Check for updates

OPEN ACCESS

EDITED BY Jie Li, Chinese Academy of Sciences, China

REVIEWED BY Jesus Garcia-Gil, University of Girona, Spain Prabhugouda Siriyappagouder, Nord University, Norway

*CORRESPONDENCE Hao Du ⊠ duhao@yfi.ac.cn

[†]These authors have contributed equally to this work

RECEIVED 13 September 2023 ACCEPTED 20 December 2023 PUBLISHED 11 January 2024

CITATION

Yang H, Zhong J, Leng X, Wu J, Cheng P, Shen L, Wu J, Li P and Du H (2024) Effectiveness assessment of using water environmental microHI to predict the health status of wild fish. *Front. Microbiol.* 14:1293342. doi: 10.3389/fmicb.2023.1293342

COPYRIGHT

© 2024 Yang, Zhong, Leng, Wu, Cheng, Shen, Wu, Li and Du. This is an open-access article distributed under the terms of the Creative Commons Attribution License (CC BY). The use, distribution or reproduction in other forums is permitted, provided the original author(s) and the copyright owner(s) are credited and that the original publication in this journal is cited, in accordance with accepted academic practice. No use, distribution or reproduction is permitted which does not comply with these terms.

Effectiveness assessment of using water environmental microHI to predict the health status of wild fish

Haile Yang[†], Jia Zhong[†], Xiaoqian Leng, Jinming Wu, Peilin Cheng, Li Shen, Jinping Wu, Pengcheng Li and Hao Du*

Key Laboratory of Freshwater Biodiversity Conservation, Ministry of Agriculture and Rural Affairs, Yangtze River Fisheries Research Institute, Chinese Academy of Fishery Sciences, Wuhan, China

Aquatic wildlife health assessment is critically important for aquatic wildlife conservation. However, the health assessment of aquatic wildlife (especially aquatic wild animals) is difficult and often accompanied by invasive survey activities and delayed observability. As there is growing evidence that aquatic environmental microbiota could impact the health status of aquatic animals by influencing their symbiotic microbiota, we propose a non-invasive method to monitor the health status of wild aquatic animals using the environmental microbiota health index (microHI). However, it is unknown whether this method is effective for different ecotype groups of aquatic wild animals. To answer this question, we took a case study in the middle Yangtze River and studied the water environmental microbiota and fish gut microbiota at the fish community level, population level, and ecotype level. The results showed that the gut microHI of the healthy group was higher than that of the unhealthy group at the community and population levels, and the overall gut microHI was positively correlated with the water environmental microHI, whereas the baseline gut microHI was species-specific. Integrating these variations in four ecotype groups (filterfeeding, scraper-feeding, omnivorous, and carnivorous), only the gut microHI of the carnivorous group positively correlated with water environmental microHI. Alcaligenaceae, Enterobacteriaceae, and Achromobacter were the most abundant groups with health-negative-impacting phenotypes, had high positive correlations between gut sample group and environment sample group, and had significantly higher abundance in unhealthy groups than in healthy groups of carnivorous, filter-feeding, and scraper-feeding ecotypes. Therefore, using water environmental microHI to indicate the health status of wild fish is effective at the community level, is effective just for carnivorous fish at the ecotype level. In the middle Yangtze River, Alcaligenaceae, Enterobacteriaceae (family level), and Achromobacter (genus level) were the key water environmental microbial groups that potentially impacted wild fish health status. Of course, more data and research that test the current hypothesis and conclusion are encouraged.

KEYWORDS

gut microbiota, environmental microbiota, microbial phenotypes, microbiota health index (microHI), aquatic wildlife, health monitoring

Introduction

Aquatic environmental microbiota could impact the health status of aquatic organisms by influencing their symbiotic microbiota (Guivier et al., 2020; Sehnal et al., 2021; Yang et al., 2022). It is well known that humans co-evolve with their symbiotic microbiota, which plays a critical role in human health (Adak and Khan, 2019; Fassarella et al., 2021; Reynoso-García et al., 2022). There is a similar situation in aquatic organisms, and the symbiotic microbiota plays a more critical role in aquatic organisms (Cabillon and Lazado, 2019; Guivier et al., 2020; Lindsay et al., 2020; Sehnal et al., 2021; Luna et al., 2022). In aquatic animals, environmental microbiota could strongly influence the symbiotic microbiota in the skin, gut, gills, and so on (Reinhart et al., 2019; Guivier et al., 2020; Kuang et al., 2020; Elsheshtawy et al., 2021; Yang et al., 2022), in which the gut microbiota has relatively more stable community components and structures (Guivier et al., 2020), weaker influences from environmental microbiota (Reinhart et al., 2019), and higher impacts on the host's health (Sehnal et al., 2021) than other symbiotic microbiota (such as the symbiotic microbiota in skin, gills, and so on). Therefore, the gut microbiota is the key point for studying the processes of aquatic environmental microbiota impacting the health status of aquatic animals.

The health status of aquatic animals may be indicated by environmental microHI (Yang et al., 2023). The health of an individual always refers to the individual not being disturbed or hurt. The health of aquatic animals could be indicated by normal gut microbiota (Sun et al., 2020; Tong et al., 2020) and a low abundance of health-negativeimpacting microbes (Qiao et al., 2019; Himani et al., 2020). The microHI (microbiota health index) is an abbreviation of the microbiota health index, which is defined as the average abundance ratio of microbes without three health-negative-impacting microbial phenotypes (Eq. 1) (Yang et al., 2023). These three health-negativeimpacting microbial phenotypes, respectively, are potential pathogenicity, mobile element content, and oxidative stress tolerance (Yang et al., 2023). The abundance ratio of microbes with potential pathogenicity indicates the potential risk of disease in the host (Himani et al., 2020). The abundance ratio of microbes that contain mobile elements (such as antibiotic resistance genes) reflects the potential risk of antibiotic pollution (Sáenz et al., 2019) or potential threats to host health (Himani et al., 2020). The abundance ratio of microbes with oxidative stress tolerance indicates the potential risk of host gut tissue damage (Qiao et al., 2019; Solomando et al., 2020). The abundance of microbes with these microbial phenotypes can be predicted and quantified based on high-throughput sequencing technology and the BugBase database (https://bugbase.cs.umn.edu/index.html) (Yang et al., 2023). Previous work identified that an individual with a low microHI gut microbiota is more likely to have a disordered microbiota and is more likely obtained from the water with a low aquatic environmental microHI (Yang et al., 2023). As a disordered microbiota always indicates an unhealthy host (Sun et al., 2020; Tong et al., 2020), the aquatic environmental microHI has the potential to indicate the health status of aquatic animals (Yang et al., 2023).

microHI =
$$\frac{\left[\left(1-P_m\right)+\left(1-P_o\right)+\left(1-P_p\right)\right]}{3}$$
(1)

where *microHI* denotes the microbiota health index; P_m denotes the abundance ratio (percent) of microbes containing mobile

elements; P_o denotes the abundance ratio (percent) of microbes with oxidative stress tolerance; and P_p denotes the abundance ratio (percent) of microbes with potential pathogenicity (Yang et al., 2023).

There is species variation in the relationship between aquatic environmental microbiota and gut microbiota in aquatic animals (Sun and Xu, 2021; Yang et al., 2022). As the gut microbiota is driven by both host-related deterministic selection and the introduction of neutral microbes from the environment (Ringø et al., 2016; Legrand et al., 2020), and the majority of the introduced gut microbiota is derived from the feeding habits of the host (Yang et al., 2022), the correlation between environmental microbiota and gut microbiota is ecotype-specific and even species-specific (Yang et al., 2022). In a previous work, the possibility of using aquatic environmental microHI to indicate the health status of aquatic animals was preliminarily identified at the community level (Yang et al., 2023). Is there any species or ecotype variation in the effectiveness of using water environmental microHI to indicate the health status of aquatic animals? Based on the fact that the gut microbiota is driven by hostrelated deterministic selection and the introduction of neutral microbes from the environment (Ringø et al., 2016; Legrand et al., 2020), and that the microHI is calculated based on the microbiota community structure (Yang et al., 2023), we hypothesized that (1) the base value of the gut microHI is species-specific, and (2) the correlation between the water environmental microHI and the gut microHI has species variation and ecotype variation. In other words, we hypothesized that there is effectiveness variation in the use of water environmental microHI to indicate the health status of aquatic wild animals at the population level and ecotype level.

