Community-promoted antibiotic resistance genes show increased dissemination among pathogens

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Abstract

Antibiotic resistance is increasing among bacterial pathogens, posing one of the most severe threats to future public health. A major contributor to the increasing resistance is the dissemination of mobile antibiotic resistance genes (ARGs) among bacterial communities. These genes are ubiquitously present in various environments and are especially diverse in the human gut and in wastewater. Despite this, the clinical implications of the prevalence of ARGs in these bacterial communities remain unclear. In this study, we aimed to investigate how the abundance of ARGs in human gut and wastewater microbiomes reflects their dissemination among important bacterial pathogens. To do this, we estimated the abundance of >30,000 ARGs, including both well-known (established) and computationally predicted (latent) genes, in >6,000 metagenomic samples. From their prevalence in human gut and wastewater, we identified four major categories of ARGs: co-promoted, human gut (HG)-promoted, wastewater (WW)-promoted, and non-promoted. Our results showed that co-promoted ARGs were, by far, most promiscuous, with an increased frequency of being carried by multiple bacterial phyla, and higher co-localization with broad host range conjugative elements. Co-

promoted ARGs were also found to be overrepresented among genes identified in multiple pathogenic species and exhibited an overall higher genetic compatibility, both with pathogens and other typical residents of the human gut and wastewater microbiomes. Taken together, our results highlight the link between the promotion of ARGs in the human gut and wastewater microbiomes and their presence in human pathogens, and, thereby, their potential risk to human health.

Introduction

The increasing antibiotic resistance among bacterial pathogens threatens human health, causing over one million yearly deaths worldwide [1]. Bacteria develop resistance as a result of changes in their genetic makeup, often through the acquisition of mobile antibiotic resistance genes (ARGs) via horizontal gene transfer [2]. Many ARGs are located on mobile genetic elements (MGEs), such as transposons and plasmids, which can make them transferable between distantly related bacteria. This facilitates efficient dissemination within and between microbial communities [3, 4]. To date, thousands of ARGs that collectively confer resistance to almost every class of clinically useful antibiotics have been described and recorded in databases such as CARD and ResFinder [5, 6]. The number of known ARGs is constantly increasing as new, often more efficient, variants are discovered, typically after they have been acquired by a pathogenic host and caused hard-to-treat infections.

Bacterial communities are known to maintain a diverse array of ARGs, collectively referred to as the resistome. Investigations of the resistome utilizing shotgun metagenomics – a method wherein DNA fragments are randomly sequenced from the genomes of all microorganisms present in a sample – have demonstrated that different ARGs have complex abundance patterns that vary between environmental contexts [7-9]. Both the human gastrointestinal tract and sewage treatment plants have been demonstrated to have particularly comprehensive resistomes, encompassing ARGs that are also frequently carried by pathogenic bacteria [10, 11]. These environments create opportunities for exchange of genetic information between bacteria, potentially under selection pressure from antibiotics and other chemicals, and have therefore been suggested as potential hotspots for the promotion of ARGs.

Even though the resistome has been studied in detail for many environments, we still know little about how the composition of ARGs reflects the risks to human health. In fact, resistance genes in bacterial communities are often maintained by commensal and non-pathogenic bacteria. However, for an ARG to affect the efficacy of antibiotic treatments, it needs to be present and induce a resistance phenotype in a bacterium able to cause infections with adverse outcomes, especially in humans. The association of ARG abundance patterns to infection control is vital for wastewater surveillance, a step that has so far been challenging [12, 13]. Recent efforts have been made to assess the risk of ARGs present in bacterial communities based on various scoring systems. Nonetheless, these methodologies are inherently constrained due to the absence of empirical data

associating gene promotion with implications for human health. This limitation hiders risk assessment, complicating interpretation and decision-making processes [14].

In this study, we aim to explore and clarify the connection between the prevalence of ARGs in human gut and wastewater bacterial communities and their potential risks to human health. To this end, we estimated the prevalence of >30,000 well-characterized (established) and computationally predicted (latent) ARGs in >6,000 shotgun metagenomes and >1.6 million bacterial genomes. Our results show that genes promoted in both human gut and wastewater microbiomes have a substantially higher potential to spread among human pathogens. We, furthermore, demonstrate that this potential is facilitated by the frequent presence of broad host range conjugative elements in their genetic context and an overall genetic compatibility with a large variety of bacterial hosts. Our results, thus, provide new insights into the link between the promotion of ARGs in bacterial communities and their dissemination among pathogens.

