1 Polysaccharide breakdown products drive degradation-dispersal cycles of
2 foraging bacteria through changes in metabolism and motility
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Astrid K.M. Stubbusch[1,2,3], Johannes M. Keegstra[4], Julia Schwartzman[5,6], Sammy Pontrelli[7], Estelle E.

Clerc[4], Samuel Charlton[4], Roman Stocker[4], Cara Magnabosco[3], Olga T. Schubert[1,2], Martin

Ackermann[1,2,8], Glen G. D'Souza[1,2]

[1]Institute of Biogeochemistry and Pollutant Dynamics, Department of Environmental Systems
Science, ETH Zurich, Zurich, Switzerland
[2]Department of Environmental Microbiology, Eawag: Swiss Federal Institute of Aquatic Science and
Technology, Dübendorf, Switzerland
[3]Geological Institute, Department of Earth Sciences, ETH Zurich, Zurich, Switzerland
[4]Institute of Environmental Engineering, Department of Civil, Environmental and Geomatic
Engineering, ETH Zurich, Zurich, Switzerland
[5]Department of Civil and Environmental Engineering, MIT, Cambridge, MA, USA
[6]Department of Biology, University of Southern California, Los Angeles, USA
[7]Institute of Molecular Systems Biology, Department of Biology, ETH Zurich, Zurich, Switzerland
[8]Laboratory of Microbial Systems Ecology, School of Architecture, Civil and Environmental
Engineering (ENAC), École Polytechnique Fédéral de Lausanne (EPFL), Lausanne, Switzerland
20
ORCID
AS: 0000-0003-4767-2712
JK: 0000-0002-8877-4881
JS: 0000-0003-4563-4835
SP: 0000-0001-6265-8842
EC: 0009-0003-6046-2856
SC: 0000-0002-0091-5134
RS: 0000-0002-3199-0508
CM: 0000-0001-6871-7188
OS: 0000-0002-2613-0714
MA: 0000-0003-0087-4819
GD: 0000-0002-9123-101X
33
34 **Corresponding authors:** Glen G D'Souza and Astrid KM Stubbusch
35 Address: Eawag, Ueberlandstrasse 133, 8600 Dübendorf, Switzerland
36 Email: glengeralddsouza@gmail.com, astubbusch@icloud.com
37 Phone: +41-58-765-5495

38
39

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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 many studies indicate that these degradation-dispersal cycles contribute to the

carbon flow in marine systems, we know little about how cells alternate between polysaccharide

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, both in its soluble polymeric form as well as on its breakdown products. We

used microfluidics coupled to time-lapse microscopy to analyze motility and growth of individual cells,

and RNA sequencing to study associated changes in gene expression. We found that 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, and this is

accompanied by high expression of genes involved in flagellar assembly, chemotaxis, and quorum

sensing. The motile cells chemotax toward polymeric but not digested alginate, likely enabling them to

find new polysaccharide hotspots. Overall, our findings reveal cellular mechanisms that might also

underlie bacterial degradation-dispersal cycles, which influence the remineralization of biomass in

marine 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 triggered cells to become motile, enabling dispersal from sessile cell

groups. The chemotactic attraction of motile cells to polymeric alginate likely allows cells to move

towards new polysaccharide hotspots. Our study sheds light on the cellular processes that drive bacterial

growth and behavior during carbon remineralization, an important process resulting in CO_2 release from

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. It can be cleaved into oligomers by endo-acting alginate

lyases or into monomers by exo-acting alginate lyases (alginate lyases reviewed here[10]), namely β-D-

mannuronic acid and α-L-guluronic acid[11]. *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]. 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). Measurements of optical density showed that the onset of growth on alginate was delayed by

about 7.5 hours compared to growth on digested alginate (Fig. S2B, D 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]. We also observed that the optical density

at stationary phase was higher when cells were grown on alginate (Fig. S2B and C). However, colony

counts did not show a significant difference in cell numbers (Fig. S3), suggesting that the increased

optical density may stem from aggregation of cells in the alginate medium, as observed for other *Vibrio*

species[7]. Overall, these findings indicate that growth on polysaccharides such as alginate in well-mixed

cultures is initially limited by the extracellular breakdown of the polysaccharide, though similar cell

numbers were reached eventually.

