Monitoring fish communities through DNA metabarcoding in the fish pass system of the second largest hydropower plant in the world

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24 Abstract: The Itaipu Hydroelectric Power Plant is the second largest in the world in power 25 generation. The artificial barrier created by its dam imposes an obstacle for fish migration. Thus, 26 in 2002, a fish pass system, named Piracema Channel, was built to allow fish to access areas upstream of the reservoir. We tested the potential of metabarcoding to monitor the impact of both 27 28 the dam and associated fish pass system in the Paraná River fish communities and to compare it with traditional monitoring methods. Using a fragment of the 12S gene, we characterized 29 30 richness and community composition based on amplicon sequence variants, operational taxonomic units, and zero-radius OTUs. We combined GenBank and in-house data for 31 taxonomic assignment. We found that different bioinformatics approaches showed similar 32 results. Also, we found a decrease in fish diversity from 2019 to 2020 probably due to the recent 33 34 extreme drought experienced in southeastern Brazil. The highest alpha diversity was recorded in the mouth of the fish pass system, located in a protected valley with the highest environmental 35 heterogeneity. Despite the clear indication that the reference databases need to be continuously 36 37 improved, our results demonstrate the analytical efficiency of the metabarcoding to monitor fish species. 38

39 Key-words: Environmental monitoring, Dam, Fish passage, Neotropical ichthyofauna.

40 Background:

The Itaipu Hydroelectric Power Plant, built at the border between Brazil and Paraguay, is the 41 second largest in the world in power generation¹, second only to the Three Gorges Power Plant 42 in China. With the formation and filling of its reservoir, in 1982^2 , the natural barrier to the 43 migration of fishes of the middle section of the Paraná River (Sete Quedas falls) was replaced by 44 45 the artificial barrier of the Itaipu dam, located 170 km downstream. This artificial barrier (196 m high) caused impacts on the adjacent fish assemblages, such as the reduction in reproductive 46 activity in the first kilometers downstream of the dam³. To allow for fish migration and mitigate 47 the environmental impact of the dam, a fish passage system known as the Piracema Channel was 48 created in 2002, linking the Paraná River to Itaipu's Reservoir⁴. However, the real contribution to 49 the reproductive success of the long-distance migratory species is still under investigation, and 50 51 this channel also allowed for the dispersal of species originally restricted to the lower Paraná River upstream and species originally restricted to the upper Paraná River downstream⁵. These 52 potential impacts are continuous and can interact with natural disturbance, such as several 53 54 droughts as which happened in 2020. In this context, monitoring the impact of both dam and fish pass system in the Paraná River fish communities is essential. 55

Fish diversity estimates in Brazilian freshwater are still imprecise due to the scarcity of complete 56 inventories⁵⁻⁷. Many species are described every year and several groups are in need of 57 taxonomic revision^{5,8}. Furthermore, traditional assessment methods for fish diversity surveys are 58 costly and time consuming, given that they depend on capture (e.g. netting, trawling) or 59 observation^{9,10} and expertise for taxonomic identification¹¹. In this sense, designing methods for 60 cost-effective monitoring fish diversity and community composition is an urgent task. Most 61 sampling efforts in Brazil have historically been primarily funded by the hydroelectric sector, 62 focusing particularly on rivers where power dams were built¹². The areas of the dam construction 63 have some of the most comprehensive knowledge of fish assemblage composition in comparison 64 with other Brazilian regions and therefore offer an ideal opportunity to compare taxonomic 65 surveys with molecular approaches. 66

A promising alternative to traditional taxonomic surveys and biomonitoring methods is the use of environmental DNA (eDNA), combined with a high-throughput sequencing approach, as in the case of metabarcoding¹³. This technique has the advantage of obtaining DNA from environmental samples, such as water, without first isolating the target organism and therefore can sample entire communities¹⁴. Metabarcoding is a powerful tool for biodiversity assessment that has been widely used for several purposes and different taxonomic groups^{15–17}, and is considered a transformative technology for the entire field¹⁸. However, some limitations, such as the relative scarcity of DNA sequences for several species, which is even more problematic in highly diverse regions such as the Neotropics¹⁹, may create constraints that hamper its full application^{20,21}.

77 The absence of a comprehensive DNA reference database may lead to a misidentification of several species. Therefore, putting together a curated and complete DNA reference database is 78 fundamental for species identification through a metabarcoding approach⁷. But, even with an 79 incomplete DNA reference database, the use of molecular units, such taxonomic units clustered 80 by similarity (operational taxonomic units - $OTUs^{22}$) or unique sequences (e.g. amplicon 81 sequence variants - ASVs²³, or zero-radius OTUs - ZOTUs²⁴) allows for diversity monitoring in 82 the context of biodiversity assessment in megadiverse biomes. Such estimates without 83 comprehensive species identification limit ecological conclusions but allowed for monitoring of 84 natural and artificial impacts^{16,25}. The metabarcoding approach has been successfully used for 85 molecular identification of several vertebrate groups in temperate regions^{26,27}, monitoring of 86 endangered species such as freshwater fish in Australia²⁸ and turtles in the United States²⁹, and to 87 describe biodiversity even with limited taxonomic identification³⁰. 88

In this context, our goal here is to describe an effective survey protocol for detecting fish assemblages through eDNA metabarcoding in an ecologically complex and highly diverse freshwater system, the Piracema channel, that connects the Paraná River with Itaipu Reservoir. For this, we used an in-house molecular database of fishes occurring in the channel complemented with GenBank sequences. We also describe fish alpha diversity and community structure in the Piracema channel system. Additionally, we compare our metabarcoding results with the traditional sampling campaigns made between 2017- 2021.

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97 Material and Methods:

Study area – Our study was conducted at the Piracema channel (Fig. 1), a fish pass system connecting the Paraná River with the Upper Paraná River floodplain (main reservoir). For the traditional taxonomic survey, we sampled three points at mouth of channel at Paraná River (Fig. 1, blue square), the main lake at the Piracema channel (Fig. 1, red circle), and the reservoir near the water intake to the Channel (Fig. 1, green triangle) between 2017 to 2021. Fish were collected monthly during the fish reproductive period (October to March), and once during the winter (July or August), employing active and passive methods (Table 1).

Table 1. Fish sampling methods at the Piracema Channel

Fish sampling method	Quantitative and qualitative aspects
Gill nets	Mesh sizes: 1 a 10 cm (adjacent knots), each one 10m long and 1 to 1,5m high
Longlines	30 hooks, 10 of each size: /10, /8 and /6, fish pieces as bait
Cast nets	Mesh sizes: 3 e 6 cm (adjacent knots)
Electrofishing	Smith-Root, backpack electrofisher, 600V, 30Hz DC

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For each point, gill nets and longlines were set out in the afternoon (16:00 h) and inspected every 108 4 hours during a 24 h cycle; cast nets were operated 3 times each mesh, after every gear 109 inspection. Boarded electrofishing was operated two times in each point, at dawn and at dusk, 110 111 covering 100m of the environment margin each time. Fish were euthanized by immersion in benzocaine solution, following current legislation³¹, and identified accordingly Britski et al.³², 112 Ota et al.³³ and Neris et al.³⁴. Fragments of muscle were collected with a scalpel, placed into 2 ml 113 tubes filled with 99.8% ethanol and stored at 4°C until processing. Voucher specimens are 114 housed in the Nupelia-UEM fish collection. 115

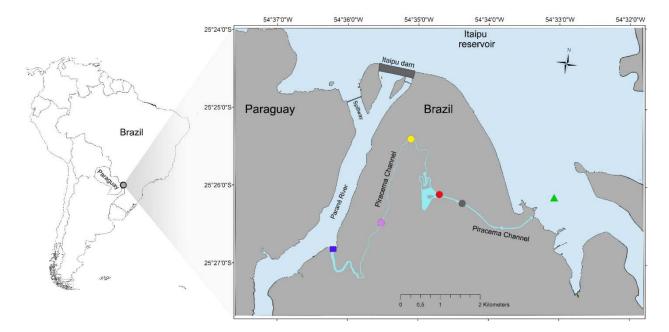


Figure 1. Sampling location. The map shows the sampling location of each collection point. We sampled one point at mouth of channel at Paraná River (blue square), four points along the Piracema Channel (circles; Bela Vista River 1 = purple, Bela Vista 2 = yellow, Brasilia stream = gray, and lake = red), and one point at Itaipu's reservoir (green triangle). Up at figure is possible to visualize the Itaipus' dam that created the reservoir. Inset panel shows the location of Itaipu's dam in relation to South America. Map was created in QGIS v.3.6.2 software ⁸⁸.

