Emergence of bacterial glass

Hisay Lama, Masahiro J. Yamamoto, Yujiro Furuta, Takuro Shimaya and Kazumasa A. Takeuchi

Abstract

Densely packed, motile bacteria can adopt collective states not seen in conventional, passive materials. These states remain in many ways mysterious, and their physical characterization can aid our understanding of natural bacterial colonies and biofilms as well as materials in general. Here, we overcome challenges associated with generating uniformly growing, large, quasi-two-dimensional bacterial assemblies by a membrane-based microfluidic device and report the emergence of glassy states in two-dimensional suspension of Escherichia coli. As the number density increases by cell growth, populations of motile bacteria transition to a glassy state, where cells are packed and unable to move. This takes place in two steps, the first one suppressing only the orientational modes and the second one vitrifying the motion completely. Characterizing each phase through statistical analyses and investigations of individual motion of bacteria, we find not only characteristic features of glass such as rapid slowdown, dynamic heterogeneity, and cage effects, but also a few properties distinguished from those of thermal glass. These distinctive properties include the spontaneous formation of micro-domains of aligned cells with collective motion, the appearance of an unusual signal in the dynamic susceptibility, and the dynamic slowdown with a density dependence generally forbidden for thermal systems. Our results are expected to capture general characteristics of such active rod glass, which may serve as a physical mechanism underlying dense bacterial aggregates.

Keywords: bacteria, glass transition, active glass, biofilm, dynamic heterogeneity, orientation glass

Significance Statement

Bacteria often live in the form of dense populations, such as biofilms. While diverse approaches have been taken to understand such aggregates, physical consequences of being dense remained largely unexplored. Here, by using a microfluidic device suitable for a uniform culture of dense bacteria, we revealed that bacteria transition from an actively swimming state to jammed states as they proliferate, through a pathway analogous to glass transitions of colloidal rods. Through analysis of both single-cell and statistical properties, we characterized the observed collective states and transitions, and identified not only similarities but dissimilarities with usual glass formers including colloids. Our model experiment of dense bacteria may impact broad contexts beyond biofilms, hinting at general characteristics of such active rod systems.

Introduction

Dense bacterial populations offer an exciting frontier of research for both physical and microbial sciences. From the biological viewpoint, dense bacterial populations, especially biofilms, abound in diverse natural environments and beyond (1, 2), being also encountered in medicine and industries often as the bane (2–4). It is therefore crucial to characterize their multifaceted aspects (1, 2), including not only extracellular substances and cell phenotypes as studied extensively in the literature, but also physical consequences of being dense aggregates, which have gained an emerging interest in physical sciences. Indeed, recent physical studies on dense bacterial populations unveiled a plethora of collective states not seen in conventional passive materials (5, 6), which however remain in many ways mysterious.

Generally, dense particle systems may undergo glass and jamming transitions as the number density is increased, showing dramatic changes in both single-particle dynamics and material properties as suspension (7–11). Even for conventional thermal systems, it is only recently that firm theoretical grounds on glass transitions started to be built for some idealized cases (12). Therefore, it constitutes an important challenge in physics to extend this understanding to active systems (13–15), i.e., systems made of motile particles akin to cells, or more generally, to athermal systems including intracellular environments. Experimentally, glassy dynamics has been reported in cytoplasm and cell extracts (16–18) as well as in mammalian tissues (19–21), posing interesting problems both in physics and biology (13–15). Concerning bacteria, by contrast, whereas a few recent studies suggested the relevance of glassy...
phenomena to the regulation of motility and three-dimensional growth in bacterial colonies (22, 23), identification of glassy states in bacterial populations remains a challenge, especially under controlled environments that are necessary for the quantitative investigation. A primary difficulty is to keep uniform growth conditions for such dense cell populations for a long time. While microfluidics is generally suitable for controlled experiments, conventional devices that deliver nutrients through channels cannot maintain uniform growth conditions for large and dense populations (24).