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, as depicted in Figure 1A and described in detail by Dal Co et al.[15]. In brief, the microfluidic

chips are made of an inert polymer (polydimethylsiloxane) bound to a glass coverslip. The PDMS layer

contains flow channels through which the culture medium is pumped continuously. Each channel is

connected to several growth chambers that are laterally positioned. The dimensions of these growth

chambers (height: 0.85 μm, length: 60 μm, width: 90-120 μm) allow cells to freely move and grow as

monolayers. The culture medium, containing either alginate or digested alginate in their soluble form,

is constantly pumped through the flow channel and enters the growth chambers primarily through

diffusion[15,16,4,17,8]. Therefore, the number of cells and their positioning within microfluidic chambers is

determined by the cellular growth rate as well as by cell movement[4]. This setup combined with time-

lapse microscopy allowed us to follow the development of cell communities over time. We found that

over 24 hours dense groups of more than 1000 cells formed on alginate, filling the entire microfluidic

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

exceeded 100 cells per chamber (Fig. 1C and D). 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. 1B). 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

more likely that the lower cell density in digested alginate is caused by cell dispersal. To test the role

of increased viscosity of polymeric alginate in causing the increased aggregation of cells, we measured

the viscosity of 0.1% (w/v) alginate or digested alginate dissolved in TR media. For alginate, the

viscosity was 1.03±0.01 mPa·s (mean and standard deviation of three technical replicates) whereas the

viscosity of digested alginate in TR media was found to be 0.74±0.01 mPa·s. Both these values are

relatively close to the viscosity of water at this temperature (0.89 mPa·s[18]) and, while they may affect

swimming behavior[19], they are insufficient to physically restrain cell movement[20]. Overall, our

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)[16], 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. (A) Schematic representation of the

setup of the microfluidic experiments. (B and C) Representative images at different time points of *V.*

cyclitrophicus ZF270 cells growing in microfluidic chambers, described in detail by Dal Co et al.[15], with (B)

alginate medium or (C) 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. (D) 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 cells 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 that the nutrient switch caused a few or no cells to disperse from small cell groups (Fig. 2B), whereas a large fraction of cells from large cell groups dispersed (Fig. 2C). In fact, the fraction of cells that dispersed upon imposition of the nutrient switch showed a strong positive relationship with the number of cells present, meaning that cells in chambers with many cells were more likely to disperse than cells in chambers with fewer 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) Positive relationship between the number of cells in the microfluidic growth chamber at the time of the switch and the fraction of cells that disperse after the switch. Each circle represents one growth chamber with colors corresponding to (B), and the line depicts a linear regression fit ($R^2 = 0.92$, slope = 0.12).

Cells growing on digested alginate are more motile and polymeric alginate acts as chemoattractant

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[4,16], 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. 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.

To understand whether alginate polymers or alginate breakdown products act as chemoattractants on

motile cells, we measured the chemotactic strength towards alginate and digested alginate using the In

Situ Chemotaxis Assay (ISCA)[21,22]. Interestingly, we found *V. cyclitrophicus* ZF270 to significantly

chemotax toward alginate (chemotactic index $I_c > 1$) but not significantly toward digested alginate (Fig.

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

polymers.

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. Black points mark the starting point of each trajectory, pink points mark the end point of each trajectory, and colored lines mark the trajectories of cells. (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.

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

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

(GSEA)[23] (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[24,25]. Both observations likely relate to the different

growth dynamics of *V. cyclitrophicus* ZF270 on digested alginate compared to alginate (Fig. S2A),

where cells in digested alginate medium reached their maximal growth rate 7.5 hours earlier and thus

showed a shorter lag time (Fig. S2D). As a consequence,, the growth rate at the time of RNA extraction

(mid-to-late exponential phase) may have differed, even though the maximum growth rate of cells

grown in alginate medium and digested alginate medium were not found to be significantly different

(Fig. S2E).

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[26,27], 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,

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 aligned with our

expectations based on 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). This role in dispersal is in agreement with a previous study that showed

induction of the quorum sensing master regulator in *V. cholerae* cells during dispersal from biofilms on

a similar time scale as here (less than an hour)[28]. Quorum sensing is known to control biofilm formation

in the well studied model system *Vibrio cholerae*[29,30] and may also orchestrate the density-dependent

dispersal in the presence of degradation products observed in our study, 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*[31,32].

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*[33,34], 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)[35]. 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)[36]. 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)[34,37]. 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[38,39]. 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,* scissors symbol), 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.[31] 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.[40]. 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")[41–43]. However, cells have been observed not only to aggregate on polysaccharide sources, but also to leave them before the source is depleted[13,43]. This cycle of biomass degradation and dispersal

has long been discussed in the context of foraging [e.g., 44,45,13,46,47], 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[48,49]. 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*[48] 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.

Growth on polysaccharides has been found to be dependent on the cell density, as increased cell density

limits the loss of degradation products and exoenzymes by diffusion[8] (Fig. 6, "Early-stage

polysaccharide breakdown"). We found that cells grown on degradation products reach their maximal

growth earlier and show 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 alginate lyases might seem surprising and wasteful in a monomer-rich

environment. One reason for this observation may be that cells primarily rely on intracellular

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

and catabolism, as has previously been observed for *E. coli* across various carbon sources[50,51]. In fact,

the majority of carbon sources are sensed by prokaryotes through one-component sensors inside the

cell[50]. In the one-component internal sensing scheme, the enzymes and transporters for the use of

various carbon sources are expressed at basal levels, which leads to an increase in pathway

intermediates upon nutrient availability. The pathway intermediates are sensed by an internal sensor,

usually a transcription factor, and lead to the upregulation of transporter and enzyme expression[50,51].

