

Substrate stiffness impacts early biofilm formation by modulating Pseudomonas aeruginosa twitching motility

Sofia Gomez Ho, Lionel Bureau, Karin John, Delphine Débarre, Sigolène

Lecuyer

▶ To cite this version:

Sofia Gomez Ho, Lionel Bureau, Karin John, Delphine Débarre, Sigolène Lecuyer. Substrate stiffness impacts early biofilm formation by modulating Pseudomonas aeruginosa twitching motility. eLife, 2023, 10.1101/2022.02.18.480999 . hal-03798581v2

HAL Id: hal-03798581 https://hal.science/hal-03798581v2

Submitted on 10 May 2023

HAL is a multi-disciplinary open access archive for the deposit and dissemination of scientific research documents, whether they are published or not. The documents may come from teaching and research institutions in France or abroad, or from public or private research centers. L'archive ouverte pluridisciplinaire **HAL**, est destinée au dépôt et à la diffusion de documents scientifiques de niveau recherche, publiés ou non, émanant des établissements d'enseignement et de recherche français ou étrangers, des laboratoires publics ou privés.

Substrate stiffness impacts early biofilm formation by modulating *Pseudomonas aeruginosa* twitching motility

- **5** Sofia Gomez², Lionel Bureau², Karin John², Elise-Noëlle Chêne¹, Delphine
- Débarre^{2*}, Sigolène Lecuyer^{1,2*}

⁷ ¹Université Lyon, ENS de Lyon, Université Claude Bernard, CNRS, Laboratoire de
 ⁸ Physique, F-69342, Lyon, France; ²Université Grenoble Alpes, CNRS, LIPhy, 38000
 ⁹ Grenoble, France

Abstract Surface-associated lifestyles dominate in the bacterial world. Large multicellular assemblies, called biofilms, are essential to the survival of bacteria in harsh environments, and 12 are closely linked to antibiotic resistance in pathogenic strains. Biofilms stem from the surface 13 colonization of a wide variety of substrates encountered by bacteria, from living tissues to inert materials. Here, we demonstrate experimentally that the promiscuous opportunistic pathogen 15 Pseudomonas geruginosa explores substrates differently based on their rigidity, leading to striking 16 variations in biofilm structure, exopolysaccharides (EPS) distribution, strain mixing during 17 co-colonization and phenotypic expression. Using simple kinetic models, we show that these 18 phenotypes arise through a mechanical interaction between the elasticity of the substrate and 19 the type IV pilus (T4P) machinery, that mediates the surface-based motility called twitching. 20 Together, our findings reveal a new role for substrate softness in the spatial organization of 21 bacteria in complex microenvironments, with far-reaching consequences on efficient biofilm 22 formation. 23

24

10

25 Introduction

- ²⁶ The transition of bacteria from a planktonic to a surface-attached state is of paramount importance
- ²⁷ in biofilm formation. In consequence, the way bacteria sense and respond to the close proximity ²⁸ of a surface has been the subject of intense scrutiny (*Dufrêne and Persat, 2020: Laventie and Ie-*
- ²⁹ *nal. 2020*). This interaction involves different aspects of bacterial motility: swimming towards the
- ³⁰ surface, but also swarming, gliding or twitching that are used by attached bacteria to explore the
- ³¹ surface collectively or individually (Wadhwa and Berg, 2022; Conrad et al., 2011). Eventually, per-
- ³² manent bacterial adhesion and microcolony structuration may arise, through mechanisms which
- essential ingredients are known (production of matrix, loss of motility), but in response to cues
- 34 that remain unclear.
- Bacteria are ubiquitous and can successfully colonize a wide range of biological tissues and
 abiotic surfaces (*Stoodley et al., 2002; Mann and Wozniak, 2012*). Different environments often re sult in different phenotypes for a given microorganism (*Dötsch et al., 2012; Cornforth et al., 2018*).
 However, although chemical signaling has long been known to impact bacterial gene regulation.
- However, although chemical signaling has long been known to impact bacterial gene regulation,
 it remains unclear how the mechanical properties of the encountered surface might impact bac
 - it remains unclear now the mechanical properties of the encountered surface might impact bac-

*For correspondence:

delphine.debarre@ univ-grenoble-alpes.fr (DD); sigolene.lecuyer@ens-lyon.fr (SL)

- 40 terial behavior (*Persat et al., 2015b*). In this paper, we investigate how the rigidity of a substrate
- 41 modifies bacterial motility, and by doing so impacts microcolony morphogenesis and early biofilm
- 42 development.

Pseudomonas geruginosa (PA) is an opportunistic rod-shaped pathogen that contaminates a 43 wide range of substrates from very soft tissues to rigid implants (Moradali et al. 2017: Change 44 2018). A particularly gifted and versatile biofilm-former, it is extremely prone to developing antibiotic resistance (Pang et al. 2019; Boucher et al. 2009) PA has developed an arsenal of techniques 46 to move on surfaces; among them is twitching motility, that allows single bacteria to translocate 47 across surfaces using type IV pili (T4P) (Majer and Wong, 2015). T4P are thin protein filaments on 48 the bacterial surface that can extend and contract by assembly and disassembly of the protein ٨0 subunit PilA. The tip of T4P acts as a promiscuous hook that can grasp most surfaces. Attachment, 50 contraction, detachment and extension cycles propel bacteria (Merz et al., 2000; Majer and Wong, 51 2015: Skerker and Berg, 2001: Talà et al., 2019). This surface motility is important for bacteria to effi-52 ciently settle on surfaces, but the exact mechanisms at play are unknown (O'Toole and Kolter, 1998: 53 Leighton et al., 2015: Craig et al., 2019). The function of T4P and the fact that it exerts forces on 54 its environment make it an obvious candidate for surface-sensing mechanisms (Merz et al., 2000) 55 Dufrêne, 2015: Sahoo et al., 2016). Recent results have shown that the polar localization of pili in 56 PA could happen in response to surface-sensing (*Cowles and Gitai, 2010*). Polarly-localized pili lead 57 to persistent rather than random displacements, as well as specific effects such as the upstream 58 migration of bacteria submitted to strong flows (Shen et al., 2012). Reversal of twitching bacteria is 50 rapidly induced upon meeting obstacles, suggesting a mechanical feedback from T4P (Kühn et al., 60 **2021**). In addition, PA can exert different types of virulence, from acute attacks to chronic infec-61 tions (Furukawa et al., 2006: Valentini and Filloux, 2016), and specific host-pathogen interactions 62 have traditionally been considered as the key players in the regulation of these virulence pathways 63 (Gellativ and Hancock, 2013). However, surface-sensing in itself has recently appeared as a poten-64 tial signal that could trigger the upregulation of virulence-associated genes (Islam and Krachler, 65 2016: Persat et al., 2015a: Sirvaporn et al., 2014), Although the global effect of surface rigidity on 66 bacterial adhesion and biofilm formation has sometimes been addressed (Saha et al. 2013: Song 67 and Rep. 2014: Song et al. 2018a) so far how the micromechanical environment experienced by individual bacteria impacts their behavior is still unclear, possibly because of the difficulty to design and control microenvironments that allow for a fine tuning of mechanical properties at the 70 bacterial scale, along with negligible changes of the chemical environment. 71 In this study, we use a home-designed microfluidic setup to investigate at the single-cell level 72 the influence of substrate rigidity on PA bacteria adhering to an open surface, under controlled flow 73 conditions. We first demonstrate that substrate elasticity strongly impacts early microcolony de-74 velopment. Focusing on single-cell behaviour, we then study quantitatively how rigidity modulates 75 bacterial motility and propose a purely mechanistic model to account for our observations. Fi-76

- nally, we demonstrate that this mechanical tuning of the motility explains rigidity-induced changes
- ⁷⁸ in early surface colonization: we explore its consequences in terms of microcolony morphology,
- ⁷⁹ matrix deposition, strain mixing and long-term gene expression.

80 Results

In order to explore *in situ* the effect of substrate rigidity on the behavior of adhering bacteria, we
 have developed an experimental approach to include mechanically well-defined hydrogel pads in
 a microfluidic channel providing controlled flow conditions and allowing confocal imaging (Fig. 1A).
 We use the biocompatible hydrogel polyacrylamide (PAA), which has been extensively used to in-

- vestigate cell-substrate interaction and mechanotransduction in mammalian cells. By varying the
- amount of bisacrylamide cross-linker during its preparation, PAA can span a biologically-relevant
- range of rigidities (from ~1 to 100 kPa) while keeping a low viscous dissipation. Several pads, with
- ⁸⁸ different's modulus values ranging from ~ 3 to 100 kPa (see Fig.1-fig. suppl. 1, and Methods and
- ⁸⁹ Materials), are used in each single experiment, and bacteria adhering on PAA and glass surfaces



Figure 1. Bacterial microcolony formation depends on substrate rigidity. (A) Experimental setup: bacteria (*P. aeruginosa* strain PAO1) are imaged in a flow cell under constant flow of minimal medium. (B) After 10h, dense, isolated colonies form on soft PAA (2.7 kPa) while bacteria are more evenly distributed on stiff PAA (84 kPa), closer to what is observed on glass. Scale bars, 20 μ m. (C) Bacterial growth is not impacted by substrate rigidity. (D) 3D reconstruction of colonies confirms their hemispherical shape on soft substrates. (E) Surface coverage is lower on soft substrates, but total volume of colonies is conserved, with a higher roughness value. (F) Fraction of area occupied by the bacteria as a function of the distance from the coverslip, showing flatter colonies on rigid substrates.

⁹⁰ are imaged with high-resolution phase-contrast and fluorescence time-lapse imaging, from very

low surface coverage up to the formation of microcolonies (1 frame/min over ~10 hours).

Substrate elasticity modifies bacterial colonization of PAA in a T4P-dependent man ner

We first focus on the effect of substrate rigidity on early microcolony formation. Straightforward observations with phase-contrast imaging show a striking impact on the shape of microcolonies 95 after a few hours (Fig. 1B and Video 1): on the softest hydrogels (< 10 kPa), bacteria form well-96 defined, dense hemispherical colonies; in contrast, on stiff hydrogels, bacteria are distributed in 97 a thin layer covering most of the surface, a morphology closer to what we observe on glass. To 98 rule out any effect due to changes in the bacterial growth rate, we quantified the division time of 00 bacteria (Fig. 1C), and the volume occupied by bacterial colonies after a few hours (Fig. 1D and E) on 100 different substrates; both were found to be unaffected, suggesting that bacteria develop and colo-101 nize substrates at the same rate irrespective of rigidity, but that the processes that drive their self-102 organization into colonies are modified. In contrast, a change in the morphology of the colonies 103 could be confirmed by quantifying the characteristic roughness of the bacterial layer, which de-104 creases as rigidity increases (Fig. 1E), and the distribution of bacteria with the distance from the 105 surface, which spreads further for soft hydrogels (Fig. 1F). To control that this is a robust phe-106 nomenon driven by substrate elasticity rather than specific chemical interactions, we reproduced 107 this assay using polyethylene glycol (PEG) hydrogels, which are chemically different from PAA but 108

can span a similar range of rigidities. We obtained very similar results regarding the phenotype of

colonies, which further confirms a role for the mechanical properties of the substrate in bacterial 110 self-organization (Fig. 1-fig. suppl. 2). Finally, because shear flow can orient polarly-attached bacte-111 ria, direct motility, disperse quorum-sensing molecules, and generally impact spatial organization 112 into colonies, we carried out experiments on hydrogels immobilized at the bottom of wells, with-113 out any agitation of the above medium. Although long-term observations are rendered difficult in 114 that case by swimming bacteria, we clearly observed the formation of denser colonies on softer 115 PAA (Fig 1-fig_suppl_3) which further demonstrates that substrate stiffness modifies bacterial 116 behavior after attachment in a wide variety of environments. 117 Since surface motility is known to be important for initial self-organization of PA, we hypothe-118 size that it could play a role in the different microcolony shapes that we observe. This link was ex-119 plored by carrying out experiments using a mutant deprived of type IV pili (T4P), and thus unable to 120 twitch on surfaces (mutant PAO1 *pilA*::Tn5. Fig.1-fig. suppl. 4). In these assays, the dependence of 121 microcolony morphology on substrate rigidity is abolished and bacteria form dense hemispherical 122 colonies on all PAA substrates. We therefore conclude that T4P-mediated surface motility ("twitch-123 ing") plays a key role in the rigidity modulation of microcolony formation by WT PAO1 on soft elastic 124

125 substrates.

126 Substrate elasticity modulates twitching motility

127 Experimental results - global motility

To quantify the coupling between the elasticity of the substrate and the twitching motility of bacte-128 ria, we analyzed time-lapse phase contrast images of adhering bacteria in flow cells. These images 129 allow segmentation of individual bacteria (SI subsection I.A) from the start of the acquisition (with 130 a few isolated bacteria per field of view) until the transition to out-of-plane growth, after which 131 individual bacteria cannot be easily separated anymore. From segmented binary images at early 132 imaging stages (<100 min), we obtain the surface coverage A(t) as the fraction of occupied pixels. 133 and the cumulative explored area S(t) as the fraction of pixels that has been explored at time t 134 (Fig. 2A). The evolution of A(t) reflects the exponential growth of initially attached bacteria on the 135 surface, as well as potential attachment and detachment events during the acquisition. However, 136 in our experiments initial surface coverage is extremely low, and at early times the number of bac-137 teria in the clean flowing medium is negligible so that attachment events are rarely observed. We 138

139 can thus consider that

$$\frac{dA}{dt} = (k_{di} - k_{de})A(t). \tag{1}$$

The bacterial division rate k_{di} does not depend on the substrate (Fig. 1C), and was measured for each experiment ($k_{di}^{-1} = 27.8 \pm 1.4$ min). Fig. 2B shows the experimental time evolution of A(t)depending on the gel rigidity, which can indeed be well described by a simple exponential (straight line in the semi-log presentation with slope $k_{di} - k_{de}$). The slope of A(t) slightly increases with gel rigidity, suggesting a higher detachment rate k_{de} on softer hydrogels at early acquisition times.

Compared to A(t), there is a quantitatively much larger dependence of the cumulative surface coverage S(t) on the substrate rigidity (Fig. 2B). We propose that this result directly reflects a change of the global motility V_g of bacteria on the surface. Indeed, neglecting bacterial surface attachment, the evolution of S can be written as

$$\frac{dS}{dt} = k_{di}A + V_g w_b N = (k_{di} + \frac{V_g}{l_b})A.$$
(2)

where *N* denotes the number of bacteria on the surface, and the typical size of a rod-shaped bacterium is $w_b l_b$ (width × length), so that the occupied area is $A = N w_b l_b$. Here we have assumed that bacteria tend to move along their major axis, neglecting reorientations, based on previous findings about the polar localization of T4P (*Cowles and Gitai, 2010; Jin et al., 2011; Kühn et al.,* **2021**) and our own observations. (Considering that bacteria can move in any direction would lower velocity values by a factor $\sqrt{\frac{l_b}{w_b}} \approx \sqrt{3}$, and not change significantly the coming discussion.) The



Figure 2. Bacterial surface motility is impaired on soft hydrogels. (A) Surface explored (dark blue) and current surface coverage (cyan) after 100 min on soft and stiff PAA surfaces. Scale bar: 20 μ m. (B) Surface coverage A(t) (broken lines) and cumulative explored area S(t) (full lines) for all tested rigidities (the initial surface coverage $\langle A(t = 0 - 10min) \rangle$ was normalized to 0.1%). Shaded areas are standard errors of the mean (84 kPa: 6 data sets from 3 independent experiments, 18.5 kPa: 10 data sets from 5 independent experiments, 2.7 kPa: 8 data sets from 4 independent experiments).(C) Global bacterial motility V_g averaged over the first 100 min, inferred from the difference between A(t) and S(t) (16 different surfaces, 6 independent experiments). The black line is the fit with the kinetic model using equation A11. with values $V_{max} = 0.77 \pm 0.35 \mu$ m.min⁻¹ and $E_0 = 84 \pm 68$ kPa. (D) Ingredients of the minimal 1D model for bacterial T4P-powered displacement.

average bacterial size was measured in each experiment ($l_{b} = 2.8 \pm 0.13 \mu m$, $w_{b} = 0.8 \pm 0.13 \mu m$). 155 For each monitored position, we determined dS/dt and A(t) experimentally. The global bacterial 156 velocity V_{a} was then estimated using Eq. 2, by averaging over the first 100 experimental time points. 157 During the first 100 minutes under flow, V_{r} exhibits a clear dependence on substrate elastic-158 ity (Fig. 2C). Motility values are close to zero on very soft substrates (3-6 kPa), and progressively 159 increase to reach $\sim 0.5 \pm 0.25 \mu$ m/min on the stiffest hydrogels tested in this study (84 kPa). We 160 wondered whether this dependence of surface motility on substrate elasticity could result from 161 intracellular modifications in response to bacterial surface-sensing. In P. geruginosa, two main 162 surface-sensing systems have been unveiled so far: the Pil-Chp system that involves T4P retraction-163 mediated force sensing *Webster et al.* (2022), and the Wsp system believed to be activated by cell 164 envelop stress ONeal et al. (2022): both systems have been shown to activate biofilm formation 165 pathways following bacterial adhesion Chang (2018). Sessility and matrix production are promoted 166 by increasing intracellular c-di-GMP levels, which production is catalyzed by diguanylate cyclases 167 (DGCs). We carried out experiments with wspR and sadC mutants, two DGCs known to be involved 168 in the surface-sensing response. Although impaired in c-di-GMP regulation, both mutants still ex-169 hibited a stiffness-dependent twitching motility (Fig.2-fig. suppl. 1). Even though other DGCs might 170 be involved, these initial results suggests that our observations at early timescales could be mainly 171 governed by the mechanical interaction of bacteria with their substrate, with gene regulation play-172 ing a secondary role. This rational motivates the simple kinetic modelling presented next. 173

174 Minimal kinetic modelling

Because a difference in twitching velocity is observed almost immediately upon attachment of bacteria onto the surface, a simple hypothesis could be that a modulation of the twitching efficiency

- arises through mechanical factors the interplay between the T4P extension/retraction mechanism
- and the linear elasticity of the substrate without the need for mechanotransduction mechanisms.
- ¹⁷⁹ To test this minimal hypothesis, we have developed a kinetic model, schematically described on

180 figure 2D.

