FISEVIER

Contents lists available at ScienceDirect

# Water Research



journal homepage: www.elsevier.com/locate/watres

# Mitigation of urbanization effects on aquatic ecosystems by synchronous ecological restoration

Hong Fu<sup>a,b,j</sup>, Pierre Gaüzère<sup>c</sup>, Jorge García Molinos<sup>d,e</sup>, Peiyu Zhang<sup>a</sup>, Huan Zhang<sup>a</sup>, Min Zhang<sup>f,\*</sup>, Yuan Niu<sup>g</sup>, Hui Yu<sup>g</sup>, Lee E. Brown<sup>b,h</sup>, Jun Xu<sup>a,i,\*\*</sup>

<sup>a</sup> Donghu Experimental Station of Lake Ecosystems, State Key Laboratory of Freshwater Ecology and Biotechnology of China, Institute of Hydrobiology, Chinese Academy of Sciences, No.7 Donghu South Road, Wuhan 430072, PR China

<sup>b</sup> School of Geography, University of Leeds, Leeds, West Yorkshire, United Kingdom

<sup>c</sup> Macrosystems ecology lab, School of Life Sciences, Arizona State University, Phoenix, United States

<sup>d</sup> Arctic Research Centre, Hokkaido University, Sapporo, Japan

<sup>e</sup> Graduate School of Environmental Science, Hokkaido University, Sapporo, Japan

<sup>f</sup> College of Fisheries, Hubei Provincial Engineering Laboratory for Pond Aquaculture, Freshwater Aquaculture Collaborative Innovation Centre of Hubei Province, Huazhong Agricultural University, Wuhan, PR China

Huaznong Agricultural University, Wunah, PR Unina

<sup>g</sup> National Engineering Laboratory for Lake Pollution Control and Ecological Restoration, Institute of Lake Environment and Ecology, Chinese Research Academy of Environmental Sciences, Beijing 100012, PR China

<sup>h</sup> Water@leeds, University of Leeds, Leeds, West Yorkshire, United Kingdom

<sup>i</sup> State Key laboratory of Marine Resource Utilization in South China Sea, Hainan University, Haikou, PR China

<sup>j</sup> University of Chinese Academy of Sciences, Beijing, PR China

# ARTICLE INFO

Keywords: Ecosystem degradation Wastewater treatment Pollution Macroinvertebrates Water quality

# ABSTRACT

Ecosystem degradation and biodiversity loss have been caused by economic booms in developing countries over recent decades. In response, ecosystem restoration projects have been advanced in some countries but the effectiveness of different approaches and indicators at large spatio-temporal scales (i.e., whole catchments) remains poorly understood. This study assessed the effectiveness of a diverse array of 440 aquatic restoration projects including wastewater treatment, constructed wetlands, plant/algae salvage and dredging of contaminated sediments implemented and maintained from 2007 to 2017 across more than 2000 km<sup>2</sup> of the northwest Taihu basin (Yixing, China). Synchronized investigations of water quality and invertebrate communities were conducted before and after restoration. Our analysis showed that even though there was rapid urbanization at this time, nutrient concentrations (NH<sub>4</sub><sup>4</sup>-N, TN, TP) and biological indices of benthic invertebrate (taxonomic richness, Shannon diversity, sensitive taxon density) improved significantly across most of the study area. Improvements were associated with the type of restoration project, with projects targeting pollution-sources leading to the clearest ecosystem responses compared with those remediating pollution sinks. However, in some locations, the recovery of biotic communities appears to lag behind nutrients (e.g., nitrogen and phosphorus), likely reflecting long-distance re-colonization routes for invertebrates given the level of pre-restoration degradation of the catchment. Overall, the study suggests that ecological damage caused by recent rapid economic development in China could potentially be mitigated by massive restoration investments synchronized across whole catchments, although these effects could be expected to be enhanced if urbanization rates were reduced at the same time.

# 1. Introduction

Almost all natural ecosystems on Earth have been disturbed by human development (Sévêque etal., 2020). Billions of dollars are invested annually to restore degraded ecosystems (Zhang et al., 2000), but many countries continue to face a dilemma between the needs of economic development and ecosystem restoration (Liu etal., 2016b). Therefore, adequate assessment of the efficiency of restoration projects

\* Corresponding author. \*\* Corresponding author.

https://doi.org/10.1016/j.watres.2021.117587

Received 23 February 2021; Received in revised form 23 July 2021; Accepted 17 August 2021 Available online 21 August 2021 0043-1354/© 2021 Elsevier Ltd. All rights reserved.

E-mail addresses: jorgegmolinos@arc.hokudai.ac.jp (J. García Molinos), zhm7875@mail.hzau.edu.cn (M. Zhang), xujun@ihb.ac.cn (J. Xu).

in maintaining and restoring natural ecosystem services in line with continuous sustainable development is needed.

There are few developing countries that have implemented as many and diverse ecosystem conservation and restoration projects in recent decades (Zhao et al., 2017), while maintaining rapid economic growth and urbanization, like China. Following the implementation of the reform and opening-up policy, China's urban population increased dramatically from 172.5 million in 1978 to 771.2 million in 2015 (Guan et al., 2018). This urban population growth has resulted in severe degradation of aquatic ecosystems as a consequence of land-use change, pollution and hydromorphological modification (Yang et al., 2019). To mitigate the severe ecosystem degradation, the Chinese government initiated major investments in eco-environmental conservation and restoration projects in 2000. The investment in total environmental restoration across the China mainland has increased from almost nothing in 1994 to 1 trillion RMB Yuan in 2014 (Zhou et al., 2017). Whilst these factors have made China one of the world's leading investors in ecosystem restoration, there is also a general perception that the national restoration policies and actions have contributed a lot to improve the status of water quality across China (Zhou et al., 2017). However, no study has yet attempted to describe the quantitative relationship between the indices of different restoration projects targeting either pollution sources (the place when pollution was generated) or pollution sinks (natural aquatic ecosystems like rivers) and ecosystem indices (nutrients, biological communities, etc.) across large space and time scale.

These investments in river restoration in China provide opportunities to enhance understanding of catchment-scale remediation schemes with varied restoration approaches, which have received comparatively less attention than restoration schemes focused on river sections (Ramchunder et al., 2012), or single types of restoration measures (Kail et al., 2015). To maintain continued, unreserved support from governmental institutions and the general public, the benefits from coordinated, large-scale ecological conservation and restoration efforts urgently need to be evaluated with communication of lessons learned to decision makers.

Here, we combine historical and present data to explore the relationship between a large set of different restoration projects (spanning a range of investments and removal amount of nutrients) and aquatic ecosystem responses in the Taihu basin (Yixing, China). Increasing urbanization intensity can complicate interpretation of aquatic environmental restoration effects over time, although impervious surface area provides a quantifiable index to incorporate this potentially confounding element into the study (Yang et al., 2019). The aim was to examine the effectiveness of ecosystem restoration using nutrients and macroinvertebrates as key indicators, spanning 10 years and across a large spatial-scale (> 2000 km<sup>2</sup>); We hypothesized that (Fig. 1): (H<sub>i</sub>) ecological damage caused by rapid economic development can be effectively mitigated by synchronous large-scale restoration projects;  $(H_{ii})$  the recovery of biotic indices would lag behind change in abiotic indices (e.g., nutrients) following the implementation of restoration projects because of the extensive pre-restoration degradation of the catchment limiting



Fig. 1. Conceptual diagram of expected changes in aquatic ecosystems over time as a consequence of restoration. To demonstrate successful restoration, response ratios of abiotic and biotic indices should increase significantly relative to their respective values in the degraded state, ideally reaching predefined target levels corresponding to the desired restoration state.

potential for rapid recolonization; and  $(H_{\rm iii})$  the choice of restoration approach can be expected to result in different effects on the ecosystem restoration with, for example, approaches such as dredging modifying physical habitat and potentially exacerbating stress and delaying recovery.

To test these hypotheses, we gathered information on several hundreds of existing restoration projects conducted in the northwest sector of the Lake Taihu basin (China) over a period of 10 years. Several restoration project indices, nutrient concentration and biological indices of benthic macroinvertebrates in aquatic ecosystems were then computed, and their local trends assessed via a moving-window approach taking into account increases in urbanization intensity and the investment made on the restoration projects. The integrated assessment of multiple data sources provides a novel and thorough analysis on the role of environmental restoration project investments on the water qualities of a watershed under the influence of urban expansion.