Here we overcome past difficulties and report a controlled experiment characterizing emergent glassy phases of dense populations of motile bacteria. We used a membrane-type microfluidic device developed in (25) (Fig. 1a), named the extensive microperfusion system (EMPS). Delivering growth medium to bacteria through a porous membrane, this device can maintain a uniform and constant growth condition for bacteria trapped in a 2D well, even if the bacterial cells are densely packed (25). The membrane may also let out substances secreted by cells, thus providing an ideal platform for studying the physical effect of crowding. In the present work, we cultured motile bacteria (E. coli, strain RP437, width ≈ 1 \( \mu \text{m} \), length varying roughly from 2 to 6 \( \mu \text{m} \); see Fig. S1 for the distribution of cell areas) in a closed 2D well (diameter 71.2 ± 0.5 \( \mu \text{m} \), depth ≈ 1.4 \( \mu \text{m} \)) supplied with growth medium (tryptone broth supplemented with surfactant; see Methods), and monitored their spatiotemporal dynamics in a region near the center. Bacteria were initially swimming actively. However, as the area fraction of bacteria, \( \phi \), increased by cell growth and division at a uniform and constant growth rate for bacteria trapped in a 2D well (Fig. 1b–e, Figs. S2 and S3), we found that the motion of the bacterial population started to be hampered rapidly (Movies S1–5), while their positions and orientations remained globally disordered in space (Fig. 1F–i). In particular, the static structure factor does not show a significant change in its shape (Fig. 2a) in the range of \( \phi \) where the bacterial motion slows down.

**Rapid dynamic slowdown**

The rapid suppression of motion can be evaluated by the difference of phase-contrast image intensities taken at two different times, \( \Delta I(r, t, \Delta t) = I(\Delta t, t) - I(0, t) \). Figure 1j–m shows that, with increasing \( \phi \), the region with \( \Delta I(r, t, \Delta t) \approx 0 \) (where bacteria hardly moved during the chosen time interval) expanded, and eventually, near \( \phi = 0.88 \), the entire population became kinetically arrested, i.e. vitrified. We also notice that this kinetic arrest took place homogeneously (see Fig. 1k, l and Movie S1), analogously to glassy systems showing the dynamic heterogeneity (8, 9).

We characterize the observed vitrification by the differential variance analysis (DVA) (26), which uses the intensity difference \( \Delta I(r, t, \Delta t) \) to analyze the structural relaxation and the dynamic heterogeneity. The structural relaxation is studied by the overlap function defined by

\[
Q(\Delta t) = 1 - \frac{V(\Delta t)}{V(\infty)},
\]

where \( V(\Delta t) = \langle \Delta I(r, t, \Delta t)^2 \rangle_{C_1} \) is the intensity variance taken over position \( r \) and reference time \( t \), and \( V(\infty) \) is evaluated by twice the variance of \( I(r, t) \). The quantity \( Q(\Delta t) \) roughly corresponds to the fraction of bacteria that did not move over lag time \( \Delta t \). More quantitatively, \( Q(\Delta t) \) was reported to behave similarly to the self-intermediate scattering function (26), a quantity often used to characterize the structural relaxation of glassy materials.

Figure 2b shows the result of structural relaxation assessed through \( Q(\Delta t) \) for different \( \phi \). For low \( \phi \), \( Q(\Delta t) \) decays to zero after a relatively short relaxation time, indicating fast structural relaxation. This corresponds to what is called the \( \alpha \)-relaxation in the literature (7–11). However, this \( \alpha \)-relaxation time increases rapidly for \( \phi \geq 0.85 \), soon exceeding the observation time. To be more quantitative, we fit the data by a stretched exponential function, \( Q(\Delta t) \sim e^{-\langle \Delta t(\phi)/\tau_\alpha \rangle^\beta} \), well known to describe the \( \alpha \)-relaxation of glassy materials (7, 8), and evaluate the relaxation time \( \tau_\alpha \) thereby (Fig. S4). The result indeed shows a rapid increase of \( \tau_\alpha \) (Fig. 2d, blue circles), by nearly two orders of magnitude in 0.78 ≤ \( \phi \) ≤ 0.88. The observed superexponential growth of \( \tau_\alpha \) indicates that our bacterial populations are a fragile glass former (7–9). It is compatible with typical growth laws documented in the literature, such as the power-law divergence

\[
\tau_\alpha \sim (\phi^2 - \phi)^{-\gamma_0},
\]

predicted by mode-coupling theories (MCT) (7, 8, 11) (Fig. 2d, blue dashed line), as well as the Vogel–Fulcher–Tamman law (7–9), \( \tau \sim \exp\left(\frac{c_0}{\phi - \phi_0}\right) \) (Fig. S6). This allows us to evaluate the glass transition point; for example, from the MCT power law Eq. (2), one obtains \( \phi^2 = 0.882(4) \), where the number(s) in the parentheses represents the uncertainty in the last digit(s) (see Methods).