This results in a positive feedback loop, which enables small changes in substrate abundance to trigger

large transcriptional responses[50,52]. Thus, the presence of alginate breakdown products may likely result

in increased expression of all components of the alginate degradation pathway, including the expression

of degrading enzymes. As the gene expression analysis was performed on well-mixed cultures in culture

medium containing alginate breakdown products, we therefore expect a strong stimulation of alginate

catabolism. In a natural scenario, where cells disperse from a polysaccharide hotspot before its

exhaustion, the expression of alginate catabolism genes may likely decrease again once the local

concentration of breakdown products decreases. However, continued production of alginate lyases

could also provide an advantage when encountering a new alginate source and continued production of

alginate lyases may thus help cells to prepare for likely future environments. Further investigations of

bacterial enzyme secretion in changing nutrient environments and at relevant spatial scales are required

to improve our understanding of the regulation of enzyme secretion along nutrient gradients.

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[13,43]. 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[53,54],

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[4,16]. 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.

Direct chemotaxis towards polysaccharides may facilitate the search for new polysaccharide sources

after dispersal. We found that the presence of degradation products not only induces cell dispersal but

also increases the expression of chemotaxis genes. Interestingly, we found that *V. cyclitrophicus* ZF270

cells show chemotaxis towards polymeric alginate but not digested alginate. This contrasts with

previous findings for bacterial strains degrading the insoluble marine polysaccharide chitin, where

chemotaxis was strongest towards chitin oligomers[53], suggesting that oligomers may act as an

environmental cue for polysaccharide nutrient sources[55]. However, recent work has shown that certain

marine bacteria are attracted to the marine polysaccharide laminarin, and not laminarin oligomers[56].

Together with our results, this indicates that chemotaxis towards soluble polysaccharides may be

mediated by the polysaccharide molecules themselves. The mechanism of this behavior is yet to be

identified, but could be mediated by polysaccharide-binding proteins as have been found in

Sphingomonas sp. A1 facilitating chemotaxis towards pectin[57]. Direct polysaccharide sensing adds

complexity to chemosensing as polysaccharides cannot freely diffuse into the periplasm, which can lead

to a trade-off between chemosensing and uptake[58]. Furthermore, most polysaccharides are not

immediately metabolically accessible as they require degradation. But direct polysaccharide sensing

can also provide certain benefits compared to using oligomers as sensory cues. First, it could enable

bacterial strains to preferably navigate to polysaccharide nutrients sources that are relatively

uncolonized and hence show little degradation activity. Second, strong chemotaxis towards degradation

products could hinder a timely dispersal process as the dispersal then requires cells to travel against a

strong attractant gradient formed by the degradation products. Overall, this strategy allows cells to

alternate between degradation and dispersal to acquire carbon and energy in a heterogeneous world with

nutrient hotspots[44,59–61].

Figure 6. Bacterial growth and regulation on patches of polysaccharides. By integrating our results with previous studies on cooperative growth on the same system, as well as results on dispersal cycles in other systems, we highlight where the specific results of this work add to this framework (bold font). When cells encounter polymer sources, the colonization of the nutrient hotspot may be aided by the 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 in a subpopulation of cells and increases the expression of chemotaxis genes. Polymeric alginate acts as chemoattractant towards new polysaccharide sources. 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[43]. 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[47,62], 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[43,63], and/or (iv) increase the genetic variation in bacterial populations[47]. 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, signaling molecules, nutrient starvation, and oxygen starvation[64]. 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[65,66] 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. At these concentrations, both alginate and digested alginate are soluble in the culture medium. 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). Plate reader assays were initiated as described previously[67]. 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 ($\sim10^5$ 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[15,16,68]. 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).

Viscosity of the alginate and digested alginate solution

We measured the viscosity of alginate solutions using shear rheology measurements. We use a 40 mm

cone-plate geometry (4° cone) in a Netzsch Kinexus Pro+ rheometer. 1200 uL of sample was placed on

the bottom plate, the gap was set at 150 um and the sample trimmed. We used a solvent trap to avoid

sample evaporation during measurement. The temperature was set to 25°C using a Peltier element. We

measure the dynamic viscosity over a range of shear rates $= 0.1 – 100$ s-1. We report the viscosity of

each solution as the average viscosity measured over the shear rates $10 – 100$ s-1, where the shear-

dependence of the viscosity was low.

