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Dysbiosis and early mortality in zebrafish larvae exposed to subclinical concentrations of streptomycin

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One sentence summary: Low concentrations of streptomycin found in aquatic environments disrupt the microbiomes of zebrafish larvae, increasing early mortality and favoring the spread of antibiotic resistance markers.

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ABSTRACT

Exposure to low concentrations of antibiotics found in aquatic environments can increase susceptibility to infection in adult fish due to microbiome disruption. However, little is known regarding the effect of antibiotic pollution on fish larvae. Here, we show that exposure to streptomycin, a common antibiotic used in medicine and aquaculture, disrupts the normal composition of zebrafish larvae microbiomes, significantly reducing the microbial diversity found in the fish. Exposure to streptomycin also significantly increased early mortality among fish larvae, causing full mortality within a few days of exposure at 10 µg/mL. Finally, we found that subclinical concentrations of streptomycin also increased the abundance of class 1 integrons, an integrase-dependent genetic system associated to the horizontal transfer of antibiotic resistance genes, in the larvae microbiomes. These results suggest that even low concentrations of streptomycin associated with environmental pollution could impact fish populations and lead to the creation of antibiotic resistance reservoirs.

Keywords: antimicrobial; pollution; 16S rRNA; metagenomic; *Danio rerio*

INTRODUCTION

The global use of antibiotics increased steadily over the past decades not only in human medicine but also in other sectors of commercial activities (Klein *et al.* 2018). For example, antibiotic consumption in livestock reached 63,151 tons in 2010 and is predicted to increase by 67% by 2030 (Van Boeckel *et al.* 2015). Antibiotic use is also rising in aquaculture as intensive farming expands as the fastest growing food sector worldwide (Henriksson *et al.* 2017). Consequently, antimicrobials of pharmaceutical origin can be found in terrestrial, freshwater and marine environments (Boy-Roura *et al.* 2018). Antibiotic residues can even be

found in atmospheric particulates near urban areas (Ferrey *et al.* 2018), contributing to the burden of antibiotic pollution (Martinez 2009).

In addition to agricultural (Li *et al.* 2018) and aquaculture runoffs (Henriksson *et al.* 2017), antibiotics are released into the environment through discharge of human sewage (Kostich, Batt and Lazorchak 2014) or industrial discharges (Bielen *et al.* 2017). Concentrations of antimicrobials found in the environment vary depending on the source, the antibiotic's persistence and the location affected. For example, antibiotic concentration found in treated sewage effluents can range from

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a few nanograms per liter for common antibiotics such as streptomycin (Kostich, Batt and Lazorchak 2014) to orders of magnitude greater than therapeutic concentrations in effluents affected by industrial plants (Cardoso, Porcher and Sanchez 2014; Larsson 2014). Even though concentrations of antimicrobials found in the environment usually do not exceed therapeutic dose (Kostich, Batt and Lazorchak 2014; Kulkarni et al. 2017), low concentrations of antibiotics can still lead to the evolution of resistant bacteria (Gullberg et al. 2011; Sandegren 2014). Indeed, antibiotic-resistant bacteria and genetic elements linked to antibiotic resistance are often detected in effluents near water treatment facilities, hospitals, pharmaceutical plants or more generally near urban areas (Gaze et al. 2011; Czekalski, Gascon Diez and Burgmann 2014). Unregulated antibiotic pollution in streams and rivers around the world is now considered to be an important driver of multidrug-resistance evolution (Wellington et al. 2013; Bengtsson-Palme, Kristiansson and Larsson 2018). Therefore, it is argued that future action plans to combat antibiotic resistance should consider the fate of antibiotics in the environment (Larsson 2014; Rosi-Marshall and Kelly 2015).

Antibiotic pollution in aquatic environments can also impact on different trophic levels in local ecosystems. For one, antibiotics can alter local microbial populations naturally present in sediments or water tables, disrupting phylogenetic structuration and nutrient cycling (Ding and He 2010; Martínez 2017; Grenni, Ancona and Barra Caracciolo 2018). Antibiotics inhibiting bacterial protein synthesis such as streptomycin can also exhibit toxic effects against unicellular algae, an important driver of aquatic ecosystem productivity (Fu et al. 2017). Similarly, low concentrations of antibiotics were shown to impact on the survival and behaviors of micro-invertebrates such as *Daphnia magna* (Flaherty and Dodson 2005) and macro-fauna such as zebrafish embryos (Bielen et al. 2017), via changes in gene transcription and expression (Zhang et al. 2016).

