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You will know by its tail: a method for quantification of heterogeneity of bacterial populations using single-cell MIC profiling†

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Severe non-healing infections are often caused by multiple pathogens or by genetic variants of the same pathogen exhibiting different levels of antibiotic resistance. For example, polymicrobial diabetic foot infections double the risk of amputation compared to monomicrobial infections. Although these infections lead to increased morbidity and mortality, standard antimicrobial susceptibility methods are designed for homogenous samples and are impaired in quantifying heteroresistance. Here, we propose a droplet-based label-free method for quantifying the antibiotic response of the entire population at the single-cell level. We used *Pseudomonas aeruginosa* and *Staphylococcus aureus* samples to confirm that the shape of the profile informs about the coexistence of diverse bacterial subpopulations, their sizes, and antibiotic heteroresistance. These profiles could therefore indicate the outcome of antibiotic treatment in terms of the size of remaining subpopulations. Moreover, we studied phenotypic variants of a *S. aureus* strain to confirm that the profile can be used to identify tolerant subpopulations, such as small colony variants, associated with increased risks for the development of persisting infections. Therefore, the profile is a versatile instrument for quantifying the size of each bacterial subpopulation within a specimen as well as their individual and joined heteroresistance.

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Introduction

Heterogeneity of an infecting bacterial population is an important consideration for the identification of an effective antibiotic treatment.^{1,2} Some of the most common infectious diseases are caused by more than one co-colonizing bacterial pathogen, including soft tissue infections, peritonitis, cystic fibrosis, urinary tract infections, and endocarditis.³ An

estimated 10% of antibiotic-resistant subpopulations is undetectable by current diagnostic tests leading to treatment failure.⁴ Moreover, polymicrobial infections doubled the number of amputations in diabetic foot infections,⁵ while polymicrobial bloodstream infections were shown to increase the mortality rates from 24 to 47%.⁶

Standard methods for antibiotic susceptibility testing fail to inform on the level of distribution of antibiotic resistance of the entire bacterial population of co-colonizing species. An example of co-infecting pathogens is *Staphylococcus aureus* and *Pseudomonas aeruginosa* abundant and prevailing in the airways and lungs of cystic fibrosis patients.^{7,8} Inadequate antibiotic regimens are likely to interfere with the population's complexity in an uncontrollable manner and may promote survival or emergence of resistant subpopulations or remove the non-pathogenic susceptible microflora.⁹

Moreover, in response to environmental stress bacteria can transiently modify their transcriptional profiles or acquire mutations. This type of stress-induced phenotypic heterogeneity has been known since the middle of the 20th century.¹⁰ These mutations often lead to the reduction in metabolic activity or changes to the cell membrane potential that result in an altered phenotypes, such as a longer lag time, increased tolerance to antibiotics. Stress-

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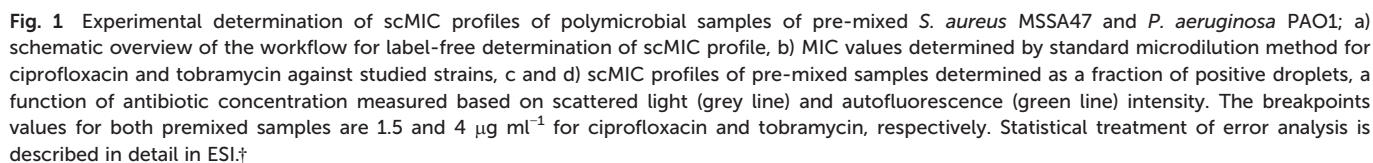
† Electronic supplementary information (ESI) available: Scheme of optical setup, exemplary autofluorescence signals, growth curves of normal colony and small colony variant phenotype, precision of scMIC curve determination, list of bacteria strains tested towards native fluorescence intensity, breakpoint concentrations of NCPs and SCVs, details regarding determination of scMIC profile. See DOI: <https://doi.org/10.1039/d2lc00234e>

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Recently, we reported that scattering can be used as a label-free detection of bacterial growth in nanoliter droplets and for the determination of scMIC.²⁷ Here, we extended the experimental setup by the implementation of an additional label-free detector of natural light emission which is specific to some bacterial species. We utilized the double detection device to investigate the scMIC profiles of phenotypically complex bacterial populations, including polymicrobial

