Ripples of Resistance: Unveiling Antimicrobial Resistance Dynamics Along Switzerland's Aare River

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Abstract

The global spread of antimicrobial resistance (AMR) is a serious public health concern, driven by widespread antibiotic use and the global environmental circulation of antibiotic-resistant bacteria and resistance genes (ARGs). Wastewater treatment plants (WWTPs) are important sources of anthropogenic AMR entering large rivers, which serve as vital water resources but facilitate downstream dissemination. The drivers and dynamics of AMR propagation along river systems remain poorly understood. As Switzerland's longest and one of its largest rivers, the Aare, situated in the upper Rhine watershed, plays a central role in the 'water castle of Europe'. This study examines the impact of WWTP discharges, some receiving high loads of hospital effluent, on ARG distribution along the 288 km Aare river-continuum. Using quantitative PCR targeting 14 ARGs conferring resistance to eight antibiotic classes, combined with *16S rRNA* gene amplicon sequencing, we conducted a high-resolution spatial survey to assess shifts in the riverine resistome and microbiome. Concentrations of trace metals and nutrients were analyzed as tracers of anthropogenic inputs. Results revealed a progressive increase in ARG abundance downstream, driven by WWTP effluents enriched in ARGs. Effluents had 70-fold higher mean ARG concentrations than upstream waters, raising downstream levels up to 141-fold. Major tributaries such as the Reuss and Limmat sustained elevated ARG levels, while passage through lakes markedly reduced concentrations. This study provides the first detailed baseline for ARG prevalence along a large river system, from pristine headwaters to pollution-affected lower reaches and insights into aquatic AMR dynamics and guidance for future monitoring.

Key words: antibiotic resistance genes, river continuum, wastewater, anthropogenic tracers, quantitative PCR, microbial community

1. Introduction

Bacterial antimicrobial resistance (AMR) has emerged as a major global health and development threat of the 21th century[1]. In 2019, an estimated 4.95 million deaths were associated with AMR, including 1.75 million deaths directly linked to bacterial AMR[1]. According to projections by the UK Government, by 2050 AMR could even be responsible for 10 million deaths, underscoring the urgent need for global action[2,3].

Although antimicrobial resistance genes (ARGs) occur in pristine environments, the levels of resistance observed today are not solely due to naturally occurring resistances, e.g. in antibiotic-producing bacteria[4–7]. Instead, the widespread use of antimicrobials in healthcare and livestock farming has significantly contributed to the spread of antibiotic-resistant bacteria (ARB) and ARGs in ecosystems[8–10]. Human and animal wastes contaminated with antibiotics, ARB and ARGs are continuously released to the environment, either untreated (e.g., agricultural run-off, aquaculture, sewer overflows) or partially treated in wastewater treatment plants (WWTPs). Generally, WWTPs receiving wastewater (WW) from various anthropogenic sources of AMR, including urban areas, hospitals, and specific industries (e.g., pharmaceutical industry and industrial slaughterhouses) are not designed for efficient micropollutant removal or disinfection[11–15]. Consequently, WWTPs have been found to be a major source of ARGs, facilitating their dissemination into aquatic environments[12,16]. While WWTPs quantitatively reduce ARG and ARB loads[17,18], they also serve as hotspots for ARG persistence, horizontal gene transfer (HGT), and bacterial adaption, with effluents exhibiting high transcriptional activity and highly mobilized ARGs[18–20].

Aquatic ecosystems may then contribute to ARG dissemination by acting as a reservoir and conduit of received resistance determinants. This enables acquisition, evolution, and recombination of

resistance genes through HGT of mobile genetic elements (MGEs) in pathogenic and clinically relevant bacteria, driving the emergence of new resistance traits[21]. Effluent discharge has well-documented impacts on the resistome of downstream river ecosystems, with ARG primarily persisting in the water column[22,23]. While ARG levels were observed to decrease over short distances (~2 km) from a point source by dilution, and potentially also via sedimentation and degradation, mass balance and fluctuations observed over longer distances (> 10 km) suggest AMR removal, local proliferation and influence of non-point sources on riverine AMR levels[24]. This highlights the role of aquatic environments in the long-term spread and evolution of AMR.

Rivers serve as critical interfaces between humans, animals, and the environment. Flowing for hundreds of kilometres, they act as conduits for WW discharge, carrying (micro)pollutants (e.g., organic chemicals, pharmaceuticals, pesticides, and detergents), as well as ARB and ARGs to downstream ecosystems across landscapes and even continents[11,12,25]. While antibiotics are generally expected to degrade, although at variable rates, the case is less clear for ARB and ARGs[26]. These can decrease through various mechanisms but can potentially also persist for decades[27] or even proliferate under suitable conditions[24]. These contaminants can re-enter the microbiomes of human and animal populations when river water is used for drinking, irrigation, and recreational activities, potentially leading to colonization or in the worst case, causing skin, gastrointestinal, urogenital, and respiratory infections with ARB[28–30].

This interconnectedness is of particular importance in Switzerland, often called the "Wasserschloss Europas" (engl. "Water castle of Europe") due to its extensive river networks and vital freshwater resources feeding into major European river systems such as the Rhine, Rhone, and Danube. Swiss rivers shape regional water quality and contribute to the broader European hydrological system,

affecting pollution levels in waters across the Swiss border[31]. Protecting these waterways is essential

for mitigating the spread of AMR, safeguarding drinking water sources, and maintaining the

ecological and public health of large parts of Europe.

Previous studies on AMR in larger river systems exposed to strong anthropogenic pressure have

mostly targeted heavily polluted sites. While recent research has focused on large European rivers

like the Danube[32] and Rhine[33], as well as rivers in China[34–37] and the USA[23,38], no comparable studies

were conducted in Switzerland. In general, studies with high-resolution longitudinal sampling and

simultaneous examination of potential point sources of contamination are lacking.

This study aimed to establish the anthropogenic impact on AMR prevalence along Switzerland's Aare

River, from its source to its confluence with the Rhine River. Originating from the Ober- and Unteraar

glaciers in the pristine alpine Grimsel region and thus lacking upstream pollution, it flows

downstream through increasingly agricultural areas and densely populated cities, where

anthropogenic impact increases. We further explored the hypothesis that WWTP effluents in general,

and those from WWTP receiving hospital WW in particular, are the primary sources of AMR relative

to agricultural sources. Additionally, the study explored potential sinks along the river course,

assessing whether AMR is removed, persists, or proliferates. Finally, we investigated whether this

system's microbial community structure drives AMR dynamics.

2. Materials and Methods

2.1. Study area and sample collection

The Aare, Switzerland's longest river, drains approximately 43% of the country (catchment size:

17'709 km^2) with a mean annual discharge of 555 m^3/s (Gauge Station Siggenthal[39]). The Aare River

originates in the high Alps, fed by glacial sources, and flows over 288 km through landscapes

characterized by increasing intensity of agricultural and industrial use, population density and

urbanization prior to its confluence with the Rhine (**Fig. 1a**). **S1.1** provides a detailed description of

the river. Switzerland's temperate climate, with year-round precipitation and distinct seasonal

variations, strongly influences discharge, with spring snowmelt being a major contributor.

Field campaign (04 to 14 March 2024) targeted expected low flow conditions prior to snowmelt. In

total, 49 samples were collected from the Aare River, its main tributaries and potential contamination

sources (**Fig. 1b, Tab. S1**). Effluents of nine municipal WWTP discharging into the Aare River and

its tributaries were sampled, seven of which also serve large hospitals (Interlaken, Thun, Bern, Biel,

Solothurn, Aarau, and Baden). Two WWTPs (Lyss and Oensingen) receive pre-treated WW from

industrial slaughterhouses. At Lyss, an additional sample was collected at the pre-cleaning step. Six

tributaries with catchments > 500 km^2 (decreasing catchment size: Reuss > Zihl > Limmat > Saane >

Kander > Emme) were sampled near their confluence with the Aare River. Samples downstream of

WWTP or tributary inflows were taken 6 – 10 km downstream, where possible, to ensure complete

mixing. For samples from the Aare, and points of WWTP discharge and tributary confluence the

distance from the source (in river kilometers, rkm) was determined from GPS coordinates of sampling

points and a shapefile of the Aare river in QGIS Maidenhead[40].