**Two-step transition**

At this point, it is worth recalling the spherocylindrical shape of the constituting entity, namely E. coli, which has both translational and orientational degrees of freedom. The relaxation of the orientational degrees of freedom can be evaluated by the orientational correlation function

\[
C_\theta(\Delta t) = \langle \cos 2(\theta(r, t + \Delta t) - \theta(r, t)) \rangle_{C_1},
\]

where \( \theta(r, t) \) represents the nematic orientation angle. Figure 2c shows \( C_\theta(\Delta t) \) for different \( \phi \). Similarly to \( Q(\Delta t) \), \( C_\theta(\Delta t) \) also decays, following the stretched exponential form \( C_\theta(\Delta t) \sim e^{-\langle \Delta t(\phi) / \tau_\theta \rangle^\beta} \) (Fig. S7), with a characteristic relaxation time \( \tau_\theta \) that increases rapidly with \( \phi \) (Fig. 2d, red squares). Importantly, we find that the orientational relaxation time \( \tau_\theta \) is larger than that of the overlap function \( \tau_\alpha \) by an order of magnitude or more, and seems to diverge at lower \( \phi \). This is underpinned by the MCT fitting (Fig. 2d, red dashed line)

\[
\tau_\theta \sim (\phi^2 - \phi)^{-\gamma_\theta},
\]

which gives \( \phi^2 = 0.852(13) \) that is significantly smaller than \( \phi^2 = 0.882(4) \). The same conclusion was reached when the data were fitted with the Vogel–Fulcher–Tamman law (Fig. S6). From \( \phi^2 < \phi^2 \), we conclude that the orientational degrees of freedom vitrify earlier than the rest, i.e. the translational degrees of freedom, the latter of which essentially governed the relaxation of the overlap function. In other words, the glass transition in our system takes place in two steps: the first being a transition to the orientational glass state at \( \phi^2 = 0.852(13) \) and the second the ultimate transition to the complete glass at \( \phi^2 = 0.882(4) \) (Fig. 2d).

**Dynamic heterogeneity**

Another hallmark of glassy dynamics is dynamic heterogeneity, which concerns nontrivial spatiotemporal correlation that develops near the glass transition (8, 9). It is often quantified by the dynamic susceptibility \( \chi_\tau \), which is essentially the variance of the structural correlation function. In DVA (26), it can be evaluated through the variance of \( Q(t, \Delta t) \), defined analogously to Eq. (1)
Formation of microdomains and collective motion

Here we scrutinize the spatial structure of the orientation field $\theta(r, t)$, which turned out to guide the motion of bacteria. Figure 4a,b shows, in their top left corner, $\theta(r, t)$ measured in the entire well, for $\phi = 0.784(7)$ (active fluid phase) and $\phi = 0.873(4)$ (orientation glass), respectively (see also Movies S8–11). Both figures indicate the formation of nematic microdomains, composed of cells that are oriented in similar directions, reminiscent of those reported earlier for growing colonies of nonmotile bacteria (27). By plotting $\langle \Delta \theta(r, t) \rangle^2$ (main panels of Fig. 4a,b), we find that these domains are surrounded largely, if not entirely, by nearly discontinuous borders (red; notice the logarithmic scale of the pseudocolor code) with neighboring domains oriented in different directions. This microdomain structure is considered to be a result of the competition between the steric interaction, which tends to align the cells locally, and the activity-driven force, which originates from self-propulsion and cell growth and tends to destabilize the ordered alignment. We characterize the distribution of microdomain areas $A$ and find an exponential distribution $e^{-A/A_0}$, with characteristic area $A_0 = 8.3(21) \mu m^2$ that hardly depends on $\phi$ (Fig. 4c and Fig. S9). The exponential distribution was also found for nonmotile bacterial colonies (27), but the characteristic area $A_0$ turned out to be much smaller in our motile bacteria, presumably because of the stronger destabilizing force due to the self-propulsion.