We sought to use this label-free detection consisting of two selective detectors to determine the antibiotic susceptibility profiles of samples consisting of pre-mixed suspensions of *S. aureus* and *P. aeruginosa*. Both opportunistic pathogens are frequently co-isolated from infections of catheters, endotracheal tubes, skin, eyes, and the respiratory tract, including the airways of people with cystic fibrosis. A range of antibiotic concentrations was added to the pre-mixed suspensions prior to experiment. Emulsification of single bacterial cells from the sample into 1 nL droplets achieving *ca.* 10% of single inoculated droplets containing at least one cell capable of proliferation. In further discussion, we refer to these droplets as “non-empty”. Approx. 30 thousand droplets were produced from each sample. Readouts of scattered light and autofluorescence intensities were taken for each droplet after incubation for 18 hours at 37 °C [Fig. 1a]. The number of droplets with multiplying bacteria was scored based on their high intensity of either signal and the fractions of positive droplets, f_+ , were calculated for each antibiotic concentration (see ESI† for



Next, we used scattering-based detection to determine the scMIC profiles of phenotypically heterogeneous samples of a



Fig. 2 Experimental determination of single-cell MIC profile for *Staphylococcus aureus* of normal colony phenotype (NCPs) and small colony variants (SCVs); a) schematic overview of the experimental workflow of SCVs triggering and b) determination of scMIC including sample analysis towards phenotypic heterogeneity, c) comparison of single-cell MIC profiles determined for NCPs (red) and SCVs (blue), d) photographs of representative NCP and SCV colonies on tryptic soy agar, respectively. Statistical treatment of error analysis is described in detail in ESI.†

clonal strain. We selected normal morphology colonies (normal colony phenotype, NCP) of strain *S. aureus* SH1000 grown under optimal conditions, as well as colonies representing the small colony variant (SCV) phenotype, which we triggered with gentamicin [Fig. 2a]. Isolated SCV colonies were characterized by phenotypic changes, including aminoglycoside tolerance, small colony size, longer lag time, and reduced pigmentation [Fig. 2d, S3†]. Both populations, consisting of cells harvested from either NCP or SCV colonies, were separately subjected to scMIC testing [Fig. 2b] and the respective scMIC profiles were determined (see Materials and methods for more details). The NCP population required $2 \mu\text{g mL}^{-1}$ for the inhibition of the entire population while the scMIC of the SCV population was determined at $32 \mu\text{g mL}^{-1}$ [Fig. 2c]. Both scMIC values determined for either NCP or SCV were consistent with the breakpoint concentrations measured for the respective populations using the standard microdilution method [Table S2†].

Distinctive scMIC profiles, consisting of two regions, were observed for NCPs and SCVs [Fig. 2c]. The region of low antibiotic concentrations (up to gentamycin of $\text{ca. } 1 \mu\text{g mL}^{-1}$) with a slowly decaying fraction of recovering bacteria $F_R(c)$ was similar for both, NCP and SCV. The second region which we referred to as a transition region started at gentamicin of $1 \mu\text{g mL}^{-1}$ and was different in both populations. For NCPs the transition region ended at $2 \mu\text{g mL}^{-1}$ and had a sharp decline, while for SCVs it reached $32 \mu\text{g mL}^{-1}$ and had a

much milder slope. Therefore, the transition region for SCVs was about 16 times broader than for NCPs. In this sense our SCVs population was 16 times more resistant than NCPs, suggesting it consists of bacterial cells of larger distribution of antibiotic susceptibility. In the low antibiotic region with a high proportion of positive droplets the $F_R(c)$ values fluctuations were observed. They could be related to the decreased stability of non-empty droplets (droplets containing bacteria) likely to be caused by the cell aggregation, which is characteristic of the strain *S. aureus* SH1000.

