



Large-scale biogeographical patterns of bacterial antibiotic resistome in the waterbodies of China



Lemian Liu^{a,b,1}, Jian-Qiang Su^{a,1}, Yunyan Guo^{a,c,1}, David M. Wilkinson^d, Zhengwen Liu^e, Yong-Guan Zhu^a, Jun Yang^{a,*}

^a Key Laboratory of Urban Environment and Health, Institute of Urban Environment, Chinese Academy of Sciences, Xiamen 361021, People's Republic of China

^b College of Biological Science and Engineering, Fuzhou University, Fuzhou 350108, People's Republic of China

^c University of Chinese Academy of Sciences, Beijing 100049, People's Republic of China

^d School of Life Sciences, University of Lincoln, Lincoln LN6 7TS, UK

^e State Key Laboratory of Lake Science and Environment, Nanjing Institute of Geography and Limnology, Chinese Academy of Sciences, Nanjing 210008, People's Republic of China

ARTICLE INFO

Keywords:

Antibiotic resistance genes
Functional biogeography
Anthropogenic activities
Community ecology
Geographical distance
Multidrug resistance gene
Lake and reservoir

ABSTRACT

Antibiotic resistance genes (ARGs) are widespread in aquatic environments, but we know little about their biogeographical distribution and occurrence at national scales. Here we analyzed the patterns of ARGs from 42 natural waterbodies (natural lakes and reservoirs) across China using high-throughput approaches. The major ARGs were multidrug genes and the main resistance mechanism was the efflux pump. Although the absolute abundance of ARGs (gene copies/L) in the south/central waterbodies was similar to the northern waterbodies, the normalized abundance of ARGs (ARGs/16S rRNA gene copy number) was higher in the south/central waterbodies than in the north (mainly because of the aminoglycoside and multidrug resistance genes). Human activities strongly correlated with the normalized abundance of ARGs. The composition of ARGs in the waterbodies of south/central China was different from that in the north, and ARGs showed a distance-decay relationship. Anthropogenic factors had the most significant effects on this spatial distribution of ARG composition, followed by the spatial, bacterial and physicochemical factors. These indicate that the ARGs exhibited biogeographical patterns and that multiple ecological mechanisms - such as environmental selection (human activities and local physicochemical parameters) and dispersal limitation - influence distribution of ARGs in these waters. In general, our results provide a valuable ecological insight to explain the large-scale dispersal patterns in ARGs, thereby having potential applications for both public health and environmental management.

1. Introduction

Antibiotics have been widely used to treat bacterial infections since the 1940s and have revolutionized global medicine (Allen et al. 2010; Marti et al. 2014). However, the widespread occurrence of microbial resistance to antibiotics has become 'possibly the best example of human-driven evolution in action' at a global scale (Gillings and Paulsen 2014), with the first cases of antibiotic resistance typically being reported only a few years after the first use of a particular antibiotic (Pruden 2014). This has been considered as a serious public health problem by some scientists for several decades, and has led to the frighteningly plausible predictions of the end of effective antibiotics in medicine (Levin and Andreasen 1999). Clearly, the control of

antibiotic resistance in medical settings, such as hospitals, is important, but it has become apparent that the wider environment has become a potentially major pool of, and source of, antibiotic resistance genes (ARGs) (Pehrsson et al. 2016). In particular, the extensive use of antibiotics by humans, and their subsequent release into aquatic ecosystems in treated and untreated sewage, hospital waste, aquaculture discharges, and agricultural runoff have made aquatic environments crucial for the occurrence, exchange, evolution and spread of ARGs (Allen et al. 2010; Pruden 2014; Rodriguez-Mozaz et al. 2015; Y.G. Zhu et al. 2017c). However, there is still a lack of fundamental data on the distributions and effects of ARGs in natural waters (Czekalski et al. 2015).

Previous biogeographical studies have largely used 16S rRNA gene (via either sequencing or fingerprinting) as a standard approach for

* Corresponding author at: Aquatic Ecohealth Group, Key Laboratory of Urban Environment and Health, Institute of Urban Environment, Chinese Academy of Sciences, Xiamen 361021, People's Republic of China.

E-mail address: jyang@iue.ac.cn (J. Yang).

¹ These authors contributed equally to this work.

determining microbial diversity and differences in composition between communities, but it can be difficult to know how variation in taxonomic composition relates to ecosystem function (Louca et al. 2016). More recently, there has been a growing consensus on the need for a better representation of functional traits in biogeography - an approach which is better developed in macro-organism studies than in microorganisms (Violle et al. 2014). One of the lines of evidence that microorganisms display spatial biogeographical patterns is the distance-decay relationship (i.e. A decline in similarity with increasing geographical distance) (Hanson et al. 2012). This relationship shows that different locations harbor microorganisms that differ in taxonomic and functional composition. In functional biogeography, some studies found no significant correlation between geographical distance and microbial functional differences, implying that the functional traits available to a specific community have no physical constraints on their dispersal (Angermeyer et al. 2016; Raes et al. 2011). However, other studies suggest microbial functional genes have a biogeographical provincialism, as these functional genes correlated to geographical distance (Haggerty and Dinsdale 2017; Kelly et al. 2014). To date, we still know little about the whether the ARGs show a distance-decay relationship, and how the four fundamental processes that drive the biogeography of microbial taxonomic communities - selection, drift, dispersal and mutation - affect the ARG biogeographical patterns (Hanson et al. 2012). Studies on the biogeography of ARGs can enhance our knowledge on the dispersal mechanism of ARG across ecosystems. It can also contribute to strategies for prevention and control of the emergence, spread, enrichment and exchange of ARGs in both natural waters and manmade ecosystems (Berendonk et al. 2015).

In this study, we used high-throughput quantitative PCR (HT-qPCR) to investigate the ARGs along a latitudinal gradient ranging from 24 to 50°N (over 2700 km) in 42 lakes and reservoirs across China. China has a greater area than all of the countries of Europe put together - because of the size of the country, this is effectively a continental scale survey of the distribution of ARGs in waterbodies. We aimed to determine the abundance and diversity patterns of ARGs at a large-scale in the Chinese lakes/reservoirs and explore how the ecological processes and mechanisms that drive these biogeographical patterns of ARGs.

2. Materials and methods

2.1. Study area, sampling and physicochemical analysis

A total of 42 Chinese lakes and reservoirs were sampled during July and August 2012 (Fig. 1 and Table S1). As the emphasis in this study was on maximizing the number of different sites across China, only one collection was taken at each lake/reservoir. In order to guarantee the representation of each sample, first, the sample site was located at the center of each lake/reservoir where the water was mixed well. Second, we collected several water samples every 50 cm in the epilimnion, and then mixed them very well. Only one mixed sample was taken at the center of each waterbody (natural lake or reservoir). These waterbodies spanned a south-north gradient of over 2700 km and ranged from approximately 24 to 50°N. The study covered several major climate types; comprising five regions based on their climate and geographical characteristics: FJ (included 5 reservoirs) - Fujian province, southeast China; CJ (9 lakes) - the lower and middle reaches of Changjiang River, China; ECC (6 lakes) - east central China, IM (13 lakes) - Inner Mongolia, north China; NEC (9 lakes) - northeast China. All of these lakes have been featured in our previous work on large scale patterns in microbial taxonomic community diversity (Ju et al., 2014; Liu et al. 2015). We further classified these five regions into two large regions, that is, south/central China (included FJ, CJ, and ECC) and north China (included IM and NEC) based on the human population, economy and climate. Water samples were collected using a polypropylene bottle. The bottle was sterilized in an autoclave and pre-rinse by lake/reservoir water. All water samples were stored in the dark at 4°C and filtered

within two hours.

