

https://doi.org/10.1093/ismejo/wrae005 Advance access publication: 14 January 2024 Original Article

Escherichia coli CRISPR arrays from early life fecal samples preferentially target prophages

Moïra B. Dion^{1,2}, Shiraz A. Shah³, Ling Deng⁴, Jonathan Thorsen^{3,5}, Jakob Stokholm^{3,4}, Karen A. Krogfelt^{6,7}, Susanne Schjørring⁶,

Philippe Horvath⁸, Antoine Allard^{9,10}, Dennis S. Nielsen⁴, Marie-Agnès Petit¹¹, Sylvain Moineau^{1,2,12,*}

¹Département de biochimie, de microbiologie, et de bio-informatique, Faculté des sciences et de génie, Université Laval, Québec, QC G1V 0A6, Canada

²Groupe de recherche en écologie buccale, Faculté de médecine dentaire, Université Laval, Québec, QC G1V 0A6, Canada

³Copenhagen Prospective Studies on Asthma in Childhood, Herlev and Gentofte Hospital, University of Copenhagen, Ledreborg Alle 34, 2820 Gentofte, Denmark ⁴Food Science, University of Copenhagen, Rolighedsvej 26, 1958 Frederiksberg, Denmark

⁵Novo Nordisk Foundation Center for Basic Metabolic Research, Faculty of Health and Medical Sciences, University of Copenhagen, Blegdamsvej 3B, 2200 Copenhagen, Denmark

⁶Department of Bacteria, Parasites and Fungi, Statens Serum Institut, Artillerivej 5, 2300S Copenhagen, Denmark

⁷Department of Science and Environment, Roskilde University, Universitetsvej 1, 4000 Roskilde, Denmark

⁸IFF Danisco, Health & Biosciences, Dangé-Saint-Romain 86220, France

⁹Département de physique, de génie physique et d'optique, Université Laval, Québec, QC G1V 0A6, Canada

¹⁰Centre interdisciplinaire en modélisation mathématique, Université Laval, Québec, QC G1V 0A6, Canada

¹¹Université Paris-Saclay, INRAE, AgroParisTech, Institut Micalis, Jouy-en-Josas 78350, France

12 Félix d'Hérelle Reference Center for Bacterial Viruses, Faculté de médecine dentaire, Université Laval, Québec, QC G1V 0A6, Canada

*Corresponding author: Sylvain Moineau, Département de biochimie, de microbiologie, et de bio-informatique, Faculté des sciences et de génie, Université Laval, Québec, QC G1V 0A6, Canada. Email: sylvain.moineau@bcm.ulaval.ca

Abstract

CRISPR–Cas systems are defense mechanisms against phages and other nucleic acids that invade bacteria and archaea. In Escherichia coli, it is generally accepted that CRISPR–Cas systems are inactive in laboratory conditions due to a transcriptional repressor. In natural isolates, it has been shown that CRISPR arrays remain stable over the years and that most spacer targets (protospacers) remain unknown. Here, we re-examine CRISPR arrays in natural *E*. coli isolates and investigate viral and bacterial genomes for spacer targets using a bioinformatics approach coupled to a unique biological dataset. We first sequenced the CRISPR1 array of 1769 *E*. coli isolates from the fecal samples of 639 children obtained during their first year of life. We built a network with edges between isolates that reflect the number of shared spacers. The isolates grouped into 34 modules. A search for matching spacers in bacterial genomes showed that *E*. coli spacers almost exclusively target prophages. While we found instances of self-targeting spacers, those involving a prophage and a spacer within the same bacterial genome were rare. The extensive search for matching spacers also expanded the library of known *E*. coli protospacers to 60%. Altogether, these results favor the concept that *E*. coli's CRISPR–Cas is an antiprophage system and highlight the importance of reconsidering the criteria use to deem CRISPR–Cas systems active.

Keywords: CRISPR, phage, bacteriophage, E. coli, virome, gut, microbiome, phage resistance

Introduction

Clustered Regularly Interspaced Short Palindromic Repeats (CRISPR–Cas) systems are defense mechanisms found in numerous bacterial and archaeal genomes. Cells carrying active CRISPR– Cas systems are protected against phages, plasmids, and other invasive genetic materials. The system is based on recognizing and then cutting of foreign genetic sequences [1, 2]. This occurs when spacers, which are short sequences present in the CRISPR array, are identical to regions (called protospacers) in the invading genetic material. Upon infection, transcribed CRISPR arrays guide Cas nuclease proteins toward protospacer sites in the phage or plasmid genome for DNA or RNA cleavage, preventing infection and making the host cell resistant [2, 3]. The structure of the CRISPR array is usually adaptive and dynamic because new spacers can be acquired over time (mostly at the 5' leader end of the array) or native spacers can be deleted.

CRISPR-Cas systems are highly diverse in terms of organization, prevalence, and activity. They are currently organized into two classes, six types and several subtypes based on the architecture of the genomic loci and the composition of cas genes [4]. Because prokaryotes carry several additional defense mechanisms, their reliance on CRISPR-Cas to evolve phage resistance varies depending on environmental conditions. It has been reported that the presence of CRISPR in microbes correlates negatively with the oxygen level and positively with temperature [5]. Other biotic factors that can be used to predict CRISPR prevalence in microbial ecosystems include viral abundance and diversity [6]. For example, in mixed culture, Pseudomonas aeruginosa preferentially uses CRISPR-Cas to defend against phages, while in pure culture, phage receptor mutants are favored [7]. When this bacterial species encounters bacteriostatic antibiotics, the proportion of CRISPR-mediated resistant cells also increases [8].

© The Author(s) [2024]. Published by Oxford University Press on behalf of the International Society for Microbial Ecology.

Received 26 December 2023. Revised: 9 January 2024. Accepted: 10 January 2024

This is an Open Access article distributed under the terms of the Creative Commons Attribution License (https://creativecommons.org/licenses/by/4.0/), which permits unrestricted reuse, distribution, and reproduction in any medium, provided the original work is properly cited.