An important consequence of the formation of microdomains is the emergence of collective motion. Since individual bacteria tend to move along the orientation field, they attempt to move collectively in each microdomain, though the motion is largely hampered in the glassy states. However, in the active fluid phase, bacteria do move in the form of microdomains or clusters (Fig. 5a,b and Movies S8 and 9), more visibly for lower $\phi$, similarly to swarming states of bacteria (28–32). The collective motion is evidenced by the presence of the velocity correlation between nearby cells (Fig. 5c), within a distance of $\approx 2 \mu m$, which is consistent with the typical size of the microdomains reported above. Note that the velocity correlation is generally absent in thermal systems, hence another athermal characteristic of our system,
and is known to play a significant role in active glassy dynamics (13, 15, 33). In our system, the velocity correlation is formed as a result of the microdomain structure of the orientations, but it is noteworthy that the observed velocity correlation indicates the presence of polar order, despite the essentially nematic nature of the steric interaction. This collective motion produces non-trivial spatiotemporal correlation in the cell positions, which presumably results in the anomalous peak of the dynamic
susceptibility $\chi^2$ found to develop for low $\phi$ (Fig. 2e). This argument is also supported by the fact that the anomalous peak is absent in the dynamic susceptibility of the orientation (Fig. 2f), $\chi^2$, which is not affected by the bacterial motion. Interestingly, a similar short-time peak was also reported in a recent simulation of particles self-propelling in discrete directions and attributed to the cooperative vibrational motion of active particles (34).

**Concluding remarks**

To summarize, we report that dense populations of motile bacteria spontaneously vitrify as the number density increases by cell growth and division. We found that this transition takes place in two steps as sketched in Fig. 2d, first from the active fluid phase to the orientation glass, where only the orientational degrees of freedom are arrested, and second to the complete glass where the remaining translational degrees of freedom are also arrested. These transitions were found to show characteristic properties of glass transitions, such as rapid slowdown, fragility, dynamic heterogeneity, and cage effects. We also showed that bacteria form nematic microdomains, leading to collective motion which seems to result in the unusual signal in the dynamic susceptibility $\chi^2$ for low $\phi$. These are attributed to the rod shape and self-propulsion of bacteria, and as such, can be characteristic of active rod systems in general.

Our analysis has revealed some other aspects that deserve further investigation. First, regarding the two-step transition, it is not trivial whether the glass transition should first occur in the orientational modes, in the translation modes, or both simultaneously. The literature is mixed in this respect; experiments and simulations of ellipsoidal colloid glass showed the two-step transition in the same order (35–38), but an MCT approach to hard ellipsoids (39) as well as simulations of an active dumbbell model (40) exhibited the two transitions in the opposite order, i.e. first translation, then orientation. The two transitions may occur simultaneously depending on the particle aspect ratio, but to our knowledge, no theory or system has demonstrated a change in the order. It is therefore an interesting open problem to clarify what physical property controls the type of the two-step transition. Second, we remark that our estimates of the MCT exponents for the dynamic slowdown, $\gamma_2 = 1.6(3)$ and $\gamma_6 = 1.5(13)$, seem to suggest the athermal nature of our system, because MCT generally gives $\gamma > 1.76$... for thermal systems (7) (see supplementary material). It is important to identify what aspect of our active system is responsible for violating this lower bound for thermal systems, and what the consequence is.

All in all, our experiment serves as a model system for investigating physical properties of dense bacterial populations, with potential relevance in various contexts such as biofilms. It also contributes to the understanding of characteristic glassy
dynamics of such active rod systems, a class of active, or athermal glass, which represents a frontier in the study of the physics of glass.