Water temperature (WT), electrical conductivity (EC), pH, dissolved oxygen (DO) and turbidity (Turb) of the epilimnion layer, water depth of sampling site, water transparency (Trans) and concentrations of chlorophyll *a* (Chl *a*), total nitrogen (TN) and total phosphorus (TP) were measured using the methods described in our previous study (Liu et al. 2015).

2.2. DNA extraction

Water samples of approximately 500 ml were collected for ARG analyses and filtered through a 0.22-μm pore size polycarbonate filter (47 mm diameter, Millipore, Billerica, MA, USA) as described previously (Liu et al. 2015). Total lake/reservoir microbial DNA was extracted directly from the filter using the FastDNA SPIN Kit and the FastPrep Instrument (MP Biomedicals, Santa Ana, CA, USA) according to the manufacturer's instructions. All DNA samples were checked for quality using the agarose gel electrophoresis and quantified using the NanoDrop ND-100 device (Thermo Fisher Scientific, Waltham, MA, USA).

2.3. High-throughput quantitative PCR of ARGs

We analyzed the ARGs in 42 lakes and reservoirs using high-throughput quantitative PCR (HT-qPCR). HT-qPCR of ARGs was performed using the SmartChip Real-time PCR (Wafergen Biosystems, Fremont, CA, USA) as described previously (Wang et al. 2014; Takara Bio USA, 2017). A total of 296 primer pairs were used to target one 16S rRNA gene, 285 ARGs (resistance to major classes of antibiotics), 8 transposase genes, one universal class I integron-integrase gene (*intI*), and one clinical class I integron-integrase gene (*cintI*) (Table S2). These 285 ARGs confer resistance to major classes of antibiotics (Table S2). All these primer sets were designed, used, and validated in the previous studies (Looft et al. 2012; Su et al., 2015; Zhu et al. 2013, YG Zhu et al., 2017c). Antibiotic resistance-gene reference sequences were harvested from (1) the Antibiotic Resistance Genes Online database (<http://ardb.cbc.umd.edu/>); (2) a National Center for Biotechnology Information (NCBI) search for resistance-gene sequences; and (3) literature searches. All qPCRs were performed in triplicate, and a non-template negative control was included. The results of the HT-qPCR were analyzed with the standard settings: (1) any wells with multiple melting peaks and/or amplification efficiency < 1.8 or > 2.2 were discarded, and (2) a threshold cycle (Ct) 31 was used as the detection limit and only sample with three replicates simultaneously (Ct < 31) were regarded as positive. The relative copy number of ARGs, MGEs and 16S rRNA gene was calculated according to a previous study (Eq. 1) (Looft et al. 2012; Muurinen et al. 2017).

$$\text{Relative gene copy number} = 10^{(31 - Ct)/(10/3)} \quad (1)$$

where Ct refers to high-throughput quantitative PCR results, 31 refers to the detection limit, and 10/3 refers to the 10-fold difference in gene copy numbers is 10/3 cycles, if the efficiency is close to 100%.

To obtained the absolute abundance of ARGs, we quantified the absolute copy number of 16S rRNA gene (gene copies/L) using a Lightcycler 480 instrument (Roche, Basel, Switzerland) with a SYBR® Green approach as previously described (Ouyang et al. 2015). A significantly high correlation between the relative copy number of 16S rRNA gene from SmartChip Real Time System and the absolute copy number of 16S rRNA gene from Roche 480 was observed ($r = 0.95$, $P < 0.01$). Therefore, we transformed the relative copy number of ARGs to absolute copy number according to a previous study (Eq. 2) (Ouyang et al. 2015).

$$\text{ARG}_{\text{SmartChip}} \text{ (or } \text{MGE}_{\text{SmartChip}}) / 16\text{S}_{\text{SmartChip}} = \text{ARG}_{\text{absolute}} \text{ (or } \text{MGE}_{\text{absolute}}) / 16\text{S}_{\text{Roche480}} \quad (2)$$

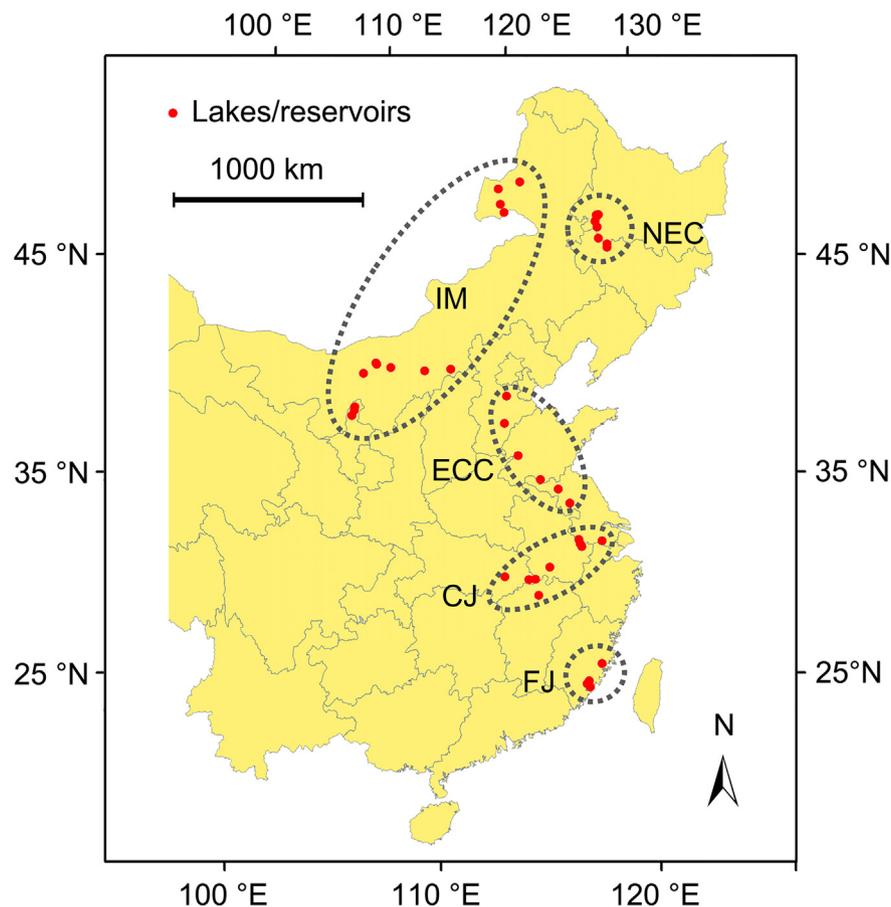


Fig. 1. The partial map of China showing the location of 42 sampling stations. FJ (included 5 reservoirs): Fujian province, southeast China; CJ (9 lakes): the lower and middle reaches of Changjiang River, China; ECC (6 lakes): east central China, IM (13 lakes): Inner Mongolia, north China; NEC (9 lakes): northeast China.

where $ARG_{SmartChip}$, $MGE_{SmartChip}$ and $16S_{SmartChip}$ are ARG, MGE and 16S rRNA gene assays from SmartChip Real Time System (relative copy number), $16S_{Roche480}$ is absolute abundance of 16S rRNA gene assays from Roche Lightcycler 480, while $ARG_{absolute}$ and $MGE_{absolute}$ are absolute abundance of ARG and MGE, respectively.

