1 Polysaccharide breakdown products drive degradation-dispersal cycles of
2 foraging bacteria through changes in metabolism and motility
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Abstract

Most of Earth's biomass is composed of polysaccharides. During biomass decomposition, polysaccharides are degraded by heterotrophic bacteria as a nutrient and energy source and are thereby partly remineralized into CO_2. As polysaccharides are heterogeneously distributed in nature, following the colonization and degradation of a polysaccharide hotspot the cells need to reach new polysaccharide hotspots. Even though these degradation-dispersal cycles are an integral part in the global carbon cycle, we know little about how cells alternate between degradation and motility, and which environmental factors trigger this behavioral switch. Here, we studied the growth of the marine bacterium *Vibrio cyclitrophicus* ZF270 on the abundant marine polysaccharide alginate. We used microfluidics-coupled time-lapse microscopy to analyze motility and growth of individual cells, and RNA sequencing to study associated changes in gene expression. Single cells grow at reduced rate on alginate until they form large groups that cooperatively break down the polymer. Exposing cell groups to digested alginate accelerates cell growth and changes the expression of genes involved in alginate degradation and catabolism, central metabolism, ribosomal biosynthesis, and transport. However, exposure to digested alginate also triggers cells to become motile and disperse from cell groups, proportionally increasing with the group size before the nutrient switch, accompanied by high expression of genes involved in flagellar assembly, chemotaxis, and quorum sensing. The motile cells chemotax toward alginate hotspots, likely enabling cells to find new polysaccharide hotspots. Overall, our findings reveal the cellular mechanisms underlying bacterial degradation-dispersal cycles that drive remineralization in natural environments.

Importance

Polysaccharides, also known as glycans, are the most abundant form of biomass on Earth and understanding how they are degraded by microorganisms is essential for our understanding of the global carbon cycle and the storage and release of CO_2 by natural systems. Although group formation is a common strategy used by bacterial cells to degrade ubiquitous polymeric growth substrates in nature, where nutrient hotspots are heterogeneously distributed, little is known about how cells prepare for dispersal from an exhausted nutrient source and re-initiate degradation of new nutrient patches. By

quantifying growth, motility and chemotaxis of individual cells and comparing gene expression changes

when populations were exposed to either polysaccharides or their degradation products in the form of

digested polysaccharides, we show that bacterial cells alter their behavior when they experience a shift

from polymeric to digested polysaccharides: After cells form groups during growth on polymers, the

exposure to degradation products made cells motile, enabling dispersal from sessile cell groups and -

guided by chemotaxis - movement towards new polysaccharide hotspots. Our study sheds light on the

cellular processes that drive bacterial growth and behavior during carbon remineralization, an important

driver of CO_2 release from biomass in natural systems.

Introduction

Polysaccharides represent the largest fraction of biomass on Earth[1,2] and are constantly degraded and

remineralized by microorganisms. Polysaccharides, also known as glycans, are long chains of

monosaccharide units produced by cells for structural support (e.g., cellulose, chitin, and alginate) or

energy storage (e.g., starch or glycogen)[1]. Heterotrophic microbes obtain nutrients and energy from the

polysaccharide breakdown products. They often use exoenzymes, either secreted or anchored in the cell

membrane[2,3], to cleave these large polymers into smaller units that can be taken up by cells. The

formation of dense cell groups is observed during growth on polysaccharides by diverse bacteria,

including the soil- and gut-dwelling *Bacillus subtilis*[3], the oligotrophic fresh-water *Caulobacter*

crescentus[4], and several representatives of the copiotrophic[5,6] marine *Vibrio spp.*[7,8]. It has been

suggested that "cooperative" growth[3], where dense cell groups reduce diffusional loss of valuable

degradation products and secreted exoenzymes, represents a general principle in polysaccharide

degradation[4,8]. Yet, how bacteria subsequently disperse from exhausted polysaccharide sources and

navigate towards new polysaccharide hotspots remains poorly understood.

To study the cellular mechanisms that govern the switch between polysaccharide degradation and

dispersal toward new polysaccharide hotspots, we worked with the ubiquitous marine

Gammaproteobacterium *Vibrio cyclitrophicus* ZF270, a degrader of the prevalent marine

polysaccharide alginate[6,9]. Alginate is a linear polysaccharide that is produced by brown algae as a cell

wall component, as well as by certain bacteria[10]. It can be cleaved into oligomers by endo-acting alginate

lyases or into monomers by exo-acting alginate lyases (alginate lyases reviewed here[11]), namely β-D-

mannuronic acid and α-L-guluronic acid[10]. *V. cyclitrophicus* ZF270 is found predominantly in the large-

particle fractions of coastal water[12], can attach and form biofilms on the surfaces of multiple

polysaccharides including alginate[13], and was shown to secrete alginate lyases during alginate

degradation[8]. In this study, we used microfluidics coupled to automated time-lapse imaging to quantify

the growth dynamics, group formation, and motility of *V. cyclitrophicus* ZF270 at the single-cell level

under constant supply of either polymeric alginate or alginate degradation product in the form of

digested alginate, as well as upon a transition from alginate to digested alginate. Furthermore, we used

RNA-sequencing to compare the gene expression of *V. cyclitrophicus* ZF270 grown on alginate and

digested alginate. We found striking responses to the form of alginate in growth rate, group formation,

motility and chemotaxis, as well as in the expression of corresponding genes. Overall, our work

provides insights into the metabolic and cellular regulation that allows cells to forage in heterogeneous

nutrient-scapes through degradation-dispersal cycles.

Results

Extracellular break down of alginate delays population growth

To probe the phenotypic and metabolic regulation of bacterial cells during the progressive degradation

of a polysaccharide source, we developed an experimental system consisting of the marine bacterium

V. cyclitrophicus ZF270 growing on alginate, a highly abundant polysaccharide in the ocean. To mimic

the local nutrient environment during the colonization of a new polysaccharide source, we used 0.1%

weight per volume (w/v) algae-derived alginate in its soluble form (in the following also simply referred

to as "alginate"). Since commercially available alginate breakdown products of specific sizes are limited

and expensive, with only one monomeric component available at a considerable cost, we simulate an

advanced stage of polysaccharide degradation by supplying cells with digested alginate (0.1% (w/v)).