To identify the effectiveness variation, we took a case study to research the water environmental microbiota and fish gut microbiota at the fish community level, population level, and ecotype level. As the third longest river in the world with rich aquatic biodiversity, the Yangtze River has received much attention in wildlife conservation (Chen et al., 2020; Zhang et al., 2020). Proposing a convenient method to monitor and indicate the health status of wildlife in the Yangtze River is urgently needed (Yang et al., 2023). In the current work, we collected a data set of the fish gut microbiota and water environmental microbiota in the middle Yangtze River. Using this data set, we further verified that the water environmental microHI could be used to indicate the health of aquatic animals at the community level. Based on the previous data set and the new data set, we tested the effectiveness of using water environmental microHI to indicate the health of wild fish at the population level and ecotype level, and we identified the key microbial groups that influenced the health status of wild fish. The current study clarifies the effectiveness of using water environmental microHI to indicate the health status of wild fish and proves a non-invasive method to monitor the health status of wild fish without harming the individuals, which would offer a new way of understanding and managing the health status of aquatic wild animals and provide a new tool for the toolkit of aquatic wildlife conservation.

Materials and methods

Sample collection

From 27 June 2022 to 14 July 2022, we collected gut samples of adult wild fish (Table 1) and water environmental DNA samples at the Jiang'an section of the middle Yangtze River (N 30°38'52", E

Group label	Species	Order	Family	Ecotype	Diet	Number of samples
L.C.	Leiocassis crassilabris	Siluriformes	Bagridae	carnivorous	Oligochaetes, little fish, small mollusks, shrimp	31
L.L.	Leiocassis longirostris	Siluriformes	Bagridae	carnivorous	Little fish, shrimp, aquatic insects	41
P.V.	Pelteobagrus vachelli	Siluriformes	Bagridae	carnivorous	Aquatic insects, shrimp, oligochaetes, little fish, small mollusks	4
P.N.	Pelteobagrus nitidus	Siluriformes	Bagridae	carnivorous	Aquatic insects, shrimp, oligochaetes, little fish	11
P.F.	Pelteobagrus fulvidraco	Siluriformes	Bagridae	carnivorous	Aquatic insects, shrimp, little fish	16
P.T.	Pseudobagrus tenuis	Siluriformes	Bagridae	carnivorous	Aquatic insects, mollusks, oligochaetes, crustaceans, little fish	6
S.A.	Silurus asotus	Siluriformes	Siluridae	carnivorous	Shrimp, little fish	
S.M.	Silurus meridionalis	Siluriformes	Siluridae	carnivorous	Shrimp, little fish	2
H.M.	Hypophthalmichthys molitrix	Cypriniformes	Cyprinidae	filter-feeding	Phytoplankton, zooplankton	4
A.N.	Aristichthys nobilis	Cypriniformes	Cyprinidae	filter-feeding	Zooplankton, phytoplankton	5
С.Н.	Coreius heterodon	Cypriniformes	Cyprinidae	omnivorous	Small mollusks, fish eggs and larvae, phytoclasts	35
X.A.	Xenocypris argentea	Cypriniformes	Cyprinidae	scraper-feeding	Bottom-attached algae, phytoclasts	14
S.C.	Siniperca chuatsi	Perciformes	Serranidae	carnivorous	Fish, shrimp	9
S.K.	Siniperca kneri	Perciformes	Serranidae	carnivorous	Fish, shrimp	5
C.B.	Coilia brachygnathus	Clupeiformes	Engraulidae	carnivorous	Little fish, shrimp	31

TABLE 1 Wild fish species in the current study.

114°20'58", located in Wuhan City, Hubei Province, People's Republic of China), along with a conventional Yangtze River fishery resources investigation. Wild fish were caught using a floating gill net, transported to the laboratory in an ice bath, and then dissected. Approximately 0.5g of gut sample, including both tissues and contents, was aseptically taken from each individual fish caught. A 1.5 L surface water sample was collected each day (including two transection sample; more details in Supplementary Table S1) using a 1.5 L sterilized bottle (rinsed three times with sampling water) and transported to the laboratory in an ice bath. To obtain eDNA samples, each water sample was filtered using a 0.2-µm membrane filter (JinTeng, Tianjin, PRC) with purified water as a negative control. All samples (including gut samples and eDNA samples) were transported on dry ice and then stored at -80°C until DNA extraction. The sampling procedures in the current study were similar to the previous work (Yang et al., 2023).

Microbiome sequencing

The fish gut microbiota and water environmental microbiota were analyzed using 16S rRNA metabarcoding (primers 338F/806R). The sequencing technical details followed those in the previous work (Yang et al., 2020, 2021). Microbial DNA was extracted from gut samples and water eDNA samples using an E.Z.N.A.[®] Stool DNA Kit (Omega BioTek, Norcross, GA, United States) according to the manufacturer's protocols. Then, the final DNA concentration and purity were determined using a NanoDrop 2000 UV-vis spectrophotometer (Thermo Fisher Scientific, Wilmington, United States), and DNA quality was checked by 1% agarose gel electrophoresis. The V3-V4 hypervariable regions of the bacterial 16S gene were amplified with the primers rRNA 338F (5'-ACTCCTACGGGAGGCAGCAG-3') and 806R (5'-GGACT ACHVGGGTWTCTAAT-3') by using a PCR thermocycler system (GeneAmp 9,700, ABI, United States). The PCRs were performed in triplicate 20-µl mixtures containing $4 \mu L$ of $5 \times$ FastPfu Buffer, $2 \mu L$ of 2.5 mM dNTPs, 0.8 µL of each primer (5 µM), 0.4 µL of FastPfu Polymerase, and 10 ng of template DNA. The PCRs were conducted using the following program: 3 min of denaturation at 95°C; 27 cycles of 30s at 95°C, 30s for annealing at 55°C, and 45s for elongation at 72°C; and a final extension at 72°C for 10 min. The quality of PCR products was tested using 2% agarose gel electrophoresis. Then, the resulting PCR products were extracted from a 2% agarose gel, further purified using an AxyPrep DNA Gel Extraction Kit (Axygen Biosciences, Union City, CA, United States), and quantified using QuantiFluorTM-ST (Promega, United States) according to the manufacturer's protocol. The standard Illumina tags were added to PCR products using another PCR program. Then the tagged PCR products were extracted, purified, and checked. The single-stranded DNA was prepared. Purified amplicons were pooled in equimolar amounts and subjected to paired-end sequencing $(2 \times 300 \text{ bp})$ on an Illumina MiSeq platform (Illumina, San Diego, CA, United States) according to standard protocols. All samples were extracted and sequenced by Shanghai Majorbio Biopharm Technology Co., Ltd. (Shanghai, China).

The data for the current study could be obtained from the China National GeneBank DataBase (CNGBdb, https://db.cngb.org/) with the accession numbers CNP0002410 and CNP0002411, which are,

respectively, continually updated aquatic wildlife gut microbiota data sets and continually updated aquatic environmental DNA data sets.

