

# Giant Number Fluctuations and Energy Spectra of 3D Bacterial Suspensions

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Giant number fluctuations, a landmark of collectively moving active particles, is universal in active systems across multiple length scales. Here, we present the first experimental study on the giant number fluctuation in 3-dimensional bacterial suspensions. Our measurements show that the number fluctuation scaled by the square root of mean number  $\Delta N/\sqrt{N}$  scales with  $N^{0.32}$  at high concentrations, confirming the theoretical predictions. Near the phase boundary, we observe a simultaneous increase of the scaling exponent and the flow induced by bacterial motions, suggesting a strong interplay between giant number fluctuations and flow. We show that this interplay spans all length scales in an active turbulence, by analyzing the kinetic energy spectra.

## I. INTRODUCTION

Active fluids exhibit many unusual behaviors beyond the expectation of equilibrium statistical mechanics [1–5]. In particular, an active fluid can exhibit anomalously large density variations, the so-called giant number fluctuations (GNF), where the standard deviation of the number of particles grows nonlinearly with the square root of the mean particle number, defying the central limit theorem of equilibrium systems [6]. Such unusual density fluctuations have been observed in a wide range of active fluids in both living and non-living systems including vibrated granular rods [7–10], swarming bacteria [11, 12] and mammalian cells [13], self-propelled cytoskeleton [14], and synthetic colloidal swimmers [15, 16]. As a result, GNF has generally been viewed as a hallmark of the emergent behaviors of active fluids.

Significant theoretical and computational advancement on GNF has been made over the past two decades since the seminal works of Toner and Tu [17–29]. Meanwhile, many important theoretical and numerical predictions have been tested in experimental realizations [7, 8, 10–15]. Of particular interest is the scaling exponent  $\alpha$ , which is defined following  $\Delta N/\sqrt{N} \propto N^\alpha$ , where  $\Delta N$  is the standard deviation of particle number and  $N$  is the mean number of particles in a subsystem of given size. Heretofore, all the existing experiments on GNF were limited to two-dimensional (2D) or quasi-2D systems. In contrast to theoretical predictions (where  $\alpha \approx 0.3$ ),  $\alpha$ 's obtained in these experiments show large variations ranging from 0.13 to 0.5. Such a large variation arises partially because of complicated particle-boundary interactions, which are hard to incorporate in theoretical studies. Moreover, the predictions of  $\alpha$  in three-dimensional (3D) wet active fluids—one of the most important classes of active fluids where hydrodynamics dominate the interparticle interactions and conserve the total momentum of systems [3]—has not been experimentally testified. Therefore, there is an imperative need for

an experimental measurement of  $\alpha$  in 3D active fluids, which are not affected by system boundaries. Such a measurement will provide not only an unambiguous experimental benchmark to testify theories of active fluids, but also experimental support on the effect of dimensionality on GNF of active fluids [3].

The rise of GNF in active fluids is usually accompanied by the transition to ordered phases with collective motions [1, 3]. For wet 3D active fluids such as bacterial suspensions, these collective motions lead to large-scale coherent flows with intermittent vortices and jets, which are often referred to as active turbulence [30–39]. Similar to GNF that manifests density fluctuations across different scales, the flow of active turbulence also exhibits scale-dependent structures. Imported from the study of classical turbulence, energy spectra are frequently measured to quantify such structures in active turbulence [31, 32, 37, 40–43]. Although both GNF and energy spectra quantify the scale-dependent dynamics of active fluids, the deep connection between these two quantities at different scales has not been experimentally investigated. Our experiment reveals the density- and scale-independent coupling between GNF and energy spectra, suggesting that GNF is governed by the strength of the motion-induced flow.