This digested alginate is prepared by treating alginate with a readily available endo-acting alginate lyase

(see Materials and Methods). This mimics an environment where degradation products like monomers

and oligomers become abundant through the action of extracellular alginate lyases. The commercially

available alginate lyase has been shown to produce alginate oligomers of progressively smaller size

over extended digestion periods[14,15]. We used Liquid Chromatography-Mass Spectrometry (LC-MS) to

analyze the composition of the digested alginate, which had been subject to 48 hours of digestion. We

found that digested alginate contained more monosaccharides than untreated alginate, which contained

more alginate molecules of higher molecular weight (Fig. S1).

Using this system, we first set out to investigate the growth dynamics of *V. cyclitrophicus* ZF270 in

well-mixed batch cultures containing either alginate or digested alginate as a sole carbon source (Fig.

S2A). We found that similar yields were reached in the alginate and digested alginate cultures as

indicated by optical density (Fig. S2B and C) and confirmed by colony counts (Fig. S3), suggesting that

the same amount of alginate supports a similar yield irrespective of its degree of depolymerization.

However, the onset of growth on alginate was delayed by about 7.5 hours compared to growth on

digested alginate (Fig. S2D and E). This is consistent with previous observations that in well-mixed

environments the lag time of bacteria growing on alginate can be reduced by the external

supplementation of alginate lyases[8,14]. Together, these findings indicate that growth on polysaccharides

such as alginate in well-mixed cultures is initially limited by the extracellular breakdown of the

polysaccharide.

Large cell groups form on alginate but not on digested alginate

To better understand how cells react to the changing degree of depolymerization of a polysaccharide

source during degradation, we investigated the growth of *V. cyclitrophicus* ZF270 at the level of single

cells on alginate and digested alginate. For this purpose, we grew the cells in microfluidic growth

chambers, described in detail by Dal Co et al.[16], where cells were provided with a continuous flow of

culture media containing either alginate or digested alginate as carbon source. We found that over 24

hours dense groups of more than 1000 cells formed on alginate, filling the entire microfluidic chamber

(Fig. 1A). In contrast, cells supplied with digested alginate grew in smaller groups that never exceeded

100 cells per chamber (Fig. 1B and C). Reconstructed cell lineages revealed that the large cell groups

on alginate formed because cells often did not disperse after division and thereby formed dense cell

groups originating from a single cell lineage (Fig. 1A). The lower cell density on digested alginate could

be caused by slower growth or by more cells leaving the chambers. As the maximum growth rate both

in bulk and on single cell level is similar in alginate and digested alginate (Fig. S2E and F), it is most

likely that the lower cell density in digested alginate is caused by cell dispersal. These observations

suggest that cells can modulate their propensity to form groups depending on the state of polysaccharide

degradation in their local environment. Similar observations were made with a different model system

(*Caulobacter crescentus* growing on the polysaccharide xylan)[14], indicating that group formation on

polymeric nutrient sources may be a general mechanism of bacteria that degrade polysaccharides

extracellularly.

Figure 1. Large cell groups form on alginate but not on digested alginate. Representative images at different time points of *V. cyclitrophicus* ZF270 cells growing in microfluidic chambers, described in detail by Dal Co et al.[16], with (A) alginate medium or (B) digested alginate medium, both in their soluble form (not visible). Cells are false-colored according to their lineage identities based on cell segmentation and tracking over 24 hours. Cells without identified progenitors are colored in dark blue. See Supplemental Videos S1 (alginate) and S2 (digested alginate) for time-lapse videos. (C) Cell numbers within microfluidic chambers supplied with alginate (orange) are substantially higher than cell numbers within microfluidic chambers supplied with digested alginate (blue) (Logistic growth regression for alginate: $R^2 = 0.99$, maximal number of cells = 1217-1564, k = 0.24-0.38 h^{-1}; for digested alginate: $R^2 = 0.86$-0.97, maximal number of cells = 47-100, k = 0.07-0.4 h^{-1}). Circles indicate the number of cells present at a given time point in each chamber ($n_{chambers} = 7$). Data for chambers with alginate originate from D'Souza et al.[8]. Lines are fits of a logistic growth regression line for each condition.

Transition from alginate to digested alginate triggers density-dependent dispersal of cells

To investigate the transition from the large cell groups formed on alginate to the small groups formed on digested alginate, we subjected *V. cyclitrophicus* ZF270 cells grown on alginate to a switch to digested alginate (Fig. 2A). Following the limited cell motility on alginate, this switch led to a rapid decrease in cell density within the growth chambers (Fig. 2B), presumably caused by cell dispersal. As we previously reported for other *Vibrionaceae* isolates, the growth rate of the cells on alginate was dependent on the local cell density: Initially the growth rate increased with cell density but then decreased at high cell densities, indicating that cell groups can benefit from the sharing of breakdown products generated by each other's exoenzymes, but also increasingly compete for nutrients[8]. This led us to investigate whether cell in larger groups, potentially experiencing stronger nutrient competition, might have a higher propensity to disperse after a switch to digested alginate than cells in smaller groups. We indeed found an inverse relationship between the number of cells at the time of the nutrient switch and the following change in cell number, meaning that cells in chambers with many cells were more likely to disperse than cells in chambers with less cells (Fig. 2C). Thus, during the transition from polysaccharides to degradation products, we found that the dispersal rate of cells depends on the size of the cell groups, likely through increased motility of cells in large groups.

Figure 2. Transition from alginate to digested alginate triggers density-dependent dispersal of cells. (A)

Representative time-lapse images of *V. cyclitrophicus* ZF270 cells (phase contrast microscopy) in microfluidic

growth chambers that were initially exposed to alginate and then switched to digested alginate. (B) Number of

cells in different chambers over time, each chamber indicated by a unique color (n = 8). The carbon source is

indicated by the colored background (orange: alginate; blue: digested alginate). See Supplemental Video 3 for a

time-lapse video. (C) Negative relationship between the difference in number of cells within the microfluidic

growth chambers before and after the nutrient switch. Each circle represents one growth chamber with colors

corresponding to (B), and the line depicts a linear regression fit ($R^2 = 0.85$, slope = -0.81, *p*-value < 0.001).

Cells growing on digested alginate are more motile and chemotax towards new polysaccharide

sources

To investigate whether increased cell motility of *V. cyclitrophicus* ZF270 underlies the dispersal of cells

observed after a shift to digested alginate, as also previously observed for *C. crescentus* cells on the

monosaccharide xylose[4,14], we quantified the motility of single cells supplied with either alginate or

digested alginate. We found that the single-cell swimming speed as well as the swimming distance were

significantly larger for cells supplied with digested alginate compared to cells supplied with alginate,

and that a larger fraction of cells was motile (Fig. 3). This confirmed increased motility as a cellular

response to the exposure to degradation products in the form of digested alginate.

