

# Decoding Benthic Macroinvertebrate Communities in Freshwater Ecosystems Leveraging Environmental DNA

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## Abstract

Benthic macroinvertebrates are key indicator groups within freshwater ecosystems, with their community being closely tied to ecosystem functioning. Environmental DNA (eDNA) technology, with its high sensitivity and non-invasive nature, provides a promising tool for studying the spatiotemporal dynamics of benthic macroinvertebrate communities, their responses to anthropogenic disturbances, and the mechanisms governing community assembly. However, current eDNA-based research on freshwater benthic communities largely concentrates on optimizing sampling and detection methods, along with environmental monitoring applications. Broader ecological investigations using eDNA data remain fragmented and limited. In particular, it remains unresolved whether benthic community ecology theories derived from morphological classification can be reliably reconstructed from eDNA-based molecular signals. We systematically summarize the application of eDNA technology in benthic macroinvertebrate ecology, encompassing diversity assessment, spatiotemporal community dynamics, cascading effects of human-induced disturbances, and multi-trophic interaction networks. Although significant empirical progress has been made, challenges persist, including limited spatiotemporal coverage and methodological constraints. Future developments should prioritize the establishment of long-term monitoring networks, integration of multidimensional data, and deeper engagement with community ecology theories. Such advances will help shift eDNA-based research from descriptive to mechanistic understanding, thereby offering scientific support for the precise management and restoration of river ecosystems.

**Keywords:** Macroinvertebrates; eDNA; Community ecology; Human disturbance; Multitrophic interactions

## 29    **1. Introduction**

30            Freshwater ecosystems are fundamental role to maintaining biodiversity and supporting human well-being  
31    (Lynch et al., 2023). However, their biological integrity is increasingly threatened by multiple stressors, including  
32    climate change, pollution, overexploitation of resources, and the spread of invasive species (Perry et al., 2024; Sayer  
33    et al., 2025). Benthic macroinvertebrates are integral components of these ecosystems, performing irreplaceable  
34    functions such as detritus decomposition (Tank et al., 2010; Yue et al., 2022). Their diversity is undergoing  
35    significant decline under intense anthropogenic disturbances, including urbanization, agricultural expansion,  
36    wastewater discharge, aquaculture, and forestry development (Li et al. 2015; Dudgeon and Strayer, 2025; Keck et  
37    al., 2025). Owing to their limited mobility and high sensitivity to environmental change, they are widely used as  
38    bioindicators across global monitoring programs (Dudgeon et al., 2006; Morse et al., 2007; Thomsen et al., 2011;  
39    Jin et al., 2025). Nevertheless, conventional assessments of benthic macroinvertebrate biodiversity which rely on  
40    field sampling, sorting, and morphological identification, are time-consuming, labor-intensive, and requires  
41    considerable taxonomic expertise (Johnson et al., 1993; Sumudumali and Jayawardana, 2021). This approach often  
42    entails extensive specimen sorting and can yield inaccurate results due to difficulties in distinguishing  
43    morphologically similar taxa, juvenile stages, and rare species. Ntislidou et al. (2020) highlighted how human error  
44    during identification can affect water quality assessments, while Haase et al. (2010) reported that 29% of specimens  
45    were overlooked during sorting, identifications varied by more than 30% among analysts, and 16% of samples led  
46    to discrepant ecological evaluation. Such methodological limitations may seriously compromise the effectiveness  
47    of river management and restoration plans.

48            In contrast, environmental DNA (eDNA) technology offers considerable technical advantages that are  
49    advancing the study of benthic macroinvertebrate biodiversity (Deiner et al., 2017; Múrria et al., 2024). Its non-  
50    invasive nature minimizes disturbance to both target species and their habitats, while its cost-effectiveness across  
51    temporal and spatial scales facilitates large-scale monitoring of biodiversity dynamics. The high sensitivity of eDNA

further allows for effective detection of rare and cryptic species, as well as early-stages biological invasions (Beng and Corlett, 2020; Takahashi et al., 2023; Altermatt et al., 2025; Sander et al., 2025; Yates et al., 2025). Consequently, eDNA is increasingly regarded as the most promising tools for efficient, large-scale species monitoring in aquatic ecosystems.

Environmental DNA (eDNA) studies are conducted more frequently in aquatic ecosystems, such as rivers, lakes, and marine waters, than in terrestrial environments, largely due to the relative ease of sample collection. Among these, freshwater ecosystems dominate the eDNA literature, representing approximately 65% of published studies, whereas marine systems account for about 25% and the remaining 10% pertain to other or multiple habitat types (Takahashi et al., 2023; Altermatt et al., 2025). Notably, more than half of these studies (around 52%) focus on fish species, while benthic macroinvertebrate communities have received comparatively less attention.

In this review, we summarize the application of eDNA technology to key ecological questions regarding benthic macroinvertebrate communities in freshwater ecosystems, including their spatiotemporal dynamics, resilience to anthropogenic disturbance, and multi-trophic interactions. We also synthesize recent technological and methodological advances, outline major international initiatives, and provide a comprehensive overview of published eDNA-based studies in benthic macroinvertebrate community ecology. Current challenges and promising future research directions are discussed. We anticipate that this work will help advance eDNA from a primarily descriptive tool toward a more mechanistic analytical framework, thereby providing actionable insights to support the monitoring and management of freshwater ecosystems under global change.

## **2. Towards Optimized eDNA-based in Benthic Community Ecology: Sampling Strategies and Methodological Validation**

eDNA originates from a variety of biological sources, such as skin, saliva, mucus, secretions, urine, blood, feces, gametes, roots, leaves, pollen, and decomposing tissues (Bohmann et al., 2014). It exists both as extracellular

74 molecules and within cellular debris. The technology enables non-invasive species detection and community  
75 assessment by capturing genetic material shed or released by organisms, followed by high-throughput sequencing  
76 (HTS) and bioinformatics analysis (Thomsen and Willerslev, 2015; Harrison et al., 2019; Gu et al., 2024). Recently,  
77 Sahu et al. (2025) provided an updated overview of the standard eDNA workflow, which outlines the complete  
78 pipeline from field sampling and DNA extraction/purification through PCR amplification, HTS, bioinformatic  
79 processing, to final taxonomic assignment.

80         The choice of environmental sample type is crucial in eDNA metabarcoding, as it directly influences taxon  
81 detection and the resulting community profile (**Table1**). Water samples are widely used for large-scale biodiversity  
82 assessment due to their ease of collection (Vourka et al., 2023); however, local benthic community composition may  
83 be underestimated because of signal dilution and hydrological transport (Wang et al., 2021). In contrast, ethanol-  
84 preserved samples (e.g., from kick-net sampling) show higher sensitivity for sessile or sedentary taxa and better  
85 reflect local biodiversity (Erdozain et al., 2019; Martins et al., 2021), though they can underrepresent rare, small, or  
86 heavily sclerotized species (Zizka et al., 2019). Sediment samples accumulate DNA from benthic organisms and are  
87 less influenced by transport, making them particularly useful for detecting arthropods, annelids, and mollusks at  
88 local scale (Ji et al., 2022). Their high spatial heterogeneity, however, requires careful sampling design and adequate  
89 replication to ensure representativeness (Sakata et al., 2021). Biofilm samples also hold promise for monitoring  
90 species in complex or structured habitats due to their strong DNA retention capacity (Rivera et al., 2021). Recent  
91 evidence suggests that combining water and detritus samples currently represents the most effective strategy for  
92 assessing riverine macroinvertebrate diversity. This integrated approach leverages complementary taxonomic  
93 coverage and outperforms single-sample methods and other combinations in capturing species richness and  
94 community composition (Shi et al., 2025).

