

Aquatic insect community structure revealed by eDNA metabarcoding derives indices for environmental assessment

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Environmental DNA (eDNA) analysis provides an efficient and objective approach for monitoring and assessing ecological status; however, studies on the eDNA of aquatic insects, such as Ephemeroptera, Plecoptera, and Trichoptera (EPT), are limited despite its potential as a useful indicator of river health. Here, we investigated the community structures of aquatic insects using eDNA and evaluated the applicability of eDNA data for calculating assessment indices. Field surveys were conducted to sample river water for eDNA at six locations from upstream to downstream of two rivers in Japan in July and November 2016. Simultaneously, aquatic insects were collected using the traditional Surber net survey method. The communities of aquatic insects were revealed using eDNA by targeting the cytochrome oxidase subunit I gene in mitochondrial DNA via metabarcoding analyses. The eDNA results revealed 63 families and 75 genera of aquatic insects, which was double than that detected by the Surber net survey (especially for families in Diptera and Hemiptera). The seasonal differences of communities were discriminated by the eDNA and Surber net survey data. Furthermore, environmental assessment indices (i.e., EPT index and Chironomidae index) calculated using the richness of operational taxonomic units at the genus level showed positive correlations with total nitrogen concentration as a surrogate parameter of the degree of organic pollution but the index calculated using the Surber net survey data did not. Our results demonstrated that eDNA analysis with higher taxonomic resolution can serve as a more sensitive environmental assessment index than the traditional method that requires biotic samples.

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3 **assessment**

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18

19 **Abstract**

20 Environmental DNA (eDNA) analysis provides an efficient and objective approach for
21 monitoring and assessing ecological status; however, studies on the eDNA of aquatic insects,
22 such as Ephemeroptera, Plecoptera, and Trichoptera (EPT), are limited despite its potential as a
23 useful indicator of river health. Here, we investigated the community structures of aquatic insects
24 using eDNA and evaluated the applicability of eDNA data for calculating assessment indices.
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26 downstream of two rivers in Japan in July and November 2016. Simultaneously, aquatic insects
27 were collected using the traditional Surber net survey method. The communities of aquatic
28 insects were revealed using eDNA by targeting the cytochrome oxidase subunit I gene in
29 mitochondrial DNA via metabarcoding analyses. The eDNA results revealed 63 families and 75
30 genera of aquatic insects, which was double than that detected by the Surber net survey
31 (especially for families in Diptera and Hemiptera). The seasonal differences of communities
32 were discriminated by the eDNA and Surber net survey data. Furthermore, environmental
33 assessment indices (i.e., EPT index and Chironomidae index) calculated using the richness of
34 operational taxonomic units at the genus level showed positive correlations with total nitrogen
35 concentration as a surrogate parameter of the degree of organic pollution but the index calculated
36 using the Surber net survey data did not. Our results demonstrated that eDNA analysis with
37 higher taxonomic resolution can serve as a more sensitive environmental assessment index than
38 the traditional method that requires biotic samples.

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40

41 **Introduction**

42 Stream ecosystems are threatened by global climate changes and anthropogenic impacts,
43 including damming, water abstraction, and land-use changes (WWF, 2016). For sustainable
44 development and resource use of freshwater, there is a need to effectively manage stream
45 environments, which requires effective methods and indicators to measure and assess
46 environmental impacts. Aquatic insects are commonly used as indicators of environmental health
47 due to their high sensitivity to deterioration of water quality. They form a core component of the
48 ecological food web in river ecosystems by feeding on producers and being preyed upon by
49 higher consumers. Thus, monitoring of aquatic insect fauna is an effective method for assessing
50 the environmental and ecological status. However, traditional survey methods such as kick net
51 and Surber net sampling are subject to limitations. First, traditional field sampling processes
52 result in data bias because the success and quality of a survey depend on the ability and skills of
53 investigators and the accessibility of sampling sites. In addition, direct sampling methods
54 inherently involve damage to natural habitat and organism. Second, the subsequent process of
55 sorting and morphological identification is time-consuming and requires expertise in taxonomic
56 identification (Baird and Hajibabaei, 2012; Haase et al., 2006). These limitations have created
57 problems in performing high-frequency and long-term biological monitoring.