Briefly (more details can be found in appendix 1), we consider a bacterium adhering onto an elastic substrate with a single effective pilus. The pilus is modelled as a rigid inextensible filament (*Beaussart et al., 2014*) and attaches to the substrate via its extremity with a typical adhesion size λ . This simple choice is motivated by microscopic observations of pilus straightening over its whole

- length during retraction **Talà et al. (2019**) and an estimation of the pilus attachment spot size of
- $\approx 1 \text{ nm}$ from traction force microscopy measurements *Koch et al.* (2022). Note, that multiple at-
- tachments of pili over extended regions to the substrate have also been suggested Lu et al. (2015).
- However, in our simple approach we do not consider this possibility.
- The cell actively retracts its pilus until it detaches from the substrate with the force dependent velocity $v_{\rm R}(F) = v^0(1 - \frac{F}{F_{\rm R}})$ (*Marathe et al., 2014*), where *F* denotes the tensile load on the pilus, $F_{\rm R}$
- the retraction stall force and v^0 the retraction speed at zero load. Assuming linear elasticity, the
- tensile load F is related to the substrate displacement u at the pilus adhesion patch by $F = Yu_{i}$
- where $Y \sim E\lambda$ and E is the Young's modulus of the substrate. Since the typical size of the bacterial
- body l_b is much larger than λ , we neglect the deformation of the substrate induced by the bacterial
- body. Instead we assume that the pilus tension leads to a forward sliding of the bacterial body with
- a linear force-velocity relationship $v_{\rm B}(F) = v^0 \frac{F}{F_{\rm B}}$ (see SI subsection II.B and (*Sens, 2013*)), reducing the substrate deformation and the load in the pilus. Here, the ratio $\eta = \frac{F_{\rm B}}{v_0}$ denotes the mobility constant of the cell on the substrate. With this model, the evolution of the pilus tension *F* is thus
- 199 given by

$$\frac{\mathrm{d}F}{\mathrm{d}t} = Y\frac{\mathrm{d}u}{\mathrm{d}t} = v_{\mathrm{R}}(F) - v_{\mathrm{B}}(F) \tag{3}$$

200 which is solved by

$$F(t) = F_0 \left(1 - e^{-\frac{Y_t 0}{F_0} t} \right)$$
(4)

²⁰¹ with the naturally arising force scale

$$F_0 = \frac{F_{\rm B}F_{\rm R}}{F_{\rm R} + F_{\rm B}} \,. \tag{5}$$

From $\frac{dx_B}{dt} = v_B(F)$ we obtain the bacterial sliding distance during the pilus retraction

(

$$x_{\rm B}(t) = \frac{F_0}{F_{\rm B}} \left[v^0 t + \frac{F_0}{Y} \left(e^{-\frac{Yv^0}{F_0}t} - 1 \right) \right] \,. \tag{6}$$

²⁰³ While retracting the pilus will detach with a rate constant k_{off} from the substrate. Assuming a force-

- independent off-rate constant $k_{\text{off}} = k_{\text{off}}^0$ (and hence a mean pilus adhesion time $(k_{\text{off}}^0)^{-1}$) and a pilus
- retraction frequency $k_{\rm p}$, we obtain an effective bacteria velocity:

$$v_{\rm eff} = k_{\rm p} \langle x_{\rm B} \rangle = k_{\rm p} k_{\rm off}^0 \int_0^\infty x_{\rm B}(t) \, e^{-k_{\rm off}^0 t} \, \mathrm{d}t = V_{\rm max} \frac{E}{E + E_0} \,. \tag{7}$$

Here, $\langle x_{\rm B} \rangle$ denotes the mean bacterial sliding distance per pilus retraction event and $V_{\rm max}$ denotes the maximum effective speed a bacterium can reach on a given substrate at infinite rigidity. It is given by

$$V_{\rm max} = v^0 \frac{k_{\rm p}}{k_{\rm off}^0} \frac{F_{\rm R}}{F_{\rm B} + F_{\rm R}} \,. \tag{8}$$

 E_0 denotes the rigidity at half-maximal speed and is given by

$$E_0 = \frac{F_{\rm B} F_{\rm R} k_{\rm off}^0}{(F_{\rm B} + F_{\rm R}) v^0 \lambda} \,. \tag{9}$$

Fitting (A11) against experimentally measured effective bacterial velocities V_g provides a quantitative description of the data for $V_{max} = 0.77 \pm 0.35 \mu m.min^{-1}$ and $E_0 = 84 \pm 68$ kPa. The error estimates for V_{max} and E_0 were calculated directly from the co-variance matrix of the fit function and the variance of residuals (chi-squared sum divided by the number of degrees of freedom) and are reflective of the wide scattering of measured velocities between different experiments. Conversely,

- we can estimate V_{max} and E_0 from values of the parameters used in the model: assuming a typical
- pilus retraction speed $v^0 = 1 \,\mu\text{m.s}^{-1}$ (*Marathe et al., 2014*), a stall force of the order $F_R = 100 \,\text{pN}$
- (*Marathe et al., 2014; Koch et al., 2022*), a pilus off-rate constant $k_{off}^0 = 1 \text{ s}^{-1}$ (*Talà et al., 2019*), a
- ²¹⁸ contact size of $\lambda = 1$ nm (*Koch et al., 2021*), a high friction surface with $F_{\rm B} = 1$ nN and a typical pilus
- retraction frequency¹ of $k_p = 0.2 \,\mathrm{s}^{-1}$ we obtain $V_{\rm max} \sim 1 \,\mu {\rm m.min}^{-1}$ and a substrate rigidity at half
- maximum speed of $E_0 = 100$ kPa, which are within 30% of the fitted values.

In addition, our model [Eq. (8)] shows that two separate effects translate a (fast) load-free microscopic pilus retraction speed v^0 into a (slow) macroscopic bacterial speed V_{max} . First, the bacterium only translocate during a fraction of the pilus cycle of extension and retraction $\frac{k_p}{k_{off}}$ over the substrate. Second, the pilus retraction speed slows down in a load dependent manner $F_R/(F_B + F_R)$. Both effects together reduce the local speed by an order of magnitude from $um.s^{-1}$ to $um.min^{-1}$.

Together this demonstrates that our experimental results on bacterial effective motility on elastic substrates can be interpreted as the result of a simple interplay between the pilus retraction mechanism, the deformation of the elastic substrate, and the friction of the bacterial body on this substrate.

230 Analysis of individual trajectories

The simple approach presented Fig. 2 yields a population-averaged value of the bacterial velocity 231 $V_{\rm g}$. Yet, bacterial populations can be heterogeneous, and moreover the model we have used Eq. (2) 232 to determine V_{a} relies on a number of strong assumptions, such as neglecting detachment and re-233 attachment events. To go further in dissecting bacterial motility on PAA substrates, we developed a 234 segmentation and tracking protocol in order to obtain the individual tracks of all the bacterial cells 235 visible over the course of the acquisiton (Fig. 3A, Video 2 and Methods and Materials for details). 236 This thorough approach allows us to measure the velocity associated with each 1-min displacement 237 step. 238

Can a characteristic twitching velocity be defined for all bacteria? or do phenotypically distinct 230 populations of slow and fast bacteria cohabit on the surface? To answer these questions, we la-240 beled each track, defined as the displacement of a bacterium between two successive division 241 events (Fig. 3A). We expected track duration to be similar to the characteristic division time shown 242 on Fig. 1C. However, we obtained a bimodal distribution, with two peaks centered at times unaf-243 fected by the substrate rigidity: one peak indeed centered on the division time (~ 27 min), and a second one that corresponds to bacteria spending 5 to 10 min on the surface before detaching 24 (Fig.3-fig. suppl. 1A). The velocity distribution corresponding to each population is similar (Fig.3-fig. suppl. 1B). This observation is consistent with a phenotypical difference between daughter cells. 247 in agreement with the results of (Laventie et al., 2018) Indeed, the duration of the track before detachment tends to be shorter on soft substrates, but the fraction of bacteria that detach from 249 the substrate (35 + 2.%) is independent of the substrate rigidity. We thus assume that at early ex-250 perimental timepoints, after moving in sync with the first daughter cell, the second one sometimes 251 detaches from the substrate (about 70% of the time). This feature might change at later timepoints 252 - bacterial tracking was interrupted at the onset of 3D spatial organization (~100 min on the softest 253 hydrogels, see next section for a full description). 254

Considering only adhering offsprings, for which full tracks were recorded, we normalized tracks 255 with respect to their initial position, yielding Fig.3A. These homogeneous radial distributions con-256 firm that shear does not influence bacterial orientation in our experiments. As expected, track 257 extension becomes larger as substrate rigidity increases. The distribution of the mean velocity of 258 tracks does not allow us to distinguish different bacterial sub-populations; it reaches higher values 250 on stiffer hydrogels, but it is broad, continuous with an exponential decay (reflecting the diversity 260 of behaviours expected in a population of cells) (Fig. 3B). In addition, for each track the standard 261 deviation of this mean velocity is comparable and proportional to its mean (Fig.3-fig. suppl. 2), 262

¹Here we assume that one single effective pilus is active during a retraction event. Using a typical pilus length of 5 μ m with retraction speed of $v_0 = 1 \mu$ m.s⁻¹ we obtain a duration of 5 s per retraction and a retraction frequency of 0.2 s⁻¹



Figure 3. Twitching motility depends on substrate rigidity and is highly distributed in the bacterial population. Analysis of full tracks (top row) and 1-min displacement steps (bottom row). (A) Individual bacterial tracks on soft (2.7 kPa), intermediate (18.5 kPa) and stiff (84 kPa) PAA during the first 3 hours after bacterial inoculation (total number of tracks is respectively 60, 123 and 175). Scale bar: 10 μm (B) Mean track velocity distribution for different values of the substrate rigidity. Only full tracks were considered (corresponding to the right peak in fig. suppl. 1). 84 kPa: 330 tracks from 2 independent experiments, 18.5 kPa: 394 tracks from 3 independent experiments, 2.7 kPa: 83 tracks from 2 independent experiments. (C) Normalized velocity distributions for the whole bacterial population on different PAA surfaces. The exponential decrease yields a characteristic active velocity V_{C} on each substrate. Displacement steps were measured every minute for 100 minutes, and two positions were acquired on each rigidity. The average of T4P-defective mutant on all surfaces is shown as a reference. (D) Characteristic active velocity values V_{C} obtained by fitting velocity distributions (6 independent experiments, 16 different surfaces). Values measured on different surfaces in a single experiment (same channel) are shown with the same symbols and connected. Black squares(circled stars) are mean values obtained for the WT(pilA mutant), and error bars show the SEM. The black line is the fit of WT values with the kinetic model-derived equation A11, with $V_{\text{max}} = 0.48 \pm 0.12 \mu \text{m.min}^{-1}$ and $E_0 = 32 \pm 30 \text{ kPa}$. (E) Average velocity values of the top 5% fastest displacement steps for different substrates. Error bars are standard errors.

²⁶³ suggesting a stochastic distribution of twitching steps within a given trajectory.

Focusing next on 1-min displacement steps, and pooling all monitored events during the first 100 min of experiments (which provides more data than only analysing full tracks within the same time window), we obtained the typical velocity distributions shown on figure 3C. These distributions further confirm that bacterial displacements are very heterogeneous, and present an exponentially decreasing tail which fitting yields a characteristic velocity V_c for the bacterial population on a given

269 substrate, i.e.:

$$N(V > V_0) = N_0 \exp\left(-\frac{V - V_0}{V_C}\right),$$
(10)

 $_{270}$ ($V_0 = 0.08 \,\mu$ m/min denotes a visual cutoff for low velocities).

To rationalize the meaning of V_{CI} we reasoned that displacement steps are the sum of a passive 271 velocity due to bacterial elongation, local reorganisations and experimental noise, and an active ve-272 locity powered by T4P. The velocity distribution obtained using a *pilA* mutant is purely exponential. 273 and was used to determine the characteristic passive motility, which does not significantly depend 274 on substrate rigidity ($V_C(pilA) = V_{C,p} = 0.044 \mu$ m/min, see Fig. 3D). In the case of motile strains, as-275 suming that active and passive displacements are incoherent, our numerical calculations (Fig. 3-fig. 276 suppl. 3) show that in the limit $V_C > V_{C,n}$, the fitted characteristic velocity V_C obtained as described 277 above reflects the active twitching motility of the population, and is not significantly affected by 278 passive movements. A detailed justification for our analysis of the probability distributions of dis-279 placement steps is given in the Methods section. 280

This analysis, which does not rely on any strong assumption, yields active velocity values for the 281 WT strain (Fig. 3D) in very good qualitative agreement with the global velocity analysis described 282 earlier (V_{a} , Fig. 2C, and Fig.3-fig. suppl. 4). Again, our kinetic model describes the data quantita-283 tively with values very close to the ones fitted and calculated in the previous subsection (V_{max} = 284 $0.48 \pm 0.12 \mu m.min^{-1}$ and $E_0 = 32 \pm 30 k Pa$). The large error bars on fitting parameters reflect the 285 dispersion of experimental measurements, despite our efforts to reproduce identical experiments. 286 However, velocity values measured in a given experiment always exhibit a similar dependence on 207 substrate rigidity, i.e. a clear increase of motility as rigidity increases. In addition, we characterized 288 the velocity of the 5 % fastest bacterial displacement steps (Fig. 3F) on each type of substrate. This 280 analysis confirms the dependence of twitching velocity on substrate rigidity, but also yields higher 200 velocity values, in good quantitative agreement with those reported in the literature using other 201 experimental approaches (*Talà et al.*, 2019), which might be biased towards more active bacte-202 ria. Finally, this approach was used to quantify bacterial motility in experiments on PEG hydrogels 203 (shown Fig. 1 - fig.supp.), which confirms the very similar bacterial behaviour on the two kinds of 294 substrates we have used (Fig. 3-fig. suppl. 5). 205

Rigidity-modulated bacterial motility governs the spatial characteristics of early surface colonization

²⁹⁸ In-plane to 3D transition of emerging colonies

To understand the way rigidity-modulated bacterial motility impacts the process of microcolony 299 formation, we studied in details the way colonies transition to out-of-plane growth. Several ex-300 perimental and theoretical approaches have been developed in the past to decipher this process: 301 for confined colonies, the switch from planar to 3D growth takes place when it becomes energeti-302 cally too costly to push neighboring cells outwards. In that case, the adhesion forces between the 303 bacteria and their underlying substrate play a key role: strongly adhering bacteria transition to 304 3D colonies earlier in their development (Duvernov et al., 2018; Grant et al., 2014). In our exper-305 iments, there is no strong vertical or lateral confinement: bacteria can move on the substrate or 306 away from it, so that cells stemming from a given progenitor do not necessarily stay in contact with 307 each other. However, the twitching velocity determines how much cells, on average, move away 308 from one another between two successive division events, thereby creating space to accommodate 300 new offsprings on the surface. 310

To investigate the possible link between twitching motility and 2D to 3D transition of growing 311 microcolonies, we sought to determine N_{i} , the number of adhered cells in a progeny (i.e. stemming 312 from successive divisions of a given bacterium) when the 2D to 3D transition takes place, as a 313 function of the twitching velocity. For softer substrates, all bacteria can be imaged, and $N_{\rm c}$ is directly 314 measured; we also determined the average number of colonies per unit area. On stiffer substrates. 315 it is impossible to track all bacteria stemming from a mother cell, since they are very motile and 316 sometimes move out of the field of view. We assume that bacteria from other progenies are equally likely to move inside the field of view, so that measuring the number of bacteria on the image at 318 t. (the time at which transition to 3D is first observed), divided by the average microcolony density 310 determined on softer substrates earlier gives a good approximation of N. Fig. 4A shows N as a 320 function of the center-of-mass characteristic velocity V_C determined above (Fig. 3D), on different 321 substrates and for 9 different experiments. $N_{\rm c}$ consistently increases with the twitching velocity, 322 indicating a strong correlation between the twitching efficiency and the shape of early colonies and 323 shedding light on our initial observations of variations in microcolony morphology as a function of 324 the substrate rigidity (Fig. 1B, D-F). 325