95 **Table1. Comparison of Common eDNA Sample Types in Freshwater Ecosystems**

Sample Type	Advantages	Limitations	Recommended Applications
Water	Captures broad-scale diversity patterns; easy to collect and process	Prone to dilution; lower resolution for site-specific communities	Regional-scale biodiversity assessment
Ethanol	Accurately reflects local community composition; effective for sessile taxa	May underrepresent rare or small-bodied taxa; potential sampling biases	Local biotic integrity evaluation
Sediment	High DNA concentration; suitable for detecting benthic organisms	High spatial heterogeneity; requires intensive replication for representativeness	Local community structure assessment and taxonomic screening
Biofilm	Long DNA retention time; effective in slow-flow or attached habitats	Methodologically novel; standardization protocols still under development	Monitoring in complex habitats or where conventional methods are limited
Detritus	Superior DNA retention; enhanced detection of key taxa (e.g., Oligochaeta, Diptera)	High operational complexity; limited spatial representativeness	Combined with water samples

96           Although water sampling remains prevalent in eDNA studies due to its operational simplicity, conducting  
97 controlled comparisons with traditional morphological methods are still essential to validate the accuracy and  
98 reliability of eDNA-based assessments (**Table 2**).

99 **Table 2. Representative studies comparing benthic macroinvertebrate richness estimates with traditional sampling or historical**  
100 **data for a geographic location to that of eDNA**

No.	eDNA Sample Types	Traditional sampling method	Habitat	eDNA efficacy finding	References
1	Water	Dip net	Pond water	●	(Thomsen et al., 2011)
2	Water	Dip net	Ditches	↑	(Beentjes et al., 2022)
3	Water	Dip net	Ditches	↑	(van der Plas et al.,2025b)
4	Water	Dip net	River	●	(Marshall and Stepien, 2020)
5	Water	Kick-sweep	Lake and river	●	(Mächler et al., 2014)
6	Water	Kick-sweep	Hypersaline lake	↑	(Campbell et al., 2023)
7	Water	Kick-sweep	Hypersaline lake	●	(Saccò et al., 2025)
8	Water	Kick-sweep	River	●	(Fernández et al., 2019)
9	Water	Kick-sweep	River	↑	(Mächler et al., 2019)
10	Water	Kick-sweep	River	●	(Brantschen et al., 2021)
11	Water	Kick-sweep	River	↑	(Pereira-da-Conceicao et al., 2021)

No.	eDNA Sample Types	Traditional sampling method	Habitat	eDNA efficacy finding	References
12	Water	Kick-sweep	River	↑	(Seymour et al., 2021)
13	Water	Kick-sweep	River	●	(Blackman et al., 2022a)
14	Water	Kick-sweep	River	●	(Keck et al., 2022b)
15	Water	Kick-sweep	Stream	●	(Reinholdt Jensen et al., 2021)
16	Water	Kick-sweep	Stream	●	(Gleason et al., 2021)
17	Water	D-net	River	●	(Múrria et al., 2024)
18	Water	D-net	River	●	(Huo et al., 2025)
19	Water	Hand net	Spring	●	(Blattner et al., 2021)
20	Water	Hand net	River	●	(Jeunen et al., 2022)
21	Water	Surber net	River	↑	(Uchida et al., 2020)
22	Water	Hess sampling	River	●	(Penaluna et al., 2026)
23	Water	Visual census, benthic trawling	River	↑	(Hata et al., 2022)
24	Water	Historical data	Lake	●	(Coghlan et al., 2021)
25	Water	Historical data	Lagoon	●	(Specchia et al., 2022)
26	Water	Historical data	River	●	(Wu et al., 2023)
27	Water	Historical data	River	●	(Qin et al., 2023)
28	Water	Historical data	River and stream	●	(Aunins et al., 2023)
29	Sediment	Surber net	River	↑	(Ji et al., 2022)
30	Sediment	Surber net	River	↑	(Zhou et al., 2022)
31	Sediment	D-net, Surber net	River	●	(Li et al., 2024)
32	Detritus	Kick-sweep	River	↑	(Ntislidou et al., 2023)
33	Detritus	D-net	River and stream	↓	(Jijón et al., 2025)
34	Ethanol, Water	D-net	Stream	↑	(Wang et al., 2021)
35	Ethanol, Water	Surber net, kick net	River	↑	(Chen et al., 2025)
36	Biofilm	Historical data	River	↑	(Rivera et al., 2021)

Notes: “●” indicates that the results are complementary; “↑” indicates that the eDNA method has obtained higher diversity data compared to the traditional sampling method; “↓” indicates that the eDNA method has obtained lower diversity data compared to the traditional sampling method.

In summary, future studies should prioritize multiple sample types, such as water, sediment, and ethanol-preserved materials, to capture complementary ecological niches and improve overall detection efficiency. Even when full integration of all sample types is not feasible, comparative analyses with traditional sampling methods can still offer a robust evaluation of eDNA’s strengths and limitations, thereby strengthening the credibility of the findings.

108 **3. Global Insights: eDNA in Setting Research Priorities and Response Indicators**

109       The assessment of ecosystem stress has long relied on diversity metrics, with  $\alpha$  diversity (intra-community  
110 diversity) and  $\beta$  diversity (inter-community differences) serving as core indicators in ecological research for decades  
111 (Daly et al., 2018; van der Plas et al., 2025a). These metrics quantify biodiversity patterns and are essential for detecting  
112 shifts in community composition across spatial and temporal scales. Here, we synthesize benthic macroinvertebrate  
113 community indices obtained from eDNA data across different countries, illustrating how environmental drivers, such  
114 as land use, water quality, and climate factors, influence biodiversity patterns in distinct regions (**Table 3**). This  
115 synthesis offers a comprehensive overview of current research trends in this field internationally.

