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1	Gyrotaxis in a steady vortical flow
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12	We show that gyrotactic motility within a steady vortical flow leads to tightly
13	clustered aggregations of microorganisms. Two dimensionless numbers,
14	characterizing the relative swimming speed and stability against overturning by
15	vorticity, govern the coupling between motility and flow. Exploration of parameter
16	space reveals a striking array of patchiness regimes. Aggregations are found to form
17	within a few overturning timescales, suggesting that vortical flows might be capable
18	of efficiently separating species with different motility characteristics.
19	
20	
21	Spatial heterogeneity, or 'patchiness', in the distribution of organisms affects important
22	ecological processes, including competition, predation, the spread of epidemics, and the
23 24	maintenance of species diversity [1]. We report on a biophysical mechanism that rapidly
24 25	implications for marine phytoplankton. These unicellular, photosynthetic organisms are
26	responsible for half of the world's oxygen production [2] and represent the base of the
27	oceans' food web [3] Patchiness in the distribution of phytoplankton is strongly coupled
28	to ecosystem productivity [4] and has been found to extend down to centimeter scale [5-
29	9].
30	
31	Active locomotion is used by many organisms to achieve and maintain advantageous
32	positions with respect to resources, predators, and each other, thereby conferring
33	enhanced fitness [10]. Although many marine microorganisms are motile, their motility is
34	often neglected because swimming speeds are typically smaller compared to ambient
35	flow speeds. Using a well-established flow model, we show that a coupling between
36	motility and vortical fluid motion can drive aggregations of gyrotactic cells, with a rich
3/ 20	diversity of steady-state cell distributions.
20 20	Motile phytoplankton often swim in a preferred direction \mathbf{k} (typically vertical to perform
40	daily migration through the water column) owing to a stabilizing torque that can arise
41	from an asymmetry in shape [11] or body density [12] or the ability to sense the
42	direction of gravity [13]. In moving fluids, cells further experience rotation due to
43	gradients in velocity and cells are said to be gyrotactic [12] (Fig. 1(a)). Modeling cells as
44	prolate ellipsoids, their swimming direction, p , is governed by [14]
45	
46	$\frac{d\mathbf{p}}{dt^*} = \frac{1}{2B} \left[\mathbf{k} - (\mathbf{k} \cdot \mathbf{p}) \mathbf{p} \right] + \frac{1}{2} \boldsymbol{\omega}^* \times \mathbf{p} + \alpha \mathbf{p} \cdot \mathbf{E}^* \cdot \left[\mathbf{I} - \mathbf{p} \mathbf{p} \right]. $ (1)

- 47 Starred quantities indicate dimensional variables: $\boldsymbol{\omega}^*$ is the fluid vorticity, \mathbf{E}^* is the rate of
- 48 strain tensor, **I** is the identity matrix, t^* is time, *B* is the characteristic time a perturbed
- 49 cell takes to return to orientation **k** if $\boldsymbol{\omega}^* = 0$, and $\boldsymbol{\alpha} = (\gamma^2 1)/(\gamma^2 + 1)$, where γ is the ratio
- 50 of the cell's major to minor axes. For phytoplankton, $B \sim 1 10$ s, with the uncertainty
- stemming from the paucity of data [15-17]. When there is no preferred swimming
- 52 direction $(B^{-1} = 0)$ Jeffery orbits [18] are recovered. Equation 1 applies to organisms
- 53 much smaller than the scale of ambient velocity gradients, which allows cells to be
- 54 modeled as point particles.
- 55 The study of particle motion in vortical flows has a rich history, partly due to its
- 56 importance in marine [19] and atmospheric [20] processes. Though many models of
- 57 vortical flow exist, the Taylor-Green vortex flow (TGV; [21]) has been widely used [19,
- 58 20, 22], largely because of its tractability. The TGV flow is a two-dimensional array of
- steady, counter-rotating vortices (Fig. 1(b)), with spacing L and maximum vorticity ω_0 at
- 60 the center of vortices. The nondimensional velocity $\mathbf{u} = [u, 0, w]$ and vorticity $\mathbf{\omega} = [0, \omega, w]$
- 61 0] fields are given by $u = -\frac{1}{2} \cos x \sin z$, $w = \frac{1}{2} \sin x \cos z$, and $\omega = -\cos x \cos z$, where
- 62 lengths, velocities and vorticities are non-dimensionalized by 1/m, ω_0/m and ω_0 ,
- 63 respectively, and $m = 2\pi/L$.
- 64 To determine how populations of gyrotactic cells might respond to vortical flows, we
- 65 computed the trajectories of individual gyrotactic organisms swimming at constant speed
- 66 $V_{\rm C}$ within a TGV flow (phytoplankton can swim at up to $V_{\rm C} = 3 \text{ mm s}^{-1}$ [7]). The
- 67 nondimensional equations of motion for a cell are then