2.4. Anthropogenic factors

The anthropogenic factors (population, GDP, aquatic production, meat production, municipal domestic sewage, number of patients diagnosed and treated, and number of residential patients) were collected or estimated according to the main administrative region covered by each basin (Table S3). These values and the explanation of these anthropogenic data collected, or estimated, in this study can be found in the supplementary information (Tables S3, S4 and S5). Given the difference in area between the 42 lake/reservoir basins, all anthropogenic factors were normalized by the area of administrative regions (km^2) where were mainly covered by each basin.

2.5. Statistics

The Bray-Curtis similarity matrix of ARGs profile was constructed based on the normalized abundance of ARGs (ARGs/16S rRNA gene copy number). Non-metric multidimensional scaling (NMDS) ordination and analysis of similarities (ANOSIM) were then used to investigate differences in Bray-Curtis similarity of composition of ARGs between sites (Clarke and Gorley, 2015). Further, the similarity percentage (SIMPER) procedure was used to identify ARGs that contributed most to the dissimilarity between sites.

The Shannon-Wiener index was calculated according to Eq. 3.

$$H = - \sum_{i=1}^n P_i \ln P_i \quad (3)$$

where H is Shannon-Wiener index, n is the total number of ARGs (richness of ARGs) and P_i is the proportion of copy number of ARG i in a sample.

We used variation partitioning to determine the relative contribution of anthropogenic, spatial, physicochemical and bacterial factors to the distribution of ARG composition based on the normalized abundance of ARGs. Before the variation partitioning, we used a forward selection procedure to select anthropogenic, spatial and physicochemical variables using the 'ordiR2step' function from vegan R package (Blanchet et al. 2008). All variables where $P > 0.05$ were eliminated in further analyses. Spatial factors were generated through the use of principal coordinates of neighbor matrices (PCNM) analysis based on the geodetic coordinates from latitude and longitude of each site (using the pcnm R package) (Borcard and Legendre 2002). Bacterial factors are the relative proportion (the number of sequences of a bacterial group/total number of sequences of all bacterial groups in a sample) of major groups of abundant and rare bacteria (Actinobacteria, Bacteroidetes, Cyanobacteria, Firmicutes, Planctomycetes, Proteobacteria, Verrucomicrobia and unclassified bacteria) (Liu et al. 2015). The abundant bacterial OTUs were defined as the OTUs with a representation of $\geq 1\%$ within a sample and had a mean relative abundance of $\geq 0.1\%$. The rare bacterial OTUs were defined as having an abundance of $< 0.01\%$ within a sample and having a mean relative abundance of $< 0.001\%$ (Liu et al. 2015).

To visualize the correlations between bacterial taxa and ARGs, we constructed a network between bacterial OTUs and ARGs by calculating all possible pairwise Spearman's rank correlations based on the

sequence number of bacterial OTUs and normalized abundance of ARGs. The bacterial OTUs were characterized by Illumina sequencing (MiSeq platform) of the 16S rRNA gene as described by Liu et al. (2015). All 16S rRNA gene sequence data in this paper have been deposited in the public NCBI Sequence Read Archive database under the accession number SRX525963. A correlation between two items was considered statistically robust if the Spearman's correlation P value was < 0.001 . All P values were adjusted using the Benjamini and Hochberg false discovery rate (FDR) procedure (the multtest R package) to reduce the chances of obtaining false-positive results with FDR-adjusted P value < 0.05 (Benjamini and Hochberg 1995). To further reduce false-positives, we restricted our analysis to bacterial OTUs and ARGs present both in $> 30\%$ of the samples. The bacteria OTUs, which we were unable to classify into family level, were excluded in this analysis. All these analyses were performed in PRIMER 7.0 (Clarke and Gorley, 2015) and R 3.2.3 (R Core Team 2015).

In total 41 lakes/reservoirs were included in the variation partitioning and network analyses, because we failed to detect ARGs by high-throughput qPCR in Lake Chaohu, and failed to obtain bacterial taxonomic OTUs by high-throughput sequencing in Lake Poyang (Liu et al. 2015).

3. Results

3.1. Spatial distribution of ARGs in waterbodies

The absolute abundance of ARGs (i.e. the copy number of ARGs per litre water) varied from 5.95×10^6 (Quansanhai Lake) to 1.85×10^9 (Shijiu Lake) with mean of $4.33 \times 10^8 \pm 7.19 \times 10^7$ standard error (s.e.) ($n = 42$). The absolute abundance of ARGs and mobile genetic elements (MGEs) (i.e. the copy number of MGEs per litre water) and the normalized abundance of MGEs (i.e. the MGEs/16S rRNA gene copy number) did not show significant difference between the southern/central and northern waterbodies (Fig. 2A and Fig. S1). However, the normalized abundance of ARGs was significantly higher in the south/central than in the north China (Fig. 2B). Moreover, the genes conferring resistance to aminoglycoside were significantly higher in the southern/central than in the northern waterbodies, in both absolute and normalized abundance. The multidrug resistance genes were significantly higher in the south/central than in the north waterbodies in normalized abundance (Table S6).

In total, 167 unique ARGs and 9 mobile genetic elements (MGEs)

were detected in this study. The number of ARGs varied from 15 (Shengjing Lake) to 57 (Hubian Reservoir) (mean = 32, s.e. = 1, $n = 42$). Similar to the normalized abundance of ARGs, ECC had the highest mean richness of ARGs, followed by FJ, CJ, IM and NEC, although there were no formally significant differences among these five regions (Fig. 2C). The Shannon-Wiener index of ARGs ranged from 0.42 to 3.10 with mean of 1.96 ± 0.11 s.e. (Fig. 2D).

The detected ARGs potentially conferred resistance to almost all major classes of antibiotics, the percentage of multidrug resistance genes was the highest (mean \pm s.e. = $45.28\% \pm 4.38\%$, $2.15 \times 10^8 \pm 4.78 \times 10^7$ copies/L), followed by the β -lactamase resistance genes ($21.42\% \pm 3.25\%$, $9.98 \times 10^7 \pm 3.25 \times 10^7$ copies/L). These ARGs encompassed three major resistance mechanisms including efflux pump, antibiotic modification or deactivation, and cellular protection (Figs. S2 and S3). However, the richness of multidrug resistance genes was the highest (7.88 ± 0.42), followed by the aminoglycoside resistance (6.17 ± 0.44) and β -lactamase resistance genes (5.07 ± 0.40) (Fig. S4).

The spatial pattern of ARG composition was not clearly separated in terms of the five regions (global $R = 0.206$, $P = 0.004$). Interestingly, most of the sites from north China (i.e. IM and NEC) were placed on the upper region of the ordination, while most of the sites from south and central China (i.e. ECC, CJ and FJ) were clustered in the lower region (global $R = 0.268$, $P = 0.001$) (Fig. 2E). The SIMPER analysis further indicated the mean dissimilarities between north and south/central China was $82.73\% \pm 2.34\%$, and the main contribution to these separations was the high abundance of the *mexF* gene (multidrug resistance gene, $42.93\% \pm 7.28\%$) and *blaCTX-M-04* gene (β -lactamase, $7.43\% \pm 4.01\%$) in the south and central China, and *fox5* gene (β -lactamase, $12.32\% \pm 3.68\%$) in the north China.