Figure 1: Sampling sites and land use within Aare River's catchment, Switzerland. Lakes are
shown in light blue, and the course of the Aare River and its tributaries is indicated in two shades
of blue. (a) A zoomed-in version of the Aare catchment, depicting land-use distribution across the
Aare and its tributary catchments. Unproductive areas are shown in grey, agricultural areas in
yellow, wooded areas in green, and settlement areas in red. (b) Map illustrating 49 sampling
locations along the Aare River, the longest river in Switzerland. Sampling includes 10 effluent
samples from municipal WWTPs receiving WW from either industrial slaughterhouses (2 samples)
or hospitals (8 samples), 26 samples from the main Aare River, and 13 samples from its tributaries
(details in Table S1).

Samples were primarily collected from bridges to capture mid-river water at ~30 cm depth. A Comet-

Combi 12 4T pump (Comet Pumps, Pfefferschwanden, DE) was used. To avoid contamination, the

pump, tubing and 5 L plastic sample bottles were rinsed thrice by pumping ~ 5-10 L of river water

prior to sampling. Per location 2 x 5 L samples were taken over ~5 minutes for biological analyses

except for WWTP effluents (1 x 5 L). Subsamples were obtained for microbial, chemical, and nutrient

analysis. A separate grab sample was obtained with a metal-free sampling bucket for trace metal

analyses. Metadata (temperature, pH, conductivity, and dissolved oxygen) were recorded in situ using

a WTW multi-parameter device, MultiLine 3630 IDS (Fisher Scientific, USA).

2.2 Flow cytometry (FCM)

Total cell count (TCC), and ratios of high nucleic acid (HNA) and low nucleic acid (LNA) cells were

determined by FCM on a NovoCyte Advanteon flow cytometer (Agilent, USA) according to the

Swiss accredited Standard method 333.1, based on SYBRgreen staining[41]. The procedure followed

the methodology described by Proctor and Besmer et al.[42]. Negative controls with sterile-filtered

Evian water were included. The gating strategy was recalibrated to ensure accurate differentiation of

microbial cells from other particles (see **S1.2, Fig. S1**).

2.3 Biomass Filtration in the Field

Aliquots of all collected water samples were filtered in the field using S-Pak mixed esters of cellulose

Membrane filters (GSWG047S6, Millipore, USA) with a pore size of 0.22 μm and a diameter of 47

mm. Filters were placed in sterilized polysulfone filter holders (Ref. No. 10461100) with a diameter

of 50 mm (Whatman, Germany) under sterile conditions. A peristaltic pump type V6-3L (Shenchen,

China) with two pump heads equipped with silicon tubing (# 35 from the manufacturer) was utilized

for the filtration process. From each sample, two filter replicates were produced and stored

immediately at -20 °C until DNA extraction. Filtration volume (**Tab. S1**) depended on the turbidity

of water samples. To avoid contamination between samples, silicon tubing was rinsed with 70%

ethanol, followed by nanopure water.

2.4 DNA Extraction

DNA was extracted from both filter replicates using the DNeasy PowerWater Kit (Qiagen, Germany)

following the manufacturer's protocol. To optimize yield, the elution volume was reduced to 65 µL

per replicate, and the extracts were subsequently pooled, resulting in a final DNA sample volume of

130 µL. DNA concentration was quantified using the Qubit 4 Fluorometer (Invitrogen, USA), and

purity was assessed with the NanoDrop One spectrophotometer (Thermo Fisher Scientific, USA) (see

Tab. S2). Blank extractions confirmed that filters and reagents were free of DNA contamination.

2.5 Quantitative PCR (qPCR)

The abundance of a total of fourteen ARGs as indicators for anthropogenic AMR input (*aadA,*

aph(3')-lb, *bla$_{CTX-M-1}$, bla$_{TEM}$*, *ermB*, *ermF*, *mcr-1*, *mecA*, *sul1*, *sul2*, *tetA*, *tetM*, *tetW*, *qnrA)* was

determined by qPCR. In addition, qPCR was performed for three genetic markers (GM): class 1

integron integrase gene *intI1* as a marker for mobile (multi-)resistance cassettes; the beta-

glucuronidase gene *uidA* as an *E. coli* specific target to assess fecal contamination, and bacterial

16S rRNA genes (*16S*) as a proxy for bacterial abundance[43,44]. qPCR assays were performed on a

LightCycler®480 (Roche, Switzerland). Standard dilutions (5 x 10^7 to 50 copies/reaction) and

negative controls (*no-template*-control: AE buffer; PCR-control: only PCR reagents without the

addition of sample) were run in technical quintuplicates, while 1:10 and 1:100 dilutions of each

pooled extract and extraction blanks were analyzed in triplicate. Plasmid and synthetic DNA

standards were used for calibration. Primer sequences, primer concentration, amount of DNA, and

cycling conditions are provided in **S1.3** and **Tab. S3 – S7**. Information on quantification standards

is provided in **Tab. S8-S9**. Gene abundances were obtained by normalizing to sample volume,

logarithmized (base 10) and reported in units of log copies/L. Calculations for LOD, LOQ, qPCR

efficiency, and PCR data analysis, as well as the relative abundance of anthropogenic AMR

indicators normalized to *16S* followed the methodology outlined by Rathinavelu et al. and

Czekalski et al.[45,46]. qPCR assays met qualitative standards for linearity ($R^2 > 0.99$) and efficiency

(90-110%), with the exception of the *tetA* gene ($R^2 = 0.981$) (see **Tab. S7**).

For each WWTP location, we calculated AMR contamination potential ($F_{EFF/US}$) and observed

WWTP impact ($F_{DS/US}$) as log-transformed ratios of gene abundance (C) in effluent (C_{EFF}) or

downstream (C_{DS}) samples, respectively, to the upstream river samples (C_{US}) (**Eq. 1, 2**).

$$F_{EFF/US} = \log_{10}(\frac{C_{EFF}}{C_{US}}) \hspace{4cm} (Eq.1)$$

$$F_{DS/US} = \log_{10}(\frac{C_{DS}}{C_{US}}) \hspace{4cm} (Eq. 2)$$

2.5 *16S rRNA* Gene Amplicon Sequencing

Microbial community composition was assessed by *16S rRNA* gene amplicon sequencing of the V3-

V4 region (**S1.4**), performed with PCR-cDNA Barcoding Kit SQK-PCB111.24 on GridION platform

using MinION flow cells R9.4.1 (Oxford Nanopore Technologies, UK). Sequencing depth was a

minimum of 30,000 reads per sample. Quality control included re-sequencing of samples that showed

matrix inhibition. Resulting FASTQ data were matched against the SILVA database (Release 138,

Ref Nr. 99) using EMU[47]. Non-bacterial sequences were removed prior to data analysis.

2.6 Metal, Chemical, and Nutrient Parameters

Total nitrogen (TN) and total organic carbon (TOC) were determined following ISO standards (TN:

ISO 11905-1:1998, TOC: ISO 1484:1979) using a Skalar San++ system (Procon AG, Switzerland)

and a Shimadzu TOC-L analyzer (Japan). Anions (fluoride, chloride, sulfate) were determined by

anion chromatography (930 Compact IC Flex, Metrohm, Switzerland). Dissolved trace metals

(Copper (Cu), Zinc (Zn) and Gadolinium (Gd)) and Phosphorus (as total dissolved Phosphorus (TDP)

and total Phosphorus (TP) in filtered and unfiltered samples respectively) were quantified by Agilent

8900 ICP-MS equipped with a PrepFAST autosampler, following a protocol adapted from Janssen et

al. (2024)[48]. Raw data were processed with MassHunter 5.1 (Agilent). Measurements were conducted

in four batches (Sept. 2024 – Apr. 2025) with procedural blanks and a certified reference material

(SLRS-6) in each batch (**Tab. S10**). Additional details are provided in **S1.5**.