To understand if the motile cells are directed by chemotaxis, we measured the chemotactic strength

towards alginate and digested alginate using the In Situ Chemotaxis Assay (ISCA)[17,18]. We found *V.*

cyclitrophicus ZF270 to significantly chemotax toward alginate (chemotactic index $I_c > 1$) but not

toward digested alginate (Fig. 3F). This suggested that the increase in motility is accompanied by

chemotaxis toward alginate. Overall, these findings suggest that once the breakdown of a

polysaccharide source makes breakdown products available in the local environment, a fraction of cells

becomes motile and disperses. These motile cells are guided by chemotaxis towards new polysaccharide

sources, allowing them to start a new degradation-dispersal cycle.

Altered gene expression in central carbon metabolism, enzyme production, secretion and

transporters, motility, and quorum sensing underlies the late-stage alginate degradation and cell

dispersal

Next, we sought to elucidate the molecular mechanisms underlying the observed phenomenological

disparities between cells cultivated on alginate and digested alginate. Due to the challenge of generating

knock-out mutants in natural isolates, we used transcriptomics to investigate the differentially expressed

genes of *V. cyclitrophicus* ZF270 under these respective conditions. To obtain a high-quality reference

genome of *V. cyclitrophicus* ZF270, we sequenced and assembled a new reference genome using

combined short and long read sequencing (BioProject PRJNA991487). We then grew cultures on either

alginate or digested alginate until mid-exponential phase. To understand which cellular functions were

affected by the expression changes, we first performed differential gene expression analysis using the

software DESeq2. Here, genes exhibiting a log2 fold expression change greater than 0.5 or smaller than

-0.5 between the two conditions, with a Benjamini-Hochberg(BH)-adjusted *p*-value below 0.01, were

considered to be differentially expressed (Table S1). Next, we investigated which KEGG categories

Figure 3. Cells are more motile on digested alginate than alginate and show chemotaxis towards alginate.

(A and B) Spatial trajectories of cells supplied with (A) alginate or (B) digested alginate in representative

microfluidic growth chambers are shown. White circles mark the starting point of each trajectory and colored

lines mark the individual trajectories. (C) Distributions of the mean single-cell swimming speeds (Nested t-test,

p-value < 0.0007, t = 4.803, df = 10, n_{cells} = 86 vs 375 in $n_{chambers}$ = 5) are shown. (D) Distributions of cell

displacement over the course of a trajectory (Nested t-test test, p-value < 0.0131, t = 4.39, df = 10, n_{cells} = 86 vs

375, and $n_{chambers}$ = 5) are shown. In (C) and (D) the red horizontal lines indicate the mean while black lines depict

the 25th and 75th quartiles of the distribution. (E) The mean fraction of motile cells in each chamber, where motile

cells are defined as cells with displacement greater than 1 μm (Mann-Whitney test on the means of five growth

chambers, p-value = 0.034). In C, D, and E, each chamber was considered as an independent replicate. (F)

Chemotactic index (I_C) quantified by In Situ Chemotaxis Assay (ISCA) (Tukey multiple comparisons of means,

95% family-wise confidence levels as error bars, *p*-value < 0.05, n = 3). Asterisks indicate statistically significant

differences. See Supplemental Videos S4 and S5 for time-lapse videos of swimming cells.

were enriched in either genes with increased or decreased expression via Gene Set Enrichment Analysis

(GSEA)[19] (Table S2) and found nine categories significantly enriched in genes with increased gene

expression on digested alginate, and three categories significantly enriched in genes with decreased

gene expression (Fig. 4A).

Cells growing on digested alginate showed expression changes in parts of the central metabolism and

translation, compared to cells growing on alginate. Specifically, two pathways of the central metabolism

were significantly enriched in genes with lower expression on digested alginate ("Valine, leucine and

isoleucine biosynthesis" and "Propanoate metabolism" with a negative normalized enrichment score

(NES), Fig. 4A, S4A and B, Table S3 and S4). We also found the gene set that encodes ribosomal

proteins significantly enriched with genes highly expressed on digested alginate (positive NES, Fig. 4A

and S4C, Table S5). This implies that cells invest proportionally more of their transcriptome into the

production of new proteins, a sign of faster growth[20,21]. Both observations likely relate to the different

growth dynamics of *V. cyclitrophicus* ZF270 on digested alginate compared to alginate.

The expression of transporters and secretion systems was generally increased in cells growing on

digested alginate, compared to cells growing on alginate. This includes genes with a function in

"Bacterial secretion systems", namely secretion systems I to VI, which mediate protein export through

the inner and outer membranes of Gram-negative bacteria (Fig. 4A and S4D, Table S6). Notably, eleven

of the thirteen General Secretion Pathway (GSP) genes (part of the type II secretion system) showed

1.1 to 3.4-fold increased expression levels on digested alginate (*p*-values < 0.01, Fig. 4A, Table S8).

The GSP is known to facilitate secretion of various extracellular enzymes like chitinases, proteases, and

lipases[22,23], therefore, the positive enrichment on digested alginate may be linked to the export of

extracellular alginate lyases. The KEGG BRITE category "Secretion system" was also enriched on

digested alginate (Fig. 4A and S4D, Table S7). It contains additional genes involved in protein export,

of which especially genes of type IV pilus assembly, conjugal transfer pilus assembly, and MSHA pilus

biogenesis were highly expressed on digested alginate, implicating them as interesting subjects to

further research in their role in protein secretion and cell attachment and detachment during

degradation-dispersal cycles of polysaccharide-degrading bacteria. Also, the gene set of transporters

and in particular ABC transporters were positively enriched (Fig. 4A, Table S10 and S11). The latter

showed enrichment especially in genes related to saccharide, iron, zinc, and phosphate transport (Fig.

S4E), which suggests that cells growing on digested alginate invest proportionally more of their

transcriptome into uptake of not only saccharides but also essential nutrients like iron, zinc, and

phosphate, which may become growth-limiting when degradation products are abundantly available.

Both motility and quorum sensing genes increased in expression in cells growing on digested alginate.