Statistical analysis

The sequence data of each sample was analyzed on the Majorbio Cloud Platform (https://cloud.majorbio.com/). Raw fastq files were demultiplexed, quality-filtered by Trimmomatic (http://www. usadellab.org/cms/index.php?page=trimmomatic) and merged by FLASH (https://ccb.jhu.edu/software/FLASH/index.shtml). Operational taxonomic units (OTUs) were clustered with a 97% similarity cutoff using USEARCH7-uparse (http://drive5.com/ uparse/), and chimeric sequences were identified and removed using (http://www.drive5.com/uchime/uchime_download. UCHIME html). The abundance of each OTU was analyzed using Usearch (http://www.drive5.com/usearch/). The taxonomy of each OTU was analyzed using the RDP Classifier Bayesian algorithm (http:// sourceforge.net/projects/rdp-classifier/) against the database of silva138/16s (http://www.arb-silva.de) with a confidence threshold of 75%. The phenotypes of each taxon were predicted using the phenotype prediction tool of BugBase (https://bugbase.cs.umn.edu/ index.html), in which the unclassified sequences (sequences that cannot be annotated as species) were ignored in the analysis. The abundance ratios of microbes with the phenotypes of potential pathogenicity, mobile element content, and oxidative stress tolerance in each sample were obtained. Then, the microHI of each sample was calculated following Eq. 1. To identify the samples with disordered gut microbiota, hierarchical clustering (along with the column diagram of microbiota) of fish gut microbial samples was conducted at the microbial species, genus, and family levels based on the Bray-Curtis distance algorithm and the average linkage method (using Qiime http://qiime.org/install/index.html). The normal branches could be recognized using a definite species and even species with similar ecotypes, as we assumed that most individuals have a normal gut microbiota (Yang et al., 2022). The samples clustered in abnormal branches (out of normal branches) could be identified as disordered ones, which always indicates unhealthy hosts (Tong et al., 2020). The statistical analysis processes in the current study followed our previous work (Yang et al., 2023).

The current water environmental microbiota and fish gut microbiota data obtained from the Jiang'an section in June 2022 were analyzed to verify the relationship between gut microHI and disordered gut microbiota and then were combined with the data that was collected in 2020 at the sampling sections of Jiayu, Xinzhou, and Hukou in a previous study (Yang et al., 2023) to verify the correlation between gut microHI and environmental microHI at the fish community level. The data that was collected in 2020 includes 219 adult gut samples (29 from the Jiayu section, 156 from the Xinzhou section, and 34 from the Hukou section) from 11 species of fish and 43 water samples from two water environment groups (13 from the Xinzhou section, 30 from 30 sampling transects of the middle Yangtze River). Then, the gut microHI base value (average gut microHI of the individuals with normal gut microbiota) of each fish species was calculated. According to a pairwise comparison, the correlation relationship between gut microHI and disordered gut microbiota (the pairwise comparison of average gut microHIs from the samples with disordered gut microbiota and the samples with normal gut microbiota) and the correlation relationship between host growth condition (fullness) (Fulton's condition, Eq. 2) (Bolger and Connolly, 1989; Hayes et al., 1995) and disordered gut microbiota (the pairwise comparison of average fullness between samples with disordered gut microbiota and samples with normal gut microbiota) were identified at the fish population level. A linear regression identified a correlation relationship between the gut microHI and environmental microHI at both the fish population level and the fish ecotype level. The gut microHI of each fish ecotype was adjusted using the fish community structure. In other words, the gut microHI of each fish ecotype was a weighted average of the gut microHI, where the individual percentage of each species in the fish community is the weighting factor. The key groups determining the microHI (with the phenotypes of potential pathogenicity, mobile element content, and oxidative stress tolerance) were identified using the non-parametric factorial Kruskal-Wallis (KW) sum-rank test and LDA (LDA \geq 3).

$$K = \frac{m}{L^3} \times 100 \tag{2}$$

where *K* denotes Fulton's condition (fullness) of an individual; *m* denotes the weight of an individual; and *L* denotes the body length of an individual (Bolger and Connolly, 1989; Hayes et al., 1995).

Results

The characteristics of the samples' sequences

In the 214 adult fish gut samples from 14 species and 30 water environmental eDNA samples, we obtained 18,260,434 clean sequences, which were clustered into 15,247 OTUs and were identified as 1 kingdom, 65 phyla, 195 classes, 459 orders, 804 families, 1884 genera, and 2,409 species. The rarefaction curve (sobs & chaos, Supplementary Figures S1–S2) showed that the sequence depth was almost sufficient.

The abundance ratios of microbes with the phenotype of potential pathogenicity, mobile element content, or oxidative stress tolerance were variable in different samples and species (Supplementary Figure S3). More than half of the adult fish gut samples had a low (lower than 40%) relative abundance of the microbes with the phenotype of Potentially Pathogenic, a low (lower than 30%) relative abundance of the microbes with the phenotype of Contains Mobile Elements, and a low (lower than 20%) relative abundance of microbes with the phenotype of Stress Tolerant. Most of Coilia brachygnathus had a high (>50%) relative abundance of the microbes with the phenotypes of Potentially Pathogenic, Contains Mobile Elements, and Stress Tolerant. In 30 water environmental eDNA samples, the relative abundance of the microbes with the phenotype of Potentially Pathogenic mainly (>93%) ranges from 10 to 30%; the relative abundance of the microbes with the phenotype of Contains Mobile Elements mainly (90%) ranges from 50 to 70%; and the relative abundance of the microbes with the phenotype of Stress Tolerant mainly (>93%) is lower than 20%.

The relationship among host health, gut microHI, and water environmental microHI at the community level

In the total of 214 fish gut microbial samples, 65 (30%) had disordered microbiota (unhealthy) (Supplementary Figures S4-S6). Approximately 66% of the fish gut microbial samples had a microHI of higher than 0.6, and approximately 15% of the fish gut microbial samples had a microHI of lower than 0.4. In the fish gut microbial samples with a microHI higher than 0.6, approximately 77% had normal microbiota (healthy). As the microHI increased, the probability of that the sample was captured from a healthy specimen also increased (Figure 1A). The average gut microHI of healthy individuals (0.71) was higher than that of unhealthy individuals (0.58). Comparing the microbial data from Jiang'an section in 2022 with the microbial data from Xinzhou section in 2020, the average environmental microHI of water samples is 0.71 (n = 30) in 2022, and 0.83 (n = 13) in 2020; the average gut microHI of all sampled individuals is 0.67 (*n* = 214) in 2022, and 0.68 (*n* = 156) in 2020; the average gut microHI of all sampled healthy fish individuals is 0.71 (*n* = 149) in 2022, and 0.73 (*n* = 115) in 2020; the percentage of healthy individuals was 69% in 2022 and 74% in 2020. Plotting the current diagram of the correlation between average fish gut microHI and average water environmental microHI with combined data (2020 and 2022), the positive correlation between the water environmental microHI and fish gut microHI at the community level was still maintained (Figure 1B).

The relationship among host health, gut microHI, and water environmental microHI at the population level

In the combined data of the fish gut microbiota from the Jiang'an section in 2022 and the Xinzhou section in 2020, 11 fish species with more than 10 individuals were analyzed. The average gut microHI of healthy individuals was generally higher than that of unhealthy individuals in each fish species (Figure 2A). The gut microHI of the healthy group was significantly higher than that of the unhealthy group in some fish species (with enough samples in both healthy and unhealthy groups) (Figure 2B). The average Fulton's condition of the healthy individuals was generally higher than the average Fulton's condition of the unhealthy individuals in the majority of fish species, although the difference in Fulton's condition was less than 9% (Figure 2C). As the water environmental microHI in 2020 was higher in 2020 than in 2022, the average gut microHI of the healthy fish group of a species was higher in 2020 than in 2022 for the majority of species (with three exceptions), although there was no significant difference between the sample groups in 2020 and 2022 (Figure 2D). The average gut microHI of healthy individuals in each species, which could be regarded as the base value of this species' gut microHI, was species-specific (Figure 2A, Supplementary Table S2) and influenced by environmental microbiota (Figure 2D). Moreover, the degree to which the average value of the gut microHI in the healthy fish group of a species was higher in 2020 than in 2022 was found to be speciesspecific (Figure 2D). In other words, the influence of environmental microbiota on the base value of the gut microHI was species-specific, although there was a random error observed (Figure 2D).

