1	Stochastic bacterial population dynamics prevent the emergence of
2	antibiotic resistance
3	
4	Helen K. Alexander*, R. Craig MacLean
5	Department of Zoology, University of Oxford, Oxford, OX1 3PS, United Kingdom
6	* Corresponding author: helen.alexander@zoo.ox.ac.uk
7	
8	Keywords: antimicrobial resistance, Pseudomonas aeruginosa, streptomycin, minimum inhibitory
9	concentration (MIC), inoculum effect, mathematical model, extinction probability, demographic
10	stochasticity, evolutionary rescue, experimental evolution
11	
12	Abstract:
13	Understanding how antibiotic exposure impacts the evolution of resistance is key to optimizing
14	antibiotic treatment strategies. The conventional approach to this problem is to measure the range
15	of antibiotic concentrations over which resistance is selectively favoured – the "mutant selection
16	window". Here we take an alternative approach by investigating how antibiotic concentration
17	impacts de novo establishment of resistance from single cells of a streptomycin-resistant
18	Pseudomonas aeruginosa strain. We show that demographic stochasticity prevents outgrowth of
19	resistant cells with >95% probability across ~90% of the range of streptomycin concentrations where
20	resistance is selectively favoured. This effect occurs because exposure to streptomycin extends lag
21	time and increases the mortality rate of streptomycin-resistant cells. Our study highlights
22	fundamental differences between the processes that drive emergence versus spread of resistance. It
23	suggests that moderate doses of antibiotics, within the traditional mutant selection window, may
24	effectively prevent emergence of resistance in cases where transmission of resistant strains is
25	negligible, for instance when using new antibiotics.
26	
27	
28	Introduction
29	
30	The emergence of antibiotic resistance is a key obstacle to successful treatment of bacterial
31	infections, with resistant infections associated with poorer clinical outcomes for patients [1]. A
32	better understanding of how antibiotic dosing drives resistance evolution could aid the design of
33	more effective treatment strategies.

### Stochastic bacterial population dynamics

2

0.	
35	A crucial question is how evolution of resistance depends on antibiotic concentration. A major step
36	towards answering this question was the formulation of the mutant selection window (MSW)
37	hypothesis, which proposes that resistance arises within a predictable range of antibiotic
38	concentrations [2] [3]. Specifically, the lower boundary of this window is the minimum inhibitory
39	concentration (MIC) of the sensitive strain, i.e. the concentration that is sufficient to prevent its
40	growth. The upper boundary, the so-called "mutant prevention concentration", was originally
41	defined as the minimal concentration at which no colonies are formed when plating a culture of at
42	least 10 <sup>10</sup> cells [2] [3], and gradually became equated with the MIC of the most resistant strain
43	available [4] [5]. The MSW hypothesis has become an important paradigm in the field and a
44	framework for suggesting dosing strategies to avoid emergence of resistance [4] [5] [6].
45	
46	In recent years, it has been recognized both on theoretical grounds and with accumulating empirical
47	evidence that higher resistance can also be selectively favoured below the MIC of a sensitive or less-

48 resistant strain [7] [8] [9] [10]. This selection occurs because sub-MIC concentrations of antibiotic

49 can already be sufficiently detrimental to render the sensitive strain less fit than the resistant strain.

50 Indeed, the "minimal selective concentration" (MSC) favouring resistance can be as low as a few

51 percent of the sensitive strain's MIC [9]. Thus, the window of antibiotic concentrations over which

52 resistance can emerge is potentially very wide.

53

34

54 Selection is expected to operate efficiently when both sensitive and resistant cell populations are 55 reasonably large, resulting in an increase in frequency of the fitter strain (i.e. of the resistant strain 56 at concentrations beyond the MSC). However, a crucial distinction must be drawn between 57 situations in which resistance is already present at appreciable frequency, and cases in which it 58 emerges de novo [11]. If a resistant strain is not already present, it must first arise (e.g. by mutation 59 or horizontal gene transfer), and then increase in frequency from a single cell to a sizeable 60 population. Although the spread of a common resistant strain is a more or less deterministic process 61 driven by selective differences, the establishment of an initially rare resistant strain is highly 62 stochastic [11]. That is, due to chance events on the individual cell level (division and death), a 63 resistance mutation present in only a few cells may be lost even when it is selectively favoured. 64 Indeed, predicting the probability that beneficial mutations escape stochastic loss when rare is a 65 classical problem in population genetics [12].

### Stochastic bacterial population dynamics

3

67 While a large body of work has addressed the question of antibiotic concentrations at which 68 resistance is selectively favoured, it is largely unknown how antibiotic concentration impacts the 69 stochastic establishment phase of de novo resistant mutants. Theory predicts that the probability of 70 establishment gradually decreases as cell division rate decreases or death rate increases, even while 71 the expected population net growth rate remains positive, albeit reduced [12]. Thus, if antibiotics 72 have detrimental effects on resistant cells below the resistant strain's MIC, it is possible that small 73 initial populations of cells are stochastically lost and thus establishment fails, despite being within 74 the MSW.

75

76 We set out to test this idea in the bacterium *Pseudomonas aeruginosa* challenged with the 77 aminoglycoside antibiotic streptomycin. We used a strain carrying a plasmid that confers 78 streptomycin resistance through production of an enzyme (aadA5) that deactivates the antibiotic 79 intracellularly. To study stochastic establishment, we tested growth in hundreds of cultures 80 inoculated with resistant cells at extremely low density (around a single cell per culture). Consistent 81 with an important role for stochasticity, under identical culture conditions some replicate cultures 82 grew and others did not. By fitting a mathematical model to these data, we were able to quantify 83 the probability that a single resistant cell establishes a large surviving population across a range of 84 antibiotic concentrations. This establishment probability began to decrease at concentrations below 85 the resistant strain's MIC (MIC<sub>R</sub>), suggesting that antibiotics are exerting detrimental effects on 86 resistant cells. This decline occurred at surprisingly low antibiotic concentrations: significant effects 87 were seen as low as  $1/16 \times MIC_{R}$ , and we estimate that establishment from single cells succeeds less 88 than 5% of the time at  $1/8 \times MIC_{R}$ . We show that these results are consistent with a simple model of 89 how inoculum size affects population growth, whereby each individual cell has a small, independent 90 chance of establishment. Collectively, our results imply that *de novo* emergence of resistance may 91 be unlikely throughout a substantial portion of the mutant selection window.

- 92
- 93

# 94 **Results**

95

## 96 *Resistance is selectively favoured over a wide range of streptomycin concentrations*

- 97
- 98 We measured the impact of antibiotic exposure on selection for resistance by directly competing
- 99 fluorescently labelled sensitive and resistant strains across a gradient of streptomycin
- 100 concentrations, ranging from 1/16 to 16 times the MIC of the sensitive strain (MIC<sub>s</sub> = 16  $\mu$ g/ml; [13]),

4

101	corresponding to 1/2048 to 1/8 times the MIC of the resistant strain (MIC <sub>R</sub> = 2048 $\mu$ g/ml; [13]).
102	Strains were inoculated at approximately equal starting frequencies, with a total density around 5 x
103	$10^5$ CFU/ml, and their final frequencies were evaluated by flow cytometry after 24h (Methods and
104	Suppl. Fig. 1). An increase in resistant frequency to significantly above 50% thus indicates selection
105	for resistance. We were unable to quantitatively estimate the selection coefficient of the resistant
106	strain, because virtually no sensitive cells were recovered after culturing with streptomycin above
107	MICs, leading to 100% final resistance frequency (Fig. 1, Suppl. Fig. 2 and Suppl. Table 1). Strikingly,
108	resistance was still strongly selected as low as $1/4 \times MIC_s$ (two-sided t-test: $p = 5e-6$ ), while
109	sensitivity was only significantly favoured in the complete absence of streptomycin ( $p = 4e-3$ ). Based
110	on the point at which final resistance frequency crosses the 50% mark, we estimate that the MSC of
111	streptomycin lies between $1/16$ and $1/8 \times MIC_s$ , in agreement with previous results for these strains
112	[13].
113	

- 114 These results demonstrate that the resistant strain, once present at an appreciable population size,
- is expected to spread by outcompeting the sensitive strain over a wide range of streptomycin
- 116 concentrations, ranging from well below (1/4x or less) to well above (at least 16x) the MIC of the

117 sensitive strain (MIC<sub>s</sub>).