The positive enrichment of the gene set containing bacterial motility proteins matched the increase in

motile cells that we observed in Figure 3E (Fig. 4A, Table S12). The set of quorum sensing genes was

also positively enriched in cells growing on digested alginate (Fig. 4A and S4F, Table S13). Quorum

sensing is known to control biofilm formation in *Vibrio cholerae*[24,25] and may orchestrate the density-

dependent dispersal in the presence of degradation products, though the particular signaling cues remain

to be uncovered. The strong cellular response to the degree of alginate depolymerization was also

emphasized by the finding that transcription factor genes were enriched among genes with decreased

expression on digested alginate (Fig.4, Table S14). The set of genes associated to KEGG's beta-lactam

resistance category was enriched in genes with increased expression on digested alginate. However,

most genes in this gene set are also associated to the KEGG pathways "Transporters", "Quorum

sensing", "Chromosome and associated genes", and "Peptidoglycan biosynthesis" and thus the

enrichment likely reflects expression changes in these categories and may relate to the difference in

growth dynamics (Fig. 4A and S4G, Table S15). Overall, the observed gene expression changes provide

insights into the molecular mechanisms underlying the substantial adaptations of *V. cyclitrophicus*

ZF270 cells to the polymeric and digested form of alginate.

Figure 4. Twelve functional gene sets are enriched in genes with increased or decreased expression in cells grown on digested alginate. Gene set enrichment analysis (GSEA) with (A) all KEGG pathways and KEGG BRITE categories as gene sets or with (B) a custom alginate utilization, flagellar assembly, and flagellum-driven chemotaxis gene set was performed comparing the gene counts of the transcriptome of *V. cyclitrophicus* ZF270 cultures grown on digested alginate and alginate (six replicates each). Gene sets with a positive enrichment score were enriched with genes with higher expression in cells grown on digested alginate relative to cells grown on alginate (BH-adjusted *p*-value < 0.05), whereas gene sets with negative enrichment scores were significantly enriched with genes with decreased expression on digested alginate. The number in brackets indicates the number of genes with unique K number per gene set (A) and the number of genes per gene set (B) within the *V. cyclitrophicus* ZF270 genome.

Growth on digested alginate is associated with increased expression of alginate catabolism, flagellar motility and chemotaxis genes

For a more fine-grained understanding of the impact of the form of alginate on the alginate catabolism, we investigated specifically the expression of genes involved in alginate degradation, uptake, and

catabolism. In the closed genome of *V. cyclitrophicus* ZF270, we identified genes that encode CAZymes

responsible for alginate degradation, namely alginate lyase genes from the PL6, PL7, PL15, and PL17

family. We identified homologs of alginate transporters (porin *kdgM*, symporter *toaA*, *toaB*, and *toaC*)

and metabolic enzymes that shunt into the Entner-Doudoroff pathway (DEHU reductase genes *dehR*,

kdgK, and *eda*), based on the known genes in the alginate degradation pathway of *Vibrionaceae*[26,27].

We found that the expression of most of these genes increased significantly on digested alginate relative

to alginate (Fig. 4B, Fig. 5A, S8A, Table S16). Surprisingly, also the expression of most alginate lyase

genes increased on digested alginate, especially the expression of secreted alginate lyase genes. This

indicates the production of "public" exoenzymes despite the abundance of monomeric and oligomeric

degradation products in the digested alginate medium.

The increased motility of cells observed upon exposure to digested alginate (Fig. 3, Fig. 4A) led us to

evaluate the expression of motility- and chemotaxis-associated genes across digested alginate and

alginate treatments. As flagella are the main mode of motility in the genus *Vibrio*[28,29], we focused on

the expression of flagella-related genes. In the genome of *V. cyclitrophicus* ZF270 we found a gene

cluster that encodes most genes involved in flagellar assembly and that was flanked by chemotaxis

genes (hereon called flagellar locus, Fig. S8B, and Table S17). Overall, 21/34 flagellar locus genes were

differentially expressed (log2 fold change > 0.5, BH-adjusted *p*-value < 0.01) and the majority (90%)

of these differentially expressed genes showed increased expression on digested alginate (Fig. 4B). The

flagellar genes *flgA*, *fliC*, and *fliH* and the chemotaxis gene *cheW* showed the strongest overexpression

(5.4, 3.8, 2.0, and 2.6-fold, respectively). The expression of flagellar biosynthesis genes in *Vibrionacaea*

occurs by a cascade of gene expression of four classes of genes (Class I - IV)[30]. We found the expression

of the master regulator of the flagellar biosynthesis regulon, the Class I gene *flrA*, to be 1.6-fold

increased on digested alginate (BH-adj. *p*-value 3e-19)[31]. The Class II regulatory genes *flrBC*,

controlling Class III genes, and *fliA*, controlling Class IV genes, showed 1.5-fold and 1.3-fold increased

expression on digested alginate (BH-adj. *p*-value 1e-12 for *flrB*, 2e-39 for *flrC*, 6e-15 for *fliA*) (Fig.

5B)[29,32]. These findings suggest that the increased phenotypic motility observed on digested alginate

(Fig. 3) is related to the upregulation of flagellar biosynthesis genes. Additionally, the expression of the

flagellum filament, encoded by *fliC* genes of the flagellar locus, was partially increased: *flaD* and *flaC*

expression were increased by 3.8 and 1.6-fold, whereas the *flaA* gene was not significantly differentially

expressed (BH-adj. *p*-value 7e-37, 2e-19, and 1e-02, respectively) (Fig. 5B, Table S6). This suggests

that cells grown on digested alginate have a *flaD*-rich filament composition, which has been shown to

alter the swimming and adhesion characteristics of bacterial cells[33,34]. Lastly, we found that most genes

involved in chemotaxis and located in the flagellar locus are highly expressed in cells grown on digested

alginate (Fig. 5C, Fig. 4B) and likely drive the chemotactic activity of *V. cyclitrophicus* ZF270 during

dispersal. Overall, our findings elucidate that cellular responses upon exposure to degradation products

manifest in increased expression of genes involved in extracellular alginate breakdown and alginate

catabolism as well as flagellar assembly and chemotaxis.