To decipher the link between N_c and V_{Cl} we have built a simple kinetic model with a single 326 unknown parameter (see appendix 2 for details). Briefly, we assume that the 2D to 3D transition 327 takes place when the area occupied by bacteria reaches a fraction of the equivalent "microcolony 328 size", defined as the characteristic area explored by bacteria in a progeny. Assuming that bacteria 329 explore the surface through a random walk with persistence (Marathe et al., 2014), the charac-330 teristic area accessible to bacteria in a microcolony over time can be written as $a(t) = a_0(1 + \alpha V_c t)$ 331 where a_0 is the area of one bacterium and α is a parameter related to the properties of the random 332 walk. Our experimental data show that not only the velocity, but also the contour length of the 333 trajectories of bacteria increases with the rigidity since the duration of these trajectories is mostly 334 constant (Fig. 4B and Fig. 3-fig. suppl. 1). Area a(t) is related, but not equal, to the area over which 335 the microcolony spreads. Indeed, bacteria are not evenly distributed within the microcolony area. 336 and we observe strong local density fluctuations. If we now consider an exponential growth of the 337 number of bacteria on the surface due to the balance of cell division and detachment, it follows 338 that the increase in the number of cells, and hence the area required to accommodate these cells 339 on the surface, grows faster than the accessible area, driving a transition to 3D growth. Expressing 340 the number of cells $N_{\rm c}$ in the microcolony at the time when this transition stochastically occurs 341 leads to the following dependence as a function of V_c : 342

$$N_c = 1 + \gamma V_C \log(N_c) \tag{11}$$

r is an unknown parameter related to the properties of the random walk and the growth rate of 343 bacteria on the surface that can be measured for each experiment. On figure 4A, we have plotted 344 the corresponding curve using the average of experimental values for γ (solid line) + their standard 346 deviation (dotted lines). We observe an excellent agreement between this simple kinetic model and 346 our experimental data over a wide range of velocities, including the T4P deficient mutant and the 347 WT strain adhering on glass. This hints that elasticity is a key factor shaping the organization of early 348 colonies on elastic substrates, and that it is the main determinant of the colony shapes observed 349 in our experiments on chemically identical substrates with varying rigidities (rather than energy 350 minimization whereby bacteria would either favour adhesion to other cells or to the substrate 351 depending on its rigidity). 352

353 Surface decoration by extracellular matrix

One consequence of the modulation of twitching efficiency by the substrate elasticity could be a

variation in matrix distribution on the surface. Indeed, P. aeruginosa can secrete an extracellular

matrix mostly composed of exopolysaccharides (EPS), which was shown to result in the deposi-

tion of "trails" by twitching bacteria on glass substrates. By mediating the attachment of the cell

³⁵⁸ body to the underlying substrate such deposits are inferred to facilitate further colonisation by



Figure 4. Spatial structuring of surface colonization is impacted by substrate rigidity through twitching velocity. (A), size of microcolony (in number of bacteria N_c) at the 2D to 3D transition as a function of the center-of-mass characteristic velocity V_C defined in Fig. 3A. Markers are experimental data from 9 different experiments, each with different substrates including glass (hence leading to higher values that in Fig. 3B). Blue dots are data obtained with the pili-deficient mutant pilA :: Tn5. Lines, kinetic model for $\langle \gamma \rangle$ (solid line) \pm its standard deviation (dotted lines). (B) Distribution of track lengths of full trajectories as a function of the substrate rigidity. (C) Distribution of occupation occurrence on each image pixel as a function of rigidity, showing a much more heterogeneous occupancy on soft substrates. (D), ConA staining (red) of EPS deposition during cell (green) exploration of the surface.

- bacteria and to impact microcolony formation (Liu et al., 2007; Zhao et al., 2013). To investigate
- matrix deposition on hydrogel substrates, we introduced a fluorescent dye (lectin concanavalin A,
- 361 see Methods and Materials) in the nutrient medium infused in our device. The main component
- of PAO1 matrix, psl (*Jackson et al., 2004*), is rich in mannose, that conA specifically binds (*Ma et al.,*
- 363 **2007**).

For high stiffness substrates where bacteria explore the surface efficiently, this staining confirms 364 that trails of matrix decorate a significant fraction of the surface; on the contrary, nearly immo-365 bile bacteria on soft substrates accumulate matrix locally, leaving most of the surface unmodified 366 (Fig.4C and D). This difference in matrix distribution is maintained at a later stage of surface colo-367 nization (Fig.4-fig. suppl. 1). While on rigid hydrogels, most of the surface is covered by bacteria-368 secreted matrix, lectin staining on soft hydrogels is only present on compact colonies separated 369 by regions completely devoid of EPS. While proper quantification of the total amount of matrix 370 produced in each case is difficult (staining efficiency might be impacted by lectin diffusion inside 371

- dense colonies), our results confirm that substrate rigidity impacts bacterial propensity to modify
- ³⁷³ hydrogel substrates via matrix deposits.

374 Substrate rigidity affects bacterial mixing

Real-life biofilms generally comprise several species: pathogens can compete or help each other (*DeLeon et al., 2014*: *Orazi et al., 2017*), and commensal strains protect organisms from detrimen-

- tal ones (*Aoudia et al., 2016*). To further investigate how modulation of surface colonization with
- ³⁷⁸ rigidity impacts the structure of forming biofilms, we studied the model co-colonization of hydro-
- ³⁷⁹ gels by two PAO1 strains constitutively expressing two different fluorescent proteins. Beside their
- ³⁸⁰ fluorescence, the two strains exhibit identical properties (motility, division rate, etc.), similar to
- that of WT PAO1. Through fluorescent confocal imaging, the strains were spectrally separated to



Figure 5. Bacterial spatial distribution is impacted by substrate rigidity. (A) Images of surfaces seeded with a 1:1 mixture of constitutively fluorescent bacteria expressing GFP or YFP show mostly monoclonal colonies on soft hydrogels, and mixed bacteria on rigid substrates (3D-rendering obtained by stacking images). Scale bar, 20 μm .(B) Spatial correlations quantified via Moran's I index.

study their spatial distribution at different stages of surface colonization. As expected, rigidity-382 modulated motility impacts the co-colonization of the hydrogels from early stages (Fig. 5-fig. suppl. 383 1): on rigid substrates, high motility promotes mixing of the offsprings of different cells, resulting in 384 a spatial distribution of the two strains close to random (a residual correlation between the colour 385 of neighbouring cells is always found due to the presence of cells that have just divided). Con-386 versely, nearly immobile cells on soft substrates exhibit strong correlations between neighbouring 387 cells, which mostly arise from a single progenitor cell. This striking difference in strain mixing dur-388 ing surface co-colonization is maintained at later stages of biofilm formation: on soft substrates. 389 guasi-monoclonal colonies with complete spatial segregation are observed, while biofilms forming 390 on rigid surfaces exhibit a close-to-random distribution of the two strains at the 10-um scale (Fig 391 5A). To quantify this effect, we have used Moran's Lindex, a statistical tool designed to quantify 392 the spatial clustering of species. It provides a measure of the local spatial correlations and takes 393 values ranging from 1 (perfectly correlated values) to -1 (perfectly anti-correlated values), with 0 394 corresponding to a spatially random distribution of the variable (see Methods and Materials for 395 details). The resulting quantitative analysis (Fig. 5B) confirms the decisive impact of rigidity on 396 the structure of mixed biofilms, with potentially far-reaching consequences on the interactions of 397 different strains in multi-species biofilms. 398

399 Surface rigidity impacts gene expression

Cell-cell communication, either via exported molecules or by direct contact is crucial during biofilm 400 development (Shrout et al., 2011). Modifications of bacterial distribution as described above could 401 thus likely impact gene regulation in surface-attached bacteria. To start addressing this complex 402 auestion, we focused on the expression level of cyclic-di-GMP (c-di-GMP), a second messenger that 403 controls the motile-to-sessile transition in *P. geruginosa* (*Rodesnev et al., 2017*). We used a posttranscriptional fluorescent reporter build on the promoter of the gene *cdrA*, which encodes an 405 exported protein involved in matrix cohesion, upregulated during biofilm formation by PAO1 (Re-406 *ichhardt et al., 2018*). The PcdrA-gfp intracellular reporter provides a measure of the integrated 407 production of CdrA with a \sim 40 min delay between expression of the gene and fluorescence detec-408 tion (*Rybtke et al., 2012*). Fig. 6 shows how *crdA* expression is modulated by rigidity on 4 substrates 409 included in the same microfluidic device. For this reporter, the degradation rate of gfp occurs over 410 several hours, and its dilution due to growth and division of bacteria occurs at the same rate on 411 all surfaces (see Fig. 1C). The increase rate of the fluorescence signal is thus a direct proxy to the 412 expression rate of gene cdrA, and thus to the changes in c-di-GMP level. 413



Figure 6. Substrate rigidity influences gene expression levels (A) Colonies grown from a modified PAO1 strain bearing the plasmidic $P_{cdrA} - gfp$ reporter imaged in phase contrast (top) and fluorescence (bottom), after 500 min under flow. Scale bars, 20 μ m. (B) Average intracellular gfp fluorescence as a function of time, on different PAA surfaces and on glass. Broken lines are duplicate positions on a given surface. (C) Average cdrA expression level, obtained from a linear fit of (B).

During a first phase of surface colonization, fluorescence remains low on all surfaces. The sig-414 nal subsequently starts increasing linearly, roughly at the same time for all surfaces (within the 415 uncertainty of fluorescence quantification, i.e. ≈ 10 minutes). This second phase ends with the on-416 set of a plateau, again around the same time for all surfaces, at the end of the exponential growth 417 of bacteria adhered on the surface, possibly as a result of oxygen depletion in the flowing medium 418 that would be sensed simultaneously on all surfaces (Fig.6-fig. suppl. 1). This linear increase in flu-410 orescence directly translates into a constant production rate of CdrA that can be compared for the 420 4 surfaces (Fig. 6C): our analysis shows a marked increase in CdrA expression with the substrate 421 rigidity. 422

423 Discussion

In this study, we have designed an experimental approach to investigate early microcolony forma-424 tion by *P. geruginosa* on hydrogels with different elastic moduli, under constant flow rate. By con-425 tinuously imaging surface-attached bacteria in situ, we show that substrate rigidity influences the 426 twitching motility of individual bacteria, therefore strongly impacting the process of microcolony 427 formation. Through two different analyses of the surface motility of the bacterial population, ei-428 ther via the global evolution of the explored area or via the tracking of individual cells, we find that 429 the characteristic twitching velocity increases with substrate stiffness (from 0.02 to 0.4 μm/min 430 when rigidity goes from \sim 3 to 80 kPa). 431 The encounter between bacteria and a substrate generates mechanical stress. Deciphering surface-432 sensing, i.e. understanding how the mechanical feedback resulting from this interaction translates 433 into chemical signals that will in turn tune bacterial behavior has been the focus of a lot of recent 434 research. It is now clear, for instance, that T4P contraction acts as a force sensor that transmits 435 signals to the bacterium at the single-cell level (Webster et al., 2022), and triggers a response that

signals to the bacterium at the single-cell level (*Webster et al., 2022*), and triggers a response that
 involves an increase in c-di-GMP level (*Armbruster et al., 2019*). Furthermore, recent results sug-

13 of 40

- 438 gest that pili can differentiate substrate rigidity, yielding a maximum response for stiffness values
- 439 ~300 kPa *Koch et al.* (2022).
- However, in our experiments a difference in bacterial motility can be observed almost immediately
- upon surface adhesion to soft or rigid PAA, and this behavior is not modified in mutants impaired
- in c-di-GMP regulation (wspR or sadC). We thus propose a physical model to account for the mod-
- ulation of the twitching motility. This 1D model is based on a force balance between (i) a pilus
- that extends, attaches and retracts with a defined frequency; (ii) the deformation of the underlying
- substrate at the pilus tip upon retraction; and (iii) the friction force due to adhesion of the bacterial body when it is dragged across the surface at the other end of the pilus. In this balance, the
- detachment rate of the pilus tip from the substrate is a key parameter in the resulting bacterial
- velocity. We have assumed a force-independent off-rate constant for the pilus. In a more complex
- scenario, the contact between the pilus and substrate may act as a slip bond or a catch bond. For
- 450 completeness we show some numerical results for slip and catch bond behavior in the SI (section
- 451 I.D), which do not increase however the quality of fit between experimental and theoretical velocity
- data. In addition, although we have explored the possibility that substrate rigidity, which is directly
- 453 correlated to the mesh size of the hydrogel network, could impact the frequency of attachment of
- T4P, this was not necessary to efficiently account for the variation in motility we observe, which we
- instead solely attribute to the elastic deformation of the substrate.
- ⁴⁵⁶ Strikingly, our minimal mechanistic model thus suggests that a variety of observed phenomena ⁴⁵⁷ (3D structure of colonies: EPS deposition on the surface: strain mixing during co-colonization) can
- 457 (3D structure of colonies; EPS deposition on the surface; strain mixing during co-colonization) can 458 all derive from a modulation of the efficiency of pili activity by the deformability of soft substrates.
- This purely mechanical model may be of particular importance for surface colonization, since the

adaptation of bacterial behaviour to the environment can thus be instantaneous - possibly a key

to PA ability to efficiently colonize extremely different microenvironments.

While this model is sufficient to account for our observations (twitching velocity, microcolony formation), it certainly does not rule out a regulatory response of the bacteria, which probably takes

- ⁴⁶⁴ place in parallel. Such a response can happen on two levels: at the single cell level, mechanotrans-
- duction processes mediated for instance by adhesion and retraction of T4P can influence gene
- expression at short timescales (%1 hour) *Armbruster et al.* (2019); Song et al. (2018b). At longer timescales, in developing microcolonies, cell-cell interactions could in turn modulate the bacterial
- timescales, in developing microcolonies, cell-cell interactions could in turn modulate the bacterial transcriptome, which depends on microcolony characteristics (e.g. shape, cell density, matrix con-
- tent) *Livingston et al.* (2022). Our attempt at quantifying c-di-GMP expression using a fluorescent
- intracellular reporter does evidence a difference in bacterial regulatory response depending on
- substrate stiffness. While the level of expression of the gene is clearly impacted by the substrate rigidity, differences in expression level are detected only 6-7 hours after the onset of surface col-
- onization, with a first phase characterized by low c-di-GMP level on all surfaces. This timeframe
- suggests that the difference in gene expression that we observe is probably not due to a direct sensing of the substrate rigidity by individual bacteria, but rather a consequence of their organiza-
- 475 sensing of the substrate rigidity by individual bacteria, but rather a consequence of their organiza-476 tion into more or less dense colonies. Of note, when the increase in c-di-GMP takes place bacteria
- have stopped twitching and immobilized into colonies. We do not observe the early increase in
- c-di-GMP described in the litterature, possibly because we initially only track a very small number
- of bacteria on the surface, and the expression signal is stochastic. Further investigating c-di-GMP expression in WT and mutant strains upon adhesion to mechanically different substrates could
- help reveal which pathways are differently activated on soft substrates.

482 Interestingly, our results show that microcolony phenotype may not be indicative of a specific c-

- di-GMP regulation. The dense colonies observed on soft hydrogels correspond to lower c-di-GMP
- levels than the flat bacteria carpets that grow on rigid substrates, a somehow counter-intuitive
- result given the paradigm that c-di-GMP production upregulates biofilm-inducing genes, in partic-
- ⁴⁸⁶ ular matrix production, while downregulating motility. Here, we describe a case when motility is ⁴⁸⁷ rendered impossible by the micromechanical properties of the environment rather than by the
- ⁴⁸⁷ rendered impossible by the micromechanical properties of the environment rather than by the ⁴⁸⁸ absence of functional pili, thus resulting in the rapid formation of compact colonies on soft sub
 - absence of functional pill, thus resulting in the rapid formation of compact colonies on soft sub-

- strates. Further exploring the density and the exact composition of the extracellular matrix in
- these colonies would be interesting since this parameter could influence the subsequent fate of
- bacteria on the surface. EPS distribution, composition and concentration may also be significant
- for the recruitment of new cells on the surface: indeed, previously deposited matrix is thought to
- 493 strengthen adhesion of *P.aeruginosa* bacteria (*Zhao et al., 2013*), and could also possibly mediate
- adhesion of other microorganisms.
- In a wider context, the process we observe could also be envisioned as a strategy to optimize bacte-
- rial colonization of mechanically heterogenous environments by ensuring accumulation of bacteria
- into dense colonies located in the softer regions of their environment, e.g. over cellular tissues. Re-
- cently, Cont et al. have shown that dense colonies were able to deform soft substrates and exert forces that could disrupt an epithelium laver (*Cont et al., 2020*): rigidity-modulated twitching could
- forces that could disrupt an epithelium layer (*Cont et al., 2020*): rigidity-modulated twitching could thus provide *Pseudomonas geruginosa* with a convenient means of targeting soft tissues for coop-
- erative disruption and subsequent invasion.
- ⁵⁰² The phenotypic differences that we report for colonies are likely to impact subsequent interactions
- ⁵⁰³ of bacteria with their environment: response to changes in nutrient or oxygen availability, and
- ⁵⁰⁴ chemical signals in general which will not efficiently penetrate inside dense colonies. This could in
- particular influence susceptibility to antibiotics, as confirmed by very recent work *Cont et al.* (2023).
- This is all the more relevant that PA can invade many different environments, and might have to
- ⁵⁰⁷ be treated differently when it settles in the lungs of cystic fibrosis patients, or on the surface of ⁵⁰⁸ rigid implants.
- ⁵⁰⁹ Finally, our data show that rigidity-modulated twitching has a striking impact on the mixing of
- ⁵¹⁰ different strains upon surface colonization. Understanding the mechanisms governing the forma-
- tion of mixed-species communities is one of the key challenges of current biofilm research. Since
- the motility modulation mechanism described here is quite general and should be marginally af-
- fected by the particulars of different strains/species moving through elongation/retraction of an
- appendage, we expect it to provide a relevant framework to study co-colonization in different me-
- ⁵¹⁵ chanical micro-environments.