Finally, antibiotics can also affect microbial populations associated with animal hosts. Such microbial populations carry different functional roles for the hosts ranging from efficient nutrient metabolism to promoting bone formation (Yan et al. 2016; Raymann and Moran 2018). The disruption of a host microbiome, known as dysbiosis, can therefore lead to important health consequences (Lynch and Pedersen 2016). For example, streptomycin, an antibiotic commonly used in human medicine and agriculture, interferes with the intestinal microbiomes, leading to increased susceptibility to enteric infection (Sekirov et al. 2008; Schubert, Sinani and Schloss 2015) and modifying lymphocytes expression (Bazett, Bergeron and Haston 2016) in murine model systems. Although the role of microbiomes in fish is still relatively unexplored (Austin and Al-Zahrani 1988; Navarrete et al. 2008), recent studies showed that antibiotics administered as prophylactic in aquaculture reduced microbial diversity in fish gut (Navarrete et al. 2008), and also increased susceptibility to pathogens (Schmidt et al. 2017). Increased susceptibility to pathogens was also observed in adult zebrafish chronically exposed to low concentrations of streptomycin, leading to increased mortality (Zhou et al. 2018).

Here, we look at the impacts of low concentrations of streptomycin on the microbiomes of zebrafish larvae. Streptomycin is extensively used in human medicine, agriculture and aquaculture and thus is an important source of antibiotic pollution. Fish larvae are especially susceptible to chemical toxicants (Zhang et al. 2015) and are commonly used to screen for products toxic in the environment (Ali, van Mil and Richardson 2011). Using

this experimental model system, we investigated whether exposure to low concentrations of the antibiotic lead to dysbiosis and early mortality in the fish larvae. We also studied whether streptomycin favored the spread of integron 1, an integrase-based genetic element that allows the capture and expression of exogenous genes often associated to antibiotic resistance genes (Gillings 2014).

MATERIALS AND METHODS

Zebrafish husbandry and maintenance

Zebrafish (*Danio rerio*) were maintained in the laboratories of the Biology Department of Bard College in accordance with standard protocols for zebrafish husbandry (Lawrence and Mason 2012). Experimental populations of zebrafish strain Et20 were raised in a 14-h light:10-h dark cycle in standard recirculating rack water kept at 28.5°C with pH ranging from 7.0 to 7.4. Strain Et20 resulted from multiple outcrossing with wildtype strains in our laboratories and present no phenotypic differences, except for the GFP staining of mantle cells. At 0 days post-fertilization—or dpf, eggs from a single mating were bleached twice in 0.5% hypochlorite solution for 4 min and once in sterile 1 × E3 media (5 mM NaCl, 0.17 mM KCl, 0.33 mM CaCl₂, 0.33 mM MgSO₄ and 0.5mg/L methylene blue). We then placed 20–30 eggs in sterile petri dishes filled with 60 mL of 1 × E3 media. Every day, 80% of the media was exchanged for fresh media to minimize contamination due to dead cell tissue.

Following larval hatching at five (dpf), we established a control population by randomly picking 24 healthy embryos and allocating each individual in the well of a 24-well plate filled with 2 mL of sterile 1 × E3 media. We then established experimental populations to be exposed to streptomycin by randomly picking and allocating a total of 72 healthy embryos among the cells of three 24-well plates. Each plate was supplemented with either one of the following streptomycin concentrations (Streptomycin sulfate salt, Sigma-Aldrich, St-Louis, MO): 0.1, 1.0, 10.0 μg/mL. Even though streptomycin can degrade in the environment under the effect of light and temperature among other things, all three streptomycin concentrations used in this were previously reported in environmental surveys (Shen et al. 2017). Again, approximately 80% of media was exchanged daily for sterile 1 × E3 media with the relevant antibiotic concentration when applicable. Streptomycin solution was prepared and stored at –10°C as per manufacturer protocols (Sigma-Aldrich, St. Louis, MO) before use. Fishes were fed generously once daily with micro-powder food containing rotifers and *Paramecium* as well as one drop of Roti Feast (Reed Mariculture, CA). We maintained the fishes in experimental conditions for 10 days and measured survival daily by gently poking immobile embryos with a sterile pipette tip to confirm the absence of movement.

At the end of the experiment, five embryos from each treatment were anesthetized on ice for 10 min and sacrificed by overdose of sterile tricaine methane sulfonate (250 mg/l) according to established euthanasia techniques (Matthews and Varga 2012). Only live fishes or fishes that had died within the last 18 h were chosen for downstream application. Fish were removed from tricaine following cessation of opercular movement (~10 min) and washed three times with nuclease free water to minimize the presence of free-living bacteria. All experimental procedures were approved by the Bard Institutional Animal Care and Use Committee (IACUC; most recent approval ID 'Perron 2018').

DNA extraction and processing

We extracted and purified microbial DNA from each sacrificed fish larvae using a modified protocol for the DNeasy Blood and Tissue Kit (QIAGEN, Germantown, MD) first described in Hang et al. (2014) and as implemented in Dahan et al. (2018). When necessary, gDNA concentrations were increased using the SpeedVac System (ThermoFisher Scientific, Asheville, NC). Purified gDNA samples were stored in nuclease free H₂O at -20°C.

Fish microbiomes were characterized via targeted gene amplification of the 16S rRNA V4 region using Golay-barcoded primers 515F and 806R (Caporaso et al. 2012). PCR conditions were previously described (Dahan et al. 2018). Following gel-purification, libraries were pooled at equimolar ratios, and sequenced on the MiSeq paired-end Illumina platform adapted for 250-bp paired-end reads (Wright Labs, Juniata College, Huntingdon, PA). All sequence reads are available at the Sequence Read Archive of the NCBI (accession number SRP139123).

