



Single-Cell Imaging Reveals That Staphylococcus aureus Is Highly Competitive Against Pseudomonas aeruginosa on Surfaces

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Pseudomonas aeruginosa and Staphylococcus aureus frequently occur together in polymicrobial infections, and their interactions can complicate disease progression and treatment options. While interactions between P. aeruginosa and S. aureus have been extensively described using planktonic batch cultures, little is known about whether and how individual cells interact with each other on solid substrates. This is important because both species frequently colonize surfaces to form aggregates and biofilms in infections. Here, we performed single-cell time-lapse fluorescence microscopy, combined with automated image analysis, to describe interactions between P. aeruginosa PAO1 with three different S. aureus strains (Cowan I, 6850, JE2) during microcolony growth on agarose surfaces. While P. aeruginosa is usually considered the dominant species, we found that the competitive balance tips in favor of S. aureus on surfaces. We observed that all S. aureus strains accelerated the onset of microcolony growth in competition with P. aeruginosa and significantly compromised P. aeruginosa growth prior to physical contact. Upon direct contact, JE2 was the most competitive S. aureus strain, simply usurping P. aeruginosa microcolonies, while 6850 was the weakest competitor itself suppressed by P. aeruginosa. Moreover, P. aeruginosa reacted to the assault of S. aureus by showing increased directional growth and expedited expression of quorum sensing regulators controlling the synthesis of competitive traits. Altogether, our results reveal that quantitative single-cell live imaging has the potential to uncover microbial behaviors that cannot be predicted from batch culture studies, and thereby contribute to our understanding of interactions between pathogens that co-colonize host-associated surfaces during polymicrobial infections.

Keywords: opportunistic human pathogens, single-cell microscopy, surface colonization, interspecies interactions, polymicrobial infections

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INTRODUCTION

Bacterial infections are frequently caused by multiple species, and such polymicrobial infections can be more virulent and more difficult to treat (Brogden et al., 2005; Peters et al., 2012). For this reason, there is great interest in understanding how pathogens interact and how their interactions affect virulence and treatment outcomes (Short et al., 2014; Filkins and O'Toole, 2015). *Pseudomonas aeruginosa* (PA) and *Staphylococcus aureus* (SA) have emerged as a particularly important model system in this context (Hotterbeekx et al., 2017; Limoli and Hoffman, 2019; Ibberson and Whiteley, 2020), as these two pathogens co-occur in multiple types of infections, including the lungs of cystic fibrosis (CF) patients and chronic wounds (Gjødsbøl et al., 2006; Dowd et al., 2008; Hubert et al., 2013; Maliniak et al., 2016).

Interactions between PA and SA have been studied at the molecular, ecological, and evolutionary levels. Molecular studies revealed that interactions between PA and SA seem to be predominantly antagonistic, whereby PA is the dominant species suppressing the growth of SA through the production of a variety of inhibitory molecules like proteases, biosurfactants, siderophores, and toxic compounds (Kessler et al., 1993; Soberón-Chávez et al., 2005; Mashburn et al., 2005; Hoffman et al., 2006; Harrison et al., 2008; Lin et al., 2018; Orazi et al., 2019; Rezzoagli et al., 2020). At the ecological level, it was shown that the strain genetic background, the spatial structure of the environment, and the relative frequency of strains impact the outcome of interactions (Radlinski et al., 2017; Niggli and Kümmerli, 2020; Barraza and Whiteley, 2021). For example, in our previous work, we showed that PA can only displace the SA strain JE2 when occurring above a certain threshold frequency but fails to invade SA populations when being initially rare (below 5%). At the evolutionary level, there is great interest to understand whether PA and SA (co-)evolve (Michelsen et al., 2016; Briaud et al., 2019; Bernardy et al., 2020; Briaud et al., 2020) and indeed, there is evidence that this is the case, with PA becoming either more (Tognon et al., 2017) or less (Michelsen et al., 2016; Limoli et al., 2017; Camus et al., 2020; Camus et al., 2021) competitive towards SA over time.

Important to note is that our understanding of PA and SA interactions is predominantly based on laboratory batch culture experiments, where large populations grow under shaken conditions. This contrasts with the environment prevailing in infections, where PA and SA frequently act as surface-colonizing pathogens, forming small microcolony aggregates and biofilms (Stoodley et al., 2001; Schleheck et al., 2009; Connell et al., 2010; Alhede et al., 2011; Bjarnsholt et al., 2013; Kragh et al., 2016; Jennings et al., 2021). It is conceivable to assume that interspecies interactions mainly take place at the front of such bacterial aggregates, and that interactions therefore occur at the local micrometer, and not the batch culture scale. Yet little is known about the dynamics and the outcome of competition between species at this scale. The single-cell study by Limoli et al. (2019) is a notable exception, where it was shown that PA modifies its motility upon sensing nearby SA cells.

In our study, we aim to explore single-cell interactions between PA and SA both at the behavioral and fitness level.

For this purpose, we performed time-lapse fluorescence microscopy, where we tracked growing microcolonies on solid agarose patches, either in mono- or mixed culture. Using automated image analysis, we quantified the time until the onset of growth of a microcolony and the number of progenies produced per founder cell and tested whether these two fitness metrices were influenced by the presence of a competitor. Next, we assessed whether there is growth directionality in mixed cultures, whereby the competing species would grow towards or away from each other. We then followed physical encounters between microcolonies of the two species and allocated the various interaction patterns observed into distinct behavioral categories. In a final experiment, we focused on PA and asked whether PA reacts to the presence of SA by changing the expression of key quorum sensing (cell-to-cell communication) genes, known to regulate competitive traits against SA (Lee and Zhang, 2014; Hotterbeekx et al., 2017). Importantly, we repeated all experiments for three SA strains (Cowan I, 6850, JE2) in competition against a single PA strain (PAO1) to test whether micro-scale interactions are SA strain-specific.

MATERIALS AND METHODS

Bacterial Strains Used and General Growth Conditions

We used fluorescently tagged variants of the *Pseudomonas* aeruginosa (PA) strain PAO1 and the untagged *Staphylococcus* aureus (SA) strains Cowan I, 6850 and JE2 (**Table S1**) for all experiments. The wound isolate PAO1 is a commonly used PA laboratory reference strain. Cowan I is a methicillin-sensitive SA (MSSA) strain that is not cytotoxic, but highly invasive and was originally isolated from a patient suffering from septic arthritis. 6850 is a MSSA as well. It is highly invasive, cytotoxic, and hemolytic, and it originates from an osteomyelitis infection. JE2 is a highly virulent, hemolytic, and cytotoxic USA300 community-acquired methicillin-resistant SA (CA-MRSA) strain that was originally isolated from a skin and soft tissue infection (**Table S1**).

For time-lapse experiments, we used a constitutively expressed green fluorescent protein (attTn7::ptac::gfp) in the chromosome of PA as a marker to distinguish PA from SA. For experiments with PA gene reporters, we used PA strains carrying constructs with promoters of interest fused to mCherry together with the housekeeping gene promoter of *rpsL* fused to *gfp* (*attTn7::lasR:: mCherry-rpsL::gfp* and *attTn7::rhlR::mCherry-rpsL::gfp*) (Jayakumar et al., 2021). Prior to imaging, bacterial overnight cultures were grown in 10 ml tryptic soy broth (TSB, Becton Dickinson) in 50 ml falcon tubes for approximately 16 hours at 37 °C and 220 rpm with aeration. After centrifugation and removal of the supernatant, we washed bacterial cells using 10 ml 0.8% NaCl solution and adjusted OD₆₀₀ (optical density at 600 nm) to obtain similar cell numbers per ml for both PA and SA. This was achieved by adjusting OD_{600} of PA to 0.35, for SA strains JE2 and 6850 to 0.65 and for Cowan I to 0.85. Samples were diluted 1:10 with 0.8% NaCl and PA-SA strain pair combinations

were mixed at a ratio of 1:1. 1.5 μ l of this mix and of the respective monocultures was used to inoculate agarose pads for microscopy.