516 Methods and Materials

517 Bacterial strains

518 Strains used in this study were Pseudomonas aeruginosa wild-type (WT) PAO1, fluorescent strains

PAO1 miniCTX-PX2-gfp and PAO1 miniCTX-PX2-eyfp (a kind gift from PBRC group, IBS, Grenoble,

- ⁵²⁰ unpublished), and PAO1 mutants pilA::Tn5, sadC::Tn5 and wspR::Tn5 obtained from the transpo-⁵²¹ son library at University of Washington (*Jacobs et al., 2003*). Strain PAO1 pCdrA-gfp was obtained
- by transforming plasmid $pCdrA::gfp^{C}$ from (*Rvbtke et al., 2012*) in our WT strain.
- Bacteria were inoculated in Luria-Bertani (LB) medium from glycerol stocks, and grown overnight at
- $_{524}$ 37 °C at 250 rpm. The next morning, 10 μ L of the stationary phase culture were diluted in 3 mL of LB
- medium and placed in a shaking incubator (37 °C, 250 rpm) for 3.5 hours, to reach mid-exponential
- phase (OD₆₀₀ = 0.6-0.8). Bacteria were then diluted to OD₆₀₀ = 0.005 in our working medium, TB:PBS,
- and inoculated into the channel. TB:PBS is obtained by mixing TB (Tryptone broth, Euromedex,
- ⁵²⁸ 10 g.L⁻¹) and PBS (w/o calcium and magnesium) with a volume ratio of 1:2. We found that this
- ⁵²⁹ minimal medium favors bacterial twitching for a few hours after adhesion.

530 Microfluidic device

- ⁵³¹ Microfluidic channels were cut into 100 μ m-thick double-sided sticky tape (7641W #25, Teraoka, ⁵³² Japan) with a die-cutter. Typically, a 5 cm-long x 1 mm-wide channel was used to bind together
- a rectangular glass coverslip bearing the hydrogel patches, and a flat 5 mm-thick slab of poly-
- ⁵³⁴ dymethylsiloxane (PDMS, Sylgard prepared by mixing crosslinker and monomer solutions 1:10 and
- baking at 65°C for 1 hour). Two channels were stuck together to obtain a height of 200 μ m, in order
- for the flow through the channel to not be significantly modified by the 25 μ m-thick PAA hydrogels.

⁵³⁷ These sticky-tape channels were first adhered onto the PDMS piece and then placed onto the de-

⁵³⁸ hydrated hydrogels. To ensure proper binding, the whole device was placed under vacuum for 30

minutes. Next, the channel was rinsed with TB:PBS (1:2) for a minimum of 1 hr, in order to rehy-

drate the hydrogels. Medium was placed in a plastic container and withdrawn into the channel

with a syringe pump (Harvard Apparatus, USA, 30 μ L/min) to avoid the formation of bubbles.

Gels and substrates preparation

Hydrogels of polyacrylamide (PAA) and polyethyleneglycol (PEG) were prepared following previ-543 ously established protocols (Tse and Engler, 2010; Beamish et al., 2010). All reagents were ob-544 tained from Sigma Aldrich and used as received: Acrylamide solution (AA, 40% in water), N.N'-545 Methylenebisacrylamide (Bis, 2% in water), Ammonium Persulfate (APS, > 98%), N.N.N'.N'- Tetram-546 ethylethylenediamine (TEMED, > 99%). Poly(ethylene glycol) diacrylate (PEGDA, $M \sim 6000$ g.mol⁻¹). E 4 7 2-Hydroxy-4'-(2-hydroxyethoxy)-2-methylpropiophenone (Irgacure 2959, 98%), Bind-silane, Sigma-548 cote. 549 Rectangular glass coverslips $(24 \times 60 \text{ mm})$ were used as substrates for gel casting. They were 550 plasma-cleaned and immersed in a solution of Bind-silane (60 uL of Bind-silane, 500 uL of 10% 551 acetic acid, 14.5 mL of ethanol) for 1 hour before being rinsed with ethanol, water, and blow-dried 552 with nitrogen before use. Round glass coverslips (12 mm diameter) were used as counter-surfaces 553

for gel casting. After plasma cleaning, they were immersed in Sigmacote for 1 hour before rinsing with acetone, ethanol and water, and blow-dried before use.

Bulk solutions of AA/Bis and PEGDA were prepared in phosphate buffer saline (PBS) and stored 556 at 4 °C until use. The final stiffness of the gels was tuned by adjusting the AA/Bis or PEGDA content 557 according to Table 1. PAA gels were obtained by adding 1 μ L of TEMED and 1 μ L of a freshly made 558 APS solution (10 w% in water) to a volume of 168 μ L of AA/Bis solution. A 3 μ L droplet of the mix-559 ture was immediately placed on the surface of a bindsilane-treated glass coverslip, sandwiched by 560 a Sigmacote-treated round coverslip, and left for curing for 1 hour in a water vapor-saturated atmo-561 sphere. After curing, the round coverslip was lifted off using the tip of a scalpel blade, resulting in a 562 circular pad of gel of thickness $\sim 25 - 30 \,\mu\text{m}$ covalently bound to the bottom rectangular coverslip 563 and exposing its free top surface. Circular gel pads were then scrapped with a razor blade in order 564 to adjust their lateral size to the width of the microfluidic channels into which they would eventually be installed. Gel pads were then copiously rinsed with ultrapure water, and left for drying in a 566 laminar flow cabinet. Up to three such pads, with different elastic properties, were prepared simul-567 taneously on the same coverslip, arranged to fit along the length of the microfluidic channel, PEG 568 gels were obtained by adding 5 µl of a 10 wt% solution of Irgacure in ethanol to 0.5 ml of PEGDA 560 solution. A 3 µL droplet of the mixture was placed in between coverslips as described above, and 570 irradiated under UV light (365 nm, 180 mW.cm⁻²) for 15 minutes for curing. Subsequent steps were 671

as described above for PAA gels.

Acrylamide (wt%)	Bis-acryl. (wt%)	PEGDA (wt%)	Modulus (kPa)
4	0.225	0.0	2.7±0.3
5	0.225	0.0	6.1 <u>+</u> 0.2
8	0.264	0.0	18.5 <u>+</u> 0.7
20	0.47	0.0	65 <u>+</u> 5.6
15	0.65	0.0	84 <u>+</u> 1.1
20	0.7	0.0	103 <u>+</u> 3.8
0	0	5	5.7 <u>+</u> 0.3
0	0	20	102 <u>+</u> 8.4

 Table 1. Hydrogel compositions and associated Young's moduli

573 Mechanical characterization

- ⁵⁷⁴ The viscoelastic properties of the various gels were characterized by AFM microrheology, using the
- ⁵⁷⁵ "contact force modulation" technique described recently and validated on hydrogels (*Abidine et al.*,
- **2015**). It allows determining elastic and loss shear moduli, G' and G'', as a function of frequency
- over the range 1 300 Hz. The Young moduli reported in table 1 have been computed as $E = 3G'_{0'}$
- with G'_0 the low frequency plateau modulus obtained by microrheology, assuming a Poisson ratio
- $\nu = 0.5$ for all gels. All gel samples displayed elastic behavior with $G' \gg G''$.
- 500 Measurements were performed on a JPK Nanowizard II AFM, with pyramidal-tipped MLCT probes
- [581 (Bruker) of spring constant 15 mN/m. Data were analyzed using a home-written software for mi-
- $_{\tt 582}$ $\,$ crorheology. 30 μm -thick gels were prepared, as described above, on round coverslips mounted at
- the bottom of 35 mm petri dishes. They were then either characterized immediately or left to dry
- to mimick the protocol used for inclusion in the flow chamber. Experiments were performed in PBS + 1 % yol, tween 20 (Sigma) with Tween used to prevent adhesion of the AEM tip to the get All
- measurements were carried out at 37°C to mimic experimental conditions with bacteria Results
- were compared with force-distance indentation curves that gave consistent results at low rigidities
- (< 20 kPa) but overestimated the rigidity for higher values (Fig.1-fig. suppl. 1a).
- $_{sso}$ Homogeneity of the gels was assessed at the μm and mm scales by multiposition measurements.
- ⁵⁹⁰ We found very good repeatability of the measurements and homogeneity of the gels at all scales
- (Fig.1-fig. suppl. 1b). Subsequent measurements were hence acquired at 3-6 different points in
- the gels and the average and standard error of the mean are provided (Table 1). Rigidity was also measured before and after drving and rehydration of the gel to check for possible damage to the
- ⁵⁹³ measured before and after drying and rehydration of the gel to check for possible damage to the ⁵⁹⁴ structure. In addition, confocal images of the surface of fluorescently labelled gels were used to
- track default on the gel surface before and after drving. We found no evidence of damage to the
- ⁵⁹⁶ hydrogel upon drying, except for very soft gels of rigidity below 1 kPa that were not used in this ⁵⁹⁷ study (Fig.1-fig. suppl. 1c).

Microscopy experiments

- ⁵⁰⁹ Diluted bacterial solution was injected into the channel, and kept without flow for 30 min to allow ⁶⁰⁰ bacteria to attach. During that time, clean tubing was connected to a syringe and filled with TB-PBS ⁶⁰¹ medium supplemented with 3 mM glucose and connected to the inlet of the device. 30 min after ⁶⁰² injecting bacteria into the device, the flow of clean medium was initiated. The flow rate was first ⁶⁰³ set at 25 μ l/min for 3 min in order to flush out unattached bacteria, and then lowered to 1 μ L/min ⁶⁰⁴ and maintained constant with a syringe pump (Pico Plus, Harvard Apparatus) throughout the ac-⁶⁰⁵ quisition (yielding a wall shear stress of 2.5 mPa). The set up was immediately placed into the
- incubation chamber (37 °C) of a Leica SP8 confocal microscope, and acquisition was started at 1 frame/minute.
- ⁶⁰⁸ For matrix staining experiments, concanavalin A (Alexa Fluor 647 conjugate, ThermoFisher Scien-⁶⁰⁹ tific) was added to the medium (3 ul/ml of a 1mg/ml stock solution) and infused in the flow cell
- for at least 30 min prior to imaging. Since the tetravalent conA interferes with the structure of the matrix, it was used either for short-term imaging of bacterial twitching at early stages (t < 1h, fig.
- matrix, it was used either for short-term imaging of bacterial twitching at early stages (t<1h, fig.
 4D), or added at the end of an acquisition to assess matrix distribution on and around colonies
- ۲۵۱۵ (t∼ 8h, Fig.4-fig. suppl. 1).
- For control experiments in wells (Fig. 1-fig. suppl. 3), the protocol was modified as follows: PAA gels were prepared as described above, but at the center of a 35 mm round glass coverslip. The coverslip was then glued (5-min epoxy, Araldite) to the bottom of the well of a 6-well plate (Costar, Corning), previously cut-out to open a circular 32-mm hole. 1 mL of diluted bacteria suspension (OD₆₀₀=0.005) were deposited on the gel, incubated for 30 min, and then carefully pipetted out, and the well was filled with 3 mL fresh medium (TB-PBS + 3 mM glucose) and kept at 37 °C. This setup
- allowed continuous imaging of bacteria on a Zeiss Axio-observer 7 inverted microscope in phase
- contrast mode (63x objective) equipped with an Orca-Flash 4.0 LT camera (Hamamatsu).
- 622

⁶²³ Surface coverage analysis and tracking of individual bacteria

624 All image processing and analysis, unless otherwise noted, was performed with Fiji using avail-

able plugins and home-written macros. In order to quantify the movements of individual bacteria,

time series of phase-contrast images were registered using the Fiji plugin "MultiStackReg" Theve-

naz et al. (1998) and segmented with the plugin "weka trainable segmentation" Arganda-Carreras

et al. (2017). The resulting segmentation was checked and corrected manually.

For the analysis of the global velocity $V_{g'}$ segmented binary images were used to estimate the sur-

face coverage A(t) and the explored surface area S(t), using a home-written MATLAB script. Briefly, at each timepoint 2 binary images were generated: one where pixels occupied by bacteria were

at each timepoint 2 binary images were generated: one where pixels occupied by bacteria were assigned the value 1, and all others zero (providing A(t)), and another image obtained by adding

all binary images up to this timepoint, so that all visited pixels were assigned the value 1 (providing and a state of the value 1) (pr

634 S(t)).

For the analysis of individual displacement steps, segmented bacteria were fitted with an ellipse, and the "analyze particle" imagel function was subsequently used to locate the center of mass of

and the "analyze particle" imageJ function was subsequently used to locate the center of mass of each bacterium. The Fiji plugin "TrackMate" *Tinevez et al.* (2017) was used to track all individual bac-

each bacterium. The Fiji plugin "TrackMate" *Thevez et al. (2017)* was used to track all individual bacteria, again followed by manual validation and correction (see Video 2). The function importTrack-

MateTracks (https://github.com/fiji/TrackMate/blob/master/scripts/importTrackMateTracks.m) was

used to import tracking data into MATLAB, and homemade scripts were used to sort data, plot

- tracks and obtain velocity distributions.
- 642 643

644

The analysis of the histograms of displacement steps were performed as follows: we assume that the measured steps are the incoherent sum of two displacement vectors, the active displace-

ment \vec{V}_a due to T4P activity, and a vector \vec{V}_p that includes passive effects due to both the noise

on measurements, and displacements resulting from bacterial growth and local crowding. First, we considered the experimental distribution obtained with the *pilA* mutant, for which $\vec{V}_a = \vec{0}$: this

allows extracting the probability distribution for $\|\vec{V}_n\|$, which can be well fitted with a decreasing

exponential with a characteristic passive velocity $V_{C,p}$: $p(\|\vec{V}_p\|) = exp(-\frac{\|\vec{V}_p\|}{V_{C,p}})$, with $V_{C,p} = 0.044 \,\mu m/min$.

 $_{650}$ We then considered the case of twitching bacteria. Here, we observed that the tail of the displace-

ment step distribution also follows a decreasing exponential trend. Based on the reasoning that

passive displacements are short-ranged and should not significantly modify the distribution for

large displacement values, we deduce that the tail of the probability distribution for $\|\vec{V}_a\|$ is a decreasing exponential, $p(\|\vec{V}_a\|) = exp(-\frac{\|\vec{V}_a\|}{V_c})$, with V_C the characteristic active twitching velocity of

₅₅₅ bacteria.

We confirmed the validity of this hypothesis by calculating the probability distribution functions.

We assumed that the distribution of measured displacement steps, $\|\vec{V}_{tot}\| = \|\vec{V}_a + \vec{V}_p\|$ is the sum of two uncorrelated exponential distributions with different scales and a random angle between

the two vectors. There is no analytical expression for this sum, hence we performed numerical

calculations of the distributions obtained in the general case. In the limit $\|\vec{V}_{tot}\| >> V_C > V_{C,p}$ an

exponential distribution is retrieved with a characteristic velocity V_C , unaffected by $V_{C,p}$ (Fig. 3-fig.

suppl. 3, left). A fitting of the range $p(\|\vec{V}_{tar}\|) < 0.3$ (which excludes the first few points that do not

⁶⁶³ follow an exponential trend) confirms that V_c is obtained accurately provided that $V_c > V_{c,p}$ (Fig.

⁶⁶⁴ 3-fig. suppl. 3, right). Below this limit, only $V_{C,p}$ is detected since active displacements are in the

⁶⁶⁵ range of passive "noisy" ones.

Experimentally, we have used a lower cutoff of $\|\vec{V}_{tot}\| > 0.08 \ \mu m/min \simeq 2V_{C,p}$ for the fitting range, to

restrict it to the exponential part of the distribution. To account for the noise in the measurement,

we have also considered that fitted values below $V_{C,p} = 0.044 \mu m/min$ were in the range [0;0.044] $\mu m/min$.