**Materials and methods**

**Culture of bacteria**

We used a motile strain of *E. coli*, RP437. As one of the standard strains for motility studies, its run-and-tumble behavior is well documented in the literature (41). First we inoculated bacteria into a sterile test tube with 5 mL of Luria–Bertani (LB) medium (containing bacto-tryptone 10 g/L, yeast extract 5 g/L, and NaCl 10 g/L). Then we incubated the liquid culture inside a shaking incubator at 37°C for 12 h at the shaking speed of 220 rpm. We resuspended 100 μL of the overnight culture into 10 mL of fresh tryptone broth (TB, containing bacto-tryptone 10 g/L and NaCl 10 g/L), and again incubated it for 4 h in the same culture condition. Finally, we measured the optical density of the liquid culture by a spectrophotometer. The optical density at wave length 600 nm, OD_{600}, was 0.4, which is equivalent to 10^6 cells/mL. We diluted the liquid culture to a concentration of OD_{600} = 0.1 (≈10^4 cells/mL) by resuspending it into a fresh TB medium containing 0.02 wt% of surfactant Tween-20.

**Fabrication of microfluidic device**

We performed experiments using a membrane-based microfluidic device, namely, the EMPS (25) (Fig. 1a). The EMPS comprises a micropatterned coverslip, a bilayer porous membrane (cellulose and polyethylene terephthalate), and a polydimethylsiloxane (PDMS) pad with an inlet and two outlets. The coverslip substrate used in the present work has an array of circular wells (Shin-Etsu Chemical) and then treated it with biotin solution. To assemble the device, we first soaked the coverslip substrate (diameter ≈ 70 μm and depth ≈ 1.4 μm) that we microfabricated. To assemble the device, we first soaked the coverslip substrate in 1 wt% solution of 3-(2-aminoethyl)aminopropyl trimethoxysilane (Shin-Etsu Chemical) and then treated it with biotin solution. We also prepared a bilayer membrane of EMPS, which comprises a biotin-coated polyethylene terephthalate porous membrane (taken from Transwell 3450, Corning, with a nominal pore size of 0.4 μm) and a streptavidin-decorated cellulose membrane (Spectra/Por 7, Repligen, Waltham, MA, molecular weight cut-off 25,000). Before the experiment, we put 1 μL of the bacterial suspension (OD_{600} = 0.1, prepared by the aforementioned method) on top of wells on the substrate and attached the bilayer membrane to confine bacteria in the wells below the membrane, by biotin-streptavidin bonding between the cellulose membrane and the coverslip. Then we placed a double-sided tape (also acting as a spacer) of thickness 100 μm on the coverslip, enclosing the micropatterned region, and attached a PDMS pad on the double-sided tape. This completes the fabrication of EMPS.

**Observation of bacteria**

The assembled EMPS device was placed on the stage of an inverted optical microscope (Leica DMi8), equipped with a 63x (NA=1.30) oil immersion objective and operated by the software Leica LasX. During the experiment, bacteria in the microfluidic wells were kept supplied with TB medium containing 0.02 wt% of surfactant Tween-20, through the bilayer porous membrane of EMPS. This medium was infused from a syringe by a pump (NE-1,000, New Era Pump Systems), at the flow rate of 60 mL/h for the first 5 min and 2 mL/h for the rest. As a result, bacteria grew and proliferated throughout each experiment. We monitored bacterial population in a well (diameter 71.2(5) μm) by phase-contrast microscopy, capturing images (pixel size 0.1724 μm, optical resolution 0.258 μm) by a charge-coupled device camera (DFC3000G, Leica). For the main experiment presented in this work, there were initially three cells in the observed well, but after roughly five hours, the number increased to ≈ 800. We then started to repeat series of acquisition of 1,000 images at an interval of 0.0263 s, preceded by automatic focus adjustment (Leica adaptive focus control, single-shot mode). We further grouped the images into sets of 500 consecutive images and statistical analysis was carried out for each group. Bacteria continued growing and proliferating during this set of image acquisitions too, leading to an increase of the area fraction φ (Fig. S2). Note that, because the timing of the transition varied a little among different samples, we used a single biological replicate for the main analysis presented in this paper. For the uninterrupted observation shown in Movie S1, phase-contrast images were acquired at a regular interval of 1 s with automatic focus adjustment (continuous mode).