# 2. Material and methods

# 2.1. Study region

The Taihu basin, located at the plain river network region in the downstream area of the Yangtze River, covers an area of approximately  $36,895 \text{ km}^2$  (Fig. 2). The basin, representing 0.4% of China's land area, is heavily populated (40 million residents) and highly industrialized, supporting 11% of China Gross Domestic Product (GDP) (Yi et al., 2017).

This study focused specifically on the upstream areas of the northwest Taihu basin, covering the whole area of Yixing city (Fig. 2). The district covers a total area of 1996.6 km<sup>2</sup> (including 242.29 km<sup>2</sup> of Lake Taihu), 16.8% of which is occupied by water bodies. The catchment has a northern subtropical monsoon climate with an average annual temperature of 16.0 °C and abundant rainfall (1177 mm a year on average,). The urban area of 66.3 km<sup>2</sup> includes rivers with a density of approximately 2.27 km/km<sup>2</sup> (Wang et al., 2017).

Yixing provides an ideal case study to test our hypotheses for two main reasons. First, it represents the typical characteristics of the wider Taihu basin (Pan and Zhao, 2007), and includes nine of the 13 main tributaries to Lake Taihu, which together account for around 60% of the total flow into the lake Taihu. Second, Yixing has spent 8.21 billion RMB (\$1.2B USD) on 440 different aquatic environment restoration projects throughout the catchment between 2007 and 2017.

# 2.2. Restoration project data

We collected data corresponding to restoration projects which were implemented and maintained between 2007 and 2017. The database contained > 440 water environmental restoration projects from the Development and Reform Commission of Jiangsu province (Jiangsu Development and Reform Commission, 2008; Jiangsu Development and Reform Commission, 2014). Of these, one hundred projects that did not provide information on specific restoration measures, the project scale, or for which the measures taken could not be quantified and converted into removal quantity of nitrogen and phosphorus, were discarded (see below for an explanation of how we calculated these parameters). Similarly, projects that did not have direct impacts on the aquatic environment (garbage disposal, drinking water treatment) were removed from further analysis. We collected data on the location of the restoration works (latitude and longitude) (Fig. 2), type of restoration projects (Figure 5), starting and completion year, specific restoration measures, project scale, and total investments (Table S1). To eliminate the effects of inflation on the project investment costs, we used 2007 as the base year and made price adjustments to that baseline for other years' investments (Table S2) (Imai, 2018). Projects were classified according to targeted pollution paths: (i) restoration projects targeting pollution sources (e.g., treatment of industrial and agricultural (farming,

aquaculture and livestock breeding etc.) wastewater or sanitary sewage), and (ii) restoration projects targeting pollution sinks (e.g., dredging of contaminated sediment, water hyacinth cultivation for removal of pollutants, harvesting of harmful blue-green algae, etc.). For the restoration projects that aimed to control wastewater pollution at source, we further divided those restoration projects into three different categories: (i) industry-focused, (ii) agricultural wastewater (mainly include livestock breeding and aquaculture in this study), and (iii) domestic sewage.

For each restoration project we calculated the removal quantity (in  $10^4$  t/a) of key nutrients including ammonia nitrogen (NH<sup>+</sup><sub>4</sub>-N), total phosphorus (TP) and total nitrogen (TN), according to different subproject categories and by reference to various national or regional standards of wastewater discharging (for formulas see Table 1 and references therein). The main principle of the removal quantities calculation was to estimate nutrient removal from the sink of the water pollution in theory Table 2.

# 2.3. Field sampling

To assess relationships between aquatic ecosystems and nutrient removal efficiency, we monitored the recovery of 63 locations by sampling each site both before (2007) and after (2017) the implementation of the restoration works. Sampling sites were located in the limnetic zone of the lakes or the rivers of Yixing and collected between July and September during both sampling campaigns (Fig. 2).

Benthic macroinvertebrate samples were collected within a 100 m reach for each site using a  $0.05 \text{ m}^2$  modified Peterson grab (three grabs per reach), and sieved *in situ* through a 250 µm mesh. The resulting sieved materials were stored in a cooler box and transported to the laboratory on the same day. In the laboratory, the samples were sorted on a white tray, and all specimens picked out and preserved in 7% formalin solution. Specimens were identified to the lowest feasible taxonomic level under a dissection microscope (Olympus® SZX10) according to several taxonomic keys (Morse et al., 1994; Wang, 2002).

Simultaneously with benthic macroinvertebrate sampling, four water samples were collected from an intermediate depth at each site, stored in an acid-cleaned plastic container (200 mL), and kept in a cool box for transportation to the laboratory. TN (mg/L), TP (mg/L) and  $NH_4^+$ -N (mg/L) were then measured using an ultraviolet spectrophotometer (PhotoLab S12, WTW Company, Munich, Germany). TP and TN were measured on the unfiltered samples, whereas  $NH_4^+$ -N was determined from samples filtered using 0.45  $\mu$ m Whatman GF/F filters (Whatman, Kent, Great Britain). All storage, preservation, and chemical analyses were performed in the laboratory following national standard analytical methods for water and wastewater (National Environmental Protection Bureau, 2002).

# 2.4. Quantification of restoration effects

*Nutrient concentrations:* We used the response ratio  $\Delta r$  proposed by Benayas et al. (2009) as a standardized effect size of restoration effects (Eq. (1)). The response ratio is dimensionless with positive values indicating an improvement of the original degraded status and negative values denoting a degradation. Given that decreasing NH<sub>4</sub><sup>+</sup>-N, TN and TP concentrations in eutrophic environments are the target of restoration, we reversed the sign of the resulting ration (- $\Delta r$ ) for all assessed nutrient parameters (NH<sub>4</sub><sup>+</sup>-N, TN and TP) to make their interpretation more intuitive and keep consistency with that of the biological indices.

$$\Delta r = -\ln(After \ Restoration \ / \ Degraded) \tag{1}$$

**Biological indices:** By referring to the applications of biological indices in the Yangtze River Basin, China (Huang et al., 2015), taxonomic richness, Shannon–Wiener index (Simpson, 1949) and percentage of Oligochaeta were selected as representative indices to describe the



**Fig. 2.** Map of the study area (Yixing, China) showing the location of the sampling sites, restoration project sites and the spatial definitions considered. Insets refer to ① the changing trend of Gross Domestic Product (GDP) in Yixing from 2002 to 2018), and ② schematic of the calculation process where each squared grid ( $250 \times 250$  m) was considered the center of a 6 km radius window containing at least three sampling sites and nine restoration project sites. Indices were then calculated for each of the 4080 windows (see Methods for details on the type of indices and their calculation).

# Table 1

Evaluation of removal quantity of nutrients include NH<sub>4</sub><sup>+</sup>-N, TN and TP.

Name of sub projects	Class of restoration	Evaluation formulas	Units	References	
	measures				
Biogas digester	Source	Vb*swine heads*pollution coefficient* nutrients	$10^{4}$	/	
		removal efficiency	t/a		
Septic tank	Source	Vs/daily output of livestock sewage*pollution	104	/	
Variation and a distant	0	coefficient* nutrients removal efficiency	t/a 10 <sup>4</sup>	(Chinese Descent) Assidence of 2020)	
stondard discharge	Source	pig population "pollution coefficient" nutrients	10	(Chinese Research Academy or, 2020)	
Standard discharge	Course	(mass of livesteely sources (deily output of	10 <sup>4</sup>	$((X_{i0} \text{ ot } a_{1}, 2014))$	
Sewage of Investock	source	(mass of investock sewage/ daily output of livestock sewage) * pollution coefficient *	10 t/a	((Ale et al., 2014))	
		nutrients removal efficiency	t/ a		
Fermentation bed	Source	area of fermentation bed*breeding	$10^{4}$	/	
		density*pollution coefficient* nutrients removal	t/a	,	
		efficiency			
Removal of net cage culture	Source	area of net cage*pollution coefficient	$10^{4}$	(Chinese Research Academy of, 2020)	
-			t/a		
Renovation of wastewater	Source	treatment scale*(influent of nutrients	$10^{4}$	(Department of Ecology and Environment of Jiangsu	
treatment & water conservation		concentration-effluent nutrients concentration)	t/a	Province, 2004; Department of Ecology and Environment of	
and zero emission projects		* 365		Jiangsu Province, 2007; State Environmental Protection	
				Agency of the People's Republic of China, 2002)	
Rural population benefited by	Source	population*pollution coefficient of rural	$10^{4}$	(Wang et al., 2010)	
sewage treatment facilities	_	people*365	t/a		
Domestic wastewater treatment	Source	treatment scale*(influent of nutrients	104	(Ministry of Environmental Protection of the People's	
		<ul><li>concentration-effluent nutrients concentration)</li><li>* 365</li></ul>	t/a	Republic of China, 2010)	
Diversion of urban rain and	Source	(area of rain and sewage diversion/per capita	$10^{4}$	(Ministry of Environmental Protection of the People's	
sewage water		occupation land)*pollution coefficient of urban people*365	t/a	Republic of China, 2010)	
Ecological forest	Sink	area of ecological forest*annual reduction of	$10^{4}$	(Sun et al., 2015)	
0		nutrients	t/a		
Surface flow wetlands	Sink	area of wetlands* nutrients removal efficiency	$10^{4}$	(Li, 2017)	
			t/a		
Dredging of contaminated	Sink	Vd*bulk density*(1-moisture content of silt) *	$10^{4}$	(Liu et al., 2016a; Qin et al., 2005; (Zhu et al., 2008))	
sediment		release coefficient of nutrients *average amount	t/a		
		of nutrients of dry matter			
Cyanobacteria salvage	Sink	dealing rate of cyanobacteria*(moisture content)	$10^{4}$	(Zhang et al., 2009; Zhou, 2012)	
		* nutrients content of dry matter*duration days	t/a		
		of cyanobacteria bloom	1 04		
Water hyacinth planting	Sink	area of water hyacinth* nutrients removal	104	(Liu et al., 2015; Zhao, 2010)	
		efficiency	t/a		

