

RESEARCH PAPER



Adaptations in gut Bacteroidales facilitate stable co-existence with their lytic bacteriophages

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ABSTRACT

Bacteriophages (phages) and bacteria within the gut microbiome persist in long-term stable coexistence. These interactions are driven by eco-evolutionary dynamics, where bacteria employ a variety of mechanisms to evade phage infection, while phages rely on counterstrategies to overcome these defenses. Among the most abundant phages in the gut are the crAss-like phages that infect members of the order Bacteroidales, in particular, genus *Bacteroides*. In this study, we explored some of the mechanisms enabling the co-existence of four phage-Bacteroidales host pairs *in vitro* using a multi-omics approach (transcriptomics, proteomics and metabolomics). These included three *Bacteroides* species paired with three crAss-like phages (*Bacteroides intestinalis* and ϕ crAss001, *Bacteroides xylanisolvens* and ϕ crAss002, and an acapsular mutant of *Bacteroides thetaiotaomicron* with DAC15), and *Parabacteroides distasonis* paired with the siphovirus ϕ PDS1. We show that phase variation of individual capsular polysaccharides (CPSs) is the primary mechanism promoting phage co-existence in Bacteroidales, but this is not the only strategy. Alternative resistance mechanisms, while potentially less efficient than CPS phase variation, can be activated to support bacterial survival by regulating gene expression and resulting in metabolic adaptations, particularly in amino acid degradation pathways. These mechanisms, also likely regulated by phase variation, enable bacterial populations to persist in the presence of phages, and *vice versa*. An acapsular variant of *B. thetaiotaomicron* demonstrated broader transcriptomic, proteomic, and metabolomic changes, supporting the involvement of additional resistance mechanisms beyond CPS variation. This study advances our understanding of long-term phage–host interaction, offering insights into the long-term persistence of crAss-like phages and extending these observations to other phages, such as ϕ PDS1. Knowledge of the complexities of phage–bacteria interactions is essential for designing effective phage therapies and improving human health through targeted microbiome interventions.

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1. Introduction

The human gastrointestinal tract (GIT) harbors a complex and diverse population of microorganisms (bacteria, archaea, and eukarya) and their viruses, which exhibit a high degree of taxonomic diversity.¹ These inhabitants have been implicated in human health and diseases since they can be directly or indirectly involved in essential physiological and metabolic functions within the GIT.^{1–4} Bacteriophages (phages), viruses that infect bacteria, are one of the most abundant entities within the gut microbiome.⁴ From an ecological standpoint, phages are commonly seen as predators or parasites that bacteria actively evade, regardless of

the mode of phage infection (lytic, lysogenic, or chronic infection).⁵ Bacteria constantly acquire new resistance mechanisms against viruses to evade infection, while phages evolve strategies to overcome these defensive mechanisms.^{6–8} This selective pressure constitutes a force influencing ecological and evolutionary dynamics within microbial communities called antagonistic coevolution.^{9–11} This phenomenon leads to diversity in both phage and bacterial populations in natural environments, resulting in a mutually beneficial outcome that secures the perpetuation of both partners without dramatic fluctuations and extinction events.¹¹ There are two main biological

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models used to explain phage-host coevolution in the human gut; the ‘arms race’ dynamic that involves continuous directional selection of mutations leading to expanded ranges of host and phage resistance and infectivity over time, and the fluctuating selection dynamic involving density-dependent fluctuating selection based on a trade-off between the benefits of resistance and its metabolic costs.^{10,12,13} The latter model appears to be favored in the gut, as it allows for localized coevolution promoting specialized bacterial resistance, narrowing the phage host range, and increasing bacterial diversity, as is usually described in this environment.¹⁴ However, specific ways in which this system operates in the human gut to perpetuate phage and bacterial survival and its implications for human health is still an area of ongoing research.

The crAss-like phages (order *Crassvirales*) are one of the most abundant groups of dsDNA-tailed phages in the human gut.¹⁵ Their bacterial hosts belong to the order Bacteroidales, which is one of the most prevalent bacterial taxonomic groups in the healthy human gut microbiota, where *Bacteroides* and *Parabacteroides* are two of the most abundant genera.^{15,16} To date, only a limited number of gut crAss-like phages have been successfully isolated in the laboratory,^{17–21} and even fewer have undergone thorough biological characterization.^{17,19} Despite their virulent nature, crAss-like phages can effectively co-exist with their host bacteria without dramatically impacting community structure or target bacterial numbers.^{19,22,23} Most species of human gut bacteria in the order Bacteroidales, with the exception of *Prevotella*, contain genetic loci encoding for several alternative capsular polysaccharides (CPSs), with some species possessing up to 13 different types.²⁴ The presence of strong capsular niche competitors places acapsular strains at a significant fitness disadvantage.²⁵ These surface structures are implicated in several functions, such as evading or modifying the host immune response, protecting bacterial cells against phage, and facilitating bacterial colonization and adhesion.^{24,26} This molecular mechanism also supports the concept of fluctuating selection dynamics between phage and host among species of Bacteroidales. Dynamic phase variation in CPS

expression has been proposed as an explanation for the rapid acquisition of phage resistance within a high proportion of the host population, as seen in the case of ϕ crAss001 infecting *B. intestinalis*.²² However, further in-depth investigations with other crAss-like phages and other phage types infecting Bacteroidales are lacking. Additionally, the co-existence observed between ϕ crAss001 and its host bacterium may not be a unique characteristic of *Crassvirales*. Many tailed phages demonstrate a similar pattern of relatively stable cohabitation with their hosts over numerous generations in the mammalian gut.¹⁵

Here, we investigated the mechanisms underlying phage-bacteria co-existence in bacteria and phage isolated from the mammalian gut through an *in vitro* multi-omics approach. We investigated four phage-bacteria pairs: three *Bacteroides* strains supporting replication of three distinct crAss-like phages (*Bacteroides intestinalis* APC919/174 and ϕ crAss001, *Bacteroides xylanisolvens* APCS1/XY and ϕ crAss002, *Bacteroides thetaiotaomicron* VPI-5482 *tdk*[−] engineered acapsular mutant [Δ cps] and DAC15), and *Parabacteroides distasonis* APCS2/PD with the siphovirus ϕ PDS1. Our findings reveal that while CPS phase variation is the primary method that promotes phage co-existence in the Bacteroidales, it is not the only mechanism. Alternative competing resistance mechanisms, which also involve phase variation but less efficiently than changes in CPS expression, can still exert sufficient quantitative effect to allow bacterial populations to survive in the presence of phages. This suggests that bacteria employ multiple strategies to resist phage predation, ensuring their co-existence in the gut microbiome.

2. Materials and methods

2.1. Bacteriophages, bacterial hosts, and culture conditions

Four phage-bacteria pairs were analysed in this study. The host strains evaluated were as follows: *Bacteroides intestinalis* APC919/174 (complete genome RefSeq accession number NZ_CP041379), *Bacteroides xylanisolvens* APCS1/XY (RefSeq: NZ_CP042282), *Parabacteroides distasonis* APCS2/PD (RefSeq: NZ_CP042285), and *Bacteroides*

thetaitaomicron VPI-5482 *tdk*[−] acapsular [Δ cps] (wild type RefSeq: NC_004663), along with their corresponding phages ϕ crAss001 (Refseq: NC_049977), ϕ crAss002 (Refseq: NC_055828), ϕ PDS1 (Genbank: MN929097), and DAC15 (RefSeq: NC_055832), respectively. Frozen pure stocks of the host strains stored in 30% glycerol at -80°C were streaked on Fastidious Anaerobe Agar (FAA) (Neogen) plates (1.5% agar w/v) and incubated at 37°C for 48 hours in anaerobic jars (Thermo Fisher Scientific) with AnaeroGenTM Compact (Thermo Fisher Scientific) to obtain isolated colonies to perform each experiment. Overnight cultures were obtained after picking a single bacterial colony grown on FAA plates and inoculating into 10 mL of Fastidious Anaerobe Broth (FAB) (Neogen) and incubating overnight at 37°C anaerobically in anaerobic jars or in a type A vinyl anaerobic chamber (Coy Laboratory Products).

2.2. Phage-bacteria in vitro co-culture experiment

Co-cultures of each host with its phage were performed in triplicate by serial sub-culturing in broth over 5 days. Controls, which were not infected, were also carried out in triplicate. This was accomplished by picking single bacterial colonies, inoculating them into 10 mL of FAB and incubating them overnight at 37°C anaerobically. The next day, 200 μL of the overnight culture was inoculated in 10 mL of fresh FAB. When bacteria reached the early-log phase ($\text{OD}_{595} = \sim 0.2\text{--}0.3$), they were infected with a corresponding phage at a multiplicity of infection (MOI) of 1 and cultured for 24 hours at 37°C in anaerobic conditions. An equal volume of growth media without phage was added to the controls. Subsequent rounds of sub-culturing were conducted daily for 5 days by introducing the previous day's co-culture into 10 mL fresh FAB medium at a ratio of 1:50. Bacterial and phage enumeration was carried out daily following the overnight incubation. This involved plating bacterial dilutions and performing plaque assay or qPCR for phage counting. Bacteria were counted by serial diluting of 500 μL of overnight culture in phosphate-buffered saline (PBS) (Sigma-Aldrich). One hundred microliters of the diluted sample was spread on FAA agar plates (1.5% agar

w/v) and incubated in anaerobic conditions at 37°C for 48 hours. The number of generations per day was calculated using the formula: $N = N_0 \times 2^n$, where n is the number of generations, N is the number of cells after 24 hours, and N_0 is the initial number of cells at the start of the day. For phage counting, the overnight culture was centrifuged to remove the cells at $4,500 \times g$ for 20 min at 4°C . The supernatant was filtered through a 0.45 μm pore syringe-mounted polyethersulfone (PES) membrane filter (Sarstedt). The quantification of ϕ crAss001 and DAC15 was performed through plaque assays by mixing 200 μL of an overnight host bacteria culture, 100 μL of SM buffer (1 M Tris HCl pH 7.5, 5 M NaCl, 1 M MgSO_4) diluted phage supernatant and 3 mL of 0.45% FAA molten overlay soft agar with the addition of MgSO_4 and CaCl_2 (1 mm final concentration) and poured onto FAA agar plates (1.5% agar w/v). Since ϕ PDS1 does not produce large enough plaques that can be counted by naked eye, and ϕ crAss002 does not produce plaques at all, both phages were quantified by qPCR with the standard curve method using the same conditions and primers described previously.^{19,27}

To assess the potential influence of the initial MOI on the infection, a range of different initial MOIs (0.001–10) was tested for each phage-host pair, following the same procedure described above for the co-culture experiment over 5 days.