58 The use of environmental DNA (eDNA) is a novel biological monitoring method that can be
59 used to overcome these limitations. Due to the simple sampling method involved (i.e., grab
60 sampling of water, soil, etc.), eDNA monitoring can reduce the sampling bias caused by
61 individual investigators (Rees et al., 2014; Smart et al., 2015). Moreover, it can minimize the
62 bias associated with accessing the site because water and suspended materials, including eDNA,
63 are transported and mixed through different environments. Therefore, eDNA can detect not only
64 lotic animals but also lentic ones (Fernández et al., 2018; Macher et al., 2018, Deiner et al.
65 2016). This sampling method can also overcome ethical issues such as habitat disturbance and
66 animal sacrifice associated with field sampling because it requires only nonbiotic samples.
67 Furthermore, DNA-based identification can quickly provide results with higher taxonomic
68 resolution than morphological identification (Carew et al., 2013; Elbrecht and Leese, 2015;
69 Hajibabaei et al., 2011; Serrana et al., 2018). In addition, the necessary skills to analyze DNA
70 can be acquired over a shorter time than those required for morphological identification.

71 Although the characteristics of eDNA remain obscure (e.g., production and degradation rates and
72 transportation dynamics), it can provide ecological information that is unobtainable via
73 traditional survey methods. For example, eDNA can be used to evaluate the diversity of animals
74 across almost the entire fauna (fish fauna: Ushio et al., 2018) or across multiple phyla (phylum in
75 eukaryotes: Deiner et al., 2016).

76 Aquatic insects such as Ephemeroptera, Plecoptera, Trichoptera (the abbreviated name of these
77 groups in combination is EPT), and Diptera have been gradually targeted in eDNA studies in
78 pursuit of effective river management (Bista et al., 2017; Fernandes et al., 2018; Fernández et al.,
79 2018; Hajibabaei et al., 2019a; Macher et al., 2018; Mächler et al., 2019). However, there is a

80 lack of information regarding the spatial and temporal differences of community structures of
81 aquatic insects revealed by eDNA (Bush et al., 2019; Roussel et al., 2015). Furthermore, studies
82 examining the possible use of eDNA data for calculating environmental assessment indices and
83 focusing on aquatic insects in stream ecosystems in a Japanese context are warranted.
84 Therefore, the present study aimed to evaluate the applicability of eDNA in environmental
85 assessments. First, we revealed the community structures of aquatic insects using eDNA and the
86 traditional Surber net survey. Subsequently, we investigated whether eDNA data and the Surber
87 net survey data can discern the spatial and temporal differences of aquatic insect communities.
88 We also evaluated the relationships between water quality parameters and assessment indices
89 derived from each method.

90
91

92 **Materials & Methods**

93

94 **2.1 eDNA sampling, filtration, and DNA extraction**

95 Field samplings were conducted at Hirose River and Natori River located at the Natori River
96 basin, Miyagi Prefecture, northeast Japan. The length of the channel of Hirose River is 45.2 km
97 and its catchment area is 315.9 km². Natori River is 55.0 km long and its catchment area is 623.0
98 km² (not including the Hirose River Basin). Sampling was conducted in July and November
99 2016 at six locations from upland to lowland regions along the two rivers (sites H1–H3 and N1–
100 N3; Figure 1 and Table S1). These are temperate rivers that originate in the mountains and flow
101 through the hills at their middle reaches and through urbanized flatlands at their lower reaches
102 and finally flow into the Pacific Ocean.

103 Water samples for eDNA analysis were collected at the same sites and on the same days. The
104 plastic bottles for eDNA sampling were sterilized with 10% chlorine bleach (Kao Corporation,
105 Tokyo, Japan), rinsed with tap water in the laboratory, and washed thrice with river water at the
106 collection site before sampling. At each site, flowing surface water was collected and transported
107 to the laboratory on ice in a cool box. Water samples were filtered on the same day using
108 vacuum filtration with 47-mm diameter glass fiber filters with a 0.7- μ m pore size (GF/F,
109 Whatman, 1 L/filter, referring to Mächler et al., 2016). These filtered samples were stored at
110 -20°C until DNA extraction. DNA was extracted from the filters through lysis using proteinase
111 K at 56°C for 30 min. Then, the supernatant obtained was subjected to phenol–chloroform–
112 isoamyl alcohol extraction and ethanol precipitation. Eventually, the elution was purified using
113 the OneStep PCR Inhibitor Removal Kit (Zymo Research, Irvine, CA, USA) with a final volume
114 of 100 μ l.