Notes: Vb, volume of biogas digester; Vs, volume of septic-tank; Vd, volume of dredging.

variation of benthic macroinvertebrate assemblages. A function of species richness and density (Nzengya and Wishitemi, 2000) was used to determine the Shannon diversity.

The Hilsenhoff Family Biotic Index (FBI) (Hilsenhoff, 1988) was applied to assess the ecological conditions of each site. FBI sore are assigned a tolerance number from 0 (very intolerant) to 10 (highly tolerant), and calculated by the following equation:  $FBI = \sum^{[}(TV_i)(n_i)]/N$ , where  $TV_i$  is the tolerance value of the *i*th taxon,  $n_i$  is the number of individuals in *i*th taxon, and N is the total number of individuals in the sample. The tolerance value of each family was obtained from Qin et al. (2014) and Wang and Yang (2004). Low FBI values reflect a higher abundance of sensitive invertebrate groups, thus a lower level of organic pollution.

We analyzed the changes in species composition between restored (2017) and degraded (2007) sites using the command beta.temp in the *R* package betapart (Baselga and Orme, 2012). This procedure computes the total dissimilarity (measured as Sørensen dissimilarity,  $\beta_{SOR}$ ), and partitions it into turnover ( $\beta_{SIM}$ ) and nestedness ( $\beta_{SNE}$ ) components (Baselga, 2012). In the context of temporal variation of communities these two components reflect (i) the substitution of some species by others through time ( $\beta_{SIM}$ ), and (ii) the loss (or gain) of species through time in a nested pattern ( $\beta_{SNE}$ ).

Biological response ratios were based on a slightly modified formula:

$$\Delta r = \ln[(After \ Restoration + 1) / (Degraded + 1)]$$
<sup>(2)</sup>

where, in this case, the degraded and restored conditions were calcu-

lated using the biological indices of benthic macroinvertebrate (taxonomic richness, Shannon diversity, percent Oligochaeta and Hilsenhoff FBI). The addition of a unit (+1) to each term in the formula was needed because some sites it registered zero values.

# 2.5. Land use data and urbanization metric

Land use data for Yixing district was derived from 30-m resolution land use maps for 2007 and 2017 (taken as surrogates for existing conditions before and after implementation of restoration) provided by the Resource and Environmental Science Data Center of the Chinese Academy of Sciences (http://www.resdc.cn) (Fig. S1). The 26 original land use categories were simplified into six categories according to the land resource classification system of China's land use/land cove change (CNLUCC), namely farm land, building land (artificial surfaces), forest land, grassland, water body and barren land (Song and Deng, 2017). The land use transformation matrix for the Yixing district across the six land use categories between 2007 and 2017 is provided in Table S3.

The impervious surface area (ISA) of Yixing has increased from 4.36% in 2007 to 10.15% in 2017. Prior research has noted that when the ISA increases to a range between 10 and 25%, the impact on aquatic environments is significant (Schueler, 1994). However, the water environment in relation to the ISA may vary depending on regional conditions (Luo et al., 2018). Thus, we used the response ratio of impervious surface area (rISA = ln (ISA<sub>2017</sub>/ISA<sub>2007</sub>)) as a co-variable in subsequent analyzes to assess confounding effects of land use change (urbanization) acting in opposition to restoration effects. Land use data and the

# Table 2

Results of GLMM and LMM for nutrients (NH<sub>4</sub><sup>+</sup>-N, TN, TP), the investments of different restoration project categories and the intensity of urbanization (expressed as the response ratio of impervious surface area (rISA)) on biological parameters (taxa richness, Shannon diversity,% Oligochaeta,  $\beta_{SOR}$ ). Variables are only given when the correlation was significant (p < 0.05). Variables shaded in gray correspond to positive correlations. s\_Livst\_inv, investment targeting agricultural sewage; s\_san\_inv, investment targeting sanitary sewage; s\_ind\_inv, investment targeting industry waste water; sinkPinvstm, investment targeting pollution sink.

taxa richness GLMM (gaussian, link="log"), $N = 3022$ , Marginal $R^2$ : 0.73								
Variables	Estimates	SE	t	Р				
(Intercept)	1.44	0.13	10.79	< 0.001				
log(NH <sub>4</sub> <sup>+</sup> -N)	0.27	0.02	14.00	< 0.001				
TN	-0.36	0.02	-14.62	< 0.001				
ТР	-0.67	0.02	-42.42	< 0.001				
s_Agric_inv	-0.28	0.04	-6.93	< 0.001				
s_san_inv	0.09	0.02	4.62	< 0.001				
s_ind_inv	-0.12	0.03	-3.63	< 0.001				
sinkPinvstm	-0.47	0.01	-43.78	< 0.001				
s_Agric_inv: rISA	0.74	0.08	9.56	< 0.001				
s_san_inv:rISA	-0.16	0.03	-4.70	< 0.001				
s_ind_inv:rISA	-0.39	0.07	-5.63	< 0.001				
Shannon diversity LMM, $N = 3022$ , Marginal R <sup>2</sup> : 0.63								
Variables	Estimates	SE	t	Р				
(Intercept)	1.76	0.07	23.66	< 0.001				
log(NH <sub>4</sub> <sup>+</sup> -N)	0.51	0.01	39.10	< 0.001				
TN	-0.09	0.02	-4.65	< 0.001				
TP	-0.56	0.01	-44.70	< 0.001				
s_Agric_inv	-0.25	0.003	-8.13	< 0.001				
s_san_inv	0.04	0.01	4.65	< 0.001				
s_ind_inv	-0.32	0.01	-25.77	< 0.001				
sinkPinvstm	-0.18	0.003	-21.88	< 0.001				
rISA:s_Livst_inv	0.41	0.07	5.98	< 0.001				
(-% Oligochaeta) LMM, $N = 3022$ , Marginal R <sup>2</sup> : 0.47								
Variables	Estimates	SE	t	Р				
(Intercept)	0.46	0.06	1.82	< 0.001				
log(NH <sub>4</sub> <sup>+</sup> -N)	-0.03	0.01	-4.33	< 0.001				
TN	-0.10	0.01	-8.19	< 0.001				
TP	-0.03	0.01	-3.90	< 0.001				
s_Agric_inv	0.05	0.02	2.72	< 0.01				
s_san_inv	-0.11	0.01	-14.65	< 0.001				
s_ind_inv	0.28	0.01	18.14	< 0.001				
sinkPinvstm	-0.26	0.01	-27.11	< 0.001				
s_Agric_inv:rISA	-0.22	0.04	-5.01	< 0.001				
s_san_inv:rISA	0.18	0.02	8.00	< 0.001				
s_ind_inv:rISA	-0.53	0.04	-14.94	< 0.001				
sinkPinvstm:rISA $\beta_{SOR}$ LMM, $N = 3022$ , Marginal R <sup>2</sup> :	0.34	0.02	10.36	<0.001				
0.31		0.1						
Variables	Estimates	SE	t	P				
(Intercept)	0.62	0.03	18.60	< 0.001				
$\log(\mathbf{NH}_4^+ \cdot \mathbf{N})$	-0.07	0.03	-14.21	<0.001				
TP	-0.04	0.01	-8.96	< 0.001				
s_Agric_inv	0.09	0.01	7.44	< 0.001				
s_ind_inv	0.03	0.01	2.84	< 0.001				
s_san_inv	0.04	0.004	8.70	< 0.001				
sinkPinvstm	-0.06	0.003	-17.81	< 0.001				
s_Agric_inv:riSA	-0.18	0.003	-6.75	< 0.001				
s_san_inv:rISA	-0.08	0.01	-9.58	< 0.001				
s_ind_inv:rISA	-0.21	0.02	-9.56	< 0.001				

impervious surface area were handled and calculated using ArcGIS 10.2 (ESRI Company, Redlands, CA, USA) and Fragstats 4.2 (McGarigal et al., 2012).