116 **Table 3. Summary of benthic macroinvertebrate community ecological metrics derived from eDNA data, their associated**  
117 **environmental drivers, and representative application studies.**

No.	Country	Environmental Variables	Response Variables	References
1	Australia	Salinity	$\alpha$ -diversity, $\beta$ -diversity	(Campbell et al., 2023)
2	Belarus	Water temperature, Artificial canals	$\beta$ -diversity, Keystone species screening, Phylogenetic diversity	(Jeunen et al., 2022)
3	Brazil	Land use change	$\alpha$ -diversity, $\beta$ -diversity	(Faria et al., 2024)
4	Canada	Land use change	$\alpha$ -diversity, $\beta$ -diversity	(Gleason et al., 2021)
5	Canada	Pollution gradient (herbicide)	$\alpha$ -diversity, $\beta$ -diversity	(Loria et al., 2025)
6	China	Anthropogenic (pollution gradient)	$\alpha$ -diversity, $\beta$ -diversity, Functional diversity	(Xu et al., 2023)
7	China	Dams, Nutrient enrichment	$\alpha$ -diversity, $\beta$ -diversity	(Li et al., 2022)
8	China	Drought	$\alpha$ -diversity, $\beta$ -diversity	(Feng et al., 2026)
9	China	Elevation	$\alpha$ -diversity, $\beta$ -diversity	(Shen et al., 2024)
10	China	Flow	$\alpha$ -diversity, $\beta$ -diversity, Functional diversity	(Li et al., 2024)
11	China	Heavy metal pollution	$\alpha$ -diversity, $\beta$ -diversity, Phylogenetic diversity	(Zhou et al., 2022)
12	China	Heavy metal pollution (Cu)	$\alpha$ -diversity, $\beta$ -diversity, Functional diversity, Phylogenetic diversity	(Yang et al., 2018)
13	China	Land use change	$\alpha$ -diversity, $\beta$ -diversity, Functional diversity	(Li et al., 2020)
14	China	Land use change	$\alpha$ -diversity, $\beta$ -diversity, Functional diversity, Phylogenetic diversity	(Wu et al., 2023)
15	China	Land use change	$\alpha$ -diversity, $\beta$ -diversity	(Li et al., 2023a)
16	China	Land use change, Pollutant gradient	$\alpha$ -diversity, $\beta$ -diversity	(Li et al., 2023b)
17	China	Land use change, Pollution gradient	$\alpha$ -diversity, $\beta$ -diversity	(Qin et al., 2023)

No.	Country	Environmental Variables	Response Variables	References
18	China	Salinity	$\alpha$ -diversity, Functional diversity, Phylogenetic diversity	(Zhang et al., 2024)
19	China	Seasonal hydrological variation (Water temp., Flow)	$\alpha$ -diversity, $\beta$ -diversity, Functional diversity	(Liang et al., 2022)
20	China	Trophic status	$\alpha$ -diversity, $\beta$ -diversity, Functional diversity	(Xiong et al., 2025)
21	China	Urbanization gradient	$\alpha$ -diversity, $\beta$ -diversity, Functional diversity	(Ji et al., 2022)
22	China	Water temperature, Elevation	$\alpha$ -diversity, $\beta$ -diversity, Functional diversity, Phylogenetic diversity	(Lu et al., 2023)
23	China	Water temperature, Elevation	$\alpha$ -diversity, $\beta$ -diversity, Phylogenetic diversity	(Lu et al., 2024)
24	China	Water temperature, Salinity	$\alpha$ -diversity, $\beta$ -diversity, Functional diversity	(Huo et al., 2025)
25	Denmark	Seasonal hydrological variation (Water temp., Flow)	$\alpha$ -diversity, $\beta$ -diversity, Functional diversity	(Reinholdt Jensen et al., 2021)
26	Ecuador	Elevation	$\alpha$ -diversity	(Jijón et al., 2025)
27	Germany	Seasonal hydrological variation (Water temp., Flow)	$\alpha$ -diversity, $\beta$ -diversity	(Hupało et al., 2022)
28	Greece	Hydrological barriers	$\alpha$ -diversity, $\beta$ -diversity, Functional diversity, Phylogenetic diversity	(Ntislidou et al., 2023)
29	Indonesia	Flow	$\alpha$ -diversity, $\beta$ -diversity	(Effendi et al., 2023)
30	Italy	Salinity, Anthropogenic pressure	$\alpha$ -diversity, $\beta$ -diversity, Functional diversity, Keystone species screening	(Specchia et al., 2022)
31	Italy	Salinity, Substrate	$\alpha$ -diversity, $\beta$ -diversity, Functional diversity	(Pinna et al., 2024)
32	Japan	Water temperature, Flow	$\alpha$ -diversity, $\beta$ -diversity	(Uchida et al., 2020)
33	Netherlands	Agricultural source (Pesticides)	$\alpha$ -diversity, $\beta$ -diversity	(Beentjes et al., 2022)
34	Netherlands	Pollution gradient (microplastic)	$\alpha$ -diversity, $\beta$ -diversity	(van der Plas et al., 2025b)
35	Spain	Anthropogenic (pollution gradient)	$\alpha$ -diversity, $\beta$ -diversity	(Múrria et al., 2024)
36	Switzerland	Anthropogenic pressure (Non-native species detection)	$\alpha$ -diversity, keystone species screening	(Blackman et al., 2022a)
37	Switzerland	Seasonal hydrological variation (Water temp., Flow)	$\alpha$ -diversity, $\beta$ -diversity	(Blackman et al., 2022b)
38	Switzerland	Urbanization gradient	$\alpha$ -diversity, $\beta$ -diversity	(Perrelet et al., 2025)
39	UK	Land use change	$\alpha$ -diversity, $\beta$ -diversity, Functional diversity	(Seymour et al., 2021)
40	USA	Habitat quality	$\alpha$ -diversity, $\beta$ -diversity	(Marshall and Stepien, 2020)
41	USA	Land use change	$\alpha$ -diversity, $\beta$ -diversity, Functional diversity	(Penaluna et al., 2026)
42	USA	Seasonal hydrological variation (Water temp., Flow)	$\alpha$ -diversity, $\beta$ -diversity	(Aunins et al., 2023)

118 International research on benthic macroinvertebrates using eDNA technology exhibits distinct regional  
119 emphases. For instance, studies in the United States and China frequently examine how environmental variables



120 such as water temperature and nutrient concentrations affect species diversity and community structure, often using  
121 classical response indices like  $\alpha$ - and  $\beta$ -diversity. In contrast, research in Australia and Brazil tends to focus on  
122 functional diversity, species composition, and related ecological processes, commonly applying functional diversity  
123 indices to track ecosystem changes. In Germany and Spain, greater attention has been given to pollutant gradients  
124 and their interactions with aquatic biota, with species richness and community structure regularly used as response  
125 metrics. Notably, as eDNA methods become more accessible and their resolution improves, an increasing number  
126 of studies are extending beyond purely taxonomic metrics to quantify ecological function. A significant advance in  
127 this direction is the “Fun-eDNA” framework recently proposed by Cantera et al. (2024). This method systematically  
128 extracts functional trait information, such as feeding group, body size, and dispersal capacity, from eDNA  
129 metabarcoding data, allowing the calculation of functional diversity indices (e.g., functional richness, evenness, and  
130 redundancy). By transcending the previous limitation of eDNA to taxonomic identification, this framework supports  
131 functional inference at a molecular level, offering considerable potential for both theoretical and applied ecology.