About 76.0% ARGs (127) were shared between the south/central and north regions (Fig. 2E inset). However, 12 ARGs were unique in the south/central Chinese waterbodies (two aminoglycoside, four β -lactamase, two MLSB (macrolide-lincosamide-streptogramin B), two tetracycline and two multidrug resistance genes) and 28 ARGs were unique in the north Chinese waterbodies (one aminoglycoside, seven β -lactamase, six MLSB, one sulfonamide, three tetracycline, five vancomycin and four multidrug resistance genes, while one ARGs which confer resistance to triclosan). Importantly, the shared ARGs account for 98.71% of the total absolute abundance of ARGs (Table S7).

Both absolute and normalized abundances of ARGs showed a significant positive correlation with the number of locations at which they

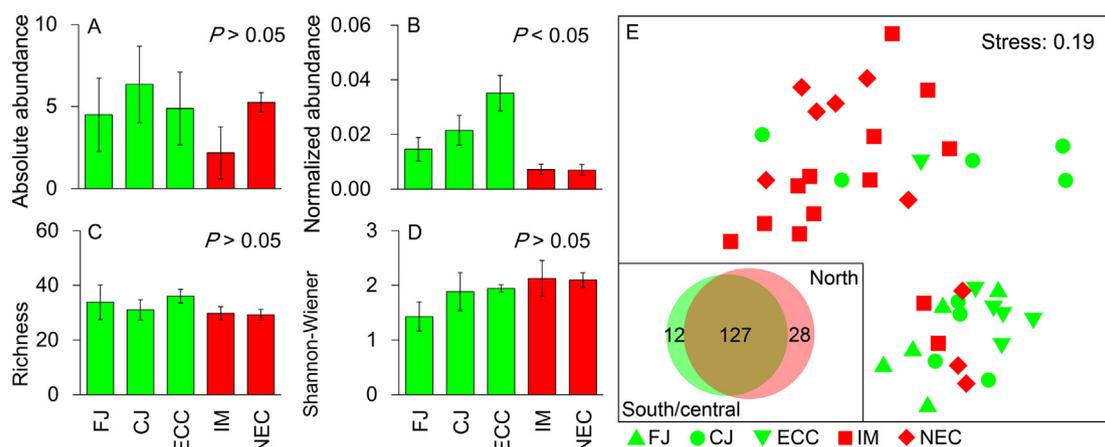


Fig. 2. Abundance, diversity and composition of ARGs in the 42 natural lakes and reservoirs across China (green - South and Central China, red - North China; see Materials and methods). Absolute abundance (10^8 copies/L, A), normalized abundance (ARGs/16S rRNA gene copy number, B), richness (C) and Shannon-Wiener diversity (D) of ARGs in five geographical regions (Data are expressed as mean \pm s.e.). One-way ANOVA was used to denote the differences among regions. Non-metric multidimensional scaling (NMDS) ordination of ARGs composition (E) and Venn diagram showing the number of ARGs that are unique and shared between south/central and north Chinese lakes/reservoirs (E inset). (For interpretation of the references to colour in this figure legend, the reader is referred to the web version of this article.)

Table 1
Spearman's rank correlations between environmental/anthropogenic parameters and normalized abundance of ARGs.

Environmental variables	ARGs/16S rRNA gene	Anthropogenic variables	ARGs/16S rRNA gene
Water depth	0.341**	Population per km ²	0.560**
Water temperature	0.616**	GDP per km ²	0.413**
Electrical conductivity	-0.182	Municipal domestic sewage per km ²	0.571**
pH	-0.242	Meat production per km ²	0.577**
Turbidity	-0.268	Aquatic production per km ²	0.610**
Water transparency	0.197	No. of patients diagnosed and treated per km ²	0.675**
Dissolved oxygen	-0.203	No. of residential patients per km ²	0.667**
Total nitrogen	0.087		
Total phosphorus	-0.364**		
Chlorophyll <i>a</i>	-0.038		

** $P < 0.01$.

were found, indicating a strong abundance-dependent effect on the biogeographical distribution of ARGs at the continental scale - the ARGs with higher abundance were detected in more samples (Fig. S5).

3.2. Relationships between anthropogenic, spatial, environmental factors and ARGs

We consider that the human-induced changes of antibiotic selection pressure between the lakes and reservoirs are one of the possible factors that drive the normalized abundance of ARGs. Seven anthropogenic factors – population, GDP, municipal domestic sewage, number of patients diagnosed and treated, number of residential patients, meat production and aquatic production were significantly, strongly and positively correlated with the normalized abundance of ARGs. However, only three physicochemical factors (water depth, water temperature and total phosphorus) showed significant correlations with the normalized abundance of ARGs (Table 1).

We found ARGs showed a significant distance-decay relationship (Fig. 3). Moreover, variation partitioning showed that pure anthropogenic factors (14.93%) had the most significant effects on the spatial distribution of ARG composition, followed by the pure spatial (8.93%), pure bacterial (5.49%) and pure physicochemical (0.44%) factors (Fig. 4). However, 43.55% of spatial variation of ARG composition was unexplained.

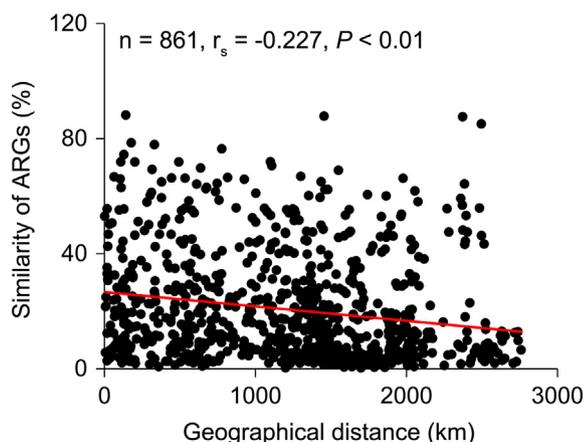


Fig. 3. Spearman's rank correlations between dissimilarity of ARGs composition and geographical distance.

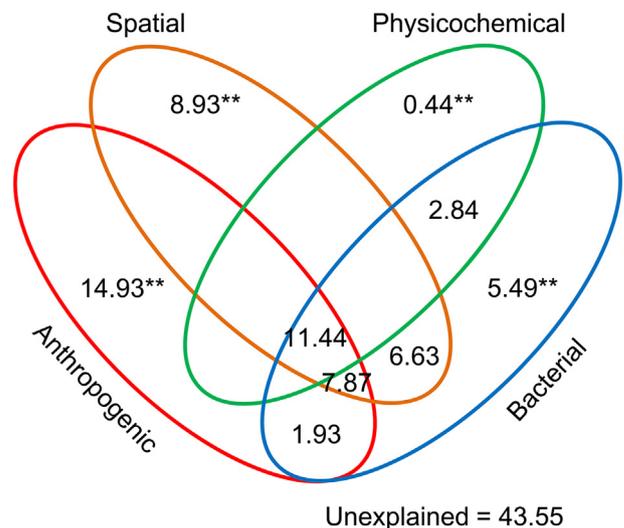


Fig. 4. Variation partitioning showing the effects (%) of anthropogenic (meat production, aquatic production, number of patients diagnosed and treated and number of residential patients), spatial (PCNMI, PCNM3 and PCNM24), physicochemical (water temperature) and bacterial (rare actinobacteria and rare firmicutes phyla) factors on spatial distribution of ARG composition in the lakes/reservoirs (** $P < 0.01$, $n = 41$, see Materials and methods). For simplicity, the explanations which $< 1\%$ were not shown except the pure physicochemical explanation.