A growth curve for all phage-host pairs under the tested conditions was performed on the fifth day of the co-culture experiment to evaluate potential variations in bacterial growth in the presence of phage. A 1:50 dilution of the fifth-day co-culture in fresh media was prepared. Aliquots of 200 μL , in triplicate, were dispensed into the wells of a flat-bottom 96-well micro test plate (Sarstedt). The plate was sealed, and the OD_{595} was measured under anaerobic conditions for 24 hours using a microtiter plate reader (Multiskan FC, Thermo Fisher Scientific). FAB was used as a negative control. All samples were analyzed in triplicate.

2.3. RNA extraction, RNA sequencing, and RNA-seq analysis

Samples for the transcriptomics analysis were collected during the bacterial early-log phase on

the fifth day of the co-culture experiment, which occurred between 5 and 7 hours after the subculture ($OD_{595} = \sim 0.2-0.3$). For each of the phages examined in the study, three control samples (consisting only of bacteria) and three phage-bacteria co-culture samples were analyzed. One milliliter of the early-log phase culture was transferred to RNase-free tubes (Thermo Fisher Scientific) and centrifuged for 3 min at $17,000 \times g$. The pellet was resuspended and lysed in 1 mL of TRIzol™ Reagent (Life Technologies), and the total bacterial RNA was extracted into a final volume of 30 μ L following the manufacturer's guidelines. The RNA samples (30 μ L) were treated with the TURBO DNA-free kit (Thermo Fisher Scientific) to remove any DNA contamination and were once again purified and concentrated using the RNeasy Mini kit (Qiagen) following the manufacturer's protocol. The quantity, purity, and integrity of the isolated RNA were measured by Qubit with the RNA High Sensitivity assay (Thermo Fisher Scientific), Nanodrop (Thermo Fisher Scientific), and Bioanalyzer 2100 (Agilent Technologies) with the Agilent RNA 6000 nano kit (Agilent Technologies), respectively. Strand-specific RNA-Seq libraries were prepared and sequenced by Genewiz (Leipzig, Germany) on the Illumina NovaSeq 6000 platform using a 2×150 bp paired-end sequencing configuration. RNA-Seq analysis was conducted following the practice guidelines proposed by Love et al. (2015).²⁸ Briefly, RNA-seq reads were mapped to host and phage reference genomes [*Bacteroides intestinalis* APC919/174 (RefSeq: NZ_CP041379) and ϕ crAss001 (Refseq: NC_049977), *Bacteroides xylanisolvens* APCS1/XY (RefSeq: NZ_CP042282) and ϕ crAss002 (Refseq: NC_055828), *Parabacteroides distasonis* APCS2/PD (RefSeq: NZ_CP042285) and ϕ PDS1 (Genbank: MN929097), and *Bacteroides thetaiotaomicron* VPI-5482 (RefSeq: NC_004663) and DAC15 (RefSeq: NC_055832)] using Bowtie2 (v2.5.1) in an end-to-end alignment mode.²⁹ SAMtools was used to process the resulting alignment SAM files to obtain sorted BAM files. A read count matrix, summarizing transcript abundance on a per-gene basis, was constructed using the *summarizeOverlaps* function from the R package GenomicAlignments.

Differential gene expression analysis was carried out on this count matrix using the DESeq2 package in R.²⁸

2.4. Ratio of sensitive-resistant bacterial colonies after phage infection

The dynamics of the sensitive/resistant cell ratios in phage-bacteria co-cultures were examined for the four bacterial strains tested in this study. Early-log phase cultures ($OD_{595} = \sim 0.2-0.3$) were infected with their respective phages at an MOI of 1 and incubated anaerobically at 37°C for 24 hours. Negative controls, consisting of bacteria without phage infection, were also included. Daily subcultures (dilution 1:50) in fresh media were performed for five consecutive days. On each day, 200 μ L of serially diluted overnight culture was mixed with 3 mL of FAB soft agar (0.45%) and 100 μ L of a high-titer phage stock ($>10^9$) of the host. These high-titer phage stocks were derived from different co-culture time points. The mixture was poured onto FAA (1.5%) plates and incubated under anaerobic conditions at 37°C for 48 hours. Overnight cultures with no extra phage added were also plated and incubated under the same conditions. The percentage of resistant colonies was determined by dividing the number of colonies counted when additional phage was added (resistant sub population) by the number of colonies counted on the plates where phage was not added (total population).

2.5. In silico analysis

Long-read sequencing data of *Bacteroides xylanisolvens* APCS1/XY from Guerin et al. (2021)¹⁹ were used to identify potential structural variants within its genome. To accomplish this, long-read DNA sequences were aligned against the reference genome using minimap2,³⁰ and putative structural variants were detected using Sniffles (v2.0.7).³¹ Reads supporting genomic structural variants were extracted using Samtools (v1.18)³² and QIIME2 (v2023.7, filter_fasta.py script).³³ These reads were then manually reviewed using Artemis (v16) by aligning them at genomic coordinates containing structural variants, as indicated by Sniffles. The relevant sequences were then searched

within all genomes using Artemis and BLASTn to check if the sequence was a common element within the genomes of interest.

A Clusters of Orthologous Groups (COGs) analysis was carried out to classify and categorize functions of gene products that significantly differed in the RNA-seq analysis to identify functional patterns and trends.³⁴ Functional annotation of orthologous groups was accomplished by RPS-BLAST (v2.14.1)³⁵ against the COG database (22 January 2018).

Visualization of gene clusters was performed using Clinker.³⁶

2.6. Proteomics analysis

Samples for the proteomics analysis were collected during the bacterial early-log phase ($OD_{595} \approx 0.2$ – 0.3) on the fifth day of co-culture. For each of the phages examined in the study, four control samples (consisting only of bacteria) and four phage-bacteria co-culture samples were analyzed. Bacterial cultures were centrifuged at $4,500 \times g$ for 15 min at 4°C . The supernatant and pellet were stored separately at -80°C , with any remaining liquid removed from the pellet before storage, prior to protein extraction.

Tryptic digests were prepared from all pellet and supernatant samples following a modified version of the manufacturer's (Protifi) protocol for S-Trap mini devices, in which dithiothreitol (DTT) and iodoacetamide (IAA) were used as the reducing and alkylating agents, respectively. The supernatants and pellets were prepared for digestion on the S-Trap devices using slightly different methods to account for the unique characteristics of each sample matrix. The supernatants (5–8 mL) were first lyophilized to reduce the sample volume followed by solubilization in $1 \times$ lysis buffer (5% sodium dodecyl sulfate [SDS], 50 mM triethylammonium bicarbonate [TEAB] pH 7.5) at a $\sim 1:10$ sample to lysis buffer ratio. The pellets contained suspended agar from the growth media which co-pelleted with the bacterial cells. To aid with agar dissolution, $2 \times$ lysis buffer (10% SDS, 100 mM TEAB pH 7.5) was used³⁷ to cover each of the pellets ($\sim 500 \mu\text{L}$) prior to pellet resuspension with vortexing. The supernatant and pellet samples were heated to 95°C for 10 min for lysis

followed by a 10 min centrifugation at $21,000 \times g$ to remove debris. Samples were reduced with 500 mM DTT (final concentration 22 mM) and incubated at 95°C for 10 min. Samples were alkylated with 500 mM IAA (final concentration 40 mM) and incubated for 30 min in the dark. The remainder of the digestion followed the manufacturer's protocol. Protein digestion was performed with $0.8 \mu\text{g}$ of MS grade trypsin (Thermo Scientific Pierce) and incubation overnight at 37°C in a wet chamber. The eluted peptides were concentrated in a vacuum centrifuge and acidified with 10% formic acid (FA) for a final percentage of 1% FA. Acidification precipitated and removed residual agar. Acidified samples were filtered with a 10 kDa MWCO PES membrane centrifugal filter. Peptide concentrations were determined with a Pierce Micro BCA assay (Thermo Scientific Pierce) following the manufacturer's instructions.

Samples were analyzed by 1D-LC-MS/MS in two separate runs, one for all supernatant samples and one for all pellet samples. Samples were run in a block-randomized design as previously described³⁸ with each block containing a full replicate of all species and conditions and two washes between blocks. For each sample, 1,000 ng of peptide was loaded onto a 5 mm, $30 \mu\text{m}$ ID C18 TRIzol™ PepMap100 pre-column (Thermo Fisher Scientific) using an UltiMate™ 3000 RSLCnano Liquid Chromatograph (Thermo Fisher Scientific) for desalting. The pre-column was switched in line with a $75 \text{ cm} \times 75 \mu\text{m}$ analytical EASY-Spray column packed with PepMap RSLC C18, $2 \mu\text{m}$ material (Thermo Fisher Scientific) which was heated to 60°C . The analytical column was connected via an Easy-Spray source to an Exploris 480 hybrid quadrupole-Orbitrap mass spectrometer (Thermo Fisher Scientific). Peptides were separated on the analytical column using a 140 min gradient as previously described³⁹ and ionized via electrospray ionization (ESI). MS^1 spectra were obtained at a resolution of 60,000 on a 380 to 1,600 m/z window and fragmented with a normalised collision energy of 27%. MS^2 spectra were obtained for the 15 most abundant ions in the MS^1 spectra using a maximum injection time of 50 ms, dynamic exclusion of 25 s, and an exclusion of

ions of +1 charge state. Roughly 100,000 MS/MS spectra were acquired per sample.

For protein identification, individual protein sequence databases were created for each phage-host pair. Each database contained all the protein sequences from the bacterial host and phage genome in addition to proteins from the cRAP protein sequence database (<http://www.thegpm.org/crap/>) containing protein sequences of common laboratory contaminants. The accession numbers for the proteomes in each database are as follows: *B. intestinalis* APC919/174 (GenBank: CP041379.1) and ϕ crAss001 (UniProt: UP000262320), *B. xylanisolvens* APCS1/XY (GenBank assembly: GCA_018279805.1) and ϕ crAss002 (UniProt: UP000595483), *P. distasonis* APCS2/PD (GenBank: CP042285.1) and ϕ PDS1 (UniProt: UP000595413), *B. thetaiotaomicron* VPI-5482 (RefSeq: NC_004663.1) and DAC15 (RefSeq: NC_055832.1). For protein identification, MS/MS spectra from each sample were searched against the appropriate database using the Sequest HT and Percolator nodes in Proteome Discoverer version 2.3.0.523 (Thermo Fisher Scientific) as previously described.⁴⁰ Only proteins identified with medium or high confidence were retained resulting in an overall False Discovery Rate (FDR) of <5%. Proteins were quantified by calculating normalized spectral abundance factors (NSAF%) as previously described.⁴⁰