Processing of 16S rRNA sequence data

Before processing the 16S rRNA sequence data, phiX control reads were removed by mapping raw sequence reads against an indexed phiX genome provided by Illumina (San Diego, CA). Sequence reads were then processed, aligned and categorized independently using the DADA2 pipeline version 1.6 (Callahan et al. 2016) available at (<https://github.com/benjjneb/dada2>) and implemented in R version 3.2.3 (<http://www.r-project.org>).

The DADA2 pipeline characterizes microbial communities by identifying unique amplicon sequence variants (ASVs) within the 16S rRNA reads (Callahan, McMurdie and Holmes 2017). In brief, forward and reverse reads are first filtered using the pipeline's recommended parameters. Filtered reads are then de-replicated (i.e. combining identical reads into unique sequences with a consensus quality profile) and de-noised (i.e. inferring and removing errors from samples) using DADA2's default parameters. After building the ASV table and removing chimeras, taxonomy was assigned using the 'assignTaxonomy' function of DADA2 trained against the SILVA ribosomal RNA gene database version 132 (Quast et al. 2013; Koo et al. 2017). A multiple alignment of all the ASVs and maximum likelihood phylogenetic tree was built using the phangorn package version 2.1.3 (Schliep 2011). Lastly, we removed sequences positively identified as either zebrafish mitochondrial DNA or chloroplast DNA.

Data visualization and statistical analyses of 16S rRNA sequence data

Patterns of diversity within the ASV tables were analyzed using a modified version of the pipeline described in Dahan et al. (2018) implemented in R and visualized in ggplot2 (Wickham 2016). A mapping file linking sample names and streptomycin concentration is provided in Supporting Information. Briefly, we identified the core microbiome among control zebrafish using the pipeline described in Dahan et al. (2018). Then, using phyloseq version 1.14.0 (available at <https://joey711.github.io/phyloseq/>), performed Principle Coordinate Analyses (PCoA) on unweighted UniFrac distance scores (Lozupone and Knight 2005) to identify differences in community composition between samples. We conducted the analyses first by comparing control microbiomes to all microbiomes exposed to different concentrations of streptomycin and then by considering each concentration individually. We then tested the effect of streptomycin

as a continuous variable on using a permutational analysis of variance using the adonis function (Oksanen et al. 2007) of vegan version 2.3.2. To confirm multivariate homogeneity of variances among and within treatments, we used vegan's package implementation of PERMDISP2 via the betadisper method (Oksanen et al. 2007).

We then analyzed community composition using phyloseq and DESeq2 differential abundance comparison (Love, Huber and Anders 2014) adapted for use with microbial count data (McMurdie and Holmes 2014). Core microbiomes were estimated using a custom pipeline (Dahan et al. 2018). Finally, we estimated alpha diversity metrics using phyloseq's estimate_richness function. More specifically, we estimated richness S, i.e. observed number of ASVs, Shannon diversity measurements (H'), as well as Pielou's evenness index (Pielou 1966) from the ASV table subsampled to the lowest sampling depth of 18,477 paired-reads. We then tested whether diversity indices changed with streptomycin concentration using linear modeling and comparing the different statistical models with Akaike's Information Criterion as implemented in R's stats package.

qPCR of 16S rRNA and int1 genes

To quantify the relative abundance of integron 1, we estimated the abundance of *int1* gene, a genetic marker of integron 1 commonly used as an indicator of antibiotic pollution (Gillings et al. 2015), in relation to the abundance of 16S rRNA genes found in each sample. We used quantitative real-time PCR conditions as described in Gaze et al. (2011). For each sample, triplicate PCR reactions using the PowerUp SYBR Green Master Mix (Applied Biosystems, Foster City, CA) were cycled using the Bio-Rad CFX96 Real-Time PCR Detection System (Bio-Rad Laboratories, Hercules, CA). An internal standard curve constructed from a serial dilution of *Escherichia coli* SK4903 harboring seven 16S rRNA copies and on average six *int1* copies was processed with each qPCR run. Strain SK4903 contains plasmid R751, which belong to the IncP1 family, known to be one of the most stably maintained plasmid (Grinter 1984; Adamczyk and Jagura-Burdzy 2003). The relative abundance of *int1* genes was normalized to 16S rRNA copies of each sample before being analyzed via linear modeling with *int1* relative abundance as the response variable and streptomycin concentrations as the explanatory variable. Statistical analysis and model assumption testing were conducted in R and visualized in ggplot2 (Wickham 2016).