We measured the viscosity of 0.1% (w/V) alginate dissolved in TR media, which was 1.03 +/- 0.01

mPa·s (reporting the mean and standard deviation of three technical replicates.). The viscosity of 0.1%

digested alginate in TR media was found to be 0.74+/-0.01 mPa·s. This means that the viscosity of

alginate in our microfluidic experiments is 36% higher than of digested alginate, but the viscosities are

close to those expected of water (0.89 mPa·s at 25 degree celsius according to Berstad and colleagues

(1988)[18]).

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[21,22] (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[21,22]. *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 x 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[69]). Additionally, KEGG Ontology

identifiers ("K numbers") were annotated with BlastKOALA (v2.2)[70]. 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[71] by dbCAN2 (v9.0)[72]. Enzymes for alginate transport

and metabolism were identified by BLASTn-search[73,74] 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*[31].

Location prediction of alginate lyases

Signal peptides were annotated using SignalP (v.5.0)[75], and LipoP (v.1.0)[76]. 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)[77] and reads were trimmed

with Trimmomatic (v0.38)[78]; the high-quality reads were mapped to the reference genome (described

above) with Bowtie2 (v2.3.5.1)[79]; binarization, sorting, and indexing were done with Samtools

(v1.10)[80]; gene counts were computed with the featureCount function of Subread (v2.0.1)[81].

Differential expression analysis was performed with DESeq2 (v1.30.1)[82]. 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[83] 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)[23] 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[23]. 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)[84]. 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

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

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

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

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

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)[85] 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 artifacts, 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 and code 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.

Citations:

1. BeMiller JN. 4 - Polysaccharides: Occurrence, Structures, and Chemistry. In: BeMiller JN, ed.
Carbohydrate Chemistry for Food Scientists (Third Edition). AACC International Press; 2019:75-
101. doi:10.1016/B978-0-12-812069-9.00004-2

2. Reintjes G, Arnosti C, Fuchs B, Amann R. Selfish, sharing and scavenging bacteria in the
Atlantic Ocean: a biogeographical study of bacterial substrate utilisation. *ISME J*.
2019;13(5):1119-1132. doi:10.1038/s41396-018-0326-3

3. Ratzke C, Gore J. Self-organized patchiness facilitates survival in a cooperatively growing
Bacillus subtilis population. *Nature Microbiology*. 2016;1(5):16022.
doi:10.1038/nmicrobiol.2016.22

4. Povolo VR, D'Souza GG, Kaczmarczyk A, Stubbusch AK, Jenal U, Ackermann M. Extracellular
appendages govern spatial dynamics and growth of Caulobacter crescentus on a
prevalent biopolymer. *bioRxiv*. Published online January 1, 2022:2022.06.13.495907.
doi:10.1101/2022.06.13.495907

5. Westrich JR, Griffin DW, Westphal DL, Lipp EK. Vibrio Population Dynamics in Mid-Atlantic
Surface Waters during Saharan Dust Events. *Frontiers in Marine Science*. 2018;5.
https://www.frontiersin.org/articles/10.3389/fmars.2018.00012

6. Takemura A, Chien D, Polz M. Associations and dynamics of Vibrionaceae in the environment,
from the genus to the population level. *Frontiers in Microbiology*. 2014;5.
https://www.frontiersin.org/articles/10.3389/fmicb.2014.00038

7. Schwartzman JA, Ebrahimi A, Chadwick G, et al. Bacterial growth in multicellular aggregates
leads to the emergence of complex life cycles. *Current Biology*. 2022;32(14):3059-3069.e7.
doi:10.1016/j.cub.2022.06.011

8. D'Souza G, Ebrahimi A, Stubbusch A, et al. Cell aggregation is associated with enzyme secretion
strategies in marine polysaccharide-degrading bacteria. *ISME J*. Published online February 22,
2023. doi:10.1038/s41396-023-01385-1

9. Wang X, Liu J, Liang J, Sun H, Zhang XH. Spatiotemporal dynamics of the total and active
Vibrio spp. populations throughout the Changjiang estuary in China. *Environmental
Microbiology*. 2020;22(10):4438-4455. doi:10.1111/1462-2920.15152

10. Wong TY, Preston LA, Schiller NL. Alginate Lyase: Review of Major Sources and Enzyme
Characteristics, Structure-Function Analysis, Biological Roles, and Applications. *Annual Review
of Microbiology*. 2000;54(1):289-340. doi:10.1146/annurev.micro.54.1.289

11. Mabeau S, Kloareg B. Isolation and Analysis of the Cell Walls of Brown Algae: Fucus spiralis, F.
ceranoides, F. vesiculosus, F. serratus, Bifurcaria bifurcata and Laminaria digitata. *Journal of*
Experimental Botany. 1987;38(9):1573-1580. doi:10.1093/jxb/38.9.1573