3.3. Correlations between the abundance of ARGs and MGEs

The correlation between absolute abundance of ARGs and MGEs was significant (Table S8). However, the normalized abundance of ARGs and MGEs did not show a significant correlation (Table S9).

For the absolute abundance, the integrase gene *intl-1* exhibited a high correlation with the gene classes conferring resistance to aminoglycoside, β -lactamase and MLSB. The transposase gene *tnpA-04* showed a high correlation with aminoglycoside, FCA (fluoroquinolone, quinolone, florfenicol, chloramphenicol, and amphenicol), MLSB, sulfonamide, tetracycline, vancomycin and multidrug resistance genes. In addition, significant correlations were observed between the transposase genes *tnpA-01*, *tnpA-02* and *tnpA-05* and some ARG groups (Table S10). For the normalized abundance, no significant correlation was observed between the integrase genes and ARG groups. However, the transposase genes *tnpA-03*, *tnpA-04* and *tnpA-05* showed high correlation with some ARG groups (Table S11). The correlation of transposase gene with ARG groups suggests that some transposase genes may contribute significantly to the dissemination of specific ARG groups.

3.4. Co-occurrence patterns between bacterial taxa and ARGs

Our network analysis showed co-occurrence relationships between the bacterial families and ARGs. Forty-nine bacterial families were strongly correlated with eight types of ARGs. The multidrug resistance genes significantly correlated with 35 bacterial OTUs which were affiliated with 24 bacterial families. The aminoglycoside resistance genes significantly correlated with 29 bacterial OTUs affiliated with 19 bacterial families. The MLSB resistance genes significantly correlated with 17 bacterial OTUs affiliated with 13 bacterial families. The β -lactamase, sulfonamide, vancomycin, tetracycline and FCA resistance genes significantly correlated with some of bacterial families (Fig. S6). Chitinophagaceae bacterial OTUs had the most edges with the ARGs, flowed by the Planctomycetaceae and Sphingomonadaceae bacteria.

4. Discussion

4.1. The biogeography of ARGs

This is the first national/continental scale study using high-throughput methods to investigate human-induced environment effects on the abundance and composition of ARGs in inland waterbodies. We found the normalized abundance and composition of ARGs exhibited biogeographical patterns in natural water ecosystems at a national scale - the ARGs normalized abundance was significantly higher in the south/central than in the north Chinese waterbodies, and the ARGs showed a significant distance-decay relationship. As a functional trait, the ARGs have a biogeography at a large spatial scale suggesting that they are most likely subjected to some ecological processes and environmental filtering (Hanson et al. 2012).

First, our data suggest that environmental selection associated with human activities may act strongly both on the spatial distribution of ARG composition and normalized abundance (Table 1 and Fig. 4). Although many antibiotics and ARGs are a natural part of microbial ecology (Allen et al. 2010; Wiener 1996), and can be shown to have existed before human antibiotic use (D'Costa et al. 2011), human activities have been demonstrated to make a strong contribution to the occurrence, spread and enrichment of antibiotics and antibiotic resistance (Jones et al. 2008; Qiao et al. 2018). A diverse mixture of antibiotics with other pollutants and their metabolites can reach the natural rivers, lakes and reservoirs through the urban wastewater, aquaculture discharges, and agricultural runoff, altering the distribution and magnitude of ARGs in both natural and man-made aquatic environments (Marti et al. 2014; YG Zhu et al. 2017a; 2017b). More importantly, it has been shown that the annual emissions of antibiotics in watersheds of south and central China was higher than in north China (Zhang et al. 2015). Therefore, the increase of human activity and intensity from north to south/central China may be an important reason for the increase of emissions of antibiotics and thus may result in the increase of normalized abundance of ARGs in the south/central lakes and reservoirs. Second, environmental selection associating with local physicochemical factors has an influence on the spatial distribution of ARG composition. These physicochemical factors may have indirect effects on ARGs due to their effect on the host bacterial abundance and community composition (Su et al., 2015). Finally, our distance-decay and variation partitioning analyses (pure spatial factors significantly correlated with the distribution of ARG composition) suggest dispersal limitation of ARGs across the waterbodies (Fig. 4). The most likely reason is the low abundance of ARGs in the natural waterbodies. For example, the mean of ARGs to 16S rRNA gene copy number ratio was about 1.5% in the waterbodies. The low abundance of functional genes (e.g. ARGs) may result in low dispersal chances of these functional genes though host species dispersal, horizontal gene transfer and gene flow across ecosystems (Liu et al. 2015; Raes et al. 2011). Moreover, the Actinobacteria and Firmicutes phyla were closely related to the ARG composition in the waterbodies. This result agrees with our previous study of urban sewage sludge composting (Su et al., 2015). However, this result also indicates that the rare bacteria may play more important roles in the spatial distribution of ARG composition compared to abundant bacteria.

It should be noted that some of our anthropogenic data refer to the corresponding values in the province where each waterbody is located. Therefore, these data may not fully and precisely reflect the characteristics of basin where each lake or reservoir is located. However, our anthropogenic data reflect the broad anthropogenic characteristics of the different geographical regions across China, and these data closely correlated with the normalized abundance of ARGs in these waters. In addition, it should be noted that the large variation (43.55%) in composition of ARGs cannot be explained by variation partitioning. We assume that this high proportion of unexplained variation of ARGs could be related to the antibiotic concentration, and other biotic and

abiotic factors not measured in this study. This might also reflect the complex mechanisms that generate the distribution of ARG composition in these waterbodies.

4.2. The abundance and composition of ARGs and MGEs

The absolute abundance of ARGs ranged from 5.95×10^6 to 1.85×10^9 copies/L with mean of $4.33 \times 10^8 \pm 7.19 \times 10^7$ s.e. in the waterbodies. Previous studies using HT-qPCR investigated ARGs in urban areas of Wenruitang River (Zhejiang, China) (Zhou et al. 2017), Jiulong River (Fujian, China) (Ouyang et al. 2015), East Tiaoxi River (Zhejiang, China) (Zheng et al. 2017) and Qiantang River (Zhejiang, China) (Xu et al., 2016). These studies found the absolute abundance of ARGs in urban areas of rivers reached 10^9 – 10^{11} copies/L, which was one to three orders of magnitude higher than that in our waterbodies. However, most of our sampling sites are located within suburban areas and in rural locations surrounding major urban areas, these have lower human activity compared to urban areas. Ouyang et al. (2015) found the absolute abundance of ARGs in a suburban site of Jiulong River was about 10^8 copies/L, which is similar with the result in this study.

MGE marker genes *int1-1* and *tnpA-04* were detected at high concentrations in most of our samples, and they exhibited high correlations with ARG groups. The *int1-1* significantly correlated with aminoglycoside, β -lactamase and MLSB resistance genes, while *tnpA-04* significantly associated with aminoglycoside, FCA, MLSB, multidrug, sulfonamide, tetracycline and vancomycin resistance genes. These indicate the important roles of these two MGE genes in ARGs transfer in the waterbodies. Previous studies using the HT-qPCR array found *tnpA-04* gene is the most enriched MGEs in treated water generated from drinking water treatment plants (DWTPs) (Xu et al., 2016) and in park soils irrigated with reclaimed water (Wang et al. 2014). While other studies also found the *int1-1* was persistent with high abundance in estuarine sediments (YG Zhu et al. 2017c). The class 1 integrons found in the clinical context are able to capture different ARGs in the form of circular cassette to construct a tandem array of ARGs (YG Zhu et al. 2017c). The persistence of integrons (*int1-1*) and the close relationship between transposons (*tnpA-04*) and ARGs in the waterbodies could further emphasize the risk of ARG transfers from the natural water to human pathogens (Djordjevic et al. 2013; Domingues et al. 2012; Gillings et al. 2008).