## II. EXPERIMENT

Here, we present our systematic experimental study of GNF and energy spectra in bulk bacterial suspensions, a premier example of 3D wet active fluids. We use genetically engineered light-powered *Escherichia coli* (*E. coli*) as our model bacteria [44]. For a typical experiment, a bacterial suspension of control volume fraction  $\phi$  is injected into a sealed chamber of  $20\text{ mm} \times 3\text{ mm} \times 140\text{ }\mu\text{m}$ . Without supply of oxygen, bacteria quickly consume all the remaining oxygen in the chamber and stop moving after 5 minutes. We then illuminate the suspension with a high-intensity microscope light, which powers bacteria at their maximal swimming speed of  $15 \pm 3\text{ }\mu\text{m/s}$  in the dilute limit. A video of the suspension is taken  $50\text{ }\mu\text{m}$  above the bottom wall of the chamber by a bright-field

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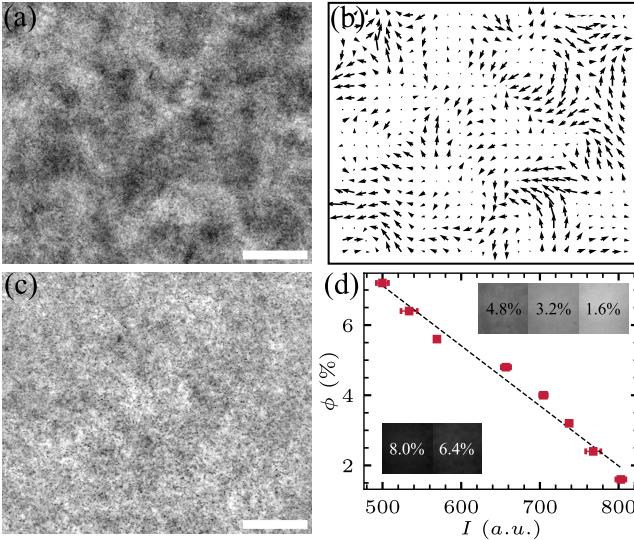


FIG. 1. (a) Active turbulence in a dense bacterial suspension (6.4%) displaying constantly varying concentration inhomogeneity. Scale bar is 85  $\mu\text{m}$ . (b) Velocity field of the active turbulence in the dense bacterial suspension ( $\phi = 6.4\%$ ) shown in (a). (c) A dilute bacterial suspension (1.6%) showing weaker density variations. (d) Volume fraction as a function of averaged pixel intensities. Inset shows bacterial suspensions of different volume fractions under the same illumination conditions.

inverted microscope at a frame rate of 30 fps and the field of view of  $420 \times 360 \mu\text{m}^2$  (Fig. 1a). We use a standard Particle Image Velocimetry (PIV) algorithm [45] to extract the 2D in-plane velocity field ( $v_x, v_y$ ) in the 3D suspension, which exhibits the characteristic chaotically-moving vortices and jets of active turbulence at high  $\phi$  (Fig. 1b).

It is challenging to directly count the number of bacteria in a 3D suspension of fast moving bacteria. Luckily, by virtue of Beer-Lambert law, the local bacterial density is monotonically correlated with the local intensity of microscope images, where darker regions correspond to higher bacterial densities (Fig. 1a and Supplementary Video). Similarly principles also appeared in other experimental works, where optical information was exploited in probing the dynamics in suspensions of bacteria and actin filaments [14, 46, 47].

To calibrate the density-intensity correlation, we prepare bacterial suspensions of different  $\phi$  and image the suspensions under the same illumination (Fig. 1c inset). The mean image intensity decreases with increasing  $\phi$  following an approximately linear relation (Fig. 1c), agreeing with the the Beer-Lambert law for samples of small thickness and weak absorptivity appropriate for our experiments. The linear density-intensity relation has also been used in previous experiments on *E. coli* suspensions [47].