Escherichia coli strains may harbor multiple CRISPR arrays within their genomes, and analyses of the associated cas genes indicated that these CRISPR-Cas systems belong to Subtypes I-E or I-F [9]. Both subtypes are rarely found in the same genome [10], and most E. coli strains only have Types I-E CRISPR-Cas system. Unfortunately, the nomenclature for E. coli CRISPR arrays is inconsistent throughout the literature. There are two arrays associated with the Types I-E system: CRISPR1 (or CRISPR 2.1), located downstream of the iap gene, and CRISPR2 (or CRISPR 2.2 and CRISPR 2.3), located between genes ygcE and ygcF [11, 12]. In some strains, a 0.5-kb AT-rich sequence splits the CRISPR2 array, hence the alternative CRISPR 2.2-2.3 nomenclature. CRISPR1 and CRISPR2 are separated by \sim 20 kb and only CRISPR1 has associated cas genes. There are also two arrays associated with the Types I-F system, designated as CRISPR3 (or CRISPR 4.1) and CRISPR4 (or CRISPR 4.2). However, they are not as well studied, but they are less diverse and less prevalent than CRISPR1 and CRISPR2.

In the E. coli K12 laboratory strain, the Types I-E CRISPR-Cas system is repressed by the histone-like nucleoid-structuring protein [13], which is a global transcriptional repressor. The system can, however, be made active by genetically engineering it to efficiently acquire new spacers as well as block phage infection and plasmid transformation through DNA interference [3, 14]. In natural isolates, the activity of the system has not been demonstrated [11]. Although the spacers are highly diverse, some E. coli CRISPR arrays have been shown to even remain stable for 42000 years [15], suggesting that the system rarely acquires new spacers. In fact, it appears that most of the CRISPR diversity in strains of the serotype E. coli O157:H7 is driven by spacer deletion rather than acquisition [16]. However, there is a negative correlation between the presence of CRISPR arrays and the pathogenic potential of E. coli strains [10]. Given the role that CRISPR-Cas systems play in immunity, an active system would likely prevent the acquisition of new genes coding for virulence factors via horizontal gene transfer. Thus, some environmental conditions might favor the activity of the CRISPR-Cas system in E. coli.

Here, we evaluated the CRISPR diversity in *E. coli* isolates that originate from a large collection of fecal samples from children under 1 year old. Using also the viral and bacterial metagenomes from the same samples and reference databases, we found targets for 60% of the spacers, significantly expanding the library of known targets and revealing that *E. coli* spacers preferentially target prophages. Combined with rare events of prophage-targeting spacers being present in the same genome as their targets, these results suggest that *E. coli* CRISPR1, despite a low activity of spacers acquisition, exhibits an antiprophage interference in vivo.

Materials and methods Bacterial strains

Escherichia coli isolates were isolated from the fecal samples of 648 children enrolled in the Copenhagen Prospective Studies on Asthma in Childhood 2010 (COPSAC2010) mother-child cohort [17]. Fecal samples were diluted and plated in aerobic, microaerophilic, and anaerobic conditions on nonselective and selective media [18]. Bacterial identification was confirmed biochemically, as described previously [18]. Altogether, 348, 467, and 954 *E. coli* isolates were obtained from fecal samples obtained 1 week, 1 month, and 1 year after birth, respectively, from 639 children. The 1769 *E. coli* isolates were stored at -80° C in 20% glycerol at the Statens Serum Institut (Copenhagen, Denmark) in 96-well plates.

PCR and Sanger sequencing

Isolates were first transferred in fresh TSB medium in 96-well plates and were incubated at 37°C overnight. To screen for the CRISPR1 locus of the *E. coli* strains, primers 5'-GATGGGTTTGAAA ATGGGAGCTGGG-3' and 5'-AGACGTATTCCGGTGGATTTGGATGG-3' were used. These primers anneal the *iap* and *cas2* genes, respectively. PCR amplification was performed with the *Taq* polymerase (Bio Basic) using the following program: 2 min at 95°C, followed by 35 cycles at 95°C for 20 s, 58°C for 40 s, 72°C for 2 min, and then 1 cycle at 72°C for 5 min. To estimate the amplicon size, 5 μ l of PCR product was migrated on a 2% agarose gel. PCR products were then sent for Sanger sequencing (CHUL sequencing platform, Quebec City) with the same primers.

Assembly and CRISPR identification

Forward and reverse nucleotide sequences were assembled using Geneious v11.1.5 and the De Novo Assemble tool. When sequences failed to assemble or the sequence quality was poor, custom internal primers (Supplementary Table S1) were designed to perform PCR and Sanger sequencing using the same program. Assembled sequences were then exported in FASTA format. CRISPR arrays were identified with CRISPRDetect v2.2 [19], using default parameters, except for the -array_quality_score_cutoff, which was set to 3.

Bioinformatics analyses

Bioinformatics analyses were conducted in a Jupyter notebook using Python3 packages and software mentioned in Supplementary Table S2. CRISPRStudio [20] was used to produce a colorcoded figure of the CRISPR array, with default parameters. Given the size of the dataset, the full color-coded figure was neither practical nor informative as a main text figure. Instead, we used a network representation to illustrate the diversity of spacers and the interconnectivity of the CRISPR arrays. First, a graph was created by identifying each isolate with a node and by linking any two isolates if they shared at least one spacer. The magnitude of each connection was measured using three similarity indices: binary Jaccard, weighted Jaccard, and Tanimoto. We compared the similarity index distributions and found they only marginally differed (see Supplementary Fig. S1). We continued our network construction with similarity values from the binary Jaccard index. The same steps were performed on 500 random datasets, which consisted of spacers randomly distributed among all isolates, with each isolate keeping the number of spacers they had in the original dataset. We looked at the distribution of the value of the similarity indices and compared them to the original dataset and estimated a 95th percentile, i.e. a value under which 95% of the similarity index measured in the random datasets are found. This estimation provided a threshold (0.12) above which any connection in the original dataset were considered as significant (see Supplementary Fig. S2). We then removed any connection with a similarity index below 0.12 to build the final network. The Infomap [21] algorithm was used to extract modules. Cytoscape v3.9.0 was used to edit the network and color the modules.