Results

Identification of promoted resistance genes in microbial communities

We detected the presence of 30,654 ARGs in a total of 5,630 human gut and 1,034 wastewater metagenomic samples. The ARGs included both established (i.e., well-known ARGs defined by matches to the ResFinder database) and latent variants (i.e., previously uncharacterized ARGs predicted computationally) [7] conferring resistance to aminoglycoside, beta-lactam, macrolide, fluoroquinolone, and tetracycline antibiotics. The prevalence of each ARG was estimated as the proportion of samples from the two different environments where each ARG was present (≥3 mapped reads in rarefied data, see Methods). This resulted in a total of 1,592 detected ARGs, which we classified into four categories (Fig. 1): 44 *co-promoted* ARGs that were present in ≥5% of human gut samples and ≥5% of wastewater samples, 112 *human gut (HG)-promoted* ARGs that were present in ≥5% of human gut samples but <5% of wastewater samples, 48 *wastewater (WW)-promoted* ARGs that were present in ≥5% of wastewater samples but <5% of human gut samples, and, finally, 1,388 *non-promoted* ARGs that were present in <5% of human gut samples and <5% of wastewater samples (but not completely absent from all metagenomes). Notably, all HG-promoted ARGs were also detected (≥1 read) in wastewater samples, albeit at lower abundances, but only a few WW-promoted ARGs (25%) were detected in any human gut samples, though these displayed, in some cases, high abundances (Supplementary Fig. 1).

Fig. 1. Prevalence of antibiotic resistance genes (ARGs) in human gut and wastewater metagenomic samples. Labels are included for established ARGs that were present in ≥25% of samples from either environment. The dashed lines show the classification of the ARGs into four groups: co-promoted (top right), human gut-promoted (top left), wastewater-promoted (bottom right), and non-promoted (bottom left).

Co-promoted and HG-promoted established ARGs together encompassed the largest diversity of resistance mechanisms (Fig. 2a). In particular, tetracycline resistance was significantly overrepresented among both co-promoted and HG-promoted established ARGs (RPG with $p=3.31\times10^{-9}$, inactivating enzymes with $p=1.87\times10^{-5}$, respectively, Fisher's exact test, Supplementary Fig. 2) while beta-lactamases were underrepresented ($p=2.31\times10^{-3}$, $p=1.00\times10^{-4}$, for co-promoted and HG-promoted established ARGs, respectively). Co-promoted ARGs also included aminoglycoside and macrolide resistance genes (*erm* and *mph* inactivating enzymes) but not any *qnr* genes, which were only found among the HG-promoted (established) and non-promoted ARGs (both latent and established).

Fig. 2. Overview of the four identified categories of antibiotic resistance genes (ARGs): Co-promoted, human gut (HG)-promoted, wastewater (WW)-promoted, and non-promoted. a Distribution of gene classes among the different categories of established and latent ARGs. b Proportion of established and latent ARGs comprising the different categories. Absolute numbers are displayed above each bar. c-f Distribution of phyla among the hosts carrying co-promoted, HG-promoted, WW-promoted, and non-promoted ARGs, respectively, divided between established and latent ARGs. The length of each bar represents the proportion of ARGs from that category that were identified in at least one genome from that phylum. Absolute numbers are displayed together with each bar.

The co-promoted ARGs almost exclusively included established genes known to be mobile and widespread (Fig. 1, Fig. 2b). This included, for example, *tet(Q)* and *erm(B)*, as well as *aph(3'')-Ib* (*strA*) and *aph(6)-Id* (*strB*), which are frequently found on mobile genetic elements [15]. By contrast, the HG-promoted and WW-promoted groups consisted of a more equal mix of established

and latent ARGs (Fig. 2b). Established HG-promoted ARGs included, e.g. *erm(F)* and *cfxA* beta-lactamases, while established WW-promoted ARGs included e.g., *tet(C)* and OXA beta-lactamases (Fig. 1). Non-promoted ARGs were highly dominated by latent resistance genes (Fig. 2b), but also included established ARGs from 21 of the 22 analyzed gene classes.

Co-promoted resistance genes have a higher potential for horizontal transfer

Resistance genes from all four categories, both latent and established, were most frequently detected in the phyla Pseudomonadota, Bacillota, and Bacteroidota (Fig. 2c–f). However, the hosts for co-promoted and HG-promoted ARGs showed a higher diversity with a more even distribution between major bacterial phyla. In contrast, hosts carrying WW-promoted and non-promoted ARGs primarily belonged to Pseudomonadota. Also, several of the co-promoted established ARGs were present in multiple host genomes from different phyla, confirming inter-phyla gene transfers.