# 2.6. Data analysis

2.6.1. Assessing spatial distribution of project indices, ecosystem indices and the response ratio of impervious surface area

Because of the well-developed floodplain river network of Yixing

district, Taihu Basin, the landform is flat, water flows slowly, and flow direction is often variable because of the influence of artificial drainage (Deng et al., 2015). Thus, we adopted a moving window approach to estimate all parameters (project, ecosystem and urbanization intensity indices) on a spatial continuum covering the whole study area. This approach is useful for summarizing local spatial trends emerging from regional dynamics (Gaüzère et al., 2016). The principle lies in calculating the metrics of interest for each cell of a squared grid ( $250 \times 250$  m, slightly less than the distance between the two nearest sampling sites to generate more windows), covering the study area, using a circular moving window centered on the centroid of each cell. In this way, the values of the different metrics attributed to each grid cell represent summaries of the neighboring restoration project sites, sampling sites and the response ratio of impervious surface area (Fig. 2).

We used a 6 km radius for the circular window (Figs. 2 and S2). The chosen window radius resulted from a compromise between incorporating the range of restoration projects and enough spatial repetition to estimate reliable linear trends in variables, and achieving an adequate coverage of the study area. This generated 4080 spatial windows, each containing at least three sampling sites and nine restoration project sites. Finally, indices (project, nutrients and biological indices) were calculated for the 4080 spatial windows based on the mean of ecosystem indices or the sum of project indices, and the response ratio of impervious surface area was then calculated for each window. This moving window approach enabled the local spatial trends of each restoration project index to be compared with the local spatial trends of aquatic ecosystem indices (Gauzere et al., 2017).

#### 2.6.2. Statistical analysis for all indices

Visual inspection of frequency histograms showed all response ratios of ecosystem indices ( $\Delta$ rNH<sup>4</sup><sub>4</sub>-N,  $\Delta$ rTN,  $\Delta$ rTP, taxonomic richness, Shannon diversity,% Oligochaeta and Hilsenhoff FBI) followed nonnormal distributions (Fig. 4). Therefore, we used Wilcoxon signed rank tests to examine whether median response ratios of ecosystem indices were significantly different from zero. Non-metric multidimensional scaling ordination (NMDS) was used to visualize invertebrate communities by site and restoration phase (before/after). Taxon density data were ordinated using Bray–Curtis similarity as the distance measure for the scaling with square-root transformation to reduce impacts of extremely high counts of individual taxa. Similarity percentage (SIMPER) analysis was used to identify which taxa contributed the most to the average Bray-Curtis dissimilarity between the two-restoration phases.

Spearman Rank correlation was used to test for significant correlations between project investment and removal quantity of NH<sub>4</sub><sup>+</sup>-N, TP, TN by project category. We also used Kruskal-Wallis tests to examine whether investments differed among different restoration project categories. Finally, the relationships between restoration projects and ecosystem recovery were assessed by fitting a generalized linear mixed model (GLMM) with a Gamma distribution (log link) or Linear Mixed Model (LMM) to each nutrient ( $\Delta rNH_4^+$ -N,  $\Delta rTN$ ,  $\Delta rTP$ ). Restoration project investment by category and the response ratio of impervious surface area (rISA) were added as fixed effects, while the number of years since the implementation of the restoration (DurationT) and the time since completion of the restoration (dt = 2017 - end year of the restoration) were used as random effects. GLMM with Gaussian distribution (log link) or LMM were applied to the biological indices ( $\Delta r$ taxonomic richness,  $\Delta r$  Shannon diversity,  $\Delta r$ % Oligochaeta,  $\beta$ SOR) with nutrients and investment of different restoration categories as fixed effects, rISA as covariate, Duration T and dt as random effects. Removal quantity of nutrient was subsequently omitted from these models because of its significant positive correlation with project investment (see Results section). To explore the interaction effect between urbanization intensity and the strength of restoration, the interaction term 'rISA\*investment of different project categories' was included in the

Prior to analysis, the investment of each restoration project category was  $log_{10}$  transformed to constrain the influence of extreme values. We compared the complex model with a null model; models were simplified

by removing non-significant terms and verifying the distribution through residuals analysis ((Crawley, 2002)). Akaike's Information Criterion (AIC) values were used to determine the most parsimonious fit. Model residuals were tested for spatial autocorrelation with Moran's



**Fig. 3.** Scatter plots showing the relationships between project investments of different restoration project categories and either of (a–f) removal quantities for the different nutrients ( $\Delta$ rNH<sub>4</sub><sup>4</sup>-N,  $\Delta$ rTN,  $\Delta$ rTP) (the marginal boxplots in a-f show the investment distribution on different restorations) and response ratio (g–l) of different restoration project categories, (a–c) restoration measures for pollution source (Spearman rank *Rs* = 0.62, 0.58, 0.55, *p* < 0.001) and pollution sink (*Rs* = 0.79, 0.85, 0.82, *p* < 0.001); (d–f) three main categories of restoration measures for pollution source, which include restoration measures for industry waste water (*Rs* = 0.92, 0.95, 0.94, *p* < 0.001), agricultural (*Rs* = 0.89, 0.89, 0.89, *p* < 0.001) and sanitary (*Rs* = 0.70, 0.68, 0.69, *p* < 0.001) sewage. Marginal effects of investment of different restoration project categories on each nutrient ( $\Delta$ rNH<sub>4</sub><sup>4</sup>-N,  $\Delta$ rTN and  $\Delta$ rTP): (g–i) restoration measures for pollution source, which include restoration measures for pollution source and sink; (j–l) three main categories of restoration source, which include restoration measures for pollution source and sink; (j–l) three main categories of restoration measures for industry waste water, agricultural and sanitary sewage. GLMM or LMM regression lines are given where a correlation was significant (*p* < 0.05). The initial unit of investment is 10<sup>5</sup> RMB, and the initial unit of the removal quantity of nutrients is 10<sup>4</sup> t/a, both were log<sub>10</sub> transformed before inclusion in models.

tests (Birk et al., 2020), which showed in all instances no autocorrelation.

All data analysis was performed in using R v 4.0.1 (R Core Team 2020, https://www.R-project.org/) using the packages: lme4 and lmerTest.

# 3. Results

3.1. Relationship between restoration project investments and nutrient removal

Spearman rank (*Rs*) correlations analysis showed a significant positive correlation between project investment and the removal quantity of nutrients (calculated as described in Table 1) across project categories (Fig. 3, Table S4). The amount of money invested by the government varied significantly with project category (Kruskal Wallis test, p <0.001). The projects attracting larger investments were, in decreasing order of magnitude: pollution source, pollution sink, sanitary sewage, industrial wastewater, agricultural sewage (Fig. 3).

# 3.2. Efficiency of restoration projects on nutrients and biological status

Restoration works were found to be efficient at recovering aquatic ecosystems from their initial degraded condition as shown by their significant effect on almost all assessed ecosystem indices. The concentration of NH<sup>+</sup><sub>4</sub>-N, TN and TP across the whole Yixing river network was significantly lower in restored (2017) than in degraded (2007) aquatic ecosystems, leading to overall positive response ratios (Fig. 4); Taxonomic richness and Shannon diversity of benthic macro-invertebrate were significantly higher in restored (2017) than in degraded (2007) sites (mean response ratio = 1.085, 0.415, *P* < 0.001, Fig. 4). Percent Oligochaeta was significantly lower in restored (2017, 17.53% ± 16.65%) than in degraded (2007, 40.78% ± 39.70%) sites. Hilsenhoff FBI of benthic macroinvertebrate communities showed no significant difference between degraded (2007) and restored (2017) ecosystems Fig. 5.