The relationship between gut microHI and water environmental microHI at the ecotype level

Considering the species-specific relationship between gut microHI and environmental microHI, the community-level fish gut microHI of each sampling location (Jiayu, Jiang'an, Xinzhou, and Hukou) was corrected using the fish community structure (based on the number of individuals, Supplementary Table S3) identified by our conventional Yangtze fishery resource investigation. The fish in the community were divided into four ecotypes: filter-feeding, scraperfeeding, omnivorous, and carnivorous (Table 1). The positive correlation between water environmental microHI and fish gut microHI at the community level was still maintained (Figure 3A),



FIGURE 1

The correlation between fish gut microHI and water environmental microHI. (A) Distribution of fish gut microHIs in the Jiang'an section of the Yangtze River (n = 214). (B) The correlation between fish gut microHI (n = 370) and water environmental microHI (n = 43).



FIGURE 2

The gut microHIs of each species at the population level. (A) The average gut microHIs of healthy (green) individuals and unhealthy (red) individuals for each fish species. (B) The gut microHI characters of healthy (_H) individuals and unhealthy (_U) individuals for each fish species. (C), The difference in values of average microHI and average fullness (Fulton's condition) between healthy and unhealthy groups of individuals. (D), The gut microHIs of healthy individuals for each fish species in 2020 and 2022. Eleven fish species groups: *C.B., Coilia brachygnathus; C.H., Coreius heterodon; H.M., Hypophthalmichthys molitrix L.C., Leiocassis crassilabris; (L).L., Leiocassis longirostris; (P).F., Pelteobagrus fulvidraco; P.N., Pelteobagrus nitidus; P.V., Pelteobagrus vachelli; (S).A., Silurus asotus; S.C., Siniperca chuatsi; (X).A., Xenocypris argentea. *p <0.05 and **p <0.01 indicate significant differences.*



FIGURE 3

The adjusted correlation between water environmental microHI and fish gut microHI at the community level and ecotype level. The overall community level (A), the ecotypes of carnivorous (B), scraper-feeding (C), filter-feeding (D), and omnivorous (E) fish groups.

while the positive correlation at the ecotype level was only maintained in the carnivorous fish group (Figures 3B-E).

The key groups that potentially impact the health status of wild fish from water environmental microbiota

When analyzing the fish gut microbiota and the water environmental microbiota, there were mainly 107 families and 106 genera that related to the phenotypes of potential pathogenicity, mobile element content, and oxidative stress tolerance (Supplementary Figures S7–S12). The most abundant groups with a negative impact on fish gut microHI were Alcaligenaceae, Clostridiaceae, Peptostreptococcaceae, Enterobacteriaceae, and Moraxellaceae at the family level, and *Achromobacter*, *Plesiomonas*, *Clostridium*, *Aeromonas*, and *Acinetobacter* at the genus level (Supplementary Figures S7–S12). A total of 26 families and 22 genera had high correlations ($r^2 > 0.6$) between the gut sample group and environment sample group, among which 17 families and 13 genera had positive correlations (slope>0.1), including Alcaligenaceae, Enterobacteriaceae, and *Achromobacter* (Supplementary Table S4).

Comparing the healthy groups and unhealthy groups of the four ecotypes (filter-feeding, scraper-feeding, omnivorous, and carnivorous), the percentage of families and genera that related to the phenotypes of potential pathogenicity, mobile element content, and oxidative stress tolerance was lower in the healthy group than in the unhealthy group (Supplementary Table S5). There were 51 families and 47 genera that had significant abundance differences between healthy and unhealthy groups (Supplementary Figures S13–S16 and Supplementary Table S5), among which 10 families and 6 genera showed positive correlations (slope > 0.1, $r^2 > 0.6$) between gut sample groups and environment sample groups in each ecotype (except the omnivorous ecotype) (Figure 4A and, Supplementary Tables S4–S5). Additionally, only two families (Alcaligenaceae, Enterobacteriaceae) and two genera (*Achromobacter, Streptococcus*) had significantly

higher abundance in the unhealthy groups than in the healthy groups of the carnivorous, filter-feeding, and scraper-feeding ecotypes (Figure 4B). The other families and genera did not have significantly higher abundances in unhealthy groups than in healthy groups.

Discussion

Gut microHI could indicate the health status of wild fish at both community and population levels

The homeostasis of the intestinal system occupies an important position in the regulation of the host immune system, health, and physiology (Su et al., 2018; Zhong et al., 2019). The metabolites of gut microbes can act as energy sources for cell metabolism, as promoters of development and immune function, and can prevent colonization by pathogenic microorganisms (Honda and Littman, 2012; Butt and Volkoff, 2019). The altered gut microbiota composition can affect fish health and is often associated with a diseased state (Sartor and Wu, 2017; Sun et al., 2020; Bozzi et al., 2021). Many studies have shown that the gut microbiota can be a monitoring indicator of multiple diseases (Xiong et al., 2019; Wang et al., 2022), abnormal physiological metabolism (Kraimi et al., 2019; Kim et al., 2020), and general health conditions (Borbón-García et al., 2017; Bozzi et al., 2021) in mammals, birds, fish, etc.

The gut microHI could reflect the microbial composition associated with tissue damage, risk of disease, or potential threats to host health (Sáenz et al., 2019; Himani et al., 2020). One of our previous studies found that, at the community level, most of the fish gut samples with low microHI were obtained from specimens with a disordered gut microbiota, which is always indicative of an unhealthy host, so the gut microHI could potentially be used to evaluate host health status (Yang et al., 2023). In the current study, the new data showed that, at the community level, (1) the individuals with normal gut microbiota were more likely to have a higher gut microHI than the



differentially abundant between the healthy (blue) and unhealthy (red) groups, as determined using the Kruskal-Wallis test. The LDA score (effect size indicates significant differences in bacterial taxa (LDA score > 3.0; alpha value p < 0.05). (**B**), Relative abundance of indicator taxa in the healthy (blue) and unhealthy (red) groups. *p < 0.05 and **p < 0.01 indicate significant differences.

10.3389/fmicb.2023.1293342

individuals with disordered gut microbiota, and (2) the individuals with a higher gut microHI were more likely to have normal gut microbiota than the individuals with a lower gut microHI (Figure 1A). In addition, the average gut microHI of healthy individuals was higher than that of unhealthy individuals. Thus, the current data support the idea that the gut microHI could be used to indicate host health status at the community level.

In the 11 fish species analyzed, the average Fulton's condition of healthy individuals was found to be generally higher than that of unhealthy individuals in the majority of fish species, although the difference in Fulton's condition was small (Figure 2C). Generally speaking, in a given species, a higher Fulton's condition in an adult fish individual always indicates a more healthy status (Nash et al., 2006), although the Fulton's condition index has some limitations (Bolger and Connolly, 1989; Hayes et al., 1995). Therefore, the delineation of a healthy group based on normal gut microbiota is supported by the growth status of individuals as indicated by Fulton's condition. Moreover, the average gut microHI of healthy individuals was generally higher than that of unhealthy individuals in each species, although there was individual variation (Figures 2A–B). This suggests that the gut microHI can be used to indicate host health status at the population level.

The correlation between gut microHI and water environmental microHI is positive at the community level, but variable at the population and ecotype levels

Environmental microbiota (Zhang et al., 2021; Zhou et al., 2021), diet composition (Gajardo et al., 2017; Wilkes Walburn et al., 2019), host genetics (Zhang et al., 2019; Callens et al., 2020), and environmental pollutants (Li et al., 2022; Zhang et al., 2023) have been demonstrated to play important roles in defining the composition of the gut microbiota of the host. In aquatic animals, the introduction of environmental microbes is known to be an extremely important force in constructing the gut microbiota (Zhang et al., 2021; Zhou et al., 2021; Yang et al., 2022). One of our previous studies found that there were low fish gut microHIs in the Yangtze River sections with low aquatic environmental microHI (Yang et al., 2023), which is consistent with the current results of a positive correlation between the gut microHI and the water environmental microHI at the community level. The current results support the idea that the water environmental microHI could be used to indicate the gut microHI at the community level. Moreover, the adjusted average fish gut microHI of the sampling locations still maintained a positive correlation between the water environmental microHI and fish gut microHI (Figure 3A). Therefore, considering that the gut microHI could indicate the health status of wild fish and that there is a positive correlation between the water environmental microHI and the fish gut microHI at the community level, the water environmental microHI could serve as a useful indicator of the health status of the wild fish community unit.