2.7. Untargeted metabolomics analysis

Untargeted metabolomics was performed on bacterial pellets collected during the bacterial early-log phase on the fifth day of the co-culture experiment. After reaching OD₅₉₅ = ~0.2–0.3, bacterial cultures were centrifuged at 4,500 × g for 15 min at 4°C. The supernatants were discarded, and the bacterial pellets were frozen in liquid nitrogen and stored at –80°C until further analysis. The extraction of metabolites and analysis were carried out by MS-Omics (Denmark) following an analytical methodology for the detection of the semi-polar metabolites. Briefly, 350 µL of ice-cold MilliQ water (~0°C), 350 µL of pre-cooled methanol (–20°C), and 350 µL of pre-cooled chloroform (–20°C) were added into cell pellet tubes. Sample tubes were vortexed for 1 min, snap-frozen in liquid nitrogen for 1 min, and thawed on ice. The freeze-thaw cycle was repeated two times to induce cell

lysis. The samples were centrifuged at 4 °C at 15,000 × g for 10 min, and the supernatant (500 µL) was transferred to a new tube. Extracts were evaporated under a gentle stream of nitrogen and reconstituted in 120 µL of mobile-phase eluent A (10 mM ammonium formate, 0.1% FA in water) with 10% content B (10 mM ammonium formate, 0.1% FA in MeOH). Reconstituted samples were diluted 1:20 and were analyzed using a Thermo Scientific Vanquish LC coupled to a Thermo Q Exactive HF MS. An ESI interface was used as ionization source. Analysis was performed in positive and negative ionization mode. The Ultra Performance Liquid Chromatography was performed using a slightly modified version of the protocol described by Doneanu et al. (2011).⁴¹ Peak areas were extracted using Compound Discoverer 3.3 (Thermo Fisher Scientific). Identification of compounds was performed at four levels: i) Level 1: identification by retention times (compared against in-house authentic standards), accurate mass (with an accepted deviation of 3 ppm), and MS/MS spectra, ii) Level 2a: identification by retention times (compared against in-house authentic standards), accurate mass (with an accepted deviation of 3 ppm), iii) Level 2b: identification by accurate mass (with an accepted deviation of 3 ppm), and MS/MS spectra, and iv) Level 3: identification by accurate mass alone (with an accepted deviation of 3 ppm). To ensure high-quality sample preparation, a quality control sample was prepared by pooling small equal aliquots from each sample to create a representative average of the entire set. This sample was treated and analyzed at regular intervals throughout the sequence.

2.8. Statistical analysis

Statistical analyses were carried out using the GraphPad Prism (v8.0.1) and R (v2023.12.1 + 402) software. Data are expressed as mean ± standard deviation (SD). Normality of data distribution was assessed by the Shapiro–Wilk test. Differences between infected and control samples in experiments (e.g., bacterial counts, % resistant vs. sensitive colonies) were evaluated using unpaired *t*-tests with Holm–Sidak method for multiple comparisons correction (*p* < 0.05). A two-way ANOVA was applied to analyze the main effects of

infection condition, time and their interaction to evaluate growth kinetics, followed by Sidak's multiple comparisons test for *post-hoc* analysis ($p < 0.05$). For RNA-seq data, differential expression analysis was performed using DESeq2, where generalized linear models were fitted for each gene. The Wald test was used to evaluate significance, and p -values were adjusted for multiple comparisons using the Benjamini–Hochberg correction with an FDR threshold set at 5%. The expression values for each CPS cluster were calculated by averaging the expression levels of their constituent genes, followed by normalization against the total expression of all CPS clusters. Differences in the relative expression of the CPS loci were analyzed assuming normal distribution using Welch's t -test. In the proteomics analysis, normalized spectral abundance factor (NSAF) values were calculated and \log_2 -transformed. Proteins were filtered to include only those with at least three non-zero values across all replicates in both conditions, and missing values were imputed using a normal distribution (width = 0.1, downshift = 2.5). Welch's t -test was performed followed by Benjamini–Hochberg correction with an FDR threshold set between 5% and 20%. Heat map was generated using z-scored data and Euclidean row and column clustering in R. Volcano plot and principal component analyses (PCA) were also performed using R. For metabolomics, t -tests corrected by the Benjamini–Hochberg method with an FDR threshold of 5% were applied. Multivariate analysis, including PCA and partial least squares-discriminant analysis (PLS-DA), was carried out on metabolites with high accuracy (level 1 and 2a) using the MetaboAnalyst (v6.0) tool.⁴²

3. Results

3.1. Stable co-existence of crAss-like phages and ϕ PDS1 with their hosts in vitro

Four phage-bacteria pairs were co-cultivated by sub-culturing for 5 days to investigate the effect of phage on survival and transcriptional responses in *Bacteroides*/*Parabacteroides* strains over time. Enumeration of bacteria and phages was carried out daily, and it was confirmed that all four phages continued to co-exist with their bacterial hosts over

the 5 days of co-culture, without a complete collapse of either the host or the phage populations (Figures 1(a–d)). The number of generations per day was estimated for each bacterium as ~6 generations over 24 hours in each experiment, resulting in 30 generations in total, over 5 days. The dynamics of infection by phage ϕ crAss001 in *Bacteroides intestinalis* APC919/174 and by ϕ PDS1 in *Parabacteroides distasonis* APCS2/PD were similar, with both phages showing an increase in titer after 24 hours of infection, reaching $\sim 5 \times 10^9$ pfu/mL and $\sim 3 \times 10^{10}$ copies/mL, respectively. In the following days, the titre decreased and stabilized at $\sim 4 \times 10^7$ pfu/mL and $\sim 5 \times 10^8$ copies/mL, respectively (Figures 1(a,c)). In the case of DAC15 infecting an engineered acapsular mutant of *B. thetaiotaomicron* VPI-5482 (*B. thetaiotaomicron* Δ cps), an increase in the titre was observed in the early days, reaching its peak at 48- and 72-hours post-infection with a titre around $\sim 3 \times 10^9$ pfu/mL, which then remained stable beyond the 5-day mark with a titre of $\sim 2 \times 10^8$ pfu/mL (Figure 1(d)). The absence of CPS phase variation in this strain, previously shown to be the major mechanism of phage evasion,^{18,24} neither prevented the crAss-like phage DAC15 from propagating at high levels nor did it result in the collapse of bacterial culture, supporting previous data that alternative cell-surface related mechanisms, limiting phage infection through phase variation or otherwise, may be involved.²⁴

On the other hand, ϕ crAss002 exhibited very different dynamics. Similar to a previous report,¹⁹ the ϕ crAss002 titer underwent a significant reduction after the initial 24 hours of infection, falling to $\sim 10^5$ genome copies/mL. However, in the following days, the ϕ crAss002 titer increased considerably, reaching its peak at 72 hours with $\sim 2 \times 10^{10}$ genome copies/mL and remaining close to this level in the subsequent days (Figure 1(b)).

Phage co-existence with their hosts was confirmed for all four phages and a titer stabilization around the 72-hour post-infection was observed in all cases except for DAC15, where a slight decrease occurred at the end of the experiment. All phages had a significant impact ($p < 0.05$) on the population of their bacterial hosts in the first days of the experiment, resulting in a reduction in the number of bacteria compared to the control (no infection). These differences diminished during the following

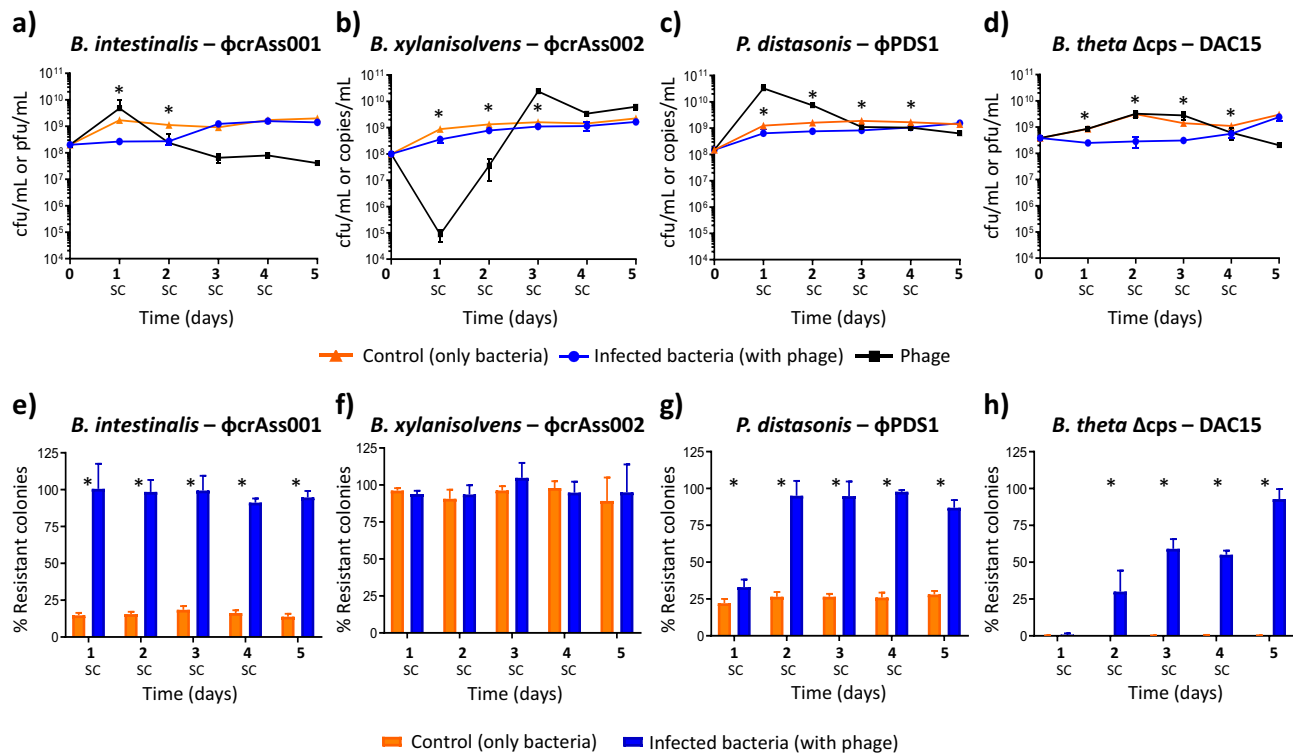


Figure 1. Bacteria-phage dynamics and changes in phage resistance during the 5-day co-culture. (a–d) enumeration of bacteria and phage during the 5-day co-culture experiments. After each 24-hour period, a new subculture (SC) was performed in fresh media at a dilution of 1:50. (a) *B. intestinalis* + ϕ crAss001, (b) *B. xylanisolvens* + ϕ crAss002, (c) *P. distasonis* + ϕ PDS1, (d) *B. theta* Δ cps + DAC15. The orange line represents bacterial counts (CFU/mL) for the control condition (uninfected bacteria), the blue line represents bacterial counts (CFU/mL) in co-culture with phage, and the black line represents phage counts (pfu/mL or copies/mL) in co-culture with bacteria. The choice of the phage enumeration method depended on the ability to produce countable plaques with a given phage-host pair (PFU/mL). Non-plaquing ϕ crAss002 and ϕ PDS1 were quantified by qPCR (copies/mL). (e–h) changes in the percentage of resistant cells during the 5 days of co-culture experiments. (e) *B. intestinalis* + ϕ crAss001, (f) *B. xylanisolvens* + ϕ crAss002, (g) *P. distasonis* + ϕ PDS1, (h) *B. theta* Δ cps + DAC15. Orange bars represent the mean percentage of resistant cells in the uninfected culture (control), while blue bars show the mean percentage of resistant cells when bacteria were exposed to phage. Error bars indicate standard deviation (SD). All experiments were carried out in triplicate. *: statistically significant differences in bacterial counts or percentage of resistant colonies when comparing the control with the infected conditions were set at $p < 0.05$.

days, and by the fifth day, the bacterial values were identical to the control. In *B. intestinalis* and *B. theta* Δ cps, a ~ 1 log reduction in bacterial colony-forming unit (CFU) counts was observed between 24 and 72 hours post-infection compared to the control (Figures 1(a,d)), which suggests that initially dominant phage-sensitive subpopulations were depleted during the first several rounds of phage infection and were later replaced by less sensitive subpopulations. For *B. xylanisolvens* and *P. distasonis*, the reduction in bacterial counts after initial infection was less pronounced (Figures 1(b,c)).