The composition of benthic macroinvertebrate communities differed significantly between degraded (2007) and restored (2017) periods (PERMANOVA, p < 0.01; final stress = 0.128, Fig. 6). SIMPER analysis identified eight species cumulatively contributing > 70% to the dissimilarity between restored (2017) and degraded (2007) invertebrate communities (Table S5). They were *Limnodrilus hoffmeisteri*, *Bellamya* 

aeruginosa, Corbicula fluminea, Branchiura sowerbyi, Parafossarulus eximius, Neocaridina denticulata, Exopalaemon modestus and Parafossarulus striatulus in decreasing order. Some sensitive species to anthropogenic pressures with low tolerance values recolonized after the restoration (2017). For example, river flies *Heptagenia* sp., *Ephemera orientalis and Ceratopsyche* sp. The partitioning of the Sørensen dissimilarity index was dominated by species turnover ( $\beta_{SIM}$ ) (mean = 0.44, SD = 0.36), implying that, in any given site, an average of 44% of the species were unique to the time (either 2007 or 2017 site assemblage). In contrast, the nestedness component ( $\beta_{SNE}$ ) was much lower (mean = 0.34, SD = 0.33), implying that weaker patterns of species losses or gains from preexisting communities have occurred between 2007 and 2017 (Fig. 4). The spatial distribution of total ( $\beta$ SOR) and nested ( $\beta$ SIM) dissimilarity can be seen in Figure 5.

# 3.3. Effects of restoration projects on aquatic ecosystem status

Examination of the marginal effect of project investment amount by category on nutrients (Fig. 3, Table 2) showed a significant correlation of decreasing river network NH<sub>4</sub><sup>+</sup>-N concentrations (i.e., increasing response ratios) with increasing investment on projects targeting pollution sources but not those targeting pollution sinks (marginal = 0.20, p < 0.001). On the contrary, decreasing TN and TP concentrations were positively correlated with increasing investment on restoration projects targeting both pollution sources and sinks (marginal  $R^2 = 0.23$ , p < 0.001; marginal  $R^2 = 0.19$ , p < 0.001). Decreasing NH<sub>4</sub><sup>+</sup>-N and TP concentrations correlated with increasing investment in both restoration projects targeting agricultural and domestic sewage, but not those targeting industry wastewater (marginal  $R^2 = 0.14$ , p < 0.001; marginal  $R^2$ = 0.19, p < 0.001). Decreasing TN concentrations were negatively correlated with the increasing investment on restoration projects targeting sanitary sewage (marginal  $R^2 = 0.32$ , p < 0.001). A significant interaction was evident between the response ratio of impervious surface area and investments of different restoration project categories and nutrient responses. For example, poor nutrient responses were associated with the growth of impervious surface area (p < 0.001, Fig. 7), but these effects were overcome where restoration projects were large but impervious area increased minimally.

For the biological indices, increased Shannon diversity and taxonomic richness over time showed significant inverse relationships with NH<sub>4</sub><sup>+</sup>-N concentrations, and a positive association with increasing investment on restoration projects targeting sanitary sewage (Shannon



**Fig. 4.** Response ratios of NH<sup>4</sup><sub>4</sub>-N, TN, TP and taxonomic richness (Richness), Shannon diversity, percent Oligochaeta, FBI of benthic macroinvertebrate in restored (2017) compared with degraded ecosystems (2007) (a,b). All response ratios differed significantly from zero (Wilcoxon signed rank tests, p < 0.001) except for Hilsenhoff FBI. The mean and SD are given alongside the overall data distribution for each metric. (c) The partition of temporal total dissimilarity ( $\beta_{SNE}$ -solid gray line) and turnover ( $\beta_{SIM}$ -dashed lines) for beta diversity of benthic macroinvertebrates in Yixing from 2007 to 2017.

H. Fu et al.



**Fig. 5.** (a) Location of different restoration project sites by category (n = 420, projects of garbage disposal and drinking water treatment were not included in the analysis) in Yixing from 2007 to 2017. (b,c) Maps showing the spatial distribution of total ( $\beta_{SOR}$ ) and nested ( $\beta_{SIM}$ ) dissimilarity for beta diversity of benthic macroinvertebrates.



**Fig. 6.** NMDS biplots showing changes in community composition of benthic invertebrates among restoration projects between their initial degraded (2007) and final restored (2017) states in the Yixing river network with indication of (a) the individual taxa (denoted by S) and (b) sampling sites (denoted by the numbers). An outlier was removed from this figure because it had only one scare species in 2007 and had 16 species in 2017. S1, *Limnodrilus hoffmeisteri*; S2, *Bellamya aeruginosa*; S3, *Branchiura sowerby*; S4, *Corbicula fluminea*; S5, *Parafossarulus eximius*; S6, *Nephtys oligobranchia*; S7, *Parafossarulus striatulus*; S8, *Neocaridina denticulata*; S9, *Semisulcospira cancelata*; S10, *Gammarus* sp.; S11, *Exopalaemon modestus*; S12, *Alocinma longicornis*; S13, *Branchiodrilus hortensis*; S14, *Cricotopus bicinctus*; S15, *Limnoperna fortunei*; S16, *Procladius* sp.; S17, *Physa* sp.; S18, *Tanypus chinensis*; S19, *Radix swinhoei*; S20, *Ceratopsyche* sp.; S21, *Heptagenia* sp.; S22, *Chironomus plumosus*; S23, *Ploypedilum scalaenum*; S24, *Propsilocerus akamusi*; S25, *Acuticosta chinensis*; S26, *Anodonta woodiana pacifica*; S27, *Anodonta woodiana elliptica*; S28, *Glossiphonia* sp.; S29, *Dicrotendipus lobifer*; S30, *Semisulcospira libertina*; S31, *Stenothyra glabra*; S32, *Unio douglasiae*; S33, *Glossiphonia complanata*; S34, *Glyptotendipes tokunaga*; S35, *Nemertea* sp.; S42, *Cricotopus sylvestris*; S43, *Tanytarsus chinyensis*; S44, *Laccophilus* sp.; S45, *Rhaphium* sp.; S46, *Glyptotendipes pallens*; S47, *Lamelligomphus* sp.; S48, *Helobdella fusca*; S49, *Glossiphonia lata*; S50, *Aciagrion* sp.; S51, *Baetis* sp.; S52, *Harnischia fuscimana*; S53, *Ephemera orientalis*; S64, *Cryptochironomus* sp.; S65, *Cercion* sp.; S66, *Calopteryx* sp.; S67, *Holorusia* sp.; S68, *Brachythemis* sp.; S65, *Cercion* sp.; S66, *Calopteryx* sp.; S67, *Holorusia* sp.; S68, *Brachythemis* sp.

marginal  $R^2 = 0.63$ , p < 0.001; richness marginal  $R^2 = 0.73$ , p < 0.001). Decreasing Oligochaeta relative abundance was associated with investment value of restoration projects both targeting agricultural and industrial wastewater (marginal  $R^2 = 0.47$ , p < 0.001). Increasing  $\beta_{\text{SOR}}$  was correlated positively with investment for all three project categories targeting pollution sources (marginal  $R^2 = 0.32$ , p < 0.001). There was evidence for significant interaction between the response ratio of impervious surface area and investments of different restoration project

categories and biological index responses. For example, poor biological responses were associated with the growth of the impervious surface area (p < 0.001, Fig. 7), but these effects were overcome where restoration projects were large but impervious area increased minimally.

#### 4. Discussion

This study has provided new insights to understand the effectiveness



Fig. 7. Interaction effect between the response ratio of impervious surface area (rISA) and investments of different restoration project categories on the response ratio of  $NH_4^4$ -N, TN, TP in natural waterbodies in Yixing. Figures are given when the interaction effect was significant (p < 0.05).

of catchment-scale restoration towards increasing water quality and biodiversity in rivers of China, building on knowledge from previous studies from a variety of ecosystems in other parts of the world (Benayas et al., 2009; Crouzeilles et al., 2016). Using a large data set comprising hundreds of different aquatic ecosystem restoration projects undertaken over the last two decades in a large urban district of China, we showed that implementation of large-scale restoration projects can, to some extent, mitigate the environmental degradation as a result of economic boom. In Yixing, recovery occurred despite ongoing rapid economic growth and urbanization, although it should be noted that the impervious surface area reached only 10.15% at the bottom end of Schueler (1994) 10–25% range for significant impacts on water quality. Further

urbanization may therefore negate the positive aspects of restoration observed to date.

Restoration led to decreases in indicators of stress, notably concentrations of main nutrients ( $NH_4^+$ -N, TN and TP) and Oligochaeta relative abundance, whereas taxonomic richness and Shannon-Weiner diversity of benthic macroinvertebrate were significantly higher across the Yixing river network. These general findings for macroinvertebrate community and water quality recovery are supported by studies which found a significant positive effect of restoration on the organism groups and water quality (Kong et al., 2020). External inputs of organic pollution from sewage have been reduced from the catchment, and measures such as sediment dredging, cyanobacteria salvage, etc. have been

implemented to reduce the internal nutrient loading. This combination of approaches has allowed dissolved oxygen concentrations to rise, gradually improving aquatic habitat and enhancing aquatic biodiversity (Mason, 2002).