‘Tails’ in the scMIC profiles

As we characterized the scMIC profiles of either NCPs or gentamicin-triggered SCVs, we wanted to investigate if they could determine their respective contribution within pre-mixed populations of a known proportion. The homogenous suspensions of NCPs and SCVs were combined into heterogeneous mixtures consisting of either 12% SCVs or 50% of SCVs and the remaining portion of NCPs, respectively [Fig. 3a]. As expected, the scMIC profiles of mixtures were combinations of profiles of either population. The profile of 12% SCV declined sharply in the transition region around $1 \mu\text{g mL}^{-1}$ to the $F_R(c)$ value of 0.08 and continued on with a minimal reduction in the shape of a tail between 1 and $8 \mu\text{g mL}^{-1}$. The height of the tail at 0.08 directly corresponded to the size of SCV subpopulation and it amounts to 8% [Fig. 3b].



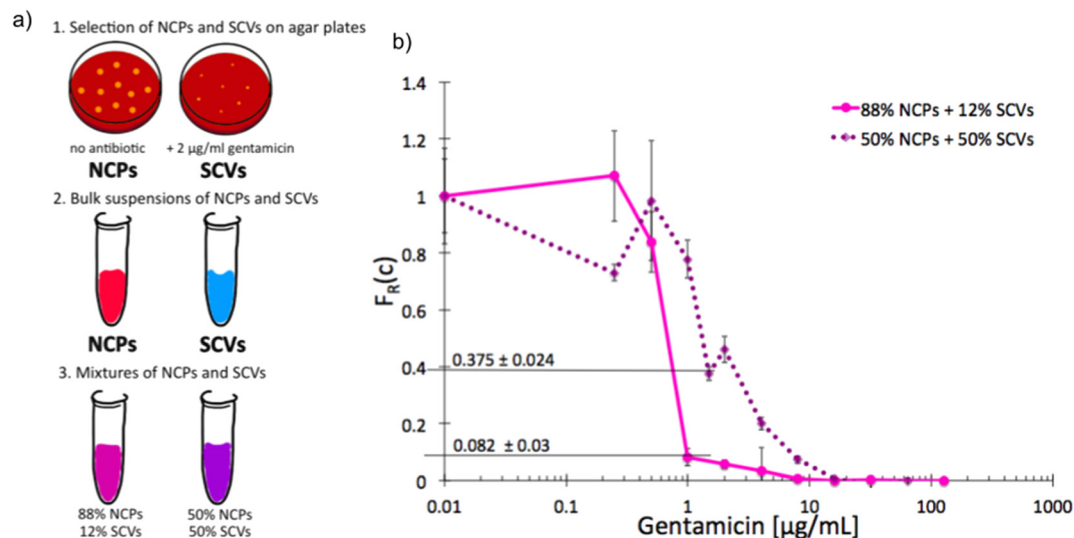


Fig. 3 Single-cell MIC measurement of mixtures of phenotypic variants of *S. aureus* SH1000; a) schematic workflow of sample preparation, b) single-cell MIC profile of phenotypically heterogeneous mixtures containing respectively 12% (solid pink) and 50% (dotted purple) small colony variants (SCV) mixed with normal phenotype cells (NCPs). Statistical treatment of error analysis is described in detail in ESI.†

Whereas the scMIC distribution for the mixture of 50% SCV was broader, yet equally sharp starting at $2 \mu\text{g mL}^{-1}$ down to $F_R(c)$ ca. 0.37 and continuing as a tail ending at $16 \mu\text{g mL}^{-1}$ [Fig. 3b]. Based on the height of the tail we estimated the SCV cells at 37% of the entire population. The difference between 50% and 37% follows from the

precision of our method that we discuss in “Method precision” below.