2.7 Hydrological Data and Total ARG Load

To evaluate discharge dynamics during the campaign weeks from source to estuary, discharge data

from nine gauging stations (**Tab. S11**) along the Aare River were obtained from the Swiss Federal

Office for the Environment[49]. These data were used to calculate mean daily discharge for each gauge

(**Tab. S11**) and then to model sampling point discharge by interpolating between gauging stations

and accounting for discharge of major tributaries (**Tab. S11, Fig. S2**). We then calculated ARG loads

(L) for each sampling point in the Aare River and each ARG by multiplying discharge values (Q_i) at

each sampling location (i) with ARG abundances according to **Eq. 3.**

$$L_{ARG}\,[copies/sec] = Q_i\,[mL/sec]\,\times C_{i,ARG}\,\,[copies/mL] \tag{Eq. 3}$$

Additional discharge and load calculations were conducted for sampling locations in four major

tributaries (Saane (rkm 134.8), Emme (rkm 196.5), Reuss (rkm 272.6), and Limmat (rkm 274.1))

using data from the gauging station closest to the confluence, and for WWTP effluent discharge using

data provided by the WWTP (**Tab. S10, S11**). Predicted ARG gene concentrations downstream of

WWTP (C_{DS}) were calculated by applying a discharge-based mixing model using measured WWTP

(Q_{WWTP}) and modeled river discharge data ($Q_{US\text{-}River}$) and measured ARG concentrations in US ($C_{US\text{-}}$

$_{River}$) and EFF (C_{WWTP}) (**Eq. 4**).

$$C_{DS} = (Q_{US\text{-}River} \times C_{US\text{-}River} + Q_{WWTP} \times C_{WWTP}) \, / \, (Q_{River} + Q_{WWTP}). \tag{Eq. 4}$$

2.8 Data Processing and Visualization

Data handling, statistical analyses and graphics were generated in R[50] using R Studio[51] and the

packages ggplot2, cowplot, stringr, scales, grid, readr, knitr, dplyr and tidyr. Species richness and

Shannon's alpha diversity index were calculated using the vegan[52] package and differences among

groups were tested using the Kruskal-Wallis' test provided in base R. Non-metric multidimensional

scaling (NMDS) was performed on *16S rRNA* abundance data with Bray-Curtis distance as the

dissimilarity measure (k = 4). Differences in microbial community composition between different

waterbodies were tested using PERMANOVA. Spearman rank correlations were performed to assess

relationships between microbiological, chemical, nutrient parameters, and concentrations of trace

metals. QGIS Maidenhead[40] was used for geographical illustration and map design. Adobe Photoshop

was used to fine-tune graphical illustrations.

3. Results

3.1 Background Characterisation of Anthropogenic Influence on the Aare River

Nutrient and metal concentrations show the general anthropogenic impact of urban and agricultural activity along the Aare River. Both TDP (total dissolved phosphorus), representing the bioavailable, short-term reactive fraction, and TP (total phosphorus), reflecting total P-load, remained < 10 µg/L upstream of rkm 118, but concentrations increased further downstream (**Fig. 2c**). WWTP effluents emerged as major point sources, with TP >100 µg/L (except WWTP Thun, 60 µg/L), while tributary P concentrations were generally similar to the Aare River at confluence.

Three trace metals were compared to further characterize anthropogenic impacts: Copper (Cu) and Zinc (Zn) as indicators of diffuse urban, agricultural, and industrial activity[53,54] and gadolinium (Gd) as a specific tracer of healthcare-related inputs[55], reflecting its use as contrast agent in magnetic resonance imaging (MRI). Copper concentrations gradually increased from 0.22 µg/L at the source to 1.07 µg/L at the confluence with the Rhine. WWTPs were point sources with up to 2.6 µg/L Cu (**Fig. 2b**). Zn followed a similar pattern (**Fig. S3**), but in contrast to Cu also increased in the downstream section between Solothurn and Aarau, suggestive of non-point sources such as road runoff and industrial emissions[53]. Gd concentrations showed a similar trend, but with more pronounced increases downstream of WWTPs receiving hospital effluent, notably Interlaken (0.0048 µg/L), Bern (0.04 µg/L), and Aarau (0.03 µg/L). WWTP effluent Gd concentrations were > 0.1 µg/L, $10^1 - 10^2$ fold higher than river background. Furthermore, elevated Gd/Cu ratios, especially in Thun, Bern, Biel and Solothurn WWTP effluents, reflect the specific point source input of Gd through WWTP of urban regions with healthcare centres relative to the more varied Cu sources, confirming Gd as a sensitive tracer of healthcare-related discharges (**Fig. S4**).

Figure 2: Dynamics of nutrients and metals along the Aare River. Left-side panels show nutrient
dynamics: (a) total organic carbon (TOC; black) and total nitrogen (TN; blue), and (c) total dissolved
phosphorus (TDP; black) and total phosphorus (TP; blue). Right-sided panels display (b) copper
concentrations and (d) gadolinium concentrations. In all panels, values along the river course are
shown as continuous lines, while values from WWTP effluents (triangles on grey bars) and
tributaries (squares on blue bars) represent input values. Flow-through lakes are indicated as black
bars on the x-axis: 1) Brienzersee, 2) Thunersee, 3) Wohlensee, and 4) Bielersee.

3.2 Trends in Microbial Community Abundance and Composition

The observed trends in nutrient and metal distribution are consistent with microbial indicators such
as TCC and high HNA fraction, confirming elevated microbial abundance at nutrient-enriched sites
(**Fig. 3a, Fig. S5**). TCC values steadily increased roughly ten-fold from the source to the Rhine,
ranging from 1.7 x 10^5 to 1.7 x 10^6 cells/mL. Tributaries draining urban areas (e.g., Saane, Reuss,
Limmat) showed elevated TCCs, whereas those from more pristine regions (e.g., Emme, Kander)
were lower than the Aare at their confluences. *16S rRNA* gene sequencing further revealed clear

spatial patterns in microbial diversity and composition (**Fig. 3b, 3c**). Beta diversity analysis using NMDS on Bray-Curtis dissimilarities demonstrated structuring by spatial proximity, with headwater (rkm 0-35) and upper Aare (rkm 35-100) communities being significantly different from more tightly clustered downstream communities (rkm >100) (PERMANOVA: $R^2 = 0.323$, $F = 1.813$, $p = 0.001$, **Fig. 3b**). Tributaries fell into the cluster of the river stretch they entered, except Zihl channel, while composition of WWTP effluent communities was clearly differentiated.

These trends in community composition are clearly visible in the phylum level taxon distribution (**Fig. 3c**). Localized shifts in the community were noted e.g., downstream of Thunersee (increase in *Patescibacteria*), around Bielersee (increased *Cyanobacteria*) and in metropolitan areas downstream of WWTPs such as Bern and Aarau, where a notable rise in the abundance of *Firmicutes* and *Campylobacterota* was observed. In WWTP effluents, *Proteobacteria* were consistently dominating, and together with *Bacteroidota* comprising up to 70% of the sequences. *Patescibacteria* were particularly abundant (> 25%) in the WWTP effluents of Thun and Biel and *Firmicutes* were enriched (> 10%) in effluents of WWTPs Interlaken, Thun, Bern, and Biel. In the Aare and its tributaries, considerable variability was observed.

Figure 3: Composition and dynamics of the bacterial community along the Aare River. (a) Total
bacterial cell count (TCC) along the river course (black line), including effluents from WWTPs along
Aare River (triangles on grey bars), and tributaries (square on blue bars). (b) NMDS ordination
showing sample distribution along the Aare River based on community composition, highlighting
three aggregates (black circles): i) WWTP effluents, ii) upper reach samples, and iii) lower reach
samples. (c) Relative abundance of the 12 most frequent bacterial phyla along the Aare River, its
tributaries (blue bars), and WWTP effluents (black bars); remaining phyla are grouped as "others".
Flow-through lakes are indicated as black bars on the x-axis: 1) Brienzersee, 2) Thunersee, 3)
Wohlensee, and 4) Bielersee.