⁶⁷⁰ Quantification of the morphology of colonies

Quantification was performed on confocal fluorescence 3D resolved images. First, signal attenua-

tion with depth was compensated by decreasing exponential fitting of the mean pixel values inside

the colony with depth, and normalization by the corresponding function. A 2D 3x3 smoothing op-

eration was then performed on each image of the z-stack, and the colonies were subsequently segmented using a simple thresholding operation: while this procedure does not permit segmen-

segmented using a simple thresholding operation: while this procedure does not permit segmentation of individual bacteria, it provides a good estimate of the 3D envelope of the colonies. The

tation of individual bacteria, it provides a good estimate of the 3D envelope of the colonies. The topology of the colonies was then quantified by calculating the roughness of this envelope using

topology of the colonies was then quantified by calculating the roughness of this envelope using

the widely-used arithmetic average roughness *Ra*

$$Ra = \frac{1}{N} \sum_{i=1}^{N} |z_i - \langle z \rangle|, \qquad (12)$$

where summation is over all 2D positions *i* in the 3D image, z_i is the height of the highest segmented

pixel at position *i* and $\langle \rangle$ is the averaging operator over all positions. The occupied volume V is

681 calculated as

$$V = px^2 \sum_{i=1}^{N} z_i,$$
 (13)

with px the pixel size. The occupied area as a function from the distance to the coverslip is the histogram of z_i values with bin size 0.5 μ m (corresponding to the vertical sampling of the 3D images).

Quantification of the mixing of two strains co-colonizing the same soft substrate, as a function of the softness

This quantification is performed both at the low density stage with isolated bacteria, and at a later

stage on maturing colonies. To this aim, we used a statistical tool, Moran's I index, designed to

quantify the spatial clustering of species and widely used in the field of ecology and geography

Moran (1950). Moran's I is a measure of the local spatial correlations that includes a notion of

spatial proximity, either in the form of a spatial cut-off for the calculation of the heterogeneity (in

other words, a characteristic distance), or a number of neighbors. It takes values ranging from

- ⁶⁹² 1 (perfectly correlated values) to -1 (perfectly anti-correlated values), with 0 corresponding to a
- spatially random distribution of the variable.

⁶⁹⁴ Considering a variable y that can take two different values (in our case, green (1) or yellow (-1)) ⁶⁹⁵ with n realisations, Moran'l is expressed as:

$$I = \frac{n}{\sum_{i=1}^{n} \sum_{j=1}^{n} w_{ij}} \frac{\sum_{i=1}^{n} \sum_{j=1}^{n} w_{ij} \left(y_i - \langle y \rangle \right) \left(y_j - \langle y \rangle \right)}{\sum_{i=1}^{n} \left(y_i - \langle y \rangle \right)^2},$$
(14)

where w_{ij} is the matrix of weights that contains the spatial information (with $w_{ii} = 0$). In our experiment, the relevant spatial scale (and hence the matrix of weights) varies greatly over time because of the change in the density of the bacteria on the surface. While at high density (maturing colonies) defining a length scale is a suitable way of testing the presence of local correlations, this is more challenging at earlier times when the distance between neighbours exhibits large stochastic variations, in particular for stiff substrates. Hence, different matrices of weights were chosen for early-stage and later-stage colonisation of the surface:

at early stages of colonisation, when the bacteria are sparse on the surface, we chose to focus
 on the nearest neighbours of each bacteria. To this aim, individual bacteria are segmented
 in the green and yellow images, and their center of mass location is collated into a list of 2D
 coordinates and colour for all bacteria in the field of view. Moran's I is then calculated based

on this list using the following weight matrix:

$$w_{ij} = \begin{cases} 1 & \text{if } j \text{ is one of the } p \text{ nearest neighbours of bacteria } i \\ 0 & \text{otherwise} \end{cases}$$

We arbitrarily chose p = 5 as a significant number of neighbours, although similar results are 708 found for values of p ranging from 4 to 10. Lower numbers are biased by cell division: at 709 the time of division, the closest neighbour is necessarily of the same strain as the hacteria 710 under consideration, so that there is always a positive correlation between them. As a result, 711 testing for mixing requires to mitigate this effect by choosing a large enough value for p. In 712 practice, we found that p = 5 was a good compromise to limit this bias while maintaining a 713 "local" approach, i.e. not considering the correlation between bacteria further apart than half 714 of the field of view (i.e. 160 um). 715

at later stages with dense, 3D colonies, individual segmentation of bacteria becomes challeng ing and the correlation measure is performed on individual pixels: first, a simple thresholding
 operation is performed on the green and yellow image, and each pixel is attributed a value:
 1 (green pixel), 0 (black pixel) or -1 (yellow pixel). From this new image, Moran's I is calculated
 using the following weight matrix:

$$w_{ij} = \begin{cases} 1 & \text{if the distance between } i \text{ and } j \ (i \neq j) \text{ is smaller than or equal to } d \\ 0 & \text{otherwise} \end{cases}$$

Again, the cut-off distance *d* is arbitrarily chosen as $5 \,\mu$ m although values between 3 and 10 μ m yield similar results: it permits limiting fluctuations by averaging over a significant number of bacteria, while maintaining a local measure of mixing. In addition, because individual bacteria cover more than one pixel in the acquired images, a number of pixels of the same colour as pixel *i* are removed to avoid correlating the bacteria with itself. In our data the average number of pixels covered by one bacteria is measured to be 40.

While there is some degree of freedom on the choice of the weight matrix, it is important to note that we use the same weight to compare data obtained on three different rigidities, hence minimising the impact of the exact chosen parameters on the comparison. In contrast, values obtained on one surface at the two different time points should not be directly compared as they have not been obtained with the same weight matrix.

733 Acknowledgments

We thank Claude Verdier for help with the AFM elasticity measurements, Denis Bartolo for help
with microfluidics, and Benoit Coasne and Benedikt Sabass for fruitful discussions on data modeling. We are extremely grateful to Ina Attree and Sylvie Elsen (IBS, Grenoble) for strains, help
and advice. We thank Tim Tolker Nielsen for the kind gift of pcdrA-gfp reporter plasmids. D.D.
was supported by the French National Research Agency (grant ANR-19-CE42-0010). The authors
acknowledge support from LabeX Tec 21 (ANR-11-LABX-0030).

707



Figure 1-figure supplement 1: Mechanical characterization of hydrogels by AFM (a) comparison of elastic moduli measured by indentation and by microrheology. Both techniques yield quantitatively similar results for gels with Young's moduli E < 20 kPa. (b) spatial homogeneity of the gels characterized by indentation measurements. For each position, separated by 1 mm, a 4x4 force spectroscopy map is taken, with a spacing between "pixels" of 3µm. (c) comparison of elastic moduli *E* measured before and after drying/rehydration of the gels.



Figure 1-figure supplement 2: Morphology of microcolonies is strongly impacted by surface rigidity on PEG hydrogels. Phase contrast images of WT PAO1 bacteria on PEG hydrogels 5h after the onset of surface colonization. Scale bars, 20 μm.



Figure 1-figure supplement 3: Substrate rigidity impacts early microcolony morphology in the absence of shear flow. Phase contrast images of WT PAO1 bacteria on PAA hydrogels at the bottom of a 6-well plate, after 150 min of incubation at 37{oC without agitation. In red: initial adhering bacteria at t0. (Scale bars, 20 μ m.)



Figure 1-figure supplement 4: In the T4P-deficient mutant PAO1 *pilA*::*Tn*5, substrate rigidity does not significantly impact colony morphology. Colonies imaged after 5 h. Scale bar 20 μm.



Figure 2-figure supplement 1: Behavior of mutants *PAO1 sadC* :: *Tn*5 and *PAO1 wspR* :: *Tn*5 on soft (2.7 kPa) and stiff (84 kPa) PAA hydrogels. Left: typical images showing initially attached bacteria (red) and total explored area after 100 minutes (white). Right: Average bacterial velocity Vg calculated using equation 2 of the main text, over the first 100 minutes of acquisition (1 experiment for each strain, 3 different positions for each gel. Error bars are SEM).



Figure 3-figure supplement 1: (A) Distribution of the path duration of bacterial tracks measured on different PAA substrates (WT PAO1). One track starts after a division event, and finishes either at the next division, or when the bacterium leaves the surface. Full lines are bimodal gaussian fits, evidencing a sub-population of bacteria that detach from the surface before dividing. (B) Mean track velocity is similar for short tracks (below the cutoff value t_c =16 min) and long tracks, for all tested substrate rigidities. (total number of analyzed tracks: 60 (2.7 kPa), 124 (18.5 kPa), 175 (84 kPa).



Figure 3-figure supplement 2: Anaysis of the mean track velocity. (A) Mean track velocity distribution for different values of the substrate rigidity. Only full tracks were considered (corresponding to the right peak in Fig.7). Considering all tracks does not significantly modify the distributions (data not shown). 84 kPa: 330 tracks from 2 independent experiments, 18.5 kPa: 394 tracks from 3 independent experiments, 2.7 kPa: 83 tracks from 2 independent experiments. (B) Standard deviation of the mean velocity for individual tracks, shown for the 2.7 kPa and 84 kPa data sets. Dotted line is a linear fit for the 84 kPa data (y=0.65x).



Figure 3-figure supplement 3: Validation of the fitting of displacement steps distributions with single exponentials. Left, calculated distributions (dots) of the sum of two uncorrelated random vectors each following a decreasing exponential distribution with characteristic velocities V_C and $V_{C,p}$, at an angle following a random distribution. $V_{C,p} = 0.044 \mu m/min$ is equal to the experimentally measured value on the *pilA* mutant. V_C spans a range similar to experimental measurements, (0.05, 0.1, 0.2, 0.3 and 0.4 μ m/min, from dark blue to cyan). Orange lines are exponential functions with the corresponding V_C . Right, characteristic velocities extracted by fitting the exponential part of the simulated distributions, as a function V_C , *with* $V_{C,p} = 0.044 \mu m/min$. The black line corresponds to y=x.



Figure 3-figure supplement 4: Statistical analysis of V_g (A) and V_c (B) for various gel rigidities. Indicated are p-values from paired two-sided Mann-Whitney non-parametric tests. n.s. denotes datasets that are not significantly different at a threshold of 0.05.



Figure 3-figure supplement 5: Characteristic twitching velocity measured on PEG hydrogels is similar to the one measured on PAA hydrogels under identical experimental conditions. V_C obtained by fitting the distribution of displacement steps on PAA or PEG hydrogels in a similar range of substrate rigidity. Error bars are SEM. Each point is the average of 2 positions from 1 experiment. (Note that for technical reasons these experiments were carried out with a wall shear stress of 6.4 mPa, about 3x higher than all other experiments on PAA).



Figure 4-figure supplement 1: EPS staining with concanavalin A highlights matrix deposits on the surface of substrates of different stiffnesses after 8h of surface colonization. Matrix deposits are more compact on soft substrates with large areas (>60%) devoid of EPS, while stiffer substrates are almost fully decorated (>% covered area). The two colours for the softer surface corresponds to two areas taken 30 minutes apart after staining, before and after acquisition of the two other datasets on larger rigidities, to rule out any significant time evolution of the staining.



Figure 5-figure supplement 1: Bacterial spatial distribution as a function of substrate rigidity. Top, images of surfaces seeded with a 1:1 mixture of constitutively fluorescent bacteria expressing GFP or YFP at two different times after the start of surface colonization (bottom : 3D rendering of a volumetric image). Bottom, Spatial autocorrelation quantified via Moran's I index at the two time points il lustrated above. The values should not be directly compared between the two time points (see SI I-C) but illustrate that the rigidity modulation of mixing is maintained over time from very sparse to large coverage of the surface.



Figure 6-figure supplement 1: Fluorescence intensity from a PAO1 strain expressing a $P_{cdrA} - gfp$ plasmid over time, showing a plateau of fluorescence expression after \approx 500 min. In our experiments, this change in behaviour might be due to a lack of oxygen in the flow upon growth of the biofilm in the microfluidic channel and occurs. Two rigidities are shown and scaled to the same final fluorescence value.

Video 1: surface colonization on 2.7 kPa (left) and 65 kPa (right) PAA hydrogels, imaged with
 phase contrast microscopy with one image/min over 6 hours. The two gels were included in the
 same microfluidic channel and imaged quasi-simultaneously. Scale bar, 20µm.

744

Video 2: Principle of the image processing for the tracking of individual bacteria. Glass sur face colonization under controlled shear flow followed over 90 minutes (1 frame/minute) by phase
 contrast microscopy. Left, the registered phase contrast image is superimposed with the center
 of mass of the detected cells after segmentation (green dots). Right, tracks of the detected cells
 (obtained with the simple LAP tracker of the Trackmate Image] plugin), color-coded as a function
 of the track final length. Scale bar, 20µm.

Appendix 1: Modeling twitching velocity on soft substrates

The principle of our modeling of rigidity-modulated twitching in 1D is shown in figure 2D (main
text), and incorporates three main ingredients: modeling of the substrate deformation (subsection
A), of cell body friction on the surface (subsection B) and of the pilus retraction dynamics (subsections C and D).

756 A. Modeling substrate deformation

751

We have based our approach on the theory of linear elasticity for the description of the substrate: 757 in this framework, the deformation of the substrate occurs over a typical length scale given by 758 the size of the adhesion, λ_i and it is proportional to the applied force. Finally, the proportionality 759 coefficient Y scales as the product of the substrate elastic modulus, E, and the adhesion size λ i.e. 760 $Y = E\lambda$. This simple relation is valid only for small displacements on rigid substrates. It is likely to 761 fail quantitatively on very soft substrates with large displacements, low cross-linker densities and 762 non-affine deformations, but is a reasonable first approximation for the simple model we propose 763 here 764

This modeling introduces characteristic length scales that depend on the part of the bacteria 765 under consideration: both the pilus and the cell body form contact with the substrate. The pilus 766 attaches at its tip over size $\lambda \approx 1$ nm, while the cell body has a typical size of $l_{\lambda} \approx 1$ µm. In addition, 767 a third length scale is the typical length of the pilus, L, which varies during retraction but is most 768 of the time > 1 μ m. Introducing these three quantities permits to simplify the description of the 769 deformation of the substrate: the pilus tension F and the displacement at the adhesion site in the 770 substrate u are linearly related by F = Yu, with Y being an effective spring constant. We model 771 the substrate as an infinite (thickness $25 \,\mu\text{m} \gg \lambda$, lateral extension $\approx 1 - 10 \,\text{mm} \gg \lambda$), isotropic, 772 elastic and incompressible half space. Furthermore, we neglect the influence of the cell body on 773 the deformation around the pilus tip since $L >> \lambda$ so that the deformation of the substrate has 774 decaved to zero at the cell body. 775

The 2D Boussinesq Green's tensor at the surface z = 0 for a point-like shear force **f** at the origin is given by *Landau and Lifschitz* (2004)

$$\mathbf{G}(\mathbf{r}) = \frac{3}{4\pi E} \left[\frac{\mathbf{I}}{r} + \frac{\mathbf{r} \otimes \mathbf{r}}{r^3} \right] \mathbf{f} \,. \tag{A1}$$

⁷⁷⁸ Considering an adhesive T4P tip of length λ and half-width *d* and using slender body approxima-⁷⁷⁹ tions, the total force *F* on the pilus for a "lengthside" displacement *u* is given by

$$F = \frac{E\lambda\pi}{3\ln\frac{\lambda}{d}}u \quad \text{with} \quad Y = \frac{E\lambda\pi}{3\ln\frac{\lambda}{d}} \approx E\lambda.$$
 (A2)

Here we have implicitly introduced a 1D setting, i.e. we will neglect the vectorial nature of forces and displacements and restrict ourselves to a 1D setting. We find, as expected, that *Y* scales linearly with λ . This holds equivalently for the cell body by replacing λ with $l_b >> \lambda$: as a result, the substrate deformation at the cell body caused by the same pilus tension *F* is of amplitude smaller by a factor $\lambda/l_b \ll 1$ and will be neglected for the sake of simplicity.

In contrast, we consider the pilus tip to be firmly attached to the substrate until detachment while the cell body can slide on the surface. Note that this asymmetry between bacterial body (macroscopic sliding over the substrate) and the supposedly small pilus/substrate contact (pointlike force deforming the substrate) is the essential difference to the pulling process described in Ref. *Simsek et al.* (2019), where the contact of the bacterial body and the pilus extremity are mechanically treated as equivalent.

791 B. Modeling cell body friction

As stated above, the model requires a description of the sliding motion of the cell body on the sub-

strate as a function of the force F applied by the pilus. We base our modeling on the theory from

- 794 Sens (2013) that considers stochastic friction by an ensemble of N elastic linkers (not necessarily
- ⁷⁹⁵ all bound at all times) between an elastic substrate and a cell, submitted to a sliding velocity v. The
- bonds are modeled as slip bonds with a critical force f^* , an off-rate constant at zero force k_{off}^0 and
- an on-rate constant k_{on} . The linkers' stiffness is k_{b} .