Preparation of Microscope Slides for Imaging

The following method was previously described and successfully used in our laboratory (Weigert and Kümmerli, 2017). To prepare agarose pads, we used standard microscopy slides (76 mm x 26 mm), standard coverslips and 'gene frames' (Thermo Fisher Scientific). Each frame is 0.25 mm thick and sticky on both sides. As a solid growth substrate for bacteria, we heated 20 ml of TSB + 1% agarose in a microwave and pipetted an excess (ca. 400 µl) of medium into the gene frame chamber. We covered the chamber with a microscope coverslip and let the TSB + 1% agarose solidify for around 20 min. at room temperature. After solidification, we removed the coverslip by carefully sliding it upwards and divided the agarose pad into four smaller pads using a sterile scalpel. Channels were introduced around each pad to allow continuous supply of oxygen during microcolony growth. Finally, we pipetted 1.5 µl PA monoculture, 1.5 µl SA monoculture and two times 1.5 µl mixed culture on the four smaller pads. After evaporation of the droplet containing bacteria (ca. 3 min.), we sealed the pads with a new coverslip. Imaging or incubation of the agarose pads at 37°C was started right after slide preparation was completed.

Microcolony Imaging in Time-Lapse And Individual Timepoint Experiments

All microscopy experiments were carried out at the Center for Microscopy and Image Analysis of the University of Zurich (ZMB) with a widefield Olympus ScanR HCS system and the Olympus cellSens software. This microscope features a motorized Z-drive, a Lumencor SpectraX light engine LED illumination system and a Hamamatsu ORCA-FLASH 4.0 V2 camera system (16-bit depth and 2048 x 2048 resolution). For all experiments, we used a PLAPON 60x phase oil objective (NA = 1.42, WD = 0.15 mm) with double digital magnification.

For time-lapse microscopy, we imaged growing microcolonies with phase contrast (exposure time 100 ms) and FITC SEM (exposure time 50 ms, excitation = BP 470 \pm 24 nm, emission = BP 515 \pm 30 nm and DM = 485). Time-lapse recording was performed with temperature in the incubation chamber set to 37°C for six hours with images taken every 10 min. We imaged one PA-SA strain combination per time-lapse experiment and repeated this on three separate days (resulting in nine experiments). On each day, we imaged at least one field of view per monoculture and at least three fields of view for co-cultures.

For imaging individual timepoints measuring gene expression with the PA gene double reporters in the presence vs. absence of SA (5 hours and 8 hours after preparation and incubation of agarose pads at 37°C), we used phase contrast (exposure time 100 ms), FITC SEM (exposure time 50 ms, excitation = BP 470 \pm 24 nm, emission = BP 515 \pm 30 nm and DM = 485) and TRITC SEM (exposure time 400 ms, excitation = BP 550 \pm 15 nm, emission BP 595 \pm 40 nm and DM = 558). We imaged both PA gene double reporters together with and without the three SA strains and the untagged controls (with and without SA) in three independent experiments. For every timepoint, we imaged four fields of view per strain combination and four blank positions (with no bacteria present) to use for average blank subtraction during image analysis (see below).

Image Analysis and Quantification of Growth and Behavioral Patterns

In a first step, we drift-corrected our time-lapse images in Fiji (Schindelin et al., 2012) using a drift correction script, published under a GNU general public license (https://github.com/fiji/ Correct_3D_Drift). The drift-corrected images were then cropped to remove black space that was created during drift correction. Next, we exported the time-lapse series with the ilastik Import Export plugin as HDF5 (https://github.com/ilastik/ ilastik4ij/). In ilastik (version 1.3.2), we created a pixel classification and object classification project in which we imported the respective HDF5 files (Berg et al., 2019). We segmented cells based on phase contrast (to distinguish cells from background) and gfp (to distinguish gfp-positive PA from gfp-negative SA cells), created the respective object predictions using a Gaussian blur with a sigma value of 0.5, the simple thresholding method with a threshold of 0.5, excluded objects smaller than 50 pixels, and exported the resulting object information. The remaining steps of our image analysis workflow were performed in R studio (version 3.6.3). First, we loaded the object predictions into a Shiny app that was programmed in our laboratory. This app allowed us to perform several steps. (1) Mark and exclude false positive cells; (2) exclude cells that exit or enter the field of view during imaging; (3) define groups of cells based on a hierarchical cluster analysis of the euclidean distance between the cells, which can be manually modified and corrected after visual inspection if necessary; and (4) calculate the center of mass for each cell group at each timestep using the formula $x_COM = sum(x_i \times a_i)/sum(a_i)$, where x_{COM} is the center of mass x_i is the cell position and a_i is the area of the cell. We used the information obtained from the Shiny app to calculate: (a) The onset of cell division for each microcolony; (b) the number of progeny cells per founder cell for each microcolony; c) the number of doublings per microcolony; d) the growth rate per microcolony; and (e) the directionality of microcolony growth over time (see detailed descriptions below).

(a) To quantify the onset of cell division for each microcolony, we used the initial cell number of a group (N_i) and calculated at which timestep of imaging that number exceeded N_i for the first time. (b) To calculate the number of progenies per founder cell N_p , we used the formula $N_p = (N_f - N_i)/N_i$, where N_f is the final and N_i is the initial cell number of a microcolony, respectively. c) To calculate the number of doublings per microcolony, we used the formula $[\ln(N_f) - \ln(N_i)]/\ln(2)$. d) To calculate the growth rate for each microcolony, we used the formula $[\ln(N_f) - \ln(N_i)]/ln(2)$. d) To calculate the growth rate for each microcolony, we used the formula $[\ln(N_f) - \ln(N_i)]/ln(2)$, we used the formula $[\ln(N_f) - \ln(N_i)]/ln(2)$. d) To calculate the growth rate for each microcolony, we used the formula $[\ln(N_f) - \ln(N_i)]/ln(2)$. d) To calculate the growth rate for each microcolony are used the formula $[\ln(N_f) - \ln(N_i)]/ln(2)$. d) To calculate the growth rate for each microcolony are used the formula $[\ln(N_f) - \ln(N_i)]/ln(2)$. d) To calculate the growth rate for each microcolony are used the formula $[\ln(N_f) - \ln(N_i)]/ln(2)$. d) To calculate the growth rate for each microcolony are used the formula $[\ln(N_f) - \ln(N_i)]/ln(2)$. d) To calculate the growth rate for each microcolony are used the formula $[\ln(N_f) - \ln(N_i)]/ln(2)$. d) To calculate the growth rate for each microcolony are used the formula $[\ln(N_f) - \ln(N_i)]/ln(2)$. d) To calculate growth directionality D_g , we used the formula $D_g = D_e/D_a$, where D_e is the euclidean distance (corresponding to the distance between the two center of masses of a colony in the first and the last frame) and where D_a is

the accumulated distance (corresponding to the sum of distances between the center of masses of a colony across all successive time points imaged). Random microcolony movements would lead to large D_a but low D_e distances, and thus lead to low directionality D_g values. In contrast, D_g values close to 1.0 would indicate high directional movement of a microcolony (see also Limoli et al., 2019).

To quantify the different behavioral growth patterns in mixed microcolonies when PA and SA came into close contact, we manually screened all the time-lapse series and counted the two most distinct events: (1) PA grows around SA until the end of the six-hour imaging period and (2) SA pushes PA aside and overgrows it.