115

116 **2.2 Library preparation and sequencing**

117 The cytochrome oxidase subunit I (COI) gene region in mitochondrial DNA was amplified from
118 extracted DNA using the universal primer for invertebrates developed by Folmer et al. (1994).
119 The primer set of LCO1490 (5'-GGT CAA ATC ATA AAG ATA TTG G-3') as the forward

120 primer and HCO2198 (5'-TAA ACT TCA GGG TGA CCA AAA AAT CA-3') as the reverse
121 primer resulted in an amplification of a 658-bp fragment. For library preparation, a three-step
122 polymerase chain reaction (PCR) was conducted. The first PCR was performed in a total volume
123 of 20 μ l PCR mixture comprising 10 μ l of TaqTM HS Low DNA (TaKaRa, Kyoto, Japan), 0.4 μ l
124 each of 10 μ M forward and reverse primers, 17.2 μ l of ultrapure water, and 2.0 μ l of template
125 DNA. The PCR conditions were as follows: 35 cycles at 94°C for 5 s, 50°C for 30 s, 68°C for 10
126 s; and a final extension at 68°C for 7 min. The fragment size of amplicons and the concentrations
127 were verified by electrophoresis using the Agilent 2100 Bioanalyzer DNA7500 kit (Agilent,
128 Santa Clara, CA, USA). PCR products were purified using the Agencourt AMPure XP (Beckman
129 Coulter, Brea, CA, USA), and the purified products were used as templates for the following
130 PCR. The second PCR was performed using Ex Taq Hot Start Version (TaKaRa, Kyoto, Japan)
131 to add the overhang sequences that required amplification with the Nextera XT Index Kit for
132 Illumina MiSeq analysis. The PCR conditions were as follows: 94°C for 2 min; followed by 5
133 cycles of 94°C for 30 s, 50°C for 30 s, 72°C for 30 s; and a final extension at 72°C for 5 min.
134 The amplicons were purified in the same manner as those obtained from the first PCR, and the
135 purified products were used as templates for the next PCR. The third PCR was performed using
136 Ex Taq Hot Start Version and Nextera XT Index Kit v2 Set A (Illumina, San Diego, CA, USA).
137 The PCR conditions were followed: 94°C for 2 min; followed by 8 cycles of 94°C for 30 s, 50°C
138 for 30 s, 72°C for 30 s; and a final extension at 72°C for 5 min. After purification, the final PCR
139 amplicons were quantified using the Qubit dsDNA High Sensitivity Kit. The sequencing of
140 prepared libraries was performed using MiSeq, according to the manufacturer's instructions.

141

142 **2.3 Bioinformatics**

143 The sequence lengths of the amplicons were 658 bp so the forward and reverse reads in our study
144 could not be merged when using the MiSeq Reagent Kit v3 (600 cycles). Elbrecht and Leese
145 (2017) have demonstrated that invertebrate species could be identified at the reverse side of the
146 COI region through an *in silico* PCR approach. Therefore, we conducted a subsequent analysis
147 using the reverse side sequence. Initially, the raw sequence reads were subjected to the
148 Trimmomatic v0.36 software to discard low-quality sequences and read sequence lengths of
149 <150 bp. The filtered reads were clustered into operational taxonomic units (OTUs) using
150 QIIME (Caporaso et al., 2010), with an identity cut-off value of 97%, which is a common
151 approach for invertebrate metabarcoding analyses (Macher et al., 2018). Subsequently, OTUs
152 with singleton sequences were removed. The most frequently occurring sequences in each OTU
153 were extracted as representative sequences. The assignment was performed against 3,433,026
154 sequences retrieved from the National Center for Biotechnology Information (NCBI) database
155 using the following search criteria: cytochrome [all fields] AND oxidase [all fields] AND
156 mitochondrion [filter]. After the assignment, eight orders (i.e., Ephemeroptera, Plecoptera,
157 Trichoptera, Diptera, Coleoptera, Odonata, Megaloptera, and Hemiptera) that mostly include
158 aquatic insect species were extracted using the QIIME script "filter_taxonomy_from_table.py."
159 Subsequently, representative sequences of the extracted OTUs were subjected to a chimera

160 check. Taxonomic identification was performed using a BLAST search and the QIIME script
161 “assign_taxonomy.py.” Because the traditional environmental assessment indices (%EPT,
162 %Diptera, and %Chironomidae) require at least family-level taxonomic identification, we
163 employed two thresholds for taxonomic identification: 97% identity for genus-level assignment
164 and 85% identity for family-level assignment. For macroinvertebrates, a threshold of 97% is
165 commonly used for species/genus-level assignment (Hebert et al., 2003) but may cause a loss of
166 sequence depth. According to our investigation and using a subset of sequence data of EPT
167 registered in NCBI, the sequence identity for the COI region of EPT was 99% at the intraspecific
168 level, 85% at the intragenus level, 83% at the intrafamily level, and 80% at the intra-order level
169 (see Text S1). Therefore, a BLAST assignment was conducted with a minimum identity of 97%
170 (assigned at the genus-level) and 85% (assigned at least at the family-level) and a maximum e-
171 value of 10^{-50} (Fernández et al., 2018). To compare the communities among samples, we
172 subsampled the number of sequences in each sample by a uniform number. According to the
173 smallest numbers of sequence reads in the samples, either 250 reads for the family-level analysis
174 or 150 reads for the genus-level analysis were randomly selected (see Table S2).