68
$$\frac{d\mathbf{p}}{dt} = \frac{1}{2\Psi} \left[\mathbf{k} - (\mathbf{k} \cdot \mathbf{p})\mathbf{p} \right] + \frac{1}{2} \boldsymbol{\omega} (\mathbf{X}) \times \mathbf{p} + \alpha \mathbf{p} \cdot \mathbf{E} (\mathbf{X}) \cdot \left[\mathbf{I} - \mathbf{p} \mathbf{p} \right], \quad (2)$$

69
$$\frac{d\mathbf{X}}{dt} = \Phi \mathbf{p} + \mathbf{u}(\mathbf{X}), \qquad (3)$$

where $\mathbf{X} = [x, y, z], \Psi = B\omega_0, \Phi = V_C m/\omega_0$, and time was non-dimensionalized by $1/\omega_0$. We neglected the effect of cells on flow.

- 72 We first considered spherical cells ($\alpha = 0$) swimming within a vertical plane (x-z), for
- 73 which equation 2 becomes $d\theta/dt = -\frac{1}{2}(\cos x \cos z + \sin \theta/\Psi)$ [12], where θ is the swimming
- 74 direction relative to the vertical (Fig. 1(a)). With these assumptions, the two parameters,
- 75 Φ and Ψ , fully control the fate of the cells. Φ measures the swimming speed relative to
- the flow speed and Ψ is a measure of orientational stability; if $\omega \Psi > 1$ the cell can be
- 77 overturned by vorticity [17] (red circles, Fig. 1(c)).
- 78 We find that the spatial distribution of gyrotactic cells in vortical flow is highly
- dependent upon Ψ and Φ . We begin by comparing trajectories of three cells with
- 80 different Ψ and Φ parameters, all initialized with the same orientation and position (Fig.
- 81 1(b)). The slow, intermediately stable red cell ($\Phi = 0.2, \Psi = 1$) spirals inwards towards a
- single point, the fast and stable green cell ($\Phi = 20, \Psi = 0.1$) rapidly finds an upward path,
- 83 whereas the slow and unstable blue cell ($\Phi = 0.5, \Psi = 100$) wanders aimlessly. These
- 84 strikingly different behaviors highlight the complex interaction between motility and flow

and suggest the existence of multiple regimes of phytoplankton aggregation in vorticalflows.