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For metabarcoding, we sampled one site at the mouth of channel at Paraná River (Fig. 1, blue square), four sites along the Piracema Channel (Fig. 1, circles), and one site at Itaipu Reservoir (Fig. 1, green triangle). Each sampling point was collected in sextuplicate. All six sites were sampled in 2019 and three sites (mouth of channel at Paraná River, lake at Piracema channel [red circle], and the reservoir) were sampled again in 2020, totaling 54 samples. All sampling sites were provided with GPS coordinates.

Sampling design for molecular analysis – We collected water by partially submerging a one litter polypropylene bottle. The objective was to sample water at the air/water interface. After water collection, bottles were closed and cleaned with a 10% sodium hypochlorite solution, following by rinsing with distilled water. We used gloves which were changed in between each new sampling replicate to reduce the risk of cross-sample contamination.

After the collection and cleaning steps, the bottles were stored in polystyrene boxes containing artificial ice to maintain the temperature of the samples at 4 to 10 °C. The samples were filtered, on the same day of collection, using nitrocellulose membranes (0.45 μ m pore) with the aid of a vacuum pump. Filters were kept in 100% ethanol under refrigerated conditions until molecular analysis was performed. All dry samples were processed at the ATGC laboratory at the Universidade Federal do Paraná (UFPR).

141 *DNA extraction* – For total DNA extraction, we kept the collected filters at room temperature to 142 allow the residual ethanol to dry completely. After dried we extract the DNA using magnetic 143 beads (microspheres surrounded by magnetite and carboxyl), which bind to DNA (carboxyl bond 144 - DNA) by the process of Solid Phase Reversible Immobilization (SPRI). The DNA extract was 145 stored at –20 °C until the amplification. The extraction and quantification processes were carried 146 out in separate rooms, as suggested by Pie et al.³⁵. We checked the DNA concentration using 147 both a spectroscope (Nanodrop, Thermo, USA) and a fluorimeter (Qubit, Invitrogen, USA).

148 PCR amplification - We targeted the 12S rRNA gene using the MiFish forward (5'-149 GTCGGTAAAACTCGTGCCAGC-3') (5'and reverse CATAGTGGGGTATCTAATCCCAGTTTG-3') primers designed by Miya et al.³⁶ to yield 163-150 185 bases long fragments. Amplification was performed in a total volume of 20 µl in GoTaqG2 151 152 system (Promega, USA), 500 nM of forward and reverse primers, and 20 ng of DNA template. 153 The PCR conditions consisted of an initial denaturation step of 2 min at 95 °C and then 25 cycles 154 of denaturation at 94 °C for 30 s, hybridization at 55 °C for 45 s, and elongation at 72 °C for 30 s, followed by a final elongation at 72 °C for 5 min and finishing at 4 °C. To avoid PCR 155 inhibition BSA (0.5 μ g/ μ l) was added to the reaction as suggested by Boeger et al³⁷. The quality 156 of amplification was verified on a 1.5% agarose gel in TBE buffer (9 mM TRIS, 9 mM boric 157 158 acid, 1 mM EDTA), stained with SYBR Safe DNA Gel Stain (ThermoFisher Scientific, 159 Country). All replicates from each sampling point were amplified to increase the chance of 160 detecting rare species. The PCR product was then diluted (20x) and used as a template for the addition of adapters in the second PCR. Indexing was performed for Illumina MiSeq sequencing 161 162 (Illumina, USA), using the above PCR system with Nextera indexes (Illumina) in a total volume of 10 µl. PCR conditions were an initial step of 95 °C for 3 min, following by 12 cycles of 163 denaturation at 94 °C for 30 s, hybridization at 55 °C for 45 s, and elongation at 72 °C for 30 s, 164

followed by a final elongation at 72 °C for 5 min and finishing at 4 °C. We checked the DNA
concentration in a Qubit[®] fluorimeter (Invitrogen, USA), normalized and pooled the PCR
products following the Illumina protocol. The samples were sequenced at GoGenetic (Curitiba,
Brazil) using Illumina MiSeq Reagent 600V3 (PE 300b). Three DNA extraction negative
controls were included for sequencing. We included three negative controls from the DNA
extraction to sequencing. The raw sequences are deposited in GenBank under Bioproject
PRJNA750895 (biosamples SAMN20500524 – SAMN20500577).

Sequence analyses and taxonomic assessment – For the amplicon sequence variants (ASVs) 172 approach, we used the Cutadapt package³⁸ in Python v.3.3³⁹ to remove primers. We then used the 173 DADA2 package²³ in R v. $4.0.2^{40}$ to quality filter reads, merge sequences, remove chimeras, and 174 to infer ASVs. We excluded reads with ambiguous bases (maxN=0). Based on the quality scores 175 of the forward and reverse sequences, each read was required to have <3 or <5 errors, 176 respectively (maxEE=c (3,5), truncQ=2). Therefore, ASVs were inferred for forward and reverse 177 178 reads for each sample using the run-specific error rates. To assemble paired-end reads, we 179 considered a minimum of 12 base pairs of overlap and excluded reads with mismatches in the 180 overlapping region. Chimeras were removed using the consensus method of "removeBimeraDenovo" implemented in DADA2. 181

For operational taxonomic units (OTUs) and zero-radios OTU (ZOTUs) analyses, we used the USEARCH/UPARSE v.11.0.667 Illumina paired reads pipeline⁴¹ to primer remove, quality filtering, dereplicate and sort reads by abundance, to infer OTUs and ZOTUs, and to remove singletons. We filtered the sequences to discard chimeras and clustered sequences into OTUs at a minimum similarity of 97% using a 'greedy' algorithm that performs chimera filtering and OTU clustering simultaneously and the UNOISE algorithm to denoised sequence as zero-radios OTUs to create or ZOTUs table^{41,42}.