The base values of the gut microHI for each species, indicated by the mean gut microHI of healthy individuals, were species-specific at the population level (Figure 2A and Supplementary Table S2), and the water environmental effects were similarly species-specific (Figure 2D). These species variations are impacted by the combined effects of host-genetic deterministic selection and the introduction of neutral microbes (Ley et al., 2011; Ringø et al., 2016; Legrand et al., 2020). The host-genetic deterministic selection depends on the host digestive system structures (Yan et al., 2016; Wilkes Walburn et al., 2019; Xiao et al., 2021). The introduction of neutral microbes depends on the habitats and diet conditions of the hosts (Ringø et al., 2016; Kashinskaya et al., 2018; Sun et al., 2019; Heys et al., 2020). Of course, part of the variations in the current work may have been caused by individual random errors. The correlation between fish gut microHI and water environmental microHI is variable at the population level.

Host-mediated environmental factors such as diet, lifestyle, xenobiotic exposure, and medications can induce substantial shifts in microbiome composition (Yatsunenko et al., 2012; Schroeder et al., 2016). Multiple studies have indicated associations between longterm diet and host microbiota (Turnbaugh et al., 2009; Albenberg and Wu, 2014; Cresci and Bawden, 2015). Diet was also the main force influencing gut microbial diversity (Falony et al., 2016; Turpin et al., 2016; Zhernakova et al., 2016). However, it only explained a small percentage of the variation in the microbiota. Thus, given the importance of diet on the gut microbiota composition, we combined the populations into four ecotypes (filter-feeding, scraper-feeding, omnivorous, and carnivorous, Table 1). The results showed that only the carnivorous group maintained a positive correlation between gut microHI and water environmental microHI (Figures 3B-E), whereas the filter-feeding, scraper-feeding, and omnivorous groups did not. Perhaps driven by the extensive and longer intestinal system (filterfeeding, 5.0-6.0 times body length; scraper-feeding, 1.5-5.0 times body length; omnivorous, 0.9-1.1 times body length) (Chen, 1998), there was a relatively stable gut microbiota structure in filter-feeding, scraper-feeding, and omnivorous species that was seldom or weakly influenced by environmental microbiota (Figures 3C-E). In contrast, the short (0.3–0.7 times body length) intestinal system of carnivorous fish (Zheng, 1987; Zhu et al., 1999) was strongly influenced by environmental microbiota; therefore, there was a positive correlation between gut microHI and water environmental microHI (Figure 3B). Of course, it is possible that the effects of sediment environmental microbiota on the gut microbiota structure of the scraper-feeding and omnivorous groups (Zhang et al., 2021; Zhou et al., 2021; Yang et al., 2022) caused the weak correlation between the gut microHI of the scraper-feeding omnivorous species and and water environmental microHI.

Alcaligenaceae, Enterobacteriaceae (family level), and Achromobacter (genus level) are the key groups that may impact the health status of wild fish from environmental microbiota in water

The groups that have a negative influence on the health status of wild fish are general in both the gut microbiota and the environmental microbiota (Al-Hisnawi et al., 2016). The aquatic habitat, as a living matrix, supports various microbes (Legrand et al., 2020; Sehnal et al., 2021). The gut of aquatic wild animals, due to sharing a large amount of microbes with the environment (Yang et al., 2022), also harbors various microbes. In the current results, there were mainly 107 families and 106 genera that related to the phenotypes of potential pathogenicity, mobile element content, and oxidative stress tolerance, and a large amount of these groups were present in most samples

(including from both healthy and unhealthy individuals) (Supplementary Figures S7–S12).

Few of the groups showed the possibility of originating from water environmental microbiota and potentially impacting the health status of wild fish. In the groups that related to the phenotypes of potential pathogenicity, mobile element content, and oxidative stress tolerance, the abundance of 26 families and 22 genera in gut sample group was strongly correlated with that in environment sample group ($r^2 > 0.6$), and the abundance correlations were positive for 17 families and 13 genera (slope > 0.1) (Supplementary Table S4). Comparing the healthy groups and unhealthy groups of four ecotypes, 10 families and 6 genera (part of previous 17 families and 13 genera) had significant abundance difference between healthy groups and unhealthy groups (Supplementary Figures S13-S16 and Supplementary Table S5), and only two families (Alcaligenaceae, Enterobacteriaceae) and two genera (Achromobacter, Streptococcus) had significantly high abundance in the unhealthy group of the carnivorous or filter-feeding ecotypes. Furthermore, Alcaligenaceae, Enterobacteriaceae (family level), and Achromobacter (genus level) were found to be the most abundant groups among the 107 families and 106 genera. Alcaligenaceae, including the genus Achromobacter, as opportunistic pathogens, are significantly more abundant in hosts with certain diseases (Swenson and Sadikot, 2015; Pan et al., 2020), and can be a bacterial bio-indicator of environmental pollution (Zolkefli et al., 2020). Enterobacteriaceae are an important normal flora in the intestine. However, some species can become opportunistic pathogens and cause infections after the host's immunity declines, resulting in a variety of diseases (Satlin and Walsh, 2017; Skuja et al., 2018; Pisani et al., 2021). In the current results, Alcaligenaceae, including the genus Achromobacter, are associated with potential pathogenicity and oxidative stress tolerance. Enterobacteriaceae are related to potential pathogenicity, mobile elements, and oxidative stress tolerance. Therefore, we believe that Alcaligenaceae, Enterobacteriaceae (family level), and Achromobacter (genus level) are the key water environmental microbial groups that potentially impact the health status of wild fish in the middle Yangtze River.

A weighted microHI, based on scientific rationale, should be a better indicator of the health status of wild fish

In one of our previous works, microHI was defined as the unweighted average abundance ratio of microbes without three health-negative-impacting microbial phenotypes, i.e., potential pathogenicity, mobile element content, and oxidative stress tolerance (Eq. 1) (Yang et al., 2023). MicroHI has shown its ability to evaluate the health status of wild fish and the possibility of using water environmental microbiota to indicate the health status of wild fish (Yang et al., 2023). However, the current unweighted microHI is only a primary indicator (microHI-1.0). The correlation between fish gut microHI and the delineation of healthy and unhealthy individuals in a species is not very accurate (little exceptions) and strong (high r2) (Figure 1A). The correlation between fish gut microHI and environmental microHI is also not very accurate (little exceptions) and strong (high r2) (Figure 3). Perhaps a weighted microHI based on scientific evidence would be the best way forward.

Although Potentially Pathogenic, Contains Mobile Elements, and Stress Tolerant are identified as health-negative-impacting

microbial phenotypes, in practice, the microorganism with these phenotypes could be beneficial if it had low abundance and low health-negative-impacting (Wiles and Guillemin, 2019). Using the microHI-1.0, this situation cannot be taken into account when assessing the health status of fish using the fish gut microHI. Then, the effectiveness of using water environmental microHI to predict the health status of wild fish could also be impacted. Therefore, an updated microHI (microHI-2.0) that considers these influences might be better. To obtain the microHI-2.0, sufficient scientific data and knowledge are needed to determine the weighting factors and adjustment parameters. More studies in this field are necessary. We hope that microHI-2.0 can be realized in the future.