Figure 5. Digested alginate increases expression of genes involved in alginate degradation, uptake and

catabolism, as well as flagellar assembly and chemotaxis. Genome-wide differential expression analysis where

the log2 fold changes of gene expression on digested alginate compared to alginate is shown for (A) alginate

lyases (*PL6, PL7, PL15, PL17*), transporters (porin *kdgM*, symporter *toaB*, symporter *toaC*), and metabolic

enzymes shunting into the Entner-Doudoroff pathway (DEHU reductase *DehR*, *kdgK*, *eda*), (B) genes of the

flagellar locus associated with flagellar assembly and (C) adjacent chemotaxis genes. Genes displayed in (B) and

(C) are part of the KEGG pathways "Bacterial motility proteins" and "Bacterial chemotaxis". Differential

expression analysis was performed to compute the Benjamini-Hochberg-adjusted Wald test *p*-value ("BH-adj. *p*-

value", text color and box outline color) and log2 fold change (box fill color) for each gene (box). For better

visibility, genes that exhibited a log2 fold gene expression change greater than 1 (i.e., doubling of expression) or

less than -1 (i.e., halving of expression) are designated maximum intensity of red or blue, respectively. Genes with

BH-adj. *p*-value smaller than 0.01 were considered significantly differentially expressed. In (A), the location of

the gene products was based on Figure 1 of Wargacki et al.[26] with the exception of the alginate lyases (*PL6*, *PL7*,

PL15, *PL17*) which were placed based on their signal peptides (S: extracellular, LS: membrane-embedded, none:

cytosolic). In (B) and (C) the gene location and depiction were based on the KEGG pathway "Flagellar assembly"

(map02040), "Bacterial chemotaxis" (map02030), and Figure 3 of Rajagopala et al.[35]. Genes without known

cellular location were omitted here but displayed in the genomic architecture in Fig. S5. Arrow: activation; dashed

arrow: modification; "flat" arrow: inhibition; OM: outer membrane; PM: periplasm; IM: inner membrane; PL:

polysaccharide lyase family; *kdgM*: oligogalacturonate-specific outer membrane porin; *toaABC*: oligoalginate

symporter; DEH: 4-deoxy-L-erythro-5-hexoseulose uronic acid; *dehR*: DEH reductase; KDG: 2-keto-3-deoxy-

gluconate; *kdgK*: KDG kinase; KDPG: 2-keto-3-deoxy-6-phosphogluconate; *eda*: KDG-6-phosphate aldolase;

GAP: glyceraldehyde 3-phosphate; ED: Entner-Doudoroff; ns: not significant, i.e. BH-adj. *p*-value > 0.01.

Discussion

On Earth organic carbon is mostly present in the form of polysaccharides[1,2], which are often in a

particulate state and form a heterogeneous resource landscape. Over the last years, the study of

extracellular bacterial degradation of polysaccharides has revealed that bacterial growth on

polysaccharides increases with increased cell density, enabling cells to benefit from the exoenzymes

and extracellular degradation products of surrounding cells otherwise lost to diffusion ("cooperative

growth")[36–38]. However, cells have been observed not only to aggregate on polysaccharide sources, but

also to leave them before the source is depleted[13,38]. This has thus far been interpreted with optimal

foraging theory, which weighs the diminishing returns of harvesting resources from a nutrient hotspot

against the predation and opportunity costs which ensue[39,40], but the cellular mechanisms that drive the

cell dispersal remain unclear.

Our work links these observations and connects them to the cellular mechanisms that underlie the

degradation-dispersal cycles of bacterial degraders, which we see as basal drivers of the biogeochemical

processing of polysaccharides in heterogeneous nutrient-scapes. When bacteria encounter a new source

of biomass, their local environment likely contains few mono- or oligosaccharides but is rich in

polysaccharides which usually require extracellular breakdown (Fig. 6, "Finding a new nutrient

source"). General concepts of how bacteria recognize the presence of large biopolymers remain elusive,

but it was proposed that "sentry" enzymes are constitutively expressed at a basal level to cleave mono-

or oligosaccharides from polysaccharides, which cells can take up and which prime their metabolism

for the degradation of the respective polysaccharide[41,42]. It is not known yet how widespread the concept

of sentry enzymes may be, but the observation of a constitutively expressed PL7 and PL15 family

alginate lyase gene in *Z. galactanivorans*[41] is mirrored in our work by the constant expression of an

extracellular PL7 family alginate lyase gene, which may act as sentry enzyme that helps to initiate

alginate degradation when cells encounter alginate.

It has been previously found that cells grow slowly until a critical cell density for efficient cooperative

polymer breakdown is reached[36], observed here as the formation of large cell groups. In this stage, cells

start to benefit from the diffusing degradation products and exoenzymes of cells in their local

environment and reach their maximal growth rate (Fig. 6, "Early-stage polysaccharide breakdown").

We found that cells grown on degradation products reached their maximal growth earlier and showed

increased expression of ribosomal biosynthesis, enzyme secretion, especially of secreted alginate

lyases, transporters, quorum sensing and expression changes in the central carbon metabolism. The

secretion of secreted alginate lyases might seem surprising and wasteful in a monomer-rich

environment. We speculate that enhanced enzyme secretion supports the continued growth of detached

bacteria's offspring that remain on the particle, as observed in the case of *V. splendus* 1AO1 during

chitin particle degradation[43]. Another reason may be that cells primarily rely on intracellular

monosaccharide levels to trigger the upregulation of genes associated with polysaccharide degradation

and catabolism, as seen in *E. coli* across various carbon sources[44,45], unable to modulate the expression

of degrading enzymes based on the extracellular presence of monomers.

We show that cells respond to the exposure to degradation products with dispersal from dense cell

groups by means of increased flagellum-driven swimming (Fig. 6, "Late-stage polysaccharide

breakdown"), decreasing their local cell number. This finding matches with previous observations of

cells leaving biopolymer particles before they are depleted[13,38]. A plausible explanation for this density-

dependent dispersal is that cells in larger groups compete with each other for nutrients and space, while

not profiting from cooperative degradation anymore due to the abundance of degradation products.

Motility has also been shown to increase the encounter rate of cells with sources of nutrients[46,47],

suggesting motility as a strategy that allows cells to escape from the ensuing competition. Previous

work in *Caulobacter crescentus* demonstrated that a flagellum knock-out mutant formed larger cell

groups, resulting in reduced growth rates due to intercellular competition[4,14]. While it would be

interesting to study non-motile mutants of *V. cyclitrophicus* ZF270, the non-tractability of natural

isolates makes direct tests of molecular mechanisms difficult. Additionally, subjecting cells separately

to the two monomeric units of alginate or oligomers of defined size could improve our understanding

of the specific molecules that trigger motility, but this was experimentally not feasible. We also found

that the presence of degradation products increased the expression of chemotaxis genes and detected

that cells chemotaxis toward polymeric alginate, but not digested alginate, analogous to previous

findings on chitin[46–48]. Even though this may first seem counterintuitive, it makes sense in the light of

the dispersal process: chemotaxis towards degradation products would prevent cells from leaving a

nutrient source, whereas chemotaxis towards polymers may increase encounters with a fresh nutrient

source[49] (Fig. 6, "Finding a new nutrient source"), on which cells can re-initiate degradation. This

strategy allows cells to alternate between degradation and dispersal to acquire carbon and energy in a

heterogeneous world with nutrient hotspots[50–53].

