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1 Chance and pleiotropy dominate genetic diversity

2 in complex bacterial environments

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17 Abstract

18 How does environmental complexity affect the evolution of single genes? Here, 19 we measured the effects of a set of mutants of Bacillus subtilis glutamate 20 dehydrogenase across 19 different environments – from homogenous single cell 21 populations in liquid media to heterogeneous biofilms, plant roots and soil 22 communities. The effects of individual gene mutations on organismal fitness 23 were highly reproducible in liquid cultures. Strikingly, however, 84% of the 24 tested alleles showed opposing fitness effects under different carbon and 25 nitrogen sources (antagonistic pleiotropy). In biofilms and soil samples, different 26 alleles dominated in parallel replica experiments. Accordingly, we found that in 27 these heterogeneous bacterial communities the fate of mutations was dictated by 28 a combination of selection and drift. The latter was driven by programmed 29 prophage excisions that occurred along biofilm development. Overall, per 30 individual condition, by the combined action of selection, pleiotropy and chance, 31 a wide range of glutamate dehydrogenase mutations persisted and sometimes 32 fixated. However, across longer periods and multiple environments nearly all 33 this diversity would be lost - indeed, considering all environments and 34 conditions we have tested, wild-type is the fittest allele.

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38 The function of most genes may be essential in some conditions, but only marginally contributing, or even redundant, in other conditions ¹⁻⁴. The effects of 39 40 mutations on organismal fitness are therefore environment-dependent, giving 41 rise to complex, pleiotropic genotype-by-environment interactions ^{5,6}. Moreover, 42 bacterial populations often do not comprise single cells, but rather have a 43 structure as in biofilms. Under this complexity: changing environments and 44 heterogeneous bacterial communities, the fate of mutations could also be 45 dictated by population bottlenecks (drift) or rapid takeover of beneficial 46 mutations in other genes (selective sweeps) $^{7-9}$. Consequently, the frequency of a 47 given gene allele may change dramatically (from perishing to fixation) with no 48 relation to its molecular function ^{10,11}.

49 We aimed at an experimental setup would examine how complex 50 bacterial growth states and environments might shape protein evolution. 51 Previous systematic mappings were based on a direct linkage between protein 52 stability and function and organismal survival, thus enabling measurements of 53 effects of mutations at the protein level ^{5,12-15}. However, how mutations in a 54 single gene-protein affect organismal fitness under varying environments and 55 conditions is largely unexplored ¹⁶. We thus chose as our model *Bacillus subtilis* 56 NCIB 3610, a non-domesticated strain capable of growing in diverse aquatic and 57 terrestrial environments ¹⁷. We explored the effects of mutations in different 58 conditions: in dispersed cells in liquid, but also in biofilms where phenotypic and 59 genetic variability prevails ¹⁸. We also mapped the effects of mutations during spore formation and germination ¹⁹ and in more complex and close to natural 60 61 environments including soil, rhizosphere and plant roots.

62 A catabolic glutamate dehydrogenase (GDH) was our model protein. This 63 enzyme is essential when amino acids such as proline serve as sole carbon-64 nitrogen sources ²⁰. However, in the presence of ammonia and glycolytic sugars, 65 GDH activity is redundant as glutamate must be synthesized rather than 66 catabolized. GDHs therefore respond to changes in carbon-nitrogen sources, and as regulators of glutamate homeostasis, are also associated with biofilm 67 development ^{21,22}. B. subtilis has two catabolic GDHs, RocG and GudB. The latter 68 69 is constitutively expressed, and is regulated via association of its hexameric form 70 ²³. GudB has also regulatory roles ^{24,25} via interactions with the transcriptional 71 activator of glutamate synthase ²⁵ and with an essential transcription 72 termination factor, NusA, that also modulates the stringent response ²⁶⁻²⁸. We 73 explored mutations in the oligomeric interface of GudB, aiming at multilateral 74 effects on GudB's enzymatic and regulatory functions.

Altogether, these choices of organism and enzyme allowed us to readily examine and quantify the fate of GudB alleles in a range of different growth conditions and environments, also mimicking natural habitats where strong evolutionary forces act ¹¹.

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80 **Experimental setup and data processing**

We anticipated that the effects of the explored mutations would be complex and condition-dependent. We thus opted for high rather than broad coverage and mapped 10 positions within a single ~150 base pairs segment that resides at GudB's oligomeric interface while choosing highly conserved (D58) as well as highly diverged positions (M48, or S61; **Table S1**). Positions were diversified using NNS (whereby N represents any of the 4 bases, and S, G or C). The resulting GudB library contained in total 320 single mutant alleles (including wild-type),
whose genomes differ, in principle, by a single mutation: 200 different amino
acid alleles and 10 stop-codons. The library therefore included synonymous
alleles whereby the same amino acid was encoded by 2 or 3 different codons.
This allele library was incorporated into the chromosome of *B. subtilis* NCIB
3610 under *gudB*'s original promoter and terminator.

93 This starting population (the initial mix, hereafter) was used to inoculate 94 cultures grown in an array of different conditions. We tested 7 different growth 95 states where the population complexity varies from single cells to community: 96 liquid, pellicles (air-liquid biofilms), spores, germinated spores, biofilms grown 97 on agar including on carbon-nitrogen gradients, and soil colonization. Up to 5 98 different carbon-nitrogen sources were used that, at least as far as the 99 phenotypes of the GudB knockout indicate, inflict different levels of selection on 100 GudB: Glutamate *plus* ammonia (GA), where Δ GudB has no growth effect; 101 glutamate *plus* glycerol (GG), arginine (A), and arginine *plus* proline (PA), where 102 Δ GudB exhibits a slight growth defect, and proline (P), where Δ GudB exhibits the 103 strongest growth defect (Fig. S1). Thus, in total, we tested 19 conditions. At each 104 condition, three to five biological replicas were performed by inoculating from 105 the same initial mix. The cultures were grown in parallel, and individually 106 analyzed. Illumina sequencing was applied to determine the frequency of each of 107 the gudB alleles in the initial mix and after growth. Following filtering (see 108 Methods), we obtained data for 244 up to 269 individual alleles per experiment 109 (Data S1 & Fig. S2).