87	A systematic exploration of Φ - Ψ parameter space revealed ten distinct, time-invariant
88	patchiness regimes (Fig. 2; at $t = 2000$). The strongest aggregation occurs when all cells
89	converge to points where the equilibrium cell orientation is such that motility exactly
90	balances flow $(d\theta/dt = dx/dt = dz/dt = 0;$ Figs. 1(c), 2(b,c); Movie 1 [23]). This can occur
91	at either a single point $(x = \pi/2, z = \cos^{-1}(-2\Phi); \text{ Fig. 2(b)})$ or two points $(x = \cos^{-1}(\pm\Gamma^{1/2}), z)$
92	= $\tan^{-1}(-2\Psi\Phi)$, $\Gamma = (16\Psi^2\Phi^4 + 4\Phi^2 - 1)/(4\Psi^2\Phi^2 - 1)$; Figs. 1(c), 2(c)) within each vortex.
93	
94	Gyrotactic cells are known to collect in downwelling regions ($w < 0$) and retreat from
95	upwelling regions ($w>0$) [12], a mechanism that was suggested to produce accumulation
96	in turbulent flows [24]. We recover accumulation in downwelling regions in the 'vertical
97	migrator' regime (Fig. 2(d)), in which cells focus into vertical bands between vortices
98	and swim upwards ($x = \pm \pi/2$, $\theta = 0$; Movie 2 [23]). Though these cells traverse both
99	upwelling and downwelling regions, convergence prevails because cells spend more time
100	in regions where swimming and flow oppose one another.
101	
102	in contrast with earlier predictions [24], accumulation in downweiling regions is only one
103	of many possible patterns of aggregation: a multitude of patterns arise in $\Psi - \Psi$ parameter
104	space (Fig. 2). Unstable cells (Ψ >1) are more susceptible to being rotated by vorticity.
105	Slow unstable cells (Φ <0.3) are unable to escape vortices, leading to closed trajectories
106	(Fig. 2(e); Movie 3 [23]). In contrast, fast unstable cells (Φ >0.3) are locally reoriented by
107	vorticity, but can escape from vortices. They weave from one vortex to the other,
108	producing diverse patterns (Fig. 2(g, 1, j); Movie 4 [23]), including some peculiar figure
109	eights (Fig. 2(h,k)). Finally, very fast unstable cells (Φ >2) have little time to be deflected
110	by vorticity and can move diagonally in addition to vertically upwards (Fig. 2(f)).
111	Although for slow swimmers (Φ <1) there exist regimes where accumulation patterns did
112	not emerge (Fig. 2(m)) or converge (Fig. 2(1)) by $t = 2000$, the diversity of accumulation
113	patterns and their occurrence over a wide range of parameter space indicate that strong
114 115	patchiness of gyrotactic cells is the norm within vortical flows, rather than the exception.
115 116	In addition to producing notabinage vertical flow can stifle vertical migration. This affect
110	In addition to producing paterimess, voltical now can strike vertical migration. This effect can be quantified using the normalized vartical migration rate $W = \frac{1}{\sqrt{2}} \frac{1}{\sqrt{2}} \frac{1}{\sqrt{2}}$
11/ 110	can be quantified using the normalized vertical inigration rate, $W = \langle dz/dt \rangle / \Phi$, defined as the net unward speed of a cell averaged over all cells and over time (t = 0, 10)
110	the net upward speed of a cell averaged over all cells and over time $(l - 0 - 10)$, normalized by Φ (Eig. 2(a)). The upward measurement of stable cells ($W(z)$) is largely.
119	normalized by Ψ (Fig. 5(c)). The upward movement of stable cells ($\Psi < 1$) is largely
120	unaffected by flow ($W \sim 1$). In contrast, vertical migration of unstable cells ($\Psi > 1$) is severally impeded ($W < 1$), showing that vertical flow, can trap surplustic cells of death
121 122	The suppression of vertical migration is in line with both a simple scaling analysis [25]
122	and simulations utilizing more complex flow fields [26]
122	and simulations utilizing more complex now news [20].
125 125	To quantify natchiness, we partitioned the domain into a 15×15 grid of boxes and
120	To quantify pateriness, we partitioned the domain into a 13×13 grid of boxes and

126 computed the box occupancy function, f(n) [27], where *n* is the number of cells in a box

127 (with mean λ). As cells accumulate in some boxes and leave others empty, the standard

deviation of f(n), σ , increases relative to its initial (Poisson) value, $\sigma_{\rm P} (=\lambda^{1/2})$. Thus, the accumulation index $N = (\sigma - \sigma_{\rm P})/\lambda$ is a measure of patchiness [27]. Fig 3(a) shows N in

- 130 $\Phi-\Psi$ space at t = 10. Cells with motility faster than the flow ($\Phi > 0.5$) and intermediate 131 stability (Ψ ~1) exhibit marked patchiness by t = 10, hence accumulation by this 132 mechanism can be rapid (within a few vortex time scales). Cells that accumulate the most 133 swiftly belong primarily to the 'vertical migrator', 'equilibrium', and 'skater' regimes 134 (Fig. 2). This is also observed by computing the time, τ required for a randomly 135 distributed population to reach a time-invariant spatial distribution. The latter was calculated by fitting N(t) with the exponential $\kappa (1 - e^{-t/\tau})$, where κ is a constant. The 136 same region of parameter space ($\Phi > 0.5$, $\Psi \sim 1$) exhibits the fastest accumulation (Fig. 137 138 3(b)). These findings are readily rationalized: to accumulate, cells must swim across 139 streamlines. Fast swimmers are able to make significant progress across streamlines, 140 while intermediate stability represents a trade-off between persistent tumbling ($\Psi >>1$), 141 which negates directed swimming, and excessive stability ($\Psi << 1$), which prevents cell 142 orientation from being perturbed by the flow.
- 143