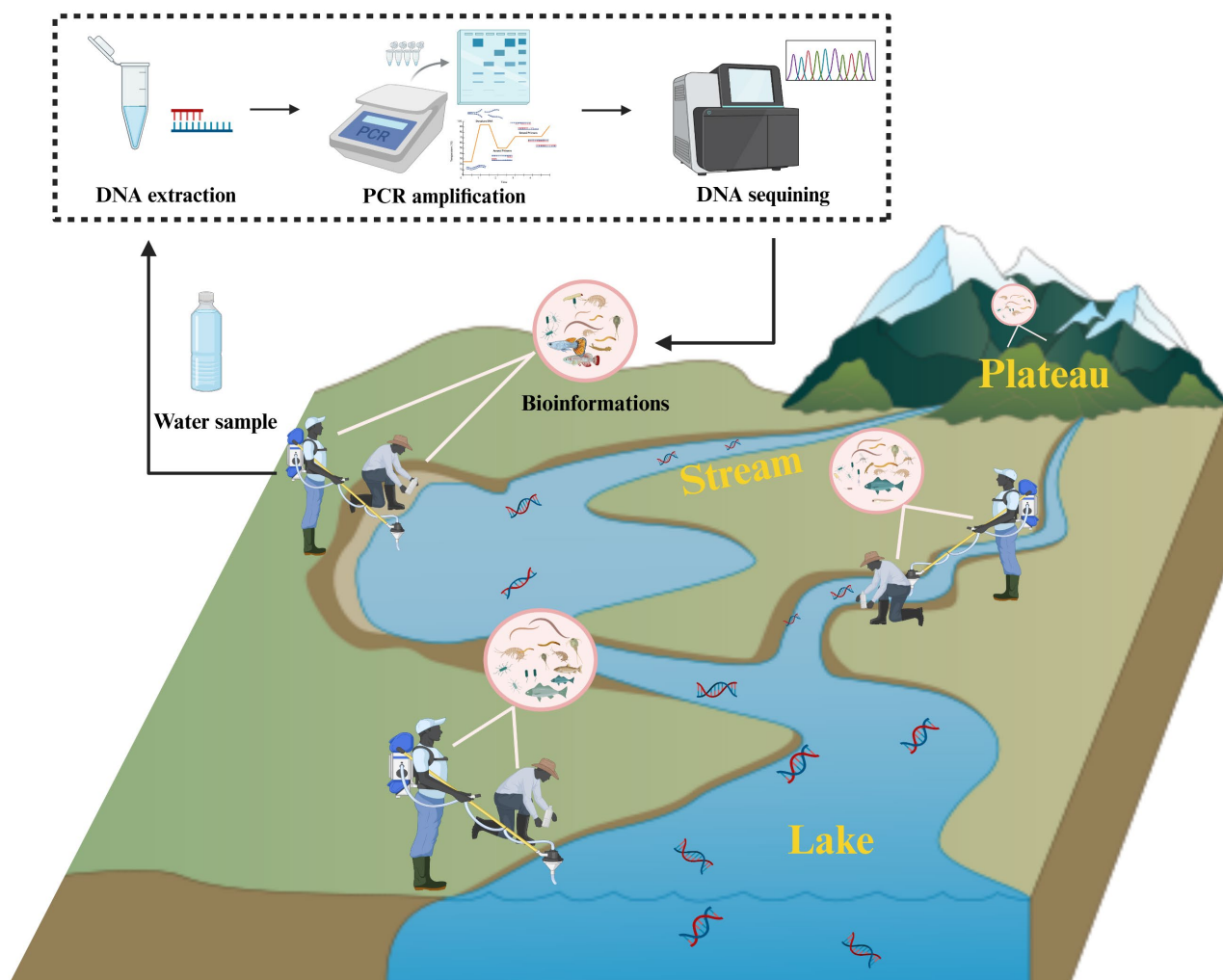
132 In summary, eDNA technology offers a holistic perspective for assessing aquatic biodiversity. The indices  
133 derived from eDNA data provide essential insights into ecosystem health, stability, and resilience, informing  
134 targeted conservation and restoration measures (Rowland et al., 2020). Ecosystem health can also be evaluated using  
135 biological indices, which are often based on the sensitivity or tolerance of certain taxa to environmental conditions  
136 and rely heavily on macroinvertebrate community composition (Sumudumali and Jayawardana, 2021). Moving  
137 forward, integrating eDNA with complementary monitoring tools and ecological models will be vital to effectively  
138 manage and mitigate the impacts of environmental degradation and climate change.

#### 139 **4. Unraveling Spatiotemporal Dynamics: eDNA Reveals Benthic Community Patterns**

140 The spatiotemporal dynamics of benthic macroinvertebrate communities in freshwater ecosystems are  
141 increasingly elucidated through eDNA approaches, capturing their nuanced responses to environmental gradients  
142 and seasonal fluctuations.

143 In the study of temporal dynamics, eDNA technology can elucidate how benthic macroinvertebrate  
144 communities respond across multiple timescales. Research on seasonal variation has revealed, for instance, that in  
145 temperate rivers, communities undergo functional group restructuring driven by winter low temperatures and summer  
146 nutrient inputs (Reinholdt Jensen et al., 2021). Similarly, in Mediterranean regions, high flows during the rainy season  
147 alter substrate stability and spatial heterogeneity, promoting a shift toward disturbances-adapted functional groups  
148 (Hupało et al., 2022). Over longer timescales, eDNA-based monitoring can track interannual trends in community  
149 dynamics. For example, Bista et al. (2017) performed annual eDNA analyses in a lake ecosystem, revealing dynamic  
150 biodiversity changes and significant interannual variations in the eDNA abundance of certain benthic  
151 macroinvertebrates, which were likely influenced by climate change and anthropogenic activities. Overall, eDNA  
152 metabarcoding effectively captures temporal restructuring of benthic communities in response to environmental  
153 drivers such as water temperature, nutrient availability, flow regimes, and substrate dynamics. The technique thus  
154 provides a sensitive means to trace how these factors shape community composition over time.

155 Spatial heterogeneity, alongside temporal dynamics, plays a pivotal role in structuring benthic  
156 macroinvertebrate communities (**Fig. 1**). Studies employing eDNA have demonstrated that community composition  
157 responds strongly to environmental gradients across multiple spatial scales, ranging from broad catchments to  
158 microhabitats. At larger (e.g., catchment) scales, complex river networks with high tributary density promote  
159 ecological niche diversification and enhance hydrological connectivity, thereby increasing both taxonomic and  
160 phylogenetic richness (Chang et al., 2025a). At intermediate (e.g., reach) scales, gradients in temperature and salinity  
161 shape distinct distribution patterns, with gradual turnover observed from freshwater inflows to coastal zones (Pinna  
162 et al., 2024). In high-altitude systems such as glacier-fed river-lake continua, eDNA has detected continuous  
163 transitions in biological assemblages for example from cold-adapted upstream specialists to widely distributed  
164 downstream taxa, which are primarily driven by thermal and turbidity gradients (Lu et al., 2023). Notably, such  
165 fine-scale community transitions can be detected across distances as short as a few hundred meters (Lu et al., 2024).