12. Hunt DE, David LA, Gevers D, Preheim SP, Alm EJ, Polz MF. Resource Partitioning and
Sympatric Differentiation Among Closely Related Bacterioplankton. *Science*. Published online
May 23, 2008. doi:10.1126/science.1157890

13. Yawata Y, Cordero OX, Menolascina F, Hehemann JH, Polz MF, Stocker R. Competition–
dispersal tradeoff ecologically differentiates recently speciated marine bacterioplankton
populations. *Proc Natl Acad Sci USA*. 2014;111(15):5622-5627. doi:10.1073/pnas.1318943111

14. Huang L, Zhou J, Li X, Peng Q, Lu H, Du Y. Characterization of a new alginate lyase from newly
isolated *Flavobacterium* sp. S20. *Journal of Industrial Microbiology and Biotechnology*.
2013;40(1):113-122. doi:10.1007/s10295-012-1210-1

15. Dal Co A, van Vliet S, Kiviet DJ, Schlegel S, Ackermann M. Short-range interactions govern the
dynamics and functions of microbial communities. *Nat Ecol Evol*. 2020;4(3):366-375.
doi:10.1038/s41559-019-1080-2

16. D'Souza GG, Povolo VR, Keegstra JM, Stocker R, Ackermann M. Nutrient complexity triggers
transitions between solitary and colonial growth in bacterial populations. *ISME J*.
2021;15(9):2614-2626. doi:10.1038/s41396-021-00953-7

17. D'Souza G, Schwartzman J, Keegstra J, et al. Interspecies interactions determine growth
dynamics of biopolymer-degrading populations in microbial communities. *Proc Natl Acad Sci U*
S A. 2023;120(44):e2305198120. doi:10.1073/pnas.2305198120

18. Berstad DA, Knapstad B, Lamvik M, Skjølsvik PA, Tørklep K, Øye HA. Accurate measurements
of the viscosity of water in the temperature range 19.5–25.5°C. *Physica A: Statistical Mechanics*
and its Applications. 1988;151(2-3):246-280. doi:10.1016/0378-4371(88)90015-5

19. Zöttl A, Yeomans JM. Enhanced bacterial swimming speeds in macromolecular polymer
solutions. *Nat Phys*. 2019;15(6):554-558. doi:10.1038/s41567-019-0454-3

20. BERG HC, TURNER L. Movement of microorganisms in viscous environments. *Nature*.
1979;278(5702):349-351. doi:10.1038/278349a0

21. Lambert BS, Raina JB, Fernandez VI, et al. A microfluidics-based in situ chemotaxis assay to
study the behaviour of aquatic microbial communities. *Nat Microbiol*. 2017;2(10):1344-1349.
doi:10.1038/s41564-017-0010-9

22. Clerc EE, Raina JB, Lambert BS, Seymour J, Stocker R. In Situ Chemotaxis Assay to Examine
Microbial Behavior in Aquatic Ecosystems. *JoVE*. 2020;(159):e61062. doi:10.3791/61062

23. Subramanian A, Tamayo P, Mootha VK, et al. Gene set enrichment analysis: a knowledge-based
approach for interpreting genome-wide expression profiles. *Proc Natl Acad Sci U S A*.
2005;102(43):15545-15550. doi:10.1073/pnas.0506580102

24. Wei Y, Lee JM, Richmond C, Blattner FR, Rafalski JA, LaRossa RA. High-Density Microarray-
Mediated Gene Expression Profiling of *Escherichia coli*. *J Bacteriol*. 2001;183(2):545-556.
doi:10.1128/JB.183.2.545-556.2001

25. Gifford SM, Sharma S, Booth M, Moran MA. Expression patterns reveal niche diversification in
a marine microbial assemblage. *ISME J*. 2013;7(2):281-298. doi:10.1038/ismej.2012.96

26. Johnson TL, Fong JC, Rule C, Rogers A, Yildiz FH, Sandkvist M. The Type II Secretion System
Delivers Matrix Proteins for Biofilm Formation by Vibrio cholerae. *J Bacteriol*.