118

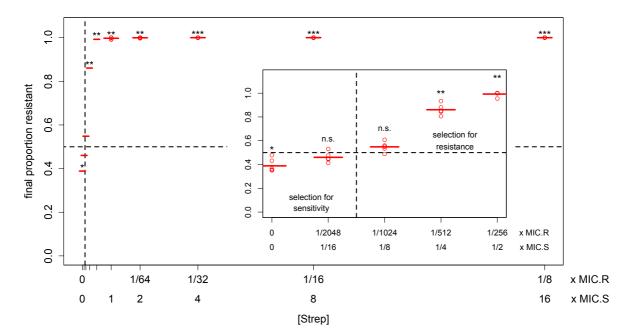




Figure 1: Evaluating selection for resistance by competition assays across streptomycin concentrations. Concentration on the x-axis is scaled by the previously measured MIC values of these strains (sensitive:  $MIC_s = 16 \ \mu g/ml$ ; resistant:  $MIC_R =$ 2048  $\mu g/ml$ ; [13]). Cultures were inoculated with a 1:1 volumetric mixture of overnight cultures of the sensitive (DsRed) and resistant (YFP) strains, at total density similar to a standard MIC test. After 24h, cultures were sampled by flow

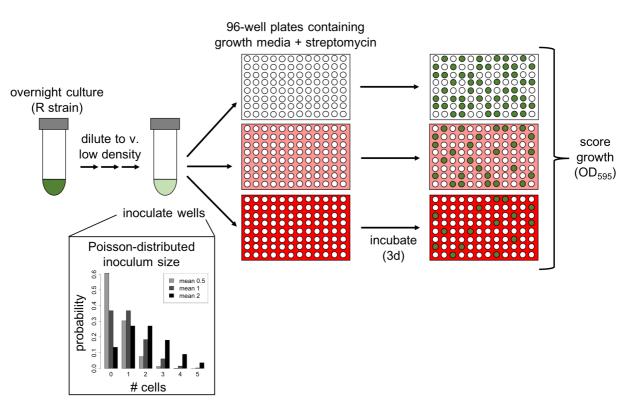
124 cytometry to estimate the proportion of each strain (**Methods**). The horizontal dashed line at 0.5 indicates the starting

# Stochastic bacterial population dynamics

125	proportion, which would be maintained if resistance were selectively neutral. Final proportions falling below this line
126	indicate selection favouring the sensitive strain, while those above this line indicate selection favouring the resistant strain.
127	The vertical dashed line indicates the approximate position of the minimum selective concentration (MSC). The inset shows
128	a zoomed-in version at low streptomycin concentration, on a log scale. Points (shown only in the inset for lower
129 130	concentrations) represent six biologically independent replicates at each concentration, with line segments indicating their mean (see also <b>Suppl. Table 1</b> ). Asterisks indicate that the mean final proportion of the resistant strain significantly differs
130	from 0.5 in a two-sided t-test at each streptomycin concentration, with a Bonferroni correction for multiple testing (n.s.: p
132	> 0.05/7; *: $p = 4e-3$ ; **: $p \le 5e-6$ ; ***: lack of variation among replicates precludes a t-test).
133	
134	
135	Outgrowth of resistance from single cells is significantly inhibited within the MSW
136	
137	Although a wide range of streptomycin concentrations select for already-established resistant
138	strains, it does not follow that streptomycin resistance will emerge <i>de novo</i> at these concentrations.
139	To test how antibiotic exposure affects establishment of the resistant strain, which we define as the
140	growth of a large population starting from a single cell, we conducted a large-scale "seeding
141	experiment" (see also [14]). In this experiment (Fig. 2), we inoculated a large number of cultures at
142	very low densities, averaging <1 to around 3 cells per culture. Importantly, the number of cells
143	inoculated into each culture is random, and can be described by a Poisson distribution (Suppl. Fig.
144	<b>3</b> ). One implication of this protocol is that many cultures are not inoculated with any cells; our
145	approach is to account for this variation statistically. We conducted seeding experiments at three
146	different inoculation densities and across a range of streptomycin concentrations (up to $1/8 \times MIC_R$ ).
147	We then scored the number of cultures showing growth based on reaching a threshold optical
148	density up to 3 days post-inoculation.
149	
150	Cell division and death in individual bacterial cells are probabilistic events, with the consequence
151	that a single cell will not necessarily give rise to a successful lineage of cells. This chance of stochastic
152	loss applies even under benign culture conditions, but is expected to be exacerbated by stressors
153	such as antibiotic exposure that reduce the probability of successful cell division compared to cell
154	death. In the context of our experiment, a culture could fail to grow either because the inoculum did
155	not contain any cells, or because cells in the inoculum failed to give rise to a surviving lineage. To
156	discriminate between these possibilities, we fit a stochastic model to infer the per-cell probability of
157	establishment (i.e. probability that a single cell produces a large number of surviving descendants
158	leading to detectable growth) from the number of replicate cultures showing growth, across
159	streptomycin concentrations ( <b>Methods</b> ). We cannot estimate an <u>absolute</u> establishment probability
160	because this quantity cannot be separated from the effect of inoculum size: inoculating on average

6

- 161 two cells, each with 50% chance of establishment, will give equivalent results to one cell with 100%
- 162 chance of establishment. Therefore, all results are expressed in terms of <u>relative</u> per-cell
- 163 establishment probability, denoted  $\tilde{p}_c$ , which is normalized by the result in streptomycin-free
- 164 culture medium. Thus,  $\tilde{p}_c$ =1 by definition in streptomycin-free conditions, while we expect
- 165 streptomycin treatment to yield  $\tilde{p}_c \leq 1$ ; however, values larger than one can arise due to sampling
- 166 error. Since the distribution of inoculum size is the same in streptomycin-treated and streptomycin-
- 167 free cultures, any significant differences in estimated  $\tilde{p}_c$  can be attributed to effects of streptomycin
- 168 on the dynamics of resistant cells.
- 169
- 170 Exposure to very low concentrations of streptomycin (up to 1/32 x MIC<sub>R</sub>) had no detectable impact
- 171 on the probability of establishment of a single resistant cell, but this probability declined sharply as
- 172 concentration increased (**Fig. 3** and **Suppl. Table 2**). Strikingly,  $\tilde{p}_c$  is already significantly reduced by
- 173 1/16 x MIC<sub>R</sub> streptomycin and is only marginally above zero (3-5%) at 1/8 x MIC<sub>R</sub>. Thus, the chance of
- de novo establishment of the resistant strain is nearly abolished under antibiotic concentrations well
- 175 below its MIC, where resistance is strongly favoured in competition with the sensitive strain
- 176 (compare Fig. 3 to Fig. 1).
- 177



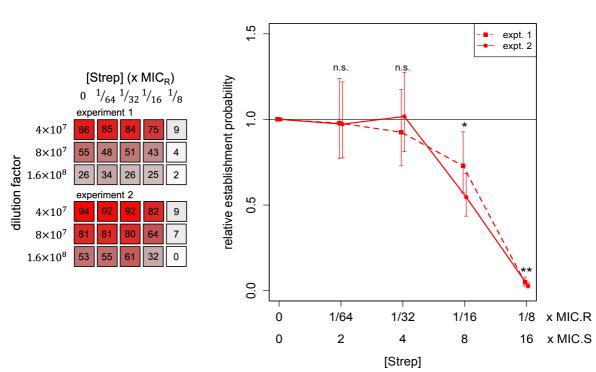


- 180 is highly diluted and used to inoculate 96-well plates containing growth media (LB) with streptomycin at various
- 181 concentrations (higher concentration indicated by darker red). The number of cells inoculated per well follows a Poisson

