



Different approaches to processing environmental DNA

samples in turbid waters have distinct effects for fish, bacterial and archaea communities.

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11 Abstract

12 Coastal lagoons are an important habitat for endemic and threatened species in California that
13 have suffered impacts from urbanization and increased drought. Environmental DNA has been
14 promoted as a way to aid in the monitoring of biological communities, but much remains to be
15 understood on the biases introduced by different protocols meant to overcome challenges
16 presented by unique systems under study. Turbid water is one methodologic challenge to eDNA
17 recovery in these systems as it quickly clogs filters, preventing timely processing of samples. We
18 investigated biases in community composition produced by two solutions to overcome slow
19 filtration due to turbidity: freezing of water prior to filtration (for storage purposes and long-term
20 processing), and use of sediment (as opposed to water samples). Bias assessments of community
21 composition in downstream eDNA analysis was conducted for two sets of primers, 12S (fish)
22 and 16S (bacteria and archaea). Our results show that freezing water prior to filtration had

23 different effects on community composition for each primer, especially for the 16S, when using a
24 filter of larger pore size (3 µm). Nevertheless, pre-freezing water samples can still be a viable
25 alternative for storage and processing of turbid water samples when focusing on fish
26 communities (12S). The use of sediment samples as an alternative to processing water samples
27 should be done with caution, and at minimum the number of biological replicates and/or volume
28 sampled should be increased.

29 Introduction

30 Coastal lagoons in California are the numerically dominant form of coastal wetland (Jacobs et
31 al., 2011; Stein et al., 2014) and are important in many other Mediterranean climates and
32 subtropical environments. These lagoons are characterized by seasonal and episodic breaching
33 (opening of the lagoon to the sea, usually by stream flow) and closure (isolation of the lagoon by
34 a high sandbar), which provide a suite of ecological services: from groundwater infiltration to
35 support of unique biodiversity (Ballard et al., n.d.). This system serves as important habitat and
36 nursery for endemic and endangered fishes and amphibians, such as the steelhead
37 (*Oncorhynchus mykiss*), red-legged frog (*Rana aurora draytonii*), and the tidewater goby
38 (*Eucyclogobius newberryi*) (Earl et al., 2010; Shaffer et al., 2004; Swift et al., 1993, 2016). Thus,
39 California lagoons are spatially and temporally variable systems with unique biodiversity and
40 biodiversity assessment challenges.

41 Coastal lagoons have been drastically reduced in numbers along the California coastline,
42 driven mostly by the impact of coastal land use for transport structures, agriculture, and
43 development. These are further exacerbated by ongoing changes in the hydrological cycles due
44 to climate change (SCWRP, 2018). While these sites are critical for endangered species
45 conservation, they are also subject to frequent invasion and their response to environmental

46 variation is poorly documented. However, monitoring of this habitat can be limited by a variety
47 of issues, ranging from limited human power and access to challenges driven by the natural
48 complexity and dynamism of these lagoons.

49 The use of environmental DNA (eDNA) has been advocated as an alternative for
50 monitoring communities and target species (Thomsen & Willerslev, 2015), and can overcome
51 and complement certain field limitations from traditional methods (e.g. seining, trapping). On-
52 site collection can be relatively fast, and therefore allow field workers to cover more ground. It
53 can also recover the DNA signal of species that are rare, cryptic and/or hard to capture by
54 traditional methods, and being non-intrusive, it offers an alternative when working with
55 endangered species for which permits are necessary (Deiner et al., 2017; Dejean et al., 2012;
56 Sard et al., 2019). In addition, metabarcoding approaches allow the investigation of multiple
57 species from a single collection (Taberlet, Coissac, et al., 2012).

58 Nevertheless, it is important to recognize that this approach also brings its own
59 limitations and biases (van der Loos & Nijland, 2021). In some circumstances, eDNA sampling
60 can be more expensive than traditional, more established methods (Smart et al., 2016). Since
61 there are no voucher specimens from collections, contamination is a major issue that needs to be
62 addressed early on, following best practices in the field (Goldberg et al., 2016). The lack of
63 voucher specimens also leads to an overdependence on the use of barcodes and genetic databases
64 for taxonomic identification, which introduces another set of biases, from misidentification to
65 lack of species representation (Taberlet, Coissac, et al., 2012). Other challenges arise from the
66 non-universality of sampling methods and downstream processing, with the probability of
67 detection varying depending on the species and their density, as well as the type of environment,

68 which affects rates of DNA degradation (Deiner et al., 2015; Nagarajan et al., 2022; Rees et al.,
69 2014; Williams et al., 2017).

70 Coastal lagoons can vary drastically in their environmental properties. One major
71 challenge is the high and variable turbidity of the water. High turbidity often occurs during high
72 stream flow, or when lagoons are closed to the ocean by a sandbar leading to an accumulation of
73 organic and inorganic matter. In this scenario, filtering turbid water on-site is problematic.
74 However, filtration is widely used for handling water samples (Laramie et al., 2015; Tsuji et al.,
75 2019). Typically, set volumes of water are run through a small pore size filter to concentrate
76 DNA before extractions. However, high concentration of fine sediment or organic matter in
77 water quickly obstructs these filters, making the filtration process time-consuming (although it
78 could actually aid recovery by binding DNA to suspended particles: Kumar et al., 2022; Liang &
79 Keeley, 2013; Torti et al., 2015).