It is well known that bacteria have many efflux pumps in their membranes. Some of these pumps are able to transport antibiotics out of the bacterial cell, allowing bacteria to survive at higher concentrations of antibiotics, and thus leading to resistance. Some efflux pumps can transport antibiotics of more than one chemical class, thereby being associated with multidrug (antibiotic) resistance (Piddock 2006). In this study, multidrug resistance genes (e.g. *qacE* Δ gene) were highly abundant in the waterbodies, perhaps due to the extensive use of different antibiotics within hospitals, communities and animal production. For example, the *qacE* Δ gene was abundant and detected in most of samples. This result is congruent with our previous finding in estuarine sediments (YG Zhu et al. 2017c). The *qacE* Δ 1 is found specifically in class 1 integrons and specify gene for resistance to quaternary ammonium compounds (Aleksun and Levy 2007). Given the serious threat of multidrug resistance genes, this highlights the need for the scientific community and public authorities to collaborate to prevent and control the development and dissemination of ARGs in aquatic environments. Moreover, we detected 13 of the 26 tetracycline resistance genes targeted on our array. The most abundant *tet* genes in this study were *tetG*, *tetB*, *tetR*, *tetX* and *tetA*. Tetracyclines (TCs) are commonly used as veterinary medicines (Jechalke et al. 2014). In the present study, we found that the proportion of tetracycline resistance genes were apparently higher in Fujian province than in four other regions (Fig. S2). The tetracycline resistance genes in Hubian Reservoir (Fujian) were two orders of magnitude higher than that in other Fujian reservoirs. Hubian Reservoir is a backup water source for Xiamen City and completely

surrounded by urban development. In the dry season, this reservoir water is supplied by Jiulong River when its water level falls below the normal value. Previous work has shown that tetracycline antibiotics were highest among the measured fourteen antibiotics (tetracyclines, sulfonamides and quinolones) in water samples along the Jiulong River (Zhang et al. 2012). Another study investigated five classes of antibiotic resistance genes along the Jiulong River and found the high concentration of *tetA* and *tetG* in water samples (Zhang et al. 2013). Therefore, the Hubian Reservoir may be influenced by the tetracycline residues and tetracycline-resistance genes from the Jiulong River water, which is a human-impact river (Zhang et al. 2013).

4.3. The co-occurrence relationship between ARGs and bacterial taxa

Previous studies used network analysis to investigate the co-occurrence relationship between ARGs and bacterial taxa and hypothesized that the co-occurrence patterns between ARGs and bacterial taxa could possibly indicate host information of ARGs (Chen et al. 2016; Li et al. 2015; B Zhu et al. 2017). For example, Li et al. (2015) found both *Escherichia* and *Bacteroides* were the host of *tetQ*, and this result has been verified in other studies (Forslund et al. 2013; Shoemaker et al. 2001; Zhang et al. 2009). In the present study, we also found co-occurrence relationships between ARGs and 23 bacterial families in the waterbodies. This result is inconsistent with a previous study on soil resistomes (B Zhu et al. 2017). In that study, the authors used network analysis and found Chitinophagaceae was a possible host of a vancomycin (*vanC-03*), a tetracycline (*tetR-02*) and a MLSB (*mphA-02*) resistance genes. However, in our study, we found Chitinophagaceae were correlated with four aminoglycoside, three MLSB (*ereA*, *erm(36)*, *ermF*), four multidrug and a FCA resistance genes. Nevertheless, the network analysis can provide information regarding the correlations of the ARGs with some bacterial taxa, but further study should verify bacterial hosts of the ARGs in these aquatic ecosystems.

4.4. Potential limitations

There are, however, potential limitations in our approach that merit further discussion. We only covered a limited number of ARG types using PCR-based approach although 285 ARGs were investigated. Therefore, some ARG types would likely be missed. Current next-generation sequencing and metagenomics approach are able to investigate ARGs across a broader spectrum and obtain a more complete picture of the ARG profiles. However, our high-capacity quantitative PCR arrays are more efficient than most of the previous studies which only targeted a few classes of ARGs (Li et al. 2015).

Moreover, it should be noted that the results from the absolute abundance and normalized abundance of ARGs did not match exactly. There are two possible reasons for this phenomenon. First, the normalized abundance of total ARGs in the waterbodies was low (mean = 1.5%, s.e. = 0.2%, n = 42), and therefore resulted in the relatively decoupled relationship between bacterial taxa and ARGs. In contrast, we found the normalized abundance of total ARGs was high during the urban sewage sludge composting (mean = 32.7%, s.e. = 3.4%, n = 23), where the bacterial showed a tight correlation with ARGs (Su et al., 2015). Second, the correlation between the 16S rRNA gene abundance from the SmartChip Real Time System and from the Roche Lightcycler 480 may have some effects on the consistency of results from the absolute abundance and from the normalized abundance of ARGs.

In this study, only one mixed sample was taken at each lake/reservoir. To make these samples as representative as possible, the sample site was located at the center of each waterbody. We then collected several water samples every 50 cm in the epilimnion, and mixed the water as one sample. There is an obvious trade-off between numbers of samples from a site and number of sites studied. This work prioritized number of sites (across China) - while this has many advantages it

obviously have disadvantages when it comes to replicates from individual lakes. There is always a risk that a single collection differs quite a bit from the mean and the values for some lakes may not be representative. Therefore, the more intensive sample strategies should be considered in future study.

In addition, it is worth noting that our sampling was performed only during the summer months (i.e. July and August). There are likely seasonal differences in bacterial abundance, diversity as well as the genes. Likewise, our study sites are located over different climate zones and would likely be affected by the different hydrothermal conditions. For example, the cyanobacteria tend to bloom in summer months and may strongly affect the whole bacterial communities so influencing the ARGs (Eiler and Bertilsson 2004; Guo et al. 2018). Therefore, our results are only strictly applicable to the time-frame in which the samples were collected, and that there is a possibility that seasonal differences in ARG composition and abundance could occur.

5. Conclusions

Our results demonstrated that, at a national (effectively continental) scale, the ARGs had a distinct biogeographical pattern in the studied Chinese lakes and reservoirs - the composition of ARGs in the waterbodies of south/central China was separated from that in the north, and the ARGs showed a significant distance-decay relationship. Anthropogenic (the highest explanation factors), spatial and physicochemical factors exhibited significant effects on this spatial distribution of ARG composition. The south/central China had higher normalized abundance of ARGs (mainly because of aminoglycoside and multidrug resistance genes) than north (northwest and northeast) China in the waterbodies as the effect of human activities. In general, the selection process associated with human activities and mobile genetic elements were closely related to the spatial distribution of ARGs in the waters, while the local physicochemical parameters and dispersal limitation also played important roles in ARG distribution. Overall, these results indicate that the root-cause of accumulation/enrichment and spread/dissemination of ARGs is intensive human activities, this threatens both public health and natural ecosystem health.

Conflict of interest

All authors have no conflict of interest to declare.

