Collective motion with anticipation: Flocking, spinning, and swarming

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We investigate the collective dynamics of self-propelled particles able to probe and anticipate the orientation of their neighbors. We show that a simple anticipation strategy hinders the emergence of homogeneous flocking patterns. Yet anticipation promotes two other forms of self-organization: collective spinning and swarming. In the spinning phase, all particles follow synchronous circular orbits, while in the swarming phase, the population condensates into a single compact swarm that cruises coherently without requiring any cohesive interactions. We quantitatively characterize and rationalize these phases of polar active matter and discuss potential applications to the design of swarming robots.

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I. INTRODUCTION

Over the last 20 years physicists have devoted significant efforts to elucidating the numerous dynamical patterns observed in populations of living organisms. Flocking patterns are a prominent example. They refer to the self-organization of an ensemble of motile individuals into a homogeneous group undergoing coherent directed motion. As initially suggested by Vicsek et al. [1], simple alignment interactions are sufficient to trigger the emergence of flocking patterns. Alignment-induced flocks have been demonstrated theoretically, numerically, and in synthetic experiments [2–7]. In addition, quantitative analyses of animal trajectories [8–11] support that the large-scale features of some animal flocks can be rationalized by simple behavioral rules at the individual level, including velocity-alignment interactions. However, inferring a model from actual data requires a minimal set of hypotheses on the possible form of the interactions [12] or on the statistical properties of the observables [8]. Until now, the overwhelming majority of the available models has been merely restrained to couplings between the instantaneous positions and orientations of neighboring individuals (see [10,13,14] for noteworthy exceptions). Investigating alternative dynamical rules may provide further insight into the collective dynamics of motile individuals.

Here, we generalize the conventional description of polar active matter [3]. We consider the dynamics of motile particles, which aligns their velocity with the orientation of their neighbors by anticipating their rotational motion. Naively, anticipation would be expected to yield more robust flocks. In striking contrast, we show that simple anticipation rules result in much richer collective behaviors: First, when motile particles strongly anticipate the orientational changes, the population simultaneously breaks a continuous and a discrete symmetry. The system self-organizes into a spinning state where all the particles follow closed circular orbits in a synchronized fashion. Second, a combination of alignment and moderate anticipation results in the emergence of a polar-liquid phase akin to the flocking pattern observed without anticipation. Finally, at the onset of the flocking-to-spinning transition, the population forms stable compact swarms despite the absence of any attractive couplings.

The paper is organized as follows: we first introduce the equations of motion of the interacting self-propelled particles. The alignment and anticipation rules are described. We then discuss the emergence of the three ordered phases (spinning, flocking, and swarming phases). They are quantitatively characterized, explained, and compared to the conventional patterns of polar active matter. We close this paper from an engineering perspective. We show that minimal anticipation rules can provide an effective and safe design strategy to interrupt the directed motion of a flock without shutting down the propulsion mechanism at the individual level.

II. MODEL: ALIGNMENT AND ANTICIPATION

We consider an ensemble of N self-propelled particles in a two-dimensional space. Particle i, located at \( \mathbf{r}_i(t) \), moves at a speed \( v_0 \) along the unit vector \( \hat{\mathbf{p}}_i \), which makes an angle \( \theta(\hat{\mathbf{p}}_i) \equiv \theta_i \) with the x axis:

\[
\dot{r}_i(t) = v_0 \hat{p}_i. \tag{1}
\]

The equation of motion of their instantaneous orientation defines their anticipation and alignment rules,

\[
\dot{\theta}_j(t) = -\frac{1}{\tau} (\sin[\theta_j - (\theta_j + \alpha \sigma_j)])_{j \in \Omega_i} + \sqrt{2 \eta_i} \xi_i(t), \tag{2}
\]

where \( \alpha \) is a scalar parameter and \( \sigma_j \equiv \dot{\theta}_j/|\dot{\theta}_j | \) is the sign of the angular velocity. The particles have a finite interaction radius \( R \), and \( \Omega_i \) denotes the ensemble of particles interacting with the \( i \)th particle. We henceforth refer to \( \sigma_j \) as the spin of particle \( j \). This quantity is akin to the spin variable defined in [14] up to a multiplicative factor, which is the local curvature of the particle trajectory. In Eq. (2), the angular noises \( \xi_i \) are uncorrelated Gaussian random variables of unit variance, and \( \tau \) is an orientational relaxation time. For the sake of simplicity, units are chosen so that \( \tau = 1 \) and \( R = 1 \), and the control parameters left are \( N, v_0, \alpha, \eta, \) and the system size \( L \).