To distinguish known spacers from new ones, we compared spacers with the CRISPR Spacer database [22] using blastn v2.9.0 [23]. Spacers were considered to be known when they matched (0 mismatch, 100% coverage) a spacer in the database; otherwise, they were considered to be new. Several databases and datasets were queried for the presence of spacer targets (protospacers). These include the NCBI Virus database (filtered for Virus; bacteriophage, Nucleotide completeness; complete, downloaded on 19 April 2021), COPSAC viromes from 1-year fecal samples [24], 46 COPSAC coliphages [25], metagenome-assembled genomes (MAGs) from COPSAC metagenomes from 1-year fecal samples [26], and the NCBI nt database for bacterial genomes (downloaded on 8 October 2021). To determine homology for results from the NCBI Virus database, COPSAC viromes, and COPSAC coliphages, both fasta36 and blastn were performed. For the bacterial genomes and COPSAC metagenomes, only blastn was used. A sequence was considered as a target when the alignment showed at most four mismatches on the full length of the 32 nucleotide-long spacer and zero gap. To avoid misidentifying spacers in CRISPR arrays as targets in COPSAC metagenomes and bacterial genomes, we removed all hits that were within 100 bp of the *E. coli*-conserved CRISPR repeat sequence "5"-CGGTTTATCCCCGCTGGCGCGGGGAACAC-3'.

The presence of prophages within the regions targeted by spacers in the bacterial genomes and COPSAC metagenomes were determined using a three-step approach. First, for bacterial genomes, we determined the gene function of every gene targeted by a spacer using the efetch() function from the Bio.Entrez Python package. Second, for each bacterial genome and COPSAC metagenome targeted by at least two spacers, the median absolute deviation (MAD) was measured. The MAD was calculated according to the position of all the spacer targets on the genome. Third, using only bacterial genomes and COPSAC metagenomes with at least 10 spacer targets, we extracted the putative prophage regions by identifying the minimum and maximum genomic positions of the spacer targets and adding 25 kb both upstream and downstream of the targets. The nucleotides of these putative prophages were pairwise compared with the phages from the NCBI Virus database, COPSAC viromes, and COPSAC coliphages with at least one spacer target using blastn to deduplicate identical sequences. Sequences were deemed to be identical when they shared 95% identity over 85% coverage. Identical sequences were then clustered using MCL [27]. One sequence per cluster was kept as a representative for the following steps. VIBRANT v1.2.1 [28] was used to confirm viral identity and type (virulent/temperate) using the default parameters, except with the -virome argument on. The viral type was manually confirmed by searching for the presence of temperate markers (mainly the serine recombinase, which is overlooked by VIBRANT). The network representation of confirmed prophages and phages were generated with the NetworkX Python package and edited in Cytoscape v3.9.0. The genome map of a representative prophage was done with the VIBRANT annotations, confirmed with the annotations of the bacterial genome on NCBI and visualized with EasyFig v2.2.5 [29].

To study the prevalence of self-targeting events, we selected E. coli-complete genomes with at least one CRISPR locus (n = 1014) from NCBI. Their corresponding spacers and repeats were extracted from the CRISPR Spacers database. Homology searches between the spacers and the genomes as well as between the repeats and the genomes were performed using blastn. The presence of spacers, both in the CRISPR array (according to the CRISPR Spacers database and if they were within 100 bp of a repeat sequence) and elsewhere in the genome, were considered to be indicative of a self-targeting event. Some genome files contained plasmid sequences. When a spacer was present only in the CRISPR array of the bacterial chromosome and in the plasmid, but not elsewhere, it was not considered to be a selftargeting event. Self-targeting events were identified in a total of 99 genomes (out of 1014, 9.8%). To verify that the self-targeting spacer targeted a prophage within the same chromosome, we checked for the presence of prophages in the 99 genomes using

PHASTER [30] and compared the prophage positions with those of the self-targeting events.

Results

Using PCR, we amplified and then Sanger-sequenced the CRISPR1 loci of 1769 E. coli isolates obtained from 639 fecal samples from children enrolled in the COPSAC2010 study. A CRISPR1 array was successfully sequenced for 1048 (59%) isolates, with arrays containing 2-32 spacers (average = 11 spacers). Then, we sought to evaluate the diversity within CRISPR arrays (spacer content) and the interconnectivity of the isolates. Traditional phylogenetic trees are poorly suited for CRISPR arrays because the sequence lengths are highly variable (due to the different numbers of spacers) and this region is prone to recombination events. The alternative approach of representing CRISPR arrays with a colorcoded figure was impractical, given the size of the dataset. Thus, we used a network-based approach to visualize the diversity of CRISPR arrays (Fig. 1). For each pair of CRISPR arrays (each CRISPR array is an isolate), a Jaccard similarity index was calculated based on the number of shared spacers, which was then used to generate a network, as shown in Fig. 1A. To group CRISPR arrays into modules, we used Infomap [21], which allowed arrays with slightly different spacer content to still cluster together. This resulted in 34 delimited CRISPR modules. Nine of them were singletons. The remaining 25 modules contained 2–408 CRISPR arrays (Fig. 1B). To appreciate the diversity of CRISPR arrays, we illustrated the spacer content of one randomly selected representative for each module (Fig. 1B).

To explore the CRISPR diversity further, we studied the spacers in the E. coli isolates. A total of 12298 spacers were extracted from all 1048 CRISPR1 arrays. These spacers corresponded to 946 unique sequences (referred to as spacer clusters) based on sequence homology. Spacer clusters contained 1–279 spacers. We were particularly interested in the specificity of the spacers, such as whether they are specific to the early life gut environment as well as the identity of their targets. We probed several databases for sequence homology. First, we compared our 946 spacer clusters with known spacers found in E. coli and in the NCBI database. Only 22% (n = 210) of the spacer clusters were considered to be new (no homology with previously sequenced spacers), whereas 78% (n = 736) of the spacer clusters were already sampled in published E. coli genomes. Spacer clusters with fewer spacers more often included new spacers: the 210 new spacer clusters only comprised 631/12 298 spacers, 5% of the total spacer dataset (Fig. 2A). As new spacer acquisition is mostly polarized at the 5' end of the CRISPR array, we compared the mean relative position along the 5'-3' CRISPR array axis of known and new spacers (Fig. 2B). We found no significant difference (Kolmogorov–Smirnov two-sample test, P-value = 0.1076) as new and rare spacers were not concentrated at the 5' end of the CRISPR array. Since new spacers do not appear to have been acquired recently, we concluded that they are not specific to the environment sampled in this study and rather represent rare E. coli spacers that had yet to be sampled.