175

176 **2.4 Aquatic insect sampling using a Surber net survey and measurement of environmental** 177 **parameters**

178 After eDNA sampling but within the same day, traditional aquatic insect collection was
179 conducted using the Surber net survey method. A Surber net of 250- μ m mesh size in a 30 \times 30-
180 cm quadrat at randomly selected riffle and pool habitats at each site (total collection area: 0.18
181 m²/reach) was used. Collected invertebrates were placed in 99.5% ethanol and morphologically
182 identified using a stereomicroscope (Leica MZ APO; Leica, Germany) by referring to the
183 identification key for the aquatic insects of Japan (Kawai and Tanida, 2018). Because
184 morphological identification was difficult for some aquatic insects, particularly Chironomidae
185 and some Baetidae individuals, population abundance and richness were summarized at the
186 family level.

187 At the same time as aquatic insect sampling, environmental parameters such as water
188 temperature, electrical conductivity (EC; TOA-DKK CM-21P; Japan), and pH (TOA-DKK HM-
189 20P; Japan) were measured in the field. River water samples of 50 ml were collected from each
190 site to obtain the concentrations of total phosphorus (TP) and total nitrogen (TN) measured in
191 our laboratory using a QuAAtro-2HR (BLTEC Corporation, Japan).

192

193 **2.5 Community structure analysis**

194 We assessed the dissimilarity in community structures of aquatic insects using eDNA or Surber
195 net survey data. Community dissimilarities were calculated based on the Sørensen index (binary
196 Bray–Curtis index) using the presence/absence of OTU data for eDNA and the detected taxa for
197 the Surber net survey using the “vegan” package (Oksanen et al., 2019) in R ver. 3.4.0 (R core
198 team, 2018). Using the ordination of dissimilarity, nonmetric multidimensional scaling (nMDS)
199 was performed to visualize the similarity in community structures and the “metaMDS” function

200 in the “vegan” package. Furthermore, the correlations between community structures and
201 environmental parameters (i.e., water temperature, EC, TN, and TP) were determined using the
202 “envfit” function in the “vegan” package.

203

204 **2.6 Environmental assessment indices**

205 The applicability of biological information obtained from eDNA to existing environmental
206 assessment indicators, namely, EPT index, Diptera index, and Chironomidae index (Reynoldson
207 and Metcalfe-Smith, 1992), was evaluated. These indices are the ratios of the number of
208 individuals/richness of EPT taxa to the total number of individuals/richness of the eight orders
209 observed in the samples (Ephemeroptera, Plecoptera, Trichoptera, Diptera, Coleoptera, Odonata,
210 Megaloptera, and Hemiptera). For the Surber net survey data, the EPT index was calculated
211 using the abundance or the richness at the family/genus-level of EPT. The Diptera/Chironomidae
212 index uses the same method as the EPT index, but using Diptera/Chironomidae instead of EPT.
213 For eDNA-analyzed samples, each index was calculated using the richness of OTUs or groups
214 by taxonomic name at the assigned family/genus-level. Here, the OTU richness refers to the
215 number of OTUs included in the sample, and the assigned family/genus richness refers to the
216 number of families/genera included in the sample (see Text S2 for formulas).

217

218

219 **Results and Discussion**

220 **3.1 The community structure of aquatic insects revealed by eDNA analysis**

221 A total of 1,235,176 sequences (50,728–168,413 sequences/sample) passed the sequence quality
222 filter (Table S2 for the detail of metabarcoding outputs). These sequences were used to create
223 OTUs based on 97% sequence identity. As a result, 90,948 OTUs were formed. Among these,
224 66,176 OTUs comprised just one sequence (singletons), which were excluded from the analysis.
225 Finally, a total of 1,169,000 sequences (47,443–161,461 sequences/sample) generating 24,773
226 OTUs were analyzed. After a BLAST search, we found that 8.0% of the total sequences were
227 assigned to aquatic insects at the family level (sequence identity \geq 85%, see Materials &
228 Methods 2.3) and only 4.1% of them were assigned at the genus level (sequence identity \geq 97%)
229 (Table S2).