**Image pre-processing**

Pre-processing of the phase-contrast images consists of the following two parts. We corrected the effect of non-uniform illumination, by normalizing the image intensity with the local threshold intensity evaluated by `adaptthresh` function of MATLAB. Also, unless otherwise stipulated, we cropped the images to the region of interest of size 259 × 214 pixels (44.7 μm × 36.9 μm) near the center of the well, to avoid influence from the well perimeter.

**Estimation of the area fraction**

The area fraction φ, i.e. the ratio of the area occupied by bacteria to the total area of the region of interest, was evaluated by binarizing the pre-processed phase-contrast images as follows. The binarization threshold was set by a method called the 3-class fuzzy c-means clustering (42). Then we evaluated the fraction of the darker pixels in the binarized images. Using all 500 images of each data set, we evaluated the mean and the standard deviation of the fraction of the darker pixels, and used them as the most probable value and the error, respectively, of the area fraction φ.

**Static structure factor**

The static structure factor S(\qv) is defined as the squared modulus of the Fourier transform of the image intensity I(\rv, t), i.e. S(\gv) = \langle |\FT[I(\rv, t)|^2] \rangle, where \FT[· · ·] is the 2D Fourier transform and \gv = (q_x, q_y) is the wavenumber vector.

**Orientation field**

The coarse-grained orientation field of bacteria, θ(\rv, t) with \rv = (x, y), was obtained by the structure tensor analysis (43). From the intensity field I(x, y) of a pre-processed phase-contrast image (here t is omitted from the argument for simplicity), the structure tensor J(x, y) is defined by

\[
J(x, y) = \begin{bmatrix}
\langle (\Delta_q^x)^2 \rangle_x & \langle (\Delta_q^x)(\Delta_q^y) \rangle_x \\
\langle (\Delta_q^y)(\Delta_q^x) \rangle_x & \langle (\Delta_q^y)^2 \rangle_x
\end{bmatrix},
\] (6)
where

\[ (g(x, y))_x = \sum_{i,j} (x', y') f_r(x - x', y - y') \]  

(7)

\[ \Delta_x f = \sum_{i,j} (x', y') \frac{\partial f_r}{\partial x} (x - x', y - y') \]  

(8)

\[ \Delta_y f = \sum_{i,j} (x', y') \frac{\partial f_r}{\partial y} (x - x', y - y') \]  

(9)

with a Gaussian kernel \( f_r(x, y) = \frac{1}{\sqrt{2\pi} \sigma^2} e^{-\frac{x^2 + y^2}{2\sigma^2}} \) and \( \sigma = 6 \) pixels = 1.03 \( \mu \)m. Then, the orientation field \( \theta(x, y, t) \) was obtained by

\[ \theta(x, y, t) = \frac{1}{2} \tan^{-1} \left( \frac{2\langle \Delta_x f \rangle \langle \Delta_y f \rangle}{\langle \Delta_x f \rangle^2 - \langle \Delta_y f \rangle^2} \right) \]  

(10)

Tracking bacteria

To track the motion of individual bacteria, we first need to carry out cell segmentation of phase-contrast images. This was done by applying MiSiC (44), a deep learning-based method for the segmentation of bacteria, iteratively to each pre-processed phase-contrast image. More specifically, after each application of MiSiC, we removed the regions where cells were detected, added small noise to the image intensity, varied parameters that set the criteria of cell detection slightly and randomly, and applied MiSiC again, unless cells were already detected in a sufficient fraction of the area or MiSiC was already applied sufficiently many times (specifically, 300 times). After this iteration, we manually corrected the segmentation result if necessary, at least for the cell we focused on and its neighbors for the results presented in Fig. 3 and Movie S6, and for all cells in the field of view, except those on the boundary, for the results in Fig. 5. Then, using the segmentation results of all images in a given data set, we applied Blair and Dufresne’s particle tracking code available at https://site.physics.georgetown.edu/matlab/, which is based on the algorithm developed by Crocker, Grier, and Weeks (45). For the results on the active fluid phase shown in Fig. 5, manual correction of the tracking result was also necessary for bacteria that happened to move over a long distance through the neighbors. Note that the analysis shown in Fig. S8 and Movie S7 did not use the method described here, but is a result of fully manual tracking.