The effect of the initial MOI at the moment of the infection was also tested (Figure S1). Over a broad range of MOIs tested (0.001–10), no

significant differences were observed in the titer of the phages and bacteria at equilibrium after 5 days of co-culture, despite an MOI-dependent difference in the time taken for the phage-host system to reach an equilibrium (Figure S1). The resulting titers, phage-host ratios at equilibrium, and the ability of phage-host pairs to co-exist stably in co-culture were, therefore, dictated by the properties of phage-host pairs and not by the initial MOI.

The kinetics of bacterial growth in both controls and infected cultures after the fifth day of co-culture were evaluated to determine if the presence of the phage had any impact when phage and bacteria ratios had stabilized. Different kinetics were observed for each phage-host pair (Figure S2). While ϕ crAss001 and DAC15 caused a delay

and reduction in the density of the stationary phase of their hosts (Figures S2A and S2D), ϕ crAss002 seemed to have a slight stimulatory effect on the growth of *B. xylanisolvens* (Figure S2B). No significant differences were observed in the kinetics of the growth curve of *P. distasonis* after ϕ PDS1 infection (Figure S2C).

3.2. Phage exposure increases the percentage of resistant host subpopulations

Phase variation or similar mechanisms leading to phenotypic differentiation and dynamic equilibrium between sensitive and resistant subpopulations^{22,24} could explain stable phage-host co-existence in the chosen phage-host pairs. Such dynamics could explain the observed variations in the titer of the phages and their hosts during the five-day experiment. To test this hypothesis, the rate of resistant cells was measured at all time points (Figures 1(e-h)). An average of ~16% of clones in a naïve culture of *B. intestinalis* were found to be resistant to ϕ crAss001 (Figure 1(e)). However, this number increased to ~97% after only 24 hours of contact with ϕ crAss001, and a high percentage of resistant clones was then maintained for the remaining days (Figure 1(e)). This correlates with the rapid drop of phage titer seen in the *B. intestinalis* – ϕ crAss001 co-culture after day 1. In the case of *P. distasonis*, the initial proportion of resistant clones was ~26% (Figure 1(g)), rising to ~35% after 24 hours of ϕ PDS1 exposure and then reaching ~95% after 48 hours, which remained similar until the end of the experiment (Figure 1(g)). The drop in phage titer was similar to ϕ crAss001, albeit stabilizing at a higher level (10^9 genome copies/mL compared to 10^7 pfu/mL) at the 72-hour timepoint.

Unlike the previous two examples, *B. thetaiotaomicron* Δ cps showed very different dynamics. The initial ratio of the clones resistant to DAC15 was ~0.3%, indicating high susceptibility to the phage. This value gradually changed to ~30% at 48 hours, reaching ~93% on the fifth day (Figure 1(h)). This result suggests that the lack of the phase-variable CPS operons necessitates the use by *B. thetaiotaomicron* Δ cps of alternative phage defense mechanisms, potentially including phase variation of other surface structures,²⁴ that require

more time to build a subpopulation of cells resistant against phage infection. Curiously, the *B. thetaiotaomicron* Δ cps-DAC15 pair showed the highest difference in bacterial counts compared to the control (no infection) on days 1–4 during the co-culture experiment, with this resistance appearing to level out on day 5, corresponding to the high level of phage resistance (~93%) achieved by this time (Figures 1(d,h)). Such delayed resistance could explain how phage DAC15 maintained high titers during the first 3 days of co-culturing.

On the other hand, the level of resistance in *B. xylanisolvens* was very high (~94%) even before contact with phage ϕ crAss002, which potentially explains the inability of ϕ crAss002 to produce visible plaques on this host bacterium (Figure 1(f)).¹⁹ After exposure to ϕ crAss002, the resistant percentage showed no significant changes over the 5 days (Figure 1(f)). This initial high resistance to ϕ crAss002 could explain the differing behavior of ϕ crAss002 in the *in vitro* co-culture experiment compared to the other phages (Figures 1(a–d)). For the other phages, a higher percentage of the naïve host cells were sensitive, allowing those phages to propagate and increase in titer during the initial stages of this experiment (Figures 1(a–d)).

3.3. Phage infection drives dynamic shifts in bacterial surface gene expression through phase variation

After confirming the ability of *Bacteroides* and *Parabacteroides* phages to co-exist with their bacterial hosts without exerting a significant penalty on host cell density, we next examined the transcriptional response of each host to persistent phage presence. RNA-seq analysis was performed for the four phage-bacteria pairs during the early-log phase of growth on the fifth day of the co-culture experiment.

The transcriptional response in the *B. intestinalis*- ϕ crAss001 pair showed statistically significant changes ($p < 0.05$) in 136 genes (28 downregulated and 108 upregulated), showing mean \log_2 fold changes ranging between -6.14 ± 0.31 and 2.35 ± 0.25 (Table S1A). Most of the genes belonged to previously reported phase-variable CPS biosynthesis clusters (operons) termed PVR7,

8, 9, 11 and 12, of which four were upregulated and one (PVR9) – downregulated (Figure 2(a)).²² The details of these gene clusters are provided in Figure S3A and Table S2A. These gene clusters shared many common elements. Among them are UpxY family transcription anti-terminators and UpxZ-family inhibitors of UpxY, preventing non-cognate UpxY anti-termination, therefore allowing transcription of only one CPS operon per cell at a time.^{43,44} Glycosyltransferases and other enzymes (acetyltransferases and aminotransferases) are also encoded by these operons that can contribute to CPS chain modifications. The transcription of these operons is likely to be driven by a putative invertible promoter region of 183–186 bp flanked

by a conserved 10 bp inverted repeat (IR) GTTCGTTTAA.²² This strongly suggests that these gene clusters, regulated by site-specific recombination (phase variation), may be relevant to determining the host's sensitivity to phage infection. As suggested previously, the downregulation of the PVR9 operon as a result of ϕ crAss001 exposure is likely to limit opportunities for infection that rely on PVR9 as a phage receptor.²² At the same time, upregulated CPS operons (7, 8, 11 and 12) may have a protective or neutral effect. Constant switching between expression of different CPS types, therefore, results in a dynamic equilibrium between resistant and sensitive sub populations, allowing the phage to persist in this

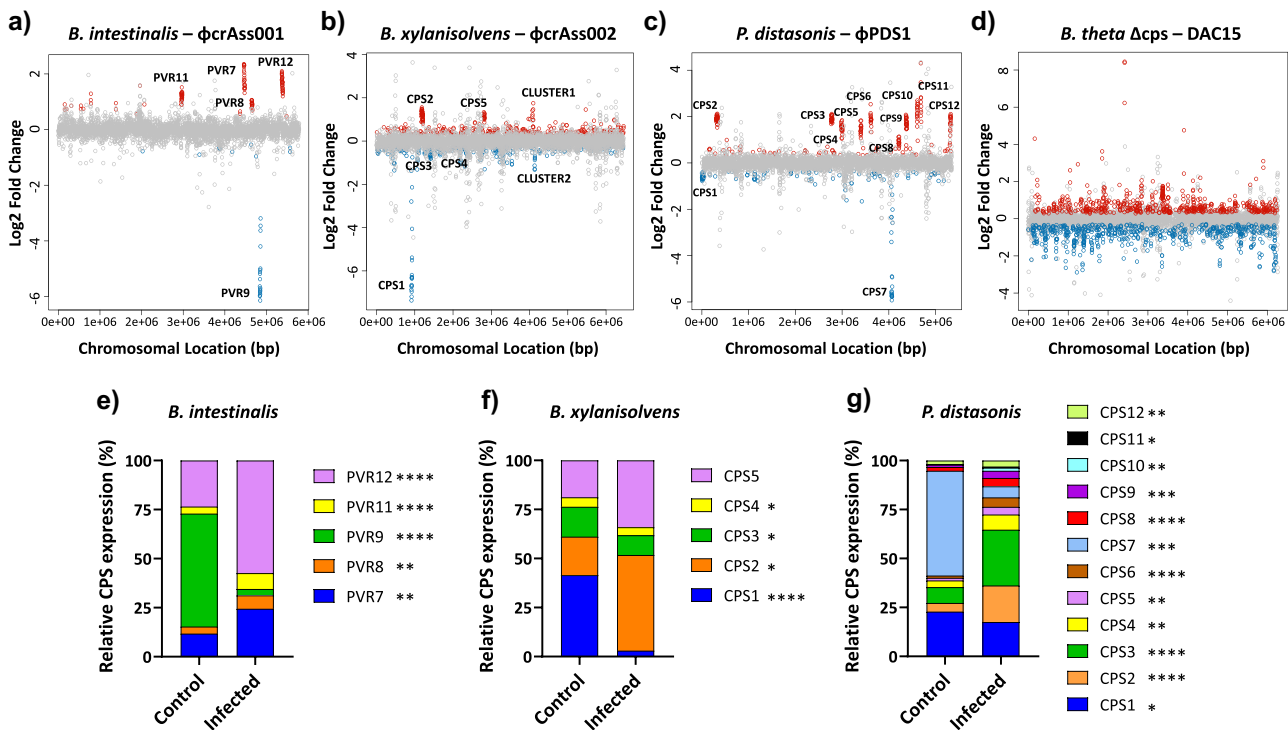


Figure 2. Transcriptional changes and CPS expression profiles after bacteria-phage co-culture. (a-d) transcriptional changes during early-log phase growth on the fifth day of each bacteria-phage co-culture: (a) *B. intestinalis* + ϕ crAss001, (b) *B. xylanisolvens* + ϕ crAss002, (c) *P. distasonis* + ϕ PDS1, (d) *B. theta* Δ cps + DAC15. Genes are plotted as dots according to their chromosomal location in each bacterial host. The log₂ fold change represents the differential expressions of genes when comparing control (uninfected) with infected bacteria. Red dots represent genes that were significantly upregulated upon phage infection, blue dots are those that were significantly downregulated, and grey dots are those that did not show statistically significant changes in expression. Gene clusters implicated in capsule biosynthesis are labelled as CPS (capsular polysaccharide) and PVR (phase-variable region). Statistical significance was determined using an FDR-adjusted p -value (FDR < 0.05). (e-g) relative expression of the different CPS loci in bacterial capsular strains under infected and uninfected conditions determined by RNA-seq data. (e) *B. intestinalis* + ϕ crAss001, (f) *B. xylanisolvens* + ϕ crAss002, (g) *P. distasonis* + ϕ PDS1. The expression values for each CPS cluster were calculated by averaging the expression levels of their constituent genes, followed by normalization against the total expression of all CPS clusters. The resulting values are presented as a percentage of the total CPS expression for each condition. Differences in the relative expression of the CPS loci were analysed assuming normal distribution using Welch's t -test. Significant differences in proportions after phage infection are indicated in the legend as follows: * $p < 0.05$; ** $p < 0.01$; *** $p < 0.001$; **** $p < 0.0001$.

mixed population. Clusters of Orthologous Groups (COGs) analysis was performed to classify and categorize the protein functions of other genes with altered expression located outside the PVR loci, revealing that most of the genes (40%) with known functions were related to inorganic ion transport and metabolism (Figure S4A, Table S3A).