RESULTS

Processing 16S rRNA V4 reads using DADA2

Overall, we sequenced the V4 region of the 16S rRNA genes to characterize the microbiomes of 10 individuals from the control group and five individuals from each of the three streptomycin treatments (i.e. 0.1 µg/mL, 1.0 µg/mL and 10.0 µg/mL) for a total of 25 individuals. We obtained a total of 3487,592 pairs of forward and reverse reads with an average read length of 250 base pairs, totaling ~1.7 G bases and a median sequencing depth per sample of 151,631 paired-reads. After removing reads mapping as phiX and trimming forward reads at 240 bp and reverse reads at 160 bp, we retained a total of 6078,045 (87.71% of initial) reads. Finally, following de-replication, de-noising and removing predicted chimeric reads, we were left with 2662,658 (76.34% of the initial) paired-reads.

Using the RDP classifier, we identified a total of 344 unique ASVs among the 25 sampled zebrafish larvae. When looking

Table 1. Relative abundance of core phyla and genera in control zebrafish larvae.

Phylum Genus	Mean %	s.d
Proteobacteria	48.82	12.39
<i>Rhodovarius</i>	5.75	0.55
<i>Bosea</i>	0.18	0.16
<i>Phenylobacterium</i>	0.37	0.28
<i>Shinella</i>	2.86	2.86
<i>Pseudomonas</i>	0.75	0.91
<i>Vibrio</i>	0.12	0.13
<i>Pelomonas</i>	22.05	14.78
<i>Xylophilus</i>	18.50	24.69
<i>Pseudacidovorax</i>	3.41	5.59
Bacteroidetes	51.18	12.39
<i>Fluviicola</i>	5.88	2.80
<i>Chryseobacterium</i>	5.50	5.86
<i>Flectobacillus</i>	39.81	14.10

specifically at the larvae from the control group, we found that 12 ASVs were shared amongst every individual (Table 1), suggesting a high instability of the microbiome at early stage of development in zebrafish. As expected from previous work (Roeselers et al. 2011; Stephens et al. 2016), we found taxa predominantly belonging to the phyla Proteobacteria and Bacteroidetes. Among the most common genera found in the core microbiome of control fishes were genera *Flectobacillus* sp., 39.81(14.1)%, and *Fluviicola* sp., 5.88(2.80)%, two Bacteroidetes that were previously associated with zebrafish (Davis et al. 2016; Falcinelli et al. 2016).

Streptomycin exposure alters microbial community composition

We first tested whether overall streptomycin exposure resulted in changes in the overall microbial community composition of zebrafish larvae. Using PCoA on unweighted UniFrac distances, we found that the two highest-ranked dimensions explained 57.24% and 14.03% of variance respectively, and that microbiomes exposed to any concentration of streptomycin clustered away from control microbiomes (ADONIS; $R^2 = 0.59$; PERM = 999; adj- $P = 0.006$; Fig. 1A). We found that streptomycin explained less variance when concentration was treated individually (ADONIS; $R^2 = 0.31$; PERM = 999; adj- $P = 0.003$; Fig. 1B), suggesting that UniFrac scores converged for microbiomes exposed to any concentration of streptomycin when compared to control fishes. We found a small difference in variance homogeneity when comparing controls to streptomycin combined together ($F_{(1,18)} = 16.99$; adj- $P > 0.01$), but no differences in sample distances to their treatment centroids when treating streptomycin concentrations individually ($F_{(2,17)} = 4.42$; adj- $P > 0.05$).

The main change in microbial community was a shift from a more or less equal mixture of Proteobacteria and Bacteroidetes in healthy microbiomes to communities dominated by Proteobacteria when fishes were exposed to streptomycin (Fig. 2A). When looking at changes in ASV specifically, we found that a total of 26 ASVs differed in abundance between control microbiomes and microbiomes exposed to streptomycin (DESeq2; adj- $P \leq 0.01$; Fig. 2B). Indeed, six ASVs, classified within four genera, increased in abundance in the presence of streptomycin, while 22 ASVs, including 11 ASVs identified as part of the core microbiome in healthy zebrafish, were significantly less abundant in the presence of streptomycin (Fig. 2B). Three ASVs that increased in abundances were identified as *Sphingomonas*, a genus