230 eDNA metabarcoding detected 93 families and 104 genera before subsampling, and after
231 subsampling, 63 families and 75 genera were detected (sequence depths were 250 reads for the
232 family-level assignment and 150 reads for the genus-level assignment; see Tables S2 and S3 for
233 the OTU tables before subsampling). A total of 26 families were common with the Surber net
234 survey results. Even after subsampling, the total number of taxa detected by the eDNA method
235 was nearly double than that detected by the Surber net survey method (Table S4). Specifically,
236 eDNA detected 27 genera of Chironomidae (Diptera); however, these genera could not be
237 distinguished by morphological identification. In addition, eDNA detected taxa that were mostly
238 distributed in riparian/terrestrial habitats (e.g., Culicidae; Diptera, Cicadidae; Hemiptera) and
239 lentic habitats (Aeshnidae and Epiophlebiidae; Odonata).

240 According to the subsampling results of eDNA, the taxa detected in both the months were 26
241 families/25 genera, those detected in July alone were 26 families/41 genera, and those detected in
242 November alone were 11 families/9 genera. Among these, we found that the three families
243 (Ephemerellidae, Chironomidae, and Simuliidae) were commonly detected among all sites in
244 both the months (see Table S5 for details of assignment results at the family level). Conversely, a
245 number of unique taxa were detected in the communities at site H1 (11 genera in July and 9
246 genera in November; see Table S6 for details of assignment results at the genus level).

247 Previous studies have reported that compared with the use of a traditional survey method, the use
248 of eDNA in lotic systems tends to enable the detection of more taxa (Macher et al., 2018); this is
249 in contrast with the case in pond systems wherein DNA transportation is very low (Hajibabaei et
250 al., 2019a). This is because DNA is transported downstream in lotic systems, which results in the
251 additional detection of upstream communities that are overlooked by traditional methods. It has
252 been reported that fish eDNA is decomposed and transported after release from organisms, with
253 a 73% decrease in eDNA concentration within 900 m downstream of the source (Nukazawa et
254 al., 2018). Even 50–250 m downstream of the source, eDNA was not reported to be detected
255 when the target organisms' abundance or biomass is small (Jane et al., 2015; Pilliod et al., 2014).

256 Thus, the DNA sampled in rivers probably includes some DNA originating from abundant
257 organisms, within an approximate distance of 1 km upstream. While the source materials of
258 eDNA differ depending on the organism [e.g., mucus for fish (Takeuchi et al., 2019), saliva for
259 terrestrial mammals (Rodgers and Mock, 2015; Ushio et al., 2017), and exuvia for aquatic
260 arthropods (Deiner and Altermatt, 2014)], the nature of eDNA in lotic systems may exist in a
261 similar manner. The interval between sampling sites in our study was approximately 3–5 km;
262 therefore, the eDNA contamination among samples was assumed to be negligible. In addition, Jo
263 et al. (2017) have shown that amplification of longer DNA fragments (719 bp vs. 127 bp) is
264 more effective in reflecting real time biological information. The length of the amplicons in the
265 present study was relatively long (658 bp); therefore, the community structures obtained using
266 eDNA were mainly based on DNA that might have been generated recently and transported from
267 a closer area. In addition, eDNA can be used to detect taxa that are usually difficult to capture
268 using the Surber net survey method in lotic systems such as terrestrial organisms (Deiner et al.,
269 2016b; Mächler et al., 2014). As shown in previous studies, terrestrial and semiterrestrial taxa
270 were also detected in our samples. These results indicate that the eDNA sampled from river
271 ecosystems provides a diverse taxonomic list that differs from the traditional Surber net sampling
272 method.

273 From the results of the family-level assignment, nine families were not detected in any eDNA
274 samples but were detected with Surber net sampling, namely, two ephemeropteran (Isonychiidae
275 and Siphonuridae), one plecopteran (Chloroperilidae), one trichopteran (Apataniidae), one
276 dipteran (Blephariceridae), and four coleopteran (Gyrinidae, Hydrophilidae, Psephenidae, and
277 Ptilodactylidae) taxa. Of these, Isonychiidae, Blephariceridae, Hydrophilidae, Psephenidae, and
278 Ptilodactylidae have mismatched sequences with the primers used in this study, which results in
279 failure of PCR primer amplification. Therefore, the primers should be modified or new primers

280 should be developed to analyze these five families. Some refined primer sets for the
281 metabarcoding of aquatic invertebrates have been developed (Elbrecht and Leese, 2017;
282 Hajibabaei et al., 2012). Hajibabaei et al. (2019b) also suggested that the use of multiple
283 universal primers enables coverage of a broader range of taxa.