B. xylanisolvens showed differences in the expression of 603 genes (233 downregulated and 370 upregulated) with a mean log₂ fold change ranging from -7.36 ± 0.14 to 1.75 ± 0.17 (Table S1B). Among them were six well-defined gene clusters (3 downregulated and 3 upregulated) (Figure 2(b)). Similar to our observations in *B. intestinalis*, four of these gene clusters primarily included genes related to CPS biosynthesis, including a UpxY family anti-termination system, glycosyltransferase enzyme genes and other genes related to the biosynthesis of polysaccharides (Figure S3B, Table S2B). Among the remaining two gene clusters, one (CLUSTER1) may represent a polysaccharide utilization locus (PUL) that includes a Sus-TonB dependent transporter system, hydrolases, phosphodiesterases, and a DeoR transcription regulator.⁴⁵ CLUSTER2 contains genes encoding an ABC transporter, a fimbriin-related protein and a TonB-dependent receptor (Figure S3B, Table S2B). Furthermore, alterations in gene expression of repetitive functions distributed throughout the genome and not associated with these gene clusters were observed, including transposases, which may have the potential to induce phase variation, Sus/Rag family and TonB-dependent transporters systems, and ATP-binding proteins (Table S1B), which encode bacterial surface molecules that have been previously reported to be associated with phage sensitivity in *Bacteroides* species.^{16,22} The COG analysis for the known functions of genes outside the CPS operons showed more diversity than *B. intestinalis*, with carbohydrate transport and metabolism (13.3%) and amino acid transport and metabolism (10%) being the most represented categories (Figure S4B, Table S3B).

To further explore the regulation of the gene clusters identified in *B. xylanisolvens*, we performed a re-analysis of the available long-read genome sequencing data¹⁹ to detect potential structural variants. We identified 46 potential

structural variants across the *B. xylanisolvens* genome (Table S4). However, only one high-confidence inversion was found to be associated with the regions of interest detected in our RNA-seq analysis. This inversion was located in the intergenic region of the first gene of the CPS1 operon (FQN58_03730 – FQN58_03810) (Figure S5), and an integrase was identified upstream of this region, suggesting potential involvement in genomic recombination within this gene cluster (Figures S5A and S5B). A 20 bp IR sequence, GTTACTTCTTAGGTAACGGA, flanked a 290 bp invertible region preceding the first gene of the CPS1 operon (Figure S5C). This identical repeat sequence was also identified in four other genomic regions, each flanking a potential invertible fragment of either 289 bp (in three) or 290 bp (in one). Three of these IRs were located in the intergenic regions next to the UpxY genes of CPS3 (FQN58_RS05950 – FQN58_RS06080), CPS4 (FQN58_RS09155 – FQN58_RS09165), and CPS5 (FQN58_RS11900 – FQN58_RS11985), and were preceded by a tyrosine-type DNA invertase. The fourth matching fragment was in the intergenic region next to a site-specific integrase (FQN58_RS05740), although no CPS genes were found nearby. Additionally, a 90% identical IR (18/20 bp) (GTTACTTTGTAGGTAACGGA) was identified in both forward and reverse orientations, flanking a 280 bp potential invertible region preceded by an integrase. Still, it was not associated with CPS clusters detected in the RNA-seq analysis. The case of the potential CPS4 is noteworthy since it appears to consist of only three genes: an UpxY, a polysaccharide biosynthesis/export family protein and an N-acetylmuramoyl-L-alanine amidase. In the gene coordinate plot with RNA-seq data, this cluster could not be easily distinguished due to the small number of genes involved and a relatively small transcriptional change (Figure 2(b)). Nevertheless, all three genes within this cluster were found to be downregulated, and the presence of the IR in the intergenic region of the UpxY family antiterminator, preceded by a DNA invertase, suggests that it may represent another phase variable cell surface operon in *B. xylanisolvens*. The remaining CPS operon

(CPS2) was preceded by a gene coding for a DUF3078 domain-containing protein, and no repeats or invertases were observed in the intergenic region, unlike in the other CPS operons (Figure S3B). In contrast, this operon includes a transposase as part of its structure. The other two gene clusters identified, which were not associated with CPS structures, were found to be preceded by a transposase (CLUSTER1) and a helix-turn-helix domain-containing protein (CLUSTER2).

The RNA-seq results of the *P. distasonis* – ϕ PDS1 pair revealed changes in the expression profile of 393 genes (131 downregulated and 262 upregulated) (Table S1C). The mean \log_2 fold change ranged between -5.92 ± 0.36 and 4.30 ± 1.46 , and 12 clusters of genes were identified (10 upregulated and 2 downregulated) (Figure 2(c)). While the two downregulated gene clusters contained the UpxY (CPS1 and CPS7), the rest of the clusters contained a gene encoding the transcriptional anti-termination protein NusG positioned at the beginning of each operon. This may represent an alternative mode of transcriptional regulation to that of the UpxY anti-terminator system observed in the other *Bacteroides* strains (Figure S3C, Table S2C). The other genes present in these clusters were sugar transferases and modification enzymes similar to the ones described in the other strains (Figure S3C, Table S2C). COG analysis revealed that most of the genes with a known function outside the CPS operons were categorized under inorganic ion transport and metabolism (20.3%) and cell envelope biogenesis (14.5%) (Figure S4C, Table S3C). We identified that 10 out of 12 CPS operons had a 17 bp consensus IR (GCTACTYRGNRAGTAGC) flanking a putative promoter region of 314 to 345 bp in size. These operons were preceded by a site-specific integrase and corresponded with those containing the anti-termination protein NusG gene in the first position of the operon. Invertible promoter regions flanked by IRs were not found in CPS1 and CPS7. After analyzing potential similarities between the IR sequences detected in the three studied strains and those found in other *Bacteroides* species, such as *B. thetaiotaomicron* or *B. fragilis*,^{24,46} it was observed that, in all cases except for *B. intestinalis*, these IR sequences contained palindromic regions with the same short sequence, GTTAC–GTAAC (with the first thymine replaced by a cytosine in the case of

P. distasonis), and were also preceded by integrase or invertase enzymes.

These results suggest that alterations in the expression of CPSs are one of the main mechanisms that explain phage-bacteria co-existence in discontinuous culture. The normalized fractions of CPS gene reads mapped to each of the strains changed after phage exposure ($p < 0.05$) (Figures 2(e–g)). The most predominant CPS type in each of the three hosts: PVR9 (57.6%) in *B. intestinalis*, CPS1 (41.5%) in *B. xylanisolvens*, and CPS7 (53.5%) in *P. distasonis* were strongly suppressed to 3.3%, 3.0% and 5.7%, respectively. Simultaneously, the expression of other CPS types was favored (Figures 2(e–g)), so that PVR12 became the highest expressed CPS (57.4%) in *B. intestinalis*, CPS2 (48.8%) in *B. xylanisolvens*, and CPS3 (28.5%) in *P. distasonis*.

However, CPS gene operons are not the only class of genes involved in bacterial response to phage persistence, as we have seen that DAC15 can also successfully propagate and persist with an acapsular host. The transcriptomics analysis of *B. thetaiotaomicron* Δ cps showed the highest number of genes with significantly altered transcription among the four phage-bacteria pairs. Overall, 1,136 changes were detected, 591 downregulated and 545 upregulated. The mean \log_2 fold change varied from -2.88 ± 0.89 to 8.44 ± 0.57 (Table S1D), showing even stronger upregulation of some of the genes compared to the data from capsular strains. Clearly defined gene clusters with a strong transcriptional response could not be identified (Figure 2(d)). Nevertheless, other groupings of non-consecutive genes were identified. Notably, an upregulation of 30S and 50S ribosomal proteins (BT_RS13645 – BT_RS13840) suggests changes in translational machinery. In addition, other gene groups exhibited changes in expression, including transposases and integrases that may be involved in DNA recombination processes, TonB-dependent transporters and Sus-like systems linked to outer membrane proteins that control the uptake of polysaccharides, and other genes encoding surface proteins, like OmpA family proteins or efflux RND transporter permeases, among others (Table S1D). Analysis of the COGs revealed a more even distribution among different categories, with no single category prominent. Carbohydrate transport and metabolism was the primary category (15.4%) of known gene functions. In comparison, cell envelope

biogenesis and outer membrane category represented 9%, positioning this category as the fourth most altered in *B. thetaiotaomicron* Δ cps (Figure S4D, Table S3D). Curiously, the most highly upregulated gene, BT_RS09760 (which corresponds to BT1927 in GenBank annotation), is located near a site-specific integrase and has previously been reported to encode a surface layer (S-layer) protein expressed in a phase-variable manner.⁴⁷ Unlike the gene operons of CPS, which were comprised of multiple genes, only two additional genes (BT_RS09750 – BT_RS09755) were upregulated in this operon. A consensus IR of 19 bp (CCGTTACCTABVRAGTAAC) preceding this operon was also detected in seven other upregulated loci close to site-specific integrases, surface proteins, and OmpA family proteins, as previously reported.²⁴ Remarkably, a nearly identical palindromic IR sequence (CCGTTACCTAAGAAGTAAC) is found next to tyrosine-type DNA invertase genes of the wild type *B. thetaiotaomicron* CPS operons: CPS1, CPS3, CPS5, and CPS6. This finding shows that these diverse operons are regulated by related mechanisms, enabling phage persistence even in the absence of phase-variable CPS.