To identify targets (protospacers), we first searched for homologies with phage genomes from three sources: (i) all NCBI phage genomes, (ii) 1-year viral metagenomes obtained from the same children as our isolates, and (iii) a collection of sequenced coliphages amplified from the isolate supernatants [25]. Spacers rarely matched sequences from viral sources. There were 34 spacer clusters that matched 93 phages that infect mostly *Salmonella* and *Escherichia* (see Supplementary Table S3). The most frequently targeted known phages were Escherichia phage P1,

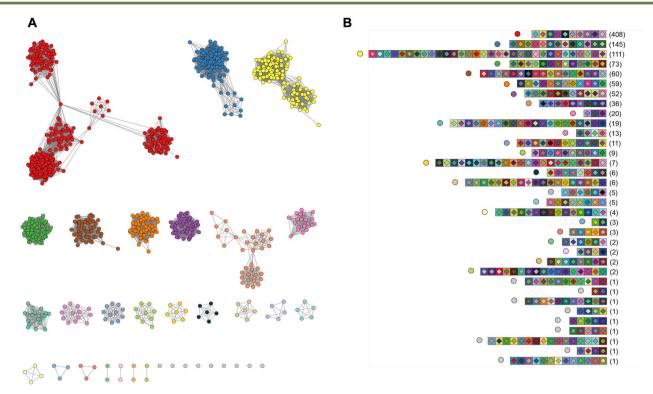


Figure 1. CRISPR diversity in *E. coli* isolates; (A) each node represents a CRISPR array (isolate) and each edge represents a level of shared spacers between two nodes; the length of the edge is indicative of the Jaccard similarity index; the color of the node represents the module the CRISPR array belongs to; (B) one random CRISPR array per module is illustrated; each colored square corresponds to a spacer; the colored dot next to the array refers to the module it represents; two spacers with the same diamond–square color combination share homologous sequences; the numbers in parentheses correspond to the number of CRISPR arrays in each module.

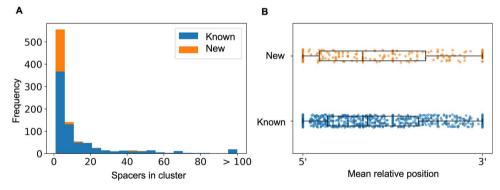


Figure 2. Specificity of the spacers to the early life gut environment; (A) distribution of the number of spacers per cluster, showing known (blue, identical to previously sampled spacers) and new (orange, unique to our dataset) clusters; (B) mean relative spacer position for each spacer cluster along the 5′–3′ axis, grouped according to known (blue, bottom) and new (top, orange) clusters.

Escherichia phage RCS47 and Salmonella phage SJ46 with seven related-protospacers. There were 31 spacer clusters that matched 37 viral contigs from viral metagenomes. In a previous study [24], we performed a host prediction for these viral contigs. *Enterobacteriaceae* was predicted for 14/37 (38%) viral contigs and there was no matching prediction for the others. Lastly, only three spacer clusters targeted three coliphages previously isolated from the same fecal samples (Escherichia phage Evi, LR597642.1; Escherichia phage ESSI2_ev239, NC_049392.1, and Escherichia phage mEp460_ev081, LR597641.1). In total, 133 phage protospacers matched the 74 spacers.

We next explored the NCBI bacterial genomes as well as bulk 1-year metagenomes from the same children as our isolates. Surprisingly, most of these unknown targets could be identified by searching for spacer matches in bacterial genomes. After removing any matches with spacers in CRISPR arrays, 474 spacer clusters were found to target 9321 bacterial genomes from the NCBI database. Roughly half were *E.* coli genomes and the rest were genomes from members of the *Enterobacteriaceae* family (*Klebsiella*, n=2031; *Salmonella*, n=1320; *Enterobacter*, n=365; *Citrobacter*, n=265; *Shigella*, n=202). Of note, the most targeted genome was *E.* coli strain RHB42-C16 (CP056933.1), with spacers targeting 87 different regions (protospacers). We also identified numerous spacer targets in MAGs from the COPSAC 1-year metagenomes: 256 spacer clusters targeted 1702 bacterial contigs. Combined with spacers targeting sequences of viral sources, this reduced the percentage of unknown targets from 95% to 41% (392/946). This apparent preference for targets in prophages could be an artifact of the more exhaustive sequence availability for bacterial genomes, where spacer targets in prophages are identified, compared to phage genomes. Bacterial genomes are >10 times larger and are more frequently found than phage genomes on NCBI. Still, we find the same inclination for temperate phages when we analyzed spacers targeting NCBI phages only. To evaluate if a bias for temperate phages exists, we compared the percentage of temperate phages on NCBI with the percentage of temperate phages targeted by E. coli CRISPR spacers. NCBI phages were chosen for this analysis because their genomes are complete, making it possible to accurately determine their lifestyle (whereas, viral contigs from the virome dataset could be incomplete). Since there may be different percentages depending on the bacterial host, we focused on phages that infect Salmonella, Escherichia, and Enterobacteria (n = 2437). These three hosts represent 80% of the targeted phages' hosts (see Supplementary Table S3). To avoid any overrepresentation, sequences were deduplicated, which resulted in 1243 genomes. We then ran VIBRANT to predict the lifestyle for each phage. VIBRANT predicted that 1058 and 175 genomes were virulent and temperate, respectively. Thus, temperate phages represent roughly 14% of all phages that infect Salmonella, Escherichia, and Enterobacteria. In comparison, 100% of the phages targeted by spacers (the subset of phages that infect the same three hosts) were temperate. Despite temperate phages being a minority in the NCBI Virus database, they represent the totality of phages being targeted by E. coli CRISPR spacers.