Nonlinear fit

All nonlinear fits presented in this work were carried out by using the Levenberg–Marquardt algorithm, known to solve nonlinear least-squares problems reliably. The uncertainties of the fitting results indicate the 95% confidence interval.

Acknowledgments

We are grateful to Y. Wakamoto and R. Okura for their help in the construction of a prototype of the experimental device used in this work. We thank A. Ikeda for his enlightening discussions on interpretations of the MCT fitting results (supplementary material), H. Yoshino and A. Ikeda for discussions on glass transitions of anisotropic particles, G. Luca for his suggestion to carry out domain analysis (Fig. 4), Y. Han for his suggestion to measure the displacements (Fig. 5), and D. Nishiguchi for frequent discussions. We acknowledge the useful codes and libraries made available by S. Panigrahi and L. Espinosa for MiSiC (44), by J. C. Crocker, E. R. Weeks, D. Blair and E. Dufresne for the particle tracking, by G. Xiong for the 3-class fuzzy c-means clustering (42), and by F. Grussu for the structure tensor method.

Supplementary Material

Supplementary material is available at PNAS Nexus online.

Funding

This work is supported in part by KAKENHI from Japan Society for the Promotion of Science (Grant Nos. JP16H04033, JP19H05800, JP20H00128, JP21K20350, JP24K00593), by Core-to-Core Program “Advanced core-to-core network for the physics of self-organizing active matter (JPJSCCA20230002), and by “Planting Seeds for Research” program and Suematsu Award from Tokyo Institute of Technology.

Author Contributions

K.A.T. conceived the project and directed the research. H.L. and T.S. constructed the device. H.L. performed the experiments. H.L. and K.A.T. analyzed the data. M.J.Y. and Y.F. contributed to the development of the project. All authors contributed to the interpretation of the results. H.L. and K.A.T. wrote the manuscript, and all authors revised or commented.

Preprints

A preprint of this article is published at DOI:10.48550/arXiv.2205.10436.

Data Availability

The data that support the findings of this study, as well as relevant microscope images and scripts, have been deposited in Zenodo at https://doi.org/10.5281/zenodo.11522483.

References

1. Flemming H-C, Wuertz S. 2019. Bacteria and archaea on Earth and their abundance in biofilms. Nat Rev Microbiol. 17(4):247–260.
2. Sauer K, et al. 2022. The biofilm life cycle: expanding the conceptual model of biofilm formation. Nat Rev Microbiol. 20(10):606–620.
3. Hall-Stoodley L, Costerton JW, Stoodley P. 2004. Bacterial biofilms: from the natural environment to infectious diseases. Nat Rev Microbiol. 2(2):95–108.
4. Vishwakarma V. 2019. Impact of environmental biofilms: industrial contributions and its remediation. J Basic Microbiol. 60(3):198–206.
5. Be’er A, et al. 2020. A phase diagram for bacterial swarming. Commun Phys. 3(1):66.
6. Aranson I. 2022. Bacterial active matter. Rep Prog Phys. 85(7):076601.
7. Götz W. 2009. Complex dynamics of glass-forming liquids: a mode-coupling theory. International series of monographs on physics. New York: Oxford University Press.
8. Berthier L, Biroli G. 2011. Theoretical perspective on the glass transition and amorphous materials. Rev Mod Phys. 83(2):587–645.
Zheng Z, Wang F, Han Y. 2011. Glass transitions in Jähne B. 1993: Szamel G, Flenner E, Berthier L. 2015. Glassy dynamics of athe Min TL, Be’er A, Dey S, Mutneja A, Karmakar S. 2022. Enhanced short time peak in - Mandal R, Bhuyan PJ, Chaudhuri P, Rao M, Dasgupta C. 2017. Doostmohammadi A, Yeomans JM. 2019. Coherent motion of - Takatori SC, Mandadapu KK. 2020. Motility-induced buckling - Angelini TE, Shimaya T, Okura R, Wakamoto Y, Takeuchi KA. 2021. Scale invariance of cell size fluctuations in starving bacteria. Commun Phys. 4(1):238. 26 Pastore R, Pesce G, Caggiomi M. 2017. Differential variance analysis: a direct method to quantify and visualize dynamic heterogeneities. Sci Rep. 7(1):43496.