110 The ratio between an allele's frequency at the end of growth and in the 111 initial mix was derived, and this ratio is referred to as the frequency coefficient

112 (FC; Data S2). Given the experimental error in determining FC values, values 113 between 0.8 and 1.2 were classified as 'neutral', FC \leq 0.8 assigned a mutation as 114 'deleterious', and FC >1.2 as 'beneficial'. Mutations with FC ≤ 0.1 were classified 115 as 'highly deleterious', and similarly, FC \geq 10 as 'highly beneficial' (see Methods). 116 Note that the number of generations in liquid growth (~ 50 generations) and 117 spores (a dormant non-replicative form of B. subtilis), for example, differs 118 fundamentally. Moreover, in pellicles and biofilms, the number of generations 119 cannot be readily determined –different cell types having different growth rates 120 ²⁹. So, while we could not calculate selection coefficients (s), one should keep in 121 mind that an FC value of 0.8 in the spores is in effect equivalent to extinction 122 across 50 generations in liquid $(0.8^{50} = 10^{-5})$.

123

124 Irreproducibility – selection versus drift

125 Our first observations indicate two contrasting scenarios. In liquid cultures, for 126 example, we observed highly reproducible FC values in biological replicas (Fig. 127 1a). Given the small sample numbers (3 replicas, 5 in few cases like the 128 rhizosphere), the observed variance may underestimate the actual variance. 129 However, the repetitively low variance levels in a range of different liquid 130 conditions, and in other replica measurements in liquid ³⁰, supports high 131 reproducibility. In biofilms, however, despite the fact that we did not bottleneck 132 any population upon inoculum, the correlation between replicas was very low 133 (Fig. 1b). The reproducibility between biological replicas indicates selection, 134 thus indicating that that protein and organismal fitness are tightly coupled. In 135 biofilms however, the lack of reproducibility indicates the dominance of drift, i.e., 136 random sampling of GudB alleles.

137 To quantify the contribution of selection versus drift in different 138 conditions, we used two criteria. Firstly, we compared the variability in FC 139 values between replicas by calculating the standard deviation (SD) per allele 140 (using the logarithm of the FC values; see Methods). The average SD value for all alleles in each experiment (\overline{SD}) is given for 7 general growth states (Fig. 1c; Fig. S3a) 141 142 & **Table S2**). As can be seen, in liquid, pellicles and spores, the \overline{SD} values 143 between biological replicas were low (< 0.06). In biofilms and bulk soil, however, the \overline{SD} values were > 0.25 indicating low reproducibility. 144

145 Secondly, if drift dictates the fate of GudB alleles, codons of the same 146 amino acid would exhibit very different FC values. The deviations between 147 synonymous codons of the same amino acid alleles were calculated, averaged for 148 all alleles in the same experiment, and then for all replica experiments per 149 condition (SD syn, in log values; Fig. 1d; Fig. S3b & Table S3). Note that the 150 \overline{SD} syn criterion holds within individual replica experiments and is thus 151 independent of the comparison of \overline{SD} between biological replicas. Nonetheless, 152 these criteria are clearly correlated (**Fig. 1c** & **d**). Overall, it appears that in 153 liquid, pellicles and spores, the FC values report the outcome of selection acting 154 on GudB alleles at the amino acid level as expected (in few alleles, selection also 155 acted reproducibly at the codon level, **Fig. S4**). In contrast, in biofilms and bulk soil we consistently observed higher \overline{SD} and \overline{SD} syn values. In some biofilm 156 experiments, the \overline{SD} syn values exceeded 3 (*i.e.*, > 10³ SD values on a liner scale). 157 158 Thus, in effect, a single codon had taken over.

Given that some conditions were selection-dominated and others aresubject to chance, we divided our analysis in two. Firstly, we analyzed selection

dominated conditions (liquid, pellicles and spores) to examine whether and how
the fate of GudB mutations change under different environments. Secondly,
conditions where drift prevailed (germination, biofilms and soil colonization)
were analyzed to reveal the relative contributions of selection *versus* chance and
the molecular mechanisms of drift.

166

167 **Pleiotropy - fitness-effects of mutations are condition-dependent**

168 While the FC values, and hence the fitness effects of mutations, were 169 reproducible under many conditions, their distribution varied widely between 170 conditions, including between carbon-nitrogen sources (Fig. S5). This indicates 171 pleiotropy - individual GudB alleles have different fitness effects in different 172 conditions. To quantify the level of pleiotropy, we compared the FC values of the 173 same GudB mutation across the 9 individual selection-dominated conditions. 174 Because the number of generations differs from one condition to another, we 175 focused on shift from beneficial to deleterious, and vice versa (sign, or antagonistic pleiotropy) as the sign of the FC values of mutations indicates their 176 177 fitness sign irrespective of generation numbers. Representative dot plots 178 comparing the FC values across 3 different liquid conditions are shown (Fig. 2a). 179 These indicate that pleiotropy is common, even when comparing liquid cultures 180 with overlapping carbon-nitrogen sources. In particular, a significant number of 181 GudB mutations show antagonistic pleiotropy (dashed squares, Fig. 2a). Indeed, 182 the Pearson correlation values for the 36 possible pair-wise comparisons of 9 183 conditions were below 0.7 (Fig. 2b). Across all selection-dominated conditions, 184 up to 84% of alleles showed antagonistic pleiotropy in one or more of the 36 pair 185 wise comparisons, and 70% of alleles showed mild or strong antagonistic

186 pleiotropy. These pleiotropic effects are far beyond experimental noise, as

187 indicated by comparison to a control sample (**Fig. 2c**).

188 Overall, the dominance of pleiotropy meant that across all protein -189 organismal fitness coupled conditions, 86% of the alleles were beneficial in at 190 least one condition. However, not a single mutation was beneficial across all 191 conditions. Further, if a mutation were to be considered deleterious if it was 192 purged in at least one condition, then 98% of the tested GudB mutations were 193 deleterious. To our knowledge, the degree of pleiotropy in protein mutations 194 across multiple environments has not been measured so far. The exceedingly 195 high degree of pleiotropy we found may relate to GudB's multiple roles, as an 196 enzyme and regulator, and also to the chosen mutated positions (oligomer 197 interface), but it may well be a general characteristic of proteins with key 198 physiological roles. In some samples, only two alleles were present at >1%199 frequency, one being wild-type (**Fig. 3** & **Data S1**). The near-fixation of relatively 200 few alleles could indicate very strong selection acting on GudB. However, the 201 high \overline{SD} and exceedingly high \overline{SD} syn values suggest fixation by chance (**Fig. 1c** & 202 **d**). What is then the nature of these few GudB 'winners', are they merely lucky? 203

204 Combined action of selection and drift in heterogeneous environments

While drift dominated in biofilms and soil colonization, curiously, wild type GudB was enriched in up to 85% of these experiments suggesting that selection does play a role (**Fig. 3**). To assess the action of selection, we compared the three biofilm areas. There appears a systematic trend, whereby enriched alleles in the edge are more likely to arise from alleles that persisted or even enriched in the center (**Fig. 4a**). Similarly, although gradient biofilms were clearly dominated by 211 drift, 75% of the enriched alleles were neutral or beneficial under liquid growth 212 with proline, a condition under which GudB experiences the strongest selection 213 (Fig. 4b). This suggested that GudB is under selection at the early stages of 214 biofilm development. Accordingly, we found that in biofilm centers, the FC values are less skewed and more reproducible (**Fig. S5**), and also, the center \overline{SD} syn 215 216 values are half than in the edge or wrinkles (**Table S3** & **Fig. S3b**). The \overline{SD} syn 217 values are obviously higher in the biofilms' center compared to liquid cultures, 218 but the trend suggests that at the onset of the biofilm's development, selection 219 acts on GudB (Table S3 & Fig. S3b).