In the case of an infinitely rigid substrate, the mean total force on the cell body $\langle F \rangle$ as a function of its velocity v is non-monotonous and is given by

$$\langle F \rangle = N f^* \frac{r_{\rm on} e^{1/\bar{\nu}} \int_0^\infty f e^{-\left(\frac{e^f}{\bar{\nu}}\right)} df}{\tilde{\nu} + r_{\rm on} e^{\left(\frac{1}{\bar{\nu}}\right)} \Gamma\left[0, \frac{1}{\bar{\nu}}\right]}$$
(A3)

with $\tilde{v} = v/v_{\beta}$, $v_{\beta} = k_{\text{off}}^0 f^*/k_{\text{b}}$, $r_{\text{on}} = k_{\text{on}}/k_{\text{off}}^0$. $\Gamma[0, x]$ is the Euler gamma function. Eq. (A3) exhibits a complex dependence of $\langle F \rangle$ on \tilde{v} that requires estimating typical values of the parameters in our experiments. Putting in numbers to obtain the typical speed v_{β} , we can estimate that

• $k_{off}^0 \approx 1 - 10 \ s^{-1}$ (slightly higher than for specific ligand/receptor bonds **Robert et al. (2007)**) • $f^* = k_{\rm B}T/x_{\beta}$ with $x_{\beta} \approx 0.1 - 10$ nm being the transition state distance between bound and unbound state as proposed by Evans **Evans (2001)** and others **Pereverzev et al. (2005)**.

• k_b is more difficult to estimate. Here we assume that bacterial adhesion is mediated by the bacteria produced extracellular matrix, of which a major constituent are exopolysaccharides. Using a worm-like chain (WLC) model for a polymer of persistence length $L_p \approx 10$ nm (as calculated for bacteria produced exopolysaccharides in *Kuik et al.* (2011)) and contour length $L_0 \approx 100$ nm (assuming a chain length of about 100 monomers with size 1 nm), the linear force-elongation relationship in the regime of weak forces *Marko and Siggia* (1995) yields a force constant $k_b \approx \frac{3kT}{2L_0L_0} \approx 6 \times 10^{-3}$ pN.nm⁻¹.

Taking extreme values this leads to typical velocities in the range $v_{\beta} = 1 - 100 \,\mu\text{m.s}^{-1}$. In our experiments the bacteria are not expected to move faster than the pilus retraction velocity (i.e. $1 \,\mu\text{m.s}^{-1}$ **Skerker and Berg (2001)**, if one excepts the case of slingshots that were not frequently observed in our experiments). Taking into account that the pilus retraction speed slows down considerably as the tension in the pilus increases, the bacterial speed during one pilus retraction is rather smaller than this maximum value. Hence, we always have $\tilde{v} = v/v_{\beta} < 1$, and Eq. (A3) can be linearized to

$$\langle F \rangle = N f^* \frac{k_{\rm on}}{k_{\rm off}^0 + k_{\rm on}} \tilde{\nu},\tag{A4}$$

In addition, the elasticity of the substrate should be considered. Ref. Sens (2013) proposes that 819 this situation is equivalent to having a system of springs in series, one stemming from the substrate 820 elasticity and the second being the collection of individual bond springs (in parallel). In this case 821 and using once again the theory of linear elasticity, the previous analysis holds if \tilde{v} is rescaled by a 822 factor $\frac{E l_b}{k_1 + E l_c}$, with E > 3 kPa the substrate Young's modulus and $l_b \approx 1 \,\mu$ m the characteristic size of 823 the bacterial cell body, $a = l_b \approx 1 \,\mu\text{m}$. Hence $El_b \geq 3 \,\text{pN.nm}^{-1} \gg k_b$ and the scaling factor $\frac{El_b}{k_b + El_b} \approx 1$, 824 so that the elasticity of the substrate does not influence the friction of the cell body. 825 In summary, we find that we can reasonably use a linear approximation for the bacterial sliding 826

⁸²⁶ In summary, we find that we can reasonably use a linear approximation for the bacterial sliding ⁸²⁷ speed in response to the pulling force due to the pilus retraction, $F = \eta v$ with η a friction coefficient. ⁸²⁸ Finally, we consider η as independent from the substrate rigidity, which is reasonable if we assume ⁸²⁹ that the number of bonds is limited by the number of molecules/appendages of the cell body that ⁸³⁰ can interact with the substrate, rather than the number of binding sites on the substrate itself ⁸³¹ (PAA mesh size $\approx 3 - 10$ nm), and that the interaction may in addition be mediated by adsorbed ⁸³² exopolysaccharides deposited by the bacteria. However, other non-linear dependencies can be ⁸³³ easily included into the modeling.

834 C. Basic modeling of pilus retraction

The relevant step during twitching which induces bacterial motion is the active pilus retraction when attached to the substrate. Here we assume, that the limiting effect for bacterial motion is the detachment of the pilus from the substrate, and not the complete retraction of the pilus by the bacterium. To understand the role in substrate rigidity on the bacterial twitching speed we will

therefore concentrate on this crucial step without describing the whole cycle of pilus dynamics, for
which the kinetics is not completely understood *Koch et al.* (2021); *Talà et al.* (2019).

⁸⁴¹ We consider the retraction of a single effective pilus pulling on the bacterial body until it detaches from the substrate. We treat the pilus as rigid and inextensible filament: assuming a force constant of 2 pN. μ m⁻¹ for the pilus elasticity **Beaussart et al.** (**2014**), a substrate rigidity of *E* = 100 kPa, an adhesion size of $\lambda = 1$ nm and a maximum force exerted by the pilus of *F*_R = 100 pN, the substrate displacement is $u \sim F_R/E\lambda = 1 \mu m$. In contrast, the pilus elongation is $\Delta L = 50$ nm and can therefore be neglected for our conditions. However it would not pose any difficulty to include the pilus elasticity into the calculations.

Let v_R be the retraction speed of the attached pilus inducing a displacement u in the substrate. At the same time the bacterium will slide forward with speed v_B , reducing the tension in the pilus and the displacement in the substrate:

$$\frac{\mathrm{d}u}{\mathrm{d}t} = v_{\mathrm{R}}(F) - v_{\mathrm{B}}(F) \quad \text{with} \quad F = Yu \,. \tag{A5}$$

Both motions (substrate displacement and bacterial sliding) are coupled via the tension in the pilus

F. Its retraction speed is described by a simple linear dependence that has been well documented

853 Marathe et al. (2014); Koch et al. (2022)

$$v_{\rm R} = v^0 \left(1 - \frac{F}{F_{\rm R}} \right) \,, \tag{A6}$$

with F_R a stall force. As established in the previous subsection, the bacterial sliding speed depends

linearly on the pilus tension with friction constant $\eta = F_B/v^0$:

$$v_{\rm B} = \frac{1}{\eta} F = v^0 \frac{F}{F_{\rm B}} \,.$$
 (A7)

 $F_{\rm B}$ denotes the force necessary to pull the bacterium at maximum retraction speed v^0 over the substrate. From Eq. A5 we recover the increase in the pilus tension over time during the retraction

$$F(t) = F_0 \left(1 - e^{-\frac{Y_0^0}{F_0}t} \right)$$
(A8)

⁸⁵⁸ with the force scale

$$F_{0} = \frac{F_{\rm B}F_{\rm R}}{F_{\rm R} + F_{\rm B}} \,. \tag{A9}$$

Incorporating solution (A8) into Eq. (A7) with $v_{\rm B} = \frac{dx_{\rm B}}{dt}$ we recover for the bacterial sliding distance during pilus retraction

$$x_{\rm B}(t) = \frac{F_0}{F_{\rm B}} \left[v^0 t + \frac{F_0}{Y} \left(e^{-\frac{Y t^0}{F_0} t} - 1 \right) \right] \,. \tag{A10}$$

While retracting the pilus will detach with a rate constant $k_{off}(F)$ from the substrate. Assuming a force independent off-rate constant $k_{off} = k_{off}^0$ the detachment times are distributed exponentially with mean $1/k_{off}^0$. Furthermore, we assume that the single effective pilus considered in our model retracts with frequency k_p and thus gives rise to an effective velocity

$$v_{\rm eff} = k_{\rm p} \langle x_{\rm B} \rangle = k_{\rm p} k_{\rm off}^0 \int_0^\infty x_{\rm B}(t) \, e^{-k_{\rm off}^0 t} \, \mathrm{d}t = V_{\rm max} \frac{E}{E + E_0} \,.$$
 (A11)

Here, $\langle x_{\rm B} \rangle$ denotes the mean bacterial sliding distance per pilus retraction event. $V_{\rm max}$ denotes the

maximum effective speed that a cell can reach on a given substrate at infinite rigidity, given by

$$V_{\rm max} = v^0 \frac{k_{\rm p}}{k_{\rm off}^0} \frac{F_{\rm R}}{F_{\rm B} + F_{\rm R}} \,. \tag{A12}$$

 E_0 denotes the rigidity at half-maximal speed and is given by



Appendix 1-figure 1: Experimentally measured velocity vs. rigidity data and least squared fits of Eq. (A11) w.r.t. to the experimental values as indicated in the legends. a: Local velocity measures. Parameters obtained by a least-square fit: $E_0 = 32 \pm 20$ kPa, $V_{\text{max}} = 0.48 \pm 0.12 \,\mu\text{m.min}^{-1}$. b: Global velocity measures. Parameters obtained by a least square fit: $E_0 = 84 \pm 68$ kPa, $V_{\text{max}} = 0.77 \pm 0.35 \,\mu\text{m.min}^{-1}$. Errorbars indicate SEM.

Appendix 1-figure 1 shows the experimental data and fitted curves, which capture well the 868 data for medium and high rigidities. The theoretical curves were fitted to all experimental values 869 (applying the statistical weight in the measured rigidities and equal weight in the velocities) using a 870 least square fit (software gnuplot Williams et al. (2019)). Assuming a typical pilus retraction speed 871 $v^0 = 0.5 - 1 \,\mu\text{m.s}^{-1}$ Marathe et al. (2014); Koch et al. (2022), a stall force of the order $F_{\text{R}} = 50 - 100 \,\text{pN}$ 872 *Marathe et al. (2014); Koch et al. (2022), a pilus off-rate constant* $k_{off}^0 = 1 \text{ s}^{-1}$ *Talà et al. (2019)* and 873 a contact size of $\lambda = 1$ nm Koch et al. (2022), a high friction surface with $F_{\rm B} = 1$ nN and a typical 87/ pilus retraction frequency ² of $k_p = 0.1 - 0.2 \text{ s}^{-1}$ we recover a $V_{\text{max}} \sim 0.1 - 1 \,\mu\text{m.min}^{-1}$ and a substrate 875 rigidity at half maximum speed of $E_0 = 10 - 100$ kPa, a range which encloses the fitted values (see 876 Appendix 1-figure 1). 877 Here we have assumed a force-independent off-rate constant for the pilus. In a more complex

Here we have assumed a force-independent off-rate constant for the pilus. In a more complex
 scenario, the contact between the pilus and the substrate may act as a slip bond or catch bond. For
 completeness we will show some numerical results for slip and catch bond behavior below, which

do not increase however the quality of fit between experimental and theoretical velocity data.

D. Force dependent detachment rate constants

Increasing the complexity of the model, we assume that the pilus detachment rate is force depen-

dent Kramers (1940); Björnham and Axner (2010); Pereverzev et al. (2005); Talà et al. (2019) and

⁸⁸⁵ takes the form

$$k_{\rm off} = k_{\rm off}^0 \left(\varepsilon e^{-\frac{F}{F_{\rm C}}} + e^{\frac{F}{F_{\rm S}}} \right) \,. \tag{A14}$$

⁸⁸⁶ ε = 0 denotes a slip bond and ε > 0 denotes a catch bond behavior. $F_{\rm C}$ and $F_{\rm S}$ denote positive ⁸⁸⁷ force constants **Pereverzev et al. (2005)**. Eq. (A14) implies that the pilus detachment times are not ⁸⁸⁸ distributed exponentially.

We now consider the evolution equation for the probability density p(u) that a pilus attached to the substrate is retracting and is thereby inducing a displacement u

$$\partial_t p = -k_{\rm off}(F)p - \partial_u j_{\rm u} \tag{A15}$$

The first term denotes (tension dependent) pilus detachment from the substrate and the second term captures the advection of the displacement due to pilus retraction and bacterial sliding. It is

²Here we assume that one single effective pilus is active during a retraction event. Using a typical pilus length of 5 μ m with retraction speed of $v_0 = 0.5 - 1 \ \mu$ m.s⁻¹ gives a duration of 5-10 s per retraction and a retraction frequency of 0.1-0.2 s⁻¹

formulated as a divergence of a flux j_{u} with

$$j_{\rm u} = \left[v_{\rm R}(F) - v_{\rm B}(F) \right] p$$
. (A16)

The pilus retraction $v_R(F)$ and bacterial sliding speed $v_B(F)$ is given by Eqs. (A6) and (A7). To facilitate

the analysis of the equations we use the transformation p(u) = p(u[F]) = P(F) and $\partial_u = Y \partial_F$ which

⁸⁹⁶ gives rise to the evolution equation

$$\partial_t P(F) = k_{\text{off}}(F)P - v^0 Y \partial_F \left[\left(1 - \frac{F}{F_R} - \frac{F}{F_B} \right) P \right]$$
(A17)

To reduce the number of parameters we introduce the timescale $t_0 = 1/k_{off}^0$, the length scale $l_0 = v^0 t_0$ and the force scale $F_0 = (F_R F_B)/(F_R + F_B)$. The adimensional quantities are then denoted $\tilde{F} = F/F_0$, $\tilde{t} = t/t_0$, and $\tilde{u} = u/l_0$. The adimensional evolution equation of $\tilde{P}(\tilde{F})$ takes the form

$$\partial_t \tilde{P} = -\kappa(\tilde{F})\tilde{P} - \mu \partial_{\tilde{F}} \left[\tilde{P}(1 - \tilde{F}) \right], \quad \text{with} \quad \tilde{F} \in [0, 1]$$
(A18)

where $\mu = Y v^0 / (F_0 k_{off}^0)$ denotes the adimensional substrate rigidity and κ denotes an adimensional force dependent off-rate, i.e. $\kappa = k_{off} / k_{off}^0$. Solving Eq. (A18) in the steady state we find

Î

$$\tilde{P} = \frac{\tilde{P}_0}{1 - \tilde{F}} e^{\frac{I(\tilde{F})}{\mu}}$$
(A19)

902 with

$$\mathcal{I}(\tilde{F}) = \varepsilon e^{-\frac{1}{\tilde{F}_{\rm C}}} \mathsf{Ei}\left(\frac{1-\tilde{F}}{\tilde{F}_{\rm C}}\right) + e^{\frac{1}{\tilde{F}_{\rm S}}} \mathsf{Ei}\left(-\frac{1-\tilde{F}}{\tilde{F}_{\rm S}}\right) \,. \tag{A20}$$

In Eq. (A20) the force constants \tilde{F}_{C} and \tilde{F}_{S} have been rescaled by F_{0} . The normalization factor P_{0} is defined by the integral condition $\int_{0}^{1} \tilde{P}(\tilde{F}) d\tilde{F} = 1$. Ei(x) denotes the exponential integral. At detachment the distribution of forces \tilde{F}_{d} is given by

$$\tilde{P}_{\rm d}(\tilde{F}_{\rm d}) = \tilde{P}_{\rm d0} \frac{\kappa(F_{\rm d})}{1 - \tilde{F}_{\rm d}} e^{\frac{I(\tilde{F}_{\rm d})}{\mu}} \tag{A21}$$

with the normalization factor \tilde{P}_{d0} determined by the integral condition $\int_0^1 \tilde{P}_d(\tilde{F}_d) d\tilde{F}_d = 1$.

⁹⁰⁷ Using the forces and bacterial sliding distance at detachment from the substrate

$$\tilde{F}_{\rm d} = 1 - e^{-\mu \tilde{t}_{\rm d}} \tag{A22}$$

$$\tilde{x}_{\rm B} = \frac{1}{\tilde{F}_{\rm B}} \left[\tilde{t}_{\rm d} + \frac{1}{\mu} \left(e^{-\mu \tilde{t}_{\rm d}} - 1 \right) \right]$$
 (A23)

we can perform the transformation $\tilde{P}_{d}(\tilde{F}_{d}) d\tilde{F}_{d} = \tilde{P}_{d}[\tilde{F}_{d}(\tilde{t}_{d})] \mu e^{-\mu \tilde{t}_{d}} d\tilde{t}_{d} = \tilde{P}_{\tilde{t}_{d}}(\tilde{t}_{d}) d\tilde{t}_{d}$ and recover the mean bacterial displacement per pilus retraction in adimensional form as

$$\langle \tilde{x}_{\rm B} \rangle = \int_0^\infty \tilde{x}_{\rm B}(\tilde{t}_{\rm d}) \tilde{P}_{\tilde{t}_{\rm d}} \,\mathrm{d}\tilde{t}_{\rm d} \,. \tag{A24}$$

Following the same argument as for Eq. (A11), the effective bacterial speed (dimensional) is then given by

$$v_{\rm eff} = k_{\rm p} l_0 \langle \tilde{x}_{\rm B} \rangle = \frac{k_{\rm p}}{k_{\rm off}^0} v^0 \langle \tilde{x}_{\rm B} \rangle \tag{A25}$$

$$= V_{\max} \int_0^\infty \left[\tilde{t}_d + \frac{1}{\mu} \left(e^{-\mu \tilde{t}_d} - 1 \right) \right] \tilde{P}(\tilde{t}_d) \, \mathrm{d}\tilde{t}_d \,, \tag{A26}$$

912 with $\mu = E/E_0$.