852 2014;196(24):4245-4252. doi:10.1128/JB.01944-14

27. Sikora AE. Proteins Secreted via the Type II Secretion System: Smart Strategies of Vibrio cholerae to Maintain Fitness in Different Ecological Niches. Heitman J, ed. *PLoS Pathog*. 2013;9(2):e1003126. doi:10.1371/journal.ppat.1003126

28. Singh PK, Bartalomej S, Hartmann R, et al. Vibrio cholerae Combines Individual and Collective Sensing to Trigger Biofilm Dispersal. *Current Biology*. 2017;27(21):3359-3366.e7. doi:10.1016/j.cub.2017.09.041

29. Jemielita M, Wingreen NS, Bassler BL. Quorum sensing controls Vibrio cholerae multicellular aggregate formation. Storz G, ed. *eLife*. 2018;7:e42057. doi:10.7554/eLife.42057

30. Waters CM, Lu W, Rabinowitz JD, Bassler BL. Quorum sensing controls biofilm formation in Vibrio cholerae through modulation of cyclic di-GMP levels and repression of vpsT. *J Bacteriol*. 2008;190(7):2527-2536. doi:10.1128/JB.01756-07

31. Wargacki AJ, Leonard E, Win MN, et al. An Engineered Microbial Platform for Direct Biofuel Production from Brown Macroalgae. *Science*. 2012;230502(January):308-314.

32. Zhang L, Li X, Zhang X, Li Y, Wang L. Bacterial alginate metabolism: an important pathway for bioconversion of brown algae. *Biotechnol Biofuels*. 2021;14(1):158. doi:10.1186/s13068-021-02007-8

33. Khan F, Tabassum N, Anand R, Kim YM. Motility of Vibrio spp.: regulation and controlling strategies. *Appl Microbiol Biotechnol*. 2020;104(19):8187-8208. doi:10.1007/s00253-020-10794-7

34. Echazarreta MA, Klose KE. Vibrio Flagellar Synthesis. *Front Cell Infect Microbiol*. 2019;9:131. doi:10.3389/fcimb.2019.00131

35. Prouty MG, Correa NE, Klose KE. The novel sigma54- and sigma28-dependent flagellar gene transcription hierarchy of Vibrio cholerae. *Mol Microbiol*. 2001;39(6):1595-1609. doi:10.1046/j.1365-2958.2001.02348.x

36. Klose KE, Mekalanos JJ. Differential Regulation of Multiple Flagellins in *Vibrio cholerae*. *J Bacteriol*. 1998;180(2):303-316. doi:10.1128/JB.180.2.303-316.1998

37. Srivastava D, Hsieh ML, Khataokar A, Neiditch MB, Waters CM. Cyclic di-GMP inhibits Vibrio cholerae motility by repressing induction of transcription and inducing extracellular polysaccharide production. *Mol Microbiol*. 2013;90(6):1262-1276. doi:10.1111/mmi.12432

38. Kim SY, Thanh XTT, Jeong K, et al. Contribution of six flagellin genes to the flagellum biogenesis of Vibrio vulnificus and in vivo invasion. *Infect Immun*. 2014;82(1):29-42. doi:10.1128/iai.00654-13

39. Nedeljković M, Sastre DE, Sundberg EJ. Bacterial Flagellar Filament: A Supramolecular Multifunctional Nanostructure. *Int J Mol Sci*. 2021;22(14):7521. doi:10.3390/ijms22147521

40. Rajagopala SV, Titz B, Goll J, et al. The protein network of bacterial motility. *Mol Syst Biol*. 2007;3(1):128. doi:10.1038/msb4100166

41. Ebrahimi A, Schwartzman J, Cordero OX. Cooperation and spatial self-organization determine rate and efficiency of particulate organic matter degradation in marine bacteria. *Proc Natl Acad Sci USA*. 2019;116(46):23309-23316. doi:10.1073/pnas.1908512116

42. Drescher K, Nadell CD, Stone HA, Wingreen NS, Bassler BL. Solutions to the Public Goods Dilemma in Bacterial Biofilms. *Current Biology*. 2014;24(1):50-55.

doi:10.1016/j.cub.2013.10.030

43. Alcolombri U, Peaudecerf FJ, Fernandez VI, Behrendt L, Lee KS, Stocker R. Sinking enhances
the degradation of organic particles by marine bacteria. *Nat Geosci*. 2021;14(10):775-780.
doi:10.1038/s41561-021-00817-x

44. Fenchel T. Microbial Behavior in a Heterogeneous World. *Science*. 2002;296(5570):1068-1071.
doi:10.1126/science.1070118

45. Preheim SP, Boucher Y, Wildschutte H, et al. Metapopulation structure of Vibrionaceae among
coastal marine invertebrates. *Environmental Microbiology*. 2011;13(1):265-275.
doi:10.1111/j.1462-2920.2010.02328.x

46. Yawata Y, Carrara F, Menolascina F, Stocker R. Constrained optimal foraging by marine
bacterioplankton on particulate organic matter. *Proc Natl Acad Sci USA*. 2020;117(41):25571-
25579. doi:10.1073/pnas.2012443117

47. McDougald D, Rice SA, Barraud N, Steinberg PD, Kjelleberg S. Should we stay or should we go:
mechanisms and ecological consequences for biofilm dispersal. *Nat Rev Microbiol*.
2012;10(1):39-50. doi:10.1038/nrmicro2695