284 In addition to the primer issue, the sequence identity threshold used for taxonomic identification
285 can be another problem for eDNA analysis. To evaluate the discrepancy between the reference
286 library and the query sequence, we investigated intraspecific, intrageneric, and intrafamilial
287 genetic identity (Text S1). As a result, an 85% identity threshold at the family level and a 97%
288 identity threshold at the genus level were employed for taxonomic assignment in this study.
289 However, this threshold might not have been achieved by some species and thus gone
290 undetected. To overcome this issue, reference sequence data should be accumulated.
291 Geographically separated intraspecies have low sequence identity of the COI gene (Takenaka
292 and Tojo, 2019). Therefore, the accumulation of genetic information of local aquatic insects and
293 the construction of a database are necessary to improve the taxonomic assignment and to avoid
294 false negative results. There is an urgent need to overcome these issues because failure to detect
295 some specific taxonomic groups could directly affect the assessment results based on the
296 richness of taxa.

297

298 **3.2 Relationships between communities and environmental parameters**

299 Community dissimilarities among all samples using the Sørensen index for eDNA analysis and
300 Chao index for Surber net survey data were plotted on nMDS axes (Figure 2). Visually, nMDS
301 showed monthly differences among communities (July or November). Ordination was
302 significantly correlated with water temperature in all three datasets (function “envfit”; eDNA at
303 the genus level: $R^2 = 0.56$, $p = 0.016$, eDNA at the family level: $R^2 = 0.81$, $p = 0.001$, and Surber
304 net survey data at the family level: $R^2 = 0.57$, $p = 0.021$; Table S7). Bista et al. (2017) have
305 shown that eDNA targeting the chironomid community (Diptera) in a lake system can
306 distinguish seasonal differences across communities. Similarly, the present study demonstrated
307 that eDNA analysis targeting the aquatic insect community in a river system revealed seasonal
308 differences.

309 In addition, nMDS showed that the uppermost site of the Hirose River (H1) was plotted in
310 isolation from the other sites in both months and for all three datasets (Figure 2). This was
311 understandable because the landscape of H1 differs from those of the other sites; it is a mountain
312 stream, and its inhabitants differ from those found in other middle and lower reaches. Indeed,
313 eDNA analysis detected multiple unique taxa at H1 that made differences among communities
314 (Table S5 and S6). Conversely, in the Surber net survey, the differences among communities
315 were revealed by the absence of taxa that could be found in the other sites, rather than the
316 presence of unique taxa at H1 (Table S4). Thus, both eDNA analysis and Surber net survey
317 clearly illustrated differences among communities depending on landscapes (mountain stream or
318 middle/lower reach), whereas the factors attributable to the differences among communities
319 differed between methods.

320 Furthermore, the ordination based on eDNA data was significantly correlated with TN
321 concentration (function “envfit”; eDNA at the genus level: $R^2 = 0.50$, $p = 0.043$, eDNA at the
322 family level: $R^2 = 0.51$, $p = 0.046$); however, no such correlation was noted for the Surber net
323 survey data. TN concentration may not only be a direct proxy for organic pollution, but also an
324 indirect proxy for site characteristics. That is because agricultural and urban land use increase in
325 the lower area in our study area, so the larger the stream order, the higher the TN concentration
326 (Table S1). Previous studies reported that eDNA can distinguish geographical changes in
327 communities in river systems for various fauna (fungi: Matsuoka et al., 2019;
328 macroinvertebrates: Hajibabaei et al., 2019b; Fernández et al., 2019). Similarly, the present study
329 showed that the community differences revealed by eDNA analysis were related to the site
330 characteristics by water quality rather than by geographical location.

331

332 **3.3 Environmental assessment indices derived from eDNA**

333 The relationship between the TN concentration and biological environmental assessment indices
334 (%EPT, %Diptera, and %Chironomidae) was examined (Figure 3) because TN could be assumed
335 to be a chemical indicator of water pollution. The results showed that %EPT and
336 %Chironomidae derived from eDNA at genus-level resolution showed sufficient effect sizes
337 with significant rank correlation with TN (Spearman’s rank correlation; %EPT: $r = -0.59$, $p =$
338 0.049 , %Chironomidae; $r = 0.69$, $p = 0.014$; Table S7). However, the eDNA and Surber net
339 survey results at family-level resolution did not show significant correlations. These results
340 indicate that when eDNA data are obtained at a higher taxonomic resolution, the sensitivity of
341 biological indices to environmental factors can be improved. Conversely, the sensitivity could be
342 impaired if biological indices are obtained using coarse taxonomic resolution.