We also investigated the targeted genes, and several phageassociated genes were identified to be among the 15 most prevalent (Fig. 3A), such as genes coding for portal proteins, tail proteins, and major capsid proteins. We then examined whether spacers preferentially target prophages in bacterial genomes and MAGs from COPSAC 1-year metagenomes. For each genome, the MAD was calculated using the genomic position of each target. The MAD is a dispersion measure, which tells us about the deviation around the median of a dataset. The smaller the MAD is, the more condensed the data are around its median and conversely, the bigger the MAD is, the more scattered the data are. In this context, the data are the position of each spacer target in a bacterial genome. This metric was used as a proxy to detect prophage regions in bacterial genomes because we hypothesized that genomes with closely located targets (low MAD values) were indicative of a prophage region. The rationale behind this is that if E. coli's CRISPR-Cas system targeted bacterial genomes, we would find spacer targets randomly distributed across the genome, resulting in a high MAD. Conversely, if E. coli's CRISPR-Cas is an antiphage system and we find targets in bacterial genomes, they would be concentrated in specific regions, where prophages are located. This would result in a low MAD. When looking at the relationship between the number of targets and the MAD (Fig. 3B), two distinct groups emerged: genomes with few targets (<10) and high MAD values and genomes with many targets (\geq 10) and low MAD values (Fig. 3B). A total of 192 genomes (144 E. coli genomes, 17 other bacterial genomes, and 31 contigs that belong to 9 MAGs) fell into the latter category and were screened to confirm the presence of a prophage in the region encompassing spacer targets. After analyzing the sequences using VIBRANT, we determined that only 3/192 genomes did not carry a predictable prophage. There was high sequence redundancy because 126 prophages (out of 189) belonged to the same species (Fig. 3C, large cluster at the top left corner, 95% identity over 85% coverage). Together with these 189 prophages, the 133 sequences of viral origin (NCBI phages, COPSAC 1-year viromes, and COP-SAC coliphages) were used to evaluate the sequence diversity and determine whether these phages were temperate or virulent (Fig. 3C). The sequence comparison revealed that some viral contigs that were identified in the viral metagenomes were identical to NCBI phages and prophages in bacterial genomes (Fig. 3C, second largest cluster). Overall, we concluded that spacers preferentially target prophages (and temperate phages), as only 18/325 phages were considered to be virulent according to VIBRANT predictions. Genomic organization and annotation of a prophage that is representative of the large cluster are illustrated in Fig. 4 along with the regions targeted by spacers.

The prophage preference of E. coli CRISPR arrays prompted us to investigate whether the arrays were targeting resident prophages, or prophages not yet acquired by the cell. We searched for the evidence of self-targeting in E. coli, where a spacer and its target are both found in the same genome. We analyzed 1014 complete E. coli genomes from NCBI and considered a self-targeting event to have occurred when a spacer sequence was present both in the CRISPR array and elsewhere in the bacterial genome. There were 99 genomes where at least one self-targeting event was identified. That is an occurrence of ~10%. In nine of these genomes, selftargeting spacers targeted both the bacterial chromosome and one or two plasmids. Most importantly, self-targeting spacers directed at a prophage were rare, with only 18 events (out of 99) identified. This result is consistent with a role of E. coli CRISPR arrays in preventing the acquisition of new prophages. Another hypothesis that is also consistent with the results is that spacers are left-overs from a warfare with previously resident prophages that have now been excised due to the spacer match. Lastly, we investigated if there were instances where E. coli spacers and MAGs or viral contigs originated from the same child. There were four instances where a prophage in a MAG was targeted by a spacer that was found in the same child at the same sampling time (1 year old). We found no viral contigs and spacers that originated from the same child at the same timepoint. These results suggest that E. coli spacers rarely match viral invaders that are found in the same gut environment, although a limited sequencing depth could hinder the detection of such an occurrence.

Discussion

Our objective was to address the question of *E*. coli CRISPR–Cas systems' activity. To investigate this, we had access to a unique combination of biological data from early life fecal samples (bacterial isolates, bacterial and viral metagenomes, and isolated phages), which we analyzed from a bioinformatics perspective. We evaluated the CRISPR content in *E*. coli isolates originating from fecal samples from children within their first year of life. The percentage (59%) of CRISPR-positive isolates was consistent with isolates from other environments, including human, animal, and water sources [31], suggesting that CRISPR prevalence may not be environment-specific at least for *E*. coli. Next, we used a network-based approach to visualize CRISPR diversity. This allowed us to define CRISPR modules according to spacer content and examine network properties, reflecting the interconnectivity of the isolates.

We then studied *E.* coli CRISPR arrays at the spacer level. The ratio of total spacers to distinct spacer sequences (12 298/946) was consistent with other studies [11, 12, 15, 32]. Two different analyses led to the same conclusion: spacers are not specific to the early life gut environment. First, most of the spacers were identical to those carried by *E.* coli genomes in the NCBI database. Second, there were only rare instances of spacers targeting a viral sequence found in the same sample (same child and timepoint). Again, these results support the hypothesis that the Subtypes I– E CRISPR–Cas system in *E.* coli is inactive because the spacers

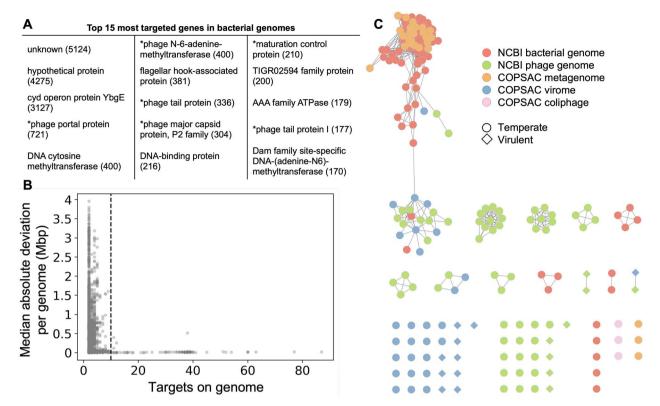


Figure 3. The *E*. coli spacers preferentially target prophages found in *Enterobacteriaceae* genomes; (A) list of the 15 most targeted genes in bacterial genomes; genes marked with an asterisk are strictly phage genes; (B) scatter plot of the number of targets and MAD for each bacterial genome targeted by spacers; the dotted vertical line at x = 10 corresponds to the minimum cut-off for bacterial genomes that were investigated for the presence of a prophage; (C) network representation of the viral sequences targeted by spacers; each node is a viral sequence, and each edge between two nodes signifies at least 95% identity over 85% coverage, which is the definition of a viral species; nodes are colored according to their origin (red, prophages in NCBI bacterial genomes; green, NCBI phage genomes; orange, prophage in COPSAC metagenomes; blue, phage in COPSAC X virome; and pink, coliphage from the COPSAC *E*. Coli supernatant), and their shapes (circle, temperate; diamond, virulent) are based on their replication mode.