80 To overcome this issue, some stakeholders have relied on a tiered filtration step
81 (prefiltration) to reduce particles and avoid clogging filters (Tsuji et al., 2019), but this approach
82 increases costs, labor and opportunities for potential contamination (Li et al., 2018; Majaneva et
83 al., 2018; Robson et al., 2016). The use of filters with bigger pore sizes (up to 20 μm) has been
84 previously tested and in cases of turbid waters is generally preferred, but requires filtering larger
85 volumes of water to capture the same amount of DNA recovered in smaller pore size filters
86 (Robson et al., 2016; Turner et al., 2014).

87 Freezing water for storage purposes prior to filtration can mitigate the issue of slow
88 filtration in the field and allow it to be done in batches in the laboratory at a later time, but this
89 type of sample storage could also introduce bias on eDNA capture and community composition
90 (Kwambana et al., 2011; Sekar et al., 2009). Freezing can disrupt cells and extrude their DNA in

91 the environment, which would then make it easier for it to pass through the filter pores and be
92 lost, an issue that has been demonstrated in certain cases (e.g. Suomalainen et al., 2006).
93 Increasing the pore size of filters to speed the filtration process could worsen this problem by
94 letting DNA in solution flow through the pores more easily.

95 When dealing with turbid waters, some stakeholders have opted to use the centrifugation
96 approach (e.g. Williams et al., 2017). Extracellular DNA (i.e. DNA not contained within a cell
97 wall) can be bound to particles (Torti et al., 2015) and consequently be captured and detected
98 following centrifugation of particles into pellets. However, the amount of water used is limited
99 by centrifuge size, usually around 15-30 mL per replicate (Doi et al., 2017; Ficetola et al., 2008),
100 which might limit recovery of diluted DNA (Deiner et al., 2015).

101 Processing sediment samples may be preferable to processing highly turbid water
102 samples. However, it is important to understand how DNA recovery from these different media
103 compare to one another. Turner et al. (2015) and Perkins et al. (2014) have shown that sediment
104 can have a higher concentration of fish eDNA and some bacteria, respectively. This may relate to
105 the organic-particle binding and sinking properties, and longer persistence of DNA in sediment
106 compared to water samples. However, as is the case with water samples, there is no consensus on
107 the rate of degradation of eDNA in soil and sediment (Dell'Anno & Corinaldesi, 2004; Levy-
108 Booth et al., 2007; Torti et al., 2015), and this will depend on multiple local biotic and abiotic
109 factors. In addition, biological communities can naturally differ between water column and
110 sediments, even when we expect some level of overlap due to both DNA sinking and suspension.

111 Previous works have compared different approaches to processing eDNA, including
112 assessment of filtration and storage methods (Hinlo et al., 2017; Takahara et al., 2015),
113 comparisons between water and sediment eDNA recovery (Sales et al., 2019; Turner et al.,

114 2015), as well as some work on turbid waters (Kumar et al., 2022; Robson et al., 2016; Williams
115 et al., 2017). But results have been contradictory, or limited to looking at just DNA
116 concentration, or at a single targeted species.

117 The goal of the present study is to compare how freezing water prior to filtration and
118 using water versus sediment samples induce and/or exacerbate biases in taxa detection for a set
119 of universal primers targeting different biological communities—12S (fish) and 16S (bacteria and
120 archaea)—in coastal lagoons. By understanding the biases introduced when processing
121 environmental samples, we will be able to inform decisions regarding experimental design for
122 monitoring such a dynamic and challenging habitat, which has invaluable importance for the
123 maintenance of ecosystem services for both wild and urban populations. We expect these results
124 will be of interest relative to eDNA sampling in other aquatic systems as well, such as rivers,
125 streams, and ponds, especially those with turbid waters.

126 Material and Methods

127 Site - Topanga Lagoon

128 To determine the variability of species detection for each protocol, water and sediment samples
129 were collected from a south-facing coastal lagoon in southern California, located in Malibu, a
130 stretch of coast that runs from Santa Monica to Point Mugu. This lagoon is part of the Topanga
131 State Park and is currently undergoing plannings for restoration. It is the only lagoon on this
132 stretch of coast that still harbors a stable population of tidewater goby (*E. newberryi*), a federally
133 endangered species, and is relatively less impacted than other lagoons in the same region. The
134 endangered southern steelhead trout (*O. mykiss*) is also found in this system during anadromy
135 when the lagoon is breached. Due to the presence of these species, Topanga lagoon has been
136 periodically surveyed by the Jacobs' lab members and collaborators such as researchers at the

137 Resource Conservation District of The Santa Monica Mountains (RCDSMM), and therefore its
138 macrobiota is regularly studied, especially the fish fauna. The lagoon was sampled on September
139 6th, 2018, at the end of the Summer season, and as is typical of this time of the year, the weather
140 was dry with no record of precipitation since June (WeatherSpark.com, n.d.). The lagoon was
141 closed to the ocean by a sandbar and the water was murky (Fig. 1), which in the author's
142 experience, such turbidity slowed filtration and easily clogged 0.45 μm cellulose nitrate filters.

143 Protocols and samples

144 A sterilized water jug was used to collect a single water sample in the lagoon, at a mid-point
145 between the mouth margin and the road bridge (Fig. 1). The sample was then placed on ice and
146 brought to the laboratory (~1 hr car ride). This method of “grab-and-hold” has proven to be
147 similarly effective as on-site filtration in a previous study (Pilliod et al., 2013). Once in the
148 laboratory, the total volume was divided for two separate protocols: centrifugation; and filtration.