144 These findings assume that the fluid vorticity is orthogonal to the preferred swimming

145 direction, k. To determine the effect of vortex orientation, we performed three-146 dimensional (3D) simulations for spherical cells ($\alpha = 0$) by extruding the TGV flow in 147 the y-direction and allowing \mathbf{k} to assume any orientation, prescribed by polar and 148 azimuthal angles (η, β) . The swimming direction was computed using equations 2 and 3. 149 When $\mathbf{k} = \mathbf{z}$ ($\eta = \beta = 0$), the x-z projection of the 3D time-invariant cell distribution is 150 identical to the 2D simulation. As one varies k, additional patchiness regimes emerge 151 compared to Fig. 2. Patchiness occurs over all orientations of k, with the exception of a 152 small region about $\mathbf{k} = \mathbf{v}$ ($\eta = \beta = \pi/2$; Fig. 4(a)), where cell orientation is unaffected by

flow ($\boldsymbol{\omega} \times \mathbf{k} \approx 0$). Thus, the proposed patch generation mechanism is robust in 3D space. 154

Phytoplankton morphology is highly diverse: many species have non-spherical cell bodies [28] or flagella that alter their effective eccentricity [29]. Elongated swimming particles in TGV flow, in the absence of a preferential swimming direction ($\Psi=\infty$), have been shown to aggregate along flow separatrices [30]. We determined how elongation

159 influences the aggregation of gyrotactic cells for ten values of Ψ and Φ (Fig. 2(a), 160 symbols), for each of them varying the cell aspect ratio, γ , from 1 to 100. Cells were

161 symbols), for each of them varying the cert aspect ratio, γ , from 1 to 100. Certs were 161 confined to the x-z plane, hence equation 2 simplifies to $d\theta/dt = \frac{1}{2}(\alpha \sin x \sin z \sin 2\theta)$

162 $\sin\theta/\Psi - \cos x \cos z$). We found that including elongation further strengthens the

- 163 conclusion that gyrotactic motility in vortical flow generates patchiness. While
- 164 elongation does not affect patch topology for some values of Ψ and Φ , it produces new
- spatial aggregations for others (Fig. S1 [23]) and can generate patchiness in some low

166 stability (*i.e.* large Ψ) regions, where spherical cells remain randomly distributed.

167 Changes in patchiness caused by cell elongation were quantified by calculating $N_E - N$, 168 the difference in *N* relative to that obtained for spherical cells (Fig. 4(b)). Out of the ten

169 values of Ψ and Φ tested, only one gave $N_E \leq N$, indicating that cell elongation

170 generally enhances patchiness. A similar conclusion was previously found in the limit of

171 $\Psi = \infty$: cells with larger γ are more likely to escape vortices and aggregate along

separatrices [30].

173

174 The influence of buoyancy, inertia, and motility on the motion of particles within vortical

- 175 flows has been studied extensively [19, 20, 22, 24, 30]. Particles that can move only 176 vertically relative to the flow, for example as a result of buoyancy, correspond to $\Psi = 0$ 177 and can not generate patchiness in unbounded flows (*i.e.* N(t) = 0; [20]). Particle inertia 178 can in principle induce patchiness [20], but phytoplankton's small size and density 179 contrast (<10% denser than seawater) preclude them from aggregating via inertia in most 180 natural flows [31]. In contrast, we have shown that a simple vortical flow can trigger 181 rapid accumulation of gyrotactic phytoplankton over a broad range of dimensionless 182 parameter space, suggesting that motility might play an important role in determining the 183 spatial distribution of these microorganisms in the environment, if this mechanism proves 184 robust in turbulent flows. Partial support for this hypothesis comes from observations that 185 motile species are more likely to be aggregated at small scales than non-motile species [7, 186 9], though alternate mechanisms including chemotaxis [32] and phototaxis [33] may also 187 be responsible.
- 188

189 An additional prediction borne out of this model is that different motility characteristics 190 may drive widely different spatial cell distributions. If verified, it would imply that the 191 interaction of motility and flow may control the success of different species in processes 192 like the competition for nutrients and sexual reproduction. One may further speculate that 193 cells could actively control their spatial distribution by adjusting their position in (Φ, Ψ) 194 space (Fig. 2) to favor or prevent aggregation, by either regulating their swimming speed 195 $(i.e. \Phi)$ [34] or altering their stability $(i.e. \Psi)$ via changes in morphology [35], chloroplast 196 position [36], or flagellar stroke [29].