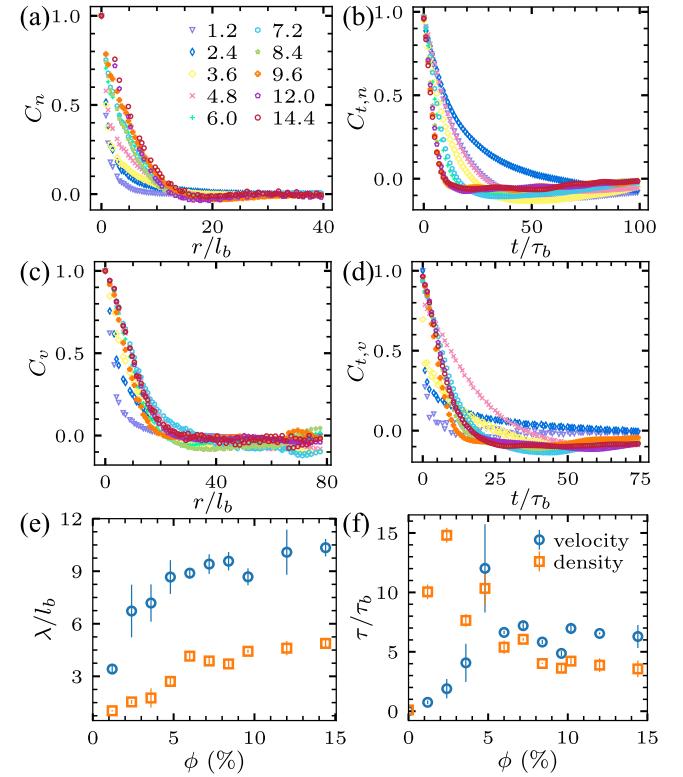


FIG. 2. **Spatial and temporal correlations of density and velocity.** (a) and (b) are density spatial correlation functions and autocorrelation functions. (c) and (d) are velocity spatial correlation functions and autocorrelation functions. (e) Correlation lengths of both density and velocity correlation functions, defined as the distance  $\lambda$  where  $C_{n(v)}(\lambda) = 1/e$ . (f) Autocorrelation times of both density and velocity autocorrelation functions, defined as the time  $\tau$  where  $C_{n(v)}(\tau) = 1/e$ . The legends in (a) indicates the volume fractions (%) of each curve, and apply to figures (a-d).

### III. RESULTS

#### A. Density fluctuations

The simple linear relation allows us to measure the spatiotemporal evolution of relative local bacterial densities and investigate density fluctuations in 3D bacterial suspensions. We first calculate the two-point density spatial correlation and the density auto-correlation for suspensions of different  $\phi$  and compare them with well-studied velocity correlations extracted from PIV (Fig. 2).

The correlation length  $\lambda$  and correlation time  $\tau$  are determined when the corresponding normalized correlation functions decay to  $1/e$  (Fig. 2c, f). Fig. 2c shows the density correlation lengths at different  $\phi$ , which characterize the scale of density inhomogeneities in suspensions.  $\lambda$  is small at low  $\phi$ , gradually increases with  $\phi$  and reaches a plateau of  $\sim 5l_b$  in high-concentration bacterial suspensions of  $\phi > \phi_c = 3.5\%$ , where  $l_b = 3 \mu\text{m}$  is the average length of bacterial body. The velocity correla-

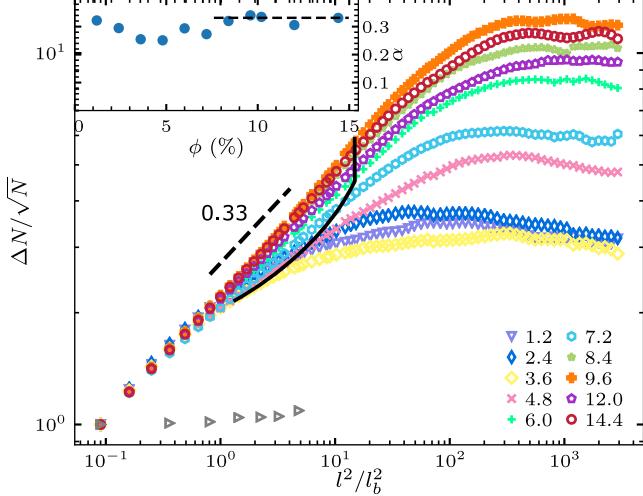


FIG. 3. **GNF at various volume fractions.** Standard deviation of particle number  $\Delta N$  plotted against subsystem size  $l^2/l_b^2$  in bacterial suspensions at volume fractions ranging from 0% to 9.6%. Black dashed line indicates the density correlation lengths at each volume fraction. Black dashed line is a guide of the eye with power 0.33. Inset: scaling exponents  $\alpha$  at different volume fractions. All  $\alpha$ 's are fitted from  $0.3 l_b$  to the correlation lengths. The dashed line in the inset indicates  $\alpha = 0.33$ .