# Stochastic bacterial population dynamics

7

- 182 distribution (examples illustrated in the inset plot for mean inoculum size of 0.5, 1, or 2 cells per well). Plates are incubated
- 183 up to 3d and optical density is measured to score growth in wells (OD<sub>595</sub> > 0.1; indicated by dark green colouring). The final
- 184 number of replicate cultures showing growth at each inoculum size and streptomycin concentration is used to estimate the
- 185 per-cell establishment probability by fitting a mathematical model (Methods).
- 186
- 187
- 188 (a) (b)



- 190 Figure 3: Establishment probability of single resistant cells, estimated from seeding experiments. (a) A visual 191 representation of the growth data, indicating the number of replicate cultures (out of 96) that grew in each test condition. 192 (b) The estimated relative per-cell establishment probability  $\tilde{p}_c$  of the resistant strain, scaled by the probability in 193 streptomycin-free medium, plotted as a function of streptomycin concentration, scaled by the MICs of the resistant and 194 sensitive strains (MIC<sub>R</sub> and MIC<sub>S</sub>, respectively). Results are shown for two separate experiments. Plotted points indicate the 195 maximum likelihood estimate of  $\tilde{p}_c$  and error bars indicate the 95% confidence interval, using the fitted model selected by 196 the likelihood ratio test (Model B' in experiment 1 and Model C' in experiment 2, both of which pool data across three 197 inoculation densities; see Suppl. Text). Significance of the streptomycin effect is determined by fitting a generalized linear 198 model to the population growth data (n.s.: not significant, p > 0.05; \* p = 0.01 in expt. 1 and 2e-7 in expt. 2; \*\* p < 2e-16 in 199 both experiments). 200 201 202 MIC depends on inoculum size
- 203

204 We hypothesized that the resistant strain's low probability of establishment even far below its MIC

205 could be explained by an inoculum size effect. Specifically, MIC is standardly measured with an

### Stochastic bacterial population dynamics

8

206 inoculation density of 5x10<sup>5</sup> CFU/mL [15], which corresponds to 10<sup>5</sup> CFU inoculated per well in our 207 experimental set-up. In contrast, in the seeding experiments above, we inoculated on the order of 1 208 CFU per well. MIC for many antibiotics is known to increase with inoculation densities higher than 209 standard [16] [17] [18]. Although less well-explored, it has also occasionally been noted that MIC can 210 decrease when lower inoculum sizes are used [19] [20]. 211 212 To test the hypothesis that MIC is inoculum size-dependent, we conducted MIC assays with 213 inoculation densities ranging over three orders of magnitude, from approximately 5x10<sup>2</sup> CFU/ml up 214 to the standard 5x10<sup>5</sup> CFU/ml (i.e. 10<sup>2</sup>-10<sup>5</sup> CFU/well). We found that MIC indeed increased with 215 inoculum size (Fig. 4). This pattern is apparent regardless of whether growth is scored at 20h, as per 216 the standard MIC assay protocol [15], or up to 3d post-inoculation, as in our seeding experiments, 217 although the number of cultures showing detectable growth and thus the observed MIC tends to 218 increase over time (Suppl. Fig. 4). 219 220 The observed increase in MIC with inoculum size could have been due either to the increase in 221 absolute number of individuals in the inoculum (i.e. CFU), or in their density (i.e. CFU per unit 222 volume). These two possibilities are not typically distinguished in the literature; however, they lead 223 to distinct interpretations. If demographic stochasticity is the dominant force, we expect absolute 224 numbers to matter, whereas if ecological effects (competition or cooperation) affect establishment,

cell density per unit volume could be more important. To disentangle these two factors, we

226 repeated the MIC assay co-varying inoculation density and culture volume. This experiment

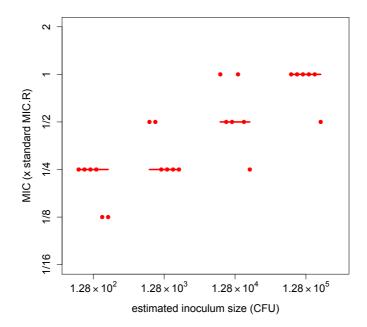
227 confirmed that absolute inoculum size has a strong effect on MIC, but revealed that when

228 controlling for absolute size, inoculation density per unit volume does not have a significant effect

on the MIC, within the range that we tested (**Suppl. Fig. 5**).

#### Stochastic bacterial population dynamics

9



### 231

Figure 4: MIC as a function of inoculum size. Test cultures at two-fold concentration steps of streptomycin were inoculated at four different densities. MIC was evaluated as the minimal tested concentration that prevented detectable growth up to 3d post-inoculation; a qualitatively similar pattern arose if growth was evaluated at 20h (Suppl. Fig. 4). The yaxis is scaled by the previously measured MIC of the resistant strain at standard inoculation density (MIC<sub>R</sub>), as before. The points represent six biologically independent replicates at each inoculum size, with the line segments indicating their median.

238

239

# 240 Population growth can be explained by an independent chance of each cell to establish

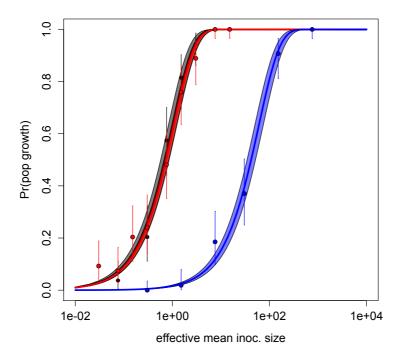
241

242 The above MIC tests revealed that the absolute number of cells in the inoculum has a strong effect 243 on whether the population eventually shows detectable growth at streptomycin concentrations 244 within the MSW. The simplest explanation for this effect is that at a given streptomycin 245 concentration, each inoculated cell has a certain probability of establishing a surviving lineage 246 (leading to detectable population growth), independently of all other cells in the inoculum. This 247 independence assumption leads to a simple stochastic model capturing the effect of inoculum size 248 on the probability of population growth (Methods, Eqn. 2). Note that this independent model of 249 cellular replication need not be true; for example, if cells secrete an enzyme that breaks down an 250 antibiotic extracellularly, then the chance that an individual cell establishes a successful lineage 251 could increase with initial cell density due to cooperative resistance. On the other hand, if cells 252 strongly compete for limiting resources or secrete toxins, the per-cell chance of establishment could 253 decrease with initial cell density.

10

255 To formally test the null hypothesis that each inoculated cell has an independent probability of 256 establishment, we again conducted high-replicate population growth experiments. These were 257 similar to the seeding experiments, but used multiple inoculum sizes spanning approximately three 258 orders of magnitude, in order to capture the full range of response (i.e. 0-100% of cultures showing 259 growth) at a given streptomycin concentration. We tested two concentrations (1/16 and  $1/8 \times MIC_{R}$ ) 260 for which growth often failed in the seeding experiment, but succeeded at standard inoculation 261 density in the MIC tests. In parallel, we tested growth in streptomycin-free media in order to 262 calibrate the "effective" mean inoculum size (Methods and Suppl. Fig. 6). 263

- 264 We found good agreement between the null model and our experimental data at all tested
- 265 streptomycin concentrations, as we would expect if establishment were independent (main
- 266 experiment, **Fig. 5**, and repeat experiments, **Suppl. Fig. 7**). More precisely, in no case did the null
- 267 model show significant deviance from the observed proportion of populations that grew (according
- to the likelihood ratio test; see **Suppl. Text** for full results), and thus we accept it as the most
- 269 parsimonious explanation for the data. At 1/16 and  $1/8 \times MIC_R$  we obtain estimates of relative
- establishment probability,  $\tilde{p}_c$ , comparable to those from the seeding experiments (**Suppl. Table 2**).
- 271





273 Figure 5: Null model of the inoculum size effect fit to population growth data. Probability of population growth of the