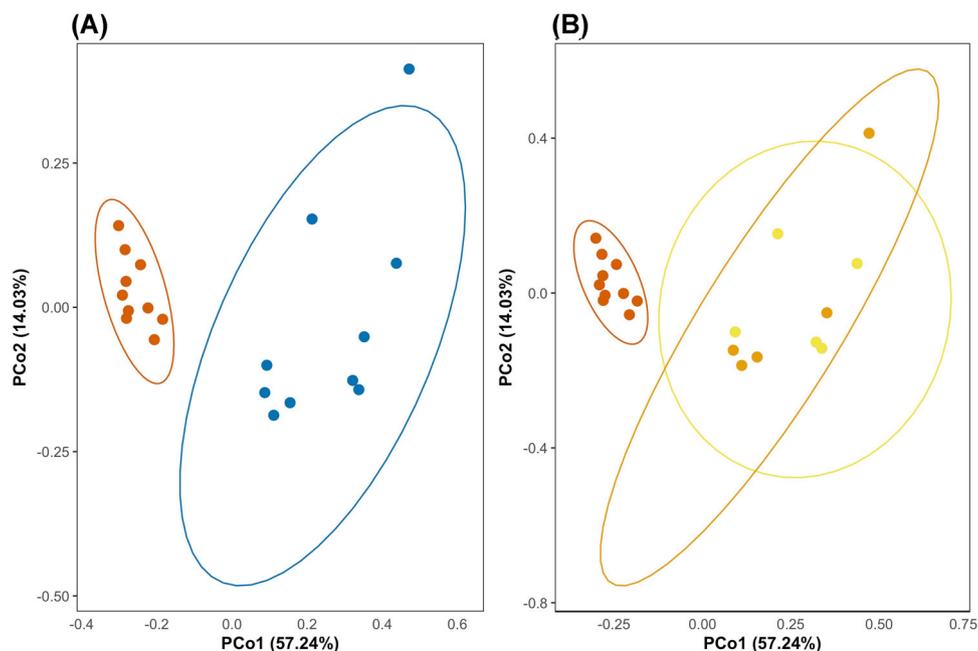


Figure 1. Principle coordinates analysis (PCoA) of zebrafish microbiota exposed to different streptomycin concentrations. (A) PCoA of unweighted UniFrac scores shows significant dissimilarity between control microbiomes and microbiomes exposed to streptomycin (ADONIS; $R^2 = 0.59$; PERM = 999; adj- $P = 0.006$). Control microbiomes are shown in vermillion while microbiomes exposed to streptomycin are in blue. (B) PCoA of unweighted UniFrac scores shows weaker dissimilarity among treatments when streptomycin is treated as a continuous variable (ADONIS; $R^2 = 0.31$; PERM = 999; adj- $P = 0.003$). Control microbiomes are shown in vermillion while microbiomes exposed to 1.0 and 10 $\mu\text{g}/\text{mL}$ streptomycin are shown in orange and yellow, respectively. Ellipses are drawn at 0.95 C.I.

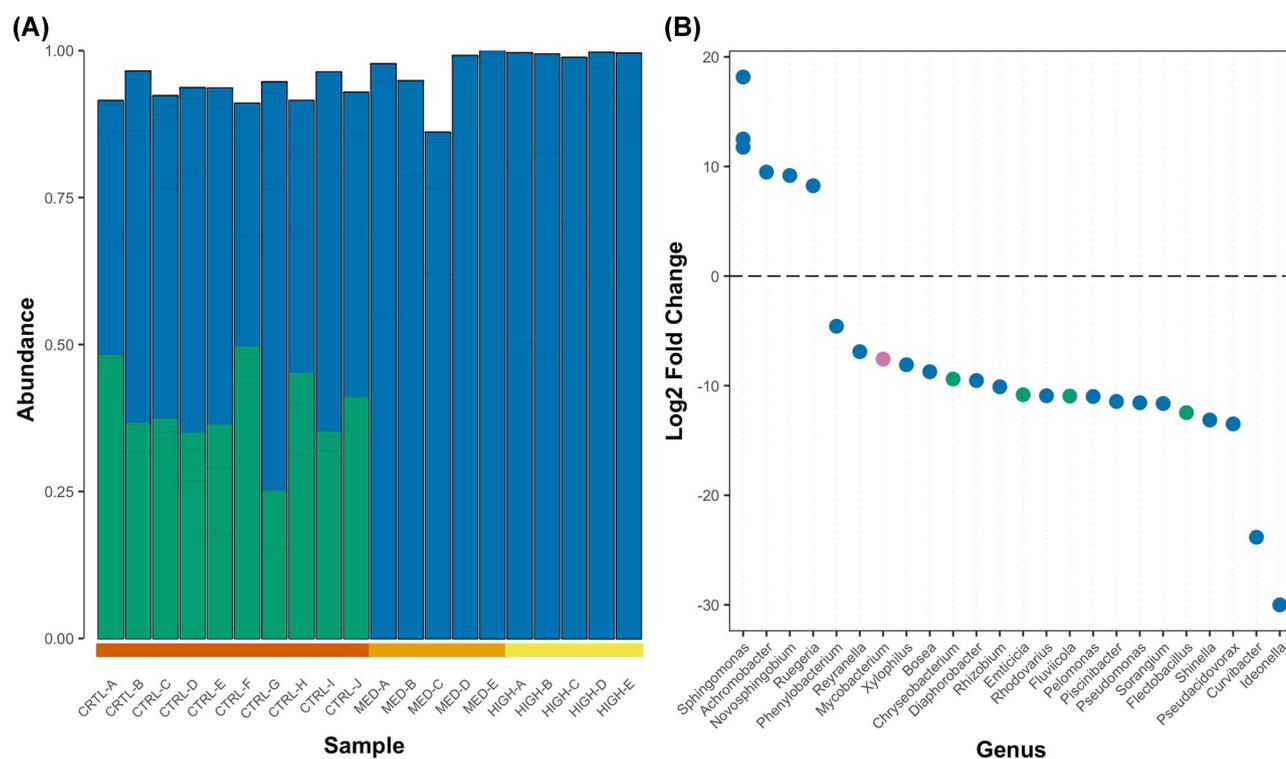


Figure 2. Taxa composition in zebrafish microbiomes exposed to different streptomycin concentrations. (A) Relative abundance of Phylum level classification. Each bar within a column represents a unique ASV. Taxa with abundance below 0.1% were removed. (B) Changes in ASVs abundance in the presence of streptomycin (DESeq2; adj-P < 0.01). Genus level classification is provided where available. Genera belonging to Proteobacteria are shown in blue while Bacteroidetes are shown in green and Actinobacteria are shown in lilac.