48. Thomas F, Barbeyron T, Tonon T, Génicot S, Czjzek M, Michel G. Characterization of the first
alginolytic operons in a marine bacterium: from their emergence in marine Flavobacteriia to their
independent transfers to marine Proteobacteria and human gut Bacteroides: Emergence and
transfer of alginolytic operons. *Environmental Microbiology*. 2012;14(9):2379-2394.
doi:10.1111/j.1462-2920.2012.02751.x

49. Dudek M, Dieudonné A, Jouanneau D, et al. Regulation of alginate catabolism involves a GntR
family repressor in the marine flavobacterium Zobellia galactanivorans DsijT. *Nucleic Acids
Research*. 2020;48(14):7786-7800. doi:10.1093/nar/gkaa533

50. Chubukov V, Gerosa L, Kochanowski K, Sauer U. Coordination of microbial metabolism. *Nat
Rev Microbiol*. 2014;12(5):327-340. doi:10.1038/nrmicro3238

51. Martínez-Antonio A, Janga SC, Salgado H, Collado-Vides J. Internal-sensing machinery directs
the activity of the regulatory network in Escherichia coli. *Trends in Microbiology*. 2006;14(1):22-
27. doi:10.1016/j.tim.2005.11.002

52. Wall ME, Hlavacek WS, Savageau MA. Design of gene circuits: lessons from bacteria. *Nat Rev
Genet*. 2004;5(1):34-42. doi:10.1038/nrg1244

53. Bassler BL, Gibbons PJ, Yu C, Roseman S. Chitin utilization by marine bacteria. Chemotaxis to
chitin oligosaccharides by Vibrio furnissii. *Journal of Biological Chemistry*.
1991;266(36):24268-24275. doi:10.1016/S0021-9258(18)54224-1

54. Meibom KL, Li XB, Nielsen AT, Wu CY, Roseman S, Schoolnik GK. The *Vibrio cholerae* chitin
utilization program. *Proc Natl Acad Sci USA*. 2004;101(8):2524-2529.
doi:10.1073/pnas.0308707101

55. Keegstra JM, Carrara F, Stocker R. The ecological roles of bacterial chemotaxis. *Nat Rev
Microbiol*. 2022;20(8):491-504. doi:10.1038/s41579-022-00709-w

56. Clerc EE, Raina JB, Keegstra JM, et al. Strong chemotaxis by marine bacteria towards
polysaccharides is enhanced by the abundant organosulfur compound DMSP. *Nat Commun*.
2023;14(1):8080. doi:10.1038/s41467-023-43143-z

57. Konishi H, Hio M, Kobayashi M, Takase R, Hashimoto W. Bacterial chemotaxis towards
polysaccharide pectin by pectin-binding protein. *Sci Rep*. 2020;10(1):3977. doi:10.1038/s41598-

937 020-60274-1

58. Norris N, Alcolombri U, Keegstra JM, et al. Bacterial chemotaxis to saccharides is governed by a trade-off between sensing and uptake. *Biophysical Journal*. 2022;121(11):2046-2059. doi:10.1016/j.bpj.2022.05.003

59. Blackburn N, Fenchel T. Influence of bacteria, diffusion and shear on micro-scale nutrient patches, and implications for bacterial chemotaxis. *Mar Ecol Prog Ser*. 1999;189:1-7. doi:10.3354/meps189001

60. Blackburn N, Fenchel T, Mitchell J. Microscale Nutrient Patches in Planktonic Habitats Shown by Chemotactic Bacteria. *Science*. 1998;282(5397):2254-2256. doi:10.1126/science.282.5397.2254

61. Smriga S, Fernandez VI, Mitchell JG, Stocker R. Chemotaxis toward phytoplankton drives organic matter partitioning among marine bacteria. *Proc Natl Acad Sci USA*. 2016;113(6):1576-1581. doi:10.1073/pnas.1512307113

62. Ronce O. How Does It Feel to Be Like a Rolling Stone? Ten Questions About Dispersal Evolution. *Annu Rev Ecol Evol Syst*. 2007;38(1):231-253. doi:10.1146/annurev.ecolsys.38.091206.095611

63. Fenchel T. Motility of bacteria in sediments. *Aquat Microb Ecol*. 2008;51:23-30. doi:10.3354/ame01182

64. Rumbaugh KP, Sauer K. Biofilm dispersion. *Nat Rev Microbiol*. 2020;18(10):571-586. doi:10.1038/s41579-020-0385-0

65. Hehemann JH, Arevalo P, Datta MS, et al. Adaptive radiation by waves of gene transfer leads to fine-scale resource partitioning in marine microbes. *Nat Commun*. 2016;7:12860. doi:10.1038/ncomms12860