220 Similarly, we tested for signatures of selection in soil colonization. As in 221 biofilms, there is a statistically significant trend, whereby enriched alleles in the 222 root are more likely to arise from alleles that were enriched in the soil (**Fig. 4c**). 223 Further, 19 amino acid substitutions were enriched in at least 10 out of the 15 224 sequenced populations, suggesting reproducibility. Of these, in two amino acid alleles, both synonymous codons enriched (D59A and D59V; \overline{SD} syn = 0.42 and 225 226 0.35; Table S4) thus indicating selection. Selection during soil colonization is 227 also manifested in the variation between biological replicas (\overline{SD} values) of alleles 228 that were enriched in root populations being on average 20 % smaller than those 229 that were not (**Fig. 4d**). Finally, stop codons were purged in all biofilms and soil 230 populations, indicating, as expected, that GudB's activity is required for B. 231 *subtilis'* survival under these conditions ^{22,23}.

Altogether, as expected ¹¹, in biofilms, and particularly in soil colonization,
both drift and selection determine the fate of GudB alleles. The drivers of drift in
biofilms were further unraveled as described in the next section.

235

236 **Drift in biofilms is driven by programmed prophage excisions**

237 Mutagenic rates in biofilms are high and mutations with a selective advantage 238 rapidly take over (genetic sweeps) ³¹. Growth in biofilms is also spatially defined, 239 giving rise to segregated lineages whereby an entire segment of the biofilm's 240 edge stems from a single cell in which a beneficial mutation had first emerged ¹¹. 241 GudB mutations that happen to be in these 'founder' cells might therefore fixate 242 along these lineages. To examine this hypothesis, we sequenced samples for 243 which enough genomic DNA was available (6 ordinary and 12 gradient biofilms, 244 and for comparison, 2 Initial Mixes, 6 liquid and 4 pellicle samples). A range of 245 single nucleotide polymorphisms (SNPs) in various loci was identified across 246 these samples (**Data S3**). We focused, however, on identifying genomic 247 mutations that were not, or scarcely observed in the Initial Mix and/or in liquid 248 samples and were thus emerged and enriched in the biofilms.

249 Foremost, we observed two large genome deletions that occurred in all 250 biofilms with a frequency approaching 100% (Figs. 5a & 5b). These deletions 251 correspond to the excision of two mobile genetic elements, or prophages, skin 252 and SP- β ³²⁻³⁴. Excision of *skin* generates a functional protein: sigK - a 253 sporulation-specific transcription factor essential for cell differentiation in B. 254 subtilis ³⁵. The excision of SP- β generates a functional CapD – an enzyme 255 mediating production of poly-y-glutamate, an essential component in capsule 256 formation and biofilm development ^{36,37}. Nearly all biofilm cells carried one of 257 these variations, and most cells carried both (**Fig. 5a & b**; **Table S5 & Data S3**). 258 Given their dominance ³⁸, these structural variations are likely to be the primary 259 cause of genetic sweeps and of GudB's drift (Fig. 5c). These prophage excisions

are also likely to occur in the soil, but the DNA recovered from these samples

261 was insufficient to allow genome sequencing.

262 Exclusively in biofilms, we also detected 59 enriched SNPs in a conserved 263 region of 16S rRNA (Table S5 & Data S3). However, B. subtilis has ten 16S rRNA 264 gene copies. Since these are essentially identical, we could not determine which 265 of these 10 paralogues carried mutations. However, per population, 98% of the 266 16S rRNA mutations occurred in the same Illumina read suggesting that one 267 paralogue was highly mutated while others remained intact (Fig. S6). At this 268 stage, the mechanism of inactivation by multiple proximal 16S mutations, and 269 how inactivation affects biofilm development, remains unclear. Large differences 270 in expression levels of 16S rRNA genes were identified in *P. aeruginosa* biofilms 271 39 , and ribosomal heterogeneity has been linked to biofilm development in *B*. 272 subtilis⁴⁰. However, to our knowledge, mutations in the 16S rRNA genes have not 273 been reported in biofilms.

274 Overall, the 16S rRNA SNPs, and the structural variations in particular, 275 seem to have a key role in biofilm development in *B. subtilis*. Accordingly, most of 276 these genetic variations were reproducible between replica experiments (**Table** 277 **S5**) suggesting that they arose during biofilm growth, and then enriched due to 278 their adaptive potential ¹¹. Thus, although selection dictated the fate of GudB 279 mutations in the early stages of biofilm development, once biofilm promoting 280 mutations appeared and rapidly took over, they drove the fixation of any GudB 281 that happened to be present in the mutated cell.

282

283 Concluding remarks

284 Pleiotropy of mutations is assumed but not at the magnitude unraveled here. 285 Environmental changes, including minute ones like addition of arginine to a 286 proline medium, completely revert the effect of up to 84% of the tested GudB 287 mutations. Pleiotropy severely restricts protein sequence space -if all tested 288 conditions are considered, only 2% of the tested GudB mutations are neutral in 289 the 9 reproducible conditions -when selection strongly acts on GudB. This 290 suggests that wild-type GudB's sequence is in fact unique in being shaped under 291 multiple constrains and environments, as also indicated by its dominance in 292 many conditions. Together, pleiotropy and drift dictate the evolution of short-293 term polymorphism (micro-evolution), but also the evolution of protein 294 sequences along long evolutionary times and across species (macro-evolution). 295 The correlation between the effects of mutations in laboratory mappings under 296 one specific condition and the natural sequence diversity is therefore limited 12. 297 Merging of data from multiple reproducible conditions does not seem to improve 298 correlation, also when applying a number of machine learning techniques 299 (stochastic gradient descent classifier, support vector machines, or random 300 forest classifier; **Fig. S7**).