Image Processing to Quantify PA Gene Expression

We segmented PA cells based on their constitutive gfp fluorescence (SA cells are non-fluorescent) using the interactive pixel and object classification workflow in ilastik (version 1.3.2) (Berg et al., 2019). We again applied a Gaussian blur with a sigma value of 0.5 and used the simple thresholding method with a threshold of 0.5. The resulting binary images (as png-files) were then exported and used as masks for mCherry and gfp quantification in Fiji. To do so, we used a custom-built script that uses the object predictions (in the form of binary images) created in ilastik, the blank images (to subtract average blank fluorescence), FITC (gfp) and TRITC (mCherry) channel images to quantify fluorescence in all predicted objects (corresponding to PA cells). This script performs the following steps: (1) Average blank subtraction in FITC and TRITC channels for each image (to correct for intensity differences across the field of view caused by microscope vignetting); (2) image cropping to a region of interest where all cells are well focused; and (3) background subtraction for both fluorescent channels in each cropped image (to correct for background autofluorescence). We imported the resulting information about the objects (corresponding to PA cells) into R studio and then performed the following steps. (1) Removing objects with an area smaller than 0.5 μ m² (which are most likely not cells); (2) adding a value of 1.0 to each integrated density value to make all datapoints positive (the integrated density is the mean grey value, corresponding to gfp or mCherry fluorescence, multiplied by the area of the cell); (3) calculating the log_{10} of all integrated density values; (4) subtracting autofluorescence of PA gene reporter strains growing alone and PA gene reporter strains growing together with SA (using the average log₁₀ fluorescence intensity of the untagged PA strain growing alone and the untagged PA strain growing together with SA from the same timepoint and experiment, respectively); and (5) plotting the 'corrected' \log_{10} integrated density for mcherry (TRITC) and gfp (FITC).

Statistical Analysis

All statistical analyses were performed with R Studio (version 3.6.3) using linear models. To test whether SA influences PA growth (onset of growth, number of progenies per founder cell, growth directionality), we used analysis of co-variance

(ANCOVA), where we fitted the culture type (PA alone, PA + Cowan I, PA + 6850, PA + JE2) as a fixed factor and the total number of microcolonies present in a field of view as a covariate. To test whether PA influences SA growth, we also used ANCOVA, but fitted SA strain genetic background (Cowan I, 6850, JE2) and presence/absence of PA as fixed factors and the total number of microcolonies present in a field of view as a covariate. Note that we log-transformed the response variable 'number of progenies per founder cell' to obtain normally distributed residues for statistical analysis.

We used Pearson's correlation coefficient to test whether the onset of growth or the number of progenies per founder cell correlated with the founder ratio of PA *vs.* SA cells within a field of view. The founder ratio and the number of progenies per founder cell were log-transformed to obtain normally distributed residues for this analysis.

To compare PA gene expression patterns across culturing conditions, we fitted the culture type (PA alone, PA + Cowan I, PA + 6850, PA + JE2) and the timepoint as fixed factors and 'experimental block' as additional factor (without interaction) to account for variation between independent experiments. The response variables (gfp and mCherry fluorescence values) were log_{10} -transformed prior to statistical analysis (see above). Finally, we used Fisher's exact test to compare whether frequencies of behavioral patterns between PA and SA differ among SA strain background (Cowan I, 6850, JE2).

For all data sets, we consulted diagnostic Q-Q plots and results from the Shapiro-Wilk test prior to statistical analysis to ensure that model residuals are normally distributed. In the results section, we consistently report the statistical test values F and t from our linear models together with the corresponding degrees of freedom as subscripts. We further report p-values with four digits after the comma. In the figures, we use symbols to display four categories of significance as follows: ***p < 0.001; **p < 0.01; *p < 0.05; n.s. (not significant) p > 0.05. Whenever necessary, p-values were corrected using the false discovery rate method.

RESULTS

P. aeruginosa Fitness Is Compromised by *S. aureus* in a Strain-Specific Manner

To address whether fitness of the two species is affected when growing together on a solid surface, we performed single-cell timelapse microscopy of *P. aeruginosa* PAO1 (PA) alone or in the presence of the *S. aureus* (SA) strains Cowan I, 6850 or JE2 (**Figure 1**). As proxies for fitness, we calculated (i) the onset of microcolony growth (i.e. time to first cell division), which reflects the lag phase; and (ii) the number of progenies per founder cell, which is a metric proportional to the doubling rate and growth rate (see **Supplementary Figure S1** for complementary analyses). In the **Supplementary Material**, we provide representative Movies S1–S4 together with **Supplementary Movie Analysis S1–S4** (showing the dynamical fitness patterns of microcolonies) for all strain combinations.



FIGURE 1 | Microscopy workflow for time-lapse fluorescence microscopy. After adjusting *P. aeruginosa* (PA) and *S. aureus* (SA) to similar cell numbers, we inoculated bacteria (each species alone or mixed 1:1) at low cell density on TSB + 1% agarose patches. Time-lapse fluorescence microscopy was carried out for six hours at 37°C with pictures taken every ten minutes. We drift-corrected and cropped the time-lapse images before cell segmentation (PA vs. SA) in ilastik (version 1.3.2). Using a Shiny app in R, we corrected segmentation errors from the exported object predictions, defined groups of cells and extracted the center of mass per group of cells for each timepoint. We then calculated the onset of growth (time to first cell division per microcolony), the number of progenies per founder cell in a microcolony, and the directionality of microcolony growth using automated scripts in R. Distinct microcolony interaction patterns between the species were manually assessed and counted.

We found that the onset of PA microcolony growth was significantly delayed in the presence of Cowan I ($t_{347} = 6.42$, p < 0.0001), but was neither affected by 6850 (t₃₄₇ = 0.59, p = 0.5564) nor by JE2 ($t_{347} = 1.90$, p = 0.0875) (Figure 2A). In contrast, the presence of all three SA strains significantly reduced the number of progenies per founder cell for PA microcolonies (ANOVA: F_{3,347} = 8.58, p < 0.0001, Figure 2C). Interestingly, there were opposing effects of the number of microcolonies (sum of PA and SA microcolonies) in the field of view on the onset of growth and number of PA progenies. While higher numbers of microcolonies led PA to start dividing earlier, the number of PA progenies was reduced when more microcolonies were present in the field of view (ANOVA for the onset of growth: $F_{1,347} = 6.94$, p = 0.0088; number of progenies: $F_{1,347}$ = 24.88, p < 0.0001). Conversely, we found no association between the two fitness measures and the founder ratio of PA vs. SA microcolonies in a field of view (Figure S2). Overall, these findings show that PA fitness is compromised by the presence of SA in a strain-specific manner.

The Onset of *S. aureus* Microcolony Growth Is Accelerated in the Presence of *P. aeruginosa*

Next, we analyzed the fitness of SA strains from the same microscopy co-culture experiments. We found that the onset of microcolony growth depended on the SA strain (ANOVA: $F_{2,318} = 8.09$, p = 0.0004) and on the presence *vs.* absence of PA ($F_{1,318} = 6.96$, p = 0.0087). Particularly, we found that the presence of PA boosted the onset of SA microcolony growth (**Figure 2B**), while a higher number of microcolonies present in a field of view delayed it ($F_{1,318} = 5.89$, p = 0.0158). The founder ratio of PA *vs.* SA microcolonies in a field of view had no effect on the onset of growth (**Figure S2**).

Comparisons of the number of progenies produced yielded a significant interaction between SA strain background and the presence vs. absence of PA ($F_{2,317} = 13.38$, p < 0.0001), which is explained by the fact that the presence of PA reduced the number of progenies of 6850 ($t_{317} = -4.27$, p < 0.0001) but not of Cowan I ($t_{317} = -0.36$, p = 0.7180) or JE2 ($t_{317} = 0.66$, p = 0.7180,



Figure 2D). While the total number of microcolonies within a field of view had no effect on SA performance ($F_{1,316} = 2.70$, p = 0.1017), we found that Cowan I and 6850 (but not JE2) produced more progeny cells per founder cell when the founder ratio of PA *vs.* SA microcolonies in a field of view was biased towards SA (Cowan I: $t_{68} = 2.91$, p = 0.0050; 6850: $t_{46} = 3.63$, p = 0.0007, **Figure S2**). This means that Cowan I and 6850 positively respond to their own cell density. Regarding the overall effect of PA on SA fitness, our results show that the presence of PA accelerates the onset of SA growth on surfaces, whereas overall microcolony growth is either not affected or reduced (for 6850).