To further assess the mobility potential for ARGs promoted in the different environments, we retrieved the taxonomic affiliations of the bacterial genomes in which they were identified and identified their lowest common ancestor (Fig. 3). Interestingly, co-promoted genes were significantly overrepresented among established ARGs identified in multiple phyla ($p=0.002$). Indeed, as many as 49% of the co-promoted established ARGs were identified in more than one phylum, compared to 17%, 20%, and 11% of HG-promoted, WW-promoted, and non-promoted established ARGs, respectively. In contrast, HG-promoted and non-promoted established ARGs were mainly restricted to the genus level, suggesting either that these genes are relatively immobile or that they have a narrow host range (Fig. 3a). However, WW-promoted ARGs, which were largely confined to the Pseudomonadota phylum, were significantly overrepresented among all ARGs found in multiple classes within the same phylum ($p=0.004$, Fig. 3a, b). Thus, while established co-promoted and HG-promoted ARGs were observed in taxonomically similar bacterial hosts, established co-promoted and WW-promoted ARGs showed similarities in their ability to spread over large phylogenetic distances. Among latent genes, however, inter-phyla transfers were much rarer. Indeed, only the HG-promoted group contained a notable proportion of latent ARGs observed in multiple phyla (14%), while inter-phyla transferred latent ARGs were (almost) completely absent from the other categories (0%, 0%, and 0.2% for co-promoted, WW-promoted, and non-promoted latent ARGs, respectively).

Fig. 3. Overview of the horizontal transfer of co-promoted, human gut (HG)-promoted, wastewater (WW)-promoted, and non-promoted antibiotic resistance genes (ARGs). a–b Taxonomic host range associated with established and latent ARGs with different promotion categories. Each bar represents a taxonomic level, displaying the mean proportions of ARGs that showed evidence of having transferred across that taxonomic distance belonging to each promotion category, after 1,000 repetitions of subsampling the groups to equal size (for details, see Methods). In panel b, co-promoted latent ARGs are excluded due to their low n. c–f Networks describing the inter-phyla transfer patterns of ARGs with different promotion categories. Each node represents a bacterial phylum, with a size that is proportional to how many inter-phyla transferred

ARGs were found in that phylum. Edges are drawn between phyla where ≥10% of ARGs were simultaneously identified, the thickness represents the proportion of inter-phyla transferred ARGs that were identified in both phyla.

Network analysis was used to explore the transfer patterns for the established and latent ARGs present in multiple phyla (Fig. 3c–f). Co-promoted ARGs were found to have consistently transferred between all the major bacterial phyla, including Actinomycetota, Bacillota, Bacteroidota, Campylobacterota, and Pseudomonadota. By contrast, HG-promoted, WW-promoted, and non-promoted ARGs showed overall lower connectivity, primarily indicating transfers between Pseudomonadota and either Bacillota, Acinomycetota, or Bacteriodota (though this was rare for non-promoted ARGs). Indeed, while inter-phyla transferred co-promoted ARGs were present in an average of 4.45 phyla, other ARG categories were much more restricted (average 2.39, 3.25, and 2.39 phyla for inter-phyla transferred HG-promoted, WW-promoted, and non-promoted ARGs, respectively; Supplementary Fig. 3).

To further explore the potential for horizontal gene transfer, we analyzed the genetic contexts of ARGs in different promotion categories. The up- and downstream sequences around the co-promoted, HG-promoted, and WW-promoted ARGs were extracted from their respective host genomes and annotated for genes associated with conjugative elements, including mating pair formation (MPF) genes and relaxases, as well as co-localized mobile ARGs (Fig. 4, Supplementary Fig. 4). There was a clear difference in the diversity of MPF genes, especially between co-promoted and WW-promoted established ARGs (p=0.001, Fisher's exact test, Fig. 4). This included, particularly, a higher frequency of class FATA MPF genes located close to co-promoted ARGs, while class T MPFs were more commonly located close to WW-promoted ARGs (Fig. 4). Among co-localized relaxases, MOB_P dominated for all promoted ARGs, however, co-promoted established ARGs were also commonly associated with MOB_V (Supplementary Fig. 4c). Co-localization of at least one other ARG was equally frequent between the established co-promoted (82.9%), WW-promoted ARGs (75%), and HG-promoted (73.8%), while it was less common for established non-promoted ARGs (28.8%, Supplementary Fig. 4e).

Fig. 4. Distribution of mating pair formation (MPF) genes co-localized with established co-promoted, HG-promoted, WW-promoted, and non-promoted ARGs. For each ARG category, bar lengths indicate the proportion of ARGs with a co-localized MPF gene type detected in at least one host genome.