In contrast, the overall Hilsenhoff FBI showed no significant difference between degraded (2007) and restored (2017) years. Despite the enormous investment in restoration, there were 25 sites showing increases in Hilsenhoff FBI scores, due to some higher tolerance taxa still remaining, and taxonomic richness in 2007 being much lower (average: 2.16) than in 2017 (average: 8.58). During the 10 years of the study, Yixing has seen its GDP increase from 42.80 billion RMB in 2006 to 155.83 billion RMB in 2017 (Fig. 2), accompanied by 45% growth of artificial surfaces (Fig. S1 and Table S3). The effects of urbanization (hydromorphology and hydrological alteration, run off pollution) are likely to have suppressed the level of biotic recovery of freshwater macroinvertebrates that may have occurred from restoration efforts in isolation by increasing the role of other stressors (Gál et al., 2019). Despite this urban cover growth, water quality of Chinese inland waters has clearly improved generally over recent decades with restoration efforts (Zhou et al., 2017).

On the other hand, we found that the response of biotic indices to restoration projects appeared to lag behind nutrients (NH<sup>+</sup><sub>4</sub>-N, TN and TP), with the standardized responses of nutrients being greater than those of biotic indices. However, compared to the degraded time-period (2007), some species that are sensitive to anthropogenic pressures (low tolerance values) recolonized after restoration (2017) including the river flies Heptagenia sp.and Ephemera orientalis. These observed increases in the Heptageniidae are in line with Pedersen et al. (2007) who reported they increased significantly in abundance after a short-term restoration (three years) at the Skjern River reaches, Denmark. The composition of benthic invertebrate communities differed significantly between degraded (2007) and restored (2017) periods. Eight species cumulatively contributed > 70% to the dissimilarity between restored (2017) and degraded (2007) invertebrate communities (Table S5). Limnodrilus hoffmeisteri (turbid worms) and Branchiura sowerbyi (crustaceans) both decreased more in restored than in degraded rivers. These species are widely used as an indicator of organic pollution throughout China (Gorni et al., 2018), thus their decreasing abundance provides important ecological evidence for restoration success alongside the water quality improvements. However, Limnodrilus hoffmeisteri was still a co-dominant species in some sampling sites in Yixing river network both in degraded and restored time-periods; something not surprising as they are widely distributed throughout global freshwater ecosystems (Armendáriz and César, 2001). In contrast, snails and clams such as Bellamya aeruginosa and Corbicula fluminea increased more in restored than in degraded. Recovery of these native snail and bivalve populations can be expected to further help improve the water quality given their roles as deposit or filter feeders that remove particulates (Zhang et al., 2014). The relative abundance of snails like Bellamya aeruginosa increased in > 20 sites over time, most likely because some native snails have been reintroduced by restoration activities in attempts to enhance algal removal. While Bellamya aeruginosa and Corbicula fluminea are common species which are widely distributed in eutrophic shallow lakes in China (Zhu et al., 2013). Although biological indices appear to lag behind abiotic indices like nutrients, sampling frequency limited our ability to elucidate more clearly the relationship between these indicators.

Even though the response ratio of taxonomic richness and Shannon diversity of benthic macroinvertebrate was significantly higher in restored (2017) than in degraded (2007) aquatic ecosystems, taxonomic richness and Shannon diversity of benthic invertebrate only showed significant positive correlation with the increasing project investment on sanitary sewage removal, the decline of  $\rm NH_4^+-N$  concentrations in Yixing river network, and the interaction effect between the response ratio of impervious surface area and project investment on agricultural

sewage removal. The muted improvements of biological indices may be due to two reasons: (i) restoration measures on pollution sink mainly include dredging of contaminated sediments, which will negatively affect the habitat of benthic macroinvertebrates; (ii) water quality in Yixing is improved but still not to a high level, and hydromorphological alterations remain throughout the catchment, limiting recovery potential. Additionally, only 14.4% of investments were targeted at pollution sinks in Yixing during 2007 to 2017. Agricultural (especially for livestock breeding and aquaculture) and domestic sewage as main sources of  $NH_4^+$ -N pollution have, however, been addressed significantly by the restoration program (Oita et al., 2016), as illustrated by the correlation between macroinvertebrate taxonomic richness, Shannon diversity and decreasing  $NH_4^+$ -N concentration, supported by the findings of Yi et al. (2018).

Although several factors can influence the outcomes of restoration, investment structure and complementarity amongst different restoration project categories appears as key factors of restoration success. Our results showed that some project categories have a disproportionate effect on nutrient recovery. Even though projects targeting both pollution sources and pollution sinks overall contributed positively to decreases in  $\rm NH_4^+-N$ , TP, TN concentrations in Yixing river network, we have showed that:

- (1) The same investment amount on restoration projects targeting pollution sources can lead to greater decreases in  $NH_4^+$ -N and TP in comparison to equivalent spending on targeting pollution sinks. This result might be driven by effective and timely actions on pollution sources, where nutrients are concentrated prior to dilution and dissipation among water and sediments in rivers and lakes. Thus, projects targeting pollution sources are the most effective way to prevent and decrease water eutrophication by  $NH_4^+$ -N and TP (Wurtsbaugh et al., 2019).
- (2) The same investment amount on restoration projects targeting agricultural sewage (especially for livestock breeding and aquaculture) can lead to greater decreases in  $NH_4^+$ -N and TP (especially for  $NH_4^+$ -N) in comparison to those spent on targeting domestic sewage. This result might also be driven by frequent agricultural activities that are one of the main nitrogen sources of the Taihu basin (Liu et al., 2020), and domestic sewage which is one of main pollution sources of TP (Qin et al., 2007).
- (3) Decreases in NH<sup>+</sup><sub>4</sub>-N and TP concentrations showed slightly negative correlation with increasing investment on restoration projects targeting pollution sink and industry waste water. This could be because additional investment in restoration projects targeting pollution sinks and industry waste water could not lead to removal of more NH<sup>+</sup><sub>4</sub>-N and TP in a proportionate way. Furthermore, there are many restoration projects on pollution sinks (except for dredging) that do not aim to remove nutrients in a direct way (Bai et al., 2020).
- (4) Decreases in TN concentrations in the Taihu river network were correlated with the increasing investment on restoration projects targeting both pollution source and pollution sink. However, deceases in TN concentrations were correlated weakly with the increasing investment on restoration projects on domestic sewage. Additional investment in sanitary sewage treatment plants may therefore not lead to removal of more TN from the waste water in a proportionate way.

The time elapsed since restoration began was also an important ecological driver underpinning ecosystem restoration success (Crouzeilles et al., 2016). Different restoration projects start on different dates by continuous planning, and so the restoration project investments towards the end of our study period may not have had a chance to exhibit their full impact. We can explore the different timelines for abiotic and biotic indices recovery after restoration in the future, if river

management agencies invest in long time-scale ater quality and biomonitoring data.

Overall, our results demonstrate that (i) investments in environmental restoration projects improved water quality and biodiversity despite urban growth (Fig. 7); (ii) investments in source control had a stronger impact on water quality than investments in restoring sinks (Fig. 3); (iii) investments in sink water quality control improved nutrient levels, albeit not as strong as investments in source controls (Fig. 3). Stakeholders should therefore plan carefully the allocation of resources and money when restoring aquatic ecosystems. Studies such as this evaluation of river catchment restoration in SE China have an important role in building the necessary trust in restoration projects for that to happen (Metcalf et al., 2015).

# 5. Conclusion

Our analysis demonstrates that, despite the unstopped expansion of urbanization, nutrient concentrations and biological indices of benthic invertebrate have improved significantly across most of Yixing catchment as a result of restoration works executed over the study period. Improvements were contingent to the type of restoration project, with some restoration approach showing disproportionate effects on response rates of ecosystem indices and projects targeting pollution-sources leading to the clearest improvements compared with those remediating pollution-sinks. However, in some locations, the recovery of biotic communities appears to lag behind that of nutrients (e.g., nitrogen and phosphorus), likely reflecting the longer time required by long-distance recolonization routes for invertebrates given the level of pre-restoration degradation of the catchment. Overall, our study suggests that ecological damage caused by recent rapid economic development could potentially be mitigated by the combined effect of massive restoration investments synchronized across whole catchments, although these effects can be expected to be muted if urbanization continues apace at the same time.