Conclusion

The current work showed that the gut microHI could indicate the community health status of wild fish and had a positive correlation with water environmental microHI. In other words, the current results support the idea that the water environmental microHI could be used to indicate the health status of wild fish at the community level. Moreover, the current study indicated that the gut microHI could indicate the population health status of wild fish, although it showed a variable correlation with water environmental microHI. When fish species were grouped into four ecotypes, only carnivorous fish retained the positive correlation between fish gut microHI and water environmental microHI. Furthermore, Alcaligenaceae, Enterobacteriaceae (family level), and Achromobacter (genus level) were the key groups introduced from the water environmental microbiota and potentially impacted the health status of wild fish in the middle Yangtze River. Therefore, this study comprehensively verifies the effectiveness of water environmental microHI as an indicator of the health status of wild fish. Our work provides a non-invasive method for the health monitoring of wild fish, offers a new path for understanding and managing aquatic wild animals' health status, and provides a new tool for the aquatic wildlife conservation toolkit. Of course, more research to develop the microHI-2.0 with weighting factors and adjustment parameters is welcome.

Data availability statement

The datasets presented in this study can be found in online repositories. The names of the repository/repositories and accession number(s) can be found in the article/Supplementary material.

Author contributions

HY: Conceptualization, Data curation, Formal analysis, Funding acquisition, Investigation, Methodology, Software, Visualization, Writing – original draft, Writing – review & editing. JZ: Data curation, Formal analysis, Visualization, Writing – original draft, Writing – review & editing. XL: Writing – review & editing, Methodology. JMW: Resources, Writing – review & editing. PC: Writing – review & editing. LS: Project administration, Writing – review & editing. JPW: Writing – review & editing. PL: Writing – review & editing. HD: Funding acquisition, Supervision, Validation, Writing – review & editing.

Funding

The author(s) declare financial support was received for the research, authorship, and/or publication of this article. This work was supported by the Central Public-Interest Scientific Institution Basal Research Fund, Yangtze River Fisheries Research Institute, Chinese Academy of Fishery Sciences [grant number YFI202201], and the Project of Yangtze Fisheries Resources Regular Monitoring after Fishing Ban [grant number CJJC-2022-01] from the Ministry of Agriculture and Rural Affairs, P. R. China.

Acknowledgments

The data supporting the findings of the current study are analyzed on the online platform of the Majorbio Cloud Platform (www. majorbio.com).

References

Adak, A., and Khan, M. R. (2019). An insight into gut microbiota and its functionalities. *Cell. Mol. Life Sci.* 76, 473–493. doi: 10.1007/s00018-018-2943-4

Albenberg, L. G., and Wu, G. D. (2014). Diet and the intestinal microbiome: associations, functions, and implications for health and disease. *Gastroenterology* 146, 1564–1572. doi: 10.1053/j.gastro.2014.01.058

Al-Hisnawi, A. A., Mustafa, J. M., Yasser, Y. K., Hussain, K. A., and Jabur, A. M. (2016). Influence of aquatic environment on microbiota of *liopropoma santi* fish in a local river in Iraq. *Karbala Int J Mod Sci* 2, 41–45. doi: 10.1016/j.kijoms.2016.01.001

Bolger, T., and Connolly, P. L. (1989). The selection of suitable indices for the measurement and analysis of fish condition. *J. Fish Biol.* 34, 171–182. doi: 10.1111/j.1095-8649.1989.tb03300.x

Borbón-García, A., Reyes, A., Vives-Flórez, M., and Caballero, S. (2017). Captivity shapes the gut microbiota of andean bears: insights into health surveillance. *Front. Microbiol.* 8:1316. doi: 10.3389/fmicb.2017.01316

Bozzi, D., Rasmussen, J. A., Carøe, C., Sveier, H., Nordøy, K., Gilbert, M. T. P., et al. (2021). Salmon gut microbiota correlates with disease infection status: potential for monitoring health in farmed animals. *Anim Microbiome* 3:30. doi: 10.1186/s42523-021-00096-2

Butt, R. L., and Volkoff, H. (2019). Gut microbiota and energy homeostasis in fish. Front Endocrinol 10:9. doi: 10.3389/fendo.2019.00009

Cabillon, N., and Lazado, C. (2019). Mucosal barrier functions of fish under changing environmental conditions. *Aust. Fish.* 4:2. doi: 10.3390/fishes4010002

Callens, M., De Meester, L., Muylaert, K., Mukherjee, S., and Decaestecker, E. (2020). The bacterioplankton community composition and a host genotype dependent occurrence of taxa shape the daphnia magna gut bacterial community. *FEMS Microbiol. Ecol.* 96:fna128. doi: 10.1093/femsec/fna128

Chen, Y. (ed.) (1998). Fauna Sinica: Osteichthys, Cypriniformes (middle). Beijing: Science Press.

Chen, T., Wang, Y., Gardner, C., and Wu, F. (2020). Threats and protection policies of the aquatic biodiversity in the Yangtze River. J. Nat. Conserv. 58:125931. doi: 10.1016/j. jnc.2020.125931

Cresci, G. A., and Bawden, E. (2015). Gut microbiome: what we do and don't know. Nutr. Clin. Pract. 30, 734–746. doi: 10.1177/0884533615609899

Elsheshtawy, A., Clokie, B. G. J., Albalat, A., Beveridge, A., Hamza, A., Ibrahim, A., et al. (2021). Characterization of external mucosal microbiomes of nile tilapia and grey mullet co-cultured in semi-intensive pond systems. *Front. Microbiol.* 12:13. doi: 10.3389/fmicb.2021.773860

Falony, G., Joossens, M., Vieira-Silva, S., Wang, J., Darzi, Y., Faust, K., et al. (2016). Population-level analysis of gut microbiome variation. *Science* 352, 560–564. doi: 10.1126/science.aad3503

Fassarella, M., Blaak, E. E., Penders, J., Nauta, A., Smidt, H., and Zoetendal, E. G. (2021). Gut microbiome stability and resilience: elucidating the response to perturbations in order to modulate gut health. *Gut* 70, 595–605. doi: 10.1136/gutjnl-2020-321747

Conflict of interest

The authors declare that the research was conducted in the absence of any commercial or financial relationship that could be construed as a potential conflict of interest.

Publisher's note

All claims expressed in this article are solely those of the authors and do not necessarily represent those of their affiliated organizations, or those of the publisher, the editors and the reviewers. Any product that may be evaluated in this article, or claim that may be made by its manufacturer, is not guaranteed or endorsed by the publisher.

Supplementary material

The Supplementary material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/fmicb.2023.1293342/ full#supplementary-material

Gajardo, K., Jaramillo-Torres, A., Kortner, T. M., Merrifield, D. L., Tinsley, J., Bakke, A. M., et al. (2017). Alternative protein sources in the diet modulate microbiota and functionality in the distal intestine of Atlantic salmon (*salmo Salar*). *Appl. Environ. Microbiol.* 83, e2615–e2616. doi: 10.1128/AEM.02615-16

Guivier, E., Pech, N., Chappaz, R., and Gilles, A. (2020). Microbiota associated with the skin, gills, and gut of the fish parachondrostoma toxostoma from the rhône basin. *Freshw. Biol.* 65, 446–459. doi: 10.1111/fwb.13437

Hayes, D. B., Brodziak, J. K. T., and O'Gorman, J. B. (1995). Efficiency and bias of estimators and sampling designs for determining length-weight relationships of fish. *Can. J. Fish. Aquat. Sci.* 52, 84–92. doi: 10.1139/f95-008

Heys, C., Cheaib, B., Busetti, A., Kazlauskaite, R., Maier, L., Sloan, W. T., et al. (2020). Neutral processes dominate microbial community assembly in Atlantic salmon, *salmo Salar. Appl. Environ. Microbiol.* 86:1. doi: 10.1128/AEM.02283-19

Himani, K., Durgesh, N. S., Anoop, S., Yogendra, S., Rup, L., and Ram, K. N. (2020). Gut microbiome of endangered *tor putitora* (ham.) as a reservoir of antibiotic resistance genes and pathogens associated with fish health. *BMC Microbiol.* 20:249. doi: 10.1186/ s12866-020-01911-7