301 Thus, along short evolutionary periods, proteins experience variable and 302 opposing selection pressures. Additionally, drift may lead to rapid fixation of 303 alleles that are marginally fit or even deleterious. The effects of drift have been 304 extensively studied initiated by Kimura's neutral theory ¹⁰. Our results quantify 305 its effects in bacterial populations and the potential effect of drift in combination 306 with selection across different environments. For example, nearly 80% of the 307 tested mutations survived or enriched sporulation, and a single spore can start a 308 whole new population. However, once the environment changes, such alleles will

- 309 be rapidly lost unless compensated by other mutations. Indeed, along macro-
- 310 evolutionary time scales, epistasis dominates gene and genome sequences ⁴¹.

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- 447

448 **Authors Contribution**

- 449 L.N.G. and D.S.T. designed experiments and wrote the manuscript. L.N.G., D.D.
- 450 and D.S.T. analysed the data. L.N.G. performed all experiments, except selection
- 451 in soil colonization that was performed in collaboration with E.K. and A.A. D.D.
- 452 and A.E. wrote the scripts used for data analysis and visualization. E.P. applied
- 453 machine learning classification.

454 Fig. 1. Selection versus chance-dominated conditions. (A) Dot-plot indicating 455 reproducible measurements of frequency coefficients of individual mutations 456 (FC values) in three parallel replica liquid cultures with proline as carbonnitrogen source. \overline{SD} is the average standard deviation between biological 457 458 replicas. The S.D. values were calculated per each amino acid allele based on 459 logFC values and averaged for all alleles in a given condition. (B) The same 460 analysis of three parallel biofilms with arginine as carbon-nitrogen source indicates low reproducibility. (C) The \overline{SD} values categorized by the 7 general 461 growth states tested here. Each point represents the \overline{SD} value between replicas 462 463 of the same experiment (the distributions of SD values per each condition are shown in Fig. S3a). (D) \overline{SD} syn represents the standard deviation between the 464 logFC values of synonymous codons. The standard deviations per allele were 465 466 averaged for all synonymous alleles in the same replica experiment, and then 467 averaged across the 3 replica experiments in a given condition (the distributions 468 of SD*syn* values per experiment are shown in **Fig. S3b**).

Fig. 2. The pleiotropic effects of alleles across different conditions. (A) A 470 471 dot-blot correlation of FC values of individual alleles in three different liquid 472 carbon-nitrogen sources. The red squares encompass alleles that show sign 473 pleiotropy - i.e., a change from beneficial to deleterious, or vice versa. (B) 474 Pairwise correlation of the FC values in all 9 different reproducible conditions 475 (average FC values of all alleles and replicas). Colors indicate the Pearson 476 correlation values (-1, negative correlation; 0, no correlation; 1, positive 477 correlation). The strongest anti-correlation was found with arginine as carbon-478 nitrogen source (black square). (C) The distribution of alleles by their level of 479 sign pleiotropy. From pale to dark): (i) Weak sign pleiotropy (changes between 480 deleterious and beneficial); (ii) Mild sign pleiotropy (changes from highly 481 deleterious to beneficial, or from highly beneficial to deleterious); and (iii) 482 Strong sign pleiotropy (changes from highly deleterious to highly beneficial, or 483 vice versa). The fraction of alleles showing mild or strong sign pleiotropy is 484 shown above the bars. The control dataset comprises 4 completely independent 485 growth experiments in liquid proline, each inoculated from a different initial mix 486 and grown on separate occasions (Fig. S8). Nonetheless, none of the alleles in 487 this control set exhibited strong pleiotropy.

489	Fig. 3. Genetic sweeps in biofilms and soil. Photographs of: (A) normal
490	biofilms; (B) gradient biofilms; (C) a scheme of soil colonization (shown biofilms
491	with proline as carbon-nitrogen source). (D) The distribution of frequency of
492	individual alleles for different growth states with proline as carbon-nitrogen
493	source. Bar widths represent allele frequency from raw read counts (Rf values;
494	Data S1). Magenta corresponds to wild type GudB. (E) Alleles with $Rf \ge 1\%$ were
495	identified and their number and sum of frequencies are shown (averages and
496	standard deviations for all experiments in a given condition). Blue designates
497	selection-dominated conditions and red drift-dominated ones, as in Fig. 1 .
498	

499 Fig. 4. The combined action of selection and chance in biofilms (red) and 500 soil colonization (green). (A) Alleles that enriched in the edge of the biofilms 501 are more likely to arise from alleles that were neutral or enriched in the center. 502 The distribution of categorized FC values from all biofilm centers (grey) 503 compared to the distribution of FC values of center alleles that were enriched in 504 the edge (red). (B) The distribution of categorized FC values of all alleles in all 505 liquid conditions (grey) compared to the distribution of FC values of liquid 506 alleles that were enriched in the edge of biofilms (red). (C) Alleles enriched in 507 the root are more likely to arise from alleles that were enriched in the bulk soil. 508 The distribution of categorized allele FC values in all soil samples (grey) 509 compared to the distribution of FC values of soil alleles that were enriched in the 510 root (green). T-tests were computed per each FC category. One, two or three 511 asterisks indicate a p-value < 0.05, < 0.001, or < 0.005, respectively. (D) The 512 distribution of SD values (variability between replica experiments, as in **Fig. 1c**) 513 of alleles enriched in one or more root populations compared to alleles that were 514 never enriched in the roots. 515

516 Fig. 5. Programmed genomic excisions drive GudB's drift in biofilms.

517 Schematic representation of *B. subtilis* genomic organization before and after the 518 excision of the prophage mobile elements SP- β (A) and skin (B) and their 519 position in the genome. (C) These excisions were absent in the initial mix yet 520 dominated biofilms and went to near fixation in the edge of gradient biofilms (for 521 frequencies in individual experiments see Table S5). Excision of the mobile 522 elements occurred in two different genomic locations within the same 523 experiment. The values were summed and averaged according to the general 524 condition shown. The details of the excisions (location and frequency) per 525 experiment are shown in **Data S3**.

527 Materials and Methods

528 Strains

B. subtilis NCIB 3610 DS7187 (gently gifted by Dr. Daniel B. Kearns ⁴²) that lacks
the ComI peptide and has high competence capacity similar to domesticated *B. subtilis* strains was recruited to this study. *Bacillus subtilis* NCIB 3610 gudB::tet
strain ²³ genomic DNA was transformed into *B. subtilis* NCIB 3610 DS7187. *B. subtilis* NCIB 3610 *ΔcomI gudB::tet* was thus isolated, and was phenotypically
and genetically tested.