Co-promoted resistance genes are more widespread in pathogens

Next, we assessed the dissemination of ARGs in different promotion categories among 15 clinically important pathogens known to be associated with high mortality [16]. We observed a strong overrepresentation of co-promoted genes among established ARGs identified in >5 pathogens (Fig. 5a, Supplementary Fig. 5). In fact, the vast majority (>90%) of the established ARGs that had spread to at least eight different pathogens were co-promoted. Accordingly, the co-promoted ARGs was the only category commonly present in pathogens from multiple phyla, including Pseudomonadota (e.g. *Escherichia coli, Klebsiella pneumoniae,* and *Salmonella enterica*), Bacteriodota (e.g. *Enterococcus spp.*, *Staphylocoocus aureus*, and *Streptococcus spp.*), and Campylobacterota (*Campylobacter jejuni*) (Fig. 5c–f). In contrast, HG and WW-promoted established ARGs were common in pathogens from Pseudomonadota, but rare in other phyla. WW-promoted ARGs were, however, somewhat more common among established ARGs identified in ≥5 pathogens (Fig. 5a), which reflects their high frequency in a limited number of specific pathogens (Fig. 5e).

Fig. 5. Distribution between pathogenic host species of antibiotic resistance genes (ARGs) among co-promoted HG-promoted, WW-promoted, and non-promoted ARGs. a–b Mean proportions of established and latent ARGs carried by ≥n pathogens belonging to each promotion category after 1,000 repetitions of subsampling the groups to equal size (for details, see Methods). In panel b, co-promoted latent ARGs are excluded due to their low n. c–f Proportion of ARGs from each category that were carried by each of the selected pathogens, divided between established and latent genes.

The lack of genetic compatibility has been suggested as a major barrier to the spread of ARGs between bacterial hosts [17, 18]. For each ARG and pathogenic species, we therefore calculated the difference in nucleotide composition (Euclidean distance in 5-mer distributions, see Methods) between the gene sequence and a representative genome. For each bacterial pathogen, we then extracted its median genetic dissimilarity to each category of ARGs. Interestingly, co-promoted established ARGs had a nucleotide composition that was on average more similar to the genomes of the included pathogens, suggesting an overall higher degree of genetic compatibility (Fig. 6, Supplementary Fig. 6). The genetic dissimilarity was significantly higher for all other established

gene categories: WW-promoted (p=2.14×10⁻⁴, compared to co-promoted, one-sided Wilcoxon's signed rank test), HG-promoted (p=3.05×10⁻⁵), and non-promoted (p=1.56×10⁻⁴). To investigate if the differences in genetic compatibility could be observed beyond the included pathogens, we identified 110 bacterial genera that were present in ≥5% of the human gut and/or wastewater metagenomic samples and estimated their nucleotide composition dissimilarity with the differently promoted established ARGs (Supplementary Fig. 7). Our results suggested that co-promoted ARGs are more genetically similar to the residents of human gut and wastewater microbiomes. Indeed, when considering the identified resident genera, co-promoted genes showed a higher average genetic compatibility with 58% of them, in comparison with 0%, 22%, and 20% for HG-promoted, WW-promoted, and non-promoted ARGs, respectively. Finally, we observed a low but significant negative correlation between the average nucleotide composition dissimilarity of an established ARG to typical residents of the human gut and wastewater and its prevalence in the metagenomes (Spearman's ϱ=-0.28, p=1.81×10⁻⁷, Supplementary Fig. 8).

Fig. 6. Genetic compatibility, estimated as the Euclidean distance between 5mer-distributions, between antibiotic resistance genes (ARGs) in different promotion categories and important bacterial pathogens. a Boxplots showing the median genetic incompatibilities between each of the included pathogens and the ARGs from different categories. *P*-values have been generated using Wilcoxon's signed rank test (one-sided), comparing co-promoted antibiotic resistance genes (ARGs) to each of the other categories. *p<0.01. b Relationship between prevalence in pathogenic hosts and genetic incompatibility with pathogenic host species. Each point represents one of the selected pathogens, as well as a category of established ARGs (co-promoted, human gut (HG)-promoted, wastewater (WW)-promoted, or non-promoted), with the *y*-axis representing the proportion of ARGs from that category that were carried by that specific species, and the *x*-axis representing the median genetic incompatibility between the included ARGs and the representative genome of that pathogen.

Discussion

In this study, we investigated ARGs highly prevalent in the human gut and wastewater microbiomes and analyzed their potential for horizontal dissemination. Based on over 30,000 resistance genes identified in over 6,000 metagenomes and almost 1.6 million bacterial genomes, we show that co-promoted ARGs, i.e., resistance genes that were re-occurring in both human gut and wastewater bacterial communities, are especially widespread among taxonomically diverse bacteria, including pathogens. In particular, co-promoted ARGs were found to co-localize with a broader diversity of conjugative elements and be more genetically compatible with their potential host bacteria, including common pathogens and typical residents of the human gut and wastewater microbiomes. This study clarifies the link between ARGs promoted in human and wastewater microbiomes and their distribution among pathogens and, thereby, their connection to human health risks.