We build a reference dataset of DNA sequences for the 205 fish taxa that have been historically recorded in the Itaipu system using the following steps. First, we looked for 12S sequences of these species in GenBank by searching for their corresponding names. We were able to find sequences for 126 species in our reference database (Table S1). Additionally, we created an inhouse database which included sequences for 42 additional species to the 79 species previously identified as present on Itaipu system but not available on GenBank. Sequences for the in-house 195 database were obtained via Sanger sequencing of tissue samples and were uploaded to GenBank (accession numbers MZ778813- MZ778856). We manually blasted all sequences against the 196 197 NCBI GenBank database to verify misidentification or problematic sequences (e.g. blasted in the 198 different family). In total, our reference database included 168 (82%) sequences from the 205 199 taxa recorded in the Itaipu system. Finally, we blasted the ASVs, OTUs, and ZOTUs sequences with our reference database to verify the taxonomic composition using the "Blastn" function of 200 the program Blast+⁴³ with an e-value of 0.001. We kept ASVs, OTUs, and ZOTUs that have 201 matched with a fish species at minimum level of 75% similarity (as these sequences are probably 202 203 fishes species not present in our reference database), and considered identified species just ASVs, OTUs, and ZOTUs that matched in a minimum level of 97% similarity. ASVs, ZOTUs 204 205 and OTUs present with a proportion > 0.01% of reads in the sum of three negative controls were discarded (13 ASVs, 1 ZOTU, and 5 OTUs). 206

Statistical analysis – We conducted all analyses in R using RStudio⁴⁴. We used the tidyverse package v. $1.3.0^{45}$ for data curation and ggplot2 v. $3.3.2^{46}$, ggfortify v. $0.4.11^{47}$, gridExtra v. 2.3^{48} , and ggpubr v. $0.4.0^{49}$ for data visualization (scripts in Appendix 1).

210 For analysis of alpha and beta diversity with metabarcoding data, we made the analysis at ASVs, OTUs, and ZOTUs level. Since the number of observed ASVs, ZOTUs, and OTUs is dependent 211 212 on the number of reads, we rarefied all samples to the lowest number of reads obtained from any 213 one plot (157 for ASVs, 147 for ZOTUs, and 219 for OTUs; Fig. S1) using the "rarefy" function with the vegan v.2.5.7⁵⁰ R package. Because in the ZOTUs table the minimum reads of a plot 214 was nine, we used the second lower value to avoid having to downsize the other samples to such 215 a low number of reads⁵¹. Because rarefying of counts is considered inappropriate for detection of 216 differentially abundant species⁵¹, even more with so different sampling depth as in our case, we 217 also calculated true effective number of ASVs, ZOTUs, and OTUs of order q=1, which is 218 equivalent to the exponential of the Shannon entropy⁵², using the function "AlphaDiversity" of 219 the Entropart v.1.6.7⁵³ R package. The effective number is more robust against biases arising 220 from uneven sampling depth than the simple counts of ASVs, ZOTUs, and OTUs^{51,54}. 221 Additionally, for alpha diversity, we also calculated the ASV, OTU, and ZOTU richness (the 222 number of ASV, OTU, and ZOTU per point), Chao1, and Fisher's alpha diversity (i.e., the 223

relationship between the number of ASV, OTU, and ZOTU in any given point and the number of
reads of each ASV, OTU, and ZOTU) using the phyloseq v.1.34.0⁵⁵ R package.

226 For beta diversity, we also used rarefaction (with "rrarefy" function of vegan package) and hill number (with "varianceStabilizingTransformation" function in DESeq2 v.1.28.1⁵⁶ R package) to 227 normalize our data. While rarefaction normalizes data by random subsampling without 228 229 replacement, the hill number transformation normalizes the count data with respect to sample size (number of reads in each sample) and variances, based on fitted dispersion-mean 230 relationships⁵⁶. We then constructed two-dimensional Principal Coordination Analysis (PCoA) 231 ordinations of the abundance (reads) and presence/absence data for both rarefied and hill 232 233 numbered data. We used the 'cmdscale' function and Bray-Curtis distances in the vegan package to assess community dissimilarity among all samples in the PCoA. We used the "envfit" method 234 235 in vegan to fit sampling localities and sample year onto the PCoA ordination as a measure of the correlation among the sampling localities with the PCoA axes. 236

For traditional survey data, we calculated the alpha diversity using the observed richness, Chao1, and ACE with the function "estimate", and Shannon index with the function "diversity" both with vegan package. We also constructed two-dimensional Principal Coordination Analysis (PCoA) ordinations of the abundance (reads) and presence/absence data, and used the "envfit" to fit sampling localities and sample year onto the PCoA ordination.

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243 **Results:**

For the traditional surveys, 4,430 fishes were collected, for a total 138 species. Most specimens

were collected at the mouth of channel at Paraná River, with 2,240 (51%) fishes belonging to

105 species. The reservoir showed the lowest number of collected specimens: 1,034 (23%) and a

total of 64 species.

248 For metabarcoding data, we obtained a total of 25,292,218 reads. After all cleaning steps, we 249 kept a total of 4,100,729 sequences belonging to 7,219 ASVs. Of these, 2,683,436 (65% of the 250 total) sequences belonging to 211 ASVs were classified into species corresponding to our reference database in the level of 75% similarity, which included other samples from additional 251 252 projects sequenced together. From our samples we kept 190 ASVs (900,865 reads), of which 121 (64%) were classified in 35 species matches at a level > 97% of similarity (Table 2), which is 253 254 certainly an underestimation of the real number of species, since the other 69 ASVs should belong to species do not present in our database. 255

Table 2. Species identified at level of > 97% similarity, the number of ASVs, ZOTUs, OTUs identified per species (possible intra-specific variation) and the number of reads per species after correction. In bold species that was not registered in one of the pipelines, five in total. It is possible to observe that the reads are more similar for the pipelines of unique sequences (ASVs

260	and ZOTUs), while the clust	ter pipeline (OTUs) keep more reads.
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	Number	Reads	Number	Reads	Number	Reads
Species	of ASVs	of	of	of	of	of
~ • • • • • • • •	01 A5 V5	ASVs	ZOTUs	ZOTUs	OTUs	OTUs
Acestrorhynchus lacustris MG755503.1	1	9151	1	2201	1	2201
Apareiodon affinis NC_015834.1	5	44738	1	44852	1	44852
Astyanax altiparanae	12	109217	2	108603	5	108551
Characidae3	16	129645	7	128540	8	128576
Characidium sp. LC036706.1	3	6357	2	6389	1	6389
Cichla kelberi	3	2738	1	799	1	799
Cichla ocellaris LC069581.1	0	0	1	2019	1	2019
Crenicichla britiskii	2	14565	0	0	0	0
Crenicichla sp.	4	28044	2	27982	4	28004
Crenicichla sp. LC069598.1	0	0	1	14350	1	14350
Eigenmannia limbata MH263669.1	3	2431	2	97	1	97
Geophagus brasiliensis C_031181.1	4	10888	2	10351	2	10351
Gymnotus carapo AP011979.1	3	10408	3	10057	2	13415
Gymnotus sylvius MN583179.1	1	3517	1	3412	3	45
Hemigrammus erythrozonus MT484070.1	3	1153	0	0	0	0
Hemigrammus marginatus MG755550.1	3	7335	1	7352	1	7352