343 Emilson et al. (2017) have reported that assessment indices (i.e., the EPT index and
344 Chironomidae index) derived from DNA metabarcoding using macroinvertebrate tissue samples
345 were highly correlated with indices derived from the morphological survey. The present study
346 demonstrated that the indices obtained from eDNA can also be used as a new assessment
347 method. Although the biological indices obtained from eDNA vs. the traditional method were
348 only compared in terms of TN concentration in the present study, new environmental indicators
349 can be developed using eDNA data by comparison with more chemical pollution indicators such
350 as biochemical oxygen demand and chemical oxygen demand.

351 While our manuscript was under review, a study that detected macroinvertebrate eDNA and
352 applied this to the environmental status assessment of a river was reported (Fernández et al.,
353 2019). In that report, the environmental assessment score from the IBMWP index was calculated
354 based on the presence/absence of indicator macroinvertebrates at family-level identification. The
355 report demonstrated that eDNA data could be used for the monitoring program that they used.
356 Conversely, the present study showed that the EPT and Chironomidae indices calculated using
357 OTU richness required genus-level resolution and showed a clearer response to organic pollution
358 compared with family-level resolution.

359

360

361 **Conclusions**

362 eDNA can be used to describe differences among community structures of aquatic insects in two
363 seasons in river systems. In addition, compared with the ordination of community derived from
364 traditional sampling methods, that derived from eDNA analysis was correlated with the degree of
365 water pollution. EPT and Chironomidae indices at the genus level derived from eDNA analysis
366 data showed significant correlations with TN concentration, whereas indices derived from Surber
367 net survey and eDNA analysis data at the family level did not. Environmental assessment indices
368 based on ecological information but not requiring biotic samples have significant advantages
369 such as they can be applied to places where the capture of organisms is restricted. In addition,
370 eDNA analysis can derive benefits related to sampling such as minimal sampling effort, high
371 taxonomic resolution, and high applicability to a broad range of species. We believe that eDNA
372 analysis is useful for monitoring the long-term trends of changes in ecological community
373 structure associated with environmental changes such as climate change and other anthropogenic
374 activities, and it facilitates environmental assessment with nonbiotic samples.

375

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Figure 1

Study area

Sampling sites in Hirose River (from the upmost site, H1, H2, H3; shown as orange circles) and in Natori River (N1, N2, N3; shown as blue circles) in northeast Japan. This map was modified using a digital map provided by the Geospatial Information Authority of Japan.