**Fig. 1. Representative illustration of spatially explicit eDNA applications for assessing benthic macroinvertebrate communities across freshwater ecosystems.**

The observed spatial patterns are driven by a combination of natural environmental gradients and anthropogenic pressures. Land-use changes, for example, indirectly alters community structure through increased nutrient loading and pollutant influx (Li et al., 2020, 2023b; Qin et al., 2023). Agricultural and urbanized areas are frequently dominated by pollution-tolerant taxa, while sensitive groups are diminished, leading to predictable spatial shifts in community composition (Li et al., 2020; Seymour et al., 2021). Overall, eDNA offers a powerful approach for mapping spatial biodiversity patterns and disentangling the complex interplay between natural heterogeneity and human impacts. Its fine spatial resolution also shows strong potential for improving predictive models of benthic community dynamics across multiple scales.

177 eDNA technology serves as a crucial tool for elucidating spatiotemporal interactions within benthic  
178 macroinvertebrate communities. For example, multi-season monitoring in the Weihe River basin demonstrated that  
179 hydrological isolation during dry periods and runoff dispersal during wet seasons drive both local adaptation and  
180 regional restructuring of these communities (Liang et al., 2022). This indicates that hydrological rhythms regulate  
181 community dynamics through coordinated spatiotemporal processes. Collectively, such findings show that eDNA,  
182 by integrating data across space and time, provides a reliable foundation for uncovering the assembly mechanisms  
183 of benthic macroinvertebrate communities.

184 Despite its considerable promise for revealing spatiotemporal dynamics, several limitations of eDNA  
185 approaches must be addressed in future applications. First, long-term time-series monitoring is often hampered by  
186 funding constraints and a lack of methodological standardization, making it difficult to systematically track  
187 community changes associated with species phenology. Second, false-negative detections require careful ecological  
188 interpretation; a taxon may go undetected due to low abundance (e.g., rare species) or low metabolic activity,  
189 resulting in an ecologically meaningful "relative absence". Furthermore, accurately describing spatiotemporal  
190 interactions in these communities necessitates the integration of multidimensional datasets (Reinholdt Jensen et al.,  
191 2021). Most current studies rely primarily on eDNA data alone and would benefit from deeper integration with  
192 hydrological models or machine learning algorithms.

## 193 **5. Unraveling Cascading Effects: Tracing Anthropogenic Stress in freshwater Ecosystems** 194 **through eDNA**

### 195 **5.1 Pollution cascades and multi-trophic effects**

196 Freshwater ecosystems face growing pressure from point-source pollutants, including mining runoff,  
197 agricultural inputs, and urban drainage. Such stressors can reduce biodiversity through direct toxicity or broader  
198 habitat degradation (Reid et al., 2019). eDNA approaches are proving effective in diagnosing the cascading effects

199 of pollutants (such as heavy metals, mining effluents, agrochemicals, and insecticides) on benthic communities  
200 across multiple trophic levels.

201         Pollutants drive systemic reorganization of benthic communities through direct toxicity or indirect alteration  
202 habitat, demonstrating characteristic ecological filtering and cascading response patterns. For instance, microcosm  
203 experiments and field monitoring show that when specific pollutants (e.g., copper and certain ions) exceed ecological  
204 thresholds, they can trigger synchronous collapse across multiple trophic levels, from bacteria and algae to benthic  
205 macroinvertebrate communities (Yang et al., 2018; Simonin et al., 2021). In contrast, the effects of agricultural non-  
206 point source pollution (e.g., nitrogen, phosphorus nutrients and pesticides) are more complex and cumulative.  
207 Evaluations of treated wastewater effluent further reveal that point-source pollution can reorganize multi-trophic  
208 communities mainly by altering trophic interactions and control pathways, rather than simply reducing overall diversity  
209 (Xiong et al., 2025). eDNA-based analyses indicate that combined stress from agricultural nutrients (e.g.,  $\text{NH}_4^+\text{-N}$  and  
210 TN) and neonicotinoid insecticides suppresses sensitive groups such as EPT taxa, thereby increasing the dominance  
211 of pollution-tolerant groups (e.g., oligochaetes). This shift alters community functional structure and simplifies food-  
212 web architecture (Beentjes et al., 2022; Xu et al., 2023). Additionally, pharmaceutical active compounds (PhACs), as  
213 emerging contaminants, can significantly disrupt the behavior and reproduction of microbial communities, benthic  
214 macroinvertebrates, and fish through chronic low-level exposure, initiating cascading effects across trophic levels  
215 (Papaioannou et al., 2023). This process may be accompanied by the spread of resistant bacteria and antibiotic  
216 resistance genes, further exacerbating the functional degradation of aquatic communities.

217         Recent studies show that eDNA metabarcoding effectively captures shifts in community composition and losses  
218 of genetic diversity under both acute stressors such as glyphosate exposure (Loria et al., 2025) and diffuse, low-intensity  
219 stressors like micro- and nanoplastics, whose impacts are often subtle and transient at the community level (van Der Plas  
220 et al., 2025b). Notably, glyphosate-induced effects are strongly amplified under eutrophic conditions, driving rapid decline  
221 of sensitive taxa (e.g., crustaceans and rotifers) and resulting in marked structural reorganization of the community.

222 Together, these findings highlight the need to incorporate both nutrient status and intra-community genetic responses into  
223 ecological risk assessments to adequately resolve pollution-driven cascades across trophic levels.

224 Despite considerable research using eDNA metabarcoding to assess pollutants impacts on benthic  
225 macroinvertebrate, several limitations persist. For example, Zhou et al. (2022), combining morphological and  
226 eDNA approaches, found that shale-gas wastewater discharge did not significantly alter benthic macroinvertebrate  
227 community structure within two years, suggesting that longer-term monitoring may be necessary to detect ecological  
228 effects of slow-release pollutants. Furthermore, emerging pollutants such as microplastics and antibiotics remain  
229 understudied. More broadly, the current lack of methodological standardization in eDNA research calls for the  
230 development of harmonized frameworks to enable large-scale, comparable ecological risk assessments.

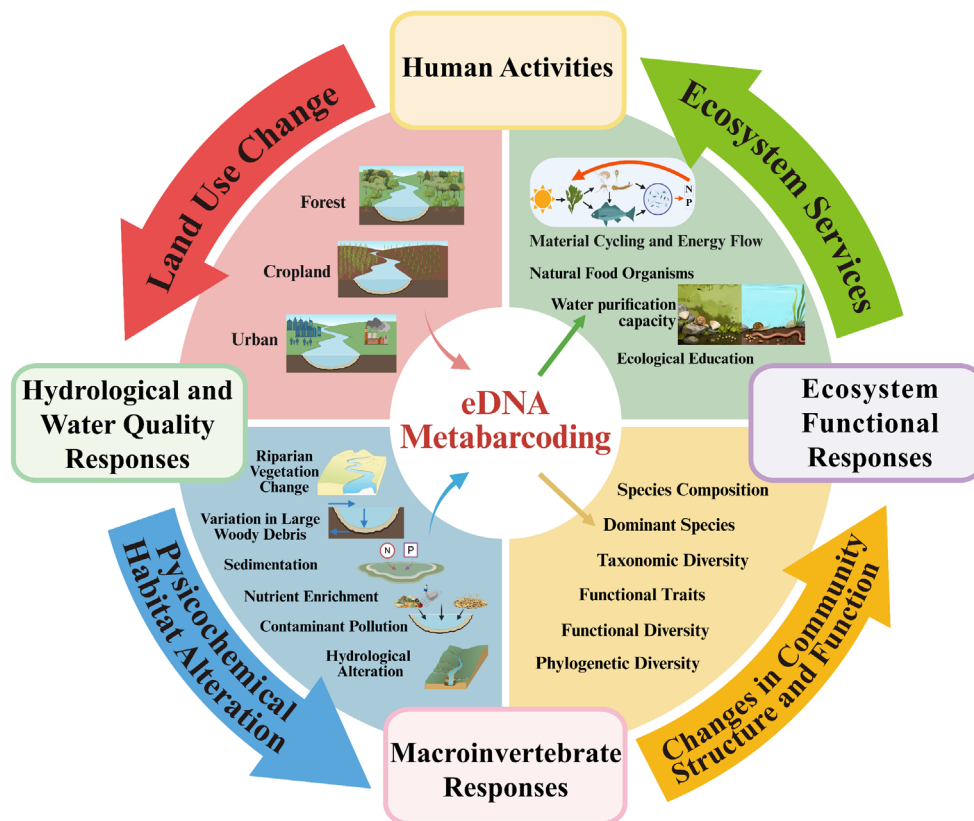
## 231 **5.2 Hydrological alteration and community reorganization**