3.3 Unique ARG Distribution Patterns Along the Aare River

Out of 14 investigated ARGs, 11 (*aadA*, *aph(3')-lb*, *bla*CTX-M-1, *bla*TEM, *ermB*, *ermF*, *sul1*, *sul2*, *tetA*, *tetM* and *tetW*) were detected in the Aare River (**Fig. 4a**). *qnrA*, *mecA*, and *mcr-1*, which confer resistance to quinolones, methicillin, and colistin, respectively, were not detected in any of the Aare River samples, but *qnrA* was found in two WWTP effluents (**Fig. 4d**). Some ARGs, e.g. those conferring resistance to aminoglycosides (*aadA*), erythromycin (*ermB*), sulfonamides (*sul1*), and tetracycline (*tetW*) were frequently detected and typically the most abundant (**Fig. 4a, Tab. S1**). In contrast, β-lactamase resistance genes (*bla*CTX-M-1, *bla*TEM) and the *E. coli* marker gene *uidA* were restricted to low-level detections ($< 10^2$ copies/mL) at hotspots with high overall ARG concentration (rkm 124 and 250).

At the source of Aare River and at rkm 30, ARGs remained undetected, while *16S* was present at ~1.9 x 10^3 copies/mL (**Fig. 4a**). From rkm 45 onwards, *16S* copies increased, consistently exceeding 1 million copies/mL close to the confluence. ARGs and the integron integrase gene *intI1* were first detected at the outflow of Brienzersee near Interlaken, although at relatively low concentrations ($10^1 – 10^3$ copies/mL) and persisted downstream. *intI1* was highly prevalent along the entire downstream river (up to 10^4 copies/mL) and, with few exceptions, the most abundant AMR indicator. Both absolute (Fig. 4a) and relative ARG abundances (**Fig. S6a**) as well as their total load (i.e. fluxes; **Fig. S7**) increased overall along the river's course. The increase was not continuous, but interrupted by marked local declines in ARG abundance at rkms 75, 130, 165 and marked peaks or increases at rkms 55, 124, 170, 206 until reaching a maximum between rkm 250 to 272 (**Fig. 4a**) with ARG diversity broadening in the lower reaches (detections of *bla*CTX-M-1, *bla*TEM).

Figure 2: Absolute abundance (log gene copies/mL) of ARGs and GMs, quantified using qPCR.
The concentration trends in the Aare River (a) are shown as line plots, while the abundance
measured in (c) tributary inflows and (d) WWTP effluents is depicted as symbols within the bars
(tributaries: blue / WWTP: gray). Flow-through lakes are indicated as bars in (b). The values are
logarithmic - 1 corresponds to 10, and 2 corresponds to 100 copies/mL.

Total load calculations showed that the flux of the *16S* marker gene increased almost 70-fold (5.84
x 10^{12} to 4.06 x 10^{14} copies/sec) while *intI1* increased 1'090-fold (1.63 x 10^9 to 1.78 x 10^{12}
copies/sec) between rkm 51 (ARG detected) and rkm 288 near the confluence. *ermB* increased
7'120-fold (8.1 x 10^7 to 5.77 x 10^{11} copies/sec), with similar trends for other widespread ARGs,
indicating the river's role in downstream ARG transport (**Fig. S7**, **Tab. S12**).

3.4 ARG Patterns of Aare River Shaped by Lakes, Large Tributaries, and WWTPs

Lakes – When ARGs were present upstream of lakes, a pattern of "resetting" ARG abundance was
observed when the river passed through a lake (**Fig. 4b**). Thunersee eliminated all ARGs, which had
reached a first peak downstream of the city of Interlaken, to levels below detection, with the

exception of *tetA*. In contrast, Wohlensee and Bielersee caused a drop in ARG abundance but did

not eliminate previously detected ARGs after lake passage, except for *tetM* and *sul2*, which were

undetectable for ~7 rkm downstream of the Bielersee outflow. Although no measurements upstream

of Lake Neuchâtel were performed, the low ARG abundance in the Zihl canal connecting Lac de

Neuchâtel and Bielersee (**Fig. 4c**) is also consistent with this pattern**.**

Tributaries – Across all sampled tributaries, the same ARGs (*aadA*, *aph(3')-lb*, bla$_{CTX-M-1}$, *bla*$_{TEM}$,

ermB, *ermF*, *sul1*, *sul2*, *tetA*, *tetM,* and *tetW*) and GMs (*16S*, *intI1,* and *uidA*) were detected as in the

Aare River (**Fig. 4c**). Meanwhile, ARGs *qnrA*, *mecA*, and *mcr-1* remained undetected in all

tributaries. A clear positive association of absolute ARG abundance with population density was

observed within the catchments, showing tributaries draining larger and more populated areas

exhibit higher ARG abundances (Spearman $\rho = 0.89$, $p = 0.033$; **Fig. S8**). Among the tributaries

with smaller catchment areas ($< 2'000$ km^2), the Kander exhibited consistently low concentrations

of all markers (< 100 copies/mL), attributed to low population density. In comparison, Saane (1'893

km^2) and Emme (976 km^2) rivers, draining more populous regions, showed elevated levels of

ARGs, reaching up to 1'000 copies/mL The Limmat River (2'412 km^2) draining the Zurich

metropolitan region showed an even higher abundance of ARGs, including *tetA*, which was

detected at even higher levels than in the Reuss river (3'426 km^2). The Reuss drains major urban

regions in the Cantons Lucerne and Zug, and large parts of central Switzerland, and is the tributary

with the largest catchment and overall highest ARG load. At the confluences with Reuss and

Limmat, Aare River displayed higher concentrations of both ARGs and GMs compared to the

inflowing tributaries, which resulted in a dilution effect downstream of the confluence (**Fig. 4c**).

Wastewater – WWTP effluents consistently contained high ARG abundances exceeding those in river water (**Fig. 4d**). In addition to *uidA* and *intI1*, 9 out of 14 ARGs were detected across all effluents, including genes conferring resistance to aminoglycosides (*aadA*, *aph(3')-lb*, sulfanomides (*sul1*, *sul2*), tetracyclines (*tetA*, *tetM*, *tetW*), and macrolides (*ermB*, *ermF*). bla_{TEM} was absent only in WWTP Thun, while $bla_{CTX-M-1}$ was found only in effluents from WWTPs Interlaken, Bern, Biel, and Aarau. *qnrA* appeared at low levels (50 – 180 copies/mL) in WWTP Interlaken and Biel. *mcr-1* and *mecA* were not detected.

The observed contamination downstream of WWTP (log DS/US ARG abundance ratio; $F_{EFF/US}$, **Fig. 5a**) mirrored the WWTP contamination potential (log EFF/US ARG abundance ratio, $F_{DS/US}$, **Fig. 5b**). Inflows of WWTP effluents, with on average 70-fold (geometric mean; average $F_{DS/US}$: 1.9) higher ARG concentrations than upstream river water ($F_{DS/US}$ range: -0.2 to 4.2, **Fig. 5b**) led to 4-fold (0.6 log) downstream increases of ARGs in the receiving river ($F_{DS/US}$ range: -0.4 to 2.1, **Fig. 5a**). Negative $F_{DS/US}$, indicating ARG abundance at or below background levels, were only observed for the effluents of the pre-clean stage of the slaughterhouse at Lyss and the smallest WWTP, Baden.

The contamination potential (log EFF/US ratio, $F_{EFF/US}$) of the effluent was similar for all genes, the Kruskal-Wallis test (p = 0.489) and pairwise Wilcoxon comparisons found no statistically significant difference between genes (**Fig. 5b**). The large WWTP in Bern (design capacity > 300 kPE), which also receives WW from the Inselspital (University Hospital) had the highest ARG contamination potential with a median $F_{EFF/US}$ of 3.1. The small WWTP of Lyss (pre-clean, <30 kPE, $F_{EFF/US}$ =0.53) and Baden (<50 kPE, $F_{EFF/US}$ =0.71) also had the lowest ARG contamination potential. However, no significant relationship (R^2 = 0.14, p = 0.29) was observed between ARG contamination potential and design capacity overall (**Fig. S9**).

Comparing $F_{DS/US}$, Kruskal-Wallis test (p = 0.223) and Wilcoxon pairwise tests again showed no

significant difference between genes. WWTP impact varied by site (**Fig. 5a**). The strongest median

contamination impact (1.66) was observed at WWTP Bern (> 300 kPE). The increase was

particularly high for *aadA* (141-fold), *aph(3')-Ib* (75-fold), *tetW* (132-fold), *tetM* (68-fold), and

ermF (61-fold). No impact (i.e. negative $F_{DS/US}$) was observed more frequently (n=14) than for

contamination potential, mainly at WWTP Baden, Oensingen and Biel.