tion length follows the exact same trend and also saturates when  $\phi > \phi_c$ . Such a quantitative similarity indicates the direct coupling between density fluctuations and collective turbulent flows, a feature we shall examine in much more details later. In the fully-developed turbulent regime above  $\phi_c$ , the velocity correlation length is about twice of the density correlation length. Fig. 2f shows the density and velocity correlation times at different  $\phi$ . The velocity and density correlation time at low  $\phi$  show a large discrepancy. When  $\phi > \phi_c = 3.5\%$ , they plateau at  $\sim 6\tau_b$  and  $\sim 4\tau_b$ , respectively, where  $\tau_b = 0.2$  s is the characteristic time scale of *E. coli* swimming.

We further examine GNF by calculating the standard deviation of bacterial number  $\Delta N$  and the mean bacterial number  $N$  in subsystems of increasing sizes [44].  $\Delta N/\sqrt{N}$  as a function of  $l^2/l_b^2$  for bacterial suspensions of different  $\phi$  is plotted in Fig. 3, where  $l$  is the side length of square subsystems. Note that the area of the subsystem  $l^2$  is linearly proportional to mean particle number  $N$  at given  $\phi$ . At small length scale, when  $l \leq \lambda_n$ , bacterial suspensions of all volume fractions examined in this study (1.2% to 14.4%) exhibit GNF, with  $\Delta N/\sqrt{N}$  increasing with subsystem size  $l^2$ . When the size of subsystems is much larger than the density correlation length  $\lambda$ , the giant fluctuation diminishes due to the spatial average over multiple dense and dilute regions.

The degree of GNF can be quantified by the scaling exponent  $\alpha$  following  $\Delta N/\sqrt{N} \sim N^\alpha$ .  $\alpha = 0$  for equilibrium systems obeying the central limit theorem, whereas

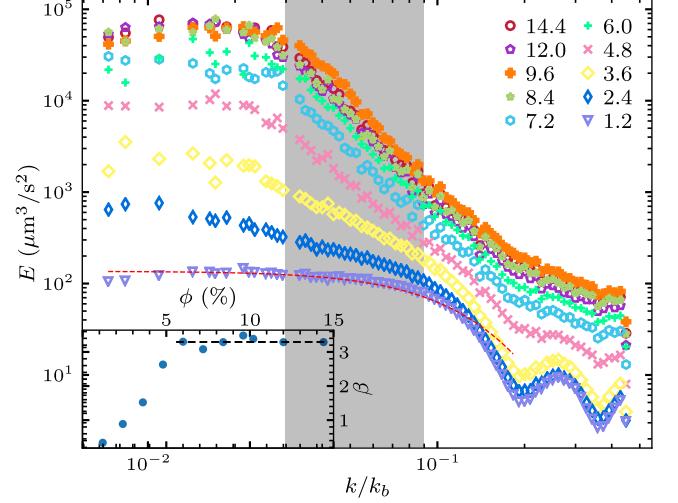


FIG. 4. Energy spectra of bacterial active turbulence at volume fractions ranging from 1.2% to 14.4%. Shaded region indicates the range over which the scaling exponents  $\beta$  are fitted. Red dashed line is a fitting of the  $\phi = 0.8\%$  curve (purple), based on the hydrodynamic model of Ref. [36]. Fitting parameters:  $n = 0.02\%$ . Inset: Scaling exponents of energy spectra ( $\beta$ ) as a function of volume fractions  $\phi$ . Dashed line is  $3.3 \pm 0.1$ .