P. aeruginosa but Not *S. aureus* Shows Directional Growth in the Presence of a Competitor

We further explored whether PA and SA show increased directional microcolony growth (away or towards each other) in the presence of a competitor. PA generally showed higher levels of directional growth than SA (**Figure 3A**), but directionality was only marginally increased in the presence of SA strains (ANOVA: $F_{3,348} = 2.46$, p = 0.0628). When repeating the analysis with a simpler statistical model testing whether PA

shows directional growth in the presence of SA overall (i.e. collapsing SA factor levels), we found indeed significantly increased directional growth ($F_{1,350} = 4.02$, p = 0.0458), but the effect size was relatively small (mean directionality \pm standard error of PA alone *vs.* with SA: 0.32 \pm 0.02 *vs.* 0.36 \pm 0.01).

For SA, directionality of growth was strain-dependent (ANOVA: $F_{2,319} = 9.37$, p = 0.0001), with JE2 growing more directional than Cowan I and 6850 (JE2 *vs*. Cowan I: $t_{319} = 4.04$, p = 0.0002; JE2 *vs*. 6850: $t_{319} = 3.24$, p = 0.0020; Cowan I *vs*. 6850: $t_{319} = 0.44$, p = 0.6620). Growth directionality was not affected by the presence of PA (ANOVA: $F_{1,319} = 0.03$, p = 0.8667, **Figure 3B**). Note that the number of microcolonies present per field of view did not have a significant effect on PA and SA growth directionality, and this covariate was thus removed from the statistical models. Overall, our analyses revealed that PA shows a weak but significant increase in directional growth in the presence of SA, whereas SA does not.

Strain-Specific Interactions Upon Physical Contact Between Microcolonies

By manually screening all the time-lapse images of our experiments, we noted two frequent behavioral interaction types



from three independent experiments per PA-SA combination, with a total of 352 and 323 microcolonies for PA and SA strains, respectively.

upon physical contact between PA and SA (**Figure 4**): (1) PA grows around SA microcolonies, which can result in a ring-like structure that remains until the termination of imaging after six hours (**Movies S2** and **S3**); and (2) PA comes in touch with SA, which results in PA growth arrest followed by SA pushing PA cells aside and (sometimes) overgrowing them completely (**Movie S4**). We detected 62 distinctive instances in which PA either grows around SA (scenario 1: 29 cases, 46.8%) or is pushed aside and overgrown by SA (scenario 2: 33 cases, 53.2%). The frequency of these two PA behavioral patterns significantly differed in interaction with the three different SA strains (Fisher's exact test p < 0.0001). While in the majority of cases, PA grew around microcolonies of Cowan I (65.0%, **Movie S2**) and 6850 (82.4%, **Movie S3**), PA was typically pushed aside and overgrown in interactions with JE2 (92.0% of all cases, **Movie S4**). These results suggest that JE2 reacts more aggressively towards PA than the other two SA strains.

P. aeruginosa Expedites the Induction of Quorum Sensing Systems in the Presence of *S. aureus*

We hypothesized that PA might sense the presence of competitors like SA and accelerate the expression of



image series and manually counted the frequency of the following two events: (1) PA grows around SA microcolonies (n = 29), and (2) SA pushes PA aside and (sometimes) overgrows PA microcolonies (n = 33). The frequency of these two types of events significantly differ across the three strain pairs (Fisher's exact test p < 0.0001). Visual examples for the two behavioral patterns can be found in the **Supplementary Movies 2-4** (event 1: Movie S2 and S3; event 2: Movie S4). As a comparison, Movie S1 shows PA growing in monoculture. Data is from three independent experiments per PA-SA strain pair.

competitive traits. To test this hypothesis, we focused on PA quorum sensing (QS) systems, which control the expression of competitive traits including the staphylolytic protease LasA and broad-spectrum toxins such as phenazines and hydrogen cyanide (Lee and Zhang, 2014). We quantified the expression of the two main QS-regulator genes *lasR* and *rhlR* together with the housekeeping gene *rpsL* (as a control) in PA cells growing as microcolonies in the presence or absence of the three SA strains.

We found that IasR gene expression depended on the presence *vs.* absence of SA strains (ANOVA: $F_{3,32168} = 697.99$, p < 0.0001), the timepoint measured (5 *vs.* 8 hours post-inoculation: $F_{1,32168} = 1394.89$, p < 0.0001), and the interaction between the two ($F_{3,32168} = 4621.03$, p < 0.0001) (**Figure 5A**). Specifically, our data shows that *lasR* is induced earlier in the presence of SA (5th hour), while gene expression profiles evened out later (8th hour). The expression of *rhlR* was similarly affected as *lasR*. There were significant effects of the presence of SA strains (ANOVA: $F_{3,3231} = 825.88$, p < 0.0001), the timepoint measured ($F_{1,32331} = 14818.31$, p < 0.0001), and an interaction between the two ($F_{3,32331} = 999.55$, p < 0.0001) (**Figure 5B**). While *rhlR* was already expressed at the first timepoint (5th hour), we also found that expression levels were generally higher in the presence of SA at both timepoints measured.

For both the *lasR* and the *rhlR* gene, we observed that the presence of Cowan I and JE2 had a greater influence on PA gene expression than the presence of 6850 (**Figure 5** and **Table S2**). In the latter case, *lasR* and *rhlR* gene expression was more similar to the pattern shown in PA monoculture. Note that the *rpsL* housekeeping gene was constitutively expressed at both timepoints and across conditions (mono *vs.* mixed culture) (**Table S2 + Figure S3**), indicating that the observed differences in QS gene expression are induced by the competitor. In sum, the PA QS gene expression data shows that the presence of SA may lead to adjustments in PA *lasR* and *rhlR* gene expression, especially in competition with Cowan I and JE2, but to a lesser extent with 6850.

DISCUSSION

Pseudomonas aeruginosa (PA) and Staphylococcus aureus (SA) frequently occur together in polymicrobial infections, and there is increasing evidence that their interactions are important for virulence, disease progression, and treatment outcome (Alves et al., 2018; Lenhard et al., 2019). Previous work explored molecular, ecological, and evolutionary aspects of PA-SA interactions. One key insight from this body of work is that PA is often dominant over SA through the production of a variety of inhibitory molecules (Kessler et al., 1993; Soberón-Chávez et al., 2005; Mashburn et al., 2005; Hoffman et al., 2006; Orazi and O'Toole, 2017). While these studies were mostly performed in vitro with planktonic batch cultures, we here used a complementary approach and studied PA-SA interactions at the single-cell level during surface-attached microcolony growth. Since PA and SA often adhere to tissues during infections, colonize medical devices and form aggregates that develop into mature biofilms (Darouiche, 2004; Chang, 2018; Zheng et al., 2018), we argue that it is important to study interspecies interactions under these conditions. Using single-cell time-lapse fluorescence microscopy, we found that SA strains (Cowan I, 6850 and JE2) are highly competitive against PA. Specifically, SA cells started to divide earlier when exposed to PA and all SA strains compromised PA growth before microcolonies came into direct contact. Meanwhile, PA had little effect on SA fitness, but reacted towards the presence of SA by showing increased directional growth and increased expression of quorum sensing (QS) regulators. There were also strain-specific patterns, with PA cells growing around microcolonies of Cowan I and 6850, while being rapidly usurped by JE2 microcolonies. Altogether, our results show that on surfaces, the competitive balance tips in favor of SA (see Table 1 for a summary of all effects).