Our results show that co-promoted ARGs, which almost exclusively encompassed well-known (established) genes, generally have a wider host range compared to other resistance genes detected in the human gut and wastewater metagenomes (Fig. 3). In fact, as many as 49% of the co-promoted established ARGs were identified in multiple bacterial phyla, suggesting at least one inter-phyla transfer event during their evolutionary history. In contrast, only 17% of HG-promoted, 20% of WW-promoted, and 11% of non-promoted established ARGs could be identified outside a single phylum. Our results, thus, suggest that co-promoted ARGs have an increased potential to spread horizontally between evolutionarily divergent bacteria. Interestingly, this potential was not clearly reflected by association with plasmids and other conjugative elements. While we found that co-promoted established ARGs were co-localized with several types of conjugative systems [19], this was also the case for resistance genes promoted solely in the human gut or wastewater bacterial communities (Fig. 4). However, co-promoted ARGs were, to a greater extent, associated with conjugative elements carrying mating pair formation (MPF) genes from class FATA, which are almost exclusively found in bacterial hosts from Bacillota and Actinomycetota [20]. The presence of co-promoted genes in these particular phyla was further supported by taxonomic analysis of their hosts, both in general (Fig. 2c) and for pathogens (Fig. 5). In comparison, co-localization of MPF_{FATA} was rare for HG-promoted and WW-promoted established ARGs, despite the strong presence of HG-promoted ARGs in Bacillota (Fig. 2d). HG-promoted and WW-promoted established ARGs were instead more commonly associated with MPF genes from class T and F. Since MPF_T and MPF_F, are most often identified in Pseudomondota [21], like the WW-promoted ARGs themselves, this could explain why these ARGs were less likely to transfer between phyla, while frequently being found among different classes from the same phylum (Pseudomonadota). Our results, thus, suggest that the higher potential to spread over large evolutionary distances observed for co-promoted ARGs is facilitated by their frequent location on specific conjugative elements that, together, provide an increased host range covering all major phyla.

Horizontally transferred genes need to be genetically compatible with their new host to confer a fitness advantage. More specifically, any acquired gene needs transcription and translation efficiency to ensure a high expression level [22]. The protein must also be successfully integrated into the cellular biochemical context to induce a sufficiently strong phenotype [23]. The lack of genetic compatibility between an acquired ARG and its new host has been suggested as a significant barrier that can prevent efficient spread between genetically diverse bacteria [24]. Interestingly, our results indicate that co-promoted establihsed ARGs generally have higher genetic compatibility with important bacterial pathogens than other resistance genes detected in the human gut and wastewater microbiome. This higher compatibility, which was estimated as the difference in nucleotide composition, varied in size but was apparent for all investigated pathogens regardless of phylum, with the exceptions of the GC-rich *Mycobacterium tuberculosis* and *Pseudomonas aeruginosa* (Fig. 5g, Supplementary Fig. 4). A similar trend could be seen for HG- and WW-promoted ARGs, which had overall higher compatibility with pathogens than non-promoted resistance genes. The overall higher genetic compatibility of co-promoted establihsed ARGs was also observed for common residents of the human gut and wastewater microbiomes, although there was significant variability between genera (Supplementary Fig. 7). Our results, thus, suggest that the higher genetic compatibility of co-promoted ARG may contribute to their ability to efficiently transfer, especially between evolutionarily divergent bacterial hosts [17]. It should, however, be pointed out that we found established ARGs in the analyzed environments that were rare in pathogens, despite having high genetic compatibility. This suggests that compatibility alone – at least not in the way it is captured in this study – is insufficient to allow widespread dissemination among clinically important bacterial pathogens.

The strong association between the promotion of ARGs in microbial communities and their increased potential to spread horizontally suggests that these processes are entwined. Indeed, the human and wastewater microbiomes both contain complex mixtures of bacteria, which allows for horizontal transfer of resistance genes between divergent hosts, sometimes under strong selection pressures [4, 25-28]. Furthermore, there is mounting evidence that these and other environmental bacterial communities act as reservoirs from which resistance genes can be recruited. This is exemplified by the many clinically relevant ARGs that have been transferred to pathogens from diverse non-pathogenic bacteria, some of which are abundant in the human gut and wastewater treatment plants [29, 30]. Our results, therefore, emphasize the need for a holistic perspective to fully understand the processes behind the mobilization, acquisition, promotion, and dissemination of antibiotic resistance genes. Interestingly, we noted that the co-promoted ARGs that were most widely spread among pathogenic species encompassed genes exclusively known to have circulated among pathogens for decades. These included, among others, the beta-lactamase TEM-1, the tetracycline ribosomal protection gene *tet(M)*, and the macrolide resistance gene *erm(B)* [31-33]. These genes all have a long history of being maintained by bacteria under selection pressure, including natural residents of the human gut and wastewater communities and/or pathogens under antibiotic treatment [34-36]. Our observations, thus, suggest that the promotion and dissemination of ARGs across environments is a slow process; indeed, these genes were all found in pathogens more than 40 years ago. It is, therefore, likely that more recent resistance genes are currently