# Data availability statement

Data are available from the Dryad Digital Repository:  $\langle https://doi. org/10.5061/dryad.547d7wm8f \rangle$  (Fu et al. 2021).

# CRediT authorship contribution statement

Hong Fu: Methodology, Formal analysis, Resources, Writing – original draft, Writing – review & editing. Pierre Gaüzère: Methodology. Jorge García Molinos: Writing – review & editing. Peiyu Zhang: Writing – review & editing. Huan Zhang: Writing – review & editing. Min Zhang: Writing – review & editing, Supervision. Yuan Niu: Formal analysis, Resources. Hui Yu: Formal analysis, Resources. Lee E. Brown: Methodology, Writing – review & editing. Jun Xu: Conceptualization, Methodology, Formal analysis, Resources, Writing – review & editing, Supervision.

# **Declaration of Competing Interest**

The authors declare that they have no known competing financial interests or personal relationships that could have appeared to influence the work reported in this paper.

# Acknowledgments

This research was supported by the Water Pollution Control and Management Project of China (Grant No. 2018ZX07208005), the International Cooperation Project of the Chinese Academy of Sciences (Grant No. 152342KYSB20190025), the National Natural Science Foundations of China (Grant No. 31872687), and the China Scholarship Council. LEB's contribution was funded by the University of Leeds. JGM was supported by JSPS KAKENHI Grant No. 19H04314.

# Supplementary materials

Supplementary material associated with this article can be found, in the online version, at doi:10.1016/j.watres.2021.117587.

# References

- Armendáriz, L.C., César, I.I., 2001. The distribution and ecology of littoral oligochaeta and aphanoneura (Annelida) of the natural and historical reserve of Isla Martín García, Río de la Plata River. Argent. Hydrobiol. 463 (1), 207–216.
- Bai, G., Zhang, Y., Yan, P., Yan, W., Kong, L., Wang, L., Wang, C., Liu, Z., Liu, B., Ma, J., 2020. Spatial and seasonal variation of water parameters, sediment properties, and submerged macrophytes after ecological restoration in a long-term (6 year) study in Hangzhou west lake in China: submerged macrophyte distribution influenced by environmental variables. Water Res. 186, 116379.
- Baselga, A., 2012. The relationship between species replacement, dissimilarity derived from nestedness, and nestedness. Glob. Ecol. Biogeogr. 21 (12), 1223–1232.
- Baselga, A., Orme, C.D.L., 2012. Betapart: an R package for the study of beta diversity. Methods Ecol. Evol. 3 (5), 808–812.
- Benayas, J.M.R., Newton, A.C., Diaz, A., Bullock, J.M., 2009. Enhancement of biodiversity and ecosystem services by ecological restoration: a-analysis. Science 325 (5944), 1121–1124.
- Birk, S., Chapman, D., Carvalho, L., Spears, B.M., Andersen, E., Argillier, C., Auer, S., Baattrup-Pedersen, A., Banin, L., Beklioğlu, M., 2020. Impacts of multiple stressors on freshwater biota across spatial scales and ecosystems. Nat. Ecol. Evol. 4 (8), 1060–1068.
- State Environmental Protection Agency of the People's Republic of China (2002). Discharge standard of pollutants for municipal wastewater treatment plant, GB 18918-2002. Retrieved from http://sthj.hd.gov.cn/main/detail/57192.
- Chinese Research Academy of Environmental Sciences (2020). Variation of nitrogen and phosphorus pollution sources and load in Taihu basin in recent 10 years, 2018ZX07208-005. Beijing.
- Crawley, M.J., 2002. Statistical computingan introduction to data analysis using S-Plus. No. 001.6424 C73.
- Crouzeilles, R., Curran, M., Ferreira, M.S., Lindenmayer, D.B., Grelle, C.E., Rey Benayas, J.M., 2016. A global meta-analysis on the ecological drivers of forest restoration. Nat. Commun. 7, 11666.
- Deng, X., Xu, Y., Han, L., Yu, Z., Yang, M., Pan, G., 2015. Assessment of river health based on an improved entropy-based fuzzy matter-element model in the Taihu Plain, China. Ecol. Indic. 57, 85–95.
- Gál, B., Szivák, I., Heino, J., Schmera, D., 2019. The effect of urbanization on freshwater macroinvertebrates-knowledge gaps and future research directions. Ecol. Indic. 104, 357–364.
- Gaüzère, P., Jiguet, F., Devictor, V., Loyola, R., 2016. Can protected areas mitigate the impacts of climate change on bird's species and communities? Divers. Distrib. 22 (6), 625–637.
- Gauzere, P., Prince, K., Devictor, V., 2017. Where do they go? The effects of topography and habitat diversity on reducing climatic debt in birds. Glob Chang Biol. 23 (6), 2218–2229.
- Gorni, G.R., Sanches, A.d.O., Colombo-Corbi, V., Corbi, J.J., 2018. Oligochaeta (Annelida: clitellata) in the Juruena river, MT, Brazil: species indicators of substrate types. Biota Neotropica 18 (4), 1–9.
- Guan, X., Wei, H., Lu, S., Dai, Q., Su, H., 2018. Assessment on the urbanization strategy in China: achievements, challenges and reflections. Habitat Int. 71, 97–109.
- Hilsenhoff, W.L., 1988. Rapid field assessment of organic pollution with a family-level biotic index. J. N. Am. Benthol. Soc. 7 (1), 65–68.
- Huang, Q., Gao, J., Cai, Y., Yin, H., Gao, Y., Zhao, J., Liu, L., Huang, J., 2015. Development and application of benthic macroinvertebrate-based multimetric indices for the assessment of streams and rivers in the Taihu Basin. China. Ecol. Indic. 48, 649–659.
- Imai, H., 2018. China's rapid growth and real exchange rate appreciation: measuring the Balassa-samuelson effect. J. Asian Econ. 54, 39–52.
- Department of Ecology and Environment of Jiangsu Province (2007). Discharge standard of main water pollutions for municipal wastewater treatment plant & key industries of Taihu area, DB 32 1072-2007. Jiangsu. Retrieved from http://hbt.jiangsu.gov. cn/art/2007/10/12/art 2483 4180299.html.
- Department of Ecology and Environment of Jiangsu Province (2004). Discharge Standard of water pollutants for dyeing and finishing of textile industry. DB 32/670-2004. Jiangsu. Retrieved from https://max.book118.com/html/2018/0218/153742083.sh tm.
- Kail, J., Brabecb, K., Poppec, M., Januschkea, K., 2015. The effect of river restoration on fish, macroinvertebrates and aquatic macrophytes: a meta-analysis. Ecol. Indic. 58, 311–321.
- Kong, X., Tian, K., Jia, Y., He, Z., Song, S., He, X., Xiang, C., An, S., Tian, X., 2020. Ecological improvement by restoration on the Jialu river: water quality, species richness and distribution. Mar. Freshw. Res. 71 (12), 1602–1615.
- Jiangsu Development and Reform Commission (2014). Water environment comprehensive management of Taihu basin, 2013–2684. Jiangsu. Retrieved from htt p://fzggw.jiangsu.gov.cn/art/2014/1/22/art\_284\_6647960.html.

Jiangsu Development and Reform Commission (2008). Water environment comprehensive management of Taihu basin. Jiangsu. Retrieved from https://jz.doci n.com/p-295389.html.

Zhu, X., Zhang, Y.W., Wang, H., et al., 2008. Study on ecological dredging of rivers and lakes in Wuxi. The first river-sea coastal ecological protection and environmental governance, river dredging technology exchange seminar 326–332.

- Y. Li (2017). Long-term operation efficiency of surface flow constructed wetland for treatment of slightly polluted river water. Master Degree, Southeast University.
- Liu, D., Yu, J., Zhong, J., Zhong, W., Fan, C., 2016a. Characteristics of nitrogen and phosphorus loading and migration in typical river networks in Taihu lake basin. China Environ. Sci. 36 (01), 125–132.
- Liu, G., Bao, X., Wu, T., Han, S., Xiao, M., Yan, S., Zhou, Q., 2015. Purification of water in Zhushan Bay of Taihu Lake with water hyacinth ecological engineering. J. Agro Environ. Sci. 000 (2), 352–360.
- Liu, L., Dong, Y., Kong, M., Zhou, J., Zhao, H., Tang, Z., Zhang, M., Wang, Z., 2020. Insights into the long-term pollution trends and sources contributions in Lake Taihu, China using multi-statistic analyzes models. Chemosphere 242, 125272.
- Liu, S., Dong, Y., Cheng, F., Coxixo, A., Hou, X., 2016b. Practices and opportunities of ecosystem service studies for ecological restoration in China. Sustain. Sci. 11 (6), 935–944.
- Luo, Y., Zhao, Y., Yang, K., Chen, K., Pan, M., Zhou, X., 2018. Dianchi lake watershed impervious surface area dynamics and their impact on lake water quality from 1988 to 2017. Environ. Sci. Pollut. Res. 25 (29), 29643–29653.Mason, C.F., 2002. Biology of Freshwater Pollution. Pearson Education.