Figure 6. Bacterial growth and regulation on patches of polysaccharides. Based on our findings we propose

a conceptual model where degradation of polysaccharides proceeds after the encounter of a polymer source by

bacterial cells that have a basal exoenzyme production of "sentry" enzymes ("Encounter of a new carbon source").

This phase is succeeded by group formation, which enables cells to benefit from exoenzymes of neighboring cells

and diffusing degradation products ("Early-stage polysaccharide breakdown"). The following phase includes

cooperative extracellular degradation of the polysaccharide source, further increasing the concentration of

available degradation products. These degradation products trigger the overexpression of alginate degrading,

importing, and catabolizing enzymes, ensuring swift polysaccharide degradation ("Late-stage polysaccharide

breakdown"). The increased pool of breakdown products also cues flagellar swimming and chemotaxis in a

subpopulation of cells towards new polysaccharide sources, thus restarting the cycle of degradation of polymeric

carbohydrates by heterotrophic bacteria. Cells and molecules are not drawn to scale. Dark red pie symbols:

intracellular and extracellular polysaccharide-degrading enzymes; orange shading: a polymeric carbon source;

blue shading: monomeric or oligomeric degradation products.

Conclusion

The heterogeneous landscape of polysaccharide hotspots in natural systems requires bacteria to effectively break down polymeric carbohydrates as well as readily ensure dispersal to new nutrient hotspots. Our findings show that the degree of depolymerization of the polysaccharide influences this decision, altering the growth dynamics, metabolic activity, and motility of cells. Our study also contextualizes the surprising finding that foraging bacteria majorly leave polysaccharide particles before the last third of the particle is consumed[38]. Dispersal from a partially degraded carbon source may serve several purposes: (i) escaping competition that ensues within large cell groups, (ii) ensuring the spread of a part of the clonal population to new environments as bet hedging strategy[40,54], here guided by chemotaxis towards new nutrient hotspots, (iii) preventing whole populations degrading a sinking marine particle or a deposited sediment particle to be buried in depth where nutrient hotspots become sparse[38,55], and/or (iv) increase the genetic variation in bacterial populations[40]. However, dispersal may also occur when a nutrient source offers a surplus of carbon while other essential nutrients become limiting, as the increased expression of iron, zinc, and phosphate transporters in cells grown on digested alginate suggested. These findings emphasize that metabolic molecules can also act as triggers of dispersal, expanding upon the current perspective of dispersal in biofilms as a reaction to dispersal cues like NO, signalling molecules, nutrient starvation, and oxygen starvation[56]. The study of bacterial motility on increasingly complex biomass particles will reveal the role of the nutrient composition of the present nutrient hotspot on the bacterial decision-making. Overall, these new insights into the cellular mechanisms and regulation that drive degradation-dispersal cycles contribute to our understanding of the microbially driven remineralization of biomass, and factors that modulate this process. The open questions of how bacteria sense polysaccharides in their environment, which cell signaling pathways integrate the presence of degradation products in the cellular decision-making of degradation and dispersal, and to what extent cell populations coordinate this decision, present an exciting avenue of further research.

Materials and Methods

Bacterial strains, media and growth assays

Vibrio cyclitrophicus ZF270 (available through Culture Collection Of Switzerland; Accession number: 2043) cells were cultured in Marine Broth (DIFCO) and grown for 18 hours at 25 °C. Cells from these cultures were used for growth experiments in Tibbles Rawling (TR) salts minimal medium[57,58] containing either 0.1% (weight/volume) algae-derived alginate (referred to as "alginate") (Sigma Aldrich, CAS-number 9005-38-3) or 0.1% (weight/volume) digested alginate. The digested alginate was produced by enzymatically digesting 2% alginate with 1 unit ml^{-1} of alginate lyase (Sigma Aldrich, CAS-number 9024-15-1) at 37 °C for 48 hours. Carbon sources were prepared in nanopure water and filter sterilized using 0.40 µm Surfactant-Free Cellulose Acetate filters (Corning, USA). Well-mixed batch experiments in alginate or digested alginate medium were performed in 96-well plates (Greiner Bio) and growth dynamics were measured using a microwell plate reader (Biotek, USA). Assays were initiated as described previously[59]. Briefly, 1 ml from a culture grown for 18 h on Marine broth was centrifuged at 5000 g in 1.5 ml microfuge tubes for 5 minutes. The supernatant was discarded and the cell pellet was subjected to two rounds of washing with the basal TR salts medium. The cell-pellet was then resuspended in 1 ml of TR salts medium and 5 µl of this suspension inoculated into 195 µl TR medium with either carbon source ($\sim 10^5$ colony forming units (CFUs ml^{-1}) in a 96-well plate (Greiner Bio). The optical density (600 nm) was then measured every 15 minutes for 40 hours. All measurements had six biological replicates.

Alginate oligosaccharide measurements

Oligosaccharide measurements were performed using liquid chromatography time of flight mass spectrometry (LC-QTOF-MS). Samples were prepared by diluting 1:20 in milliQ water and 5 µL of sample was injected per measurement. Chromatographic separation was performed using an Agilent 1290 stack, using an Agilent HILIC-Z column (2.7 µm particles, 2.1 x 50 mm). Mobile phase A contained 10% acetonitrile (Fisher Scientific) and 0.1% medronic acid (Agilent), and Mobile phase B contained 90% acetonitrile and 0.1% medronic acid. The separation was performed as follows: Mobile phase B 100% for 1 minute, gradient to 30% phase B over 3 minutes, 30% phase B for 30 seconds, and

equilibration of 100% phase B for 5 minutes. The flow rate was 400 μL min^{-1} at 30 °C. Samples were

measured using an Agilent 6520 mass spectrometer in negative mode, in 4 GHz high-resolution mode.

Data analysis was performed in Agilent Quantitative Analysis software.

Microfluidics and time-lapse microscopy

Microfluidic experiments and microscopy were performed as described previously[14,16,60]. Cells were

imaged within chambers of a PDMS (Sylgard-Dow) microfluidic chip that ranged in size from 60-120

× 60 × 0.56 μm (*l* × *b* × *h*). Within these chambers, cells can attach to the glass surface and experience

the medium that diffuses through lateral flow channels. Imaging was performed using IX83 inverted

microscope systems (Olympus, Japan) with automated stage controller (Marzhauser Wetzlar,

Germany), shutter, and laser-based autofocus system (Olympus ZDC 2). Chambers were imaged in

parallel on the same PDMS chip, and phase-contrast images of each position were taken every 8 or 10

minutes. The microscopy unit and PDMS chip were maintained at 25 °C using a cellVivo microscope

incubation system (Pecon GmbH).