535

536 **GudB allele library construction**

537 We performed site directed mutagenesis in 10 codons (amino acids: M46, L48, 538 K52, D58, D59, S61, K63, T66, Y68, S75) of the gudB gene cloned in the 539 pDG_GudB plasmid 23 . The codons were mutated to NNS (N = all bases & S = C or 540 G) whereby the 20 standard amino acids and 1 stop codon is encoded. The codon 541 mutagenesis was done in one step PCR protocol and independently for each 542 position. Thus, we created 10 libraries, each containing 20 different amino acid 543 alleles (non-synonymous, missense mutations), 1 stop-codon (nonsense), and 11 544 synonymous alleles (alternative codons encoding the same amino acid). All 545 mutagenic PCRs were performed with Kapa HiFi HotStart Ready Mix (Kapa 546 Biosystems) following manufacturers conditions (**Table S6** shows the sequence 547 of all primers). The 10 PCR products were purified and used to transform the E. 548 coli T10 strain (Thermo Fisher Scientific). Clones were pulled together after an 549 overnight growth on LB + Ampicillin (100 μ g/ml) agar plates at 37°C. At this 550 stage, 4 to 6 clones per library were isolated and analyzed by sequencing. Total 551 plasmid DNA from these library transformations was extracted and also 552 analyzed by sequencing. Each of the 10 libraries contained, after transformation, 553 at least 10^5 clones, corresponding to \geq 1000-fold coverage per allele. 554 Approximately 10 µg of plasmid DNA, from each library, was linearized (XhoI, 555 New England Biolabs, following manufactures conditions), purified, and used to 556 transform the *B. subtilis* NCIB 3610 gudB::tet ∆comI strain. Transformations were 557 performed as described ²³. After transformation, overnight growth on in + 558 Spectinomycin $(100 \mu g/ml)$ + Glucose (0.5 mg/ml) agar plates was used as 559 selection. The resulting cells were pulled together and kept at -20° C in 50%

560 glycerol. In total, 10 *B. subtilis* libraries were constructed in parallel and each 561 contained, after transformation, at least 10^4 clones (≥ 100 -fold coverage per 562 allele). Genomic DNA extraction of each library was performed (GenElute -563 Sigma). The integrity of the mutagenic process was verified by sanger 564 sequencing the *amyE::gudB* locus indicating that mutations were observed only 565 in the diversified codon.

566

567 Selection and growth conditions

10 ml of LB with Glucose (0.5%), ammonium sulfate (0.5%) and spectinomycin 568 569 $(100 \ \mu g/ml)$ cultures were inoculated with 1 ml of each library stock. The 570 cultures were grown overnight at 37° C with shaking 500 µl of the overnight 571 culture was used to inoculate 3 ml of LB plus glucose (0.5%) and ammonium 572 sulfate (0.5%). The cultures were incubated at 37°C with shaking and once the 573 $0.D_{.600}$ reached 0.8 they were mixed equally and used as the starting population 574 (Initial Mix). A fraction of the cells at this stage were harvested by centrifugation 575 and stored for genomic DNA purification. In total, three different initial mixes were used for the experiments described here. Initial Mix #1 was used to 576 577 inoculate most liquid conditions (4 carbon-nitrogen sources), pellicles and 578 gradient biofilms. Initial Mix #2 was used to inoculate 1 liquid condition, spores, 579 germination and biofilms, and Initial Mix #3 was used to inoculate bulk soil 580 (**Data S1**). Detailed selection conditions are listed below:

581 For selection under liquid serial passages 100 ul of the Initial Mix was used to 582 inoculate 10 ml cultures of MS medium 23 with glucose (0.5%) plus ammonium 583 sulfate (0.5%), glutamate (0.5%) plus glycerol (0.5%), proline (0.5%), arginine 584 (0.5%) or proline (0.25%) plus arginine (0.25%). The cultures were incubated at 585 37° C with shaking until 0.D_{.600} reached 1 – 1.5, after which 100 μ l was used to 586 inoculate 10 mL of fresh medium. The serial passages were done every 24 hours 587 when proline (0.5%), arginine (0.5%) or proline (0.25%) plus arginine (0.25%)588 where used as carbon-nitrogen sources, and every 12 hours when glucose 589 (0.5%) plus ammonium sulfate (0.5%), or Glutamate (0.5%) plus glycerol 590 (0.5%), were applied. In total, all liquid passages were maintained for 591 approximately 50 generations.

592 For selection in pellicles, 100 ml of media (b), (c) and (d) were inoculated with 593 $100 \,\mu$ l of the initial mix cells. The culture was incubated at 30°C without shaking,

594 for 5 days.

For selection of spores and germinated spores, three ml of the Initial Mix was 595 596 used to inoculate 25 ml of Difco Sporulation Medium (DSM) in 250 ml flasks and 597 incubated at 37°C with 150 rpm shaking until 0.D.600 reached 0.4. This culture 598 was used to inoculate 250 ml of fresh DSM in 1L flasks. The cultures were 599 incubated 48h at 37°C with 150 rpm shaking. Cells were subsequently harvested 600 by centrifugation and stored at 4° C over night. After, cells were re-suspended 601 with 200 ml of cold deionized sterile water (dW) and incubated for 30 min at 602 4° C. Cells were harvested and re-suspended with 200 ml of cold distilled water (dW) and incubated overnight at 4°C with slow orbital agitation, to kill all 603 604 planktonic of vegetative cells. The culture was harvested, re-suspended in 30 ml 605 of dW and heated to 80°C for 20 min. Finally, spores were harvested, re-606 suspended in 10 ml of dW, and stored at -20° C. To germinate these spores, they 607 were diluted 1000 times in phosphate-buffered saline solution and 100 μ l of this 608 suspension was used to inoculate LB plus glucose (0.5%) agar plates (10 plates). 609 Approximately 10,000 colonies were obtained and pulled together.

610 For selection in biofilms, MS agar (1.5%) plates supplemented with different 611 carbon-nitrogen sources were prepared. For gradient biofilms, gradient agar 612 plates were prepared. First, square plates (12x12 cm) with MS agar (1.5%)613 medium were poured. After the agar solidified, an area of 2x14 cm was removed 614 from the top of the plate. In this area, a solution of either Proline 5%. Arginine 615 5%, monosodium glutamate 5% or Glycerol 5% in 1.5% agar was poured into 616 the removed section. For the glutamate plus glycerol gradient biofilm, two 617 opposite areas of the agar plate were removed. Into one, a solution of 618 monosodium glutamate (5%) in 1.5% agar was poured, and into the other, 619 glycerol (5%) plus 1.5% agar solution (see **Fig. S9a** for a graphic representation 620 of the agar plates preparation). All gradient agar plates were incubated for 24 h 621 at room temperature before use. We also calibrated the place in the gradient 622 plate where we inoculated the cells such that we observed growth after 1 night 623 incubation at 30° C (Fig. S9b). For growth in biofilms and gradient biofilms, 5 μ l 624 of the Initial Mix were used as inoculum. Plates were incubated for 4 days at

625 30°C and 2 more days at room temperature. The colony was then dissected in 3 areas (center, wrinkle and edge) for normal biofilms, and in 2 areas (center and upper) for gradient biofilms (illustrated in **Fig. S9c-g**). After selection in all the above-mentioned conditions the biomass was harvested and storage at -20°C. All growth experiments were performed in triplicate by inoculating with the same Initial Mix.