undergoing this evolutionary process, which may increase their future clinical relevance. We, thus, argue that actions aiming to reduce the dissemination of ARGs between microbial communities should be considered when designing management strategies to meet the growing threat of antibiotic resistance.

Bacteria harbor a vast diversity of resistance genes, of which only a small part is represented in current sequence repositories [7, 37]. Our results indicate that there is a wide range of uncharacterized resistance genes – here referred to as latent ARGs [7] – that can commonly be detected in both the human gut and wastewater bacterial communities. Furthermore, many of the detectable latent ARGs were found to be promoted in human and/or wastewater bacterial communities. Interestingly, when looking at the genetic compatibility between latent ARGs and pathogens, we again found that promoted genes were generally more compatible with their hosts (Supplementary Fig. 6), suggesting that they are likely under similar selection pressures as established resistance genes. Several of the promoted latent ARGs were also present close to mobile genetic elements and were found to have spread between taxonomically different bacteria. This included pathogens such as *E. coli*, suggesting that some latent ARGs may constitute emerging risks for human health. Here, it is important to emphasize that even though our analysis was based on 33,944 ARGs identified in 1,631,873 bacterial genomes, which, even if it expands current databases almost tenfold [5], is still limited and only reflects a small part of the resistome. Similarly, the analysis of human gut and wastewater microbiomes was based on 6,000 metagenomic samples, thus not reflecting the total diversity of these microbial communities. Thus, this suggests that there are potentially other, not yet discovered, resistance genes promoted in human and wastewater microbiomes.

Conclusions

This study clarifies the link between the promotion of ARGs in bacterial communities and their dissemination among pathogens. Specifically, ARGs that frequently recur in human gut and wastewater environments were shown to be especially mobile, with approximately half of these genes having undergone inter-phyla transfers at least once. Moreover, these genes were significantly overrepresented among resistance genes carried by multiple pathogens and exhibited a higher degree of genetic compatibility with both these pathogens and other common residents of the analyzed bacterial communities. Consequently, this study provides empirical evidence concerning the correlation between the prevalence of ARGs in human gut and wastewater microbial communities and the potential implications for human health. The findings will therefore enhance the interpretation of wastewater surveillance, bridge existing knowledge gaps in risk assessments, and enable more informed and effective policymaking to address the escalating threat of antibiotic-resistant bacteria.

Methods

Retrieval of metagenomic data

Human gut metagenomic samples were downloaded from the MGnify database (2024-07-23) [38]. Here, only data generated using Illumina platforms was included to ensure comparability. Samples were excluded if the study had less than five sequencing runs in total or if unique links to fastq files in the European Nucleotide Archive API (ENA) [39] were missing. Finally, samples representing the gut microbiome of infant humans were removed due to the differences between infant and adult gut microbiome composition. Wastewater samples were retrieved from the National Center for Biotechnology Information Sequence Read Archive (NCBI SRA) [40] based on accession IDs provided by Munk et al. 2022 [41] and Hendriksen et al. 2019 [42]. This resulted in a total of 5,694 human gut samples as well as 1,108 wastewater samples. Next, BBDuk from BBMap v39.06 [43] was used to quality control the samples, with trim quality 20, minimum length 60, and left and right trimming of the raw files. Finally, all samples containing less than 5,000,000 reads after the quality control were discarded to ensure sufficient sequencing depth, after which 5,630 human gut samples and 1,034 wastewater samples remained.