Author contributions

J.Y. and Y.G.Z. designed the research. J.Y. and L.L. collected the samples from lakes and reservoirs. L.L., J.Q.S. and Y.G. performed the laboratory experiments. L.L., J.Q.S., J.Y. and Y.G.Z. contributed the new reagents/analytic tools, and analyzed the data. All authors discussed the interpretation of the results and wrote the manuscript. J.Y. conceived the initial concept.

Acknowledgements

We thank Kuanyi Li, Feizhou Chen, Lingyu Yu, Zheng Yu, Shunchun Yao and Bin Xue for field sampling assistance. This work was funded by the National Key Research and Development Program of China (2017YFA0605203), and the National Natural Science Foundation of China (31500372, 31370471, 21210008 and 41571130063), the K. C. Wong Education Foundation of Hong Kong.

Appendix A. Supplementary data

Supplementary data to this article can be found online at <https://doi.org/10.1016/j.envint.2018.05.023>.

References

- Alekshun, M.N., Levy, S.B., 2007. Molecular mechanisms of antibacterial multidrug resistance. *Cell* 128, 1037–1050.
- Allen, H.K., Donato, J., Wang, H.H., Cloud-Hansen, K.A., Davies, J., Handelsman, J., 2010. Call of the wild: antibiotic resistance genes in natural environments. *Nat. Rev. Microbiol.* 8, 251–259.
- Angermeyer, A., Crosby, S.C., Huber, J.A., 2016. Decoupled distance-decay patterns between *dsrA* and 16S rRNA genes among salt marsh sulfate-reducing bacteria. *Environ. Microbiol.* 18, 75–86.
- Benjamini, Y., Hochberg, Y., 1995. Controlling the false discovery rate: a practical and powerful approach to multiple testing. *J. Roy. Stat. Soc. B* 57, 289–300.
- Berendonk, T.U., Manaia, C.M., Merlin, C., Fatta-Kassinos, D., Cytryn, E., Walsh, F., Bürgmann, H., Sørum, H., Norström, M., Pons, M.N., 2015. Tackling antibiotic resistance: the environmental framework. *Nat. Rev. Microbiol.* 13, 310–317.
- Blanchet, F.G., Legendre, P., Borcard, D., 2008. Forward selection of explanatory variables. *Ecology* 89, 2623–2632.
- Borcard, D., Legendre, P., 2002. All-scale spatial analysis of ecological data by means of principal coordinates of neighbour matrices. *Ecol. Model.* 153, 51–68.
- Chen, Q., An, X., Li, H., Su, J., Ma, Y., Zhu, Y.-G., 2016. Long-term field application of sewage sludge increases the abundance of antibiotic resistance genes in soil. *Environ. Int.* 92–93, 1–10.
- Clarke, K.R., Gorley, R.N., 2015. PRIMER v7: User Manual/Tutorial. PRIMER-E Ltd, Plymouth, UK.
- Czekalski, N., Sigdel, R., Birtel, J., Matthews, B., Bürgmann, H., 2015. Does human activity impact the natural antibiotic resistance background? Abundance of antibiotic resistance genes in 21 Swiss lakes. *Environ. Int.* 81, 45–55.
- D'Costa, V.M., King, C.E., Kalan, L., Morar, M., Sung, W.W., Schwarz, C., Froese, D., Zazula, G., Calmels, F., Debruyne, R., 2011. Antibiotic resistance is ancient. *Nature* 477, 457–461.
- Djordjevic, S.P., Stokes, H.W., Roy, C.P., 2013. Mobile elements, zoonotic pathogens and commensal bacteria: conduits for the delivery of resistance genes into humans, production animals and soil microbiota. *Front. Microbiol.* 4, 86.
- Domingues, S., Harms, K., Fricke, W.F., Johnsen, P.J., Da, S.G., Nielsen, K.M., 2012. Natural transformation facilitates transfer of transposons, integrons and gene cassettes between bacterial species. *PLoS Pathog.* 8, e1002837.
- Eiler, A., Bertilsson, S., 2004. Composition of freshwater bacterial communities associated with cyanobacterial blooms in four Swedish lakes. *Environ. Microbiol.* 6, 1228–1243.
- Forslund, K., Sunagawa, S., Kultima, J.R., Mende, D.R., Arumugam, M., Tynas, A., Bork, P., 2013. Country-specific antibiotic use practices impact the human gut resistome. *Genome Res.* 23, 1163–1169.
- Gillings, M.R., Paulsen, I.T., 2014. Microbiology of the anthropocene. *Anthropocene* 5, 1–8.
- Gillings, M.R., Krishnan, S., Worden, P.J., Hardwick, S.A., 2008. Recovery of diverse genes for class 1 integron-integrases from environmental DNA samples. *FEMS Microbiol. Lett.* 287, 56–62.
- Guo, Y.Y., Liu, M., Liu, L.M., Liu, X., Chen, H.H., Yang, J., 2018. The antibiotic resistome of free-living and particle-attached bacteria under a reservoir cyanobacterial bloom. *Environ. Int.* 117, 107–115.
- Haggerty, J.M., Dinsdale, E.A., 2017. Distinct biogeographical patterns of marine bacterial taxonomy and functional genes. *Glob. Ecol. Biogeogr.* 26, 177–190.
- Hanson, C.A., Fuhrman, J.A., Hornerdevine, M.C., Martiny, J.B.H., 2012. Beyond biogeographic patterns: processes shaping the microbial landscape. *Nat. Rev. Microbiol.* 10, 497–506.
- Jechalke, Sven, Heuer, Holger, Siemens, Jan, Amelung, Wulf, Smalla, Kornelia, 2014. Fate and effects of veterinary antibiotics in soil. *Trends Microbiol.* 22, 536–545.
- Jones, K.E., Patel, N.G., Levy, M.A., Storeygard, A., Balk, D., Gittleman, J.L., Daszak, P., 2008. Global trends in emerging infectious diseases. *Nature* 451, 990–993.
- Ju, L.H., Yang, J., Liu, L.M., Wilkinson, D.M., 2014. Diversity and distribution of freshwater testate amoebae (protozoa) along latitudinal and trophic gradients in China. *Microb. Ecol.* 68, 657–670.
- Kelly, L.W., Williams, G.J., Barott, K.L., Carlson, C.A., Dinsdale, E.A., Edwards, R.A., Haas, A.F., Haynes, M., Lim, Y.W., Mcdole, T., 2014. Local genomic adaptation of coral reef-associated microbiomes to gradients of natural variability and anthropogenic stressors. *Proc. Natl. Acad. Sci. U. S. A.* 111, 10227–10232.
- Levin, S.A., Andreasen, V., 1999. Disease transmission dynamics and the evolution of antibiotic resistance in hospitals and communal settings. *Proc. Natl. Acad. Sci. U. S. A.* 96, 800–801.
- Li, B., Yang, Y., Ma, L., Ju, F., Guo, F., Tiedje, J.M., Zhang, T., 2015. Metagenomic and network analysis reveal wide distribution and co-occurrence of environmental antibiotic resistance genes. *ISME J.* 9, 2490–2502.
- Liu, L.M., Yang, J., Yu, Z., Wilkinson, D.M., 2015. The biogeography of abundant and rare bacterioplankton in the lakes and reservoirs of China. *ISME J.* 9, 2068–2077.