Applying a discharge-based mixing model, the expected downstream ARG concentration matched

observations closely (R^2 = 0.88, p = 0.005, m=0.80, **Fig. S10**), confirming that the observed

increases are due to the WWTP effluent. The impact of WWTP of urban centers thus consistently

resulted in a marked increase in ARG concentrations, loading, and relative abundance in the Aare

River. This is most pronounced downstream of WWTP Interlaken and Bern (**Fig. 4a, d**) but is also

contributing to the cumulative increases downstream of Biel (rkm 164-170), Solothurn (rkm 196-

206), and up to Aarau (rkm 206-272).

Figure 5: Observed contamination (DS to US ratio) and contamination potential (EFF/US ratio) of the absolute abundance of ARGs and GMs with increasing design capacity (PE) of WWTPs. For evaluation of contamination potential and the impact of WWTP effluent on ARG and GM abundance in receiving river, log10-transformed ratios of ARG abundance were calculated for (a) downstream relative to upstream, and (b) WWTP effluent relative to the upstream river. Results were ordered by the design capacity (population equivalent, PE) of WWTPs. Three WWTPs located on tributaries of the Aare river were also included: Baden (Limmat), Lyss (Alte Aare), and Oensingen (Dünnern). The latter two receive wastewater from slaughterhouses in addition to communal wastewater.

3.6 Correlation between ARG and other indicators of anthropogenic impact

To identify factors associated with variations in the abundance of ARGs and microbial marker genes (*intI1*, *16S* and *uidA*), we calculated Spearman rank correlations with microbiological and chemical parameters and concentrations of nutrient and trace metals. Correlation analyses were performed separately on samples collected i) along the Aare River (**Fig. 6**) and ii) on effluents from 10 WWTPs along the Aare River and its tributaries (**Fig. S11**). To capture spatial trends along the river continuum, "km from source" (distance) was included as a parameter.

In river water samples, we observed mostly positive correlations between parameters, with the exception of fluoride, sulfate, pH, DOC, and LNA abundance (**Fig. 6**). Distance emerged as a strong predictor of most parameters, including ARG abundance. Strong positive correlations ($\rho > 0.7$) were observed between distance and the abundance of *aph(3')-lb*, *ermB*, *intI1*, *sul1*, *sul2*, *tetA*, and *tetW* but also TP, TDP, chloride, nitrate, and TN ($\rho > 0.8$). Trace metals, particularly Cu, also increased with distance.

Across the dataset, most ARGs showed highest correlation coefficients with one another ($\rho > 0.85$), except for β-lactamase genes (*bla*$_{CTX-M-1}$ and *bla*$_{TEM)}$, which displayed only moderate correlations with other ARGs ($\rho = 0.4$–0.7) but co-occurred strongly with each other ($\rho > 0.8$). *16S, TCC* and *uidA* abundances correlated with ARG occurrence, supporting their roles as indicators of total bacterial load and faecal contamination. Notably, Gd correlated highly with nearly all ARGs ($\rho > 0.65$), with the rare β-lactamases being the only exception.

Figure 6: Heatmap of Spearman correlation matrix including quantified genes, physicochemical parameters, and biological indicators across all Aare River samples. Red: indicates strong positive correlation (ρ = 1), white: indicates no correlation (ρ = 0), blue: strong negative correlation, as one variable increases, the other decreases (ρ= -1). Black boxes include biological parameters, including FCM-derived cell counts as well as ARGs and GMs.

Correlation analysis of WWTP effluent samples revealed very strong correlations among most ARGs, including *bla*$_{CTX-M-1}$ and *bla*$_{TEM}$, while *tetA* consistently exhibited weak correlations (**Fig. S11**). ARGs correlated mostly positively with nutrients, metals, and bacterial abundance measures, but less closely (mostly ρ < 0.5) compared to river water samples. Distance was not a strong or consistent predictor for ARGs or other parameters in the WWTP effluent dataset.

4. Discussion

4.1 WWTPs are key contributors to the ARG content of the Aare river

Overall, observations on the River Aare were consistent with urban sources and WWTP being the

key contributors to ARG accumulation. For all WWTP sites studied, effluent was shown to be a

source of ARG relative to the receiving river water (**Fig. 5a, b**), which sufficiently explains the

downstream ARG increase observed near major urban centres (**Fig. S10**). This remains the case even

in the lower reaches of the Aare, e.g. at Aarau, where upstream contamination has accumulated

considerably. Accumulation of ARG also clearly exceeded the total bacteria (2.5 median DS/US

enrichment of *16S*) showing that ARG contamination is a specific effect distinct from bacterial cell

inputs.

Along the river continuum, two main hotspots of locally increasing ARG content at Bern and

Interlaken highlighted how source and sink processes can drastically affect downstream dynamics. In

both cases, sharp peaks in ARG absolute and relative abundance downstream of the WWTP were

short-lived, as the Aare entered a lake just after each city (**Fig. 4b, d, Fig. S6b, d**). The most

pronounced effect was observed in Bern, which was also the largest WWTP sampled. On an average,

effluent showed a 3.1 log higher ARG content compared to upstream water, resulting in the highest

downstream increase (1.6 log). WWTP Interlaken, despite operating at only one-fifth the capacity of

Bern, showed an even higher mean effluent ARG content relative to upstream water (3.2 log). While

partly explained by the comparatively pristine upstream water, the fact that Interlaken is a hotspot for

international tourism may have contributed to the much higher absolute abundances of ARG in

Interlaken compared to e.g. Thun (**Fig. 4d, 5b**). According to the local tourist association, more than

3.5 million overnight stays per year add substantially to the local population

(https://www.interlaken.ch/info-service/ueber-interlaken-tourismus/berichte-

statistiken/jahresberichte), with WW from visitors, residents, and the regional hospital all processed

by the WWTP.

Our results support earlier observations on the importance of WWTP discharge for ARG levels in

smaller Swiss rivers with proportionally high wastewater impact[24], and demonstrate this also for a

major river system along its full length.

4.2 ARG as a component of anthropogenic impact – common sources and selection

Other markers of human activity such as P and Cu (with sources in agriculture and urban wastewater)

increased in parallel with the AMR markers (**Fig. 2, 4**). The increase in Gd (MRI contrast agent)

provided an indication of healthcare activity impacts on the river, and its strong correlation with

ARGs indicates an additional contribution of large hospitals to the ARG input from major urban

centres (**Fig. 2, 6**). The overall strong correlations of ARG concentrations in river water with other

indicators of anthropogenic impact (**Fig. 6**) highlight ARGs as additional markers of anthropogenic

contamination, reflecting the cumulative influence of sustained inputs from human activity along the

river continuum.

While common sources of ARGs, metals, and nutrients primarily from wastewater discharge may

largely explain the observed correlations, more specific interactions such as selective effects of

chemical wastewater contamination should be considered[56–58]. Downstream of WWTPs, elevated

concentrations TDP closely mirrored ARG abundance. We observed a close correlation between

abundant ARG and dissolved phosphorus concentrations (TDP, $\rho > 0.7$, median 0.79), contrasting

with a weaker relationship of TDP with general bacterial abundance (*16S:* $\rho = 0.53$, TCC: $\rho = 0.43$).

A weaker relationship was observed between ARG and TN ($\rho > 0.65$, median 0.70) or TP ($\rho > 0.5$,

median 0.64). While P has previously been used as a proxy for bacterial growth[59], it is not considered

a selective agent for AMR. However, TDP increase was consistent with the observed enrichment of

eutrophic phyla (*Proteobacteria*, *Bacteriodota*, *Firmicutes*, *Campylobacterota*[60]), which include

clinically relevant genera at sites with high ARG prevalence. Such eutrophic phyla may more frequently be carriers of ARGs compared to oligotrophic phyla such as *Actinobacteriota*, *Planctomycetota*, and *Chloroflexi*[60] which in turn declined under high P loads. The strong co-occurrence of ARGs and P likely reflects shared anthropogenic sources, such as wastewater and additional diffuse agricultural inputs, and nutrient-induced alterations of the microbial community, rather than a direct selective effect. Along the river (rkm 30 – 195), TP and ARG levels in agriculturally impacted areas show neither a WWTP-independent increase nor a decline, suggesting persistence of AMR markers or background inputs from diffuse sources (**Fig. 2a**).