the upper bound  $\alpha = 0.5$  corresponds to a system with maximal density fluctuations. We extract  $\alpha$  by fitting the experimental curves with power-law relations only below the density correlation length  $\lambda$ . The inset of Fig. 3 shows  $\alpha$  as a function of  $\phi$ . At all volume fractions,  $\alpha$  takes a value of  $0.30 \pm 0.03$ . Notably, at high volume fractions, when  $\phi \geq \phi_h = 6.4\%$ ,  $\alpha$  approaches a narrower plateau of  $0.33 \pm 0.01$ . The plateau value quantitatively agrees with the theoretical prediction of  $\alpha = 1/3$  for 3D suspensions of polar-ordered self-propelled particles with hydrodynamic interactions [20]. As such, our experiments provide the first quantitative verification of the theory of GNF in 3D wet active fluids. The observation of GNF at volume fractions lower than  $\phi_c$  suggests that even in a dilute suspension below the turbulent transition, bacteria swim in a correlated fashion, in qualitative agreement with the prediction of the kinetic theory [48].

## B. Energy spectra

Similar to GNF, the velocity field of active turbulence also shows scale-dependent structures, which are often characterized by the energy spectrum of turbulent flows,  $E(k)$  [44].  $E(k)$  measures the kinetic energy density at different scales in terms of wavenumber  $k = 2\pi/l$ . It is related to the mean kinetic energy density  $E = \langle v_x^2 + v_y^2 \rangle/2$  via  $E = \int_0^\infty E(k)dk$ . Fig. 4 shows  $E(k)$  of bacterial suspensions at different  $\phi$ . The low  $\phi$  energy spectrum has been theoretically predicted for uncorrelated pusher

swimmers [36] as Eq. 1

$$E_k = 4\pi n \kappa^2 \left[ \frac{1}{3} + \frac{\cos(kl)}{(kl)^2} - \frac{\sin(kl)}{(kl)^3} \right] \frac{\epsilon^4 k^2}{l^2} K_2^2(k\epsilon) \quad (1)$$

where  $\kappa \approx 400 \mu\text{m}^3/\text{s}$  is the dipole strength of *E. coli*,  $l \approx 1.9 \mu\text{m}$  is the dipolar length,  $\epsilon$  is a factor describing the distance over which the regularisation acts and  $K_2$  is the modified Bessel function of the second kind. **New parameters:** length  $4.27 \pm 1.10 \mu\text{m}$ , radius  $0.65 \pm 0.09 \mu\text{m}$ , volume  $1.46 \pm 0.63 \mu\text{m}^3$ , swimming velocity  $15 \pm 3 \mu\text{m}/\text{s}$ . The red dashed curve in Fig. 4 shows the best fitting of our experimental data with the theoretical prediction. As can be seen, the low  $\phi$  prediction of  $E(k)$  qualitatively agrees with our measurements. Particularly,  $E(k)$  is flat in the small  $k$  limit, a feature arising from the superposition of the dipolar flow fields of uncorrelated bacteria [36]. However, the fitting parameter  $n$  for the best fitting is found to be 0.02%, 40 times smaller compared to the parameter in real experiment.

With increasing  $\phi$ ,  $E(k)$  at small  $k$  increases sharply. In the turbulent regime at high  $\phi$ , the kinetic energy is all concentrated at scales much larger than the size of single bacteria  $l_b$ , even though the turbulent flows are completely powered by single bacterial swimming at  $l_b$ . Weak peaks are observed for high  $\phi$  suspensions at low  $k$ . This overall trend of  $E(k)$  with increasing  $\phi$  qualitatively agrees with large-scale simulations [26, 36].

We also extract the scaling exponent  $\beta$  from  $E(k) \sim k^{-\beta}$  by fitting the energy spectra at intermediate and large  $k$  (between  $0.03k_b$  and  $0.09k_b$ ), where a significant change of  $E(k)$  with  $\phi$  occurs and  $E(k)$  exhibits good power-law relations. Similar to the trend of  $\alpha$ ,  $\beta$  also increases with  $\phi$  and saturates at  $3.2 \pm 0.1$  at high  $\phi \geq \phi_h$ . The saturated scaling exponent is quantitatively the same as that of the active turbulence of high-concentration suspensions of sperm at large  $k$  [42]. The exponent is also consistent with experiments on confined dense suspensions of *B. subtilis* and swarming *S. marcescens* on agar substrates, where the scaling of  $E(k) \sim k^{-8/3}$  has been reported at large  $k$  [31, 43]. Since the characteristic vortex size quantified by the velocity correlation length is linearly proportional to the system size  $L$  [34], when  $k$  is significantly smaller than  $2\pi/L$ ,  $E(k)$  would decrease, exhibiting the non-monotonic trend shown in [31, 43]. The large system size of  $L = 140 \mu\text{m}$  of our experiments allows us to probe the small  $k$  limit predicted by theories and simulations without the influence of system boundaries.