This phase variation, regulating CPS and other surface structures, apparently generates frequent and reversible changes within specific hypermutable loci, introducing phenotypic diversity into clonal populations and rendering different populations resistant or sensitive to phage infections.⁴⁸ This explains the co-existence of phage and bacteria when they are in co-culture.

3.4. Protein-level changes in gene expression in response to phage exposure

Proteomic analysis of bacterial cells collected during the early-log phase after five days of daily subcultures with phage exposure did not reveal large differences between phage-infected cultures and negative controls across the bacterial strains tested. Over 2,000 proteins were identified in each strain (Tables S5A, S5B, S5C and S5D). Principal Component Analysis (PCA) showed no distinct clustering based on experimental conditions (Figure S6), indicating minimal global changes in protein expression due to phage infection. At the individual protein level, only a few proteins showed statistically significant differences

with a 5% FDR threshold (Tables S5A, S5B, S5C and S5D). In *B. intestinalis*, four proteins were differentially abundant: a N-acetyltransferase (WP_007663187.1) belonging to PVR8, an UpxZ family transcription anti-terminator antagonist (WP_115501761.1) and an ATP-grasp domain (WP_115501769.1) both linked to PVR9, and a hypothetical protein (WP_115502230.1) that was not part of the PVR operons (Table S5A). *B. xylanisolvens* exhibited differences in the abundance of six proteins: an acyltransferase (WP_055234723.1) associated with CPS2, a polysaccharide export protein (WP_087318117.1) belonging to CPS3, two UpxY family transcription anti-terminators with identical protein sequences (WP_008025447.1) linked to CPS1 and CPS5, and a glycosyltransferase (WP_008643375.1) and a Wzz/FepE/Etk N-terminal domain-containing protein (WP_008643393.1), both belonging to CPS1 (Table S5B). *P. distasonis* had only one differentially abundant protein, an electron transport complex subunit D (WP_005855664.1) that was not part of the CPS operons (Table S5C), while *Bacteroides thetaiotaomicron* Δ cps showed differences in five proteins, members of the PorV/PorQ family (WP_008765856.1), an AAA family ATPase (WP_008766137.1), a BT1926 family outer membrane beta-barrel protein (WP_008766273.1), and two with unknown function (WP_011109276.1 and WP_008762274.1) (Table S5D). When applying a less stringent FDR of 20%, additional differentially abundant proteins were identified (Tables S5A, S5B, S5C and S5D). Comparison with RNA-seq data revealed consistency in some findings. For example, in *B. intestinalis*, 5 of 9 significantly altered proteins coincided with their corresponding genes, also significant in RNA-seq, which belonged to different CPSs: PVR8, PVR9, PVR11. Similarly, in *B. xylanisolvens*, 11 of 12 proteins matched RNA-seq data, all belonging to the identified clusters with strong transcriptional responses: CPS1, CPS2, CPS3, and CPS5. No relevant changes were observed in *P. distasonis*, while *B. thetaiotaomicron* Δ cps showed 10 out of 19 proteins coinciding with RNA-seq data, including the highly responsive WP_008766274.1 – BT_RS09760 and WP_008766273.1 – BT_RS09755, among others (Tables S5A, S5B, S5C and S5D).

Supernatant analysis revealed no significant findings for the capsular bacteria (Figure S7, Tables S6A, S6B and S6C); however, for *B. thetaiotaomicron* Δ cps,

PCA identified two distinct clusters and 155 differentially expressed proteins at a 5% FDR (Figure 3, Table S6D).

Surprisingly, 64 of these proteins corresponded to significant changes in the RNA-seq dataset, including the strongly transcriptionally regulated WP_008766274.1 – BT_RS09760, WP_008766274.1 – BT_RS07605 and WP_011109276.1 – BT_RS22595, all of which are associated with the S-layer/OmpA system.²⁴ Among the proteins not significant in RNA-seq, WP_005677723.1 (encoded by BT_RS00855), a Fe-S protein homologous to NifU, showed the most significant difference in protein abundance. Additionally, 22 of the differentially abundant proteins were ribosomal proteins (30S and 50S), while another 22 were nutrient uptake proteins belonging to the Sus/Rag systems and TonB-dependent transporters, underscoring the relevance

of these proteins in the response to DAC15 infection, which also was observed in the RNA-seq analysis. Interestingly, these findings suggest that the acapsular strain exhibits more pronounced proteomic differences compared to the capsular strain, as we also observed in the transcriptomic analysis. While we expected more significant changes in the bacterial pellet, the absence of CPSs may increase interaction with the environment and phages, leading to greater protein release into the supernatant due to lysis.

3.5. Intracellular bacterial metabolome reacts to phage exposure with more relevant changes seen in the acapsular strain

To explore whether phage exposure induces metabolic adaptations in the bacterial population that it attacks, we performed untargeted intracellular

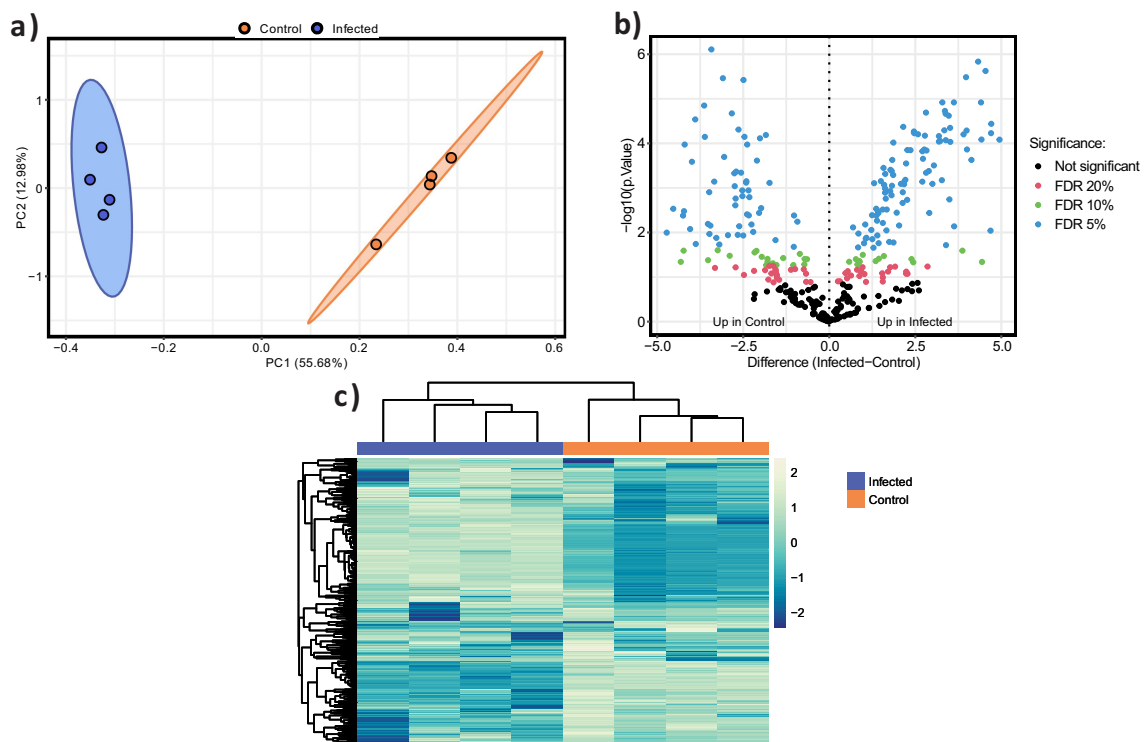


Figure 3. Differential proteomics profile in the supernatant obtained during the early-log phase on the fifth day of the co-culture experiment with *B. theta* Δ cps and DAC15. (a) Principal component analysis (PCA) of proteomic profiles of *B. theta* Δ cps supernatant between control (orange) and DAC15-infected samples (blue colour). Ellipses represent 95% confidence intervals, and the explained variances of the two principal components are shown in brackets along the axes. (b) volcano plot representing the significantly differentially abundant proteins. Dots represent individual proteins upregulated in control (left) and DAC15 infection (right) conditions. Y-axis shows non-adjusted *p*-values, however, dots are coloured based on corrected *p*-values (Benjamini-Hochberg FDR) representing different thresholds: 5% (blue), 10% (green), 20% (red), and non-significant (black). (c) heatmap displaying hierarchical clustering of protein abundance profiles after z-score transformation. Row and columns are clustered based on Euclidean distance, showing differential protein expression between control and infected conditions. Columns are coloured by sample condition with blue representing infected samples and orange representing control. Yellow indicates higher relative protein abundances, while blue represents lower relative abundance.

metabolomics on bacterial pellets during the early-log phase after 5 days of co-culture. A total of 1,722 entities were detected by LC-MS, of which 201 were accurately identified with high confidence based on chemical standards (levels 1 and 2a) (Tables S7A and S7B). Univariate analysis, after 5% FDR correction, showed significant changes in the intensity of only two metabolites, 3-methyl-2-oxovaleric acid and 4-methyl-2-oxovaleric acid, in *B. thetaiotaomicron* Δ cps. Both metabolites decreased after DAC15 exposure, with reductions of 3.1- and 5.1-fold, respectively (Figures 4(a,b)). No significant differences were observed among the metabolites of the other tested strains. 3-methyl-2-oxovaleric acid and 4-methyl-2-oxovaleric acid are catabolic products of isoleucine and leucine produced during branched-chain amino acid degradation, which suggests that modulation of this metabolic pathway may occur in response to phage exposure in this strain. This alteration in the metabolism of iso- and leucine catabolic products could indicate a metabolic shift aimed at redirecting resources or energy metabolism in response to the stress induced by DAC15 infection itself, or a population-level biosynthetic response to it (production of additional defensive cell surface structures).

Subsequently, a multivariate analysis was performed to survey global differences in the bacterial metabolome after phage exposure. PCA revealed that samples of *B. thetaiotaomicron* Δ cps were clearly separated by phage infection (Figure 4(c)), while this grouping was not observed for the samples from the capsular strains (Figure S8). A tendency to group could be observed in the metabolome samples of *B. intestinalis* and *P. distasonis* depending on whether they were infected or not, but the small number of replicates did not provide sufficient statistical power. Next, a partial least square discriminant analysis (PLS-DA), a supervised method with full awareness of the class labels, was applied. The metabolome samples were grouped according to whether the cultures were infected or not in all cases. However, only the model for *B. thetaiotaomicron* Δ cps showed good model fit, quality, and accuracy after cross-validation ($R^2 = 0.97$, $Q^2 = 0.82$), which confirmed that significant changes occurred in the metabolome of this strain after infection with

DAC15 (Table S8). Variable Importance in Prediction (VIP) scores, which estimate the contribution of a given predictor to a PLS regression model, were calculated and are shown in Figure 4(d) and Table S9. The compounds that contributed most strongly to the separation between the sample groups in *B. thetaiotaomicron* Δ cps were 4-methyl-2-oxovaleric acid, 1,3-dimethyluracil, NAD⁺, N-acetylserine, urocanic acid, 3-methyl-2-oxovaleric acid, 2-oxo-3-phenylpropanoic acid, serine phosphate, S-adenosylmethionine (decreased in infection), and 1-aminocyclopropanecarboxylic acid and cis-aconitic acid (increased in infection). Most of these metabolites are related to amino acid pathways, which suggests that the *B. thetaiotaomicron* Δ cps population responds to DAC15 infection with metabolic changes possibly improving bacterial survival despite the phage attack.