631 For selection in soil and plant roots, the Initial Mix was generated as above-632 mentioned except that the process was scaled up (instead of 3 ml, 10 ml of 633 culture was prepared per library). In total, 200 ml of the Initial Mix $(0.D_{.600} = 0.8)$ 634 was applied. This LB culture was washed three times (by means of centrifugation 635 and re-suspension) with 100 ml half strength Hoagland solution 43. Since 636 Hoagland's solution is not isotonic, the washes resulted in death of about a third 637 of the *B. subtilis* cells. Thus, handling the samples at this stage was performed as 638 fast as possible. After the final wash, the cells were re-suspended in half strength 639 Hoagland solution to a final $O.D_{.600}$ of 0.1. Natural soil was collected at the Ha-640 Masrek Reserve, Israel (31.793 N, 35.042 E), sifted through 2 mm sieve and 641 autoclaved three times for 30 min at 121° C. A total of five pots (size 10 x 8 x 5 642 cm) with autoclaved natural soil were drenched with the Initial Mix suspended 643 in half strength Hoagland Solution ⁴³. These potted soils drenched with bacterial 644 suspensions were used to plant tomato seedlings grown first in sterile 645 conditions. Seeds of tomato (Solanum lycopersicum L.; cv. Micro-Tom) were 646 surface-sterilized with 70% ethanol for 5 minutes and, 10 minutes with 3%647 bleach with 0.01% Tween 20. Surface-sterile seeds were germinated on sterile 648 filter paper (Whatman, catalog # 1001-085) saturated with half strength 649 Hoagland Solution for 7 days (23°C and 16 hours photoperiod). Six tomato 650 seedlings were transferred to each pot and grown for one month (21°C, 16h 651 light, 8h dark) with drenching with half strength Hoagland twice a week. Plants 652 were subsequently harvested from the five pots. Roots and rhizosphere samples 653 were collected for each replica experiment consisting a pool of six roots. First, 654 the plants were carefully removed from the soil. Roots were then cut out from 655 the plants and vortexed in 20 ml of washing solution (0.85% NaCl) for 30 s. This step was repeated one more time with a fresh washing solution. The combined 656 657 root washing solutions (40 ml) was centrifuged for 30 min at 3000 rpm and the

resulted pelleted samples corresponding to the rhizosphere were frozen in liquid
nitrogen and stored at -80°C. The washed roots were blotted in filter paper and
stored at -80°C until further use. Finally, bulk soil without roots was also stored

661 at -80°C.

662

663 Genomic DNA extraction

664 All samples, including pellicle, spores, biofilm and gradient biofilm samples, were 665 defrosted and re-suspended in 10 ml of dW. The samples were sonicated at 40%power, VibraCell, Sonics, for 10 min at 60 s intervals. Cells debris was harvested 666 667 by centrifugation (13,000 g for 20 min). Genomic DNA from all samples was 668 extracted using the GenElute Bacterial Genomic DNA Kit (Sigma-Aldrich) 669 generally following the manufacturer's instructions, with the exception of the 670 soil, rhizosphere soil and plant roots samples. For these samples, the PowerSoil 671 DNA Isolation kit of Mo Bio was used, following its manufacturer's instructions.

672

690

673 Illumina sample preparations

674 The mutagenized gudB fragment (from amino acids 45 to 81) was amplified 675 using the primers GudB In For (5'-676 CTCTTTCCCTACACGACGCTCTTCCGATCTnnnnnnCCCCGAAGAGGTATACGAATTGT 677 TAAAAGAG), and GudB_In_Rev (5'-678 CTGGAGTTCAGACGTGTGCTCTTCCGATCTCGCCTTTCGTTGGACCGAC). То the 679 GudB_In_For primer, 6 N's were added to increase the sequence variability 680 between amplicons. PCRs were performed with the KapaHiFi HotStart Ready Mix 681 (Kapa Biosystems) using approximately 100 ng of genomic DNA as template and 682 following manufacturer's instructions. Using 10 μ l of the PCR as template, a 683 second PCR was performed to add the Illumina adaptor sequence, using primers 684 (5'-GudB_Out_For 685 AATGATACGGCGACCACCGAGATCTACACTCTTTCCCTACACGACGC) and 686 GudB Out Rev (5'-687 CAAGCAGAAGACGGCATACGAGATTGTTATACGTGACTGGAGTTCAGACGTGTGC). 688 The Illumina index (underlined) was changed in the GudB_Out_Rev primer to 689 different Illumina indexes. Each condition was differently barcoded. All PCRs

were purified using the Agencourt AMPure XP (Beckman Coulter). The

691 concentration of PCR products was verified using Qu-bit assay (Life

692 Technologies).

693 Analysis of the Illumina reads

694 DNA samples were run using the Illumina NextSeq 150-bp paired-end kit. The 695 FASTQ sequence files were obtained for each run and customized using MatLab 696 8.0 and Python 3.6 scripts designed to count the number of each individual allele 697 in each sequenced sample. We filtered the reads to exclude any reads that have 698 mutations outside the mutagenized codons. All codons encoding for the wild-699 type amino acid were summed in one and assigned as WT. All other codons were 700 counted independently. The unprocessed read counts are shown in **Data S1**. 701 Further filtering excluded alleles with < 100 counts in the Initial mix to avoid 702 statistical uncertainty with respect to FC values. In total, we obtained data for up 703 to 269 individual alleles per condition out of the originally introduced 320 704 alleles. Per condition, a minimum of 380,000 reads was obtained. Thus, in 705 average, we obtained 1500 reads per allele.