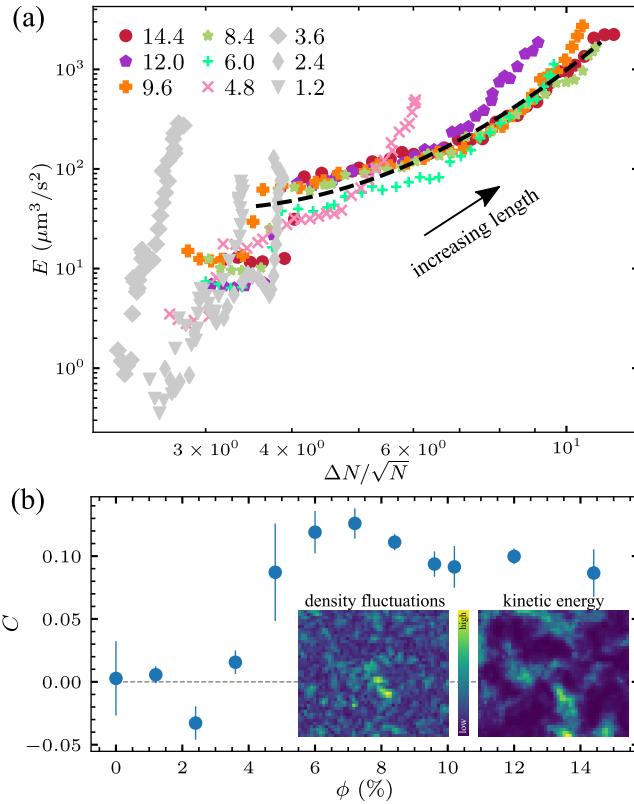
Although the scaling in the small  $k$  limit is strongly affected by the system size, the scaling in the large  $k$  limit seems to be universal with  $\beta \approx 3$  from different experiments. To the best of our knowledge, no theoretical prediction has been made on this interesting scaling behavior for 3D wet active fluids. Giomi investigated the energy spectra of 2D active nematics by combining numerical simulations with mean-field theories and showed  $E(k) \sim k^{-4}$  in the large  $k$  limit [41]. The result has also been confirmed recently by a hydrodynamic theory [37].

For isotropic turbulence in  $d$  dimension, the energy spectra can be written as  $E(k) = C_d k^{d-1} \langle \mathbf{v}(\mathbf{k}) \cdot \mathbf{v}(-\mathbf{k}) \rangle_{k=|\mathbf{k}|}$  [31, 36], where  $C_d k^{d-1}$  is the surface area of  $d$ -sphere and  $\langle \mathbf{v}(\mathbf{k}) \cdot \mathbf{v}(-\mathbf{k}) \rangle_{k=|\mathbf{k}|}$  is the Fourier transform of the velocity-velocity spatial correlation function. If we assume that the velocity correlation function for 2D active nematics is quantitatively similar to that of 3D bacterial suspensions independent of the dimensionality of systems, the mean-field theory would then predict the correct scaling of  $E(k) \sim k^{-3}$  for 3D active turbulence. This hypothesis is certainly highly non-trivial and needs further theoretical investigation. Note that although we measure the energy spectra of 2D in-plane flows of 3D suspensions, the scaling of the spectra is the same as the spectra of 3D turbulent flows [49]. Taken together, our study provides the first systematic experiments on the evolution of  $E(k)$  over a large range of  $\phi$  in bulk bacterial suspensions. The results quantitatively verify the existing theory on low-concentration active fluids and serve as a crucial benchmark for future theoretical development on the spectral properties of 3D active turbulence.