Equations (1) and (2) have a simple physical meaning. When \( \alpha = 0 \), they reduce to the continuous-time version of the seminal Vicsek model [15]: the particles interact via effective torques that promote alignment with the instantaneous orientation of the neighboring particles. Note that neither...
the momentum nor the angular momentum is conserved in Eqs. (1) and (2). When $\alpha > 0$, the mean orientation of the neighbors is anticipated in a simple fashion. If the orientation of the $j$th particle rotates in the clockwise (resp. anti-clockwise) direction, an effective torque promotes the alignment along the direction $\theta_j - \alpha$ (resp. $\theta_j + \alpha$). $\alpha$ is a constant angle used to anticipate the future orientation of the neighboring particles.

In order to gain a better insight into this interaction scheme, we expand the sine functions and use elementary algebra to recast Eq. (2) as

$$\dot{\theta}_i(t) = -\cos \alpha \langle \sin(\theta_i - \theta_j) \rangle_{j \in \Omega},$$

$$- \sin \alpha \left( \sin \left[ \theta_i - \left( \theta_j + \sigma_j \frac{\pi}{2} \right) \right] \right)_{j \in \Omega},$$

$$\sqrt{2} \eta(t),$$

which is the linear superposition of two models: a continuous-time Vicsek model and an $\alpha = \frac{\pi}{2}$ model. The second term on the right-hand side of Eq. (3) promotes alignment with the acceleration of the surrounding particles. As the magnitude of the translational velocity is constant, the acceleration of the surrounding particles. As the magnitude of the angular noise increases, the acceleration of the surrounding particles. As the magnitude of the angular noise increases, the acceleration of the surrounding particles.

Aiming at describing the impact of anticipation, we have numerically validated our numerical scheme.

III. RESULTS AND DISCUSSION

A. Case $\alpha = 0$: Alignment-induced flocking

Here, we briefly recall the phenomenology of the model when $\alpha = 0$, which is equivalent to a continuous-time version of the classical Vicsek model [15]. For high noise values, the self-propelled particles undergo uncorrelated persistent random walks. The population forms an isotropic and homogeneous gas depicted in Figs. 1(a) and 1(b). Decreasing the noise value below $\eta_0 = 0.45$, the rotational symmetry of the particle orientations is spontaneously broken: a macroscopic fraction of the population propels in the same direction along straight trajectories (see Figs. 1(c) and 1(d) and Supplemental Material Movie 1 [18]). This flocking transition is quantitatively captured by the sharp variation of the order parameter $\Pi \equiv |\langle \mathbf{p}_j \rangle_i \rangle| (\text{the mean polarization}) in Fig. 1(e).

Note, however, that, unlike what is found in simulations of discrete-time Vicsek models (in the dilute limit) [16,17], we do not observe the propagation of band-shape excitations at the onset of collective motion. The reason for this discrepancy is purely technical. Based on previous numerical observations [17] using an angular noise as in Eq. (3), the emergence of localized bands is expected to occur only in extremely large systems. The relatively small size of our largest simulations ($L = 16, N = 2048$) explains why solitonic bands do not form here at the onset of collective motion.

The simulations in the case $\alpha = 0$ reproduce the salient features of the Vicsek flocking transition and, therefore, validate our numerical scheme.

B. Case $\alpha = \frac{\pi}{2}$: Anticipation-induced spinning

We now investigate the anticipation model corresponding to $\alpha = \frac{\pi}{2}$. Not surprisingly, in the high-noise regime, the
The particles randomize their orientation at the same time and subsequently self-organize again to rotate synchronously, possibly along a different direction. This long-time dynamics results in a bimodal structure of $P(\sigma(\tau))$, shown in Fig. 2(e), and \textit{a posteriori} justifies defining the spin order parameter as the modulus of the most probable value of $\Sigma(t)$.