- resistant strain is plotted as a function of effective mean inoculum size ( $\overline{N}_{eff}$ , calibrated by the results in streptomycin-free
- 275 media; see **Suppl. Fig. 6**). Black: streptomycin-free; red: streptomycin at 1/16 x MIC<sub>R</sub>; blue: 1/8 x MIC<sub>R</sub>. The solid line shows
- 276 the best fit of the null model (using the maximum likelihood estimate [MLE] of  $\tilde{p}_c$ ) and the shaded area corresponds to the
- 277 95% confidence interval (Cl). Points and error bars indicate the MLEs and 95% CIs in the full model, i.e. treating each

### Stochastic bacterial population dynamics

11

- inoculum size separately (here MLE simply equals the proportion of experimental replicate cultures showing growth).
- According to the likelihood ratio test, the null model deviance from the full model is not significant at any streptomycin concentration (streptomycin-free: p=0.55;  $1/16 \times MIC_R$ : p=0.28;  $1/8 \times MIC_R$ : p=0.71).
- 281
- 282

# 283 Sub-MIC streptomycin concentrations extend lag phase and induce cell death

284

285 To explain the reduced establishment probability of the resistant strain when exposed to sub- $MIC_{R}$ 286 concentrations of streptomycin, we hypothesized that even these low doses must compromise cell 287 division rate and/or viability. As a simple test of this idea, we measured the relative abundance of 288 dead cells in cultures of the resistant strain grown under various concentrations of streptomycin. We 289 found that the fraction of dead cells after 7h of treatment, as determined by propidium iodide 290 staining, increases from an average of 3-4% in streptomycin-free conditions to >20% at  $1/8 \times MIC_R$ 291 streptomycin (Fig. 6a, Suppl. Fig. 8 and Suppl. Table 3). Note that this is a conservative measure of 292 cell death, because this assay only detects cells that have compromised membrane permeability,

- and does not detect, for example, lytic cell death.
- 294

To gain further insight into how the toxic effects of streptomycin impact the population dynamics of the resistant strain, we quantified viable cell density (colony-forming units in plated samples) over the first few hours of sub-MIC<sub>R</sub> streptomycin treatment. In this experiment, cultures were inoculated with approximately 100 CFU/well, such that populations are small enough for demographic stochasticity to be relevant, yet large enough to be detectable by plating samples.

300

301 We found that streptomycin treatment has a significant effect on population growth (ANOVA, main 302 effect: p < 2e-16), and this effect varies over time (ANOVA, interaction term: p < 2e-16). Following 303 inoculation into fresh medium, the resistant strain experienced a lag phase of approximately 2 304 hours. Control cultures that were exposed to streptomycin-free medium then exited lag phase and 305 began to grow exponentially. The lowest tested dose of streptomycin  $(1/32 \times MIC_R)$  had no 306 significant effect on these dynamics (Dunnett's test: p = 0.87); however,  $1/16 \times MIC_{R}$  was already 307 sufficient to slow population growth (p = 4e-4). Nonetheless, all replicate populations (48/48 at each 308 concentration, pooled across all sampled treatment plates) eventually grew, as detected by OD, at 309 streptomycin concentrations up to  $1/16 \times MIC_{R}$ . Meanwhile, higher doses of streptomycin had 310 dramatic effects (1/8 x or 1/4 x MIC<sub>R</sub> vs. control: p < 1e-4), with cultures exhibiting an extended lag 311 phase of at least 7-8 hours, in which viable cell density initially declined. After further incubation (up 312 to 3 days), 25% of cultures (15/60) exposed to  $1/8 \times MIC_R$  eventually showed growth to saturation,

### Stochastic bacterial population dynamics

12

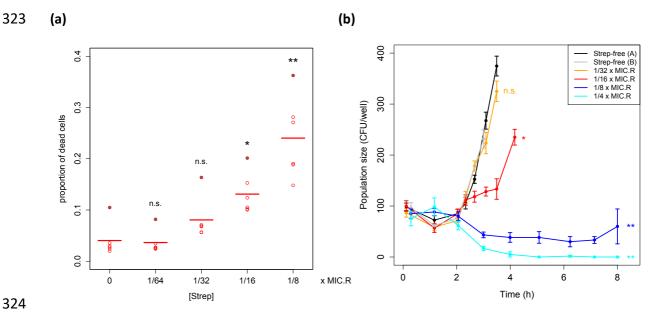
while the remaining 75% of cultures (45/60) failed to reach detectable OD. At  $1/4 \times MIC_{R}$ , no viable

cells were sampled in most replicates from 4h on, and only 1/60 replicates reached detectable OD

- within 3 days.
- 316

In summary, these population dynamics experiments reveal that sub-MIC streptomycin treatment
has the effect of extending the lag phase, before populations eventually either grow to saturation or
die out. The failure of many populations to establish at sub-MIC<sub>R</sub> concentrations can be explained by
a significant fraction of cells dying rather than successfully dividing, which can lead to stochastic loss
of initially small populations.

322



325 Figure 6: Sub-MIC<sub>R</sub> streptomycin treatment has profound effects on resistant cell dynamics. (a) Proportion of dead cells 326 after 7h treatment in sub-MIC<sub>R</sub> streptomycin. The proportion of dead cells was estimated using live-dead staining and 327 flow cytometry (Methods). Note that this is a minimal estimate of the proportion of dead cells, since propidium iodide 328 staining only detects cells with compromised membranes. The points represent six treatment replicates at each 329 concentration and the line segments indicate their mean. Differences from the streptomycin-free control cultures were 330 assessed using a one-way ANOVA followed by a post-hoc Dunnett's test (n.s.: not significant, p > 0.05; \*: p = 9e-3, \*\*: p < 331 1e-4). The significant effects do not change if we exclude an outlier replicate (shaded-in points) showing consistently 332 elevated dead fractions (Suppl. Table 3). (b) Viable cell population dynamics in sub-MIC<sub>R</sub> streptomycin. The plotted points 333 with connecting lines indicate mean population size of six independent replicate cultures per streptomycin concentration 334 and per sampling time point (or twelve replicates for each streptomycin-free control; see Methods); the error bars indicate 335 standard error. See Suppl. Fig. 9 for a version showing all individual replicates. Population size was estimated by plating 336 and truncated when plated colonies became too dense to count. Significance of each streptomycin concentration 337 compared to the streptomycin-free control was assessed by a post-hoc Dunnett's test (n.s.: not significant, p=0.87; \* p=338 4e-4; \*\* p < 1e-4).

- 339
- 340

Stochastic bacterial population dynamics

13

# 341 **Discussion**

342

343 In order for a novel resistant strain to emerge, it must successfully expand from a single cell to form 344 a large population of cells. At the level of individual cells, division and death occur as probabilistic 345 events, and demographic stochasticity should therefore be critical to the fate of newly arisen 346 resistant strains. Our key finding is that these stochastic effects impose a significant barrier to the 347 emergence of antibiotic resistance. This result was highlighted by our seeding experiment, in which 348 cultures were inoculated with approximately one resistant cell at streptomycin concentrations well 349 below the MIC of the resistant strain. Strikingly, the chance of population growth was significantly 350 reduced by streptomycin concentrations as low as 1/16 x MIC<sub>R</sub>, and nearly abolished (<5% chance) 351 by  $1/8 \times MIC_R$  (Fig. 3). This result can be explained by detrimental effects of sub-MIC antibiotic 352 exposure, including a significant mortality rate on the individual cell level (Fig. 6), leading to frequent 353 stochastic extinction of small populations even at concentrations where the expected net population 354 growth rate is positive (albeit slowed). One way to quantify the importance of stochasticity is to 355 measure the difference between the range of antibiotic concentrations at which resistance is 356 selectively favoured and the range at which resistance is likely to establish when rare. In this case, 357 we estimate that resistance is selectively favoured, and will on average grow, between 1/512 x and 1 358 x MIC<sub>R</sub>, or 1/4 x and 128 x MIC<sub>s</sub> (Fig. 1 and Fig. 4); in contrast, resistance is very unlikely to emerge 359 from a single cell at concentrations  $\geq 1/8 \times MIC_{R}$  (Fig. 3). Therefore, stochastic effects impose a 360 major constraint on the emergence of resistance across nearly 90% of the mutant selection window. 361 362 It is possible that survival of resistant lineages, while appearing stochastic at the macro-scale, is 363 governed by intrinsic differences among cells in their phenotypic level of resistance [21] [22]. Our