232 Nearly half of global river volume is affected by flow regulation and/or fragmentation (Grill et al., 2015).  
233 Hydraulic infrastructure such as dams alters hydrological regimes and connectivity, driving adaptive restructuring  
234 in benthic macroinvertebrate communities (Chan et al., 2025). Reservoirs and cascading dams dampen downstream  
235 hydrological pulse; for example, eDNA monitoring in the Wujiang River has shown a shift from fast-flow-adapted  
236 taxa (e.g., EPT groups) to standing water-adapted Chironomidae (Shen et al., 2024). In drought-prone regions,  
237 prolonged high water levels in reservoirs can exacerbate intermittent streamflow, increasing the risk of habitat  
238 desiccation and prompting a transition from seasonal dynamic to stagnant-water, steady-state benthic communities  
239 (Li et al., 2020). Furthermore, dams can impede the dispersal of benthic larvae (Ntislidou et al., 2023). Moving  
240 beyond single-stressor studies, Li et al. (2022) used eDNA metabarcoding to reveal the synergistic effects of dams  
241 and nutrient enrichment on multitrophic community cascades. eDNA-based evidence also indicates that  
242 hydrological connectivity buffers multitrophic biodiversity loss during hydrological extremes (e.g., drought),  
243 constraining community reorganization beyond the influence of flow regulation alone (Feng et al., 2026).

244 In summary, future research should integrate hydrological models with eDNA dynamic monitoring and  
245 quantify threshold effects of water infrastructure on the resilience of benthic macroinvertebrate communities.  
246 Moreover, applying eDNA to decipher complex stressor interactions (such as synergistic, antagonistic, or  
247 cumulative effects) will be essential for informing science-based management decisions.

248 **5.3 Effects of land-use change and habitat homogenization**

249 Land-use change is a major driver of biodiversity loss in freshwater ecosystems (Foley et al., 2005). Human  
250 activities such as urbanization, agricultural expansion, and deforestation cascade to alter the structure and function  
251 of benthic macroinvertebrate communities, largely by increasing nutrient loading (Faria et al., 2024) and reducing  
252 habitat heterogeneity (Birk et al., 2020). eDNA offers an effective means of tracking these changes, providing  
253 insights into how biodiversity and ecosystem function shift across land-use types changes that ultimately affect  
254 human well-being (Fig. 2).



255  
256 **Fig. 2. Responses of benthic macroinvertebrate eDNA to land-use change across river basins.**

257           Accumulating eDNA evidence reveals a consistent pattern: along gradients from forested to agricultural  
258 and urban landscapes, benthic macroinvertebrate communities are reshaped by nutrient enrichment, contaminant  
259 inputs, and habitat fragmentation, which in turn alter multidimensional biodiversity and interaction network  
260 architecture. Forest cover is generally associated with greater community stability, whereas intensive agriculture  
261 tends to reduce sensitive EPT taxa and increase the relative dominance of Oligochaeta and Chironomidae (Li et al.,  
262 2018; Marshall and Stepien, 2020; Li et al., 2023a; Faria et al., 2024; Zhu et al., 2024). Urbanization and habitat  
263 homogenization further reduce  $\beta$ -diversity and functional evenness, leading to compositional convergence and  
264 lower functional redundancy (Li et al., 2020; Seymour et al., 2021). These trends are observed across climatic zones  
265 and spatial scales—from headwater streams to mainstems and large catchments—and extend to the simplification  
266 of multitrophic network, reflected in weakened link density and connectance (Li et al., 2023b; Qin et al., 2023).  
267 Beyond broad land-use categories, recent watershed-scale studies also indicate that forest age structure and internal  
268 heterogeneity can mediate freshwater community responses independently of overall land-use type (Penaluna et al.,  
269 2026).

270           Importantly, these findings derived from eDNA align with robust morphology-based evidence confirming  
271 land use change as a widespread driver of benthic community homogenization and functional erosion (Pratiwi et  
272 al., 2024; Vidal-Abarca Gutiérrez, 2024; Xie et al., 2024), while offering greater sensitivity and taxonomic breadth.  
273 By resolving finer-scale community across land-use types and detecting cascading effects across trophic levels (Ji  
274 et al., 2022), eDNA provides a quantitative means to trace the causal pathway from land-use change, through altered  
275 habitat conditions and community structure, to changes in ecosystem functioning and services delivery.

276           Emerging evidence further suggests that multitrophic network structure exhibits a stronger response to land  
277 use than do conventional diversity metrics. Therefore, research on riverine ecosystem responses to anthropogenic  
278 pressures should prioritize the analysis of these interaction networks, which can offer a more accurate reflection of  
279 a river's ecological condition.



## 280    **6. Navigating Multitrophic Networks with Environmental DNA**

281           Understanding the structure and dynamics of multitrophic interaction networks, particularly food webs, is  
282   essential for assessing ecosystem stability, function, and resilience to disturbances (Delmas et al., 2019; Thompson  
283   et al., 2012). Benthic macroinvertebrates occupy pivotal positions within these networks, linking basal resources  
284   (e.g., detritus, algae, microbes) to higher trophic levels (e.g., fish).

285           In the Shaying River and other subtropical river systems, eDNA-based multitrophic monitoring has shown  
286   that land-use change reduces the functional evenness of benthic macroinvertebrates and alters their associations  
287   with microbial and algal groups (Li et al., 2020, 2023b; Qin et al., 2023). These shifts decrease organic-matter  
288   decomposition efficiency and simplify food web architecture. Similarly, Wu et al. (2025) used eDNA in eutrophic  
289   lakes to clarify the functional roles of benthic communities and explore regulatory pathways such as bottom-up or  
290   top-down control. In extreme habitats like hypersaline lakes, pronounced salinity gradients support unique  
291   assemblages of salt-tolerant taxa and markedly reduced network modularity (Zhang et al., 2024). Notably, the  
292   spatiotemporal dynamics of food-web structure do not always align with patterns of species diversity (Blackman et  
293   al., 2022b), suggesting that understanding ecological networks assembly requires moving beyond taxonomic  
294   richness to examine the dynamic configuration of functional groups and their interactions.