Honda, K., and Littman, D. R. (2012). The microbiome in infectious disease and inflammation. *Annu. Rev. Immunol.* 30, 759–795. doi: 10.1146/annurev-immunol-020711-074937

Kashinskaya, E. N., Simonov, E. P., Kabilov, M. R., Izvekova, G. I., Andree, K. B., and Solovyev, M. M. (2018). Diet and other environmental factors shape the bacterial communities of fish gut in an eutrophic lake. *J. Appl. Microbiol.* 125, 1626–1641. doi: 10.1111/jam.14064

Kim, M., Yun, K. E., Kim, J., Park, E., Chang, Y., Ryu, S., et al. (2020). Gut microbiota and metabolic health among overweight and obese individuals. *Sci. Rep.* 10:19417. doi: 10.1038/s41598-020-76474-8

Kraimi, N., Calandreau, L., Zemb, O., Germain, K., Dupont, C., Velge, P., et al. (2019). Effects of gut microbiota transfer on emotional reactivity in japanese quails (*coturnix japonica*). *J. Exp. Biol.* 222:jeb202879. doi: 10.1242/jeb.202879

Kuang, T., He, A., Lin, Y., Huang, X., Liu, L., and Zhou, L. (2020). Comparative analysis of microbial communities associated with the gill, gut, and habitat of two filter-feeding fish. *Aquac Rep* 18:100501. doi: 10.1016/j.aqrep.2020.100501

Legrand, T. P. R. A., Wynne, J. W., Weyrich, L. S., and Oxley, A. P. A. (2020). A microbial sea of possibilities: current knowledge and prospects for an improved understanding of the fish microbiome. *Rev. Aquac.* 12, 1101–1134. doi: 10.1111/raq.12375

Ley, R., Spor, A., and Koren, O. (2011). Unravelling the effects of the environment and host genotype on the gut microbiome. *Nat. Rev. Microbiol.* 9, 279–290. doi: 10.1038/nrmicro2540

Li, W., Chen, X., Li, M., Cai, Z., Gong, H., and Yan, M. (2022). Microplastics as an aquatic pollutant affect gut microbiota within aquatic animals. *J. Hazard. Mater.* 423:127094. doi: 10.1016/j.jhazmat.2021.127094

Lindsay, E. C., Metcalfe, N. B., Llewellyn, M. S., and Fenton, A. (2020). The potential role of the gut microbiota in shaping host energetics and metabolic rate. *J. Anim. Ecol.* 89, 2415–2426. doi: 10.1111/1365-2656.13327

Luna, G. M., Quero, G. M., Kokou, F., and Kormas, K. (2022). Time to integrate biotechnological approaches into fish gut microbiome research. *Curr. Opin. Biotechnol.* 73, 121–127. doi: 10.1016/j.copbio.2021.07.018

Nash, R. D. M., Valencia, A. H., and Geffen, A. J. (2006). The origin of fulton's condition factor - setting the record straight. *Fisheries* 31, 236–238. doi: 10.1577/1548-8446-31-5

Pan, L., Han, P., Ma, S., Peng, R., Wang, C., Kong, W., et al. (2020). Abnormal metabolism of gut microbiota reveals the possible molecular mechanism of nephropathy induced by hyperuricemia. *Acta Pharm. Sin. B* 10, 249–261. doi: 10.1016/j. apsb.2019.10.007

Pisani, A., Rausch, P., Ellul, S., Bang, C., Tabone, T., Marantidis Cordina, C., et al. (2021). Dop 25 association of enterobacteriaceae with crohn's disease subtypes during remission. *J. Crohn's Colitis* 15:S063. doi: 10.1093/ecco-jcc/jjab073.064

Qiao, R., Sheng, C., Lu, Y., Zhang, Y., Ren, H., and Lemos, B. (2019). Microplastics induce intestinal inflammation, oxidative stress, and disorders of metabolome and microbiome in zebrafish. *Sci. Total Environ.* 662, 246–253. doi: 10.1016/j. scitotenv.2019.01.245

Reinhart, E. M., Korry, B. J., Rowan-Nash, A. D., and Belenky, P. (2019). Defining the distinct skin and gut microbiomes of the northern pike (*esox lucius*). *Front. Microbiol.* 10:2118. doi: 10.3389/fmicb.2019.02118

Reynoso-García, J., Miranda-Santiago, A. E., Meléndez-Vázquez, N. M., Acosta-Pagán, K., Sánchez-Rosado, M., Díaz-Rivera, J., et al. (2022). A complete guide to human microbiomes: body niches, transmission, development, dysbiosis, and restoration. *Front System Biol* 2:1403. doi: 10.3389/fsysb.2022.951403

Ringø, E., Zhou, Z., Vecino, J. L. G., Wadsworth, S., Romero, J., Krogdahl, Å., et al. (2016). Effect of dietary components on the gut microbiota of aquatic animals. A neverending story? *Aquac. Nutr.* 22, 219–282. doi: 10.1111/anu.12346

Sáenz, J. S., Marques, T. V., Barone, R. S. C., Cyrino, J. E. P., Kublik, S., Nesme, J., et al. (2019). Oral administration of antibiotics increased the potential mobility of bacterial resistance genes in the gut of the fish *piaractus mesopotamicus*. *Microbiome* 7:24. doi: 10.1186/s40168-019-0632-7

Sartor, R. B., and Wu, G. D. (2017). Roles for intestinal bacteria, viruses, and fungi in pathogenesis of inflammatory bowel diseases and therapeutic approaches. *Gastroenterology* 152, 327–339.e4. doi: 10.1053/j.gastro.2016.10.012

Satlin, M. J., and Walsh, T. J. (2017). Multidrug-resistant enterobacteriaceae, pseudomonas aeruginosa, and vancomycin-resistant enterococci: three major threats to hematopoietic stem cell transplant recipients. *Transpl. Infect. Dis.* 19, 1–29. doi: 10.1111/tid.12762

Schroeder, B. O., Bäckhed, F., Wallenberg, L., Institute Of Medicine, D.O.M.A, Göteborgs, U., Gothenburg, U., et al. (2016). Signals from the gut microbiota to distant organs in physiology and disease. *Nat. Med.* 22, 1079–1089. doi: 10.1038/nm.4185

Sehnal, L., Brammer-Robbins, E., Wormington, A. M., Blaha, L., Bisesi, J., Larkin, I., et al. (2021). Microbiome composition and function in aquatic vertebrates: small organisms making big impacts on aquatic animal health. *Front. Microbiol.* 12:567408. doi: 10.3389/fmicb.2021.567408

Skuja, V., Derovs, A., Pekarska, K., Rudzite, D., Lavrinovica, E., Piekuse, L., et al. (2018). Gut colonization with extended-spectrum β -lactamase-producing enterobacteriaceae may increase disease activity in biologic-naive outpatients with ulcerative colitis: an interim analysis. *Eur. J. Gastroenterol. Hepatol.* 30, 92–100. doi: 10.1097/MEG.00000000000989

Solomando, A., Capó, X., Alomar, C., Álvarez, E., Compa, M., Valencia, J. M., et al. (2020). Long-term exposure to microplastics induces oxidative stress and a proinflammatory response in the gut of *sparus aurata* linnaeus, 1758. *Environ. Pollut.* (1987) 266:115295. doi: 10.1016/j.envpol.2020.115295

Su, Y., Wu, P., Feng, L., Jiang, W., Jiang, J., Zhang, Y., et al. (2018). The improved growth performance and enhanced immune function by dl-methionyl-dl-methionine are associated with nf-kb and tor signalling in intestine of juvenile grass carp (*ctenopharyngodon idella*). *Fish Shellfish Immunol.* 74, 101–118. doi: 10.1016/j.fsi.2017.12.051