Mason, C.F., 2002. Biology of Presinvater Pollution. Pearson Education. McGarigal, K., Cushman, S., Ene, E., 2012. Spatial pattern analysis program for categorical and continuous maps. Computer software program produced by the authors at the University of Massachusetts, Amherst. FRAGSTATS v4. Available at the following web site: http://www.umass.edu/landeco/research/fragstats/fragstats html

Metcalf, E.C., Mohr, J., Yung, L., Metcalf, P., Craig, D., 2015. The role of trust in restoration success: public engagement and temporal and spatial scale in a complex social-ecological system. Restor. Ecol. 23 (3), 315–324.

Morse, J., Yang, L., Tian, L., 1994. Aquatic insects of China Useful for Monitoring Water Quality Nanjing. The University of Chicago Press, People's Republic of China.

- National Environmental Protection Bureau, 2002. Standard Methods for The Examination of Water and Wastewater (Version 4). China Environmental Science Publish Press, Beijing, China.
- Nzengya, D.M., Wishitemi, B.E.L., 2000. Dynamics of benthic macroinvertebrates in created wetlands receiving wastewater. Int. J. Environ. Stud. 57 (4), 419–435.
- Oita, A., Malik, A., Kanemoto, K., Geschke, A., Nishijima, S., Lenzen, M., 2016. Substantial nitrogen pollution embedded in international trade. Nat. Geosci. 9 (2), 111–115.
- Pan, X.Z., Zhao, Q.G., 2007. Measurement of urbanization process and the paddy soil loss in Yixing city, China between 1949 and 2000. Catena 69 (1), 65–73.

Pedersen, M.L., Friberg, N., Skriver, J., Baattrup-Pedersen, A., Larsen, S.E., 2007. Restoration of Skjern river and its valley-short-term effects on river habitats, macrophytes and macroinvertebrates. Ecol. Eng. 30 (2), 145–156.

 Qin, B., Xu, P., Wu, Q., Luo, L., Zhang, Y., 2007. Eutrophication of Shallow Lakes with Special Reference to Lake Taihu, China. Springer, pp. 3–14.
 Qin, B., Zhu, G., Zhang, L., Luo, L., Gao, G., Binghe, G., 2005. Models of endogenous

- Qin, B., Zhu, G., Zhang, L., Luo, L., Gao, G., Binghe, G., 2005. Models of endogenous nutrient release from sediments in large shallow lakes and their estimation methods. Sci. China Ser. D Earth Sci. 35 (S2), 33–44.
- Qin, C.Y., Zhou, J., Cao, Y., Zhang, Y., Hughes, R.M., Wang, B.X., 2014. Quantitative tolerance values for common stream benthic macroinvertebrates in the Yangtze river delta, Eastern China. Environ. Monit. Assess. 186 (9), 5883–5895.

Ramchunder, S.J., Brown, L.E., Holden, J., 2012. Catchment-scale peatland restoration benefits stream ecosystem biodiversity. J. Appl. Ecol. 49 (1), 182–191.

Schueler, T., 1994. The importance of imperviousness. Watershed Prot. Tech. 1 (3), 100–101.

Sévêque, A., Gentle, L.K., López-Bao, J.V., Yarnell, R.W., Uzal, A., 2020. Human disturbance has contrasting effects on niche partitioning within carnivore communities. Biol. Rev. 95 (6), 1689–1705.

Simpson, E.H., 1949. Measurement of diversity. Nature 163 (4148), 688–688. Song, W., Deng, X., 2017. Land-use/land-cover change and ecosystem service provision

in China. Sci. Total Environ. 576, 705–719.

- Ministry of Environmental Protection of the People's Republic of China (2010). Relevant requirements of accounting for main water pollutions emission reduction in 2010. Beijing. Retrieved from http://www.mee.gov.cn/gkml/hbb/bwj/201206/t2012 0605\_230944.htm.
- Sun, H., Zhang, J., Shan, Q., Wang, Q., Chen, G., Wu, H., 2015. Preliminary analysis of the source reduced and sink increased for agricultural non-point source pollution by forest in Lake Taihu watershed: a case study of shelter belt in Yixing City. J. Lake Sci. 27 (02), 227–233.
- Wang, B.X., Yang, L.F., 2004. A study on tolerance values of benthic macroinvertebrate taxa in eastern China. Acta Ecol. Sin. 24 (12), 2768–2775.
- Wang, H., 2002. Studies on Taxonomy, Distribution and Ecology of Microdrile Oligochaetes of China, with Descriptions of Two New Species from the Vicinity of the Great Wall Station of China, Antarctica. Higher Education (HEP), Beijing.
- Wang, Q., Zhang, Q., Wu, Y., Wang, X.C., 2017. Physicochemical conditions and properties of particles in urban runoff and rivers: implications for runoff pollution. Chemosphere 173, 318–325.
- Wang, W., Hu, M., Tang, X., 2010. Calculation of discharge coefficient of rural domestic sewage in Taihu basin. J. Ecol. Rural. Environ. 26 (06), 616–621.
- Wurtsbaugh, W.A., Paerl, H.W., Dodds, W.K., 2019. Nutrients, eutrophication and harmful algal blooms along the freshwater to marine continuum. Wiley Interdiscip. Rev. Water 6 (5), e1373.
- Xie, F., Cao, L., Wang, Z., Ming, X.H., Zhao, Y.W., et al., 2014. Estimation of pollutant production and discharge from livestock and poultry industries in Taihu lake region. Bull Soil Water Conserv 34 (2), 128–133.
- Yang, K., Pan, M., Luo, Y., Chen, K., Zhao, Y., Zhou, X., 2019. A time-series analysis of urbanization-induced impervious surface area extent in the Dianchi lake watershed from 1988 to 2017. Int. J. Remote Sens. 40 (2), 573–592.
- Yi, Q., Chen, Q., Hu, L., Shi, W., 2017. Tracking nitrogen sources, transformation, and transport at a basin scale with complex plain river networks. Environ. Sci. Technol. 51 (10), 5396–5403.
- Yi, Y., Sun, J., Yang, Y., Zhou, Y., Tang, C., Wang, X., Yang, Z., 2018. Habitat suitability evaluation of a benthic macroinvertebrate community in a shallow lake. Ecol. Indic. 90, 451–459.

Zhang, N., Li, G., Yu, J., Ding, M., Xu, L., 2009. Preliminary analysis of the main characteristics of cyanobacteria bloom in Taihu lake. Environ. Monit. China 25 (01), 71–74.

- Zhang, P., Shao, G., Zhao, G., Le Master, D.C., Parker, G.R., Dunning, J.B., Li, Q., 2000. China's forest policy for the 21st century. Science 288 (5474), 2135–2136.
- Zhang, X., Liu, Z., Jeppesen, E., Taylor, W.D., 2014. Effects of deposit-feeding Tubificid worms and filter-feeding bivalves on benthic-pelagic coupling: implications for the restoration of eutrophic shallow lakes. Water Res. 50, 135–146.
- Zhao, D., 2010. Study on the Purification of Vallisneria Spinulosa for Aquatic Plants to Different Eutrophic Water. Master thesis. Anhui Agricultural University, Anhui.

Zhao, J., Yang, Y., Zhao, Q., Zhao, Z., 2017. Effects of ecological restoration projects on changes in land cover: a case study on the Loess Plateau in China. Sci. Rep. 7, 44496.

- Zhou, B., 2012. The Effect of Cyanobacteria Salvage on Nitrogen and Phosphorus in Water and Algae Growth. Master thesis. Nanjing Normal University, Nanjing.
- Zhou, Y., Ma, J., Zhang, Y., Qin, B., Jeppesen, E., Shi, K., Brookes, J.D., Spencer, R.G.M., Zhu, G., Gao, G., 2017. Improving water quality in China: environmental investment pays dividends. Water Res. 118, 152–159.
- Zhu, J., Lu, K., Liu, X., 2013. Can the freshwater snail Bellamya aeruginosa (Mollusca) affect phytoplankton community and water quality? Hydrobiologia 707 (1), 147–157.