The key (and rather unexpected) finding of our study is that SA dominates PA on surfaces, which opposes the frequently observed result of PA inhibiting and outcompeting SA in planktonic batch cultures (Kessler et al., 1993; Soberón-Chávez et al., 2005; Mashburn et al., 2005; Hoffman et al., 2006; Orazi et al., 2019). One reason for why SA could be more competitive on surfaces is that this bacterial species has a non-motile lifestyle and might thus be well adapted to rapidly colonize surfaces outside and inside a host (Zheng et al., 2018; Schilcher and Horswill, 2020). In contrast, PA is a flagellated motile bacterium that first engages in surface sensing to then alter its lifestyle and gene expression profile (Chang, 2018; Armbruster et al., 2019). Surface sensing takes time and is likely associated with metabolic costs, which could put PA at a disadvantage compared to SA.

What could be the mechanisms deployed by SA to suppress PA? The fact that all SA strains started to divide earlier in the presence of PA suggests that SA can sense the presence of the competitor (Cornforth and Foster, 2013) and accelerate metabolism to trigger an earlier onset of growth (Leinweber et al., 2018). Although the mechanism by which SA senses competition remains to be elucidated, our observation of an earlier onset of growth indicates that SA engages in resource competition, as predicted for interactions between pathogens competing for limited host resources (Frank, 1996). Moreover, our observation that SA inhibits PA prior to microcolonies coming into contact suggests that SA further engages in interference competition via diffusible compounds to displace its competitor. Candidate inhibitory compounds released by SA are the phenol-soluble modulins (PSMs). PSMs are amphipathic surfactant peptides that can lyse eukaryotic and certain prokaryotic cells, they are pro-inflammatory, play a role in biofilm formation and promote SA spreading on surfaces (Tsompanidou et al., 2013; Cheung et al., 2014). PSMs are produced by virtually all SA strains, and they have previously been suggested to play a role in surface interactions with PA (Limoli et al., 2019). Moreover, for two out of the three SA strains (Cowan I and 6850, but not JE2), we found that they had more progeny per founder cell at higher ratios of SA vs. PA founder cells within a field of view. Cowan I and 6850 therefore positively respond to their own cell density, suggesting that beneficial



monocultures of PA after five hours but evens out across treatments after eight hours. In comparison, the housekeeping gene *rpsL* is more homogeneously expressed across all treatments and time points (**Supplementary Figure S3; Table S2**). The box plots show the median (bold line) with the first and the third quartiles. The whiskers cover the 1.5* inter-quartile range (IQR) or extend from the lowest to the highest value if they fall within the 1.5* IQR. ***p < 0.001. Data is from three independent experiments. **Supplementary Figures S4** and **S5** show the expression of *IasR*, *rhIR*, and *rpsL* from both timepoints for each of the three independent experiments as density plots.

sharing of secreted compounds could be involved and further accelerate growth.

When PA and SA cells came into contact, we saw that JE2 showed a particularly aggressive response towards PA. While we do not know whether contact-dependent interference mechanisms were involved, it was astonishing to see how PA cells were simply pushed aside and sometimes completely disappeared from the microscope field of view. It is known that community-acquired methicillin-resistant *S. aureus* strains, such as JE2, produce particularly high levels of PSMs (Wang et al., 2007; Li et al., 2010). If PSMs were indeed involved in competition on surfaces with PA, this (together with faster growth compared to Cowan I and 6850) could explain why JE2 was the most aggressive SA strain towards PA. While Cowan I showed

	Growth onset	Overall growth	Directionality of growth	Behavior when mixed colonies come into close contact	lasR gene expression	<i>rhl</i> R gene expression
PA + Cowan I	Delayed*** Accelerated**	Reduced ***	۲. ۲.	35% SA overgrows PA, 65% PA grows around SA	Timepoint 1 increased, Timepoint 2 decreased	Timepoint 1 increased, Timepoint 2 decreased
PA + 6850	n.s. Accelerated**	Reduced *** Badi Icad ***	о * с	18% SA overgrows PA, 82% PA grows around SA	Timepoints 1 and 2 increased	Timepoints 1 and 2 increased
PA + JE2	n.s. Accelerated**	Reduced ***	*	92% SA overgrows PA, 8% PA grows around SA	Timepoint 1 increased, Timepoint 2 decreased	Timepoints 1 and 2 increased
PA SA *** <i>p</i> < 0.001; **	*p < 0.01; *p < 0.	05; n.s., not signi	ficant.			

TABLE 1 | Summary of effects in mixed microcolonies compared to monoculture microcolonies

intermediate competitiveness and was not inhibited in overall growth by its competitor, 6850 suffered from competition by PA, which was reflected in a significantly reduced number of progenies per founder cell. As 6850 was already growing poorly in monoculture microcolonies, it is probably at a greater disadvantage in mixed microcolonies with PA compared to Cowan I and JE2.

We now turn to PA and ask why this otherwise very competitive pathogen is comparatively weak against SA on surfaces. PA features many interference traits that could harm its competitor, including LasA protease, pyocyanin, and HQNO (Hotterbeekx et al., 2017). However, these interference compounds are regulated by QS and are only deployed once a certain cell density is reached (Venturi, 2006; Boyer and Wisniewski-Dyé, 2009). Hence, it might be that PA is simply not ready for competition during microcolony formation. Nonetheless, PA was not idle and managed to suppress 6850, the slowest growing SA strain, which occurred prior to microcolonies coming into contact with one another. This suggests that, not only SA, but also PA secretes at least some inhibitory compounds early on during competition. At later stages of microcolony formation, when coming into close contact with SA, we observed that PA reacted to the presence of Cowan I and 6850 microcolonies and started to grow around them. The observed pattern is reminiscent of the exploratory motility phenotype described by Limoli et al. (2019). While the resolution of our time-lapse movies was not high enough to follow specific cell-cell interactions, it did not seem that the potential exploratory motility was associated with any form of contact-dependent killing, and the benefit of this behavior thus remains to be further explored. Finally, our results indicate that PA also seems to sense its SA competitors and to mount a response through the earlier induction of QS. Interestingly, this induction was more prominent in response to Cowan I and JE2 than towards 6850, the weakest SA strain, suggesting that PA adjusts its response relative to the aggressiveness of a competitor. However, more detailed mechanistic work is required to understand potential mechanisms of competition sensing via QS.

It is important to note that there are microbial consortia known to grow better in co-culture compared to monoculture, or sometimes even exclusively grow in co-culture. In some cases, this can be attributed to the cross-feeding of metabolites, whereby species A benefits from a metabolite released by species B (Ramsey and Whiteley, 2009; Pande et al., 2014; Adamowicz et al., 2018). The fact that none of the strains we have used grew better in the presence of the other species, suggests however, that it is unlikely that PA and/or SA produce and release molecules that are beneficial for the other species.