Appendix 1-figure 2 shows exemplarily the off-rate constants for force independent, slip and catch bond behavior (Appendix 1-figure 2a) and the effective velocity of a slip-bond and catchbond model along with a force independent detachment in comparison to the measured bacterial velocity using the local velocity analysis (Appendix 1-figure 2b). Thereby we chose arbitrarily a slipbond constant $F_{\rm S} = 1.1F_0$ corresponding for example to the case of a high friction substrate with



Appendix 1-figure 2: Comparison of various force dependencies of the pilus detachment rate constant k_{off} as indicated in the legend. b: Comparison of bacterial velocities obtained by models with various complexity with experimentally measured values (local velocity analysis) as indicated in the legend (parameters were fit w.r.t experimental mean values). The model parameters for the force-independent model are as in Appendix 1-figure 1a. For the slip and catch bond model the fixed parameters are $E_0 = 32$ kPa, $\tilde{F}_S = 1.1$, $\epsilon = 2$ (catch bond), $\tilde{F}_C = 0.1$ (catch bond). For the slip and catch bond model V_{max} was estimated from least square fits: $V_{max} = 0.98 \,\mu\text{m.min}^{-1}$ (slip) and $V_{max} = 1.15 \,\mu\text{m.min}^{-1}$ (catch). Error estimates are expected to be of the same order of magnitude as for the force independent model (see Appendix 1-figure 1a).

 $F_{\rm R} = F_{\rm S} = F_{\rm B}/10$, i.e. as used previously $F_{\rm R} = F_{\rm S} = 100 \, {\rm pN}$ and $F_{\rm B} = 1 \, {\rm nN}$. The catch-bond force 918 constant was chosen to be small, i.e. $F_S \ll F_0$, following the idea of Ref. Talà et al. (2019) that pilus-919 substrate attachment is stabilized for small pilus tension. Futhermore, we chose $\epsilon = 2$, i.e. pilus 920 detachment at zero loads is three times faster than for a slip-bond model. Fixing $E_0 = 32 \text{ kPa}$ [ob-921 tained from fitting the force-independent model (see Appendix 1-figure 1a)]., the theoretical curves 922 with the force-dependent off-rate constant were fitted using a least square fit in the parameter V_{max} . The catch-bond behavior captures qualitatively better the velocities at low rigidities but neither 924 slip-bond nor catch-bond seem to perform better than the simple analytical force-independent 925 detachment model for medium and high rigidities. 926

Appendix 2: Influence of bacterial motility on the onset of biofilm verticalization

A. Simple kinetic model

As described in the main text, we propose a simple kinetic model to capture the 2D to 3D transition of bacteria in growing microcolonies over time, i.e. we assume that colony verticalization results 931 from a competition between bacterial division and motility rather than from a competition be-932 tween adhesion forces between bacteria or between bacteria and the substrate. We thus assume 933 that there is no strong difference in the binding energy of a cell to the substrate as a function of its rigidity, and that this energy is slightly higher than that of binding to another cell. Bacteria thus 935 prefer adhering to the substrate in all cases but can easily adhere to other cells if needed. Based 936 on this assumption, we consider two key features of surface colonization to describe the 2D to 3D 937 transition: 038

• Growth: initially, at time t = 0, one bacterium is attached to the surface. The number of bacteria N grows exponentially with time as:

$$N(t) = e^{\frac{t}{t_0}} . (S.27)$$

The characteristic time scale, t_0 , accounts both for the growth and for the occasional detachment of bacteria from the surface. *De novo* attachment of bacteria to the surface is neglected. Furthermore, t_0 is assumed to be constant over time and across the different surfaces.

• Movement: bacteria explore the surface with a characteristic velocity V_{CM} and perform a random walk (we consider time and length scales larger than the persistence length/time of bacterial twitching motion). These displacements result in a spreading of the colony over a characteristic area a(t) following a diffusive process:

$$a(t) = a_0 + \alpha V_{CM} t \tag{S.28}$$

where a_0 is the area of one bacterium and α is a phenomenological parameter related to the properties of the random walk.

From the two equations above, it is clear that the number of bacteria attached to the surface grows faster than the size of the corresponding colony. Therefore, at a critical time t_c corresponding to a critical number of bacteria N_c on the substrate, the area available to bacteria for spreading on the

surface will be completely occupied. i.e. $a_0 N_c = a_0 N(t_c) = a(t_c)$ and thus

$$N_c = 1 + \gamma V_{CM} \ln(N_c) \tag{S.29}$$

where we have substituted $\ln(N_c)$ for t_c on the *r.h.s.* of Eq. (5.29) and $\gamma = \alpha t_0/a_0$. Solving this equation permits to obtain N_c as a function of V_{CM} and one unknown parameter, γ . Note that when $V_{CM} = 0$, a situation in which the bacteria do not move at all, the 2D to 3D transition occurs at the first division, i.e. as soon as $N_c > 1$.

958

It should be noted that here, just as V_{CM} is a characteristic velocity and not the mean speed of 959 the bacteria (see main text, Fig. 3A), that the characteristic area a(t) accessible to the bacteria in 960 the colony at time t is not necessarily equal to the whole colony area: first because of their finite 961 center-of-mass velocity V_{CM} ; secondly, because the local density may restrict their movement and 962 the accessible surface. This effect is difficult to quantify because the fluctuations of density inside 063 the colony area may be, depending on V_a, much greater than the ones encountered in the case of 964 the Brownian diffusion of particles. Indeed, some bacteria remain static while others explore the 965 surface extensively (main text, Fig. 3E). Another reason is that upon division, the two daughter cells 966 are touching and there is hence a systematic fluctuation of density upon division. Therefore the 967 area accessible for bacteria is rather an effective measure, which cannot be directly derived from 069

microscopic diffusion processes only. While other expressions could be used, this one is the sim-969 plest that can be proposed and matches our experimental data sufficiently well. One justification 970 is that the underlying assumption that the velocity of bacteria is not affected by the local density 971 (retaining the linear scaling of a(t) with V_{CM}) is justified in the assessed situation where groups of 972 closed-packed bacteria never exceed 5-8 cells before the 2D to 3D transition occurs. However, as 973 a comparison, sub- and super-linear scalings will be compared with experimental data in the next section 975 976 To further analyze the microscopic meaning of γ , we note that it is the inverse of a velocity and 977 is related to the compactness of the colony, with higher values indicating a sparser distribution 978 of bacteria with a lesser probability that growing/twitching bacteria will encounter several others 970

and move to 3D because of local crowding. However it is misleading to compare it to values that could be derived from random walks with persistence because of the above-mentioned discrepancy between the colony area and the area accessible to bacteria for further spreading. Relating γ to experimentally measured quantities on the cell movement would require a detailed analysis of

the cell density fluctuations on the surface which is beyond the scope of this paper.

B. Comparison of experimental data with the model

All available data from which characteristic velocities were extracted (main text, Fig. 3B) were analysed and included, with the exception of one data point on glass due to the presence of an air bubble on the surface before the onset of the $2D \rightarrow 3D$ transition. The characteristic velocities for each experiment and each rigidity were taken from Fig. 3B. The characteristic number of bacteria N_c per colony was estimated as follows.

• First, for low- and medium-ridigity surfaces (2.7 kPa, 6.1 kPa and 18.5 kPa), colony formation 991 from isolated bacteria was monitored over time until the 2D to 3D transition occurs. The 992 number of bacteria on the surface stemming from the initial isolated bacteria were then 993 counted, and the count for all the colonies were averaged to calculate N. In addition, the 994 average number of colonies forming in the observed area up to that point was also measured. 995 For higher-rigidity surfaces, the movement of bacteria is too large to keep track of all bacteria 996 stemming from the same progenitor as they mix or leave the field of view, while others are 997 incoming. As a result, N_a was calculated by counting the total number of bacteria in the field 998 of view at the time of the onset of the 2D to 3D transition, and dividing this number by the 999 estimated number of colonies as measured on low-rigidity surfaces. It should be noted that 1000 in this case, the simple model presented above is not valid as it considers only one isolated 1001 colony, and can be expected to yield overestimated values of $N_{\rm o}$. Furthermore, our evaluation 1002 method of the number of colonies in the field of view may be prope to error so we used a 1003 "blind" evaluation procedure performed before the count of bacteria in the field of view, to 1004 avoid possible biases. A change of 1 (compared to a mean value around 4) in the number 100 of colonies used to normalise the total number of bacteria provides a good estimate of the 1006 error bars on each individual data point, and is comparable to the spread of the data points 1007 (see Appendix 2-figure 1). When several surfaces of the same rigidity have been measured in 1008 one experiment, the different N_c values are averaged. 1009

The above cited procedure produced one doublet (V_{CM}, N_c) for each rigidity of each experiment. To match our simple model with the experimental data, Eq. (S.29) can be used to calculate N_c as a function of V_{CM} for a given value of γ . However, a direct fit of the experimental data is difficult as there is no analytical solution to Eq. (S.29). Instead, an experimental value of γ was calculated from each experimental point using the expression

$$\gamma_{exp} = \frac{N_c - 1}{V_{CM} \ln(N_c)} \tag{S.30}$$

An average experimental value is then calculated, along with a standard deviation, $\langle \gamma_{exp} \rangle = 56.8$ min. μ m⁻¹ and $\delta \gamma = 11.2$ min. μ m⁻¹.



Appendix 2-figure 1: (a) comparison between the experimental data (markers) and the kinetic model (lines). Blue dots are data obtained with the pili-deficient mutant pilA : Tn5. Inset, residual of the fit of all the experimental data points as a function of the exponent value for V_{CM} , indicating that the best fit is achieved for a value of or close to 1. (b) γ_{exp} values for all data points (black disks) and their average (red line). Gray squares are data points with $v_0 < 2 \,\mu$ m.min⁻¹, showing a similar distribution and thus ruling out a significant bias at high velocities. (c) same dataset as in a but the surface density at the transition is plotted; (d) same dataset as in a but the time at the transition is plotted. The same model is used to describe the data, but converted into the proper quantities.

Appendix 2-figure 1a shows the experimental data $(N_c - 1)/Log(N_c)$ as a function of V_{CM} (each marker corresponds to a different experiment), and the corresponding theoretical straight lines with slopes $\langle \gamma_{exp} \rangle$ (solid line), and $\langle \gamma_{exp} \rangle$ +/- 1 standard deviation (dotted lines). To assess the deviation from the curve at high velocities, $\langle \gamma_{exp} \rangle$ was also calculated from all data points with $v_0 <$ $2 \,\mu$ m.min⁻¹ but the change in the value is minimal (58.2 min. μ m⁻¹ instead of 56.8 min. μ m⁻¹, see Appendix 2-figure 1b).

Our strongest assumption in this modelling is the expression of a(t) as a function of V_{CM} [Eq. (S.28)]: an obvious *a posteriori* evidence for its correctness is that the derived equation fits our data well over more than one decade in velocity. To strengthen our point, however, we have also calculated similar curves using an exponent for V_{CM} ranging from 0.5 to 1.5 (steps of 0.1, Appendix 2-figure 1a, inset): the comparison with experimental data indicates that reasonable agreement is only obtained for exponent values between 0.8 and 1.1, at most.

Finally we would like to point out that the data do not collapse as well when plotting the density 1030 of bacteria, or the time of the $2D \rightarrow 3D$ transition (Appendix 2-figures 1c and 1d). A likely explana-1031 tion is that the initial number of bacteria on the surface varies between different datasets, a bias 1032 that is cancelled when plotting the number of bacteria instead of the density or the time at the 1033 onset of the transition. The same model is used with the same average parameter and spread, 1034 but converted into the proper quantities; for the density, the curves in Appendix 2-figure 1a are 1035 multiplied by the average number of colonies per observed area (3 colonies), and divided by the 1036 image area (26121 μ m²); for the transition time, the logarithm of the number of bacteria per colony 1037 at the transition is multiplied by the typical growth time of the number of bacteria on the surface 1038 (\sim 40 min). This time incorporates both the division time (\sim 30 min) and the departure of a fraction 1039 of the bacteria from the surface. 1040

1041 References

Abidine Y, Laurent VM, Michel R, Duperray A, Palade LI, Verdier C. Physical properties of polyacrylamide gels probed by AFM and rheology. Europhys Lett. 2015 FEB; 109(3). doi: {10.1209/0295-5075/109/38003}.

Aoudia N, Rieu A, Briandet R, et al. Biofilms of Lactobacillus plantarum and Lactobacillus fermentum: Effect
 on stress responses, antagonistic effects on pathogen growth and immunomodulatory properties. Food
 Microbiol. 2016; 53(Pt A):51–59. https://doi.org/10.1016/j.fm.2015.04.009.

Arganda-Carreras I, Kaynig V, Rueden C, Eliceiri KW, Schindelin J, Cardona A, Sebastian Seung H. Trainable
 Weka Segmentation: a machine learning tool for microscopy pixel classification. Bioinformatics. 2017;
 33(15):2424–2426.

Armbruster CR, Lee CK, Parker-Gilham J, de Anda J, Xia A, Zhao K, Murakami K, Tseng BS, Hoffman LR, Jin F,
 Harwood CS, Wong GC, Parsek MR. Heterogeneity in surface sensing suggests a division of labor in Pseudomonas aeruginosa populations. eLife. 2019; 8.

Beamish JA, Zhu J, Kottke-Marchant K, Marchant RE. The effects of monoacrylated poly(ethylene glycol) on the
 properties of poly(ethylene glycol) diacrylate hydrogels used for tissue engineering. J Biomed Mater Res A.
 2010 FEB; 92A(2):441-450. doi: {10.1002/jbm.a.32353}.

Beaussart A, Baker AE, Kuchma SL, El-Kirat-Chatel S, O'Toole GA, Dufrêne YF. Nanoscale adhesion forces of Pseudomonas aeruginosa type IV pili. ACS nano. 2014; 8(10):10723–10733.

 Initial Björnham O, Axner O. Catch-Bond Behavior of Bacteria Binding by Slip Bonds. Biophysical Journal. 2010 Sep;

 1059
 99(5):1331–1341. https://doi.org/10.1016/j.bpj.2010.06.003, doi: 10.1016/j.bpj.2010.06.003.

Boucher HW, Talbot GH, Bradley JS, et al. Bad bugs, no drugs: no ESKAPE! An update from the Infectious
 Diseases Society of America. Clin Infect Dis. 2009; 48(1):1–12. https://doi.org/10.1086/599017.

1062 Chang CY. Surface sensing for biofilm formation in Pseudomonas aeruginosa. Front Microbiol. 2018; 8:2671.

Conrad JC, Gibiansky ML, Jin F, Gordon VD, Motto DA, Mathewson MA, Stopka WG, Zelasko DC, Shrout JD, Wong
 GCL. Flagella and Pili-Mediated Near-Surface Single-Cell Motility Mechanisms in P. aeruginosa. Biophys J.
 2011 APR 6; 100(7):1608–1616. doi: {10.1016/j.bpj.2011.02.020}.

Cont A, Rossy T, Al-Mayyah Z, Persat A. Biofilms deform soft surfaces and disrupt epithelia. eLife. 2020; 9.
 https://doi.org/10.7554/elife.56533.

Cont A, Vermeil J, Persat A. Material Substrate Physical Properties Control Pseudomonas aeruginosa Biofilm
 Architecture. mBio. 2023; p. e03518–22. https://journals.asm.org/doi/abs/10.1128/mbio.03518-22, doi:
 10.1128/mbio.03518-22.

Cornforth DM, Dees JL, Ibberson CB, Huse HK, Mathiesen IH, Kirketerp-Møller K, Wolcott RD, Rumbaugh KP, Bjarnsholt T, Whiteley M. Pseudomonas aeruginosa transcriptome during human infection.
 Proc Natl Acad Sci USA. 2018; 115(22):E5125–E5134. https://www.pnas.org/content/115/22/E5125, doi: 10.1073/pnas.1717525115.

Cowles KN, Gitai Z. Surface association and the MreB cytoskeleton regulate pilus production, localization and function in Pseudomonas aeruginosa. Mol Microbiol. 2010 Jun; 76(6):1411–1426.

Craig L, Forest KT, Maier B. Type IV pili: dynamics, biophysics and functional consequences. Nature Reviews
 Microbiology. 2019; 17(7):429–440. https://doi.org/10.1038/s41579-019-0195-4, doi: 10.1038/s41579-019-0195-4.

DeLeon S, Clinton A, Fowler H, Everett J, Horswill AR, Rumbaugh KP. Synergistic interactions of Pseudomonas
 aeruginosa and Staphylococcus aureus in an in vitro wound model. Infection and Immunity. 2014 Nov;
 82(11):4718–4728.

 Dötsch A, Eckweiler D, Schniederjans M, et al. The *Pseudomonas aeruginosa* transcriptome in planktonic cultures and static biofilms using RNA sequencing. PloS One. 2012; 7(2):e31092. https://journals.plos.org/ plosone/article?id=10.1371/journal.pone.0031092.