Motility assays

Cells were grown for 10 hours in Marine Broth (DIFCO) after which 10 μl of culture was used to

inoculate culture tubes (Greiner) containing 5 ml of TR medium with either 0.1% alginate or 0.1%

digested alginate. After 6 hours of growth at 25 °C, 2 μl of cell suspension was inoculated into

microfluidic growth chambers. Cells within six replicate chambers were then imaged with the phase-

contrast channel at a high frame rate (125 Hz, i.e., frames s^{-1}) using the same microscopy setup described

above.

Chemotaxis assays

To assess whether polymeric alginate and digested alginate attract *Vibrio cyclitrophicus* ZF270, we

used the *In Situ* Chemotaxis Assay[17,18] (ISCA), a microfluidic device consisting of a 5 × 5 array of

microwells that can be individually loaded with solutions of different chemicals (110 μl each). Once

the ISCA is deployed in an aqueous environment, the chemicals diffuse out of the wells through a small

port, creating chemical gradients which will guide chemotactic bacteria inside the wells of the

device[17,18]. *Vibrio cyclitrophicus* ZF210 was plated on Marine Agar (BD Difco) from a glycerol stock

and grown for 16 h at 27 °C. A single colony was then incubated in 10% Marine Broth (BD Difco) in

0.22-μm filtered artificial seawater (Instant Ocean, Spectrum Brands) and grown overnight at 27 °C and

180 rpm. The culture was diluted down to 1×10^6 cells ml^{-1} in 0.22 μm filtered artificial seawater (Instant

Ocean, Spectrum Brands) to perform the chemotaxis experiment. Both chemoattractants (alginate and

digested alginate) were diluted in sterile seawater (35 g l^{-1}; Instant Ocean, Spectrum Brands) at a final

concentration of 0.1% and then filtered with a 0.2 μm filter (Millipore) to remove particles and potential

contaminants. Within the ISCA, one full row of five wells was used per chemoattractant as technical

replicates. The chemoattractants were injected in triplicate ISCA with a sterile 1 ml syringe (Codau)

and needle (27 G, Henke Sass Wolf). A last row containing 0.2 μm-filtered seawater acted as negative

control accounting for cells swimming in the device by random motility only. Experiments were

conducted by incubating the ISCAs for 1 h in the diluted *Vibrio cyclitrophicus* ZF270 culture. Upon

time completion, a sterile syringe and needle were used to retrieve the content of the wells and

transferred to 1 ml microfuge tubes resulting in a pooling of a row of 5 wells containing the same

sample. Sample staining was performed with SYBR Green I (Thermo Fisher) and the chemotactic

response was quantified by counting cells using flow cytometry. The strength of the chemotactic

response was determined by the mean chemotactic index (I_C), defined as the ratio of the number of cells

found in each chemoattractant to the number of cells in control wells containing filtered seawater (so

that attraction corresponds to $I_C > 1$).

Culturing and harvesting cells for transcriptomics

Cells were grown for 18 hours in Marine Broth (DIFCO) after which 1 ml of culture was centrifuged at

5000 g in 1.5 ml microfuge tubes for 5 minutes. The supernatant was discarded and the cell pellet was

subjected to two rounds of washing with the basal TR salts medium. The cell pellet was then

resuspended in 1 ml of TR salts medium and 250 μl of this suspension were used to inoculate 100 ml

flasks (Schott-Duran) containing 10 ml of TR medium with either 0.1% alginate or 0.1% digested

alginate. This was done in parallel for six flasks. Once cultures in the flasks reached mid-exponential

phase (10 hours and 15 hours after inoculation for digested alginate and alginate, respectively) and had

approximately the same OD (0.39 for digested alginate and 0.41 for alginate), 2 ml of cultures were

harvested for RNA extraction. Samples were stabilized with the RNprotect reagent (Qiagen) and RNA

was extracted using the RNeasy mini kit (Qiagen).

Sequencing and gene annotation of Vibrio cyclitrophicus ZF270

Long read sequencing using the Oxford Nanopore Platform (Long read DNA sequencing kit) and short

read sequencing using the Illumina platform (Illumina DNA Prep kit and IDT 10bp UDI indices, and

sequenced on an Illumina NextSeq 2000 producing 2x151bp reads) was performed by the Microbial

Genome Sequencing Center, Pittsburgh, USA (MiGS), to create a new closed reference genome of *V.*

cyclitrophicus ZF270 (BioProject PRJNA991487). Annotation of this genome was done with RASTtk

(v2.0, Rapid Annotation using Subsystem Technology tool kit[61]). Additionally, KEGG Ontology

identifiers ("K numbers") were annotated with BlastKOALA (v2.2)[62]. Dedicated annotation of alginate

lyase genes was performed by homology search for proteins belonging to the PL5, PL6, PL7, PL14,

PL15, PL17, PL18, PL31, PL36, or PL39 family[63] by dbCAN2 (v9.0)[64]. Enzymes for alginate transport

and metabolism were identified by BLASTn-search[65,66] of gene sequences of *Vibrio splendidus* 12B01,

which were previously identified as minimum genetic prerequisites for alginate utilization and enabled

alginate degradation when cloned into *E. coli*[26].

Location prediction of alginate lyases

Signal peptides were annotated using SignalP (v.5.0)[67], and LipoP (v.1.0)[68]. SignalP discriminated

between 1) Sec/SPI: "standard" secretory signal peptides transported by the Sec translocon and cleaved

by Signal Peptidase I (SPI), 2) Sec/SPII: lipoprotein signal peptides transported by the Sec translocon

and cleaved by Signal Peptidase II (SPII), and 3) Tat/SPI: Tat signal peptides transported by the Tat

translocon and cleaved by SPI. LipoP discriminates between 1) SPI: signal peptide, 2) SpII: lipoprotein

signal peptide, and 3) TMH: n-terminal transmembrane helix. All predictions were in agreement, apart

from one PL7 (gene 1136176.5.peg.4375) which was predicted by LipoP as cytoplasmic and by SignalP

as equally likely cytoplasmic as containing a lipoprotein signal peptide.