149 The centrifugation protocol followed Doi et al. (2017) using five replicates of 50 mL
150 tubes (with 27 mL water samples each). Besides extracting the pellet, we also included a
151 filtration step of the supernatant using a 0.45 μm filter. For the filtration protocol, water was
152 separated into ten 500 mL bottles (Fig. 2) and these were used in two separate protocols. Half
153 (i.e. five bottles) were frozen in the -20°C for three days before thawing for filtration (hereafter
154 referred to ‘pre-freezing (PF) protocol’), and half was not frozen and immediately filtered
155 (hereafter referred to ‘no freezing (NF) protocol’). Filtration was done in two sequential steps for
156 both treatments (pre- and no freezing) using an adapted vacuum pump in the pre-PCR room of
157 the laboratory (Fig. S1). First, the water sample was filtered using a 3 μm filter, then the filtrate
158 was passed through a 0.45 μm filter. All filters used in this work were cellulose nitrate. Here,

159 however, we will focus only on the results from the first filtration step of the water filtration
160 protocol (3 μm filters). More details on that are further explained in the supplemental material.

161 Surficial sediment (within the first 5 cm) was also collected at the same location where
162 water was sampled (hereafter referred to ‘sediment (Sed) protocol’). We used five collection kits,
163 and each kit consisted of three 2 mL tubes (15 tubes total). Subsamples of these were pooled in
164 triplicates prior to DNA extraction following instructions as defined by the CALeDNA program
165 (<https://ucedna.com/methods-for-researchers>). These were also kept on ice during field work and
166 stored in a -80°C freezer upon arrival at the laboratory until DNA extractions.

167 In order to test if freezing water is a viable process to store and manage water samples,
168 we compared the results of pre-freezing water prior to filtration with the no freezing protocol.
169 Results from the sediment protocol were compared against both filtration protocols: (pre-
170 freezing versus no freezing) to test if sediments can be used as an alternative to water samples.

171 DNA Extraction

172 DNA from sediments and filters were extracted following the PowerSoil extraction protocol.
173 Filters were chopped into thin strips before being added to the bead tubes, and sediment
174 triplicates were pooled in small batches to reach 0.25-0.3 g before processing. We used the soil
175 extraction kit on the filters as well to reduce potential PCR inhibition caused by the water
176 turbidity (Kumar et al., 2022), but also to limit the number of variables in the research design by
177 adding another extraction protocol.

178 Contamination best practices

179 Care was taken to avoid contamination both in the field and the lab. Before collection, bottles
180 and the water jug were cleaned and bleached and then handled with clean gloves on site.
181 Extractions and PCR were done in a separate pre-PCR room. Utensils and bench top were

182 cleaned with 10% bleach, followed by 70% ethanol. Forceps and scissors for handling filters
183 were seared and cleaned with bleach and ethanol after dealing with each sample. PCR reagents
184 were prepared in a clean, PCR-free, positive pressure hood. Sediment samples were collected
185 with new 2 mL cryotubes and following field protocol as recommended by the CALeDNA
186 program. Blanks were made for the field collection, laboratory filtration and PCR (5 blanks in
187 total) and included in the library for sequencing.

188 Sequencing

189 Library preparation followed CALeDNA protocols ([https://ucedna.com/methods-for-](https://ucedna.com/methods-for-researchers)
190 [researchers](https://ucedna.com/methods-for-researchers)). Metabarcoding libraries were generated for bacteria and archaea (16S rRNA), fish
191 (12S rRNA) and metazoans (CO1). Sequences for each primer can be found at Table 1. All
192 libraries consisted of triplicate PCR reactions. PCR products were visualized using gel
193 electrophoresis, and for each barcode, PCR triplicates were pooled by sample. After bead
194 cleaning, all markers were pooled by sample and tagged for sequencing (single indexing).
195 Libraries were pooled and run on a MiSeq SBS Sequencing v3 in a pair-end 2x300 bp format
196 [Technology Center for Genomics & Bioinformatics (TCGB), UCLA] with a target sequencing
197 depth of 25,000 reads/sample/metabarcoding. Two sequencing runs were conducted, but the CO1
198 primer was still below the sequencing depth threshold and therefore its results will not be
199 discussed here (see Figs. S2-3). For each run, our library was pooled with different samples from
200 different collaborators to maximize efficiency of the sequencing run.

201 Bioinformatics and data pre-processing

202 Sequence data was bioinformatically processed in Hoffman2, the High Performance Computing
203 cluster at UC Los Angeles, using the Anacapa Toolkit (Curd, Gomer, et al., 2018) with default
204 settings. Briefly, reads are demultiplexed and trimmed for adapters (cutadapt, Martin, 2013) and

205 low-quality reads (FastX Toolkit, *FASTX-Toolkit*, n.d.). Dada2 (Callahan et al., 2016) is used to
206 denoise, dereplicate, merge and remove chimeras, and the resulting clean Amplicon Sequence
207 Variants (ASVs) have their taxonomy assigned using Bowtie2 (Langmead & Salzberg, 2012),
208 matched to a custom reference library (CRUX, Curd, Kandlikar, et al., 2018). Confidence levels
209 are determined by the BLCA algorithm (Gao et al., 2017) to generate a table of best taxonomic
210 hits, from super-kingdom to species level. The pipeline was designed to process not only paired,
211 but also unmerged and unpaired reads.