Creation of the ARG reference database

A total of 2,015,254 bacterial genomes were downloaded from NCBI Assembly (2024-03-24) [44]. To ensure the robustness of the downstream analysis, 383,381 genomes that did not pass NCBI's taxonomy check and/or where contamination was suspected, were removed from consideration. The remaining 1,631,873 genomes were screened for ARGs using fARGene v0.1 [45], a software that uses probabilistic models to identify ARGs in bacterial genomes. In addition to known resistance determinants, fARGene also predicts uncharacterized homologs with high accuracy, as previous studies have shown [46-49]. fARGene was executed using 22 profile hidden Markov models built to identify genes conferring resistance to six major classes of antibiotics: aminoglycosides (AAC aminoglycoside acetyltransferases, APH aminoglycoside phosphotransferases), beta-lactams (class A, B1/B2, B3, C, and D beta-lactamases), macrolides (Erm 23S rRNA methyltransferases, Mph macrolide 2'-phosphotransferaes), tetracyclines (tetracycline efflux pumps, inactivating enzymes, and ribosomal protection genes (RPGs)), and fluoroquinolones (*qnr*). For each model, previously reported threshold scores were used [46-50], and each hit above the specified threshold was recorded as a putative ARG and stored for further analysis. This resulted in 6,834,484 hits representing 195,722 unique gene sequences.

The ARGs predicted by fARGene were complemented with 4,150 gene sequences downloaded from the CARD database (2024-06-17) [5] that represented the same resistance gene classes as the fARGene sequences. To remove ARGs whose open reading frame overlapped with insertion sequences (ISs), we used BLASTx v2.15.0 [51] to align the resistance genes against 8,417 ISs from the ISFinder database (2024-06-18) [52, 53]. Genes with ≥80% sequence identity and ≥ amino acid overlap to any IS were discarded, resulting in the exclusion of 18 sequences. Next, all gene

sequences were clustered at 90% nucleotide similarity using VSEARCH v2.28.1 [54] to remove redundancy. We then used BLASTp v2.15.0 [51] to align the proteins encoded by the resulting centroid sequences against sequences from the ResFinder database of acquired ARGs (downloaded 2024-06-19) [6]. Genes displaying ≥90% sequence identity and ≥70% coverage to any ResFinder sequence were classified as "Established", otherwise they were classified as "Latent". In total, this resulted in a reference database of 720 established ARGs and 33,224 latent ARGs.

Quantification of ARGs in metagenomes

To quantify ARGs in different environments, metagenomic forward reads were aligned to the reference ARG database using DIAMOND blastx v2.1.19 [55]. To avoid double counting, only forward reads were considered. Reads displaying ≥95% sequence identity, ≥20% read coverage, and a minimum match length of 20 amino acids were considered matches. For reads matching multiple ARGs, the best match was selected based on 1) sequence identity, 2) alignment length, and 3) e-value. If all these were equal, one of the matching ARGs was randomly selected. To ensure comparability, each sample was rarefied to 5,000,000 reads. To visualize the relative abundance of specific ARGs in different environments, the rarefied counts were further normalized to log counts per million bacteria (logCPM$_\mathrm{B}$). Here, the normalized abundance x_{ij} of gene i in metagenome j was calculated as $x_{ij} = \log\left(\frac{r_{ij}+1}{B_j} \times 10^6\right)$, where r_{ij} is the number of reads matching gene i, while B_j is the total number of reads mapping to bacteria in sample j. The number of reads mapping to bacteria was calculated by applying Kraken2 to each of the quality-controlled metagenomic samples, with parameters –paired --confidence 0.1 --use-names [56].

Analysis of ARGs promoted in different environments

Next, we estimated the presence of each ARG in the human gut and wastewater by calculating the proportion of samples from each environment where the ARG was present, here defined as ≥3 matching reads. Based on these results, the ARGs were divided into four categories. Here, ARGs that were present in ≥5% of both human and wastewater samples were classified as co-promoted, while genes that were present in ≥5% of samples from only one type of environment were classified as human gut (HG)-promoted or wastewater (WW)-promoted, respectively. Finally, ARGs that were not recorded in any metagenomic sample were removed from consideration and the remaining genes that were present in <5% of both human and wastewater samples were classified as non-promoted. For each ARG, the taxonomy of the bacteria identified as hosts was recorded at each taxonomic level. To further increase the reliability of the results, we required more ≥ 3 observations of any taxa to record it for ARGs that were identified in ≥3 genomes (otherwise, the required number of observations was the same as the identified number of host genomes). Based on the observed host taxonomy, we then identified for each ARG at the taxonomic level at which the most recent common ancestor for all identified host species could be found, and used this information to estimate the taxonomic host range associated with established and latent ARGs from the different categories. Briefly, for each taxonomic level we first subsampled each category

of ARGs down to the size of the smallest category, after which we calculated the proportion of subsampled ARGs where the most recent common ancestor was found at that level. This procedure was repeated 1,000 times and the mean over the observed proportions was calculated. Here, categories with <10 genes were excluded from the analysis as to not remove too much data during the subsampling step. To assess the perceived overrepresentations of specific ARGs among genes with a given host-range, we performed a permutation test by randomly shuffling the labels assigning each ARG to a category, and repeated the procedure described above. This was then repeated 1,000 times, after which p-values were calculated as the proportion of generated values that were at least as extreme as the true observation.