Table 2. Mean (s.d) of alpha diversity metrics.

Treatment	Observed	Shannon	Pielou
Control	40.0 (8.49)	1.70 (0.22)	0.46 (0.058)
1 $\mu\text{g}/\text{mL}$	26.4 (12.72)	0.88 (0.58)	0.27 (0.15)
10 $\mu\text{g}/\text{mL}$	20.2 (7.89)	0.98 (0.12)	0.34 (0.068)

commonly isolated from environmental samples using streptomycin as a selective agent (Vanbroekhoven et al. 2004). These results taken together suggest that streptomycin exert important selective pressures on the microbiomes of zebrafish larvae, increasing the relative abundance of bacteria resistant or tolerant to the antibiotic.

Streptomycin exposure alters alpha diversity in fish microbiome

In addition to impact on community composition, we also tested whether streptomycin affected the overall taxonomic diversity present in zebrafish larvae. Overall, we found that taxonomic diversity was lower in microbiomes exposed to streptomycin (Table 2; Fig. 3). Interestingly, while the total number of observed ASVs decreased linearly as streptomycin concentration increased ($F_{(1,18)} = 7.58$; adj-P = 0.03; Fig. 3A), we found that taxonomic diversity indices that accounts for ASV evenness, even though still lower in treated microbiomes, were slightly larger at 10 $\mu\text{g}/\text{mL}$ than at 1 $\mu\text{g}/\text{mL}$. The quadratic relationship showed by

the Shannon index ($F_{(1,18)} = 7.58$; adj-P = 0.03; Fig. 3B) and Pielou's metric ($F_{(1,17)} = 9.97$; adj-P = 0.01; Fig. 3C) as streptomycin concentration increases suggests that even a small concentration of the antibiotic is sufficient to disrupt the microbiome diversity, allowing relatively few taxa to dominate the microbiomes.

Streptomycin increases *int1* abundance in fish microbiome

To investigate the possible link between streptomycin exposure and the spread of integrons in microbial communities, we also measured the abundance of *int1* relative to the number of 16S rRNA copy numbers in each microbiome. We found that exposure to streptomycin had a significant effect on the relative abundance of *int1* in larvae microbiomes ($F_{(3,20)} = 4.33$; $P < 0.016$; Fig. 4). The effect of streptomycin was especially strong at 0.1 and 1.0 $\mu\text{g}/\text{mL}$ where we found on average one *int1* gene copy number for every four 16S rRNA copy number, a relative abundance 19 times higher than in control populations. Even though *int1* relative abundance was on average 13 times higher in fish exposed to 10 $\mu\text{g}/\text{mL}$ than in control fish, this difference was not significant due to the large variance in *int1* relative abundance in high streptomycin concentration ($P = 0.21$). Using a dissimilarity matrix comparing PCoA to *int1* abundance, we found that *int1* weakly correlate with UniFrac clustering between the different samples (ADONIS; $R = 0.12$ $P = 0.068$; perm = 999; Fig. S1, Supporting Information), indicating that increases in *int1* relative copy number was likely associated with one or few ASVs rather than whole community shifts.