- Looft, T., Johnson, T.A., Allen, H.K., Bayles, D.O., Alt, D.P., Stedtfeld, R.D., Sul, W.J., Stedtfeld, T.M., Chai, B., Cole, J.R., Hashsham, S.A., Tiedje, J.M., 2012. In-feed antibiotic effects on the swine intestinal microbiome. *Proc. Natl. Acad. Sci. U. S. A.* 109, 1691–1696.
- Louca, S., Parfrey, L.W., Doebeli, M., 2016. Decoupling function and taxonomy in the global ocean microbiome. *Science* 353, 1272–1277.
- Marti, E., Variatza, E., Balcazar, J.L., 2014. The role of aquatic ecosystems as reservoirs of antibiotic resistance. *Trends Microbiol.* 22, 36–41.
- Muurinen, J., Karkman, A., Virta, M., 2017. High throughput method for analyzing antibiotic resistance genes in wastewater treatment plants. In: Keen, P.L., Fugère, R. (Eds.), *Antimicrobial Resistance in Wastewater Treatment Processes*. John Wiley & Sons, Inc., New York, USA.
- Ouyang, W.Y., Huang, F.Y., Zhao, Y., Li, H., Su, J.Q., 2015. Increased levels of antibiotic resistance in urban stream of Jiulongjiang River, China. *Appl. Microbiol. Biotechnol.* 99, 5697–5707.
- Pehrsson, E.C., Tsukayama, P., Patel, S., Mejía Bautista, M., Sosasoto, G., Navarrete, K.M., Calderon, M., Cabrera, L., Hoyosarango, W., Bertoli, M.T., 2016. Interconnected microbiomes and resistomes in low-income human habitats. *Nature* 533, 212–216.
- Piddock, L.J., 2006. Multidrug-resistance efflux pumps - not just for resistance. *Nat. Rev. Microbiol.* 4, 629–636.
- Pruden, A., 2014. Balancing water sustainability and public health goals in the face of growing concerns about antibiotic resistance. *Environ. Sci. Technol.* 48, 5–14.
- Qiao, M., Ying, G.G., Singer, A.C., Zhu, Y.G., 2018. Review of antibiotic resistance in China and its environment. *Environ. Int.* 110, 160–172.
- R Core Team, 2015. *R: A Language and Environment for Statistical Computing*. R Foundation for Statistical Computing. <http://www.R-project.org>.
- Raes, J., Letunic, I., Yamada, T., Jensen, L.J., Bork, P., 2011. Toward molecular trait-based ecology through integration of biogeochemical, geographical and metagenomic data. *Mol. Syst. Biol.* 7, 473.
- Rodríguez-Mozaz, S., Chamorro, S., Marti, E., Huerta, B., Gros, M., Sánchez-Melsió, A., Borrego, C.M., Barceló, D., Balcazar, J.L., 2015. Occurrence of antibiotics and antibiotic resistance genes in hospital and urban wastewaters and their impact on the receiving river. *Water Res.* 69, 234–242.
- Shoemaker, N.B., Vlamakis, H., Hayes, K., Salyers, A.A., 2001. Evidence for extensive resistance gene transfer among bacteroides spp. and among Bacteroides and other genera in the human colon. *Appl. Environ. Microbiol.* 67, 561–568.
- Su, J.Q., Wei, B., Ou-Yang, W.Y., Huang, F.Y., Zhao, Y., Xu, H.J., Zhu, Y.G., 2015. Antibiotic resistome and its association with bacterial communities during sewage sludge composting. *Environ. Sci. Technol.* 49, 7356–7363.
- Takara Bio USA, 2017. *SmartChip™ RealTime PCR Cycler/SmartChip qPCR Software User Manual*. http://takarabio.com/xxclt_jbcGetAttachment.jsp?cltemId=137365&minisite=10022§Id=30054.
- Violle, C., Reich, P.B., Pacala, S.W., Enquist, B.J., Kattge, J., 2014. The emergence and promise of functional biogeography. *Proc. Natl. Acad. Sci. U. S. A.* 111, 13690–13696.
- Wang, F.H., Qiao, M., Su, J.Q., Chen, Z., Zhou, X., Zhu, Y.G., 2014. High throughput profiling of antibiotic resistance genes in urban park soils with reclaimed water irrigation. *Appl. Environ. Microbiol.* 48, 9079–9085.
- Wiener, P., 1996. Experimental studies on the ecological role of antibiotic production in bacteria. *Ecol. Evol.* 10, 405–421.
- Xu, L., Ouyang, W., Qian, Y., Su, C., Su, J., Chen, H., 2016. High-throughput profiling of antibiotic resistance genes in drinking water treatment plants and distribution systems. *Environ. Pollut.* 213, 119–126.
- Zhang, X.X., Zhang, T., Fang, H.H.P., 2009. Antibiotic resistance genes in water environment. *Appl. Microbiol. Biotechnol.* 82, 397–414.
- Zhang, X., Zhang, D., Zhang, H., Luo, Z., Yan, C., 2012. Occurrence, distribution, and seasonal variation of estrogenic compounds and antibiotic residues in Jiulongjiang River, South China. *Environ. Sci. Pollut. Res.* 19, 1392–1404.
- Zhang, S., Lv, L., Zhang, Y., Zhang, H., Yu, X., Zhang, S., 2013. Occurrence and variations of five classes of antibiotic resistance genes along the Jiulong River in Southeast China. *J. Environ. Biol.* 34, 345–351.
- Zhang, Q.Q., Ying, G.G., Pan, C.G., Liu, Y.S., Zhao, J.L., 2015. Comprehensive evaluation of antibiotics emission and fate in the river basins of China: source analysis, multimedia modeling, and linkage to bacterial resistance. *Environ. Sci. Technol.* 49, 6772–6782.
- Zheng, J., Gao, R., Wei, Y., Chen, T., Fan, J., Zhou, Z., Makimilua, T.B., Jiao, Y., Chen, H., 2017. High-throughput profiling and analysis of antibiotic resistance genes in east Tiaoxi River, China. *Environ. Pollut.* 230, 648–654.
- Zhou, Z.C., Zheng, J., Wei, Y.Y., Chen, T., Dahlgren, R.A., Shang, X., Chen, H., 2017. Antibiotic resistance genes in an urban river as impacted by bacterial community and physicochemical parameters. *Environ. Sci. Pollut. Res.* 24, 23753–23762.
- Zhu, B., Chen, Q., Chen, S., Zhu, Y.G., 2017. Does organically produced lettuce harbor higher abundance of antibiotic resistance genes than conventionally produced? *Environ. Int.* 98, 152–159.
- Zhu, Y.G., Johnson, T.A., Su, J.Q., Qiao, M., Guo, G.X., Stedtfeld, R.D., Hashsham, S.A., Tiedje, J.M., 2013. Diverse and abundant antibiotic resistance genes in Chinese swine farms. *Proc. Natl. Acad. Sci. U. S. A.* 110, 3435–3440.
- Zhu, Y.G., Gillings, M., Simonet, P., Stekel, D., Banwart, S., Penuelas, J., 2017a. Microbial mass movements. *Science* 357, 1099–1100.
- Zhu, Y.G., Reid, B.J., Meharg, A.A., Banwart, S.A., Fu, B.J., 2017b. Optimizing Peri-Urban ecosystems (PURE) to re-couple urban-rural symbiosis. *Sci. Total Environ.* 586, 1085–1090.
- Zhu, Y.G., Zhao, Y., Li, B., Huang, C.L., Zhang, S.Y., Yu, S., Chen, Y.S., Zhang, T., Gillings, M.R., Su, J.Q., 2017c. Continental-scale pollution of estuaries with antibiotic resistance genes. *Nat. Microbiol.* 2, 16270.