To assess the presence of ARGs in pathogenic bacteria, the identified host species were cross-referenced against a list of 15 species associated with high mortality rates in clinical infections compiled based on Ikuta et al. 2022 [16]. We then analyzed whether established and latent genes from each category were overrepresented among ARGs carried by $\geq n$ different pathogenic host species, where n was any number between zero and 15 (the number of included pathogens). Here, for each n the different categories were again subsampled down to the size of the smallest category and the proportion of subsampled genes from each category that were carried by $\geq n$ different pathogens was calculated. As with the previous analysis, we excluded categories with <10 genes, and extracted the mean proportions over 1,000 iterations. Here, we again performed additional assessment of the results by performing 1,000 additional rounds of the analysis, each time randomly permuting the labels in the dataset. These results were then visualized and compared to the curve produced by the true observations (Supplementary Fig. 5).

Genetic context analysis

To identify mobile genetic elements (MGEs) associated with environment-specific ARGs, we implemented genetic context analysis. For each of the co-promoted, HG-promoted, WW-promoted, and non-promoted ARGs, we first retrieved genetic regions of up to 10,000 base pairs up- and downstream of the ARG from the genomes in which it was originally identified using GEnView v0.2 [57]. Here, if an ARG was identified in more than 1,000 genomes, the context was only retrieved from 1,000 randomly selected genomes to improve computational efficiency. The retrieved contexts were then screened for genes associated with MGEs as well as co-localized mobile ARGs. To identify genes involved in plasmid conjugation, the genetic regions were translated in all six reading frames using EMBOSS Transeq v6.5.7 [58] and screened with 124 hidden Markov models from MacSyfinder Conjscan v2.0 [59] using HMMER v3.1b2 [60]. Next, co-localized mobile ARGs were identified by using BLASTn v2.10.1 [61] to align the genetic regions against the ResFinder database [6], and finding the best (highest sequence identity) among overlapping hits with >90% nucleotide identity to a known ARG.

Analysis of genetic incompatibility

Next, we estimated the genetic incompatibility between ARGs from the identified categories and the included pathogens. This was computed by first selecting representative genomes for each pathogen from NCBI RefSeq [62]. The nucleotide sequences of the ARGs and genomes were divided into 5-mers, and their respective 5-mer distributions were computed. Then, we calculated the Euclidean distances between the 5-mer distributions of the ARGs in each category and each representative genome in turn. For each ARG category and pathogen, the resulting distributions of gene-genome 5-mer distances were visualized as boxplots, and the median distances were extracted and visualized together with the prevalence in that pathogenic host species.

We then expanded this analysis to estimate the genetic incompatibility between differently promoted ARGs and bacteria that are specific to human gut or wastewater. The human gut- and wastewater-specific bacteria were identified based on analysis of each quality-controlled metagenomic sample with Kraken2, with parameters –paired --confidence 0.1 --use-names [56]. For each sample, we identified bacterial genera to which $>0.01\%$ of the total reads mapped, which were considered present in that sample. Then, we identified genera which were present in $\geq 5\%$ of human gut samples and $\geq 5\%$ of wastewater samples, which were classified as universal, genera which were present in $\geq 5\%$ of human gut samples and $<5\%$ of wastewater samples, which were classified as human gut-specific, and genera which were present in $<5\%$ of human gut samples and $\geq 5\%$ of wastewater samples, which were classified as wastewater-specific. For each of the included genera, we then randomly selected up to ten individual species, discarding genera for which <5 species were available, for which representative genomes were obtained from NCBI RefSeq [62]. Finally, the genetic incompatibility between each of these genomes and the genes from each category was computed as described above.

Data availability

All raw data analyzed in this study has been retrieved from public repositories including NCBI [40, 44] and MGnify [38]. Accession numbers of the analyzed genomes and metagenomes are provided in Supplementary Data 1.

Code availability

Scripts used to perform the analyses and generate all figures presented in this paper are available via GitHub at https://github.com/davidgllund/promoted_args_hg_ww.

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Author contributions

D.L., A.J., M.H., J.S.I-D., J.B-P., and E.K. designed the study and developed the approach. D.L. downloaded the bacterial genomes and generated the reference ARG database. D.L. and M.H. retrieved the metagenomic data and analyzed it for ARG abundance. D.L. performed all additional computational analysis. All authors. discussed the results and their implications. D.L. and E.K. drafted and edited the manuscript. All authors edited and approved the final manuscript.

Competing interests

The authors declare no competing interests.