212 Taxonomic tables with a bootstrap confidence cutoff score of 0.6 were used for
213 downstream analyses. Except when noted, all bioinformatic analyses mentioned beyond this
214 point were performed using R v.3.6.2 (R Core Team, 2018) in RStudio v.1.2.1335 (RStudio
215 Team, 2020). Decontamination was done separately for each primer set and each run (since the
216 dataset was pooled with different combinations of samples for sequencing). We used the package
217 metabar (Zinger et al., 2020) to lower tag-jumping and remove contaminants through detection
218 of ASVs whose relative abundance is highest in negative controls. We also ran a modification of
219 the gruinard pipeline (https://github.com/zjgold/gruinard_decon), including only steps 4 (site
220 occupancy modeling) and 5 (dissimilarity between replicates), since previous steps were
221 redundant with the metabar decontamination steps. Lastly, taxa classified as "Not_found",
222 "Unclassified", "*Canis lupus*", "*Bos taurus*", and "*Homo sapiens*" were removed from the final
223 tables before being merged and used in downstream analyses.

224 Diversity analysis

225 We used the laboratory's own sampling record and the Global Biodiversity Information Facility
226 database (Gbif.Org, 2022) to manually check the 12S primer final taxonomic table. The number
227 of species captured by each protocol was visualized using Venn Diagrams (package

228 VennDiagram, Chen, 2018). Species rarefaction curves were made for each metabarcode to
229 inspect the level of species saturation in each protocol replicate. The slope of each curve was
230 calculated using the rareslope function in the vegan package (Oksanen et al., 2019), and the
231 confidence interval was calculated using pairwiseCI (Schaarschmidt & Gerhard, 2019) with
232 confidence level at 95%. Rarefaction curves were plotted using the ggrare function from the
233 ranacapa package (using step = 5).

234 Differential abundance

235 The raw dataset was analyzed using DESeq2 and ALDEx2 to look at differential abundance
236 between protocols (Fernandes et al., 2013; Love et al., 2014), and the output of both analyses
237 were compared to look for consistency in results. While DESeq2 uses a series of negative-
238 binomial generalized linear models to count data and estimate the log₂-fold change in
239 abundance, ALDEx2 (ANOVA-like Differential Expression) accounts for community
240 composition when calculating differential expression by performing a scale invariant
241 transformation on read counts, which are modeled as distributions of posterior probabilities
242 sampled from a Dirichlet distribution. This method has been found to produce more consistent
243 and reproducible results (Nearing et al., 2022).

244 For DESeq2, the default testing framework was used (test = “Wald”, fitType =
245 “parametric”), which includes the Benjamini-Hochberg multiple inference correction. The
246 sfType option was defined as poscounts since this estimator is able to handle zeros. The log₂
247 fold change of each pairwise comparison for which there were significant differences in
248 abundances was plotted. For ALDEx2, we performed a two sample T-test for each pairwise
249 comparison: between the different filtration protocols (pre- versus no freezing), and between
250 each filtration protocol and sediment protocol (i.e. PF vs. Sed; and NF vs. Sed). We used 500

251 Monte Carlo samples (`mc.samples = 500`) for estimation of the posterior distribution. We plotted
252 both the between- (M) versus within-condition fold change (W) (MW) and Bland-Altman (BA)
253 plots using the `aldex.plot` function.

254 Beta diversity

255 For the beta diversity analysis, samples were standardized by using either the eDNA index
256 (Kelly et al., 2019) or by rarefying them as a way to equalize sequencing effort and minimize
257 stochasticity and bias. For the eDNA index, we followed the Wisconsin double standardization
258 method in the `vegan` package. The `custom_rarefaction` function in the R package `ranacapa`
259 (Kandlikar, 2020) was used to rarefy the dataset with 10 replicates.

260 For the 12S primer, samples were rarefied to 20 000 reads. Three sediment replicates
261 were excluded due to very low read numbers (<100). For the 16S, samples were rarefied to
262 15 000 and one sediment replicate that had ~5000 reads was excluded. The number of reads per
263 taxa for each protocol replicate was plotted using the `phyloseq` package (McMurdie & Holmes,
264 2013), for both the raw and rarefied dataset.

265 The rarefied dataset followed a Constrained Analysis of Principal Coordinates (CAP)
266 using the `capscale` function in `vegan` and Bray-Curtis distance. This ordination method, which
267 can be used with non-Euclidean dissimilarity indices, explains the ordination of assemblage
268 composition based on species abundances. The difference in community composition for each
269 treatment was then analyzed using a PERMANOVA and Bray-Curtis dissimilarity, followed by a
270 pairwise PERMANOVA comparison (all with the `vegan` package). P-values were adjusted using
271 the FDR (False Discovery Rate) approach.

272 Results

273 Sequencing

274 The first run generated a total of 6 407 371 reads: 3 817 216 reads for the 12S primer, 2 393 627
275 for 16S, and 196 528 for CO1. In the second run there were a total of 9 088 496 reads: 6 685 673
276 reads for the 12S metabarcode, 1 904 283 reads for 16S and 498 540 for the CO1. For the 12S
277 and 16S primers, we were able to reach our threshold of 25 000 reads/sample in most cases,
278 while that was not the case for all except one sample of the CO1 primer. Because of this
279 limitation on the number of reads/sample, the CO1 metabarcode will not be discussed further in
280 the main paper (but check the supplemental material for more details).

281 Bioinformatics and data pre-processing

282 The number of reads per sample after decontamination and combining both runs is illustrated in
283 Figure S3. We manually checked the final taxonomic tables of each separate run for the 12S
284 primer to look for signs of contamination and evaluate how well the bioinformatic
285 decontamination steps worked (metabaR and gruinard). The taxonomic tables for the 12S primer
286 have substantially less species than the 16S, and the local fish fauna is relatively well known,
287 making the process more tractable.