Transcriptomic analysis: Sequencing, pre-processing, differential expression analysis, and functional

analysis

Sequencing (12 M reads, 2 x 50 bp) of the isolated RNA was performed by MiGS after rRNA depletion

using RiboZero Plus (Ilumina). cDNA libraries were prepared using an Illumina DNA Prep kit and IDT

10 bp UDI indices, and sequenced on an Illumina NextSeq 2000. Preprocessing of the raw reads was

carried out as follows: Quality control was performed with FastQC (v0.11.9)[69] and reads were trimmed

with Trimmomatic (v0.38)[70]; the high-quality reads were mapped to the reference genome (described

above) with Bowtie2 (v2.3.5.1)[71]; binarization, sorting, and indexing were done with Samtools

(v1.10)[72]; gene counts were computed with the featureCount function of Subread (v2.0.1)[73].

Differential expression analysis was performed with DESeq2 (v1.30.1)[74]. In brief, DESeq2 normalizes

the raw read counts with normalization factors ("size factors") to account for differences in sequencing

depth between samples. Subsequently, gene-wise dispersion estimates are computed for each gene

separately using maximum likelihood, and then shrunk toward the values predicted by the dispersion-

mean dependence curve to obtain final dispersion values. Finally, DESeq2 fits a negative binomial

model to the read counts and performs significance testing using the Wald test. Here reported *p*-values

result from the Wald test of read counts from the digested alginate condition compared to the alginate

condition and were adjusted for multiple testing by Benjamini-Hochberg correction[75] as implemented

in the p.adjust function of base R (v4.1.2). The reported log2 fold changes indicate the log2(DESeq2-

normalised reads in digested alginate condition / DESeq2-normalised reads in alginate condition) for

each gene.

Visualization of gene maps was performed in R with the ggplot2 package (v3.4.0) and the extension

gggenes (v0.4.1) by David Wilkins.

For systematic functional analysis we performed gene set enrichment analysis (GSEA)[19] using the fgsea

function of the fgsea package (v1.20.0) with minimal number of unique genes per gene set "minSize"

= 5 and number of permutations "nPermSimple" = 1000000. In brief, GSEA takes the full gene list

ranked by log2 fold change and annotated with K numbers as input and determines whether the member

genes of any KEGG pathway are randomly distributed throughout the ranked gene list or whether they

are primarily found at the top or bottom[19]. This is quantified by the enrichment score (ES), which

corresponds to a weighted Kolmogorov-Smirnov-like statistic. The ES of each gene set is normalized

to the mean enrichment of random samples of the same size to account for the size of the set, yielding

the normalized enrichment score (NES). To estimate the significance level of the enrichment score, the

p-value of the observed enrichment score is calculated relative to a null distribution that was computed

from permuted data. The estimated significance level was adjusted to account for multiple hypothesis

testing. As gene sets we chose all KEGG pathways and KEGG BRITE categories (as noted in Table S1

in column "KEGG_pathway") within all genes of *V. cyclitrophicus* ZF270 annotated with a KEGG

Ontology identifier ("K number"). Visualization of differential expression levels in KEGG pathways

was performed with the R package pathview (v1.35.0)[76]. We also formed gene sets of the genes

associated with alginate utilization (see "*Sequencing and gene annotation of Vibrio cyclitrophicus*

ZF270*"*, Table S16), of the genes of the flagellar locus that map to the KEGG pathways "Bacterial

motility proteins", and of the genes of the flagellar locus that map to the KEGG pathways "Bacterial

chemotaxis" (as noted in Table S17) within all genes of *V. cyclitrophicus* ZF270.

Image analysis

Cells within microscopy images were segmented and tracked using ilastik (v1.3) ("pixel classification

workflow" and "tracking with probabilities workflow"). Phase contrast images were used for alignment,

segmentation, tracking and linking. Images were cropped at the boundaries of each microfluidic

chamber. The lineage identity of each single cell was assigned by ilastik's tracking plugin and visualized

by coloring the segmented cells, respectively. The growth rate of each cell was computed as the change

of cell area over time, i.e., via a linear regression of the single-cell area over the time between

consecutive cell divisions, based on ilastik's segmentation and tracking output. Cells that were tracked

over less than three frames were excluded. Measurement of swimming speeds and displacement of cells

was performed using ilastik (v1.4), ImageJ (v2.3) and Trackmate (v7.5.2). Briefly, cell-segmentation

("pixel classification workflow" in ilastik) and tracking ("animal tracking workflow" in ilastik) were

performed using the high frame rate phase contrast images in ilastik (v1.4). Cell trajectories and

properties were then computed using the output of the ilastik workflow in Trackmate.

Dispersal analysis

657 For the analysis on the dispersal of cells (Fig. 2), we computed the cell number as the total number of

658 cells within a microfluidic chamber. The change in the number of cells was computed by subtracting

659 the number of cells before the medium switch (i.e., average number of cells between t = 1.9 to 2.1 h)

660 from the number of cells after the medium switch (i.e., average number of cells between t = 3.9 to 5.5

661 h).

Datasets and statistical analysis

All batch experiments were replicated 3 to 6 times. Growth curves were analyzed in Python (v3.7) using

the *Amiga* package (v1.1.0)[77] and GraphPad Prism (v8, GraphPad Software). The microscopy dataset

consisted of eight chambers each, corresponding to the eight replicates shown in Figure 1 and Figure 2.

These were grouped into two biological replicates wherein each biological replicate was fed by media

through a unique channel in a microfluidic chip. Cells with negative growth rates were excluded from

the analysis after visual curation, as they represented artefacts, mistakes in segmentation or linking

during the tracking process, or non-growing deformed cells. Each chamber was treated as an

independent replicate. Comparisons were considered statistically significant when $p < 0.05$ or when the

False Discovery Rate (FDR)-corrected q was smaller than 0.05. FDR corrections were applied when

multiple t tests were performed for the same dataset. Measures of effect size are represented by the R^2

or eta^2 value. All statistical analyses were performed in GraphPad Prism v9.0 (GraphPad Software,

USA), R v4.1.2, RStudio v1.1.463 (Posit, USA).

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All further data is deposited on ERIC Open (https://opendata.eawag.ch/,

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accessible on https://github.com/Microbial-Systems-Ecology/publications upon publication.

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Author Contributions: GD, AS, JS, and JK conceived the research along with MA. GD designed and

performed the experiments. AS developed and implemented the algorithms to compute single-cell

growth rates and visualize lineage identities from segmented and tracked cells. AS analyzed and

visualized the RNA-seq data. JK conceived and implemented the cell dispersal analysis. EC conducted

the ISCA experiment. SP carried out the mass spectrometry measurements. AS and GD analyzed the

data with inputs from JK, JS, CM, MA, and RS. AS wrote the manuscript with inputs from GD, OS,

MA, JK, JS, EC, SP, CM, and RS.

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