We advocate the view that studying pathogen interactions on surfaces mimics more closely potential interactions in infections. Our study is just an initial step towards a better understanding of how pathogens, such as PA and SA, might grow and interact on host-associated surfaces in infections. There are several aspects that should be considered in future work. First, we know that the relative frequencies of PA and SA impact competitive interactions in planktonic cultures (Niggli and Kümmerli, 2020). It would thus be important to test the effect of relative species frequencies on competition on surfaces in a more systematic way than we have explored here. One possible outcome could be positive frequencydependent competition behavior: with a high initial SA frequency, PA would probably grow very poorly, while at a high initial frequency, PA might be more competitive, potentially able to keep SA at bay. Such insights could reveal so-called 'order effects' that are relevant for polymicrobial infections, whereby the pathogen species that colonizes the host niche first is more competitive than a later arriving species. Another aspect that should be investigated in more detail is the PA gene expression profile in the presence of SA. Studies on transcriptional responses of PA towards SA exist for planktonic culturing conditions, biofilms, and in vivo growth (Mashburn et al., 2005; Miller et al., 2017; Tognon et al., 2019), and it would be important to know how the results compare to transcriptome profiles of single cells on surfaces. Interesting PA candidate genes are not only related to QS (as studied here), but also genes involved in stress response or virulence, all of which could trigger competition sensing and responses towards a competitor (Cornforth and Foster, 2013). Furthermore, little is known about secreted compounds from SA that inhibit PA. Identifying the involvement of PSMs and possibly other SA inhibitory molecules is essential to understand how SA suppresses PA on surfaces. Finally, while we looked at the early stages of microcolony formation, it would be interesting to look at strain dynamics in more mature biofilms, for instance by using flow chambers combined with confocal microscopy 3D analysis, where experiments can be run for longer without the cells overgrowing each other, which frequently occurs after prolonged hours of microcolony growth using our agarose patches.

Taken together, our work shows that the two human opportunistic pathogens P. aeruginosa (PA) and S. aureus (SA) influence each other at the single-cell level on surfaces in manifold ways. While both species seem to be able to sense competition, SA was more competitive, showing both signs of resource competition by starting to grow earlier, and interference competition, presumably through diffusible compounds reducing the growth of PA. Crucially, SA is much more competitive on surfaces than would be anticipated from planktonic batch culture experiments. Since PA and SA colonize host tissues in the context of infection, we provide new hints on the competitive strengths of these two important pathogens that often co-exist in infections. Moreover, our results with a panel of genetically distinct SA strains suggests that the virulence potential of SA strains might play a role in competition with PA, with JE2 being the most virulent and most competitive strain on surfaces. Overall, we propose that time-resolved quantitative live imaging has the potential to uncover novel

REFERENCES

- Adamowicz, E. M., Flynn, J., Hunter, R. C., and Harcombe, W. R. (2018). Cross-Feeding Modulates Antibiotic Tolerance in Bacterial Communities. *ISME J.* 12 (11), 2723–2357. doi: 10.1038/s41396-018-0212-z
- Alhede, M., Nørskov Kragh, K., Qvortrup, K., Allesen-Holm, M., van Gennip, M., Christensen, L. D., et al. (2011). Phenotypes of Non-Attached Pseudomonas

interspecies interactions in an ecologically relevant context. This approach may not only be useful to further our insights on interactions between PA and SA but may significantly improve our understanding of interactions between any two or more species infecting a host based on a surface-colonizing lifestyle.

DATA AVAILABILITY STATEMENT

The original contributions presented in the study are included in the article/**Supplementary Material**. Further inquiries can be directed to the corresponding authors.

AUTHOR CONTRIBUTIONS

SN and RK designed research. SN performed research. TW wrote the image analysis scripts. SN and RK analyzed data and wrote the paper with input from TW. All authors contributed to the article and approved the submitted version.

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SUPPLEMENTARY MATERIAL

The Supplementary Material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/fcimb.2021. 733991/full#supplementary-material

Aeruginosa Aggregates Resemble Surface Attached Biofilm. *PloS One* 6, e27943. doi: 10.1371/journal.pone.0027943

Alves, P. M., Al-Badi, E., Withycombe, C., Jones, P. M., Purdy, K. J., and Maddocks, S. E. (2018). Interaction Between Staphylococcus Aureus and Pseudomonas Aeruginosa Is Beneficial for Colonisation and Pathogenicity in a Mixed Biofilm. *Pathog. Dis.* 76, fty003. doi: 10.1093/ femspd/fty003

- Armbruster, C. R., Lee, C. K., Parker-Gilham, J., de Anda, J., Xia, A., Zhao, K., et al. (2019). Heterogeneity in Surface Sensing Suggests a Division of Labor in Pseudomonas Aeruginosa Populations. *eLife* 8, e45084. doi: 10.7554/elife.45084
- Barraza, J. P., and Whiteley, M. (2021). A Pseudomonas Aeruginosa Antimicrobial Affects the Biogeography But Not Fitness of Staphylococcus Aureus During Coculture. *mBio* 12, e00047–21. doi: 10.1128/mBio.00047-21
- Berg, S., Kutra, D., Kroeger, T., Straehle, C. N., Kausler, B. X., Haubold, C., et al. (2019). Ilastik: Interactive Machine Learning for (Bio)Image Analysis. *Nat. Methods* 16, 1226–1232. doi: 10.1038/s41592-019-0582-9
- Bernardy, E. E., Petit, R. A.III, Raghuram, V., Alexander, A. M., Read, T. D., and Goldberg, J. B. (2020). Genotypic and Phenotypic Diversity of Staphylococcus Aureus Isolates From Cystic Fibrosis Patient Lung Infections and Their Interactions With Pseudomonas Aeruginosa. *mBio* 11, e00735–20. doi: 10.1128/mBio.00735-20
- Bjarnsholt, T., Alhede, M., Alhede, M., Eickhardt-Sørensen, S. R., Moser, C., Kühl, M., et al. (2013). The *In Vivo* Biofilm. *Trends Microbiol.* 21 (9), 466–474. doi: 10.1016/j.tim.2013.06.002
- Boyer, M., and Wisniewski-Dyé, F. (2009). Cell-Cell Signalling in Bacteria: Not Simply a Matter of Quorum. *FEMS Microbiol. Ecol.* 70 (1), 1–19. doi: 10.1111/ j.1574-6941.2009.00745.x
- Briaud, P., Bastien, S., Camus, L., Boyadjian, M., Reix, P., Mainguy, C., et al. (2020). Impact of Coexistence Phenotype Between Staphylococcus Aureus and Pseudomonas Aeruginosa Isolates on Clinical Outcomes Among Cystic Fibrosis Patients. *Front. Cell. Infect. Microbiol.* 10, 266. doi: 10.3389/ fcimb.2020.00266
- Briaud, P., Camus, L., Bastien, S., Doléans-Jordheim, A., Vandenesch, F., and Moreau, K. (2019). Coexistence With Pseudomonas Aeruginosa Alters Staphylococcus Aureus Transcriptome, Antibiotic Resistance and Internalization Into Epithelial Cells. Sci. Rep. 9, 16564. doi: 10.1038/s41598-019-52975-z
- Brogden, K. A., Brogden, K. A., Guthmiller, J. M., Guthmiller, J. M., Taylor, C. E., Taylor, C. E., et al. (2005). Human Polymicrobial Infections. *Lancet* 365, 253– 255. doi: 10.1016/S0140-6736(05)70155-0
- Camus, L., Briaud, P., Bastien, S., Elsen, S., Doléans-Jordheim, A., Vandenesch, F., et al. (2020). Trophic Cooperation Promotes Bacterial Survival of Staphylococcus Aureus and Pseudomonas Aeruginosa. *ISME J.* 14, 3093– 3105. doi: 10.1038/s41396-020-00741-9
- Camus, L., Briaud, P., Vandenesch, F., and Moreau, K. (2021). How Bacterial Adaptation to Cystic Fibrosis Environment Shapes Interactions Between Pseudomonas Aeruginosa and Staphylococcus Aureus. *Front. Microbiol.* 12, 617784. doi: 10.3389/fmicb.2021.617784
- Chang, C. Y. (2018). Surface Sensing for Biofilm Formation in Pseudomonas Aeruginosa. Front. Microbiol. 8, 2671. doi: 10.3389/fmicb.2017.02671
- Cheung, G. Y. C., Soo Joo, H., Chatterjee, S. S., and Otto, M. (2014). Phenol-Soluble Modulins - Critical Determinants of Staphylococcal Virulence. FEMS Microbiol. Rev. 38 (4), 698–719. doi: 10.1111/1574-6976.12057
- Connell, J. L., Wessel, A. K., Parsek, M. R., Ellington, A. D., Whiteley, M., and Shear, J. B. (2010). Probing Prokaryotic Social Behaviors With Bacterial 'Lobster Traps'. *mBio* 1, e00202–10. doi: 10.1128/mBio.00202-10
- Cornforth, D. M., and Foster, K. R. (2013). Competition Sensing: The Social Side of Bacterial Stress Responses. *Nat. Rev. Microbiol.* 11 (4), 285–293. doi: 10.1038/nrmicro2977
- Darouiche, R. O. (2004). Treatment of Infections Associated With Surgical Implants. N. Engl. J. Med. 14 (350), 1422–1429. doi: 10.1056/NEJMra035415
- Dowd, S. E., Sun, Y., Secor, P. R., Rhoads, D. D., Wolcott, B. M., Garth, A. J., et al. (2008). Survey of Bacterial Diversity in Chronic Wounds Using Pyrosequencing, DGGE, and Full Ribosome Shotgun Sequencing. BMC Microbiol. 8, 43. doi: 10.1186/1471-2180-8-43
- Filkins, L. M., and O'Toole, G. A. (2015). Cystic Fibrosis Lung Infections: Polymicrobial, Complex, and Hard to Treat. *PloS Pathog.* 11, e1005258. doi: 10.1371/journal.ppat.1005258
- Frank, S. A. (1996). Models of Parasite Virulence. Q. Rev. Biol. 71 (1), 37–78. doi: 10.1086/419267
- Gjødsbøl, K., Christensen, J. J., Karlsmark, T., Jørgensen, B., Klein, B. M., and Krogfelt, K. A. (2006). Multiple Bacterial Species Reside in Chronic Wounds: A Longitudinal Study. Int. Wound J. 3 (3), 225–231. doi: 10.1111/j.1742-481X.2006.00159.x
- Harrison, F., Paul, J., Massey, R. C., and Buckling, A. (2008). Interspecific Competition and Siderophore-Mediated Cooperation in Pseudomonas Aeruginosa. *ISME J.* 2 (1), 49–55. doi: 10.1038/ismej.2007.96