295           By providing high-resolution, multitrophic data, eDNA metabarcoding deepens our understanding of  
296   freshwater food webs, revealing changes in connectivity and stability often missed by conventional methods. As a  
297   holistic assessment tool, eDNA technology offers the empirical foundation needed to diagnose ecosystem health  
298   under global change and to guild effective conservation strategies.

## 299    **7. Enhanced Detection of Species via eDNA: From Rarity to Invasion**

300           Accurate species detection is fundamental to effective ecological monitoring and biodiversity conservation.  
301   In freshwater ecosystems, both rare native species and invasive alien species (IAS) pose critical yet distinct

302 management challenges, with the former often requiring protection and the latter demanding early control (Yates et  
303 al., 2025). Conventional survey methods frequently fail to reliably detect these taxa due to their low abundance,  
304 cryptic behavior, or early stages of colonization. eDNA technology offers distinct advantages for detecting rare or  
305 endangered species (Sigsgaard et al., 2015; Coutts et al., 2022) as well as for tracking biological invasions (Ficetola  
306 et al., 2008; Hartle-Mougiou et al., 2023).

## 307 **7.1 Monitoring Rare and Cryptic Species with eDNA**

308 Monitoring rare and elusive taxa remains a persistent challenge in ecological assessments due to their low  
309 abundance, cryptic behaviors, and often benthic or burrowing life histories —traits that frequently lead to under-  
310 detection by traditional methods such as visual surveys, benthic trawling, and morphological identification. Owing  
311 to its high sensitivity, eDNA has emerged as a transformative tool capable of detecting trace genetic material shed  
312 by organisms into the surroundings (Keck et al., 2022; Çevik and Çevik, 2025; Chang et al., 2025b).

313 For taxa that are difficult to detect, whether due to low abundance, cryptic habits, or morphological  
314 ambiguity, eDNA significantly increases detection probability and reduces false-negative rates (Johnsen et al., 2020;  
315 Giribet et al., 2023; McCulloch et al., 2025; Zieritz et al., 2025). Its non-invasive nature also supports repeated and  
316 ethical sound monitoring of threatened species, aiding in tracking of population declines (Hata et al., 2022) and  
317 refining estimates of contemporary distributions (Lor et al., 2020; Preece et al., 2021). Together, these applications  
318 deliver timely, actionable data to guide conservation efforts.

319 However, several limitations affect the detection of rare taxa. Detectability can be influenced by low rates  
320 of DNA shedding, constrained environmental dispersion, or PCR primer bias (Coghlan et al., 2021; Schmidt et al.,  
321 2021). These issues are particularly evident in complex substrates or lotic (flowing) environments, where DNA  
322 undergoes rapid dilution and degradation. Such constraints represent the current sensitivity boundaries of eDNA  
323 methods under specific conditions, rather than an inherent flaw of the approach.

## 324 7.2 Early Detection and Surveillance of Invasive Species with eDNA

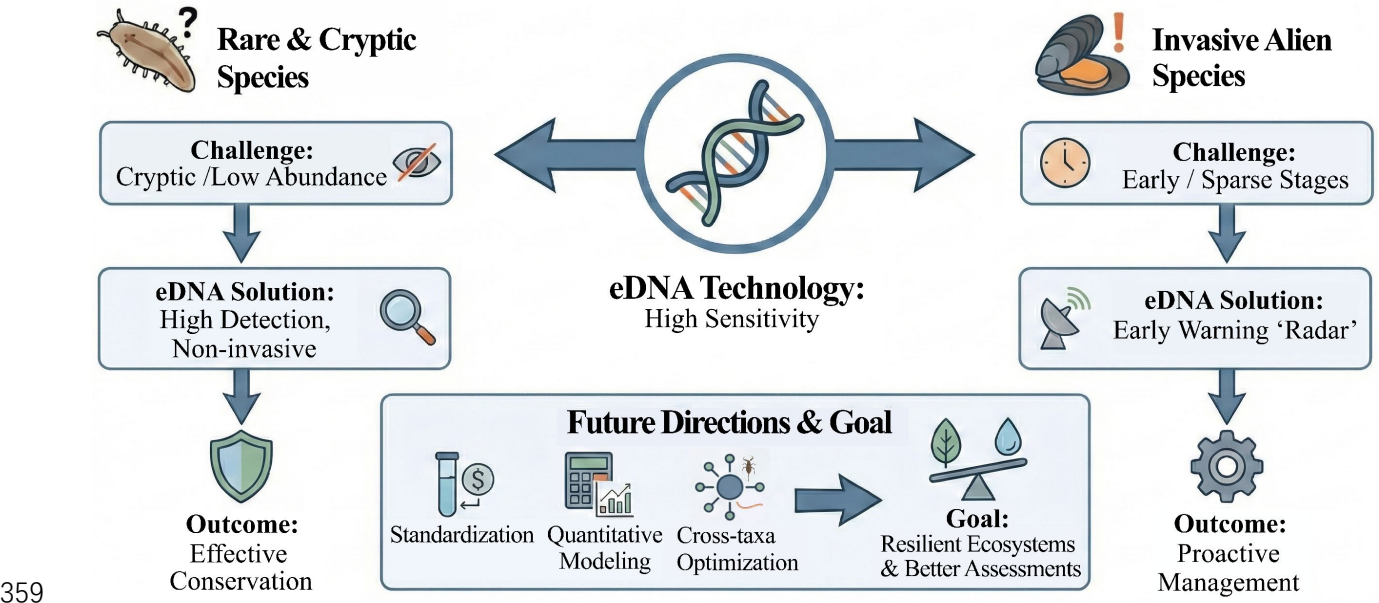
325 Invasive alien species represent a major threat to freshwater biodiversity, disrupting native communities  
326 through competitive exclusion, habitat modification, and resource monopolization (Rahel and Olden, 2008; Pyšek  
327 et al., 2020). Early detection and ongoing surveillance are therefore critical for effective management. eDNA has  
328 become a widely adopted molecular tool for this purpose, offering particular utility during the initial stages of  
329 invasion when traditional surveys often fail.

330 Often described as a "molecular radar," eDNA can detect target DNA during the earliest stages of invasive,  
331 when populations are sparse and not yet visible, or in frontline colonization zones such as ports and water intake  
332 points, offering a substantially earlier warning than traditional visual surveys (Thomsen et al., 2011; Goldberg et  
333 al., 2013; Dougherty et al., 2016; Jeunen et al., 2022). It can also rapidly characterize macroinvertebrate  
334 communities in poorly studied or inaccessible areas, providing initial evidence of invasive presence and securing a  
335 critical time window for rapid response (Mauro et al., 2025). Moreover, eDNA enables simultaneous screening of  
336 multiple potential invasion sites or extensive water bodies, covering locations that are logistically challenging for  
337 conventional methods (Clusa et al., 2021). Notably, quantitative eDNA approaches have been successfully applied  
338 to monitor invasive mussels in large-scale water diversion systems (Yang et al., 2025), further demonstrating the  
339 practical utility of this technology in invasion surveillance.