### C. Density-flow coupling

Both GNF and energy spectra probe the scale dependence of active turbulence. The former measures density fluctuations at different scales, whereas the latter considers flow energies across scales. Although both properties have been extensively studied, the coupling between the two has not been explicitly examined so far. The GNF curves and energy spectra shown in Fig. 3 and 4 show similar characteristics, including the rapid increase at small length scale and plateaus at large length scale. Such similarity suggests that a density-independent correlation between density fluctuations and kinetic energies may exist across all different length scales. Indeed, when we plot density fluctuations  $\Delta N/\sqrt{N}$  against the corresponding kinetic energies  $E$  at the same scale in Fig. 5a, all the  $\Delta N/\sqrt{N}$ - $E$  pairs fall onto a universal curve over two orders of magnitude in scales extending from the size of single PIV boxes to the vortex size in the turbulent regime, regardless of the specific volume fractions of the samples. In contrast,  $\Delta N/\sqrt{N}$ - $E$  shows much larger scattering for low  $\phi$  suspensions with random swimming bacteria. Although it is not surprising that density fluctuations correlate with kinetic energies in general as both measure different aspects of the same active turbulence, the collapse of data from samples of different volume fractions is still quite unexpected. The results show that the coupling between density fluctuations and turbulent flows occur at every scale of active turbulence in a quantitative same fashion. Such a scale-invariant coupling is independent of the volume fraction of bacterial suspensions. A phenomenological fitting of the universal coupling between density fluctuations and kinetic energy is shown in Fig. 5a by a black dashed line.

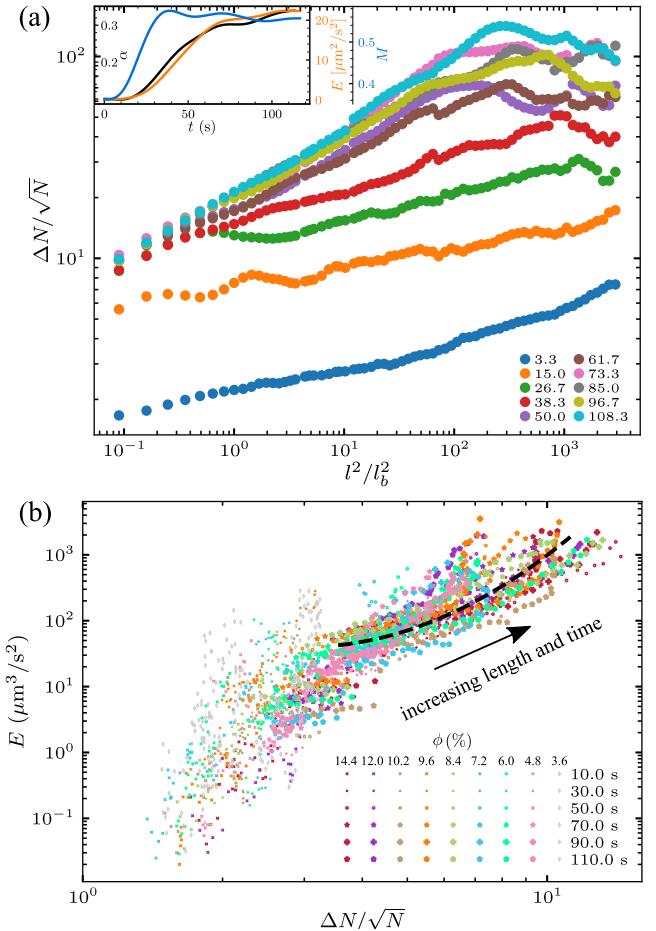
To further illustrate such an unusual coupling in real



**FIG. 5. The coupling between GNF and kinetic energy spectra.** (a) Energy spectra  $E$  plotted against number fluctuations ( $\Delta N/\sqrt{N}$ ) at corresponding length scales at volume fractions ranging from 1.2% to 14.4%. Black dashed line is a phenomenological fitting of the universal coupling between GNF and energy spectra across different volume fractions ( $\log E = 2.2(\log G)^2 - 4.8 \log G + 6.3$ , where  $G = \Delta N/\sqrt{N}$ ). Black arrow indicates the direction of increasing length scale. (b) Local correlations  $C$  between density fluctuations and kinetic energy at volume fractions ranging from 0% to 14.4%. The correlations are calculated by averaging over 1000 frames. Inset: a snapshot of a density fluctuation field and a kinetic energy field.