288 For the run that was pooled with samples from Palmyra Atoll, the output still retained
289 some tropical reef and pelagic fish and elasmobranch species that are not found in coastal
290 lagoons in California. We can expect that tag-jumping contamination is also present in the other
291 sequencing runs and primers as well. Interestingly, eight out of 28 of those tropical species (ca.
292 28%) were found exclusively on the sediment protocol and not the filtration protocols (e.g.
293 *Acanthurus achilles*, *Scarus altipinnis*, *Lutjanus russellii*).

294 Barplots for both the raw and rarefied dataset (Figs. S3-4, respectively) show that
295 sediment replicates had greater variability amongst themselves, both in number of reads and
296 community composition, compared to the replicates of either filtration protocols. Water
297 replicates were more consistent within and between protocols, and had an overall higher number
298 of reads than the sediment replicates.

299 Diversity

300 After the decontamination steps (metabaR and gruinard) and removing specific, uninformative
301 ASVs (as listed above), the total number of species assigned to 12S was 39, distributed in 20
302 orders and 22 families. Of these 39 species, only four had been previously recorded for the site
303 (Table S1). For 16S, the total number of taxa assigned to species was 2 625, distributed in 45
304 phyla and 335 families.

305 We also noticed some dubious taxonomic assignments. For example, for the 12S primer,
306 we had one hit for *Fundulus diaphanus*, which is a species of killifish native to the northeast of
307 North America. However, the californian species *F. parvipinnis* has been previously documented
308 in Topanga by lab members sampling at the site. Similarly, there were two hits for *Phoxinus*
309 *phoxinus*, which has a European distribution with a closely related North American counterpart,
310 *P. eos*, although this species has not been identified in collections from Topanga lagoon. Another
311 dubious identification occurred for two species of *Odontesthes*, *O. incisa* and *O. smitti*, which
312 were among the most abundant hits in our dataset but are native to the southwest Atlantic. These
313 two species, however, are South American relatives of topsmelt (*Atherinops affinis*), commonly
314 found in coastal lagoons and estuaries in California (Table S1).

315 The Venn Diagram (Fig. 3) shows that even though the sediment protocol had lower
316 numbers of reads overall (Figs. S2-3), they had the highest number of species recovered (12S

317 primer: N=27, 19 unique; 16S primer: N=1 929, 1 178 unique). The species overlap between
318 protocols for the 12S was only 1.2% (n=1), and for the 16S primer it was 3.5% (n=402).

319 Species rarefaction curves also show that the sediment protocol is further from reaching
320 saturation compared to both filtration protocols, for both the 12S and 16S primers (Fig. 4),
321 although there was more variation between the replicates for the 12S sediment protocol. For the
322 12S primer, there is a significant difference in the slope of the species curves between the
323 sediment and no freezing protocols (Fig. 5, Table 2), while for the 16S, all pairwise comparisons
324 showed significant differences.

325 Differential Abundance

326 In the DESeq2 analysis, there was no significant difference between species abundance for the
327 12S primer in any of the protocols' pairwise comparisons. In ALDEx2, the tidewater goby (*E.*
328 *newberryi*) was the only species with an effect size > 1 in the comparisons of filtration protocols
329 (NF and PF) against the sediment protocol, being underrepresented in the latter (Fig. 6). The
330 proportion of overlap of the 95% CI of the effect size was not zero, but they were small
331 nonetheless (NF: 0.010; PF: 0.014) (Table S4).

332 In the case of the 16S primer, the DESeq2 analysis showed no significant difference in
333 comparison between the filtration protocols (NF versus PF). However, there were significant
334 differences in the pairwise comparisons between filtration protocols and sediment protocol (NF
335 vs. Sed; PF vs. Sed; Fig. 6, Tables S2-3). The five most differentially abundant species with
336 highest abundance in the filtration protocols, relative to the sediment, were representatives of the
337 families Aphanizomenonaceae, Comamonadaceae and Flavobacteriaceae (in both NF and PF
338 comparisons); and Hemiselmidaceae and Geminigeraceae (PF protocol only). These comprise

339 groups of cyanobacteria (Aphanizomenonaceae) and algae (Hemiselmidaceae and
340 Geminigeraceae), as well as environmental bacteria (Comamonadaceae and Flavobacteriaceae).

341 The most differentially abundant species with highest abundance in the sediments were
342 representatives of the families Catenulaceae, Fragilariaceae and an archaea assigned to the
343 Thaumarchaeota phylum (in both NF and PF comparisons); plus Woeseiaceae and Elphidiidae
344 (NF protocol only); and Anaerolineaceae and Desulfobacteraceae (PF protocol only). These
345 comprise groups of diatoms (Catenulaceae and Fragilariaceae), environmental bacteria
346 (Woeseiaceae, Anaerolineaceae and Desulfobacteraceae) and archaea (Thaumarchaeota), and
347 foraminiferans (Elphidiidae).

348 In the ALDEx2 analysis, there were many species that showed an effect size > 1 (Fig. 8),
349 but only a few showed proportions of overlap < 0.001 (none in the NF versus PF comparison)
350 (Table S5). For the species that fall within these criteria (effect > 1 , overlap < 0.001), in the
351 comparisons between filtration protocols and sediment protocol (NF vs. Sed; PF vs. Sed), there
352 were 61 species overrepresented in the NF protocol compared to 39 in the sediment; and 73
353 species overrepresented in the PF protocol compared to the sediment (there was no species
354 overrepresented in the sediment that met the criteria above) (Table S5). Taxa overrepresented in
355 both filtration protocols (NF and PF) relative to sediment are in the families Comamonadaceae,
356 Flavobacteriaceae and Rhodobacteraceae. Both Comamonadaceae and Flavobacteriaceae were
357 also represented in the DESeq2 results and comprise groups of environmental bacteria.
358 Rhodobacteraceae are heterotrophic bacteria usually found in association with algae in marine
359 environments (Bischoff et al., 2021). In the NF protocol, Cryomorphaceae is the most
360 represented family, with seven species, and comprises aquatic bacteria generally found in
361 locations rich in organic carbon (Bowman, 2014).