Hemiodus orthonops	4	2695	1	2478	1	2478
Hemisorubim platyrhynchos JF898664.1	2	1727	1	1687	1	1687
Hoplias intermedius KU523584.1	2	11902	1	11274	1	11274
Hoplias sp1	8	46648	5	46823	3	46869
Hoplias sp2	3	4876	2	4915	2	4915
Hyphessobrycon amandae MT484069.1	1	1010	0	0	0	0
Hypostomus affinis KT239013.1	4	14468	2	15871	2	338
Hypostomus albopunctatus	11	89735	2	93752	7	94997
Hypostomus ancistroides	3	5240	3	5239	2	19725
Hypostomus commersoni	2	10371	2	10395	2	10396
Hypostomus gymnorhynchus JN855752.1	3	4110	1	3196	1	3195
Leporellus vittatus LC104399.1	1	1728	1	1654	1	1654
Leporinus elongatus NC_034281.1	1	1375	1	1319	1	1319
Leporinus lacustris	1	3010	1	2973	1	2973
Leporinus octofasciatus	7	112049	3	113810	12	113954
Loricaria sp Mato Grosso KR478070.1	1	78	1	71	0	0
Loricaria sp Orinoco KR478071.1	0	0	1	17	1	17
Megaleporinus obtusidens NC_034945.1	2	8293	1	8228	1	8228
Myloplus tiete	1	915	1	865	1	865
Oligosarcus sp LC145855.1	6	17868	2	18059	1	18034
Oreochromis niloticus MN255618.1	8	9673	4	9799	5	10110
Piaractus mesopotamicus NC_024940.1	0	0	1	291	3	325
Pimelodella cristata MH286807.1	3	5080	2	4922	3	4924
Pimelodus albicans JF898707.1	2	3586	2	3427	1	3424
Pimelodus ornatus JF898680.1	0	0	0	0	1	3
Pirinampus pirinampu	1	453	1	428	1	428
Plagioscion squamosissimus MT080739.1	2	3431	1	3348	1	3348
Platydoras armatulus NC_025585.1	2	3264	2	3151	2	3151
Poecilia reticulata NC_024238.1	3	7459	1	5957	1	5957
Prochilodus lineatus NC_024939.1	8	79673	1	80216	5	80504
Pseudoplatystoma corruscans NC_026846.1	5	18627	3	16746	5	16736
Pseudoplatystoma reticulatum NC_033859.1	19	8235	0	0	3	104040
Pterodoras granulosus AY264087.1	1	16	0	0	1	15
Rhamdia quelen EU179824.1	2	28623	2	26811	8	29105
Salminus brasiliensis NC_024941.1	1	140	1	138	1	138
Salminus hilarii	1	1153	1	1229	1	1229
Satanoperca jurupari LC069577.1	0	0	1	82	1	82
Schizodon borellii	3	10511	2	10611	5	10628
Trachelyopterus galeatus JX899742.1	1	288	1	269	1	340
Zungaro jahu EU179830.1	2	2368	2	2361	1	2361
Total	190	900865	87	879416	123	986767

262 For the OTU and ZOTU analyses, after all cleaning steps, we obtained 2.835,679 and 2.819,524 reads belonging to 855 OTUs and 215 ZOTUs, respectively (considering samples from other 263 264 projects). Of these, 2,660,872 (94%) and 2,657,978 (94%) reads belonging to 136 OTUs and 94 265 ZOTUs, respectively, were classified into species corresponding to our reference database in the 266 level of 75% similarity. For our samples, we kept 123 OTUs (986,767 reads) and 87 ZOTUs (879,416 reads). As the OTUs analysis already classified the sequences by 97% similarity, these 267 268 probably correspond to the number of species present in our samples (89% of species sampled in five years with traditional surveys). Yet, only 37 species belonging to 42 OTUs and 34 species 269 270 belonging to 46 ZOTUs at > 97% similarity were assigned in both analyses. Eighty-one (66%) OTUs and 41 (47%) ZOTUs were identified as a fish species with a similarity lower than 97%, 271 272 representing species not present in our reference database.

All the alpha diversity measures of ASVs, OTUs, and ZOTUs per sampling point varied, with the point at mouth of the channel at Paraná River, in 2019, showing the highest diversity for all molecular units and the lake at Piracema Channel in 2020 the lowest (Fig. 2). For the traditional surveys, the variation was more random, but the mouth of channel at Paraná River also had the highest diversity (Fig. S2).

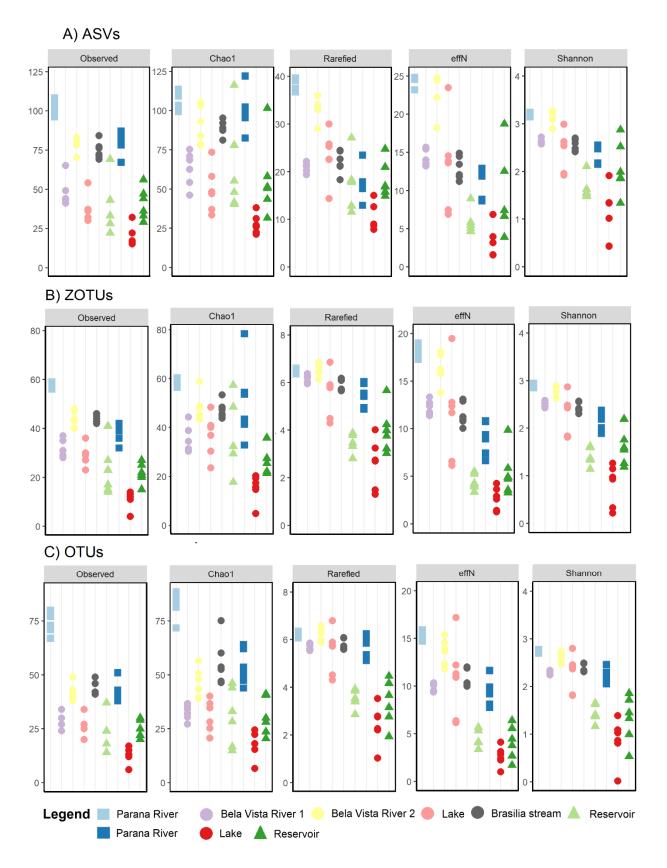


Figure 2. Alpha diversity estimation for A) ASVs, B) ZOTUs, C) OTUS. Alpha diversity varied
by location and by sampling year. Each point is one of the replicates sampled. Colors and
symbols represent collection points (mouth of channel at Paraná River = blue square, Itaipu's
reservoir = green triangle, and Piracema Channel = circles [Bela Vista River 1 = purple, Bela
Vista 2 = yellow, Brasilia stream = gray, and lake = red]), and tone represent year of collection
(light = 2019, dark = 2020).

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Fish communities varied among sampled sites. For the abundance based in hill numbers, the first 286 axis of the PCoA separated the samples by year (envfit: $R^2 = 0.30$ [ZOTUs], 0.34 [ASVs], and 287 0.36 [OTUs], p < 0.001), except for the Itaipu's reservoir, with the positive values associated 288 with 2019 and the negative values associated with 2020 (Fig. 3). The second axis separated the 289 samples by locality with some overlap (envfit: $R^2 = 0.95$ [ZOTUs] - 0.96 [ASVs and OTUs], p < 290 0.001; Fig. 3). For the presence/absence data also based in hill numbers, the overlap was higher 291 but vet the separation by vear (envfit: $R^2 = 0.30$ [ZOTUs], 0.32 [OTUs], and 0.41 [ASVs], p < 292 0.001) and locality (envfit: $R^2 = 0.91$ [ZOTUs] - 0.92 [ASVs and OTUs], p < 0.001) was similar 293 294 to the abundance data (Fig.3). The results of rarefied data were similar with more overlap among sampling points (Appendix 1, Fig. S3). For traditional surveys, the points are clustered by 295 localities (envfit: $R^2 = 0.60$ for abundance and 0.87 for presence/absence data, p < 0.001), but not 296 by year (p >0.05, Fig. S4). 297