In addition, the individual circular trajectories observed in this spinning phase are not robust: they continuously form and disrupt, and neither $\Pi$ nor $\Sigma$ converges to 1 as $\eta$ goes to 0 [Fig. 2(e)]. At any given time, a small yet finite fraction of particles does not follow synchronous circular trajectories even for vanishingly small noise. These unusual strong fluctuations in the weak-noise limit, together with the circular motion of the particle, will be better understood in the next section, upon inspecting the phase diagram of the general $\alpha$ model.

Finally, we stress that at the onset of global spinning motion, the spatial distribution of the particles is heterogeneous. The population concentrates into a large, denser region surrounded by a more dilute ensemble of spinning particles. This spontaneous breaking of the translational invariance is quantified by measuring the variations of $\delta r \equiv |\langle r_i \rangle - \langle r_i \rangle_{0}|$ as a function of the angular noise [Fig. 2(f)]. This spatial heterogeneity emerges in the absence of explicit couplings between the positional degrees of freedom, which is a very persistent feature of all the active-particle models involving any form of short-range interactions [19].

C. Case $\alpha > 0$: From flocking to spinning and swarming

We now discuss the general case, for which the anticipation angle $\alpha$ can take any positive value. The collective behavior of the population is summarized by a phase diagram that divides the $(\eta,\alpha)$ plane into three regions as shown in Fig. 3. The boundaries of the phase diagram correspond to the points where the order parameter, $\Pi$ or $\Sigma$, goes from 0 to a finite value. When $0 < \alpha < \pi/2$, for very high noise values the system forms an isotropic and homogeneous gas. Upon reducing $\eta$, the population undergoes two subsequent phase transitions toward collective motion. A flocking transition occurs at $\eta = \eta_{\text{FS}}(\alpha)$: the polarization increases sharply, while the global spin remains vanishingly small [Fig. 3(b)]. However, unlike all the Vicsek-like models with periodic boundary conditions, the global polarization does not align with one of the principal axes of the simulation box. In contrast, it makes a finite angle, $\psi$, which is set by the anticipation angle $\alpha$ [Figs. 3(c) and 3(d)].

Upon further reducing the noise amplitude below $\eta = \eta_{\text{FS}}(\alpha)$, the flock self-organizes into a spinning phase. As opposed to the case where $\alpha = \pi/2$ examined in the previous section (Sec. II B), we show in Fig. 3(b) that $\Sigma$ increases sharply and saturates to 1 together with $\Pi$, as $\eta$ goes to 0 deep in the spinning phase. The individual circular trajectories are now robust. They do not disrupt and reform intermittently; all the particles endlessly follow synchronous circular orbits [see Figs. 3(e) and 3(f)].

In the spinning phase, the curvature $\kappa$ of the particle orbits has a constant value [Fig. 3(g)]. This result can be explained by considering the zero-noise limit. When $\eta = 0$, a perfectly polarized state with all particles aligned is a solution of Eq. (3) for any value of $\alpha$, provided that the particles rotate at
the same angular velocity $\dot{\theta} = \sin \alpha$. As the particle velocity $v_0$ is a constant, the corresponding trajectories are circles of curvature $\kappa = \sin \alpha / v_0 \tau$. As shown in Fig. 3(b), this mean-field argument correctly accounts for our numerical results when $0 < \alpha < \pi / 2$.

Figure 3(a) shows that the phase behavior of the population is qualitatively different when $\alpha$ approaches and exceeds $\pi / 2$. The polar-liquid/flocking phase no longer exists above $\alpha \tau \sim 0.9(\pi / 2)$. The system undergoes a single transition from an isotropic to a spinning state. For $\alpha \tau \sim 0.9(\pi / 2)$ the transition occurs through a tricritical point, where the flocking, the spinning, and the isotropic phases coexist. In addition, above $\alpha \tau$ the spinning phase displays marked qualitative differences from the one characterized above for small values of $\alpha$: the polarization and the spin do not saturate to 1 as $\eta \to 0$ (as previously noted for $\alpha = \pi / 2$). In addition, the curvature of the trajectories deviates from the naive mean-field prediction [Fig. 3(h)].