362 In the sediments, when compared to the NF protocol, the most overrepresented family is
363 the Anaerolineaceae with 12 species, followed by Desulfobacteraceae (5), Nonionidae (2) and
364 Elphidiidae (2). Anaerolineaceae, Desulfobacteraceae and Elphidiidae were also found in the
365 DESeq2 results. The first two comprise environmental bacteria, while the latter is a family of
366 foraminiferans (such as the Noniodidae family).

367 Beta diversity

368 For the rarefied dataset, there were no discernible differences in assemblage composition in the
369 CAP analysis for the 12S primer (Fig.9A). One sediment replicate is driving most of the
370 difference (CAP1=86%) with the overrepresentation of many tropical species, likely tag-jump
371 contaminants. The PERMANOVA result showed a relatively weak effect overall ($R^2 = 0.38$; $p =$
372 0.05), with the pairwise test showing a relatively weak effect for the NF versus PF comparison
373 ($R^2 = 0.20$), followed by stronger effects in the sediment comparisons (NF vs. Sed: $R^2 = 0.41$; PF
374 vs. Sed: $R^2 = 0.31$) (Table 3). The lack of significant differences between the filtration and
375 sediment protocols could have been driven by the loss of three sediment replicates when
376 rarefying the dataset.

377 When using the eDNA index, the CAP analysis for the 12S primer showed that many of
378 the species driving the differences in assemblage composition were the tropical species that are
379 coming from the tag-jumping contamination (Fig. S5A). For example, we see overrepresentation
380 in the sediment protocol of *Stegastes nigricans* and *Caranx melampygus*; and in the NF protocol,
381 *Sphyraena barracuda*. Nevertheless, we also see some other species that are known to be found in
382 the lagoon, such as the *Eucyclogobius newberryi*, being mostly overrepresented in the filtration
383 protocols (NF and PF) compared to the sediments; and *Gila orcutii*, overrepresented in the NF
384 protocol. Two species of dubious taxonomic assignment are also overrepresented in the

385 sediment: *Phoxinus phoxinus* (as discussed in the previous ‘Diversity’ section); and
386 *Acanthogobius flavimanus*, which is a species of goby native to Asia, but that has been recorded
387 previously in California estuaries (Nico et al., 2022). The PERMANOVA result showed a
388 weaker effect than in the rarefied dataset ($R^2 = 0.21$; $p = 0.055$), including for the pairwise
389 comparisons (Table 4).

390 For the rarefied 16S primer dataset, the different protocols showed discernible differences
391 in assemblage composition in the CAP analysis. The first axis explains most of the total variation
392 (CAP1=96%), with the tidewater goby being the most underrepresented in the sediment
393 compared to the filtration protocols (NF and PF), especially in the NF protocol (Fig. 9B). The
394 sediment protocol was also slightly overrepresented by a few other species. One of them was
395 identified as *Nitrosopelagicus brevis*, a species of uncultured “Candidatus”, ammonia-oxidizing
396 archaea (Thaumarchaeota) found mainly in the epi- and upper mesopelagic environments of the
397 open oceans (Santoro et al., 2015). There are also two species of *Monomorpha*, (*M. pyrum* and
398 *M. pseudonordstedti*) that belong to the Euglenaceae family, a group of eukaryotic flagellates
399 found in freshwater environments. Lastly, there is *Elphidium williamsoni*, a foraminifera
400 belonging to the family Elphidiidae found in tidal flats of the North Sea. CAP2 identifies the
401 variation (14%) differentiating the filtration protocols, with the most distinguishing species being
402 the *Guillardia theta*, a species of flagellate algae belonging to the family Geminigeraceae,
403 overrepresented in the PF protocol. Most of these species were also recovered in the DESeq2 and
404 ALDEx2 analyses, with the exception of both *Monomorpha* species. The PERMANOVA
405 result showed a relatively large effect ($R^2 = 0.69$; $p = 0.001$), as well as for all the pairwise
406 comparisons ($R^2 > 0.5$; Table 3).

407 The species represented in the rarefied dataset differ from the ones found when using the
408 eDNA index for the 16S primer. Most of the community assemblage difference in the eDNA
409 index (CAP1=85%) is driven by differences between filtration and sediment protocols, with six
410 species being underrepresented in the latter: *Burkholderiales bacterium* TP637, *Curvibacter* sp.
411 UKPF8, *Diaphorobacter ruginosibacter* and *Verminephrobacter aporrectodeae*
412 (Comamonadaceae); *beta proteobacterium Mzo1* (Oxalobacteraceae); and *Stella humosa*
413 (*Peptostreptococcaceae*). Most of these species were also recovered in the DESeq2 and ALDEx2
414 analyses, with the exception of *V. aporrectodeae*. The PERMANOVA result also showed
415 relatively high effect ($R^2 = 0.44$; $p = 0.001$), but for the pairwise comparisons the effect between
416 the filtration protocols (NF vs. PF) was relatively small ($R^2 = 0.15$; Table 4).