Metals such as Cu and Zn may act as potential agents of selective pressure[61] which could explain observed correlations with ARG (**Fig. 6**, **Fig. S9**). However, the measured Cu and Zn concentrations in both effluents and river water (**Fig. 2b**, **Fig. S3, Table S1**) remained below experimentally established thresholds required for co-selection of ARGs[62]. Nonetheless, based on reports of higher levels of Cu and other metals in influent and primary clarification, selection in the upstream wastewater stream cannot be excluded[63,64]. In effluents, despite low concentrations (< 10 µg/L), Cu and Zn correlated positively with ARGs (ρ ~0.7 and $\rho \geq 0.6$, respectively), with the strongest associations observed for *sul2* (Cu: $\rho > 0.75$), *aph(3')-Ib* and *ermF* (Zn: both $\rho = 0.71$). Correlations with *E. coli* marker *uidA* and bacterial marker *16S* were ~0.1 lower, indicating stronger linkage of Cu to ARG-carrying subpopulations than to the overall microbial community. These findings align with Miao et al.,[59] who also reported strong positive associations between heavy metals (Cu, Cr, Mn) and ARG prevalence, as have several similar studies[61,65,66].

4.3 ARG contamination of the Aare compared to other European rivers

Can we expect a continuous accumulation of AMR along the course of major rivers? Our analysis of

ARG abundance in Aare River showed their absence near the source, a consistent increase towards

the middle and lower reaches, with pronounced regional hotspots. In contrast, while Danube

headwaters were likewise less impacted, no correlation between ARGs and distance (rkm) was

observed[36]. The Rhine likewise did not exhibit a general increase with distance; instead, regionally

restricted impacts were observed, particularly in the Hochrhein-Oberrhein section, attributed to the

river's size and buffering capacity against localized anthropogenic inputs[33]. These patterns indicate

that AMR contamination does not accumulate uniformly along rivers. Instead, source and sink

dynamics can counterbalance each other, especially in larger river systems.

By comparing selected marker genes that were also measured in the Rhine and Danube studies, the

AMR contamination in the three river systems can be compared. Even at the hotspot identified

downstream of Bern (rkm 124), ARG relative abundance in the Aare (*intI1*: 1.7×10^{-2}, *sul1*: 3×10^{-3}

copies/16S) remains below the highest values reported in the Rhine (*intI1*: 6.9×10^{0}, *sul1*: 1.07×10^{-2}

copies/16S)[33]. Although different β-lactamase genes were assessed in the Rhine, the relative

abundance at the Aare hotspot (*bla*$_{TEM}$: 5.7×10^{-5}) was comparable to *bla*$_{SHV}$ levels reported in the

Rhine (6×10^{-5} , [33]). In addition, β-lactamases in the Rhine occurred sporadically and were closely

associated with WW inputs[33], aligning with our observations for *bla*$_{CTX-M-1}$.

Comparing absolute abundances, *intI1* ($4.22 - 6.69$ $\log_{10}$ copies/100 mL) and *sul1* ($3.55 - 6.5$ $\log_{10}$

copies/100 mL) were consistently detected at high levels across all Danube samples[32]. With the

exception of the more pristine Aare headwaters, where these genes were below detection, we

observed a comparable range (*intI1*: $3.67 - 6.03$ $\log_{10}$ copies/100mL; sul1: $2.89 - 5.28$ $\log_{10}$

copies/100mL) in the Aare. This comparison suggests that AMR contamination of the mid and lower

reaches of the Aare is similar or slightly lower compared to larger European rivers, underscoring the substantial anthropogenic influence on the Aare.

In addition, load calculations showed that the Aare river acts as a major conveyor of accumulating resistance genes, with only lake passages interrupting (**Fig. S7**). This further emphasizes the role of the Aare in ARG dissemination within and beyond Switzerland's borders, as 8.1×10^9 to 5.8×10^{11} copies/sec of the measured ARG are discharged into the Rhine river, which enters Germany roughly 50 km downstream of the confluence.

4.4 ARG Dynamics in a complex watershed

Tributaries exerted a moderate but notable effect on the ARG content of the Aare river. ARG content of Aare tributaries largely reflect the size of their catchments and associated human populations. The Emme, despite elevated ARG levels, contributed only modestly due to its relatively low discharge; the subsequent rise in ARGs downstream could be more likely attributed to effluent from WWTP Solothurn. In contrast, the Reuss and Limmat, both with large catchments and daily discharges >100 m^3/sec (each almost 50% of the Aare discharge before their confluence) exerted a much stronger hydrological and microbial influence. Their combined effect was dilution of ARG levels (**Fig. 4**), while maintaining a high total ARG load (**Fig. S7**), as Aare's discharge nearly doubled downstream of these confluences (from 270 to 515 m^3/sec, **Fig. S2**). These observations underscore the critical role of absolute discharge volume on ARG distribution and dissemination.

Lakes, in contrast, behaved consistently as sinks, producing a pronounced "reset effect" on riverine ARG abundance, documented here for the first time. After passage through lakes, ARG concentrations decreased by up to 1.5 log units, with certain genes becoming undetectable in the lake

outflow. These reductions appear further enhanced by high water residence time, with near-complete

removal in Thunersee (648 days[67]) compared to partial reduction in Bielersee (58 days[67]), despite

similar upstream ARG concentrations (**Fig. 4**). Notably, Wohlensee, a reservoir with a retention time

of only 2.1 days[68], substantially attenuated the high ARG concentrations entering from downstream

of Bern. Earlier work by Czekalski et al. linked WWTP capacity in the catchment with ARG content

in Swiss lakes[69], showing that Swiss lakes are not insulated from upstream ARG inputs. The present

study adds that lakes are simultaneously major ARG sinks. In large lakes with long water residence

time, this likely reflects gradual replacement of the river microbiome by lake-adapted communities

over time scales of months to years. The rapid response observed in Wohlensee suggests additional

sink mechanisms acting on shorter timescale (days) such as sedimentation (of particle-bound cells or

cell aggregates), predation, or UV-mediated decay of resistant communities under reduced flow

conditions.

Conclusion

Integration of genetic AMR markers with microbial and chemical parameters along the full course of the Aare River enabled us to capture spatial dynamics, identify hotspots and sinks, and link resistance patterns to anthropogenic inputs. Our findings demonstrate that ARG levels are shaped by WW discharges and cumulative anthropogenic pressures along the entire river, positioning the Aare as a critical interface for potential AMR transmission. Our results further highlight the dynamic impact of lakes and tributaries on the main river. While tributaries can influence ARG levels through both inputs and dilution, lakes function as effective sinks that substantially reduce ARG concentrations. The behaviour observed in Wohlensee highlights how in-lake ARG dynamics remain poorly understood and require further investigation.

We also demonstrate that reporting absolute abundance (concentration), relative abundance, and total load (fluxes) of ARGs are each important parameters to assess the AMR situation of a river system and should be reported in such studies to improve comparability. This comprehensive dataset establishes a valuable baseline not only for Switzerland but also for comparable river systems, emphasizing the need for targeted management interventions. In the context of ongoing and planned nationwide upgrades in WW treatment for organic micropollutant removal by 2040, the effectiveness of these measures should be evaluated also with respect to ARG reduction in receiving waters. Moving forward, studies that explicitly address transmission potential and risk assessment are essential. Collectively, these findings highlight that rivers function not only as passive recipients but also as active conduits of AMR, with implications that extend beyond national borders.