In the case of capsular strains, the cross-validation of these models showed poor model fit and accuracy (Table S8). These findings, along with our previous results, confirm that the metabolites detected using this approach may not be sufficiently relevant to explain significant changes in the metabolome of these strains. Focusing this metabolomics analysis on polar metabolites could yield more relevant results for these strains, especially considering the critical role of sugars in producing CPS.

4. Discussion

Despite being constantly exposed to external perturbing factors, such as dietary changes, antibiotic therapy or infectious diseases, the gut microbiome is characterized by its resilience to perturbations and its high temporal stability.^{49,50} This stability has also been reported for the gut virome together with a pronounced inter-individual variability, with virulent crAss-like phages being one of the most prevalent entities in the human gut virome capable of long-term co-existence with their Bacteroidales hosts.^{23,51} Here, we focused on four lytic gut phages infecting bacteria of the order Bacteroidales to identify possible new mechanisms for this co-existence. Two distinct phylogenetic phage groups were included in our analysis: crAss-like phages ϕ crAss001 infecting *B. intestinalis*,

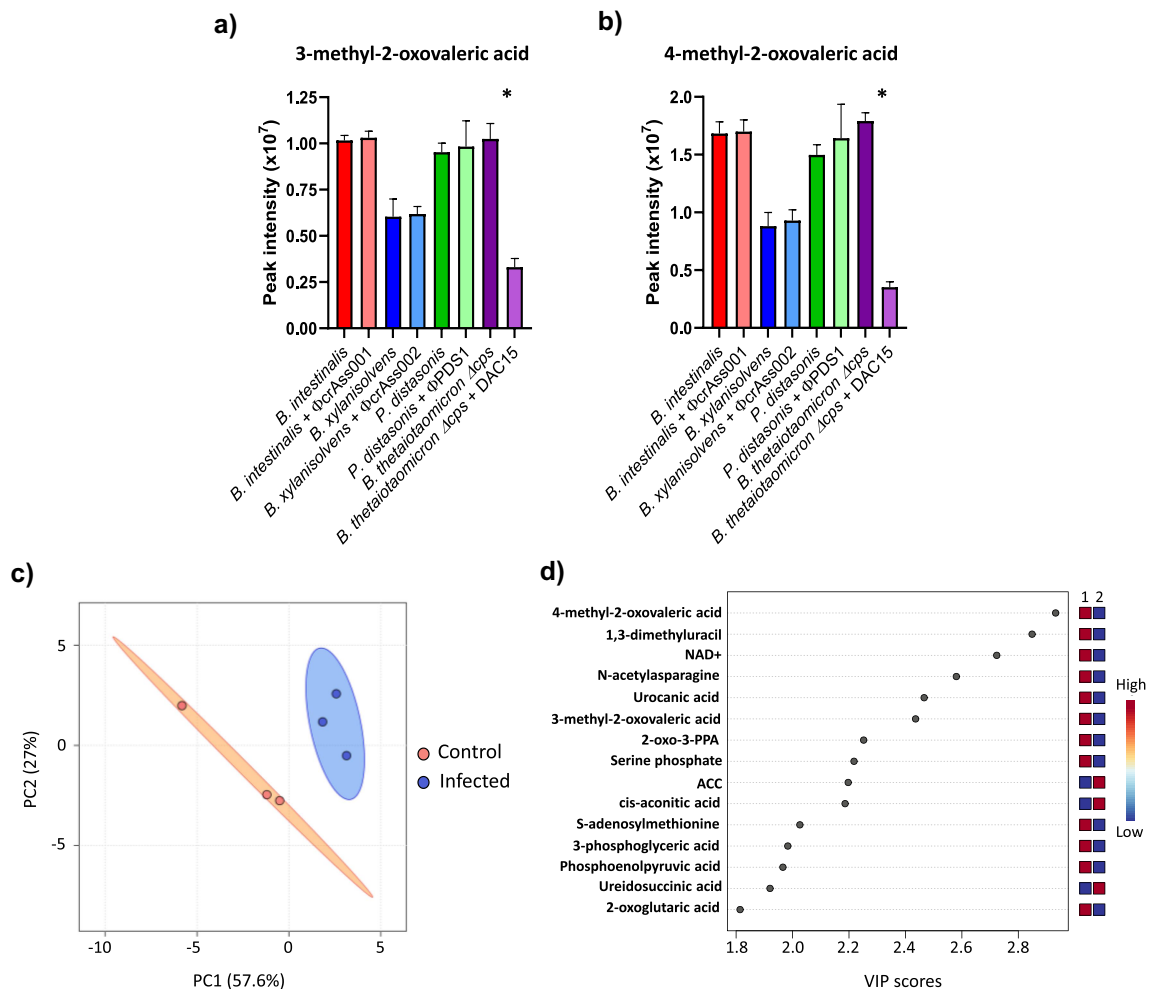


Figure 4. Intracellular metabolomic changes in *B. thetaiotaomicron* Δcps in response to DAC15 infection. (a-b) LC-MS peak intensities of 3-methyl-2-oxovaleric acid (a) and 4-methyl-2-oxovaleric acid (b) for each phage-host pair, obtained from untargeted intracellular metabolomics analysis performed at early-log phase of growth on the fifth day of the co-culture experiment. Bars represent the mean of triplicate biological samples, with error bars indicating the standard deviation (SD). * significant differences were set up at $p < 0.05$. (c) Principal component analysis (PCA) of intracellular metabolomic profiles from *B. thetaiotaomicron* Δcps and DAC15 co-cultures, based on compounds detected with accuracy levels 1 and 2a. Orange indicates the control condition (uninfected), while blue is the infected condition. Ellipses represent 95% confidence intervals, and the explained variances of the two principal components are shown in brackets along the axes. (d) important features identified by partial least squares-discriminant analysis (PLS-DA) in *B. thetaiotaomicron* Δcps bacterial pellets collected at the early-log phase on the fifth day of the co-culture. The coloured boxes on the right indicate the relative peak area of each corresponding metabolite in the groups under study, with red indicating higher peak areas and blue indicating lower. Group 1 corresponds to the control condition, and group 2 corresponds to the DAC15 infected samples. 2-oxo-3-PPA: 2-oxo-3-phenylpropanoic acid; ACC: 1-aminocyclopropanecarboxylic acid.

ϕ crAss002 - *B. xylanisolvens*, DAC15 - *B. thetaiotaomicron*, and siphovirus-like phage ϕ PDS1 - *P. distasonis*, as well as hosts with two different phenotypes (capsular and acapsular).

The expectation is that in a simple monoculture model like ours, lytic phages should exert a strong top-down control on the host population density favoring the emergence of resistant mutants. These mutants can completely or partially replace the ancestral sensitive population, depending on the cost of resistance, which in turn leads to the

reduction or complete extinction of the phage population.⁵² Our co-culture experiment revealed that phage persistence results in the co-existence of sensitive and resistant bacterial populations in an equilibrium, mainly due to phase variation of surface antigens. This allows bacterial survival due to the presence of resistant subpopulations, and phage propagation due to the continued presence of sensitive subpopulations. The initially low percentage of phage-resistant host cells increased to over 90% when in contact with the phages. A 100% resistant

population would not be expected given that the phages continued to propagate, so some of our results showing values close to 100% (Figures 1(e-h)) are explained by the limitations of plating assays. It is important to note that this resistance was not only effective against the initial phage stock, but also to phages that had evolved during the co-culture experiment, that is, no obvious coevolution of phages with their bacterial hosts during the experiment was observed. The changes observed were more attributable to bacterial adaptation rather than phage-host coevolution. This change in the proportion of phage-resistant host cells may explain why the growth kinetics of bacteria after 5 days of co-culture did not reveal any noticeable lysis. Since phages continued to propagate on the small sensitive subpopulations, this could not lead to detectable difference of growth rates or culture density.

The observation of stable co-existence in our phage-bacteria pairs can be explained by the eco-evolutionary model of fluctuating selection dynamics, where density-dependent fluctuating selection operates through a trade-off between the benefits of resistance and its associated metabolic costs,^{10,13} promoting prolonged co-existence of phage and host. Phage ecology is highly variable across different environmental microbiomes. For instance, in the ocean, phage-host pair dynamics are best explained by the ‘kill-the-winner’ model, with 20–40% of ocean bacteria being lysed by phage every day.⁵³ In contrast, in soil microbiomes, lysogeny is favored, with ecological dynamics being best explained by the ‘piggyback the winner’ model. Under this model, phages lysogenize their hosts during periods of high microbial abundance and growth rates, allowing them to multiply in parallel with their host.^{53,54} An extension of this model, accounting for the absence of true lysogeny, could also explain some of the behavior observed in our study. Recently, the existence of phage P1-like vegetative replication without producing phage progeny has been proposed for the prototypical crAssphage – the founding member of the order *Crassvirales*.⁵⁵ It is possible that phages like prototypical crAssphage and ϕ crAss002, both incapable of forming plaques in their host cultures, may instead form transient pseudo-lysogens by replicating their genomes via this vegetative replication

pathway. This model could explain the unusual behavior observed in ϕ crAss002 during our co-culture experiment, where phage numbers initially dropped sharply on day 1 but then increased from day 2 onward (Figure 1(b)). Given that the host population for this phage is largely resistant from the beginning, some phages could be entering a “dormant state” within the host, perhaps forming pseudo-lysogenic/carrier state relationships. We speculate that these phages could potentially replicate their genomes through vegetative replication, allowing for phage persistence without immediate host lysis. The increase in phage numbers after the subculture on day 2 could reflect the reactivation of these dormant phages. Thus, a combination of fluctuating selection dynamics (dynamic equilibrium between sensitive and resistant subpopulations) with ‘piggyback-the-winner’ dynamics (transient pseudolysogeny) could in theory be occurring in this phage-host system. Different eco-evolutionary models might be useful in describing certain aspects of phage-host co-existence in complex microbial communities such as the human gut microbiome.⁵³ However, even the very simplistic single phage-host pair systems used in this study highlight the complexity of phage-host interactions and the inadequacy of a simple model to capture them.¹³ The relevance of correctly capturing these dynamics lies in their potential involvement in microbiome alterations in human gut diseases such as inflammatory bowel disease, where bacterial phase variations driven by phages and host inflammation have been described to contribute to bacterial functional plasticity in disease states.⁵⁶