These results indicate that the tail represents the size of SCVs subpopulation, whether it is low height corresponding to the low SCV content or high with a larger size subpopulation of tolerant bacteria. Consequently, the shape



Fig. 4 Single-cell MIC profiles of *S. aureus* dispersed from the biofilm. a) Schematic workflow of biofilms formation together with NCPs and SCVs selection in the absence (top) and presence (bottom) of gentamicin; b) heterogeneity profiles for NCPs (solid red) and SCVs (dotted blue) isolated from a biofilm cultured for 24 hours in the presence of gentamicin; c) scMIC profiles of biofilms cultured in the presence (dotted purple) and absence (solid pink) of gentamicin; d) the tails of scMIC profiles shown in c). Statistical treatment of error analysis is described in detail in ESI.†



gentamicin. The biofilm was dispersed and subjected to scMIC experiments in three technical replicates measured one by one. Fig. S4† shows the heterogeneity profiles of all repetitions. Droplets without antibiotics or containing a low concentration of gentamicin were affected by more significant standard deviations. It is related to decreasing droplet stability with an increasing number of bacteria cells which we observed to be particularly visible for droplets containing *S. aureus* SH1000 cells as they tend to aggregate. These data show the highest deviation from the mean, 38%, in the transition region for concentration $c = 0.5 \mu\text{g ml}^{-1}$. The remaining concentrations give the deviation of F_R smaller than 7%. Our numerical analysis revealed the following SCVs concentrations in repetitions no 1–3: $17.9 \pm 1.4\%$, $31.3 \pm 2.9\%$, $29.8 \pm 3.0\%$, respectively. The average of obtained results is 26% and the standard deviation is 7.3%. The relative error is $7.3/26 = 28\%$.

It is worth mentioning that performing repetitions is not the only way to estimate the method's reproducibility. As we expect, all scMIC curves should be monotonically decaying functions. Therefore, nonmonotonicity around each point (*i.e.* how much the fraction of resistant bacteria is bigger than the measurement for the nearest lower concentration) also indicates the magnitude of the error. From this perspective, performing the repeated experiment in Fig. S4† was unnecessary. But it tests whether the method replicability can be inferred from the nonmonotonicity: notice that the value of nonmonotonicity of each of the scMIC curves from Fig. S4† is of the order of the differences between the different curves. That shows how the method reproducibility can be estimated from every single scMIC curve.

As we discussed above, the error could, in principle, be estimated from the nonmonotonicity of a scMIC curve. Such an estimation requires many measurements for concentrations around a given scMIC value. Because we also deal with situations with a relatively small number of measured concentrations in the region of interest (*e.g.* only two measured point around $\text{scMIC} = 0.5$ in Fig. 3b, dotted curve), we use the experiment with the highest number of measurements, Fig. S4,† to estimate the relative error due to coalescence, which is 28% as discussed above.

Assuming a similar precision of the method (*e.g.* the relative error 28%) for about twice larger content (50% of SCV's which corresponds to Fig. 3b) this precision explains the discrepancy between the prepared sample of 50% of SCV's and the measurement based on a height of the tail (37%) in the discussion of Fig. 3b.

We justify the usage of the above relative error in case of different samples as follows. We assume that the error due to the coalescence of droplets is proportional to the fraction of nonempty droplets. As a result, the relative error of scMIC does not depend on scMIC value. Assuming also that the coalescence is on a similar level for different samples, we used the experiment with best statistics (three scMIC curves in Fig. S4†) to estimate the relative error due to coalescence in Fig. 3b.

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recovering bacteria fraction, $F_R(c)$, and scMIC profile determination.

Autofluorescence intensity detection

Autofluorescence signal was detected in positive droplets by illumination with a 473 nm laser beam delivered by a fiber, collimated with a lens (focal length 19 mm), and focused by a 20 \times microscope objective. Native fluorescence light is collected by the objective, reflected on a dichroic mirror (a cut-off wavelength of 490 nm), passes a bandpass filter (central wavelength of 530 and 43 nm of bandwidth), pinhole, and targets a photomultiplier (Hamamatsu H5783-20). In this configuration, similar to a confocal microscope, only fluorescence originating from the focal volume will reach the photomultiplier. Thus, background autofluorescence (*e.g.*, of PDMS) will be rejected. The optical setup shown in Fig. S1 \dagger was used for simultaneous detection of scattered and autofluorescence light intensity.