364 model is still valid in case of such phenotypic heterogeneity, but the per-cell establishment

probability must then be interpreted as an average among cells (**Suppl. Text**). Further investigation

- of the cell-level factors that influence the fate of lineages represents an interesting avenue for futurework.
- 368

Standardized MIC assays are a widely used and crucial tool for repeatable evaluation of the
susceptibility of bacterial strains to antibiotics [15]. However, the discrepancy between the antibiotic
concentration required to reliably prevent growth of a larger population (as in a standard MIC assay)
and the concentration that is sufficient to drastically reduce the probability of growth of a smaller
population, highlights that MIC depends on inoculum size (Fig. 4). To clarify this effect, we rigorously
quantified the probability of population growth from inoculum sizes spanning three orders of

### Stochastic bacterial population dynamics

14

375 magnitude, in the presence and absence of streptomycin. We found that a simple null model, in 376 which each individual cell independently has some antibiotic concentration-dependent probability of 377 establishing a surviving lineage, is sufficient to explain our data (Fig. 5). In other words, we need not 378 invoke any interactions among cells (e.g. competition or cooperation) to explain the low probability 379 of emergence of resistance when starting from small inoculum sizes. Moreover, this simple 380 stochastic "numbers game" provides a unifying explanation for all of the preceding results. When only a single cell is inoculated, as in the seeding experiments, outgrowth is very unlikely at as little as 381 382  $1/8^{\text{th}}$  of this strain's standard MIC. However, the cumulative effect of many individuals, each with a 383 small chance of establishment, virtually guarantees population growth from a sufficiently large 384 inoculum size, as in standard MIC assays.

385

386 The "inoculum effect" has previously been recognized primarily in reference to the reduced efficacy 387 of antibiotic treatment (i.e. increase in MIC) as bacterial population density increases beyond that 388 used in standard MIC assays [16] [17]. This effect can arise via various mechanisms, including 389 titration of antibiotics out of the media [17] and enzymatic inactivation of antibiotics by resistant 390 cells [23] [24] [18] [25] [13]. In contrast to this density-dependent, cooperative regime at high 391 bacterial density, the inoculum effect that we found at very low bacterial population size is a 392 stochastic phenomenon driven by cells acting effectively independently. There are hints of this low 393 inoculum size effect in earlier literature [19] [20], but to our knowledge we are the first to provide a 394 rigorous explanation in terms of stochastic population dynamics. These two types of inoculum effect 395 are distinct: the cooperative growth regime arises for certain antibiotics at high bacterial 396 <u>concentrations</u> (cells/ml), whereas we expect the stochastic growth regime to arise very generally at 397 low absolute numbers of cells. That is, we expect the functional relationship between probability of 398 population growth and absolute inoculum size (Eqn. 2) to hold widely (with the scaling parameter  $\tilde{p}_c$ 399 depending on the bacterial strain and antibiotic); however, the relationship between MIC and 400 inoculum size will also depend on the relationship between  $\tilde{p}_c$  and antibiotic concentration. 401 Moreover, measured MIC will become increasingly variable at low inoculum size, calling for a more 402 precise measure of inhibition, such as the IC<sub>99</sub> [26]. In these two facets of the inoculum effect, cells 403 at the limit of low density (cells/ml) where cooperative interactions are abolished [18] are 404 distinguished from low absolute numbers of cells, where stochastic effects become relevant. Both 405 types of inoculum effect raise important considerations for antibiotic dosing. Cooperative growth 406 has implications for treatment of high-density bacterial infections [24] [27] [28], whereas the 407 stochastic effects we studied here could have implications for dosing strategies to avoid emergence 408 of de novo resistant mutants.

### Stochastic bacterial population dynamics

15

409

410 Although the role of demographic stochasticity in the fate of *de novo* mutations has long been 411 recognized in theoretical population genetics, until very recently it had never been addressed 412 empirically [12]. Our study joins a small handful of others that have now experimentally quantified 413 establishment probability from single cells [29] [26] [30] [31]. Two of these studies [26] [31] likewise 414 addressed establishment of bacterial cells across antibiotic concentrations, but using different 415 methods to ours (see Suppl. Text for a more detailed comparison). Bacterial evolution of resistance 416 to antibiotic treatment is also a prime example of the more general phenomenon of evolutionary 417 rescue, whereby adaptation prevents extinction of populations facing severe environmental change 418 [32]. Indeed, our equation for the probability of population growth (Eqn. 2) is identical in form to 419 general approximations for the probability of rescue [33] (see **Suppl. Text** for further discussion). 420 More broadly, the concepts and methods developed here could be applied to a variety of situations 421 where growth depends on establishment of rare cells and is thus highly stochastic, for instance the 422 onset of invasive bacterial infections [34] or the outgrowth of bacteria in food products from small 423 initial contaminants [35].

424

425 In summary, our study highlights the stochastic nature of *de novo* emergence of antibiotic 426 resistance, and sheds new light on the question of what concentrations of antibiotics drive 427 resistance evolution. Importantly, we found that concentrations within the traditional mutant 428 selection window (MSW), above the MIC of the sensitive strain (MICs) but still well below the MIC of 429 the resistant strain ( $MIC_R$ ), can be sufficient to severely limit *de novo* establishment of resistance 430 from single cells. Thus, our findings suggest that moderate doses (between  $MIC_s$  and  $MIC_R$ ) may be 431 more effective than previously thought at preventing emergence of resistance, in cases where the 432 chance of transmitted resistance is negligible. Taken together with our own and previous evidence 433 of sub-MICs selection for resistance [7] [8] [9] [10] [13], our work contributes to an emerging picture 434 that resistance tends to arise at a range of concentrations shifted lower than the traditional MSW. 435 Importantly, however, we studied establishment of the resistant strain in isolation; it remains to be 436 determined how interactions with an initially dominant sensitive population shape establishment of 437 resistance during antibiotic treatment. A key direction for future work will thus be to integrate the 438 dual effects of competition and direct inhibition by the antibiotic, which operate with varying 439 strengths across antibiotic concentrations [6]. By better understanding these limiting forces on the 440 evolution of resistance, we may ultimately be able to design effective dosing strategies that avoid 441 emergence of resistance.

Stochastic bacterial population dynamics

### 16

# 443 Methods

444

### 445 Bacterial strains, media and culture conditions

446

447 Bacterial strains: We used a set of *Pseudomonas aeruginosa* PA01 strains described previously [13]. 448 These include both streptomycin-sensitive and -resistant strains, which are isogenic except that the 449 resistant strains carry the clinically derived, non-conjugative plasmid Rms149 [36]. Streptomycin 450 resistance is conferred by the *aadA5* gene, which is located within a class 1 integron, and codes for 451 an enzyme that adenylates streptomycin [37]. For both plasmid carriers (resistant) and non-carriers 452 (sensitive), we have strains with a chromosomal YFP label, a DsRed label, or unlabelled. The 453 competition experiment was conducted with the DsRed-labelled sensitive and YFP-labelled resistant 454 strains. The live-dead staining experiment was conducted with the unlabelled resistant strain. All 455 other experiments were conducted with the YFP-labelled resistant strain. The minimum inhibitory 456 concentrations of streptomycin in LB broth were previously determined to be  $MIC_s = 16 \mu g/ml$  for 457 the sensitive strain and  $MIC_{R} = 2048 \mu g/ml$  for the resistant strain [13].