- Hoffman, L. R., Déziel, E., D'Argenio, D. A., Lépine, F., Emerson, J., McNamara, S., et al. (2006). Selection for Staphylococcus Aureus Small-Colony Variants Due to Growth in the Presence of Pseudomonas Aeruginosa. *Proc. Natl. Acad. Sci.* U. S. A. 103 (52), 19890–19955. doi: 10.1073/pnas.0606756104
- Hotterbeekx, A., Kumar-Singh, S., Goossens, H., and Malhotra-Kumar, S. (2017). In Vivo and In Vitro Interactions Between Pseudomonas Aeruginosa and Staphylococcus Spp. Front. Cell. Infect. Microbiol. 7, 106. doi: 10.3389/ fcimb.2017.00106
- Hubert, D., Réglier-Poupet, H., Sermet-Gaudelus, I., Ferroni, A., Le Bourgeois, M., Régis Burgel, P., et al. (2013). Association Between Staphylococcus Aureus Alone or Combined With Pseudomonas Aeruginosa and the Clinical Condition of Patients With Cystic Fibrosis. J. Cystic Fibrosis 12 (5), 497–503. doi: 10.1016/j.jcf.2012.12.003
- Ibberson, C. B., and Whiteley, M. (2020). The Social Life of Microbes in Chronic Infection. *Curr. Opin. Microbiol.* 53, 44–50. doi: 10.1016/j.mib.2020.02.003
- Jayakumar, P., Thomas, S. A., Brown, S. P., and Kümmerli, R. (2021). Pseudomonas Aeruginosa Reaches Collective Decisions via Transient Segregation of Quorum Sensing Activities Across Cells. bioRxiv. doi: 10.1101/2021.03.22.436499
- Jennings, L. K., Dreifus, J. E., Reichhardt, C., Storek, K. M., Secor, P. R., Wozniak, D. J., et al. (2021). Pseudomonas Aeruginosa Aggregates in Cystic Fibrosis Sputum Produce Exopolysaccharides That Likely Impede Current Therapies. *Cell Rep.* 34, 108782. doi: 10.1016/j.celrep.2021.108782
- Kessler, E., Safrin, M., Olson, J. C., and Ohman, D. E. (1993). Secreted LasA of *Pseudomonas Aeruginosa* Is a Staphylolytic Protease. J. Biol. Chem. 268 (10), 7503–7508. doi: 10.1016/S0021-9258(18)53203-8
- Kragh, K. N., Hutchison, J. B., Melaugh, G., Rodesney, C., Roberts, A. E. L., Yasuhiko, I., et al. (2016). Role of Multicellular Aggregates in Biofilm Formation. *mBio* 7, e00237. doi: 10.1128/mBio.00237-16
- Lee, J., and Zhang, L. (2014). The Hierarchy Quorum Sensing Network in Pseudomonas Aeruginosa. *Protein Cell* 6 (1), 26–41. doi: 10.1007/s13238-014-0100-x
- Leinweber, A., Weigert, M., and Kümmerli, R. (2018). The Bacterium Pseudomonas Aeruginosa Senses and Gradually Responds to Interspecific Competition for Iron. *Evolution* 72 (7), 1515–1528. doi: 10.1111/evo.13491
- Lenhard, J. R., Smith, N. M., Quach, C. D., Nguyen, T. Q., Doan, L. H., and Chau, J. (2019). Bacterial Brothers in Arms: Cooperation of Staphylococcus Aureus and Pseudomonas Aeruginosa During Antimicrobial Exposure. J. Antimicrob. Chemother. 74 (9), 2657–2665. doi: 10.1093/jac/dkz247
- Li, M., Cheung, G. Y. C., Hu, J., Wang, D., Soo Joo, H., DeLeo, F. R., et al. (2010). Comparative Analysis of Virulence and Toxin Expression of Global Community-Associated Methicillin-Resistant Staphylococcus Aureus Strains. J. Infect. Dis. 202 (12), 1866–1876. doi: 10.1086/657419
- Limoli, D. H., and Hoffman, L. R. (2019). Help, Hinder, Hide and Harm: What Can We Learn From the Interactions Between Pseudomonas Aeruginosa and Staphylococcus Aureus During Respiratory Infections. *Thorax* 74 (7), 684–692. doi: 10.1136/thoraxjnl-2018-212616
- Limoli, D. H., Warren, E. A., Yarrington, K. D., Donegan, N. P., Cheung, A. L., and O'Toole, G. A. (2019). Interspecies Interactions Induce Exploratory Motility in Pseudomonas Aeruginosa. *eLife* 319, e47365. doi: 10.7554/eLife.47365
- Limoli, D. H., Whitfield, G. B., Kitao, T., Ivey, M. L., Davis, M. R.Jr., Grahl, N., et al. (2017). Pseudomonas Aeruginosa Alginate Overproduction Promotes Coexistence With Staphylococcus Aureus in a Model of Cystic Fibrosis Respiratory Infection. *mBio* 8, e00186–17. doi: 10.1128/mBio.00186-17
- Lin, J., Cheng, J., Wang, Y., and Shen, X. (2018). The Pseudomonas Quinolone Signal (PQS): Not Just for Quorum Sensing Anymore. Front. Cell. Infect. Microbiol. 8, 230. doi: 10.3389/fcimb.2018.00230
- Maliniak, M. L., Stecenko, A. A., and McCarty, N. A. (2016). A Longitudinal Analysis of Chronic MRSA and Pseudomonas Aeruginosa Co-Infection in Cystic Fibrosis: A Single-Center Study. J. Cystic Fibrosis 15 (3), 350–356. doi: 10.1016/j.jcf.2015.10.014
- Mashburn, L. M., Jett, A. M., Akins, D. R., and Whiteley, M. (2005). Staphylococcus Aureus Serves as an Iron Source for Pseudomonas Aeruginosa During *in Vivo* Coculture. J. Bacteriol. 187 (2), 554–566. doi: 10.1128/JB.187.2.554-566.2005
- Michelsen, C. F., Khademi, S. M. H., Johansen, H. K., Ingmer, H., Dorrestein, P. C., and Jelsbak, L. (2016). Evolution of Metabolic Divergence in Pseudomonas Aeruginosa During Long-Term Infection Facilitates a Proto-Cooperative