417 Discussion

418 Standardized protocols to process eDNA are under development (e.g. Bohmann et al., 2021), but
419 to implement these efficiently it is necessary to compare biases in taxa detection associated with
420 different protocols. Here, we have explored the detection biases in community composition
421 introduced by freezing water samples prior to filtration (for storage purposes), and the use of
422 sediment samples as an alternative to sampling turbid waters. We find that pre-freezing water
423 does incur some effect on the recovery of community composition, but most strongly for the 16S
424 primer, while having a relatively small effect for the 12S when using larger, cellulose nitrate
425 pore size filters (3 μm). The sediment protocol was able to recover eDNA from organisms that
426 inhabit the water column, however, due to high variability in read abundance among replicates,
427 we suggest increasing the number of biological replicates and/or the volume sampled in the field.

428 Tag-jumping contamination

429 Contamination concerns are usually centered around pre-sequencing, during the field and wet
430 laboratory work. These are of fundamental importance and care should be taken by sterilizing
431 equipment and using negative controls. However, previous literature shows that the sequencing
432 phase can be another source of contamination, generating up to 10% of contaminated reads by
433 tag-jumping (Larsson et al., 2018; Schnell et al., 2015), which can skew analyses of taxa
434 abundance and composition towards the rare taxa. There are ways to help minimize this issue by
435 making use of dual indexing (Kircher et al., 2012)—although see Caroe and Bohmann (2020) for
436 a library approach without dual indexing—, and amplification of positive controls. The latter can
437 be used to track the rate and level of contamination after sequencing to guide read cutoffs on
438 samples (Deiner et al., 2017; Port et al., 2016).

439 Bioinformatics and data pre-processing

440 We relied on a bioinformatic approach developed by the metabar package, adapted from Esling
441 et al. (2015), to reduce the issue of contamination from tag-jumping, since it does not rely solely
442 on the use of positive controls (which we lacked in this analysis) to make the estimated cutoff
443 thresholds. However, after manually checking the fish dataset (12S primer), the final taxonomic
444 tables still contained reads assigned to taxa that are not found in coastal lagoons in California
445 (Table S1). Some of it might be contamination from tag-jumping, although we cannot rule out
446 the possibility that for a few of these species the eDNA could have come from local aquaria, as
447 some are known in the pet trade (e.g. *Acanthurus achilles*). We also cannot disregard the
448 limitations of the reference database, especially related to the absence of estuarine and lagoonal
449 taxa that may lead to dubious assignments to non-local related species (Nagarajan et al., 2022).
450 Due to inability to completely remove potential tag-jump contaminants from the dataset, we can

451 expect a bias towards the rare taxa that will inflate diversity metrics in our samples for all primer
452 sets.

453 The sediment protocol generally showed higher variability among replicates compared to
454 filtration protocols for both primer sets, both in number of reads and community composition
455 (Fig. S3-4). The greater consistency of filtration replicates is a result of the single source for the
456 water samples (the large jug). This means that potentially, multiple biological replicates could be
457 sampled at the site and pooled for fewer downstream analysis, optimizing cost and labor (Sato et
458 al., 2017; Dickie et al., 2018), however, since this was not the focus of this work this design
459 would require further investigation. Sediment replicates were done by individually sampling the
460 bottom of the lagoon. Although replicates were done a few centimeters apart, the bottom of the
461 lagoon appears to have small-scale heterogeneity. The spatial variation of soil and sediment
462 samples is recognized in the literature (Perkins et al., 2014; Taberlet, Prud'Homme, et al., 2012),
463 and can be caused by sediment composition but also by the flow dynamic and distribution of
464 eDNA in the water column. While this variability has been shown to occur for water samples as
465 well in lentic environments (Harper et al., 2019 and references therein), the heterogeneity of
466 water replicates in this system still requires further investigation.

467 The sediment protocol also had an overall lower number of reads compared to the
468 filtration protocols for both primer sets (Fig. S3). The lower number of reads seems to go against
469 the expectations that eDNA can be more concentrated in sediments (Dell'Anno & Corinaldesi,
470 2004; Harper et al., 2019; Turner et al., 2015). This could be due to a few issues, some of which
471 may interact. First, it could be related to a faster degradation and/or turn-over rates of eDNA in
472 the sediment, which are determined by the soil and eDNA characteristics, as well as enzymatic
473 and microbial activities (Levy-Booth et al., 2007; Pietramellara et al., 2009; Torti et al., 2015).

474 The overall lower abundance of eDNA in the sediments could also be driven by increased
475 inhibition (Buxton et al., 2017; Pawlowski et al., 2022). Even though we used a specific soil
476 extraction kit for both sediment and filtration protocols, the purification steps in the protocol
477 could still not have been enough to reduce inhibition in the sediment as well as for the water
478 samples. Lastly, this could have been driven by the much lower volume of sediment used: 0.25-
479 0.3 g versus 500 mL for water samples.

480 There is also the fact that this type of environment is affected by scouring (purging of
481 sediment to the ocean) during high precipitation events and increased flow of freshwater.
482 However, since the sediment collection was done out of the rainy season and the lagoon was
483 closed by a sandbar with no signs of scouring, we are confident that this was not a factor that
484 could have caused the decreased ability to recover eDNA from the sediments. Therefore, we
485 expect that this difference in read abundance between sediment and water samples would be
486 more related to the other factors mentioned above, such as eDNA degradation and turn-over
487 rates, inhibition, and different processed volumes. Considering both the high variability and the
488 lower sequencing throughput of the sediment replicates, we advise using a modified sampling
489 protocol, e.g. the one developed by Taberlet, Prud'Homme, et al. (2012) that includes increasing
490 the number of replicates and mixing larger volumes before downstream processing.