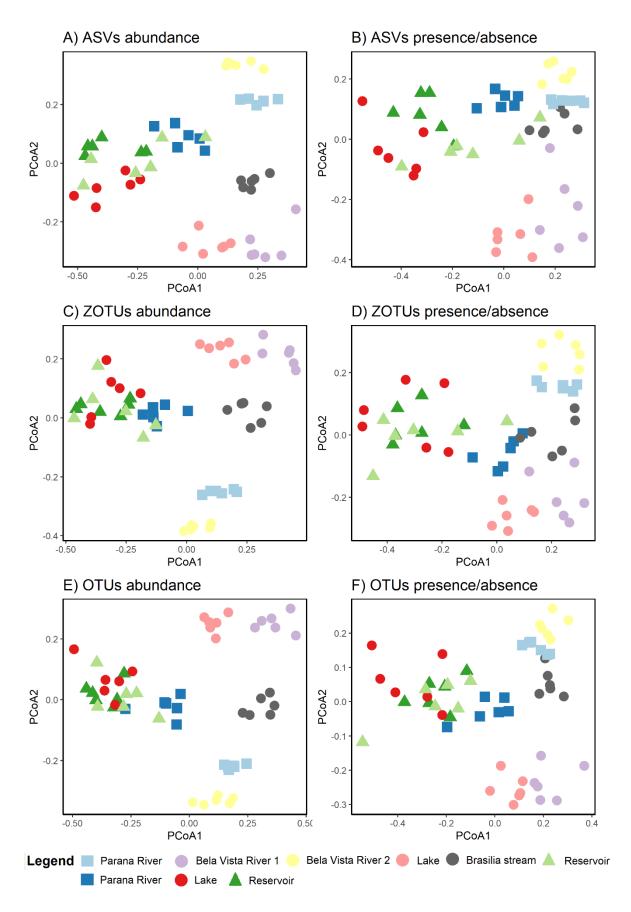


Figure 3. Principal Coordinates Analysis (PCoA) of fishes' communities from Itaipu based in 299 300 hill numbers for A) ASVs abundance, B) ASVs presence/absence, C) ZOTUs abundance, D) 301 ZOTUs presence/absence, E) OTUs abundance, and F) OTUs presence/absence. The axis 1 302 separated mainly the samples by year, while the axis 2 separated samples mainly by locality. 303 Each point is one of the replicates sampled. Colors and symbols represent collection points (mouth of channel at Paraná River = blue square, Itaipu's reservoir = green triangle, and 304 305 Piracema Channel = circles [Bela Vista River 1 = purple, Bela Vista 2 = yellow, Brasilia stream = gray, and lake = red]), and tone represent year of collection (light = 2019, dark = 2020). 306 307

308 Discussion:

309 Our results support mounting evidence that eDNA analysis provides a cost-effective alternative 310 to characterize fish biodiversity. We also demonstrate that different bioinformatic approaches 311 show similar results in terms of alpha and beta diversity, supporting the use of molecular approaches to monitor biodiversity even with incomplete taxonomic identification. However, a 312 serious caveat for using these molecular methods for biodiversity assessments is the scarcity of 313 comprehensive taxonomic reference databases, especially for the tropical regions of the globe. 314 315 Here, we also highlight these caveats for the Neotropical fish database, which are taxonomically limited, limiting the identification of several species. With a complete reference database, eDNA 316 317 could detect mostly fish community and also fish species that are poorly or non-represented by conventional methods, as suggested by our results. 318

319 We identified 35 species with ASVs, 37 with OTUs, and 34 with ZOTUs approaches at >97% similarity. However, many other ASVs, OTUs, and ZOTUs were identified at <97% similarity, 320 representing species not present in our database. We created our database based on the historical 321 taxonomic survey of Piracema Channel that may prevent identification of species that had not 322 been recorded by conventional fish survey methods. However, the use of a database without 323 curatorship can include spurious species identifications, such as species unlikely to be physically 324 present at sampling sites¹⁰. That occurs because when the database does not contain the sequence 325 326 of a certain species, the sequences will match with the closest species in that database, which can occur in a completely different environment (e.g. marine), beyond other factors that also 327 contribute to registering spurious species, such as misannotated sequences⁵⁷ or low variability in 328 the target sequenced region¹⁰ that will sign any species with such similar sequence. For instance, 329 our sequence for Prochilodus lineatus is identical to other Prochilodus species, such as P. harttii 330 and P. costatus. Furthermore, there are many species undescribed, making it impossible to 331 identify them. A recent compilation to list Paraná state fish species included 42 undescribed 332 species⁵, and this number may be underestimated due to the presence of crypt species and 333 334 sampling biases.

Even with the previously mentioned limitations, the use of molecular units such as amplicon sequence variants $(ASVs)^{58}$, operational taxonomic units $(OTUs)^{22}$, and zero-radius OTUs

(ZOTUs)²⁴ allows for assessing of genetic diversity and enables comparison among multiple 337 sites⁵⁹, space-time dynamics¹⁶ and evaluate natural and anthropogenic impacts⁶⁰. For instance, 338 vertebrate populations from freshwater ecosystems are declining at alarming rates (83% decline 339 since $(1970)^{61}$, and their conservation and management are a priority for global biodiversity⁶². 340 The Neotropical region harbors one of the largest freshwater biodiversity, with an estimated 341 9,000 described fish species (around 30% of total freshwater species)¹¹. The increasing 342 construction of dams is threatening fish populations over the entire planet $^{63-65}$ but specially in 343 Neotropical countries such as Brazil^{5,66,67}, and effective ways to monitor fish biodiversity to 344 345 understand its impact is essential.

As observed with the use of conventional ichthyofauna monitoring methods⁶⁸, the number of 346 species, ASVs, OTUs, ZOTUs, or 12S gene sequence readouts identified in our study showed a 347 348 variation between the two sampling occasions (2019 and 2020). Such variations in fish assemblages can be related to a series of factors, both biotic (ecological characteristics of the 349 350 species, for example) and abiotic (variations in water quality, and other environmental factors). In addition, physical characteristics of the environment such as total water volume and 351 352 hydrological characteristics can also play a key role in the ecology and occurrence of fish species⁶⁸. For instance, the recent extreme drought experienced in southeastern Brazil⁶⁹ may 353 have impacted fish assemblages. Our results showed a decrease of alpha diversity in 2020 in both 354 355 mouth of channel at Paraná River (blue squares) and the lake (red circles; Fig. 2). In addition to 356 the direct effects caused by this type of climatic phenomenon, such as the reduction in the 357 volume of water, indirect effects such as reduced oxygen concentration in the water and food availability can cause severe impacts on fish's communities^{68,70,71}. Such effects were more 358 evident at the mouth of channel at Paraná River, where the water level dropped 7 m from 2019 to 359 360 2020. At the reservoir, alpha diversity did not vary as water level fluctuation was less evident as a result of a stable environment due to the large size of this water body (green triangles; Fig. 2). 361 362 However, the traditional survey in Piracema Channel was unable to significantly detect the diversity variation throughout the period of the study (Fig. S2), highlighting the high sensibility 363 of eDNA metabarcoding for monitoring. 364

Among sampling points, the highest alpha diversity was recorded in those collected in mouth of channel at Paraná River, while the lowest alpha diversity was registered in the lake (Fig. 2).

Habitat heterogeneity is recognized as a main factor supporting functional and phylogenetic 367 diversity, which is often reflected in the taxonomic richness of the fish communities⁷². Mouth of 368 369 channel at Paraná River, the entrance of the Piracema Channel, is in a protected valley, where the 370 riparian vegetation is conserved, allowing the colonization by a diversified flora and fauna. 371 Besides this, the confluence with the Paraná River produces adjacent lotic and lentic 372 microhabitats, supporting a higher alpha diversity when compared to the main lake or the water 373 intake of the Channel, which are lentic and uniform environments. Such pattern of fish diversity / limnologic gradients meets the patterns previously assessed for the reservoir tributaries⁷³. 374

The beta diversity showed that in 2020, with the event of the extreme drought, a homogenization 375 of fish assemblage happened (Fig. 3 & Fig. 4). Both samples from the mouth of channel at 376 Paraná River (blue squares) and the lake (red circles) cluster together with the reservoir in both 377 378 years. The Itaipu's Reservoir was filled in 1982 and the Piracema Channel (a fish pass), connecting the region just downstream from Itaipu Dam to the Itaipu Reservoir, was opened 38 379 380 years later. Both events allowed the dispersion of species (including non-native species) in both directions promoting the homogenization of communities from upper and lower Paraná 381 River^{5,74,75}. Our results show the importance of the closest rivers and streams for system diversity 382 and resilience, as the mostly community variation was found in the Boa Vista River and Brasilia 383 384 Stream (Fig. 3).