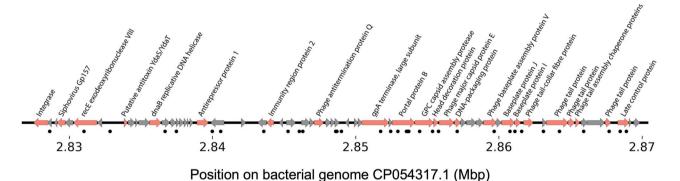


Figure 4. Genome map of a representative prophage of the large cluster in Figure 3; all arrows represent genes; the gray arrows are genes coding for hypothetical proteins or proteins of unknow functions; the black dots under the genome map are regions targeted by spacers, and the numbers are the coordinates of the prophage in the bacterial genome.

present in *E.* coli are not tailored to the viral invaders encountered by the bacterial population in its direct environment. This is also consistent with another study, where they recovered *E.* coli spacers from a 42 000-year-old mammoth specimen where some matched with present-day *E.* coli spacers [15].

Our work massively expands the known targets of *E*. coli spacers, revealing that most target prophages. In previous studies, the percentage of spacers matching a target was consistently low, ranging from 0.6% to 12% [11, 12, 15, 16, 31]. Here, we found a target for 60% of the 946 spacer clusters. The vast majority of these targets are found within prophage regions in *Escherichia* chromosomes. It has been previously reported that

E. coli spacers target prophages [12, 33], but the high prevalence of this occurrence reported in the present study is new. Many targets were identified in bacterial and viral metagenomes from the same cohort. Even though these targets were not samplespecific as already mentioned, these results support the need to combine culturomics and high-throughput sequencing to elucidate spacer targets and to better understand phage–bacteria interactions. We also showed that for phage genomes in the NCBI Virus database, the preference for spacer targets in prophages is not caused by an overrepresentation of temperate phages. Quite the contrary, temperate phages are only a minority in the database but represent the spacer targets for phages that infect *Escherichia*, Salmonella, and Enterobacteria. Still, there are methodological and biological reasons that could explain the high prevalence of spacer targets in prophages, and they should be further explored. First, searching for spacer targets in bacterial genomes inevitably increases the potential to find a hit, simply because the bacterial genome database is much larger (more than one million genomes, averaging a few million bp in size) than its phage counterpart (around 30000 genomes, averaging a hundred thousand bp in size). It would be interesting to examine this with another bacterial species whose CRISPR-Cas system mainly targets virulent phages and to verify whether we also find spacer targets in bacterial genomes. Instead, we argue that this predominance of spacer targets in prophages is the result of specificities in the interactions of E. coli and its coliphages. Mathieu and colleagues [25] previously looked at virulent and temperate coliphages originating from the same samples used in this study and found that temperate phages are more prevalent but less infectious than virulent phages. Perhaps the encounter rate is higher for temperate phages, making it more advantageous for E. coli to tailor its CRISPR-Cas system against temperate phages.

The high prevalence of spacer targets in prophages is not specific to E. coli, as it has been shown for other bacteria, such as P. aeruginosa [34], Flavobacterium columnare [35], Paenibacillus larvae [36], and Streptococcus pyogenes [37]. More broadly, a vast bioinformatics analysis led by Shmakov and colleagues [38] showed that, in bacterial and archaeal genomes, nearly all spacers are predicted to match sequences of the mobilome (mobile genetic elements, such as plasmids and prophages). In that same study, self-targeting spacers were nearly absent, which suggested that there is a strong selection against them. This result differs from what we observed in our E. coli isolates, as we measured a 10% prevalence of genomes with self-targeting records. The rate of self-targeting events for spacers matching known sequences has not been thoroughly investigated for a broad range of bacteria [39], making it difficult to evaluate if the rate measured in E. coli is indicative of an active system or not. However, in Streptococcus thermophilus, a model bacterium used for the study of an active Type II-A CRISPR-Cas system, it was found that 7% of the spacers matched the chromosome [40].

The E. coli is a fascinating model for the study of the biology of CRISPR-Cas systems because despite being the first organism in which a CRISPR array was identified in 1987 [41], we continue to uncover new aspects of its mechanism. Evidence suggests that the canonical function of its system is inactive: spacer acquisition is not observed in laboratory conditions (due to a well-characterized repressor) and spacers are not specific to invading DNA from the environment (at least not detected using the relatively shallow sequencing depth of many metagenome studies). However, other observations do not support this hypothesis. For example, there is a general mutational bias toward deletion in prokaryotic genomes [42], which makes it difficult to explain how E. coli CRISPR arrays that are up to 32 spacers long and from an inactive system could persist and avoid deletion. In addition, we uncovered the targets for 60% of the spacer clusters and found that spacers almost exclusively match prophages. Yet, spacers targeting prophages present in the same bacterial chromosomes are very rare. This could mean that the system effectively protects from temperate phage integration. In E. coli, spacers could effectively alter lysogenization, induction, or prophage curing in specific conditions. For example, spacers played a role in protection from a temperate phage (phage lambda) in E. coli in laboratory conditions with an overexpressed CRISPR-Cas system [43]. The possibility of CRISPR playing a role in protection is reminiscent of the role of the P. aeruginosa CRISPR–Cas system and its interaction with the temperate phage DMS3. This CRISPR–Cas system was first thought to be inactive in protecting against phage infection [34], but it was later found to be active in low nutrient and mixed culture conditions [7, 44]. A CRISPR–Cas system and DMS3 interplay that is not mediated by interference is necessary to alter biofilm formation [45, 46]. This suggests that CRISPR–Cas systems may play other roles beyond their canonical function in temperate phage–bacteria interactions.

Altogether, these results provide a new perspective on the diversity and potential activity of the CRISPR1 system in *E. coli*. With 60% of the spacers matching prophages, we support the idea that its CRISPR–Cas system is an antiprophage system as proposed recently [47]. We hypothesize that studying *E. coli* and temperate phages in diverse conditions, such as in environments that mimic natural ecosystem, may provide additional knowledge about CRISPR-mediated interactions.

Acknowledgements

We are grateful to all members of the EarlyVir consortium for insightful discussions. We express our most sincere gratitude to the children and families of the COPSAC2010 cohort for their support and commitment. We thank Amanda Toperoff and Michi Waygood for editorial assistance.

Supplementary material

Supplementary material are available at The ISME Journal online.

Conflicts of interest

None declared.