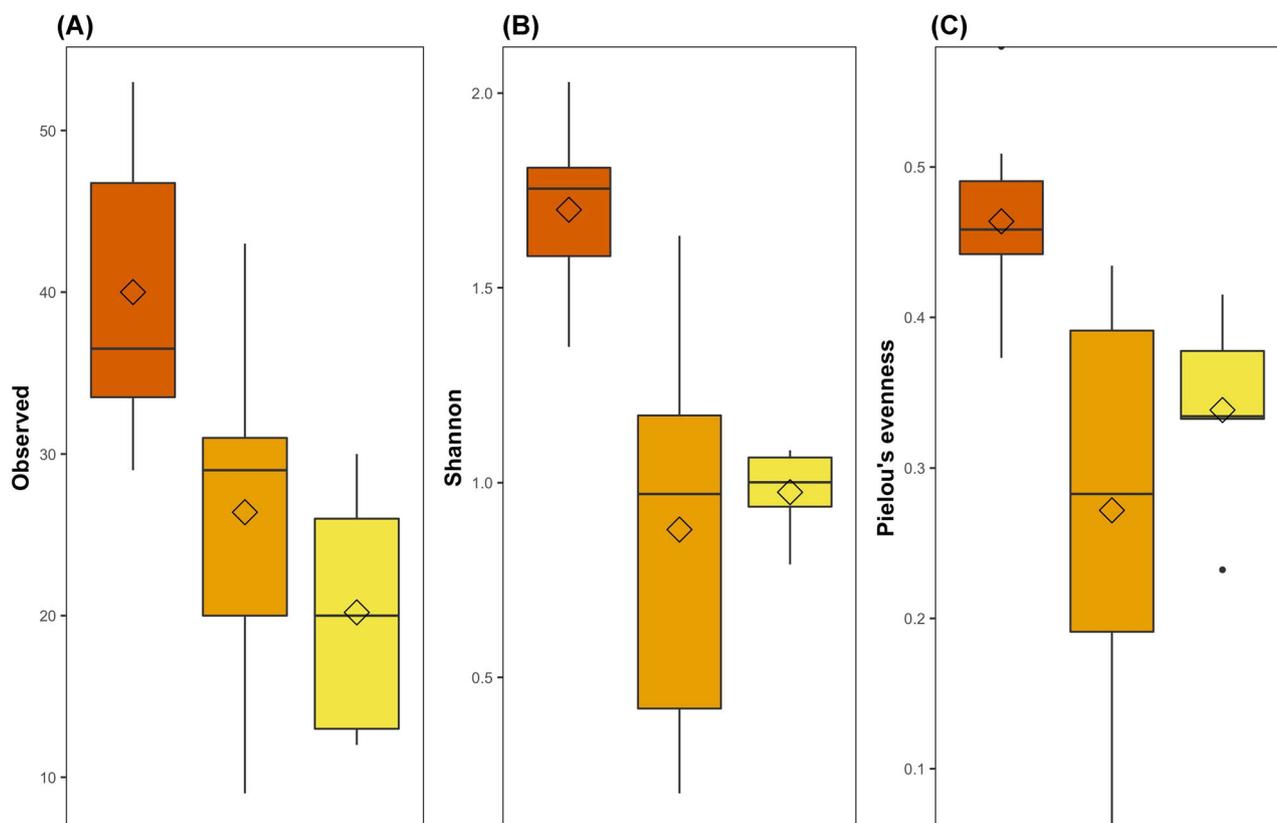


Figure 3. Alpha diversity metrics in zebrafish microbiomes exposed to different streptomycin concentrations. (A) Observed ASVs ($F_{(1,18)} = 7.58$; adj- $P = 0.03$). (B) Shannon diversity ($F_{(1,18)} = 7.58$; adj- $P = 0.03$). (C) Pielou's evenness ($F_{(1,18)} = 7.58$; adj- $P = 0.03$). All metrics are plotted against streptomycin concentration with controls microbiomes shown in vermillion and microbiomes exposed to 1.0 and 10 $\mu\text{g/mL}$ in orange and yellow, respectively. Diamonds represent population mean while lines represent median and hinges indicate first and third quartiles.

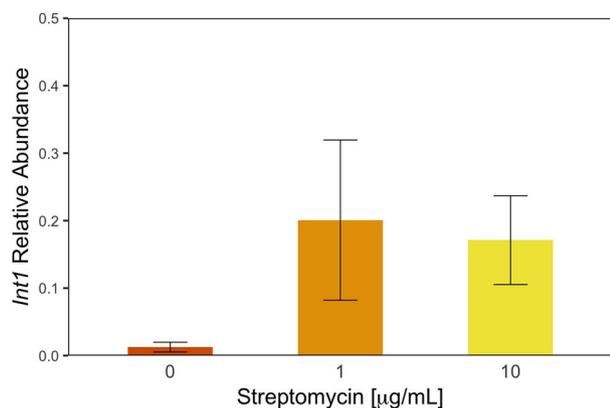


Figure 4. *Int1* relative abundance in zebrafish microbiomes exposed to different concentrations of streptomycin. Relative abundance of *int1* is calculated as the proportion of *int1* copy numbers per 16S rRNA copies and was higher in microbiomes exposed to streptomycin ($F_{(3,20)} = 4.33$; $P < 0.016$). Error bars are standard error of the mean and controls microbiomes are shown in vermillion while microbiomes exposed to 1.0 and 10 $\mu\text{g/mL}$ streptomycin are shown in orange and yellow, respectively. For consistency, 0.1 $\mu\text{g/mL}$ treatment was removed from figure.

Exposure to streptomycin increases early mortality in zebrafish larvae

Finally, we investigated whether streptomycin affected larval survival by monitoring the growth of 24 independent larvae ex-

posed to 0, 0.1, 1.0 and 10 mg/mL for a period of 10 days. We found that early mortality was strongly affected by the presence of streptomycin ($\chi^2_{(3, N=240)} = 161$; $P < 0.001$). While we observed a total of five deaths in control fishes, we observed mortality in nine larvae out of 24 fish exposed to 10 $\mu\text{g/mL}$ of streptomycin by the fourth day of treatment and complete mortality by day seven. Fish exposed to 0.1 and 1.0 $\mu\text{g/mL}$ also showed increased mortality compared to the control population with a total of 11 (45.8%) and 20 (83.3%) mortalities respectively by the end of the experiment. Whether increased mortality resulted from dysbiosis or the possible toxicological effect of streptomycin (see Owens et al. (2009)) remains to be determined. Because streptomycin is commonly used in standard zebrafish husbandry protocols to minimize contamination without apparent effect on survival (Nusslein-Volhard and Dahm 2002; Melancon et al. 2017), however, we believe that the decrease in survival rate observed in this study is, at least partly, explained by the consequences of dysbiosis.