458

459 Media and antibiotics: For culture media we used LB broth containing 5g/L NaCl (Sigma-Aldrich, 460 product no. L3022). For plating to assess colony-forming units we used LB Agar, Vegitone, containing 461 5g/L NaCl and 15g/L agar (Sigma-Aldrich product no. 19344). Streptomycin was prepared as stock 462 from streptomycin sulfate salt (Sigma-Aldrich product no. S6501), stored according to supplier 463 directions, and added to media on the day of experiments. Bacterial cultures were diluted in 464 phosphate buffered saline (PBS) prepared from tablets (Sigma-Aldrich product no. P4417). Note that 465 treatment cultures were generally set up with 90% media plus 10% inoculating culture by volume; 466 thus, the final concentrations of LB and streptomycin in the treatments are 90% of the prepared 467 media values denoted on plots. 468

469 <u>Culture conditions</u>: All cultures were incubated at 37°C, shaking at 225rpm. Overnight cultures were
470 inoculated directly from freezer stocks into 2ml of LB and then incubated for approximately 16h.
471 Overnight cultures were then diluted in PBS and used to inoculate treatment plates. All experimental
472 treatments were conducted in 200µl cultures in flat-bottom 96-well microtitre plates unless
473 otherwise noted.

474

475 Competition experiment

### Stochastic bacterial population dynamics

17

477 Competitions were carried out between the YFP-labelled resistant strain and the DsRed-labelled 478 sensitive strain, in streptomycin-free media and streptomycin at 1/16 to  $16 \times MIC_s$  in 2-fold 479 concentration steps. Previous work demonstrated that these two markers have equivalent effects on 480 bacterial fitness on LB-agar [13]. Treatment cultures were inoculated with a 1:1 volumetric mixture 481 of overnight cultures of the two strains, with expected total inoculum size around 10<sup>5</sup> CFU. We 482 tested six replicate mixed cultures along with two of each pure culture as controls. After 24h, 483 treatment cultures were diluted 500-fold and 66µl of each diluted culture was sampled by flow 484 cytometry (BD Accuri C6 Flow Cytometer, fast fluidics, discarding events with forward scatter FSC-H 485 < 10,000 or side scatter SSC-H < 8000). By gating based on fluorescence and correcting for the 486 overlap of each strain into the opposing gate using the pure culture controls (see Suppl. Text and 487 Suppl. Fig. 1), we estimated the final frequency of each strain. 488 489 Seeding experiment 490

491 <u>Experimental protocol</u>: An overnight culture was serially diluted up to 10<sup>6</sup>-fold, then in three further
492 independent steps, diluted to 4 x 10<sup>7</sup>-, 8 x 10<sup>7</sup>, and 1.6 x 10<sup>8</sup>-fold. Each of the latter diluted cultures
493 was used to inoculate an entire 96-well test plate for each tested streptomycin concentration, at
494 20µl per well. Test plates were incubated and scored for growth after approximately 1, 2, and 3d.
495

496 <u>Scoring culture growth</u>: We evaluated culture growth by measuring optical density (OD<sub>595</sub>) using a
497 BioTek Synergy 2 plate reader. Lids on 96-well plates were briefly removed for the reading;
498 comparison to controls mock-inoculated with PBS indicated that contamination was rare. We set a
499 threshold OD of 0.1 to score as growth, as background OD in controls was typically <0.05. OD</li>
500 showed a clearly bimodal distribution after a sufficiently long growth period; thus, the precise choice
501 of threshold is unimportant.

502

503 <u>Estimating establishment probability</u>: From the number of replicate cultures showing growth by 3d 504 at each condition, we obtained a likelihood-based point estimate and confidence interval for the 505 single-cell establishment probability, relative to streptomycin-free controls, by fitting a 506 mathematical model to the data (see section "Mathematical model of establishment" below).

507

508 <u>Generalized linear model</u>: To evaluate the effect of streptomycin on establishment, culture growth 509 data was additionally fit with a generalized linear model using the built-in R function 'glm'. Number 510 of replicate cultures showing growth was treated as binomial data, with inoculating dilution factor

# Stochastic bacterial population dynamics

511	and streptomycin concentration taken as explanatory variables, using the complementary log-log
512	link function (see <b>Suppl. Text</b> for details).
513	
514	MIC assay with varying inoculation density
515	
516	Test cultures were inoculated with 20 $\mu$ l of overnight culture diluted 10 <sup>3</sup> -, 10 <sup>4</sup> -, 10 <sup>5</sup> -, or 10 <sup>6</sup> -fold.
517	These dilution factors were chosen such that the highest inoculation density was expected to be
518	similar to a standard MIC test (5 x 10 $^{5}$ CFU/ml; [15]); actual density was estimated by plating out
519	diluted overnight culture on LB-agar and counting colony-forming units. We tested six replicates per
520	inoculation density and per streptomycin concentration, ranging from 1/16 x MIC $_{\!R}$ to 2 x MIC $_{\!R}$ in
521	two-fold steps, plus streptomycin-free controls. Culture growth was evaluated by OD at
522	approximately 20h, 2d, and 3d, according to the same procedure as for the seeding experiments. In
523	rare cases, MIC within a replicate was ambiguous due to growth failing in a culture at lower
524	streptomycin concentration but succeeding at the next highest step before being definitively
525	abolished. In these cases, we took MIC to be the higher concentration, at and beyond which no
526	further growth was observed.
527	
528	Testing the null model of the inoculum size effect
529	
530	To test the relationship between inoculum size and probability of population growth, we evaluated
531	growth at selected streptomycin concentrations using many different dilution factors for
532	inoculation: five in streptomycin-free conditions and 6-10 in each streptomycin concentration. These
533	dilution factors were chosen for each streptomycin concentration based on the results of previous
534	experiments, with the aim of capturing the range over which the proportion of replicate cultures
535	showing growth increases from near 0 to near 1. In each case, 54 replicate cultures were evaluated
536	per inoculum size. Culture growth was evaluated daily by OD, according to the same procedure as
537	for the seeding experiments, up to 5d. The final number of replicate cultures showing growth was
538	used for model fitting (see section "Mathematical model of establishment" below).
539	
540	Fraction of dead cells by live-dead staining
541	
542	For this experiment, we used the resistant strain with no fluorescent label, to avoid interfering with
	• • • • • • • • • • • • • • • • • • • •

- 543 the signal from the live-dead stain. We inoculated treatment cultures (six replicates per
- 544 streptomycin concentration) with 10<sup>3</sup>-fold diluted overnight culture, as in the MIC assay at standard

### Stochastic bacterial population dynamics

19

545 inoculation density. After 7h of treatment, we diluted test cultures a total of 100-fold and stained 546 with thiazole orange and propidium iodide (BD Cell Viability Kit, product no. 349483). In parallel, we 547 diluted and stained media and heat-killed cultures as controls. We sampled 50µl per diluted culture 548 using flow cytometry (with the same settings as in the competition experiment). The staining and 549 flow cytometry steps were carried out in groups containing one replicate per concentration plus 550 controls, to avoid potentially toxic effects of stain exposure over prolonged times (Suppl. Text). To 551 better discriminate cells from background in the flow cytometry data, we first gated on events 552 according to forward and side scatter before defining clusters of dead (membrane-compromised) 553 and intact cells based on fluorescence; see Suppl. Text and Suppl. Fig. 8 for details.

554

### 555 Viable cell density dynamics

556

557 We tracked the number of viable cells over time in streptomycin-free media and at 1/32, 1/16, 1/8, 558 and  $1/4 \times MIC_{R}$  streptomycin. An independent test plate was used for sampling at each time point. 559 Six replicate cultures per concentration, per test plate (twelve replicates for streptomycin-free) were 560 inoculated with  $20\mu$ l of 5 x  $10^5$ -fold diluted overnight culture. Streptomycin treatments were split 561 across two sets of plates, each of which included a streptomycin-free control; set A additionally 562 included the two lower streptomycin concentrations and set B included the two higher. These sets 563 were sampled at generally different time points. At each sampling time, undiluted cultures were 564 plated in arrays of 4µl spots, repeated across 5 square LB-agar plates for a total sampled volume of 565 20µl per culture. LB-agar plates were incubated at 37°C for the rest of the day of the experiment, 566 moved to the bench overnight to prevent overgrowth, then incubated again the next day at 37°C 567 until colonies were visible, but still separated, for counting. Population size in each replicate culture 568 was then estimated by scaling up total CFU count by a factor 10 (since 1/10<sup>th</sup> of the culture volume 569 was sampled). Comparing streptomycin-free controls from sets A and B indicated that the plate set 570 effect was non-significant (ANOVA: p = 0.103); thus, controls were pooled for further analysis of the 571 streptomycin effect.