One characteristic of gut Bacteroidales is their ability to alter their surfaces dynamically to gain a survival advantage in their environments, protecting themselves from host and environmental factors.⁵⁷ This includes phase variation (expression of alternative types) of CPS driven by invertible promoters alternating between “ON” and “OFF” orientations.^{58,59} On top of that, NusG-like anti-termination factors,⁶⁰ and trans locus inhibitors⁴³ were proposed to ensure the expression of only one type of CPS at a time in any given cell. This mechanism has been reported to modify phage susceptibility and allow for prolonged phage persistence in some strains of *B. intestinalis* and

B. thetaiotaomicron.^{22,24} Our transcriptomics analysis confirms this previously proposed model and extends it to two new phage-host pairs in *B. xylanisolvens* and *P. distasonis*, where operons associated with alternative CPS types showed the strongest transcriptional responses, with some being upregulated (protective) and others downregulated (permissive) (Figures 2(a-d)). Specifically, our results for *B. intestinalis* are consistent with those obtained in Shkoporov et al. (2019),²² which reported strong repression of the PVR9 CPS operon and the upregulation of CPS operons associated with loci PVR7, PVR8 and PVR11 during the stationary phase of a phage-bacteria co-culture. All these loci share an IR sequence (GTTCTGTTTAA), flanking their putative invertible promoters. The only difference from this previous study is that PVR12, which also shares this promoter sequence, has been upregulated in our study, showing a stronger response than PVR8 and PVR11, while Shkoporov et al. (2019)²² did not observe any significant change for this locus. This difference could be attributed to the different timing of sample collection in the experiments, as well as the stochasticity of phase variation and the following selection of phage-resistant variants. When infected by their corresponding phages, *B. xylanisolvens* and *P. distasonis* showed down- and upregulation of CPS operons, mediated by invertible promoters flanked by conserved IR sequences (GTTACTTCTTAGGTAACGGA for *B. xylanisolvens* and GCTACTYRGNRAGTAGC for *P. distasonis*). One remarkable difference between these IR sequences and the one from *B. intestinalis* is that they contain internal palindromic regions and are preceded by integrase or invertase enzymes, which suggests that while the overall mechanism is broadly similar, specific variations might be present. When compared to other *Bacteroides* species, such as *B. thetaiotaomicron* or *B. fragilis*,^{24,46} these IR sequences are more similar and even share a common pattern involving a short fragment (GTTAC), with *P. distasonis* replacing the first thymine with a cytosine. This indicates that the *B. intestinalis* system differs more from those of the other strains. Additionally, these palindromic IR sequences were found not only near CPS operons but also in other genomic regions, where they were also preceded by recombinases,

indicating a conserved regulatory mechanism dependent on site-specific recombination.

Common elements are found at the beginning of most of these CPS loci, including genes encoding proteins of the UpxY and UpxZ families.^{43,60} While UpxY-like proteins positively regulate transcription of their respective CPS biosynthesis operon by preventing premature transcription termination in the untranslated region,⁶⁰ UpxZ-like proteins repress transcription of non-cognate CPS.⁴³ In the case of *P. distasonis*, only two of the twelve CPS loci have a copy of a UpxYannotated. The rest of the CPS loci are shown to contain transcriptional regulator/anti-termination protein NusG. Although these transcriptional factors may serve a function similar to that of the UpxY system, as genomic analysis of other *P. distasonis* strains has shown,⁶¹ alternative NusG-mediated mechanisms of transcriptional regulation are also possible. CPS loci are usually preceded by recombinase genes and invertible promoters, turning them into hotspots for site-specific recombination and enabling dynamic ON/OFF switching of their expression, which in turn regulates the permissiveness to phage infection. Therefore, gene clusters downregulated during phage infection may be responsible for susceptibility to phages, while those that are upregulated could have a protective or neutral function. Such dynamic equilibrium between resistant and sensitive bacterial subpopulations helps to maintain phage persistence. However, this is not the only mechanism utilized by *Bacteroidales* to facilitate phage persistence, as evidenced by our observations during the propagation of the crAss-like phage DAC15 in an acapsular *B. thetaiotaomicron* strain. A diverse transcriptional response after DAC15 infection was observed. BT_RS09760 (also labelled as BT1927) was the most highly upregulated gene, located near a site-specific integrase and previously reported to encode an S-layer protein homologue, one out of several phase-variable homologues encoded in the *B. thetaiotaomicron* genome.⁴⁷ The authors described that in a wild type *B. thetaiotaomicron* culture, the expression of BT1927 was very low (~1:1,000 cells), but when induced, it increased bacterial resistance to complement-mediated killing. Our results reflect a strong upregulation after DAC15 infection in the acapsular strain, both at the level of mRNA and protein, with the protein being present in both the cell pellet and the culture supernatant. This suggests the

existence of at least one alternative phase-variable adaptative mechanism of phage resistance in *Bacteroides*, when multiple alternative CPS are no longer available. Another study also observed upregulation of a different S-layer protein homologue when the strain was infected with a different phage, ARB25.²⁴ Similarly to the mechanism of CPS phase variation, the S-layer operons are also preceded by promoters flanked by consensus IR sequences (CCGTTACCTABVRAGTAAC), which were detected in other 7 upregulated loci next to site-specific integrase, S-layer surface protein and OmpA family protein genes, as previously reported.²⁴ This IR sequence also has a small internal repeat with the common element CGTTA shared with the IR sequences of the CPS operons in the other *Bacteroides* species previously discussed. In fact, four of the CPS operons (CPS1-BT_RS01830, CPS3-BT_RS02920, CPS5-BT_RS08400, CPS6-BT_RS08750) in *B. thetaiotaomicron* wild type contain a nearly identical IR sequence (CCGTTACCTAAGAAGTAAC) flanking their invertible promoter and neighbored by tyrosine-type DNA invertases. This finding shows that the mechanism of phase variation of these different cell surface operons might be similarly regulated, supporting phage persistence even in the absence of CPS. The highly diverse transcriptional responses observed when alternative CPS cannot be switched on suggest the existence of multiple layers of cell surface-related phage resistance mechanisms, each of which being perhaps less efficient than CPS switching. But acting together, they can produce a sufficient quantitative effect to allow the population to survive in the presence of phage. This hypothesis can be further supported by the longer time required for this strain to achieve a high percentage of resistant cells compared to the capsulated strains (Figures 1 (e-h)). The availability of the engineered acapsular strain allowed us to investigate alternative adaptive responses to phage infection, apart from CPS phase variation. While including the wild type *B. thetaiotaomicron* strain could have been insightful, previous comprehensive studies by Porter et al. (2020) have already provided a detailed characterization of the role of multiple capsules in phage resistance in this strain.²⁴

The differential expression of alternative CPS types may also cause some intermediate metabolism changes in the bacterial population under attack

from phage. Here, we analyzed the potential metabolic changes associated with persistent phage infection in populations of *Bacteroides* spp. host strains. Previous studies that have investigated changes in the metabolome profile caused by phage predation have been conducted with *in vivo* models, focusing on how it affects the gut microbiome composition and the gut metabolome of mammalian hosts,^{62,63} rather than on the impact on the individual bacterial host. The acapsular strain of *B. thetaiotaomicron* showed the highest differences in its intracellular metabolome. Most of the affected metabolites were related to amino acid metabolism, including isoleucine, leucine, phenylalanine and histidine. This was exemplified by the reduction of 3-methyl-2-oxovaleric acid, 4-methyl-2-oxovaleric acid, 2-oxo-3-phenylpropanoic acid, and urocanic acid, which are intermediate compounds in the degradation pathways of these amino acids.^{64,65} This reduction may indicate that amino acid degradation, which can be used as a source of energy by bacterial hosts, was downregulated in the presence of phage. Another relevant change was the DAC15-driven depletion of NAD⁺, an essential co-factor of numerous redox reactions. The depletion of NAD⁺ has been previously described as a defense mechanism used against phage infection,^{7,66–68} depriving the phage of this essential co-factor and impeding phage propagation. It appears that the absence of CPS forces bacteria to use alternative mechanisms to defend against phage, which may explain these changes. On the other hand, the capsulated strains showed fewer changes in their metabolome, possibly due the ability to deploy a less costly type of phage resistance mechanism – the alternative CPS types. Also, the technique used for our analysis was focused on semi-polar compounds. This may have resulted in certain relevant compounds not being extracted, such as those with high polarity, including polar sugars serving as intermediates in CPS biosynthesis.

5. Conclusions

This study represents an additional step in our understanding of the mechanisms of the long-term persistence of virulent bacteriophages in their host cultures, including the persistence of crAss-like and non-crAss phages in *Bacteroides* and *Parabacteroides*. Phase-variable CPS expression plays an essential role in the

bacterial ability to co-exist with phages. Other gene products, including but not limited to the alternative types of S-layer proteins, regulated by phase variation also appear to enable efficient co-existence with phages, suggesting cooperative relationships between different mechanisms of resistance. In the absence of CPS, higher transcriptomic, proteomic, and metabolomic changes are observed since more factors are involved in achieving the equilibrium between bacterial and host populations. Further research is necessary to delve deeper into these aspects, enhancing our understanding of phage-bacteria interactions in the crowded environment of the gut.

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Disclosure statement

No potential conflict of interest was reported by the author(s).

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Conceptualization: A.N.S. and C.H.; Methodology: A.C.M, A. N.S, C.H., M.K. and R.P.R.; Investigation: A.C.M, C.B., J.L. M. and C.A.T.; Formal Analysis: A.C.M, C.B. and J.L.M.; Writing – original draft: A.C.M.; Writing – review and editing: A.C.M., A.N.S., C.H. and C.B.; Supervision: A.N.S., C. H. and M.K.; Project administration: L.A.D.; Funding acquisition: A.C.M., A.N.S., C.H., M.K. and R.P.R. All authors read and approved the final manuscript.

Data and material availability statement

All data generated or analyzed during this study are included in this published paper and its supplementary material. Raw sequencing data for transcriptomics are available from NCBI Sequence Read Archive (SRA) repository under BioProject PRJNA1131275. Proteomics data including raw mass spectrometry files and databases are available from ProteomeXchange Consortium via the PRIDE⁷⁰ partner repository with the dataset identifier P×D057336. Phages ϕ crAss001 (DSM 109066) and ϕ PDS1 (DSM 112067), and their hosts *B. intestinalis* APC919/174 (DSM 108646) and *P. distasonis* APCS2/PD (DSM 112022) are available from the DSMZ culture collection. Phage ϕ crAss002 and its host *B. xylanisolvens* APCS1/XY are available upon request from our research group. Phage DAC15 and the *B. thetaiotaomicron* VPI-5482 capsule mutant are available on request from Prof. A.J. Hryckowian (University of Wisconsin, USA).

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