66. Tibbles BJ, Rawlings DE. Characterization of nitrogen-fixing bacteria from a temperate saltmarsh lagoon, including isolates that produce ethane from acetylene. *Microb Ecol*. 1994;27(1). doi:10.1007/BF00170115

67. D'Souza G, Waschina S, Pande S, Bohl K, Kaleta C, Kost C. Less is more: Selective advantages can explain the prevalent loss of biosynthetic genes in bacteria. *Evolution*. 2014;68(9):2559-2570. doi:10.1111/evo.12468

68. Mathis R, Ackermann M. Response of single bacterial cells to stress gives rise to complex history dependence at the population level. *Proc Natl Acad Sci USA*. 2016;113(15):4224-4229. doi:10.1073/pnas.1511509113

69. Brettin T, Davis JJ, Disz T, et al. RASTtk: A modular and extensible implementation of the RAST algorithm for building custom annotation pipelines and annotating batches of genomes. *Sci Rep*. 2015;5(1):8365. doi:10.1038/srep08365

70. Kanehisa M, Sato Y, Morishima K. BlastKOALA and GhostKOALA: KEGG Tools for Functional Characterization of Genome and Metagenome Sequences. *Journal of Molecular Biology*. 2016;428(4):726-731. doi:10.1016/j.jmb.2015.11.006

71. Cheng D, Jiang C, Xu J, Liu Z, Mao X. Characteristics and applications of alginate lyases: A review. *International Journal of Biological Macromolecules*. 2020;164:1304-1320. doi:10.1016/j.ijbiomac.2020.07.199

72. Zhang H, Yohe T, Huang L, et al. dbCAN2: a meta server for automated carbohydrate-active enzyme annotation. *Nucleic Acids Res*. 2018;46(W1):W95-W101. doi:10.1093/nar/gky418

73. NCBI Resource Coordinators. Database resources of the National Center for Biotechnology
Information. *Nucleic Acids Res*. 2016;44(Database issue):D7-D19. doi:10.1093/nar/gkv1290

74. Altschul SF, Gish W, Miller W, Myers EW, Lipman DJ. Basic local alignment search tool. *J Mol
Biol*. 1990;215(3):403-410. doi:10.1016/S0022-2836(05)80360-2

75. Almagro Armenteros JJ, Tsirigos KD, Sønderby CK, et al. SignalP 5.0 improves signal peptide
predictions using deep neural networks. *Nat Biotechnol*. 2019;37(4):420-423.
doi:10.1038/s41587-019-0036-z

76. Juncker AS, Willenbrock H, von Heijne G, Brunak S, Nielsen H, Krogh A. Prediction of
lipoprotein signal peptides in Gram-negative bacteria. *Protein Sci*. 2003;12(8):1652-1662.

77. Andrews S. FastQC: A Quality Control Tool for High Throughput Sequence Data [Online].
Available online at: http://www.bioinformatics.babraham.ac.uk/projects/fastqc/. Published online
2010. http://www.bioinformatics.babraham.ac.uk/projects/fastqc/

78. Bolger AM, Lohse M, Usadel B. Trimmomatic: a flexible trimmer for Illumina sequence data.
Bioinformatics. 2014;30(15):2114-2120. doi:10.1093/bioinformatics/btu170

79. Langmead B, Salzberg SL. Fast gapped-read alignment with Bowtie 2. *Nat Methods*.
2012;9(4):357-359. doi:10.1038/nmeth.1923

80. Danecek P, Bonfield JK, Liddle J, et al. Twelve years of SAMtools and BCFtools. *Gigascience*.
2021;10(2):giab008. doi:10.1093/gigascience/giab008

81. Liao Y, Smyth GK, Shi W. featureCounts: an efficient general purpose program for assigning
sequence reads to genomic features. *Bioinformatics*. 2014;30(7):923-930.
doi:10.1093/bioinformatics/btt656

82. Love MI, Huber W, Anders S. Moderated estimation of fold change and dispersion for RNA-seq
data with DESeq2. *Genome Biology*. 2014;15(12):550. doi:10.1186/s13059-014-0550-8

83. Benjamini Y, Hochberg Y. Controlling the False Discovery Rate: A Practical and Powerful
Approach to Multiple Testing. *Journal of the Royal Statistical Society: Series B
(Methodological)*. 1995;57(1):289-300. doi:10.1111/j.2517-6161.1995.tb02031.x

84. Luo W, Brouwer C. Pathview: an R/Bioconductor package for pathway-based data integration
and visualization. *Bioinformatics*. 2013;29(14):1830-1831. doi:10.1093/bioinformatics/btt285

85. Midani FS, Collins J, Britton RA. AMiGA: Software for Automated Analysis of Microbial
Growth Assays. Silva-Rocha R, ed. *mSystems*. 2021;6(4):e00508-21.
doi:10.1128/mSystems.00508-21