572

## 573 Mathematical model of establishment

574

575 <u>Model</u>: We used a simple stochastic model to describe population (i.e. bacterial culture) growth as a

576 function of inoculum size and streptomycin concentration. We denote by  $p_w$  the probability that a

577 population inoculated with a small number of cells grows to a large size (detected by OD as

578 described above). Among a set of *n* independent replicate populations (inoculated at the same

20

579	density and cultured in the same conditions), the number of populations showing growth is thus
580	described by a Binomial $(n,p_w)$ distribution. Under the assumptions that (i) the number of cells in the
581	inoculum that establish surviving lineages is Poisson-distributed with some mean $lpha$ (an assumption
582	supported by the observed distribution of colony-forming units counted in highly diluted cultures;
583	Suppl. Fig. 3), and (ii) population growth is observed provided at least one cell establishes, we have
584	the relationship:
585	
586	$p_w = 1 - e^{-\alpha} \tag{Eqn. 1}$
587	
588	Relative per-cell establishment probability $\widetilde{p}_c(x)$ is generally calculated as the ratio of the mean
589	number of established cells at any given streptomycin concentration x to that in the absence of
590	streptomycin: $\tilde{p}_c(x) = \alpha(x)/\alpha(0)$ .
591	
592	In the full model (statistically speaking), $p_{\scriptscriptstyle W}$ , or equivalently $lpha,$ is taken to be distinct for each
593	inoculum size and each streptomycin concentration. Simpler nested models make additional
594	assumptions (see Suppl. Text for details). In particular, the null model of the inoculum size effect
595	supposes that cells behave independently of one another, such that the true per-cell probability of
596	establishment, $p_c$ , depends only on streptomycin concentration. Then the mean number of
597	established cells, $lpha$ , is directly proportional to the mean inoculum size, $\overline{N}$ (where the actual
598	inoculum size is Poisson-distributed about this mean). In this null model, the probability of
599	population growth can be expressed as:
600	
601	$p_w = 1 - e^{-\overline{N}p_c}$
602	$= 1 - e^{-\overline{N}_{eff}\tilde{p}_c} $ (Eqn. 2)
603	
604	where $\overline{N}_{eff} = \alpha(0) = \overline{N}p_c(0)$ denotes the "effective mean inoculum size", defined as the mean
605	number of established cells in the absence of streptomycin; and the relative establishment
606	probability at streptomycin concentration x can simply be expressed as $\tilde{p}_c = p_c(x)/p_c(0)$ . Note
607	that by scaling up $\overline{N}_{eff}$ by the dilution factor(s) applied to the inoculating culture, we have an
608	estimate of bacterial density in this culture, determined equivalently to the historical "most
609	probable number" method [38] [39]. This definition of effective inoculum size based on cells that

610 grow in benign conditions is no different in principle to the usual quantification of "viable" cells

611 according to successful formation of a colony; we simply assess growth in liquid rather than on solid

# Stochastic bacterial population dynamics

21

612	medium. We also note that if cells (or possibly clumps of cells acting as individual units) are
613	heterogeneous, then $p_c$ should be interpreted as the mean establishment probability (Suppl. Text).
614	
615	Likelihood-based model fitting and comparisons: These stochastic models were fit to experimental
616	population growth data using likelihood-based methods. Specifically, we obtained a maximum
617	likelihood estimate and a 95% confidence interval (determined by the range of parameter values
618	that would not be rejected by a likelihood ratio test at 5% significance level) on the parameter $p_w$ ,
619	which can be transformed to an estimate for $lpha.$ In the case of relative establishment probability,
620	$\tilde{p}_c(x) = \alpha(x)/\alpha(0)$ , the (profile likelihood) confidence interval takes into account the uncertainty
621	in both numerator (i.e. results at streptomycin concentration x) and denominator (i.e. results in
622	streptomycin-free conditions). The fit of nested models is compared using the likelihood ratio test
623	(LRT) at 5% significance level, i.e. a $\chi^2$ test on model deviance with degrees of freedom equal to the
624	difference in number of fitted parameters between the two models.
625	
626	To test the null model of the inoculum size effect, we assume that mean inoculum size $\overline{N}$ is precisely
627	inversely proportional to the dilution factor applied to the inoculating culture, i.e. we neglect any
628	experimental error in the dilution steps relative to one another. Effective mean inoculum size, $ar{N}_{eff}$ ,
629	is estimated by fitting Eqn. 2 to population growth data in streptomycin-free media. Per-cell relative
630	establishment probability $ ilde{p}_c$ then remains as the single free parameter to fit at each tested
631	streptomycin concentration. The goodness of fit of the null model (Eqn. 2) is assessed for each test
632	concentration separately. Here, the relevant comparison in the LRT is to the full model (Eqn. 1),
633	where a distinct parameter $p_w$ (or equivalently $lpha$ ) is estimated for each inoculum size, without any
634	assumed relationship among them.
635	
636	All model fitting was implemented in R, version 3.3.1 (The R Foundation for Statistical Computing,
637	2016).
638	
639	
640	Acknowledgements: H.K.A. was supported by an Early Postdoc. Mobility Fellowship
641	(P2EZP3_165188) and an Advanced Postdoc.Mobility fellowship (P300PA_177789) from the Swiss
642	National Science Foundation. R.C.M. was supported by Wellcome Trust Grant 106918/Z/15/Z.
643	
644	Author contributions: H.K.A. and R.C.M. conceived of the study and designed experiments. H.K.A.
615	carried out experiments and data analysis with advice from $P \in M \cup K \setminus A$ developed the

645 carried out experiments and data analysis with advice from R.C.M. H.K.A. developed the

646	mathematical model, wrote the code and carried out model fitting. H.K.A. and R.C.M. wrote the					
647	paper.					
648						
649	Data availability: The data generated in this study will be deposited in a public repository upon					
650	manuscript acceptance.					
651						
652	Code	e availability: Custom R scripts for likelihood-based model fitting and comparisons will be made				
653	free	y available upon manuscript acceptance.				
654						
655						
656	Ref	erences				
657						
	[1]	World Health Organization, "Antimicrobial Resistance: Global Report on Surveillance," WHO Press, Geneva, 2014.				
	[2]	X. Zhao and K. Drlica, "Restricting the selection of antibiotic-resistant mutants: a general strategy derived from fluoroquinolone studies," <i>Clin. Infect. Dis.</i> , vol. 33 (Suppl 3), pp. S147-S156, 2001.				
	[3]	X. Zhao and K. Drlica, "Restricting the selection of antibiotic-resistant mutant bacteria: measurement and potential use of the mutant selection window," <i>J. Infect. Dis.</i> , vol. 185, pp. 561-565, 2002.				
	[4]	K. Drlica and X. Zhao, "Mutant selection window hypothesis updated," <i>Clin. Infect. Dis.,</i> vol. 44, pp. 681-688, 2007.				
	[5]	R. Canton and MI. Morosini, "Emergence and spread of antibiotic resistance following exposure to antibiotics," <i>FEMS Microbiol. Rev.</i> , vol. 35, pp. 977-991, 2011.				
	[6]	T. Day, S. Huijben and A. F. Read, "Is selection relevant in the evolutionary emergence of drug resistance?," <i>Trends Microbiol.</i> , vol. 23, pp. 126-133, 2015.				
	[7]	MC. Negri, M. Lipsitch, J. Blázquez, B. R. Levin and F. Baquero, "Concentration-dependent selection of small phenotypic differences in TEM beta-lactamase-mediated antibiotic resistance," <i>Antimicrob. Agents Chemother.</i> , vol. 44, pp. 2485-2491, 2000.				
	[8]	A. Liu, A. Fong, E. Becket, J. Yuan, C. Tamae, L. Medrano, M. Maiz, C. Wahba, C. Lee, K. Lee, K. P. Tran, H. Yang, R. M. Hoffman, A. Salih and J. H. Miller, "Selective advantage of resistant strains at trace levels of antibiotics: a simple and ultrasensitive color test for detection of antibiotics and genotoxic agents," <i>Antimicrob. Agents Chemother.</i> , vol. 55, no. 3, pp. 1204-1210, 2011.				
	[9]	E. Gullberg, S. Cao, O. G. Berg, C. Ilbäck, L. Sandegren, D. Hughes and D. I. Andersson, "Selection of resistant bacteria at very low antibiotic concentrations," <i>PLoS Pathogens</i> , vol. 7, p. e1002158, 2011.				
	[10]	D. I. Andersson and D. Hughes, "Evolution of antibiotic resistance at non-lethal drug concentrations," <i>Drug Resistance Updates,</i> vol. 15, pp. 162-172, 2012.				