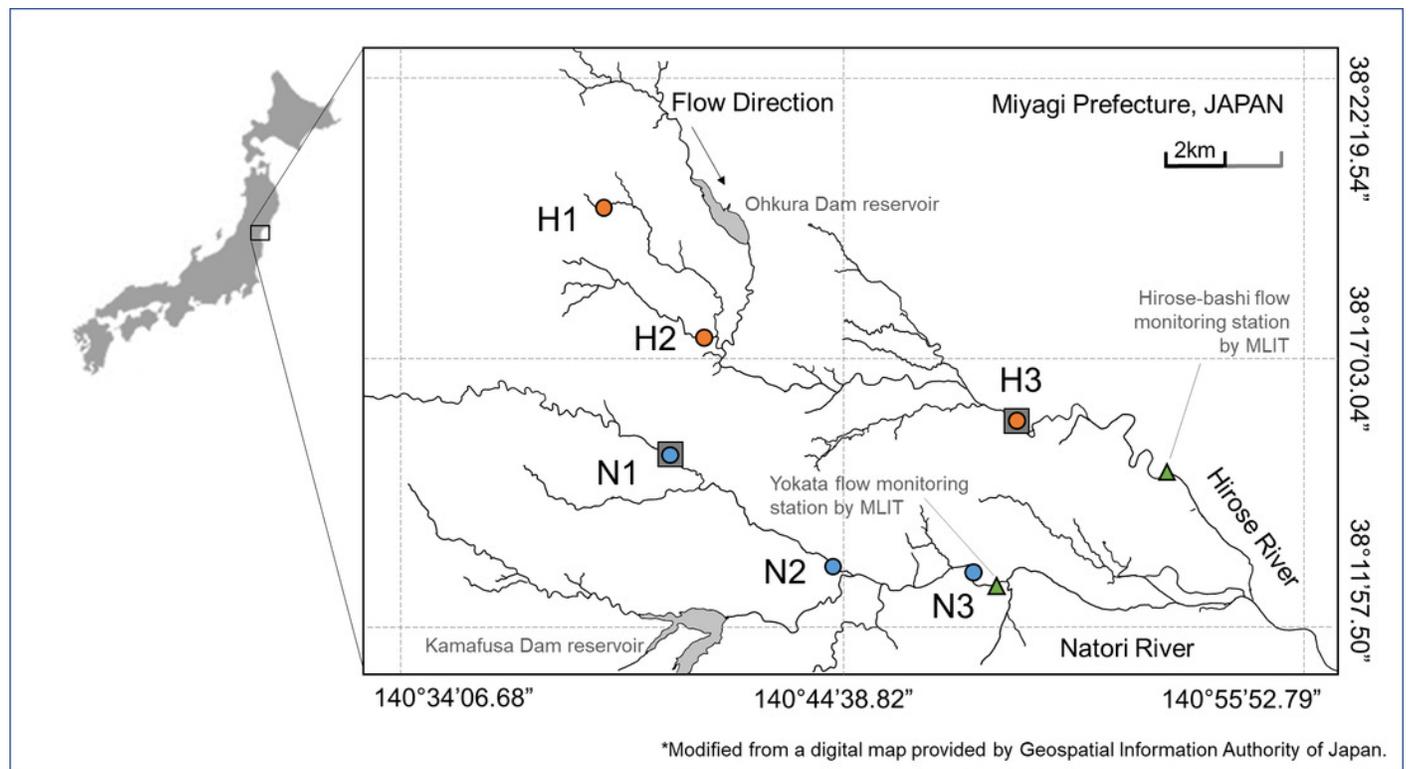


Figure 2

Non-metric multidimensional scaling (NMDS) using Sørensen dissimilarity index (P/A data).

Each panel shows the communities derived from (A) eDNA at genus level identification, (B) eDNA at family level identification, (C) Surber net (at family level identification). eDNA data were based on OTU richness with subsampled by 150 reads depth for genus level and 250 reads depth for family level. The same month is enclosed by ellipses (orange: July, blue: November). Site name and month (j: July, n: November) are displayed. The environmental parameters are shown by arrows (solid: $R^2 > 0.5$ with p-value < 0.05 , dotted: $R^2 \leq 0.5$ with p-value ≥ 0.05). The length of the arrow is proportional to the correlation between parameters and the community ordination.

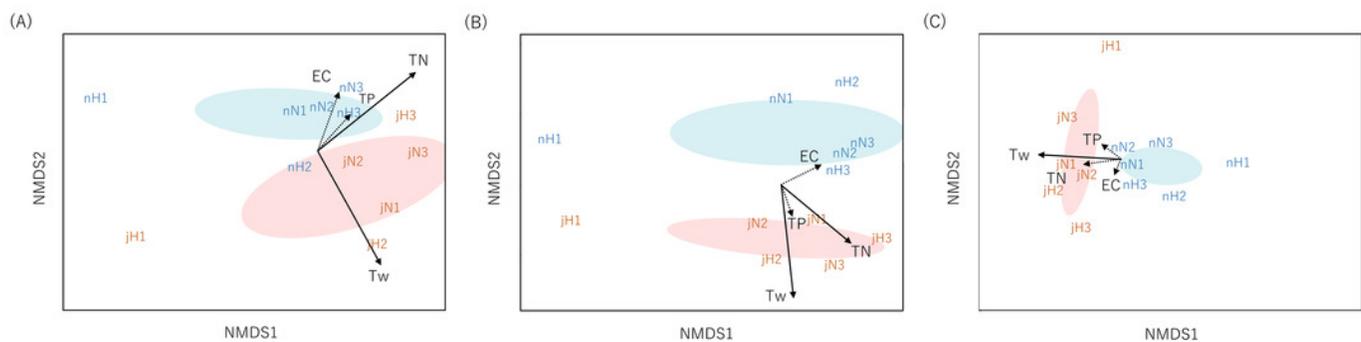


Figure 3

Relationships between biological assessment indices and TN concentration.

The first (panels A-C), second (D-F), and third (G-I) rows show %EPT, %Diptera, and %Chironomidae, respectively. The first, second, and third columns show results based on eDNA at genus level identification, eDNA at family level identification, and Surber net survey, respectively. The indices are calculated using OTU richness data for eDNA (subsampled) and abundance data for Surber net data. Seasonal differences are represented by colors (orange: July, blue: November) and river differences are represented by plot styles (circle: Hirose River, triangle: Natori River), respectively.