385 Although eDNA metabarcoding is a powerful tool for biodiversity, as it has been widely used for 386 different purposes and different taxonomic groups, including identification and quantification of Neotropical ichthyofauna^{16,76,77}, many issues can hamper the metabarcoding results^{7,10,78,79}. Shaw 387 et al.¹⁰ drew attention to methodological considerations related to the eDNA sampling process 388 for freshwater fishes. According to them, the number of replicates is extremely important to 389 obtain accurate data. Specifically, they demonstrated that the collection of two eDNA replicates 390 per point were insufficient to detect less abundant taxa; however, adopting five replicates must 391 have a 100% detection rate. In addition, the eDNA sampling water column obtained is more 392 effective in detecting fish communities than sediment eDNA¹⁰. Here, we collected six replicates 393 394 per sampling point on the water surface. Furthermore, our rarefaction analysis clearly shows that many individual samples have a very low sequencing depth, but considering the replicates all our 395 396 sampled localities reach the asymptote (Fig. S1).

397 The bioinformatic methodological choices can also affect the metabarcoding results. Here, we used three pipelines that showed the best results compared with other approaches⁸⁰. We used 398 both OTU-level clustering at 97% level, with UPARSE⁴¹, and the unique sequences with zero-399 radius ZOTU-level denoising, with UNOISE3²⁴, and ASV-level Divisive Amplicon Denoising 400 Algorithm 2, with $DADA2^{23}$. Both the OTUs and the ZOTUs are created using in USEARCH⁸¹. 401 The initial steps as merging, filtering, and deduplicating are the same for both approaches, with 402 403 just the last step been different. The third approach generated ASVs through a parametric model, based in Q-scores to calculate a substitution model, estimating a probability for each possible 404 base substitution, to infer true biological sequences from reads as implemented in DADA 2^{23} . 405 Although we recorded some variation in the number of reads and "species" registered in each 406 407 pipeline, the results are very similar, highlighting their robustness.

Another potential bias in the results is data treatment. Here we used several data normalizations 408 for both alpha and beta diversity. Although historically more used, rarefied data is biased to 409 detect differentially abundant species⁵¹ and the hill numbers are considered the best approach for 410 metabarcoding data⁵⁴. Also, due to PCR biases, variation in the copy number of 12S genes per 411 cell/genome, as well as differences in size and biomass across the targeted organisms can 412 compromise a straightforward interpretation of OTU reads as an abundance measure⁸²⁻⁸⁴. 413 However, rare (low abundant) ASVs, ZOTUs and OTUs are more likely to be an artefact (both 414 erroneous sequence or because of cross-talk⁸⁵) and the true sequences are more stochastically 415 distributed due to the intrinsic low occurrence and detection probability^{86,87}. Therefore, analyses 416 that weight more the most abundant molecular units could be preferable. As each method has its 417 418 own biases, we present here both approaches.

Finally, it is important to highlight that, in general, molecular data derived from "environmental 419 sequencing" should be seen as complementary to, rather than as competing with, traditional 420 taxonomic studies. Indeed, a confluence of both lines of evidence is highly warranted, as it will 421 be necessary to overcome their respective shortcomings. For instance, we have shown here that 422 423 many species occurring in the Itaipu fish pass system have no genetic data to be identified. To improve the species detection with metabarcoding it is crucial to enhance the genetic reference 424 database through traditional inventories. Indeed, the metabarcoding approach is an intricate web 425 426 of feedback loops with the species taxonomy and ecology.

427 **Conclusion:**

428 Despite the clear indication that the reference databases need to be continuously fed with additional information on species that occur in the region, our results demonstrate the analytical 429 430 efficiency of the metabarcoding approach for monitoring fish species in the Itaipu's fish pass system. In addition, the methodology allowed, even when the specific identity of the ASVs, 431 OTUs, and ZOTUs were below 97% similarity with the species in our database, to carry out 432 estimates of species alpha and beta diversity. The use of such a methodology enables the 433 434 monitoring of the fish community with sufficient sensitivity to detect changes due to some natural or anthropogenic event. 435

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442 **References:**

- de Souza Dias, V., Pereira da Luz, M., Medero, G. M. & Tarley Ferreira Nascimento, D.
 An overview of hydropower reservoirs in Brazil: Current situation, future perspectives and
 impacts of climate change. *Water* 10, 592 (2018).
- Patias, J., Zuquette, L. V. & Rodrigues-Carvalho, J. A. Piezometric variations in the
 basaltic massif beneath the Itaipu hydroelectric plant (Brazil/Paraguay border): Right
 Buttress Dam. *Bull. Eng. Geol. Environ.* 74, 207–231 (2015).
- 449 3. Agostinho, A. A. Pesquisas, monitoramento e manejo da fauna aquática em
 450 empreendimentos hidrelétricos. in *SEMINÁRIO SOBRE FAUNA AQUÁTICA E O SETOR*451 *ELÉTRICO BRASILEIRO* 38–59 (Brasil, 1994).
- 4. Makrakis, S., Gomes, L. C., Makrakis, M. C., Fernandez, D. R. & Pavanelli, C. S. The
 Canal da Piracema at Itaipu Dam as a fish pass system. *Neotrop. Ichthyol.* 5, 185–195

454 (2007).

455	5.	Dos Reis, R. B., Frota, A., Depra, G. D. C., Ota, R. Rú. & Da Graca, W. J. Freshwater
456		fishes from Paraná State, Brazil: an annotated list, with comments on biogeographic
457		patterns, threats, and future perspectives. Zootaxa 4868, 451-494 (2020).
458	6.	Becker, R. A., Sales, N. G., Santos, G. M., Santos, G. B. & Carvalho, D. C. DNA
459		barcoding and morphological identification of neotropical ichthyoplankton from the
460		Upper Paraná and São Francisco. J. Fish Biol. 87, 159–168 (2015).
461	7.	Milan, D. T. et al. New 12S metabarcoding primers for enhanced Neotropical freshwater
462		fish biodiversity assessment. Sci. Rep. 10, 1-12 (2020).
463	8.	Agostinho, A. A., Pelicice, F. M. & Gomes, L. C. Dams and the fish fauna of the
464		Neotropical region: impacts and management related to diversity and fisheries. Brazilian
465		J. Biol. 68, 1119–1132 (2008).
466	9.	Bonar, S. A., Hubert, W. A. & Willis, D. W. Standard methods for sampling North
467		American freshwater fishes. (2009).
468	10.	Shaw, J. L. A. et al. Comparison of environmental DNA metabarcoding and conventional
469		fish survey methods in a river system. Biol. Conserv. 197, 131–138 (2016).
470	11.	Reis, R. E. et al. Fish biodiversity and conservation in South America. J. Fish Biol. 89,
471		12–47 (2016).
472	12.	Baumgartner, G. et al. Peixes do baixo rio Iguaçu. (Eduem, 2012).
473	13.	Taberlet, P., Bonin, A., Coissac, E. & Zinger, L. Environmental DNA: For biodiversity
474		research and monitoring. (Oxford University Press, 2018).
475	14.	Taberlet, P., Coissac, E., Pompanon, F., Christian, B. & Willerslev., E. Towards next-
476		generation biodiversity assessment using DNA metabarcoding. 33 , 2045–2050 (2012).
477	15.	Ritter, C. D. et al. The pitfalls of biodiversity proxies: Differences in richness patterns of
478		birds, trees and understudied diversity across Amazonia. Sci. Rep. 9, (2019).
479	16.	Sales, N. G. et al. Space-time dynamics in monitoring neotropical fish communities using
480		eDNA metabarcoding. Sci. Total Environ. 754, 142096 (2021).