Sun, F., Wang, C., Chen, L., Weng, G., and Zheng, Z. (2020). The intestinal bacterial community of healthy and diseased animals and its association with the aquaculture environment. *Appl. Microbiol. Biotechnol.* 104, 775–783. doi: 10.1007/s00253-019-10236-z

Sun, F., Wang, Y., Wang, C., Zhang, L., Tu, K., and Zheng, Z. (2019). Insights into the intestinal microbiota of several aquatic organisms and association with the surrounding environment. *Aquaculture* 507, 196–202. doi: 10.1016/j.aquaculture.2019.04.026

Sun, F., and Xu, Z. (2021). Significant differences in intestinal microbial communities in aquatic animals from an aquaculture area. *J Mar Sci Eng* 9:104. doi: 10.3390/jmse9020104

Swenson, C. E., and Sadikot, R. T. (2015). Achromobacter respiratory infections. Ann. Am. Thorac. Soc. 12, 252–258. doi: 10.1513/AnnalsATS.201406-288FR

Tong, Q., Cui, L., Du, X., Hu, Z., Bie, J., Xiao, J., et al. (2020). Comparison of gut microbiota diversity and predicted functions between healthy and diseased captive *rana dybowskii*. *Front. Microbiol.* 11:2096. doi: 10.3389/fmicb.2020.02096

Turnbaugh, P. J., Ridaura, V. K., Faith, J. J., Rey, F. E., Knight, R., and Gordon, J. I. (2009). The effect of diet on the human gut microbiome: a metagenomic analysis in humanized gnotobiotic mice. *Sci. Transl. Med.* 1, 6ra14–14r. doi: 10.1126/scitranslmed.3000322

Turpin, W., Espin-Garcia, O., Xu, W., Silverberg, M. S., Kevans, D., Smith, M. I., et al. (2016). Association of host genome with intestinal microbial composition in a large healthy cohort. *Nat. Genet.* 48, 1413–1417. doi: 10.1038/ng.3693

Wang, H., Gou, W., Su, C., Du, W., Zhang, J., Miao, Z., et al. (2022). Association of gut microbiota with glycaemic traits and incident type 2 diabetes, and modulation by habitual diet: a population-based longitudinal cohort study in chinese adults. *Diabetologia* 65, 1145–1156. doi: 10.1007/s00125-022-05687-5

Wiles, T. J., and Guillemin, K. (2019). The other side of the coin: what beneficial microbes can teach us about pathogenic potential. *J. Mol. Biol.* 431, 2946–2956. doi: 10.1016/j.jmb.2019.05.001

Wilkes Walburn, J., Wemheuer, B., Thomas, T., Copeland, E., O'Connor, W., Booth, M., et al. (2019). Diet and diet-associated bacteria shape early microbiome development in yellowtail kingfish (*seriola lalandi*). *Microb. Biotechnol.* 12, 275–288. doi: 10.1111/1751-7915.13323

Xiao, F., Zhu, W., Yu, Y., He, Z., Wu, B., Wang, C., et al. (2021). Host development overwhelms environmental dispersal in governing the ecological succession of zebrafish gut microbiota. *Npj Biofilms Microbiomes* 7, 5–12. doi: 10.1038/s41522-020-00176-2

Xiong, J., Nie, L., and Chen, J. (2019). Current understanding on the roles of gut microbiota in fish disease and immunity. *Zool. Res.* 40, 70–76. doi: 10.24272/j. issn.2095-8137.2018.069

Yan, Q., Li, J., Yu, Y., Wang, J., He, Z., Van Nostrand, J. D., et al. (2016). Environmental filtering decreases with fish development for the assembly of gut microbiota. *Environ. Microbiol.* 18, 4739–4754. doi: 10.1111/1462-2920.13365

Yang, H. L., Du, H., Qi, H. F., Yu, L. X., Hou, X. D., Zhang, H., et al. (2021). Effectiveness assessment of using riverine water eDNA to simultaneously monitor the riverine and riparian biodiversity information. *Sci. Rep.* 11:24241. doi: 10.1038/ s41598-021-03733-7

Yang, H., Leng, X., Du, H., Luo, J., Wu, J., and Wei, Q. (2020). Adjusting the prerelease gut microbial community by diet training to improve the postrelease fitness of captivebred *acipenser dabryanus*. *Front. Microbiol.* 11:488. doi: 10.3389/fmicb.2020.00488

Yang, H. L., Wu, J. M., Du, H., Zhang, H., Li, J. Y., and Wei, Q. W. (2022). Quantifying the colonization of environmental microbes in the fish gut: a case study of wild fish populations in the Yangtze River. *Front. Microbiol.* 12:828409. doi: 10.3389/fmicb.2021.828409

Yang, H., Zhou, Q., Guo, X., Liu, Z., Zhang, H., and Du, H. (2023). Using the health index of the environmental microbiota to predict the health status of aquatic wildlife. *Ecol. Indic.* 151:110346. doi: 10.1016/j.ecolind.2023.110346

Yatsunenko, T., Rey, F. E., Heath, A. C., Warner, B., Reeder, J., Kuczynski, J., et al. (2012). Human gut microbiome viewed across age and geography. *Nature* 486, 222–227. doi: 10.1038/nature11053

Zhang, H., Kang, M., Shen, L., Wu, J., Li, J., Du, H., et al. (2020). Rapid change in Yangtze fisheries and its implications for global freshwater ecosystem management. *Fish Fish* 21, 601–620. doi: 10.1111/faf.12449

Zhang, S., Li, P., He, S., Xing, S., Cao, Z., Zhao, X., et al. (2023). Combined effect of microplastic and triphenyltin: insights from the gut-brain axis. *Env Sci Ecotechnol* 16:100266. doi: 10.1016/j.ese.2023.100266

Zhang, X., Li, X., Lu, J., Qiu, Q., Chen, J., and Xiong, J. (2021). Quantifying the importance of external and internal sources to the gut microbiota in juvenile and adult shrimp. *Aquaculture* 531:735910. doi: 10.1016/j.aquaculture.2020.735910

Zhang, Z., Li, D., Xu, W., Tang, R., and Li, L. (2019). Microbiome of co-cultured fish exhibits host selection and niche differentiation at the organ scale. *Front. Microbiol.* 10:2576. doi: 10.3389/fmicb.2019.02576

Zheng, B. (ed.) (1987). Animal atlas of China: Fish. Beijing: Science Press.

Zhernakova, A., Kurilshikov, A., Bonder, M. J., Tigchelaar, E. F., Schirmer, M., Vatanen, T., et al. (2016). Population-based metagenomics analysis reveals markers for gut microbiome composition and diversity. *Science* 352, 565–569. doi: 10.1126/science. aad3369

Zhong, J., Feng, L., Jiang, W., Wu, P., Liu, Y., Jiang, J., et al. (2019). Phytic acid disrupted intestinal immune status and suppressed growth performance in on-growing grass carp (*ctenopharyngodon idella*). Fish Shellfish Immunol. 92, 536–551. doi: 10.1016/j. fsi.2019.06.045

Zhou, L., Qu, Y., Qin, J. G., Chen, L., Han, F., and Li, E. (2021). Deep insight into bacterial community characterization and relationship in the pond water, sediment and the gut of shrimp (*penaeus japonicus*). *Aquaculture* 539:736658. doi: 10.1016/j. aquaculture.2021.736658

Zhu, X., Zheng, B., and Dai, D. (eds.) (1999). Fauna Sinica: Osteichthys, Silurifomes. Beijing: Science Press.

Zolkefli, N., Ramli, N., Mohamad-Zainal, N. S. L., Mustapha, N. A., Yusoff, M. Z. M., Hassan, M. A., et al. (2020). Alcaligenaceae and chromatiaceae as pollution bacterial bioindicators in palm oil mill effluent (pome) final discharge polluted rivers. *Ecol. Indic.* 111:106048. doi: 10.1016/j.ecolind.2019.106048