197 One must, however, be very cautious in extending findings from an idealized flow model 198 to realistic flows. While the steady TGV flow is often used as a crude analogue for 199 turbulence [20, 22], the latter is time-dependent, fully three-dimensional, and 200 incorporates a range of scales, including larger-scale fluid motion that can disperse 201 aggregations formed at smaller scales [37]. Therefore, in the same spirit as studies that 202 examined the motion of inertial particles in TGV flow [20], the results presented here 203 open new hypotheses that await to be tested with more realistic flow models (e.g. direct 204 numerical simulation) or in laboratory experiments.

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FIG. 1 (a) Gyrotactic microorganisms, such as the toxic marine phytoplankton

260 *Heterosigma akashiwo* shown here (diameter $\approx 14 \,\mu$ m), swim in a direction, θ , set by a

balance of torques. The torque due to the fluid (T_F) tends to rotate the cell, whereas the

torque due to cell asymmetry (T_A) – for example bottom heaviness – tends to restore the

263 cell to its preferential orientation, **k**. $V_{\rm C}$ is the swimming speed.

(b) Three cells, with different Ψ and Φ , initialized at the same location and orientation (*x* $z = z = -\pi/2$; $\theta = \pi/4$; white arrow) in a TGV flow follow very different trajectories. Trajectories correspond to (Ψ , Φ) = (0.1, 20)(green); (1, 0.2)(red); (100, 0.5)(blue). The dimensionless TGV velocity and vorticity fields are shown by arrows and by shading, respectively. The domain is doubly periodic. The v axis extends into the page

respectively. The domain is doubly periodic. The y axis extends into the page.

269 (c) The most intense cell accumulation occurs when cells converge to equilibrium points, 270 where total cell velocity $V_T = (dx/dt, dz/dt) = (0, 0)$. Shown here is the 'equilibrium

double' regime ($\Psi = 1.1, \Phi = 0.25$). White crosses are numerical predictions of the

equilibria, pink circles are analytical results. Arrows and shading show $V_{\rm T}$ and $|V_{\rm T}|$,

- 273 respectively. Red circles denote regions where vorticity can overturn cells ($\omega \Psi > 1$).
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FIG. 2 (a) Parameter space of gyrotactic swimming in TGV flow, showing different patchiness regimes. Each square represents one of 900 simulations. In each simulation the trajectories of 400 randomly initialized cells were integrated until t = 2000. Ten distinct patterns emerge (b-k), not including the cases in which accumulation does not occur (m) or has not converged (l). For the 'equilibrium' regimes (b,c), all cells reside at the equilibrium points. The symbols in (a) correspond to (Ψ , Φ) values that are analyzed in Fig. 4(b) to investigate the role of cell elongation.



302 FIG. 3 (a) The degree of patchiness at time t = 10, quantified by the accumulation index

N, for the same matrix of simulations as in Fig. 2. N = 0 corresponds to a random

304 (Poisson) distribution, while N > 0 indicates aggregation.

305 (b) The time, τ , required for cells to reach a time-invariant spatial pattern.

306 (c) The normalized vertical migration rate, W. Vertical migration is unhindered when W =307 1 and entirely suppressed by flow when W = 0.



FIG. 4 (a) Gyrotactic cells swimming within a 3D TGV flow form aggregations for almost any vortex orientation. Plotted here is the accumulation index, *N*, at t = 10 and Φ = $\Psi = 1$, as a function of the polar and azimuthal angles, η and β .

326 (b) Cell elongation (aspect ratio $\gamma > 1$) produces an increase in patchiness in most regimes,

327 compared to the case of spherical cells (γ =1). Shown is the change in the accumulation

328 index, $N_{\rm E} - N$, due to elongation, at t = 2000. Symbols and colors correspond to Φ, Ψ

329 values and aggregation regime color in Fig. 2(a), respectively. Representative cell

distributions for each case can be found in Fig. S1 [23].