DISCUSSION

The possible effects of antibiotic pollution found in rivers and streams on the microbial populations associated with animal hosts are mostly unknown. Here, we show that low concentrations of streptomycin associated with antibiotic pollution can result in important changes in the microbial communities associated with zebrafish larvae, an important model system. In addition, we found that low concentrations of

streptomycin increased the relative abundance of integron 1 in the host microbiomes and increased the onset of early mortality of larvae. These results suggest that even low concentrations of antibiotics found in polluted aquatic environments could have important consequences on animal populations.

Compared to mammals, fishes show a wider variability in microbiome composition, differing greatly within a species depending on diet and environmental conditions (Bolnick et al. 2014; Smith et al. 2015; Schmidt et al. 2017). In line with previous findings, we found that the larval zebrafish microbiomes are highly heterogeneous and likely influenced by their environment. For example, many of the genera identified as part of the core microbiomes in healthy zebrafish larvae were previously associated with aquatic environments rather than with animal hosts. When considering ASVs that were present in at least 8 out of 10 larvae, we found additional core taxa such as *Fusobacterium* sp. that are associated with zebrafish larvae and other teleost fishes (Roeselers et al. 2011).

Given the lack of information on fish microbiomes, it is hard to predict the possible effect of antibiotics on fish microbiomes. The few studies done on the topic found that fathead minnows recovered quickly from changes in gut microbiomes following exposure to triclosan (Narrowe et al. 2015), while other studies showed that exposure to different antibiotics caused early mortality in adult black molly, *Poecilia sphenops*, without significant changes in microbiome composition (Schmidt et al. 2017) and in adult zebrafish (Zhou et al. 2018). Here, despite high heterogeneity at the genus level, we found that streptomycin had a predictable effect on the overall community composition of zebrafish larvae (Fig. 2). Namely, we found that even the lowest concentrations of streptomycin caused a reduction in overall diversity and caused a shift from a mixture of Proteobacteria and Bacteroidetes in healthy fish to communities dominated by Proteobacteria in fish exposed to the antibiotic. Even though zebrafish microbiomes normally develop towards communities dominated by Proteobacteria, this shift is not expected until much later in the life cycle of the animal (Stephens et al. 2016).

The most abundant Proteobacteria in our study was identified as *Sphingomonas*, a genus not normally associated with healthy microbiomes in either larvae or adult zebrafish (Stephens et al. 2016). In fact, the *Sphingomonas* genus contains species known to be pathogenic in humans and to harbor resistance against streptomycin (Vanbroekhoven et al. 2004). Similarly, we observed an increase of integron 1 genetic elements in microbiomes treated with streptomycin, an important indicator of antibiotic resistance in the environment (Gillings et al. 2015). We found that the presence of integrons did not correlate specifically with any ASV in our study, suggesting that integrons either spread horizontally across multiple ASVs or that we did not have the statistical power to detect such correlation. Either way, our results thus demonstrate that even low concentrations of antibiotics could contribute to the creation of environmental reservoirs of pathogenic bacteria, pathobionts and antibiotic resistance in animal hosts (Perron, Quessy and Bell 2008).

Finally, mortality following antibiotic exposure was previously observed in fishes. In two cases, mortality was due to the host's increased susceptibility to pathogens (Schmidt et al. 2017; Zhou et al. 2018). While it is likely that the increase in early mortality observed in this study is due at least in part to dysbiosis, the effect could be due to streptomycin toxicity (Klis et al. 2014). While a previous studies showed only limited acute toxicity of streptomycin on *Daphnia* (Wollenberger, Halling-Sørensen and Kusk 2000) and no acute effects on mammals and birds (Seyler

and Extension 1994), zebrafish embryos at 3 dpf absorb more drug than at any other developmental periods (Zhang et al. 2015) and could therefore be more sensitive to the effect of streptomycin.

The extent to which low concentrations of antibiotics affect fishes and other aquatic organisms remains to be fully investigated. For example, further study is required to understand whether exposure to low concentrations of antibiotics conveys similar effects in adult. Still, our results demonstrate that antibiotic pollution in aquatic environments can not only contribute to the burden of antibiotic resistance but also can have important consequences on the health of exposed organisms. The significant increase in early mortality observed in this study could have a dramatic effect on endangered species with precariously small population size.

SUPPLEMENTARY DATA

Supplementary data are available at [FEMSLE](#) online.

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Conflicts of interest. None declare.

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