706

707 Data Analysis

708 The frequency of each allele (f_i) was calculated as the ratio between the number 709 of reads for allele i divided by the total number of reads. The allele frequency 710 coefficient (FC_i) was subsequently calculated as the ratio of after selection (f_i) 711 divided by the frequency of the same allele in the initial mix (Fig. S2 & Data S2). 712 Normalization by the number of wild-type reads rather than by the total number 713 of reads gave essentially identical FC values for the majority of samples. 714 However, in the few samples where wild-type frequency was significantly 715 reduced after selection, normalization by wild-type reads resulted in high noise 716 and large biases including large changes in sign (higher sign pleiotropy). FC 717 values relate to fitness logarithmically, and thus logFC values were compared. To 718 this end, all FC's equal to zero had to be changed, and we opted for a tenth of the 719 minimum FC value found amongst all experiments. For the liquid, pellicles, 720 biofilms, spores and germinated spores experiments (Data S2, sheet 1) the 721 zeros were changed to 4.2×10^{-6} . For the bulk soil experiments (**Data S2, sheet** 722 2) zeros were changed to 1.14×10^{-5} . The logarithm of all FC values was 723 calculated and was also used to derive mean FC values. The logFC values were

724then used to calculate: (i) the standard deviation for all alleles across conditions725 $(\overline{SD};$ the standard deviation between logFC values observed per each allele in726replica experiments were averaged for all alleles measured in a given condition);727(ii) the standard deviation between synonymous codons within the same replica728experiment (deviations between logFC values of synonymous codons of the same729amino acid allele were calculated, averaged for all alleles in the same experiment,730and then for all replica experiments per condition).

731

732 **Defining the limits of neutrality**

733 From all conditions tested here, only in glucose plus ammonia the GudB 734 knockout had no growth effect (Fig. S1). Hence, this condition is largely neutral, 735 and the variation observed in FC values would primarily be the outcome of noise. 736 The standard deviation between 3 biological replicas was calculated per allele, 737 and these values spanned over the range of 0.002 to 0.199. We rounded this 738 number to 0.2. Thus, by the strictest measure, FC values between 0.8 and 1.2739 were classified as 'neutral'. Accordingly, FC ≤ 0.8 unambiguously assigned a 740 mutation as 'deleterious', and FC >1.2 as 'beneficial'.

741

742 **Genome sequencing**

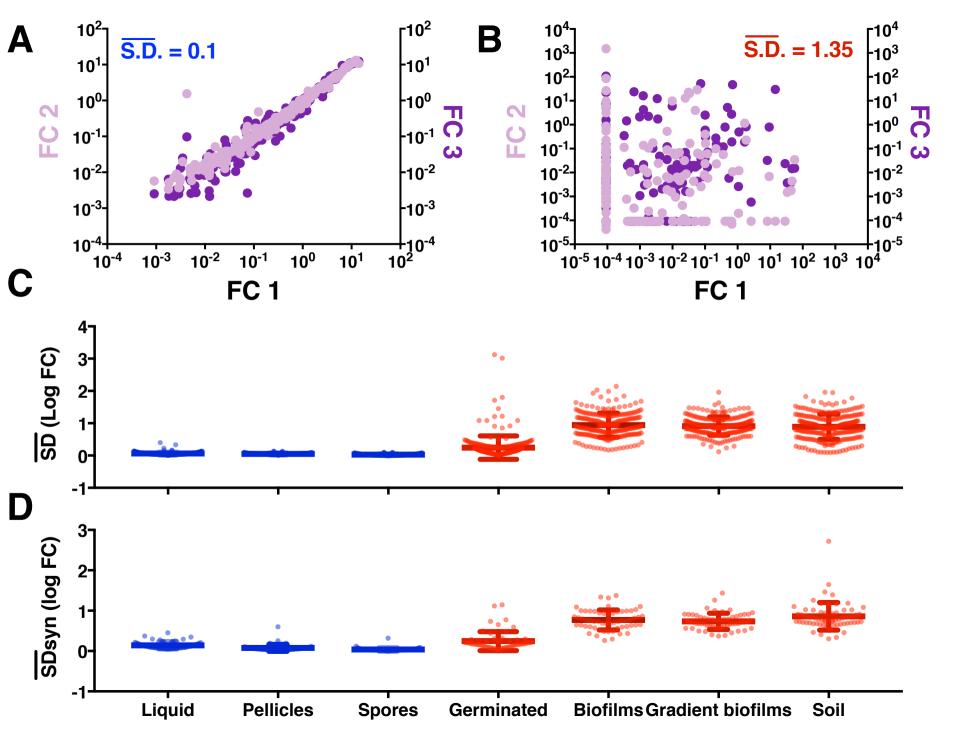
743 We sequenced the genomic DNA of all biofilm populations for which we had ≥ 1 744 µg of DNA after extraction (6 normal and 12 gradient biofilm). For comparison, 745 we also sequenced Initial mix populations 1 and 2, 6 Liquid and 4 pellicle 746 populations. The Illumina HiSeq2500 platform was used, with 2x125 base pairs 747 read length. We obtained a total of 300 million reads. The reads were assembled 748 using as reference the *B. subtilis* NCIB 3610 genome (NCBI Accession number: 749 CP020102). Overall, 95% of all reads were successfully mapped to the reference 750 genome with minimal coverage of x300 for all samples analyzed. The Breseq 751 program was used to identify genomic variants, including single nucleotide 752 polymorphisms (SNPs) and insertion-deletion polymorphisms (INDELs) ⁴⁴ (**Data** 753 **S3** & **Table S5**).