In order to elucidate this rich phase behavior, we now investigate the linear stability of the two ordered states. Let us first consider an ensemble of particles forming a polar liquid oriented along $\theta = 0$. Far from a transition point, the angles weakly deviate from $\theta_i = 0$ and the angular dynamics reduces to

$$\dot{\theta}_i(t) \sim \sin \alpha \{\sigma_i\}_{j \in \Omega} - \cos \alpha \{\theta_i - \theta_j\}_{j \in \Omega} + \sqrt{2\eta} \xi_i. \quad (4)$$

In this polarized but nonspinning state, the individual spins undergo uncorrelated fluctuations. Therefore, for any finite interaction radius, the first term on the right-hand side acts as an additional angular noise that impedes the velocity alignment, in stark contrast with the intuitive picture that one could have about the effect of anticipation. The magnitude of this effective angular noise increases with $\alpha$, which qualitatively explains why the flocking transition occurs for smaller values of $\eta$ as $\alpha$ increases. In addition, it readily follows from Eq. (4) that the effective angular stiffness, $\cos \alpha$, is negative for $\alpha > \pi / 2$. Hence the homogeneous polar-liquid phase is linearly unstable to angular fluctuations. In agreement with the phase diagram shown in Fig. 3(a), no polar-liquid phase exists for $\alpha > \pi / 2$.

Let us now repeat the same analysis for the spinning phase. Introducing the angular fluctuations $\delta \theta_i \equiv \theta_i \pm t \sin \alpha$ deep in the ordered phase, the orientational dynamics reduces to

$$\dot{\delta \theta}_i(t) \sim - \cos \alpha (\delta \theta_i - \delta \theta_j)_{j \in \Omega} + \sqrt{2\eta} \xi_i. \quad (5)$$

Again, we find that the orientational fluctuations are stabilized for $\alpha < \pi / 2$ and amplified otherwise. This stability analysis is consistent with the strong fluctuations of the individual trajectories found for $\alpha = \pi / 2$: the spinning state is linearly unstable for large anticipation angles. The ordered spinning phase numerically found for $\alpha \gtrsim \pi / 2$ is therefore stabilized by the nonlinear nature of the spin-angle interactions and cannot be captured by a linear-stability analysis alone. The flocking and the spinning phases are separated by another dynamical state where the polar-liquid phase condenses into compact swarms undergoing coherent directed motion. This behavior is characterized and illustrated in Figs. 4(a) and 4(b), respectively (see also Supplemental Material Movie 4 [18]). This swarming phase exists in a narrow range of the noise amplitude $\eta$, typically a few percent above $\eta_{FS}$. Unlike the band-shape patterns observed in the standard Vicsek models, these compact swarms are not surrounded by a sea of isotropic particles but freely propagate in an empty space. To our knowledge, this condensation is the first evidence of
the self-organization of motile particles into compact swarms without any attractive interactions.

We close this section by posing two questions that readily arise from the above analysis. First, the nature of the transition between the four phases (gas, flocking, spinning, and swarming states) remains to be elucidated (critical, first order, or sharp crossover). Answering this question would require studying much larger systems and a systematic finite-size analysis [17], which goes beyond the scope of the present paper. Second, a natural question concerns the robustness of these phases with respect to the noise structure. We have focused here on the case of an angular noise. Whereas the present results are expected to hold for a vectorial noise as well, the impact of more complex multiplicative noises, such as that found to be relevant to locust flight, remains an open question.

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IV. CONCLUSION AND PERSPECTIVES

We conclude this paper by addressing potential applications of our findings to multiagent robotics. In this context, biomimetic strategies have always been appealing for achieving collective intelligence. However, man-made programmable units could exploit a virtually infinite number of alternative strategies to accomplish emergent tasks. In this paper, we have demonstrated that simple artificial interaction rules to devise functional robotic swarms.

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