[11] P. Abel zur Wiesch, R. Kouyos, J. Engelstädter, R. R. Regoes and S. Bonhoeffer, "Population biological principles of drug-resistance evolution in infectious diseases," *Lancet Infect. Dis.*, vol. 11, pp. 236-247, 2011.

- [12] Z. Patwa and L. M. Wahl, "The fixation probability of beneficial mutations," J. R. Soc. Interface, vol. 5, pp. 1279-1289, 2008.
- [13] I. Frost, W. P. J. Smith, S. Mitri, A. San Millan, Y. Davit, J. M. Osborne, J. M. Pitt-Francis, R. C. MacLean and K. R. Foster, "Cooperation, competition and antibiotic resistance in bacterial colonies," *ISME J.*, vol. 12, pp. 1582-1593, 2018.
- [14] I. Levin-Reisman, I. Ronin, O. Gefen, I. Braniss, N. Shoresh and N. Q. Balaban, "Antibiotic tolerance facilitates the evolution of resistance," *Science*, vol. 355, pp. 826-830, 2017.
- [15] European Committee for Antimicrobial Susceptibility Testing (EUCAST), "Determination of minimum inhibitory concentrations (MICs) of antibacterial agents by broth dilution," *Clin. Microbiol. Infect.*, vol. 9, pp. 1-7, 2003.
- [16] I. Brook, "Inoculum effect," Rev. Infect. Dis., vol. 11, pp. 361-368, 1989.
- [17] K. I. Udekwu, N. Parrish, P. Ankomah, F. Baquero and B. R. Levin, "Functional relationship between bacterial cell density and the efficacy of antibiotics," *J. Antimicrob. Chemother.*, vol. 63, pp. 745-757, 2009.
- [18] T. Artemova, Y. Gerardin, C. Dudley, N. M. Vega and J. Gore, "Isolated cell behavior drives the evolution of antibiotic resistance," *Molec. Syst. Biol.*, vol. 11, p. 822, 2015.
- [19] K. C. Haltalin and J. D. Nelson, "In vitro susceptibility of Shigellae to sodium sulfadiazine and to eight antibiotics," *JAMA*, vol. 193, pp. 705-710, 1965.
- [20] S. E. McLinn, J. D. Nelson and K. C. Haltalin, "Antimicrobial susceptibility of Hemophilus influenzae," *Pediatrics*, vol. 45, pp. 827-838, 1970.
- [21] I. El Meouche, Y. Siu and M. J. Dunlop, "Stochastic expression of a multiple antibiotic resistance activator confers transient resistance in single cells," *Sci. Rep.*, vol. 6, p. 19538, 2016.
- [22] T. Bergmiller, A. M. C. Andersson, K. Tomasek, E. Balleza, D. J. Kiviet, R. Hauschild, G. Tkačik and C. C. Guet, "Biased partitioning of the multidrug efflux pump AcrAB-TolC underlies longlived phenotypic heterogeneity," *Science*, vol. 356, pp. 311-315, 2017.
- [23] E. A. Yurtsev, H. X. Chao, M. S. Datta, T. Artemova and J. Gore, "Bacterial cheating drives the population dynamics of cooperative antibiotic resistance plasmids," *Molec. Syst. Biol.*, vol. 9, p. 683, 2013.
- [24] N. M. Vega and J. Gore, "Collective antibiotic resistance: mechanisms and implications," *Curr. Opin. Microbiol.*, vol. 21, pp. 28-34, 2014.
- [25] R. A. Sorg, L. Lin, G. S. van Doorn, M. Sorg, J. Olson, V. Nizet and J.-W. Veening, "Collective resistance in microbial communities by intracellular antibiotic deactivation," *PLoS Biol.*, vol. 14, p. e2000631, 2016.
- [26] M. F. Schenk, I. G. Szendro, J. Krug and J. A. G. M. de Visser, "Quantifying the adaptive potential of an antibiotic resistance enzyme," *PLoS Genet.*, vol. 8, p. e1002783, 2012.
- [27] H. R. Meredith, J. K. Srimani, A. J. Lee, A. J. Lopatkin and L. You, "Collective antibiotic tolerance: mechanisms, dynamics and intervention," *Nat. Chem. Biol.*, vol. 11, pp. 182-188, 2015.
- [28] J. Karslake, J. Maltas, P. Brumm and K. B. Wood, "Population density modulates drug inhibition and gives rise to potential bistability of treatment outcomes for bacterial infections," *PLoS Comput. Biol.*, vol. 12, p. e1005098, 2016.
- [29] D. R. Gifford, J. A. G. M. de Visser and L. M. Wahl, "Model and test in a fungus of the probability that beneficial mutations survive drift," *Biol. Lett.*, vol. 9, p. 20120310, 2012.
- [30] D. R. Gifford and R. C. MacLean, "Evolutionary reversals of antibiotic resistance in experimental populations of Pseudomonas aeruginosa," *Evolution*, vol. 67, pp. 2973-2981, 2013.

- [31] J. Coates, B. R. Park, D. Le, E. Simsek, W. Chaudhry and M. Kim, "Antibiotic-induced population fluctuations and stochastic clearance of bacteria," *eLife*, vol. 7, p. e32976, 2018.
- [32] H. K. Alexander, G. Martin, O. Y. Martin and S. Bonhoeffer, "Evolutionary rescue: linking theory for conservation and medicine," *Evol. Appl.*, vol. 7, pp. 1161-1179, 2014.
- [33] G. Martin, R. Aguilée, J. Ramsayer, O. Kaltz and O. Ronce, "The probability of evolutionary rescue: towards a quantitative comparison between theory and evolution experiments," *Phil. Trans. R. Soc. B*, vol. 368, p. 20120088, 2013.
- [34] R. E. Moxon and P. A. Murphy, "Haemophilus influenzae bacteremia and meningitis resulting from survival of a single organism," *PNAS*, vol. 75, no. 3, pp. 1534-1536, 1978.
- [35] K. Koutsoumanis, "A study on the variability in the growth limits of individual cells and its effect on the behavior of microbial populations," *Int. J. Food Microbiol.*, vol. 128, pp. 116-121, 2008.
- [36] A. S. Haines, K. Jones, M. Cheung and C. M. Thomas, "The IncP-6 plasmid Rms149 consists of a small mobilizable backbone with multiple large insertions," *J. Bacteriol.*, vol. 187, pp. 4728-4738, 2005.
- [37] D. Sandvang, "Novel streptomycin and spectinomycin resistance gene as a gene cassette within a class 1 integron isolated from Escherichia coli," *Antimicrob. Agents Chemother.*, vol. 43, pp. 3036-3038, 1999.
- [38] W. G. Cochran, "Estimation of bacterial densities by means of the "most probable number"," *Biometrics*, vol. 6, pp. 105-116, 1950.
- [39] M. A. Hurley and M. E. Roscoe, "Automated statistical analysis of microbial enumeration by dilution series," *J. Appl. Bacteriol.*, vol. 55, pp. 159-164, 1983.