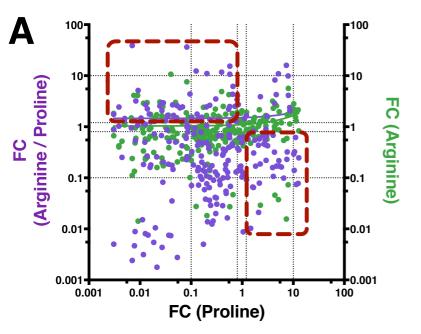
754

755 **Comparison of FC values and to GudB's natural sequence variability**

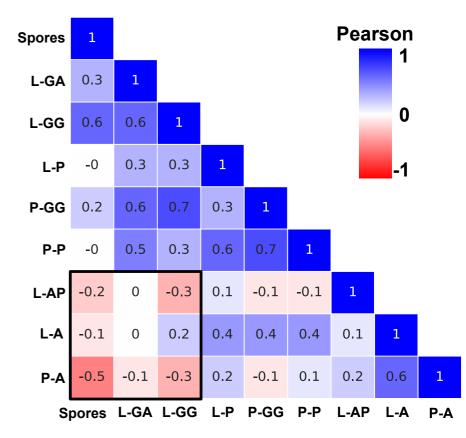
756 We examined whether the FC values for individual mutations, in individual 757 conditions, might predict whether or not a certain sequence exchange is 758 observed, or not, amongst the sequences of extant GDHs. To this end, we 759 constructed a number of different support vector machines (SVM) classification 760 models with a variety of kernels (such as linear, Gaussian, polynomial etc.). The 761 feature vector of each GudB allele was composed from the normalized FC values 762 from specific condition. The values from replica experiments of the highly 763 reproducible liquid conditions were averaged prior to training. Based on the 764 multiple sequence alignment containing 1013 GDH sequences, we divided the 765 GudB mutations in our dataset into 3 categories, which were then utilized as the 766 prediction labels: (1) mutations seen in less than 5 natural GDH sequences 767 (classified as 'not present', 66% of mutations), (2) mutations observed in 5 - 49 768 sequences ('rare', 19%) and (3) mutations present in \geq 50 sequences ('frequent', 769 15%). Introducing class weights into the loss function compensated the 770 unbalanced nature of the dataset. For each feature combination of a varied 771 length, we built an SVM classification model and assessed its accuracy using 3-772 fold cross validation. Additionally, in order to reduce noise, assuming that our 773 data belong to linear space, we extracted the first ten principal components of 774 the feature matrix and used them as the new feature vectors for a model 775 construction. To examine if our relatively high (>0.6) model accuracy was 776 distributed uniformly across different classes, for each model and genotype, we 777 recorded the predicted values during 3-fold cross-validation. Moreover, for each 778 condition combination, and for each kernel, we built 100 different models and 779 recorded the number of times each of the genotypes was predicted correctly.

780

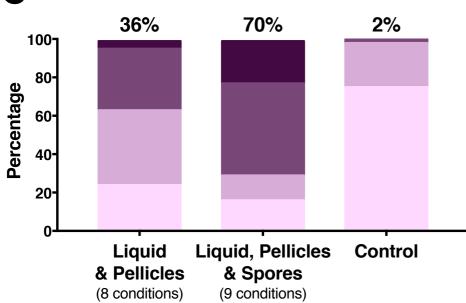




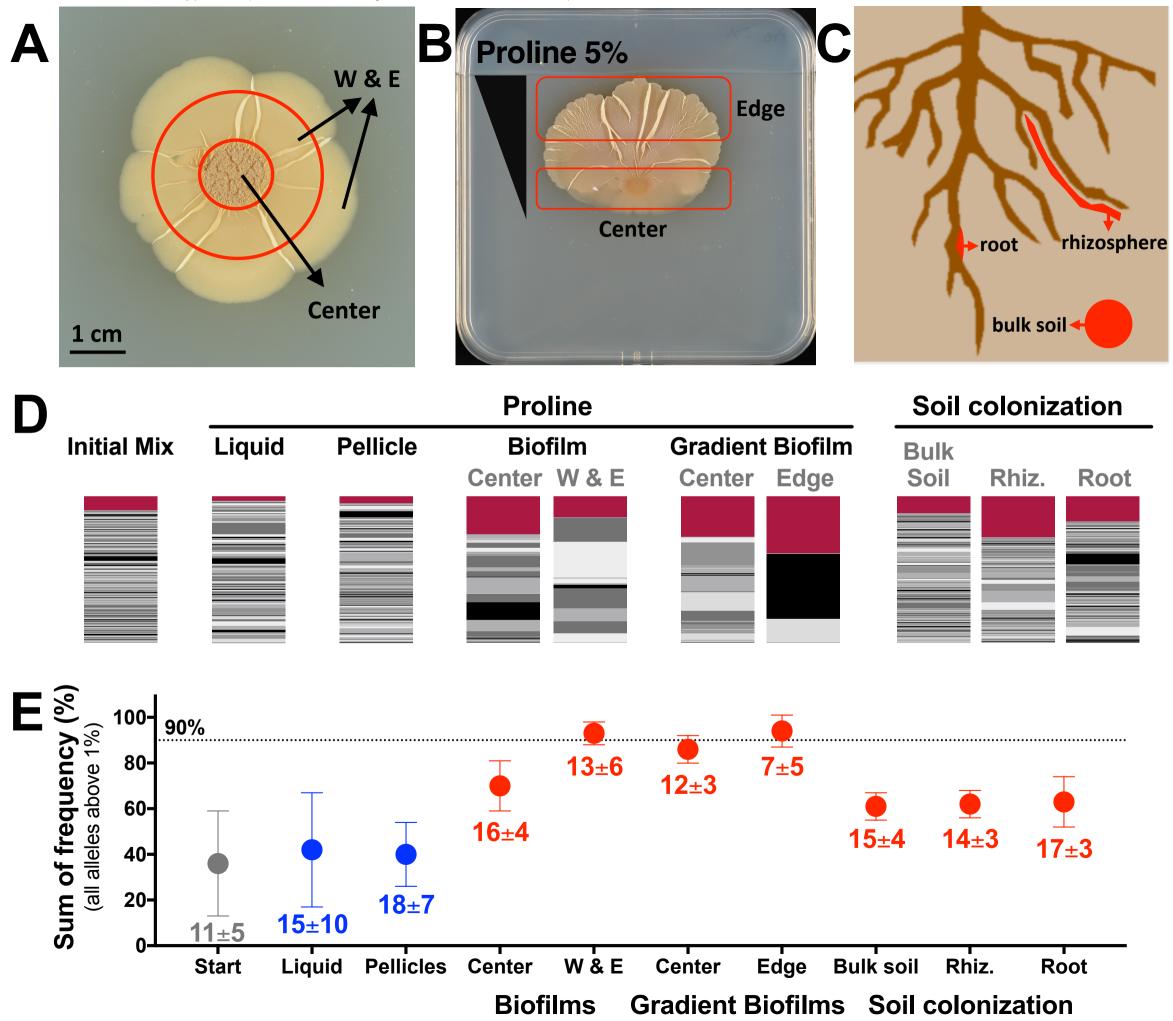
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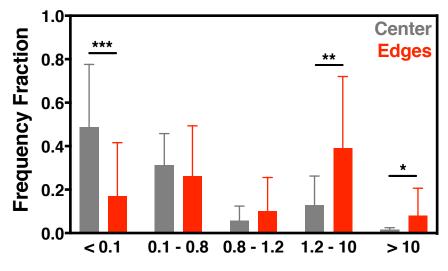


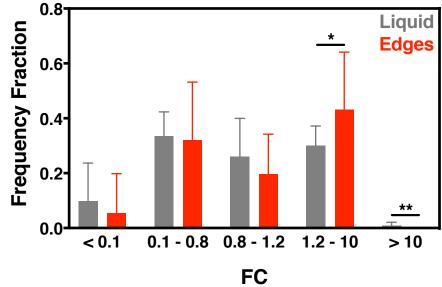
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