Acknowledgments

We thank the staff at the WWTPs in Interlaken, Thun, Bern, Biel, Lyss, Centravo AG at Lyss,

Solothurn, Aarau, Oensingen and Baden for allowing and supporting effluent sampling at their

facilities. We thank Antonia Zuber and Urs Lergster for the technical assistance with DNA extractions

and qPCR (both Eawag, Switzerland). We thank Patrick Kathriner for performing TN, and TOC

measurements, Marcel Walter for performing ICP-MS measurements, at the Aquatic Geochemistry

lab at Eawag in Kastanienbaum and AuA-Lab (Analytik- und Ausbildungslabor) at Eawag Dübendorf

for measuring anions by ion chromatography. We thank Dr. Martin Schmid and James Runnalls (both

Eawag, Switzerland) for assisting with access to additional metadata from MetoSchweiz/Idaweb and

Federal Office for the Environment (Bundesamt für Umwelt; BAFU) hydrological data, respectively.

We thank Prof. Damien Bouffard (Eawag, Kastanienbaum, Switzerland) and Dr. Massimiliano Zappa

(WSL, Switzerland) for valuable feedback and input on our discharge model for the Aare river. We

thank Andreas Scheidegger (Statistics, Data Science & Modeling, Eawag, Dübendorf, Switzerland)

for consultations regarding statistical analysis and Tara Behnsen (Surface Waters- Research and

Management, Eawag, Kastanienbaum, Switzerland) and Rosi Siber (Statistics, Data Science &

Modeling, Eawag, Dübendorf, Switzerland) for consultation regarding QGIS visualization and

analysis.

Author Contributions

Conceptualization: DLW & HB (general study and all components except trace metals); DJJ (trace metal component). **Funding acquisition**: HB. **Investigation**: DLW; JA & NJA (trace metal component). **Methodology**: KB, SR, DLW, HB; DJJ (trace metal component). **Data curation**: DLW, HB. **Formal analysis**: DLW; JA & DJJ (trace metal component). **Visualization**: DLW. **Software**: DLW, SR, HB. **Writing – original draft**: DLW. **Writing – review & editing**: all authors. **Supervision**: HB, KB; DJJ (trace metal component). **Project administration**: DLW, HB. **Resources**: HB; DJJ (trace metal component).

Data Availability

Raw sequence reads of bacterial 16S rRNA gene amplicons are available from the NCBI Sequence Read Archive under BioProject PRJNA1375744. All other processed data required to reproduce the above findings are available in the manuscript and the supplementary tables.

Competing interests

The authors have no competing interests to declare.

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Supplementary Material for:

Ripples of Resistance: Unveiling Antimicrobial Resistance Dynamics

Along Switzerland's Aare River

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S1. Supplementary Material and Methods

S1.1 Description of the study area

The Aare is Switzerland's longest river, passes through the cantons Bern, Solothurn, and Aargau, draining approximately 43 % of the country (catchment size: 17'709 km^2). Originating in a largely untouched alpine region (**Fig. 1a**), fed by the Unteraar and Oberaar glaciers, the Aare River flows through a mountain valley with several villages before entering Brienzersee (29.81 km^2) and shortly after Thunersee (47.74 km^2) - two alpine lakes, shaped by the alpine climate and inflows from numerous mountain creeks, but also the cities of Interlaken and Thun. Downstream, the river traverses foothill landscapes of the Swiss Alps, transitioning into a mix of agricultural land and scattered settlements, with urban centres concentrated around Bern and Biel. Between them lies Wohlensee (3.65 km^2), a slow-flowing hydropower reservoir built in a Molasse gorge, and Bielersee (39.51 km2), one of three large Jura lakes in the agriculturally productive Seeland region, fed by the Zihl canal and Jura mountain streams. From Thun to Solothurn, the river passes mainly agricultural areas, interspersed with small towns. East of Solothurn is the confluence with the Emme, a minor tributary by discharge but with a 900 km² catchment area, over half of which is farmland, originating in the Emmental Alps. Beyond Solothurn, agricultural land becomes less dominant as settlements become denser and more industrialized, particularly in the canton Aargau, before the Aare meets two major tributaries: The Reuss (draining central Switzerland) and the Limmat (draining the northeastern Alps), join the Aare before it discharges into the Rhine near Koblenz. At the confluence, the Aare River contributes more discharge than the upper Rhine.

S1.2 Flow Cytometry (FCM)

Measurements were performed on a NovoCyte Advanteon flow cytometer (Agilent, USA). For each measurement, 50 mL aliquots were taken from the original sample and stored at 4 °C until measurement (storing up to 3 days post sample collection). To prevent nozzle clogging, visibly turbid samples were passed through a 40 µm nylon Cell Strainer (Falcon, Thermo Fisher Scientific, USA) pre-measurement. As negative control, sterile-filtered Evian water purchased from the local store was used. The gating strategies for determination of total cell counts (TCC) and the high nucleic acid (HNA) and low nucleic acid (LNA) fractions are displayed in **S2, Fig. S1**.

S1.3 Quantitative PCR (qPCR)

qPCR assays met established quality criteria for linearity ($R^2 > 0.99$) and efficiency (90 - 110%), except for tetA ($R^2 = 0.981$) (**Tab. S7**). Slopes between -3.1 and -3.6 correspond to efficiencies between 90% and 110%, which were considered acceptable (**Tab. S7**). However, only minimal effects on measurement results are expected for the diluted samples, therefore results were normally evaluated. Negative and extraction controls were either negative or showed Cp values above the lowest standard (50 copies/reaction). Values above the LOQ were not extrapolated but set to zero to ensure consistency across genes. Outliers were excluded when standard deviations of triplicate Cp exceeded 0.5, which occurred only for values >LOQ.

Corresponding primer and fluorescence probe sequences for Taq Man assays (LightCycler 480 Probes Master; Roche, Switzerland) were adapted from internal protocols and Rathinavelu et al. (2025), except for *ermB*, *bla~CTX-M-1~*, and *qnrA*, which were analyzed using SYBR Green assay (LightCycler 480 SYBR Green Master; Roche, Switzerland). As standards for the genes *bla~CTX-M-1~*, *ermB*, *sul1*, *sul2*, *tetM*, *tetW*, *qnrA*, *intI1* and *16S rRNA*, plasmid harboring cloned inserts of the target genes, each carrying the respective target indicator gene, was utilized. Synthetic gBlock DNA, published by Rathinavelu et al. (2025) [1] served as standards for the genes *aadA*, *aph(3')Ib*, *bla~TEM~*, *ermF*, *mcr1*, *mecA*, *tetA*, and *uidA*. All sequences and cycling conditions are summarized in **Tables S3-S7**.

S1.4 16S rRNA Gene Amplicon Sequencing

Amplicons were generated using the PCR cDNA barcoding kit SQK-PCB111.24 (Oxford Nanopore Technologies, UK). Undiluted DNA extracts served as templates for amplification of 16S rRNA gene (V3-V4 region) following a modified two-step PCR protocol.

PCR-1 (25 µL) contained 12.5 µL PCR mix, 9.5 µL nuclease-free water, 1 µL of each primers 341F and 806R (10 µM), and 1 µL template DNA (2 µL if concentration <10 ng/µL). Cycling conditions were 95 °C for 30 s, followed by 30 cycles of 95 °C for 15 s, 55 °C for 15 s, 65 °C for 30 s, and a final extension at 65 °C for 6 min, yielding ~460 bp amplicons. PCR-2 (25 µL) used 5 µL of PCR-1 product, 12.5 µL PCR mix, 6.75 µL nuclease-free water, and 0.75 µL barcoded primers (cDNA barcoding kit). Cycling conditions were 95 °C for 30 s, followed by 15 cycles of 95 °C for 15 s, 62 °C for 15 s, 65 °C for 40 s, and a final extension at 65 °C for 6 min, generating ~620 bp fragments. PCR products were purified with AMPure XP magnetic beads (1 µL per 1 µL PCR reaction), washed twice with 70% ethanol, and eluted in 10 µL Tris-HCl buffer (pH 8). Purified amplicons were quantified with a NanoDrop One spectrophotometer (Thermo Fisher Scientific, USA), diluted to 10 ng/µL, and pooled ($\leq$ 100 ng total DNA per flow cell) for sequencing.