Negative droplets are visible due to the autofluorescence of culturing medium, while the presence of a large number of bacteria cells causes a pronounced increase in the fluorescence intensity. The intrinsic fluorescence of bacteria is linked with fluorescent amino acids (tryptophan, tyrosine, and phenylalanine) and coenzymes such as nicotinamide adenine dinucleotide hydride (NADH).³² It's important to emphasize that the detection of bacteria from *S. aureus* genus is possible by optical setup modification. The excitation and spectrum of this bacteria lies in the UV spectral range (centered at approx. 280 nm and 330 nm, respectively).^{32,33} Thus, the detection of autofluorescence of *S. aureus* would require a different light source, and changing bandpass filter placed in front of the photomultiplier.

Author contributions

N. P. performed experiments, M. Z. designed the project, K. M. analyzed the data, N. P. and J. B. developed the autofluorescence-based detection of bacteria growth in the droplets, N. P., M. Z., K. M., J. B., P. G. wrote the manuscript.

Conflicts of interest

The authors declare no competing interests.

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Notes and references

- 1 A. Leligdowicz and M. A. Matthay, Heterogeneity in sepsis: new biological evidence with clinical applications, *J. Crit. Care*, 2019, **231**(23), 1–8.
- 2 C. Pereira, J. Larsson, K. Hjort, J. Elf and D. I. Andersson, The highly dynamic nature of bacterial heteroresistance impairs its clinical detection, *Commun. Biol.*, 2021, **41**(4), 1–12.
- 3 K. A. Brogden, J. M. Guthmiller and C. E. Taylor, Human polymicrobial infections, *Lancet*, 2005, **365**, 253–255.
- 4 K. M. Kuper, D. M. Boles, J. F. Mohr and A. Wanger, Antimicrobial susceptibility testing: a primer for clinicians, *Pharmacotherapy*, 2009, **29**, 1326–1343.
- 5 S. A. S. Hitam, S. A. Hassan and N. Maning, The Significant Association between Polymicrobial Diabetic Foot Infection and Its Severity and Outcomes, *Malays. J. Med. Sci.*, 2019, **26**, 107.
- 6 F. E. McKenzie, Case mortality in polymicrobial bloodstream infections, *J. Clin. Epidemiol.*, 2006, **59**, 760–761.
- 7 K. Bisht, J. Baishya and C. A. Wakeman, *Pseudomonas aeruginosa* polymicrobial interactions during lung infection, *Curr. Opin. Microbiol.*, 2020, **53**, 1–8.
- 8 A. J. Fischer, *et al.* Sustained Coinfections with *Staphylococcus aureus* and *Pseudomonas aeruginosa* in Cystic Fibrosis, *Am. J. Respir. Crit. Care Med.*, 2021, **203**, 328–338.
- 9 M. K. Wieneke, *et al.* Association of Diverse *Staphylococcus aureus* Populations with *Pseudomonas aeruginosa* Coinfection and Inflammation in Cystic Fibrosis Airway Infection, *mSphere*, 2021, **6**, 1–22.
- 10 J. Bigger, Treatment of Staphylococcal Infections with Penicillin by Intermittent Sterilisation, *Lancet*, 1944, **244**, 497–500.
- 11 B. C. Kahl, Small colony variants (SCVs) of *Staphylococcus aureus*—a bacterial survival strategy, *Infect., Genet. Evol.*, 2014, **21**, 515–522.
- 12 C. Vulin, N. Leimer, M. Huemer, M. Ackermann and A. S. Zinkernagel, Prolonged bacterial lag time results in small colony variants that represent a sub-population of persisters, *Nat. Commun.*, 2018, **9**, 1–8.
- 13 G. Loss, *et al.* *Staphylococcus aureus* Small Colony Variants (SCVs): News From a Chronic Prosthetic Joint Infection, *Front. Cell. Infect. Microbiol.*, 2019, **0**, 363.
- 14 L. Tuchscher, B. Löffler and R. A. Proctor, Persistence of *Staphylococcus aureus*: Multiple Metabolic Pathways Impact the Expression of Virulence Factors in Small-Colony Variants (SCVs), *Front. Microbiol.*, 2020, **0**, 1028.
- 15 F. Peyrusson, *et al.* Intracellular *Staphylococcus aureus* persists upon antibiotic exposure, *Nat. Commun.*, 2020, **111**(11), 1–14.

