>
<td>Similar to $V = 6$</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>$V$ = 1</th>
<th>$V$ = 2</th>
<th>$V$ = 3</th>
<th>$V$ = 4</th>
<th>$V$ = 5</th>
<th>$V$ = 6</th>
<th>$V$ = 7</th>
<th>$V$ = 8</th>
<th>$V$ = 9</th>
</tr>
</thead>
</table>

### TABLE III

**ABC flow projected onto the $z = 0$ plane for $V < 1$, when $G = 1$.**

<table>
<thead>
<tr>
<th>$\alpha$ = 1</th>
<th>Similar to $V = 0.4$</th>
</tr>
</thead>
<tbody>
<tr>
<td>$\alpha$ = 0.5</td>
<td>Similar to $V = 0.2$</td>
</tr>
<tr>
<td>$\alpha$ = 0</td>
<td>Similar to $V = 0.2$</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>$V$ = 0.1</th>
<th>$V$ = 0.2</th>
<th>$V$ = 0.3</th>
<th>$V$ = 0.4</th>
<th>$V$ = 0.5</th>
<th>$V$ = 0.6</th>
<th>$V$ = 0.7</th>
<th>$V$ = 0.8</th>
<th>$V$ = 0.9</th>
</tr>
</thead>
</table>
### TABLE IV

ABC flow projected onto the $z = 0$ plane for values of $V \in [1, 9]$, when $G = 1$.

<table>
<thead>
<tr>
<th>$\alpha = 1$</th>
<th>Similar to $V = 2$</th>
</tr>
</thead>
<tbody>
<tr>
<td>$\alpha = 0.5$</td>
<td>Similar to $V = 5$</td>
</tr>
<tr>
<td>$\alpha = 0$</td>
<td>$V = 1$ $V = 2$ $V = 3$ $V = 4$ $V = 5$ $V = 6$ $V = 7$ $V = 8$ $V = 9$</td>
</tr>
</tbody>
</table>

### TABLE V

ABC flow for values of $V \in [1, 9]$, when $G = 4$.

<table>
<thead>
<tr>
<th>$\alpha = 1$</th>
<th>Similar to $V = 3$</th>
</tr>
</thead>
<tbody>
<tr>
<td>$\alpha = 2/3$</td>
<td>Similar to $V = 6$</td>
</tr>
<tr>
<td>$\alpha = 0$</td>
<td>$V = 1$ $V = 2$ $V = 3$ $V = 4$ $V = 5$ $V = 6$ $V = 7$ $V = 8$ $V = 9$</td>
</tr>
</tbody>
</table>

### TABLE VI

ABC flow for values of $G \in [1, 9]$, when $V = 4$.

<table>
<thead>
<tr>
<th>$\alpha = 1$</th>
<th>Similar to $G = 1$</th>
</tr>
</thead>
<tbody>
<tr>
<td>$\alpha = 2/3$</td>
<td>Similar to $G = 4$</td>
</tr>
<tr>
<td>$\alpha = 0$</td>
<td>$G = 1$ $G = 2$ $G = 3$ $G = 4$ $G = 5$ $G = 6$ $G = 7$ $G = 8$ $G = 9$</td>
</tr>
</tbody>
</table>

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[26] See supplemental material at .. for the full range of parameter values that have been used in simulations of the abc and tgv flows.