481 17. Zinger, L. *et al.* Body size determines soil community assembly in a tropical forest. *Mol.*482 *Ecol.* 28, 528–543 (2019).

- 18. Baird, D. J. & Hajibabaei, M. Biomonitoring 2.0: a new paradigm in ecosystem
 assessment made possible by next □ generation DNA sequencing. *Mol. Ecol.* 21, 2039–
 2044 (2012).
- 486 19. Zinger, L. *et al.* Advances and prospects of environmental DNA in neotropical rainforests.
 487 *Adv. Ecol. Res.* 62, 331–373 (2020).
- Cilleros, K. *et al.* Unlocking biodiversity and conservation studies in high □ diversity
 environments using environmental DNA (eDNA): A test with Guianese freshwater fishes. *Mol. Ecol. Resour.* 19, 27–46 (2019).
- 491 21. Sales, N. G., Wangensteen, O. S., Carvalho, D. C. & Mariani, S. Influence of preservation
 492 methods, sample medium and sampling time on eDNA recovery in a neotropical river.
 493 *Environ. DNA* 1, (2019).
- Blaxter, M. *et al.* Defining operational taxonomic units using DNA barcode data. 1935–
 1943 (2005). doi:10.1098/rstb.2005.1725
- 496 23. Callahan, B. J. *et al.* DADA2: high-resolution sample inference from Illumina amplicon
 497 data. *Nat. Methods* 13, 581–583 (2016).
- 498 24. Edgar, R. C. UNOISE2: improved error-correction for Illumina 16S and ITS amplicon
 499 sequencing. *BioRxiv* 81257 (2016).
- 500 25. Muha, T. P., Rodriguez-Barreto, D., O'Rorke, R., Garcia de Leaniz, C. & Consuegra, S.
 501 Using eDNA metabarcoding to monitor changes in fish community composition after
 502 barrier removal. *Front. Ecol. Evol.* 9, 28 (2021).
- Kitano, T., Umetsu, K., Tian, W. & Osawa, M. Two universal primer sets for species
 identification among vertebrates. *Int. J. Legal Med.* 121, 423–427 (2007).
- 505 27. Stoeckle, M. Y., Soboleva, L. & Charlop-Powers, Z. Aquatic environmental DNA detects
 506 seasonal fish abundance and habitat preference in an urban estuary. *PLoS One* 12,
 507 e0175186 (2017).

508 509	28.	Bylemans, J. <i>et al.</i> An environmental DNA based method for monitoring spawning activity: A case study, using the endangered Macquarie perch (Macquaria australasica).
510 511 512 513	29.	 <i>Methods Ecol. Evol.</i> 8, 646–655 (2017). De Souza, L. S., Godwin, J. C., Renshaw, M. A. & Larson, E. Environmental DNA (eDNA) detection probability is influenced by seasonal activity of organisms. <i>PLoS One</i> 11, e0165273 (2016).
514 515	30.	Ritter, C. D. <i>et al.</i> Locality or habitat? Exploring predictors of biodiversity in Amazonia. <i>Ecography (Cop.).</i> 42 , 321–333 (2019).
516 517	31.	CFMV- Resolução n ^o 1000 de 11 de maio de 2012 - Dispõe sobre procedimentos e métodos de eutanásia em animais e dá outras providências. (2012).
518 519	32.	Britski, H. A., de Silimon, K. Z. de S. & Lopes, B. S. Peixes do Pantanal: manual de identificação, ampl. <i>Brasília, DF, Embrapa Informação Tecnológica</i> (2007).
520 521 522	33.	Ota, R. R., Deprá, G. de C., Graça, W. J. da & Pavanelli, C. S. Peixes da planície de inundação do alto rio Paraná e áreas adjacentes: revised, annotated and updated. <i>Neotrop. Ichthyol.</i> 16 , (2018).
523 524	34.	Neris, N., Villalba, F., Kamada, D. & Viré, S. Guía de peces del Paraguay/Guide of fishes of Paraguay. (2010).
525 526 527	35.	Pie, M. R. <i>et al.</i> Development of a real-time PCR assay for the detection of the golden mussel (Limnoperna fortunei, Mytilidae) in environmental samples. <i>An. Acad. Bras. Cienc.</i> 89 , 1041–1045 (2017).
528 529 530	36.	Miya, M. <i>et al.</i> MiFish, a set of universal PCR primers for metabarcoding environmental DNA from fishes: detection of more than 230 subtropical marine species. <i>R. Soc. open Sci.</i> 2, 150088 (2015).
531 532 533	37.	Boeger, W. A. <i>et al.</i> Testing a molecular protocol to monitor the presence of golden mussel larvae (Limnoperna fortunei) in plankton samples. <i>J. Plankton Res.</i> 29 , 1015–1019 (2007).
534	38.	Martin, M. Cutadapt removes adapter sequences from high-throughput sequencing reads.

- 535 *EMBnet. J.* **17**, 10–12 (2011).
- 39. Van Rossum, G. & Drake, F. L. Python 3 References Manual. Scotts Valley CA:
 CreateSpace. (2009).
- 40. R Core Team. R: the R project for statistical computing. 2019. URL: https://www. rproject. org/[accessed 2020-03-30] (2020).
- 540 41. Edgar, R. C. UPARSE: highly accurate OTU sequences from microbial amplicon reads.
 541 *Nat. Methods* 10, 996–998 (2013).
- Edgar, R. C. & Flyvbjerg, H. Error filtering, pair assembly and error correction for nextgeneration sequencing reads. *Bioinformatics* 31, 3476–3482 (2015).
- 43. Camacho, C. *et al.* BLAST+: architecture and applications. *BMC Bioinformatics* 10, 421
 (2009).
- 546 44. Team, Rs. RStudio: integrated development for R. *RStudio, Inc., Boston, MA URL*547 *http://www.rstudio. com* 42, 84 (2015).
- 548 45. Wickham, H. tidyverse: Easily Install and Load "Tidyverse" Packages (Version R
 549 package version 1.1. 1). (2017).
- 550 46. Wickham, H. ggplot2: elegant graphics for data analysis. (Springer, 2016).
- 47. Tang, Y., Horikoshi, M. & Li, W. ggfortify: unified interface to visualize statistical results
 of popular R packages. *R J.* 8, 474 (2016).
- 48. Auguie, B. & Antonov, A. gridExtra: Miscellaneous functions for "grid" graphics
 (Version 2.2. 1)[Computer software]. (2016).
- 49. Kassambara, A. & Kassambara, M. A. Package 'ggpubr'. (2020).
- 556 50. Oksanen, J. *et al.* Vegan: community ecology package. R package version 1.17-4.
 557 *http//cran. r-project. org>. Acesso em* 23, 2010 (2010).
- 558 51. McMurdie, P. J. & Holmes, S. Waste not, want not: why rarefying microbiome data is
 inadmissible. *PLoS Comput. Biol.* 10, e1003531 (2014).
- 560 52. Jost, L. Entropy and diversity. *Oikos* **113**, 363–375 (2006).