- 16 H. Nicoloff, K. Hjort, B. R. Levin and D. I. Andersson, The high prevalence of antibiotic heteroresistance in pathogenic bacteria is mainly caused by gene amplification, *Nat. Microbiol.*, 2019, **4**, 1–13.
- 17 H. Nicoloff, K. Hjort, B. R. Levin and D. I. Andersson, The high prevalence of antibiotic heteroresistance in pathogenic bacteria is mainly caused by gene amplification, *Nat. Microbiol.*, 2019, **4**, 504–514.
- 18 D. Belikova, A. Jochim, J. Power, M. T. G. Holden and S. Heilbronner, “Gene accordions” cause genotypic and phenotypic heterogeneity in clonal populations of *Staphylococcus aureus*, *Nat. Commun.*, 2020, **11**(11), 1–15.
- 19 J. Q. Boedicker, L. Li, T. R. Kline and R. F. Ismagilov, Detecting bacteria and determining their susceptibility to antibiotics by stochastic confinement in nanoliter droplets using plug-based microfluidics, *Lab Chip*, 2008, **8**, 1265–1272.
- 20 L. Baraban, *et al.* Millifluidic droplet analyser for microbiology, *Lab Chip*, 2011, **11**, 4057–4062.
- 21 X. Liu, *et al.* High-throughput screening of antibiotic-resistant bacteria in picodroplets, *Lab Chip*, 2016, **16**, 1636–1643.
- 22 T. S. Kaminski, O. Scheler and P. Garstecki, Droplet microfluidics for microbiology: Techniques, applications and challenges, *Lab Chip*, 2016, **16**, 2168–2187.
- 23 A. M. Kaushik, *et al.* Accelerating bacterial growth detection and antimicrobial susceptibility assessment in integrated picoliter droplet platform, *Biosens. Bioelectron.*, 2017, **97**, 260–266.
- 24 T. Artemova, Y. Gerardin, C. Dudley, N. M. Vega and J. Gore, Isolated cell behavior drives the evolution of antibiotic resistance, *Mol. Syst. Biol.*, 2015, **11**, 822.
- 25 F. Lyu, *et al.* Phenotyping antibiotic resistance with single-cell resolution for the detection of heteroresistance, *Sens. Actuators, B*, 2018, **270**, 396–404.
- 26 O. Scheler, *et al.* Droplet-based digital antibiotic susceptibility screen reveals single-cell clonal heteroresistance in an isogenic bacterial population, *Sci. Rep.*, 2020, **10**, 3282.
- 27 N. Pacocha, *et al.* High-Throughput Monitoring of Bacterial Cell Density in Nanoliter Droplets: Label-Free Detection of Unmodified Gram-Positive and Gram-Negative Bacteria, *Anal. Chem.*, 2020, **93**, 843–850.
- 28 Y. C. Wu, *et al.* Autofluorescence imaging device for real-time detection and tracking of pathogenic bacteria in a mouse skin wound model: preclinical feasibility studies, *J. Biomed. Opt.*, 2014, **19**, 085002.
- 29 W. K. Philipp-Dormston and M. Doss, Comparison of Porphyrin and Heme Biosynthesis in Various Heterotrophic Bacteria, *Enzyme*, 1973, **16**, 57–64.
- 30 C. D. Cox and P. Adams, Siderophore activity of pyoverdine for *Pseudomonas aeruginosa*, *Infect. Immun.*, 1985, **48**, 130–138.
- 31 A. M. Edwards, Phenotype switching is a natural consequence of *Staphylococcus aureus* replication, *J. Bacteriol.*, 2012, **194**, 5404–5412.
- 32 P. M. Molyneux, S. Kilvington, M. J. Wakefield, J. I. Prydal and N. P. Bannister, Autofluorescence Signatures of Seven Pathogens: Preliminary in Vitro Investigations of a Potential Diagnostic for *Acanthamoeba Keratitis*, *Cornea*, 2015, **34**, 1588–1592.
- 33 H. K. Peguda, *et al.* The Autofluorescence Patterns of *Acanthamoeba castellanii*, *Pseudomonas aeruginosa* and *Staphylococcus aureus*: Effects of Antibiotics and Tetracaine, *J. Pathog.*, 2021, **10**, 894.