 The primer pair 341f (5'-TTT CTG TTG GTG CTG ATA TTG CCC TAC GGG NGG CWG CAG-3') and 806r (5'-ACT TGC CTG TCG CTC TAT CTT CGG ACT ACH VGG GTW TCT AAT-3') targeted the V3 and V4 regions. Core primer sequences were originally published by Klindworth et al. and subsequently adapted in-house for improved coverage of wastewater microbiomes. Five samples failing initial library QC, likely due to matrix-related inhibition, were successfully resequenced after 10-fold dilution. Prior to sequencing, barcoded DNA libraries were pooled in equimolar ratios, and sequenced to a minimum depth of 30'000 reads per sample, with most libraries yielding 40'000 to 76'000 reads in accordance with the manufacturer's quality standards. Reads with a Q value <10 were rejected. Median Q values per run varied between 13 and 15 per run.

FASTQ data was analyzed using the EMU software (version 3.4.5) in the MiniConda environment (version 3) using the SILVA database (Release 138, Ref Nr. 99). Data was imported into R (version 2024.04.2) and formatted for use with the Phyloseq package (by McMurdie et al.[2]) with a custom script and analyzed in R (see **Chapter 2.8**). Prior to analysis, non-bacterial sequences (such as mitochondria, chloroplasts, and eukaryotes) as well as unassigned reads, were removed.

S1.5 Metal, Chemical, and Nutrient Parameters

For nutrient analysis, 50 mL aliquots were collected in Falcon tubes to measure nutrients (TN, TP) and dissolved anions. Additionally, 10 mL aliquots were collected in glass vials for the analysis of TOC and acidified with 32% HCl in the field. Samples were stored at 4°C. For the measurement, using a Shimadzu (Japan) TOC-L analyzer, 100 mg/L C NPOC working standard solution, 100 mg/L N TN standard solution, and control solution NPOC/TN (QC 5 ppm) were prepared according to ISO 1484:1979 and ISO 11905-1:1998. Blank vials containing nanopure water enriched with 32% HCl were used. Samples and calibration standards for TP were prepared according to ISO 15681-1:2003 followed by measurements using flow-injection analysis Skalar San ++ complete system (Procon AG, Switzerland).

For metal trace quantification, samples were dried at 100 °C, digested with 0.25 mL sub-boiled HNO_3 at 120 °C for $\geq$ 6 h, re-dried, and re-dissolved in 5 mL 0.2 M HNO_3. Aliquots were transferred into acid-cleaned 15 mL polypropylene tubes for mass spectrometry analysis using an Agilent 8900 ICP-MS/MS instrument. Helium, H_2, and O_2 gases were used in a collision cell or as reaction gasses to minimize interferences Measurements were conducted in four batches (Sept. 2024 – Apr. 2025) with procedural blanks and certified reference material (SLRS-6) see Tab. S10. Calibration curves were prepared gravimetrically, and samples and standards were corrected via internal standardization using Rh. Additional details on analytical parameters and modes are presented in Janssen et al. (2024)[3]

S2. Supplementary Figures

Figure S1: Example scatter plots of the fluorescence intensity of the particles determined by flow cytometry and the gate used for the total cell count determination (SG bacteria) and high and low nucleic acid content (HNA/LNA) of three samples. FL1-SG-H: Fluorescence intensity at the wavelength of the SYBRgreen-DNA complex. FL3-PI-H: Intensity of the nonspecific autofluorescence. Each point corresponds to a cell detection, with colors ranging from blue to red indicating accumulated signals. FCM-patterns of populations downstream of the discharge point (DS: #14) is considerably altered compared to the location upstream of the WWTP (US: #12). It shows similarities to the effluent of the treatment plant (EFF: #13).

Figure S2: Modeled mean daily discharge of the Aare river at sampling points. Gauge discharge values (orange diamonds) were obtained for the day of sampling the nearby river reaches, see **Table S12**. Values for sampling locations (blue) were obtained by constructing a discharge model using interpolation and taking into account input from major tributaries, see **Table S13**.

Figure S3: Zinc (Zn) concentrations along the Aare River. Continuous black lines represent Zn concentrations in river water along the river course. Triangles on grey bars indicate concentrations in WWTP effluents, and squares on blue bars show inputs from tributaries. Flow-through lakes are shown as black bars on the x-axis: 1) Brienzersee, 2) Thunersee, 3) Wohlensee, and 4) Bielersee. Major tributaries and corresponding WWTPs are labled on top.

Figure S4: Gd/Cu ratios along the Aare River. Continuous black lines represent the ratio of Gd/Cu concentrations in river water along the river course, as an indicator of relative impact of healthcare-related activities. Triangles on grey bars indicate ratios in WWTP effluents, and squares on blue bars show ratios in the inputs from tributaries. Flow-through lakes are shown as black bars on the x-axis: 1) Brienzersee, 2) Thunersee, 3) Wohlensee, and 4) Bielersee. Major tributaries and corresponding WWTPs are labled on top.

Figure S5: Percentage of bacterial cells with low and high nucleic acid (LNA and HNA) content along the Aare River. Fractions of LNA and HNA cells are shown along the river course (LNA: black and HNA: blue lines), in WWTP effluents (triangles on grey bars), and in tributaries (squares on blue bars). Flow-through lakes are indicated as black bars on the x-axis: 1) Brienzersee, 2) Thunersee, 3) Wohlensee, and 4) Bielersee. Raw data is provided in **Table S1**.

Figure S6*: Relative abundance (gene copies/16S rRNA gene copy) of ARGs and GMs, quantified using qPCR*. The concentration trends in the Aare River (a) are shown as line plots, while the abundance measured in (c) tributary inflows and (d) WWTP effluents is depicted as symbols within the bars (tributaries: blue / WWTP: gray). Flow-through lakes are indicated as bars in (b). Raw data is provided in **Table S1**.

Figure S7: Calculated total ARG and GM load along Aare River. Line plots show log10-transformed mean discharge rates of ARGs and GM, normalized to river discharge. Discharge data from gauging stations along the main course and tributary inflows were used as the basis for a simple interpolation model to calculate discharge at each sampling point (see **Table S12, S13**). ARG and discharge data were then used to calculate total loads (**eq. 3**). Flow-through lakes are indicated as black bars on the x-axis: 1) Brienzersee, 2) Thunersee, 3) Wohlensee, and 4) Bielersee.

Figure S8: Positive association between catchment population density and the summed up abundance of ARG and GM in Aare River tributaries. Population density refers to inhabitants per km² of each tributary. The blue line shows the fitted linear regression ($R^2 = 0.92$, $p = 0.0333$), indicating a strong correlation (Spearman $\rho = 0.886$) between the sum of ARG and GM abundances and population density in the tributaries' catchment.

Figure S9: Relationship between median enrichment of ARGs and GMs and design capacity of WWTPs (population equicvalents). Each point represents one WWTP, labeled by location. The blue line indicates linear regression fit ($R^2 = 0.14$), with corresponding Spearman's rank correlation ($\rho = 0.479$) and p-value (p = 0.29) on the plot. A positive but non-statistical significant trend suggests slightly higher enrichment of ARGs and GMs in effluent relative to the upstream river with increasing WWTP size.

Figure S10: Relationship between calculated and observed downstream ARG concentrations in the Aare River. Calculated gene concentrations (calc. C_{DS}) derived from the discharge-based mixing model using measured WWTP and river flow data [m^3/s], by calculating $C_{DS} = (Q_{US\text{-}River} * C_{US\text{-}River} + Q_{WWTP} * C_{WWTP}) / (Q_{River} + Q_{WWTP})$. Observed values (obs. C_{DS}) represent the sum of gene concentrations measured downstream of the WWTP effluent inlet. A strong positive association was found between calc. C_{DS} and obs. C_{DS} concentrations (Spearman $\rho = 0.71$, p = 0.14), and the linear regression explained most of the variance ($R^2 = 0.88$, p = 0.005, y = 77'500 + 0.804 * x). Together, these findings indicate that ARG levels downstream of WWTPs are closely linked to anthropogenic inputs from treated WW effluents.

Figure S11: Heatmap of Spearman correlation matrix, including quantified genes, physicochemical parameters, and biological indicators between all WWTP effluent samples. Red: indicates positive correlation (ρ = 1), white: indicates no correlation (ρ = 0), blue: negative correlation (ρ= -1).

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