space, we measure the *local* correlation of density fluctuations and kinetic energy at a randomly chosen small scale of  $l = 2.5l_b$ . The local density fluctuation  $\delta N(\mathbf{r}, t)$  and local kinetic energy  $E(\mathbf{r}, t)$  at position  $\mathbf{r} = (x, y)$  and time  $t$  are extracted from the image intensity field and the PIV velocity field, respectively (Fig. 5b inset) [44]. The normalized correlation between  $\delta N(\mathbf{r}, t)$  and  $E(\mathbf{r}, t)$  is then computed at different  $\phi$  (Fig. 5b). At low  $\phi$  with random swimming bacteria, the correlation is weak fluctuating around zero, which then increases with  $\phi$  as bacterial suspensions transition to active turbulence. A constant positive correlation is found in the turbulent regime when  $\phi \geq \phi_c$ . The real-space measurement provides a concrete example of the coupling between density fluctuations and turbulent flows at small scales.

Moreover, we find the same density-independent cou-



**FIG. 6. GNF kinetics and density flow coupling at transient state.** (a) Kinetics of GNF at the onset of active turbulence in a  $\phi = 9.6\%$  suspension (legends indicates time in unit second). Inset: temporal evolution of scaling exponent  $\alpha$ , total kinetic energy  $E$  and the area fraction of the regions with strong alignment of local velocities  $M$  in the same sample. (b) Kinetic energy and number fluctuations at corresponding time and length scales during the onset of active turbulence at volume fractions ranging from 3.6% to 14.4%. Black dashed line is the phenomenological fitting of the universal coupling, the same as in Fig. 5a.

pling in the kinetic process during the transition towards bacterial turbulence. Taking the advantage of the light-powered bacteria, we trigger the onset of bacterial turbulence by suddenly turning on the light illumination on high-concentration bacterial suspensions at  $t = 0$  [39]. The density fluctuations and the energy spectra are then monitored as a function of  $t$  before the suspensions reach the steady turbulent state. The temporal evolution of GNF curves is shown in Fig. 6. Over time, the scaling exponent gradually increases and shows a saturation at  $\sim 60$  s. The evolution of the scaling exponent  $\alpha$  is shown as a function of  $t$  in Fig. 6 inset. The growth of both  $\alpha$  and the large scale kinetic energy  $E$  are significantly delayed compared with the formation of collective

flows, which is quantified by the area fraction of the regions with strong alignment of local velocities  $M$  [39, 44]. The finding provides strong experimental support to an important prediction of the kinetic theory of active fluids [50, 51], where the density fluctuation arises due to the nonlinear coupling between collective flows and particle densities. At the onset of hydrodynamic instability in the linear regime, the density should stay uniform as confirmed by our experiments at early times.

More importantly, we also analyze the the coupling between  $\Delta N/\sqrt{N}$  and  $E(k)$  during the transition towards active turbulence for bacterial suspensions of different  $\phi$ . As shown in Fig. 6b, when  $\phi > \phi_c$ , all the data at different  $t$ ,  $\phi$  and length scale  $l$  collapse into the same master curve obtained from the steady-state measurements [44].

This kinetic measurements further suggest that the rise of GNF is governed by the kinetic energy at each corresponding scale.

In summary, we have measured GNF and energy spectra simultaneously in 3-dimensional bacterial active turbulence. The scaling exponent  $\alpha$  we measured is in quantitative agreement with the theoretical prediction of polar-ordered self-propelled particle suspension model. Our detailed analysis on the turbulent flow fields reveals a strong coupling between flow kinetic energy and GNF, spanning all the length scales down to several swimmer length and up to the vortex size. Our measurements will contribute to the advancement of quantitative understanding of collective motions and GNF in bacterial active turbulence.

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