Funding

This work was supported by the Joint Programming Initiative "Healthy Diet for a Healthy Life," including the Danish Agency for Science and Higher Education and the Canadian Institutes of Health Research (Team grant on Intestinal Microbiomics, Institute of Nutrition, Metabolism, and Diabetes, grant number 143924). M.B.D. is recipient of graduate scholarships from the Fonds de Recherche du Québec-Nature et Technologies (259257) as well as Sentinel North and is a recipient of the Goran-Enhorning Graduate Student Research Award from the Canadian Allergy, Asthma, and Immunology Foundation. This research was enabled in part by support provided by Calcul Québec (www.calculquebec. ca) and Digital Research Alliance of Canada (alliancecan.ca). J.T. is supported by the BRIDGE-Translational Excellence Programme (bridge.ku.dk) at the Faculty of Health and Medical Sciences, University of Copenhagen, funded by the Novo Nordisk Foundation, grant agreement no. NNF18SA0034956. AA acknowledges financial support from the Sentinelle Nord initiative of the Canada First Research Excellence Fund and from the Natural Sciences and Engineering Research Council of Canada (project 2019-05183). S.M. holds the Tier 1 Canada Research Chair in Bacteriophages (950-232136).

Data availability

Spacers from COPSAC E. coli strains are available as a supplemental fasta file. Bacterial strains are available upon request.

References

- Barrangou R, Fremaux C, Deveau H et al. CRISPR provides acquired resistance against viruses in prokaryotes. Science 2007;315:1709–12. https://doi.org/10.1126/science.1138140
- Garneau JE, Dupuis M-È, Villion M et al. The CRISPR/Cas bacterial immune system cleaves bacteriophage and plasmid DNA. Nature 2010;468:67–71. https://doi.org/10.1038/nature09523
- Brouns SJJ, Jore MM, Lundgren M et al. Small CRISPR RNAs guide antiviral defense in prokaryotes. Science 2008;321:960–4. https:// doi.org/10.1126/science.1159689
- Makarova KS, Wolf YI, Iranzo J et al. Evolutionary classification of CRISPR–Cas systems: a burst of class 2 and derived variants. Nat Rev Microbiol 2020;18:67–83. https://doi.org/10.1038/ s41579-019-0299-x
- Weissman JL, Laljani RMR, Fagan WF et al. Visualization and prediction of CRISPR incidence in microbial trait-space to identify drivers of antiviral immune strategy. ISME J 2019;13:2589–602. https://doi.org/10.1038/s41396-019-0411-2
- Meaden S, Biswas A, Arkhipova K et al. High viral abundance and low diversity are associated with increased CRISPR-Cas prevalence across microbial ecosystems. Curr Biol 2022;32:220– 7.e5. https://doi.org/10.1016/j.cub.2021.10.038
- Alseth EO, Pursey E, Luján AM et al. Bacterial biodiversity drives the evolution of CRISPR-based phage resistance. Nature 2019;574:549–52. https://doi.org/10.1038/s41586-019-1662-9
- Dimitriu T, Kurilovich E, Łapińska U et al. Bacteriostatic antibiotics promote CRISPR-Cas adaptive immunity by enabling increased spacer acquisition. Cell Host Microbe 2022;30:31–40.e5. https://doi.org/10.1016/j.chom.2021.11.014
- Xue C, Sashital DG. Mechanisms of type I-E and I-F CRISPR-Cas systems in Enterobacteriaceae. EcoSal Plus 2019;8:10.1128. https:// doi.org/10.1128/ecosalplus.esp-0008-2018
- García-Gutiérrez E, Almendros C, Mojica FJM et al. CRISPR content correlates with the pathogenic potential of *Escherichia* coli. PLoS One 2015;**10**:e0131935. https://doi.org/10.1371/journal. pone.0131935
- Touchon M, Charpentier S, Clermont O et al. CRISPR distribution within the Escherichia coli species is not suggestive of immunity-associated diversifying selection. J Bacteriol 2011;193: 2460–7. https://doi.org/10.1128/JB.01307-10
- Díez-Villaseñor C, Almendros C, García-Martínez J et al. Diversity of CRISPR loci in Escherichia coli. Microbiology 2010;156:1351–61. https://doi.org/10.1099/mic.0.036046-0
- Pul Ü, Wurm R, Arslan Z et al. Identification and characterization of E. coli CRISPR-cas promoters and their silencing by H-NS. Mol Microbiol 2010;75:1495–512. https://doi.org/10.1111/ j.1365-2958.2010.07073.x
- 14. Westra ER, Pul Ü, Heidrich N *et al*. H-NS-mediated repression of CRISPR-based immunity in *Escherichia coli* K12 can be relieved by the transcription activator LeuO. *Mol Microbiol* 2010;**77**:1380–93. https://doi.org/10.1111/j.1365-2958.2010.07315.x
- Savitskaya E, Lopatina A, Medvedeva S et al. Dynamics of Escherichia coli type I-E CRISPR spacers over 42 000 years. Mol Ecol 2017;26:2019–26. https://doi.org/10.1111/mec.13961
- Yin S, Jensen MA, Bai J et al. The evolutionary divergence of Shiga toxin-producing Escherichia coli is reflected in clustered regularly interspaced short palindromic repeat (CRISPR) spacer composition. Appl Environ Microbiol 2013;79:5710–20. https://doi. org/10.1128/AEM.00950-13
- Bisgaard H, Vissing NH, Carson CG et al. Deep phenotyping of the unselected COPSAC2010 birth cohort study. Clin Exp Allergy 2013;43:1384–94. https://doi.org/10.1111/cea.12213