561	53.	Marcon.	E.,	Herault.	В.	&	Marcon.	M.	E.	Package	'entropart'.	(2021)	

- 562 54. Mächler, E., Walser, J.-C. & Altermatt, F. Decision making and best practices for
 563 taxonomy-free eDNA metabarcoding in biomonitoring using Hill numbers. *BioRxiv*564 (2020).
- 565 55. McMurdie, P. J. & Holmes, S. phyloseq: an R package for reproducible interactive 566 analysis and graphics of microbiome census data. *PLoS One* **8**, e61217 (2013).
- 567 56. Love, M. I., Huber, W. & Anders, S. Moderated estimation of fold change and dispersion
 568 for RNA-seq data with DESeq2. *Genome Biol.* 15, 550 (2014).
- 569 57. León, A., Reyes, J., Burriel, V. & Valverde, F. Data quality problems when integrating
 570 genomic information. in *International Conference on Conceptual Modeling* 173–182
 571 (Springer, 2016).
- 572 58. Callahan, B. J., McMurdie, P. J. & Holmes, S. P. Exact sequence variants should replace
 573 operational taxonomic units in marker-gene data analysis. *ISME J.* 11, 2639–2643 (2017).
- 574 59. Stahlhut, J. K. *et al.* DNA barcoding reveals diversity of hymenoptera and the dominance
 575 of parasitoids in a sub-arctic environment. *BMC Ecol.* 13, (2013).
- 60. Gillet, B. *et al.* Direct fishing and eDNA metabarcoding for biomonitoring during a 3-year
 survey significantly improves number of fish detected around a South East Asian
 reservoir. *PLoS One* 13, e0208592 (2018).
- 579 61. Barrett, M. et al. Living planet report 2018: Aiming higher. (2018).
- 580 62. Díaz, S. M. *et al.* The global assessment report on biodiversity and ecosystem services:
 581 Summary for policy makers. (2019).
- 582 63. Dudgeon, D. Asian river fishes in the Anthropocene: threats and conservation challenges
 583 in an era of rapid environmental change. *J. Fish Biol.* **79**, 1487–1524 (2011).
- 584 64. Dudgeon, D. Multiple threats imperil freshwater biodiversity in the Anthropocene. *Curr.*585 *Biol.* 29, R960–R967 (2019).
- 586 65. He, F. *et al.* Disappearing giants: a review of threats to freshwater megafauna. *Wiley*587 *Interdiscip. Rev. Water* 4, e1208 (2017).

588 589	66.	Agostinho, A. A., Thomaz, S. M. & Gomes, L. C. Threats for biodiversity in the floodplain of the Upper Paraná River: effects of hydrological regulation by dams. (2018).
590 591 592	67.	Santana, M. L., Carvalho, F. R. & Teresa, F. B. Broad and fine-scale threats on threatened Brazilian freshwater fish: variability across hydrographic regions and taxonomic groups. <i>Biota Neotrop.</i> 21 , (2021).
593 594	68.	Matthews, W. J. <i>Patterns in freshwater fish ecology</i> . (Springer Science & Business Media, 2012).
595 596	69.	de Oliveira Bueno, E., Alves, G. J. & Mello, C. R. Hydroelectricity water footprint in Parana hydrograph region, Brazil. <i>Renew. Energy</i> 162 , 596–612 (2020).
597 598 599 600	70.	Camacho Guerreiro, A. I., Amadio, S. A., Fabre, N. N. & da Silva Batista, V. Exploring the effect of strong hydrological droughts and floods on populational parameters of Semaprochilodus insignis (Actinopterygii: Prochilodontidae) from the Central Amazonia. <i>Environ. Dev. Sustain.</i> 23 , (2021).
601 602 603	71.	Jespersen, H., Rasmussen, G. & Pedersen, S. Severity of summer drought as predictor for smolt recruitment in migratory brown trout (Salmo trutta). <i>Ecol. Freshw. Fish</i> 30 , 115–124 (2021).
604 605 606	72.	Pool, T. K., Grenouillet, G. & Villéger, S. Species contribute differently to the taxonomic, functional, and phylogenetic alpha and beta diversity of freshwater fish communities. <i>Divers. Distrib.</i> 20 , 1235–1244 (2014).
607 608	73.	Oliveira, E. F. de, Goulart, E. & Minte-Vera, C. V. Fish diversity along spatial gradients in the Itaipu Reservoir, Paraná, Brazil. <i>Brazilian J. Biol.</i> 64 , 447–458 (2004).
609 610 611	74.	Daga, V. S. <i>et al.</i> Homogenization dynamics of the fish assemblages in Neotropical reservoirs: comparing the roles of introduced species and their vectors. <i>Hydrobiologia</i> 746 , 327–347 (2015).
612 613 614	75.	Vitule, J. R. S. Introdução de peixes em ecossistemas continentais brasileiros: revisão, comentários e sugestões de ações contra o inimigo quase invisível. <i>Neotrop. Biol. Conserv.</i> 4 , 111–122 (2009).

615 616	76.	Mariac, C. <i>et al.</i> Species level ichthyoplankton dynamics for 97 fishes in two major river basins of the Amazon using quantitative metabarcoding. <i>Mol. Ecol.</i> (2021).
617 618	77.	Jackman, J. M. <i>et al.</i> eDNA in a bottleneck: obstacles to fish metabarcoding studies in megadiverse freshwater systems. <i>Environ. DNA</i> (2021).
619 620	78.	Bessey, C. <i>et al.</i> Maximizing fish detection with eDNA metabarcoding. <i>Environ. DNA</i> 2 , 493–504 (2020).
621 622 623	79.	Evans, N. T. <i>et al.</i> Fish community assessment with eDNA metabarcoding: effects of sampling design and bioinformatic filtering. <i>Can. J. Fish. Aquat. Sci.</i> 74 , 1362–1374 (2017).
624 625	80.	Prodan, A. <i>et al.</i> Comparing bioinformatic pipelines for microbial 16S rRNA amplicon sequencing. <i>PLoS One</i> 15 , e0227434 (2020).
626 627	81.	Edgar, R. C. Search and clustering orders of magnitude faster than BLAST. <i>Bioinformatics</i> 26 , 2460–2461 (2010).
628 629 630	82.	Elbrecht, V. & Leese, F. Can DNA-based ecosystem assessments quantify species abundance? Testing primer bias and biomass—sequence relationships with an innovative metabarcoding protocol. <i>PLoS One</i> 10 , e0130324 (2015).
631 632 633	83.	Pawluczyk, M. <i>et al.</i> Quantitative evaluation of bias in PCR amplification and next- generation sequencing derived from metabarcoding samples. <i>Anal. Bioanal. Chem.</i> 407 , 1841–1848 (2015).
634 635	84.	Holman, L. E., Chng, Y. & Rius, M. How does eDNA decay affect metabarcoding experiments? <i>Environ. DNA</i> (2021).
636 637	85.	Edgar, R. C. UNCROSS2: identification of cross-talk in 16S rRNA OTU tables. <i>BioRxiv</i> 400762 (2018).
638 639	86.	MacArthur, R. H. <i>Geographical ecology: patterns in the distribution of species</i> . (Princeton University Press, 1984).
640 641	87.	Leray, M. & Knowlton, N. Random sampling causes the low reproducibility of rare eukaryotic OTUs in Illumina COI metabarcoding. <i>PeerJ</i> 5 , e3006 (2017).

- 642 88. Team, Q. D. QGIS geographic information system. Open Source Geospatial Found. Proj.
- 643 *Versão* **2**, (2015).