Interspecies Interaction. ISME J. 10 (6), 1323-1336. doi: 10.1038/ ismej.2015.220

- Miller, C. L., Van Laar, T. A., Chen, T., Rajasekhar Karna, S. L., Chen, P., You, T., et al. (2017). Global Transcriptome Responses Including Small RNAs During Mixed-Species Interactions With Methicillin-Resistant Staphylococcus Aureus and Pseudomonas Aeruginosa. *MicrobiologyOpen* 6, e00427. doi: 10.1002/ mbo3.427
- Niggli, S., and Kümmerli, R. (2020). Strain Background, Species Frequency, and Environmental Conditions Are Important in Determining Pseudomonas Aeruginosa and Staphylococcus Aureus Population Dynamics and Species Coexistence. Appl. Environ. Microbiol. 86, e00962–20. doi: 10.1128/AEM.00962-20
- Orazi, G., and O'Toole, G. A. (2017). Pseudomonas Aeruginosa Alters Staphylococcus Aureus Sensitivity to Vancomycin in a Biofilm Model of Cystic Fibrosis Infection. *mBio* 8, e00873–17. doi: 10.1128/mBio.00873-17
- Orazi, G., Ruoff, K. L., and O'Toole, G. A. (2019). Pseudomonas Aeruginosa Increases the Sensitivity of Biofilm-Grown Staphylococcus Aureus to Membrane-Targeting Antiseptics and Antibiotics. *mBio* 10, e01501–19. doi: 10.1128/mbio.01501-19
- Pande, S., Merker, H., Bohl, K., Reichelt, M., Schuster, S., De Figueiredo, L. F., et al. (2014). Fitness and Stability of Obligate Cross-Feeding Interactions That Emerge Upon Gene Loss in Bacteria. *ISME J.* 8 (5), 953–962. doi: 10.1038/ismej.2013.211
- Peters, B. M., Jabra-Rizk, M. A., O'May, G. A., William Costerton, J., and Shirtliff, M. E. (2012). Polymicrobial Interactions: Impact on Pathogenesis and Human Disease. Clin. Microbiol. Rev. 25 (1), 193–213. doi: 10.1128/CMR.00013-11
- Radlinski, L., Rowe, S. E., Kartchner, L. B., Maile, R., Cairns, B. A., Vitko, N. P., et al. (2017). Pseudomonas Aeruginosa Exoproducts Determine Antibiotic Efficacy Against Staphylococcus Aureus. *PloS Biol.* 15, e2003981. doi: 10.1371/ journal.pbio.2003981
- Ramsey, M. M., and Whiteley, M. (2009). Polymicrobial Interactions Stimulate Resistance to Host Innate Immunity Through Metabolite Perception. *Proc. Natl. Acad. Sci. U. S. A.* 106 (5), 1578–1583. doi: 10.1073/pnas.0809533106
- Rezzoagli, C., Granato, E. T., and Kümmerli, R. (2020). Harnessing Bacterial Interactions to Manage Infections: A Review on the Opportunistic Pathogen Pseudomonas Aeruginosa as a Case Example. J. Med. Microbiol. 69 (2), 147– 161. doi: 10.1099/jmm.0.001134
- Schilcher, K., and Horswill, A. R. (2020). Staphylococcal Biofilm Development: Structure, Regulation, and Treatment Strategies. *Microbiol. Mol. Biol. Rev.* 84, e00026–19. doi: 10.1128/MMBR.00026-19
- Schindelin, J., Arganda-Carreras, I., Frise, E., Kaynig, V., Longair, M., Pietzsch, T., et al. (2012). Fiji: An Open-Source Platform for Biological-Image Analysis. *Nat. Methods* 9 (7), 676–682. doi: 10.1038/nmeth.2019
- Schleheck, D., Barraud, N., Klebensberger, J., Webb, J. S., McDougald, D., Rice, S. A., et al. (2009). Pseudomonas Aeruginosa PAO1 Preferentially Grows as Aggregates in Liquid Batch Cultures and Disperses Upon Starvation. *PloS One* 4, e5513. doi: 10.1371/journal.pone.0005513
- Short, F. L., Murdoch, S. L., and Ryan, R. P. (2014). Polybacterial Human Disease: The Ills of Social Networking. *Trends Microbiol.* 22 (9), 508–516. doi: 10.1016/ j.tim.2014.05.007

- Soberón-Chávez, G., Lépine, F., and Déziel, E. (2005). Production of Rhamnolipids by Pseudomonas Aeruginosa. *Appl. Microbiol. Biotechnol.* 68 (6), 718–725. doi: 10.1007/s00253-005-0150-3
- Stoodley, P., Wilson, S., Hall-Stoodley, L., Boyle, J. D., Lappin-Scott, H. M., and Costerton, J. W. (2001). Growth and Detachment of Cell Clusters From Mature Mixed-Species Biofilms. *Appl. Environ. Microbiol.* 67 (12), 5608–5613. doi: 10.1128/AEM.67.12.5608-5613.2001
- Tognon, M., Köhler, T., Gdaniec, B. G., Hao, Y., Lam, J. S., Beaume, M., et al. (2017). Co-Evolution With Staphylococcus Aureus Leads to Lipopolysaccharide Alterations in Pseudomonas Aeruginosa. *ISME J.* 11 (10), 2233–2243. doi: 10.1038/ismej.2017.83
- Tognon, M., Köhler, T., Luscher, A., and Van Delden, C. (2019). Transcriptional Profiling of Pseudomonas Aeruginosa and Staphylococcus Aureus During *in Vitro* Co-Culture. *BMC Genomics* 20 (1), 30. doi: 10.1186/s12864-018-5398-y
- Tsompanidou, E., Denham, E. L., Becher, D., de Jong, A., Buist, G., van Oosten, M., et al. (2013). Distinct Roles of Phenol-Soluble Modulins in Spreading of Staphylococcus Aureus on Wet Surfaces. *Appl. Environ. Microbiol.* 79 (3), 886– 895. doi: 10.1128/AEM.03157-12
- Venturi, V. (2006). Regulation of Quorum Sensing in Pseudomonas. FEMS Microbiol. Rev. 30 (2), 274–291. doi: 10.1111/j.1574-6976.2005.00012.x
- Wang, R., Braughton, K. R., Kretschmer, D., Thanh Huy, L. B., Queck, S. Y., Li, M., et al. (2007). Identification of Novel Cytolytic Peptides as Key Virulence Determinants for Community-Associated MRSA. *Nat. Med.* 13 (12), 1510– 1514. doi: 10.1038/nm1656
- Weigert, M., and Kümmerli, R. (2017). The Physical Boundaries of Public Goods Cooperation Between Surface-Attached Bacterial Cells. Proc. R. Soc. B: Biol. Sci. 284 (1858), 20170631. doi: 10.1098/rspb.2017.0631
- Zheng, Y., He, L., Asiamah, T. K., and Otto, M. (2018). Colonization of Medical Devices by Staphylococci. *Environ. Microbiol.* 20 (9), 3141–3153. doi: 10.1111/ 1462-2920.14129

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