340 Nevertheless, detecting invasive species at their earliest stages remains inherently challenging. There is  
341 often a lag between initial colonization and population establishment, during which only a few individuals may be  
342 present and shed DNA below current detection thresholds. A further temporal delay can occur between initial DNA  
343 deposition and the accumulation of a quantifiable signal, reducing detectability during this critical window. Since  
344 eradication efforts are most effective when undertaken early in the invasion process (Fonseca et al., 2023; Harper  
345 et al., 2018), enhancing the sensitivity of eDNA detection methods is essential.

346 A further obstacle to routine implementation is the lack of standardized protocols across jurisdictions for  
 347 eDNA sampling, laboratory processing, primer selection, and data interpretation. This inconsistency hinders data  
 348 comparability and limits the regulatory adoption of eDNA evidence in policy-making (Rishan et al., 2023).  
 349 Addressing these standardization gaps is critical for integrating eDNA into operational invasive species surveillance  
 350 programs.

351 In summary, eDNA-based biomonitoring holds transformative potential for detecting both rare native  
 352 species and invasive taxa, thereby improving biodiversity assessments and enabling proactive ecosystem  
 353 management. Key challenges include detection limits in low-density populations, primer biases, and the absence of  
 354 methodological standardization. To fully integrate eDNA into conservation and regulatory practice, future work  
 355 should prioritize cross-taxa primer optimization, quantitative modeling of detection thresholds, and the international  
 356 harmonization of sampling and analytical protocols. By bridging the gap between molecular signals and ecological  
 357 insight, eDNA technology can play a pivotal role in safeguarding freshwater biodiversity amid accelerating global  
 358 change (Fig. 3).



360 Fig. 3. eDNA as a dual-purpose surveillance tool for benthic macroinvertebrates ecology.

## 361 **8. From Descriptive to Mechanistic: How eDNA is Unraveling Benthic macroinvertebrate Communities**

362 This review synthesizes current applications of eDNA technology in studying benthic macroinvertebrate  
363 communities across global freshwater ecosystems. The continuous and scalable biodiversity monitoring enabled by  
364 eDNA metabarcoding in diverse habitats (stream, rivers, lakes, etc.) highlights its transformative role. This approach  
365 advances the field by:

366 (1) Enabling accurate, high-throughput detection of multi-taxa communities, with particularly effective for  
367 low-abundance, cryptic, or taxonomically challenging organisms, thus supporting robust assessment of "cryptic  
368 biodiversity" often missed by traditional methods;

369 (2) Capturing spatiotemporal ecosystem dynamics rapidly, including community succession and turnover  
370 driven by seasonal/hydrological cycles or environmental gradients;

371 (3) Detecting early signals of anthropogenic disturbance and evaluating associated ecological impacts, such  
372 as biodiversity loss from industrial/agricultural pollution, habitat homogenization due to dams, and functional  
373 simplification of communities following land-use change;

374 (4) Resolving the structure and dynamics of multitrophic interaction networks, revealing how pollutants or  
375 habitat alterations affect food-web stability through trophic cascades, or uncovering unique community assembly  
376 patterns in extreme environments.

377 In recent years, eDNA research has progressively shifted from species identification and community  
378 description toward a more mechanistic understanding of ecological processes. Breakthroughs have been emerged  
379 on several key fronts. For instance, recent studies have parameterized particle-size distributions to simulate the  
380 eDNA transport in aquatic systems, thereby improving estimates of its physical detectability (Jo, 2025). Others have  
381 shown that data-preprocessing protocols and statistical modeling strategies substantially affect the ecological

382 interpretability of quantitative eDNA results (Bylemans et al., 2025). These advances reflect a broader paradigm  
383 shift: eDNA is evolving from a detection tool into a probe for ecological mechanisms, opening new avenues to  
384 unravel the assembly rules and functional dynamics of benthic macroinvertebrate communities.

385         Nevertheless, conceptual validation is required to bridge gaps between eDNA signals (e.g., detection rates,  
386 relative read abundance) and the true presence, absence or biomass of source organisms in benthic habitats,  
387 especially for rare species, juvenile stages, or metabolically inactive individuals. It is also important to recognize  
388 that eDNA constitutes indirect, extracellular genetic material shed by macroinvertebrates. Consequently, any  
389 eDNA-based estimates inherently involve uncertainties, including false negatives and false positives. These errors  
390 can arise at any stage of the eDNA metabarcoding workflow, from sampling and molecular processing to  
391 bioinformatic analysis, and should be systematically calibrated through co-located eDNA and morphological  
392 sampling, as well as refined through future technical improvements (Çevik and Çevik, 2025).

393         Significant technical challenges remain. Integrating eDNA into routine biomonitoring requires further  
394 validation of standardized protocols and adaptive modification of existing bioassessment indices (e.g., indices of  
395 biological integrity). Detection accuracy can be enhanced through strategies such as seasonal sampling, automated  
396 monitoring, and optimized primer design/reference databases design. The resulting large-scale quantitative datasets  
397 are essential for capturing fluctuating interspecific interactions and seasonal dynamics in benthic communities.  
398 Integrating these data with local hydrological models or machine-learning algorithms offers a promising path to  
399 model complex interactions, predict community dynamics from time-series data, and ultimately establish scientific  
400 baselines for ecosystem conservation.

401         Although further technical and methodological refinements are needed, recent national-level initiatives,  
402 such as those in China, demonstrate how eDNA monitoring can be institutionalized through standardized protocols  
403 and integrated databases (Wang et al., 2022; Chinese Society for Environmental Sciences (CSES), 2023a, 2023b,

2023c; Jiangsu Provincial Administration for Market Regulation (JPAMR), 2023). These efforts highlight the potential for eDNA to become embedded within formal ecological assessment and policy systems.

In conclusion, eDNA is fundamentally reshaping how we understand and conserve freshwater ecosystems. By addressing current methodological limitations and deepening research into ecological mechanisms, eDNA holds strong promise as a core tool for routine monitoring of benthic macroinvertebrate communities. Especially under intensifying global change, its efficiency and sensitivity provide crucial technical support for diagnosing river health and guiding ecological restoration, thereby advancing science-informed decision-making “from DNA to ecosystem management”.

## Author Contributions

**Yajing Zhang:** Conceptualization, Methodology, Investigation, Formal analysis, Visualization, Writing – original draft, Writing – review and editing; **Hui Yang:** Investigation, Data curation; **Xiwen Liang:** Investigation, Data curation; **Min Zhang:** Investigation, Data curation; **Wenwen Zhu:** Investigation, Data curation; **Hongxu Yang:** Investigation, Data curation; **Wenze Fan:** Investigation, Data curation; **Kun Li:** Conceptualization, Methodology, Writing – original draft, Writing – review and editing, Supervision, Resources, Funding acquisition.

## 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.

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## Data Availability

Data will be made available on request.

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