Dufrêne YF. Sticky microbes: forces in microbial cell adhesion. Trends Microbiol. 2015; 23(6):376–382. https://doi.org/10.1016/j.tim.2015.01.011.

Dufrêne YF, Persat A. Mechanomicrobiology: how bacteria sense and respond to forces. Nat Rev Microbiol. 2020; 18:1–14.

Duvernoy MC, Mora T, Ardré M, et al. Asymmetric adhesion of rod-shaped bacteria controls microcolony morphogenesis. Nat comm. 2018; 9(1):1–10. https://doi.org/10.1038/s41467-018-03446-y.

Evans E. Probing the relation between forceâlifetimeâand chemistry in single molecular bonds. Annual review
 of biophysics and biomolecular structure. 2001; 30(1):105–128.

Furukawa S, Kuchma SL, O'Toole GA. Keeping Their Options Open: Acute versus Persistent Infections. Journal of Bacteriology. 2006 Feb; 188(4):1211–1217.

 Gellatly SL, Hancock REW. Pseudomonas aeruginosa : new insights into pathogenesis and host defenses . Pathogens and Disease. 2013 04; 67(3):159–173. https://doi.org/10.1111/2049-632X.12033, doi: 10.1111/2049-632X.12033.

Grant MAA, Waclaw B, Allen RJ, Cicuta P. The role of mechanical forces in the planar-to-bulk transition in growing Escherichia coli microcolonies. J Roy Soc Inter. 2014 AUG 6; 11(97). doi: {10.1098/rsif.2014.0400}.

Islam S, Krachler AM. Mechanosensing regulates virulence *in Escherichia coli* O157:H7. Gut Microbes. 2016;
 7(1):63–67. https://doi.org/10.1080/19490976.2015.1121365.

 Jackson K, Starkey M, Kremer S, Parsek M, Wozniak D. Identification of psl, a locus encoding a potential exopolysaccharide that is essential for Pseudomonas aeruginosa PAO1 biofilm formation. J Bacteriol. 2004 JUL;
 186(14):4466-4475. doi: {10.1128/JB.186.14.4466-4475.2004}.

Jacobs MA, Alwood A, Thaipisuttikul I, Spencer D, Haugen E, Ernst S, Will O, Kaul R, Raymond C, Levy R, Chun Rong L, Guenthner D, Bovee D, Olson MV, Manoil C. Comprehensive transposon mutant library of Pseu domonas aeruginosa. Proc Natl Acad Sci USA. 2003; 100(24):14339–14344. https://www.pnas.org/content/
 100/24/14339, doi: 10.1073/pnas.2036282100.

Jin F, Conrad JC, Gibiansky ML, Wong GCL. Bacteria use type-IV pili to slingshot on surfaces. Proc Natl Acad Sci USA. 2011; 108(31):12617–12622. https://www.pnas.org/content/108/31/12617, doi: 10.1073/pnas.1105073108.

Koch MD, Black ME, Han E, Shaevitz JW, Gitai Z. <i>Pseudomonas aeruginosa</i> distinguishes
 surfaces by stiffness using retraction of type IV pili. Proceedings of the National Academy of
 Sciences. 2022; 119(20):e2119434119. https://www.pnas.org/doi/abs/10.1073/pnas.2119434119, doi:
 10.1073/pnas.2119434119.

Koch MD, Fei C, Wingreen NS, Shaevitz JW, Gitai Z. Competitive binding of independent extension and retraction motors explains the quantitative dynamics of type IV pili. Proceedings of the National Academy of Sciences. 2021 Feb; 118(8):e2014926118. https://doi.org/10.1073/pnas.2014926118, doi: 10.1073/pnas.2014926118.

Kramers HA. Brownian motion in a field of force and the diffusion model of chemical reactions. Physica. 1940 Apr; 7(4):284–304. https://doi.org/10.1016/s0031-8914(40)90098-2, doi: 10.1016/s0031-8914(40)90098-2.

Kühn MJ, Talà L, Inclan YF, Patino R, Pierrat X, Vos I, Al-Mayyah Z, Macmillan H, Negrete J, Engel JN, Persat A.
 Mechanotaxis directs Pseudomonas aeruginosa twitching motility. Proc Natl Acad Sci USA. 2021; 118(30).
 https://www.pnas.org/content/118/30/e2101759118, doi: 10.1073/pnas.2101759118.

Kuik J, Vincent S, Leeflang B, Kroon-Batenburg LMJ, Kamerling J. Conformational analysis in aqueous solution and estimation of the persistence length of exopolysaccharides produced by Lactobacillus helveticus
 Lh59 and Streptococcus macedonicus Sc136. Canadian Journal of Chemistry. 2011 02; 84:730–742. doi: 10.1139/v06-055.

Landau LD, Lifschitz EM. Theory of Elasticity, vol. 7 of Theoretical Physics. 3 ed. Butterworth Heinemann; 2004.

Laventie BJ, Jenal U. Surface sensing and adaptation in bacteria. Annu Rev Microbiol. 2020; 74:735–760.

Laventie BJ, Sangermani M, Estermann F, Manfredi P, Planes R, Hug I, Jaeger T, Meunier E, Broz P, Jenal U. A

Surface-Induced Asymmetric Program Promotes Tissue Colonization by Pseudomonas aeruginosa. Cell Host
 Microbe. 2018 Dec; p. 1–20.

Leighton TL, Buensuceso RNC, Howell PL, Burrows LL. Biogenesis of Pseudomonas aeruginosa type IV pili and regulation of their function. Environmental Microbiology. 2015 Nov; 17(11):4148–4163.

Liu Y, Yang CH, Li J. Influence of extracellular polymeric substances on Pseudomonas aeruginosa transport and deposition profiles in porous media. Environ Sci Technol. 2007 JAN 1; 41(1):198–205. doi: 1138 {10.1021/es061731n}.

 Livingston J, Spero MA, Lonergan ZR, Newman DK. Visualization of mRNA Expression in Pseudomonas aeruginosa Aggregates Reveals Spatial Patterns of Fermentative and Denitrifying Metabolism. bioRxiv. 2022; https://www.biorxiv.org/content/early/2022/03/12/2022.03.11.484052, doi: 10.1101/2022.03.11.484052.

- Lu S, Giuliani M, Harvey H, LL B, Wickham R, Dutcher J. Nanoscale Pulling of Type IV Pili Reveals Their Flexibility and Adhesion to Surfaces over Extended Lengths of the Pili. . 2015; 108(12):2865–2875.
- Ma L, Lu H, Sprinkle A, Parsek MR, Wozniak DJ. Pseudomonas aeruginosa PSI is a galactose- and mannose-rich
 exopolysaccharide. J Bacteriol. 2007 NOV; 189(22):8353–8356. doi: {10.1128/JB.00620-07}.
- Maier B, Wong GCL. How Bacteria Use Type IV Pili Machinery on Surfaces. Trends Microbiol. 2015; 23(12):775–788. https://doi.org/10.1016/j.tim.2015.09.002.
- Mann EE, Wozniak DJ. Pseudomonas biofilm matrix composition and niche biology. FEMS Microbiol Rev. 2012
 JUL; 36(4):893–916. doi: {10.1111/j.1574-6976.2011.00322.x}.
- Marathe R, Meel C, Schmidt NC, Dewenter L, Kurre R, Greune L, Schmidt MA, Müller MJI, Lipowsky R, Maier B, Klumpp S, Bacterial twitching motility is coordinated by a two-dimensional tug-of-war with di-
- rectional memory. Nature Communications. 2014 May; 5(1). https://doi.org/10.1038/ncomms4759, doi:
- 1153 10.1038/ncomms4759.
- Marko JF, Siggia ED. Statistical mechanics of supercoiled DNA. Physical Review E. 1995; 52(3):2912.
- Merz AJ, So M, Sheetz MP. Pilus retraction powers bacterial twitching motility. Nature. 2000; 407(6800):98–102.
 https://doi.org/10.1038/35024105.
- **Moradali MF**, Ghods S, Rehm BHA. Pseudomonas aeruginosa Lifestyle: A Paradigm for Adaptation, Survival, and Persistence. Front Cell Infect Microbiol. 2017; 7:39. https://doi.org/10.3389/fcimb.2017.00039.
- Moran PA. Notes on continuous stochastic phenomena. Biometrika. 1950; 37(1/2):17–23.
- **ONeal L**, Baraquet C, Suo Z, Dreifus JE, Peng Y, Raivio TL, Wozniak DJ, Harwood CS, Parsek MR. The Wsp system of <i>Pseudomonas aeruginosa</i> links surface sensing and cell envelope stress. Proceedings of the
- 1162
 National Academy of Sciences. 2022; 119(18):e2117633119. https://www.pnas.org/doi/abs/10.1073/pnas.

 1163
 2117633119, doi: 10.1073/pnas.2117633119.
- Orazi G, O'Toole GA, Dunman P. *Pseudomonas aeruginosa* Alters *Staphylococcus aureus* Sensitivity to Vancomycin in a Biofilm Model of Cystic Fibrosis Infection. mBio. 2017; 8(4):e00873–17. https://journals.asm.org/doi/abs/ 10.1128/mBio.00873-17. doi: 10.1128/mBio.00873-17.
- **O'Toole GA**, Kolter R. Flagellar and twitching motility are necessary for Pseudomonas aeruginosa biofilm development. Mol Microbiol. 1998 Oct; 30(2):295–304.
- Pang Z, Raudonis R, Glick BR, Lin TJ, Cheng Z.Antibiotic resistance in Pseudomonas aerug-1170inosa: mechanisms and alternative therapeutic strategies.Biotechnology Advances.2019;117137(1):177–192.https://www.sciencedirect.com/science/article/pii/S0734975018301976,doi:1172https://doi.org/10.1016/j.biotechadv.2018.11.013.
- Pereverzev YV, Prezhdo OV, Forero M, Sokurenko EV, Thomas WE. The Two-Pathway Model for the Catch-Slip
 Transition in Biological Adhesion. Biophysical Journal. 2005 Sep; 89(3):1446–1454. https://doi.org/10.1529/
 biophysj.105.062158, doi: 10.1529/biophysj.105.062158.
- **Persat A**, Inclan YF, Engel JN, et al. Type IV pili mechanochemically regulate virulence factors in *Pseudomonas aeruginosa*. Proc Natl Acad Sci USA. 2015; 112(24):7563–7568. https://doi.org/10.1073/pnas.1502025112.
- Persat A, Nadell CD, Kim MK, et al. The Mechanical World of Bacteria. Cell. 2015; 161(5):988–997. https:

 1179
 //doi.org/10.1016/j.cell.2015.05.005.
- Reichhardt C, Wong C, Passos da Silva D, Wozniak DJ, Parsek MR. CdrA Interactions within the Pseudomonas
 aeruginosa Biofilm Matrix Safeguard It from Proteolysis and Promote Cellular Packing. MBio. 2018 Sep; 9(5).
- **Robert P**, Benoliel AM, Pierres A, Bongrand P. What is the biological relevance of the specific bond properties revealed by single-molecule studies? Journal of Molecular Recognition: An Interdisciplinary Journal, 2007:
- revealed by single-molecule studies? Journal of Molecular Recognition: An Interdisciplinary Journal. 2007;
 20(6):432–447.
- Rodesney CA, Roman B, Dhamani N, Cooley BJ, Touhami A, Gordon VD. Mechanosensing of shear by Pseudomonas aeruginosa leads to increased levels of the cyclic-di-GMP signal initiating biofilm development.
 Proc Natl Acad Sci USA. 2017 JUN 6; 114(23):5906–5911. doi: {10.1073/pnas.1703255114}.
- **Rybtke MT**, Borlee BR, Murakami K, et al. Fluorescence-based reporter for gauging cyclic di-GMP levels in Pseudomonas aeruginosa. Appl Environ Microbiol. 2012; 78(15):5060–5069.

- Saha N, Monge C, Dulong V, et al. Influence of Polyelectrolyte Film Stiffness on Bacterial Growth. Biomacromol.
 2013; 14(2):520–528. https://doi.org/10.1021/bm301774a.
- **Sahoo PK**, Janissen R, Monteiro MP, et al. Nanowire Arrays as Cell Force Sensors To Investigate Adhesin-Enhanced Holdfast of Single Cell Bacteria and Biofilm Stability. Nano Letters. 2016; 16(7):4656–4664. https:
- 1194 //doi.org/10.1021/acs.nanolett.6b01998.
- 1195 Sens P. Rigidity sensing by stochastic sliding friction. EPL (Europhysics Letters). 2013; 104(3):38003.
- Shen Y, Siryaporn A, Lecuyer S, et al. Flow Directs Surface-Attached Bacteria to Twitch Upstream. Biophys J. 2012; 103(1):146–151. https://doi.org/10.1016/j.bpj.2012.05.045.
- **Shrout JD**, Tolker-Nielsen T, Givskov M, Parsek MR. The contribution of cell-cell signaling and motility to bacterial biofilm formation. MRS Bull. 2011 May; 36(5):367–373.
- Simsek AN, Braeutigam A, Koch MD, Shaevitz JW, Huang Y, Gompper G, Sabass B. Substrate-rigidity dependent
 migration of an idealized twitching bacterium. Soft Matter. 2019; 15(30):6224–6236. https://doi.org/10.1039/
 c9sm00541b, doi: 10.1039/c9sm00541b.
- Siryaporn A, Kuchma SL, O'Toole GA, Gitai Z. Surface attachment induces Pseudomonas aeruginosa virulence.
 Proc Natl Acad Sci U S A. 2014 Nov; 111(47):16860–16865.
- Skerker JM, Berg HC. Direct observation of extension and retraction of type IV pili. Proc Natl Acad Sci USA.
 2001; 98(12):6901–6904.
- Song F, Ren D. Stiffness of Cross-Linked Poly(Dimethylsiloxane) Affects Bacterial Adhesion and Antibiotic Susceptibility of Attached Cells. Langmuir. 2014; 30(34):10354–10362. https://doi.org/10.1021/la502029f, doi: 10.1021/la502029f.
- Song F, Wang H, Sauer K, Ren D. Cyclic-di-GMP and oprF Are Involved in the Response of *Pseudomonas aerug-inosa* to Substrate Material Stiffness during Attachment on Polydimethylsiloxane (PDMS). Front Microbiol. 2018; 9:110. https://doi.org/10.3389/fmicb.2018.00110.
- Song F, Wang H, Sauer K, Ren D. Cyclic-di-GMP and oprF Are Involved in the Response of Pseudomonas aeruginosa to Substrate Material Stiffness during Attachment on Polydimethylsiloxane (PDMS). Frontiers in Microbiology. 2018: 9. https://www.frontiersin.org/article/10.3389/fmicb.2018.00110, doi: 10.3389/fmicb.2018.00110.
- Stoodley P, Sauer K, Davies D, Costerton J. Biofilms as complex differentiated communities. Annu Rev Microbiol.
 2002; 56:187–209. doi: {10.1146/annurev.micro.56.012302.160705}.
- **Talà L**, Fineberg A, Kukura P, Persat A. Pseudomonas aeruginosa orchestrates twitching motility by sequential control of type IV pili movements. Nat Microbiol. 2019; 4(5):774–780.
- **Thevenaz P**, Ruttimann UE, Unser M. A pyramid approach to subpixel registration based on intensity. IEEE transactions on image processing. 1998; 7(1):27–41.
- Tinevez JY, Perry N, Schindelin J, Hoopes GM, Reynolds GD, Laplantine E, Bednarek SY, Shorte SL, Eliceiri KW.
 TrackMate: An open and extensible platform for single-particle tracking. Methods. 2017; 115:80–90.
- Tse JR, Engler AJ. Preparation of hydrogel substrates with tunable mechanical properties. Curr Protoc Cell Biol.
 2010; 47:10.16. https://doi.org/10.1002/0471143030.cb1016s47.
- Valentini M, Filloux A. Biofilms and Cyclic di-GMP (c-di-GMP) Signaling: Lessons from Pseudomonas aerugi nosaand Other Bacteria. Journal of Biological Chemistry. 2016 Jun; 291(24):12547–12555.
- Wadhwa N, Berg HC. Bacterial motility: machinery and mechanisms. Nature Reviews Microbiology. 2022 Mar;
 20(3):161–173.
- Webster SS, Mathelié-Guinlet M, Verissimo AF, Schultz D, Viljoen A, Lee CK, Schmidt WC, Wong GCL, Dufrêne
 YF, O'Toole GA, Kline KA. Force-Induced Changes of PilY1 Drive Surface Sensing by Pseudomonas
 aeruginosa. mBio. 2022; 13(1):e03754–21. https://journals.asm.org/doi/abs/10.1128/mbio.03754-21, doi:
 10.1128/mbio.03754-21.
- Williams T, Kelley C, many others, Gnuplot 5.2: an interactive plotting program; 2019. http://gnuplot.sourceforge.
 net/.
- 1236Zhao K, Tseng BS, Beckerman B, et al. Psl trails guide exploration and microcolony formation in Pseudomonas1237aeruginosa biofilms. Nature. 2013; 497(7449):388–391. https://doi.org/10.1038/nature12155.