- Stokholm J, Thorsen J, Chawes BL et al. Cesarean section changes neonatal gut colonization. J Allergy Clin Immunol 2016;138:881– 889.e2. https://doi.org/10.1016/j.jaci.2016.01.028
- Biswas A, Staals RHJ, Morales SE et al. CRISPRDetect: a flexible algorithm to define CRISPR arrays. BMC Genom 2016;17:356. https://doi.org/10.1186/s12864-016-2627-0
- Dion MB, Labrie SJ, Shah SA et al. CRISPRStudio: a userfriendly software for rapid CRISPR array visualization. Viruses 2018;10:602. https://doi.org/10.3390/v10110602
- Rosvall M, Bergstrom CT. Maps of information flow reveal community structure in complex networks. Proc Natl Acad Sci USA 2008;105:1118–23. https://doi.org/10.1073/pnas.0706851105
- Dion MB, Plante P-L, Zufferey E et al. Streamlining CRISPR spacerbased bacterial host predictions to decipher the viral dark matter. Nucleic Acids Res 2021;49:3127–38. https://doi.org/10.1093/ nar/gkab133
- Camacho C, Coulouris G, Avagyan V et al. BLAST +: architecture and applications. BMC Bioinform 2009;10:421. https://doi. org/10.1186/1471-2105-10-421
- Shah SA, Deng L, Thorsen J et al. Expanding known viral diversity in the healthy infant gut. Nat Microbiol 2023;8:986–98. https://doi. org/10.1038/s41564-023-01345-7
- Mathieu A, Dion M, Deng L et al. Virulent coliphages in 1year-old children fecal samples are fewer, but more infectious than temperate coliphages. Nat Commun 2020;11:378. https:// doi.org/10.1038/s41467-019-14042-z
- Li X, Stokholm J, Brejnrod A et al. The infant gut resistome associates with E. Coli, environmental exposures, gut microbiome maturity, and asthma-associated bacterial composition. Cell Host Microbe 2021;29:975–87.e4. https://doi.org/10.1016/j. chom.2021.03.017
- Van Dongen S. Graph clustering via a discrete uncoupling process. SIAM J Matrix Anal Appl 2008;30:121–41. https://doi. org/10.1137/040608635
- Kieft K, Zhou Z, Anantharaman K. VIBRANT: automated recovery, annotation and curation of microbial viruses, and evaluation of viral community function from genomic sequences. Microbiome 2020;8:90. https://doi.org/10.1186/ s40168-020-00867-0
- Sullivan MJ, Petty NK, Beatson SA. Easyfig: a genome comparison visualizer. Bioinformatics 2011;27:1009–10. https://doi. org/10.1093/bioinformatics/btr039
- Arndt D, Grant JR, Marcu A et al. PHASTER: a better, faster version of the PHAST phage search tool. Nucleic Acids Res 2016;44:W16– 21. https://doi.org/10.1093/nar/gkw387
- Sheludchenko MS, Huygens F, Stratton H et al. CRISPR diversity in E. Coli isolates from Australian animals, humans, and environmental waters. PLoS One 2015;10:e0124090. https://doi. org/10.1371/journal.pone.0124090
- Fu Q, Su Z, Cheng Y et al. Clustered, regularly interspaced short palindromic repeat (CRISPR) diversity and virulence factor distribution in avian *Escherichia coli*. Res Microbiol 2017;**168**:147–56. https://doi.org/10.1016/j.resmic.2016.10.002
- Alonso CA, de Toro M, de la Cruz F et al. Genomic insights into drug resistance and virulence platforms, CRISPR-Cas systems and phylogeny of commensal E. coli from wildlife. Microorganisms 2021;9:999. https://doi.org/10.3390/microorganisms9050999
- Cady KC, White AS, Hammond JH et al. Prevalence, conservation and functional analysis of Yersinia and Escherichia CRISPR regions in clinical Pseudomonas aeruginosa isolates. Microbiology 2011;157: 430–7. https://doi.org/10.1099/mic.0.045732-0
- 35. Laanto E, Ravantti JJ, Sundberg LR. Prophages and past prophage-host interactions revealed by CRISPR spacer content

in a fish pathogen. Microorganisms 2020;8:1919. https://doi. org/10.3390/microorganisms8121919

- Stamereilers C, Wong S, Tsourkas PK. Characterization of CRISPR spacer and protospacer sequences in *Paenibacillus lar*vae and its bacteriophages. Viruses 2021;**13**:459. https://doi. org/10.3390/v13030459
- Le Rhun A, Escalera-Maurer A, Bratovič M et al. CRISPR-Cas in Streptococcus pyogenes. RNA Biol 2019;16:380–9. https://doi. org/10.1080/15476286.2019.1582974
- Shmakov SA, Sitnik V, Makarova KS et al. The CRISPR spacer space is dominated by sequences from species-specific mobilomes. mBio 2017;8:e01397–17. https://doi.org/10.1128/ mBio.01397-17
- Wimmer F, Beisel CL. CRISPR-Cas systems and the paradox of self-targeting spacers. Front Microbiol 2020;10:3078. https://doi. org/10.3389/fmicb.2019.03078
- Horvath P, Romero DA, Coûté-Monvoisin A-C et al. Diversity, activity, and evolution of CRISPR loci in Streptococcus thermophilus. J Bacteriol 2008;190:1401–12. https://doi.org/10.1128/JB. 01415-07
- 41. Ishino Y, Shinagawa H, Makino K et al. Nucleotide sequence of the *iap* gene, responsible for alkaline phosphatase isoenzyme conversion in *Escherichia* coli, and identification of the gene

product. J Bacteriol 1987;**169**:5429–33. https://doi.org/10.1128/jb.169.12.5429-5433.1987

- Kuo C-H, Ochman H. Deletional bias across the three domains of life. *Genome Biol Evol* 2009;1:145–52. https://doi.org/10.1093/gbe/ evp016
- Edgar R, Qimron U. The Escherichia coli CRISPR system protects from λ lysogenization, lysogens, and prophage induction. J Bacteriol 2010;192:6291–4. https://doi.org/10.1128/JB.00644-10
- Westra ER, Van Houte S, Oyesiku-Blakemore S et al. Parasite exposure drives selective evolution of constitutive versus inducible defense. Curr Biol 2015;25:1043–9. https://doi. org/10.1016/j.cub.2015.01.065
- Zegans ME, Wagner JC, Cady KC et al. Interaction between bacteriophage DMS3 and host CRISPR region inhibits group behaviors of Pseudomonas aeruginosa. J Bacteriol 2009;191:210–9. https://doi. org/10.1128/JB.00797-08
- Cady KC, O'Toole GA. Non-identity-mediated CRISPRbacteriophage interaction mediated via the Csy and Cas3 proteins. J Bacteriol 2011;193:3433–45. https://doi.org/10.1128/ JB.01411-10
- Song S, Semenova E, Severinov K et al. CRISPR-Cas controls cryptic prophages. Int J Mol Sci 2022;23:16195. https://doi.org/10.3390/ ijms232416195