491 Diversity

492 Considering that contamination through tag-jumping could be inflating the numbers of rare
493 species in the dataset, the steepness and lack of a plateau for many of the species rarefaction
494 curves could be artificial (Fig. 4). This is especially evident for the 12S primer since we were
495 able to manually investigate the taxonomy tables (Table S1). However, this lack of a plateau is
496 an expected outcome from environmental samples (Alberdi et al., 2018), and has been shown to

497 occur more acutely in a coastal lagoon in California when compared to other environments in
498 California (Shirazi et al., 2021)—albeit the authors were looking specifically at plants and fungi.
499 The high number of species recovered from the sediment for the 16S primer (Fig. 3) is likely
500 driven by the recovery of a rich and complex sediment biota that is not paralleled in the water
501 column.

502 The low taxonomic assignment to the species level for some of the dubious fish species
503 found in our dataset, e.g. *Phoxinus phoxinus*, *Odontesthes* spp. And *Sebastes pachycephalus*, also
504 highlight the need to expand barcoding efforts to the local estuarine taxa to improve reference
505 databases. On the other hand, *Fundulus diaphanus*, the northeastern killifish, did receive a few
506 high taxonomic scores at the species level, which merit further consideration for biomonitoring
507 of coastal lagoons in the region.

508 Pre-freezing water prior to filtration had an effect on the species curves of the 16S primer
509 dataset, but not so much on the 12S. This could be explained by how differently eDNA
510 molecules are found in the environment for these two different groups of organisms, and how
511 freezing and thawing water would impact them. In the case of the fish fauna, the DNA that is
512 shed from the organisms would be either found within cells, or adsorbed to colloids (Liang &
513 Keeley, 2013; Torti et al., 2015; Turner et al., 2014). Even if cell walls were to disintegrate from
514 the freezing and thawing process, they could still release intact mitochondria (which range from
515 1-8 μm in length) that could still be captured by our 3 μm pore size filters. On the other hand,
516 bacteria and archaea, which are prokaryotic and often single celled organisms, would have their
517 DNA released directly to the medium and pass through the larger pore size filters ($>0.2 \mu\text{m}$).
518 Nevertheless, this freezing effect on cell walls has been shown to not always occur and likely be
519 species-dependent (Sekar et al., 2009; Suomalainen et al., 2006).

520 Differential abundance

521 Pre-freezing water did introduce bias in species abundance compared to the no freezing protocol,
522 although the effect was distinctively larger for the 16S than for the 12S primer dataset when
523 using larger, cellulose nitrate pore size filters (3 μm). Our results seem to align to other reports,
524 where it was shown that freezing had differential effects on detection and relative abundance of
525 different prokaryotic taxa (Kwambana et al., 2011; Sekar et al., 2009; Suomalainen et al., 2006).
526 The smaller effect seen for the 12S dataset could be related to the different qualities and
527 properties of eDNA in the environment, as mentioned in the previous section ('Diversity').
528 Larger, mitochondrial molecules from Eukaryotes may be more easily captured in the filters
529 compared to the smaller celled, prokaryotic organisms targeted by the 16S primer.

530 Naturally, due to eDNA precipitation and resuspension, we expect to capture some
531 community overlap between water and surficial sediment samples, however abundances should
532 be different as consequence of the origin and fate of the eDNA in the environment and the
533 processes acting on it throughout (Torti et al., 2015). As expected, in the differential abundance
534 analyses, we see overrepresentation of families of algae (Hemiselmidaceae and Geminigeraceae),
535 environmental and aquatic bacteria (Comamonadaceae, Flavobacteriaceae, Cryomorphaceae and
536 Rhodobacteraceae), and cyanobacteria (Aphanizomenonaceae) in the water samples (NF and PF
537 protocols). There is also significantly higher representation of presumptively benthic diatoms
538 (Catenulaceae and Fragilariaceae), environmental bacteria (Anaerolineaceae and
539 Desulfobacteraceae) and foraminiferans (Elphidiidae and Nonionidae) in the sediment. In
540 addition, the types of environmental bacteria most abundant in the sediments were typical of soil
541 and sediments elsewhere. Of particular note are those from anoxic environments (e.g.
542 Anaerolineaceae and Desulfobacteraceae) as lagoon sediments in this region are often dark and
543 sulfide-rich, consistent with anoxia and sulfur-cycling.

544 In the family Flavobacteriaceae there are important pathogens of fish and humans that
545 belong to the genus *Flavobacterium*. Suomalainen et al.(2006) found that *F. columnare* was
546 more susceptible to having its cell walls disrupted to freezing due to high amounts of DNAases,
547 lyases and proteases, likely connected to its pathogenicity, which then led to lower rates of DNA
548 recovery. The species found in our dataset was *F. johnsoniae*, a species not known to be
549 pathogenic–albeit with low species taxonomic score. Given that there was no difference in
550 abundance for this species in our NF and PF protocols, which is in contrast with the results for
551 the pathogenic species, *F. columnare*, this might relate to a true non-pathogenic organism.
552 However, considering that the endangered northern tidewater goby often achieves high
553 abundance in this lagoon, more detailed assessment of the *Flavobacterium* species inhabiting this
554 site would be of interest.

555 The other species assignment that draws our attention is the archaea Candidatus
556 *Nitrosopelagicus brevis* (Thaumarchaeota), which was shown as more abundant in sediment than
557 water samples. As mentioned earlier, this is a pelagic species, normally found in the open ocean
558 worldwide. Although coastal lagoons are subject to marine input, the relatively high
559 concentration in sediment is unexpected and merits inquiry, especially considering that the
560 confidence in its taxonomic assignment was low across reads. More likely, this represents a new
561 environmental archaea that is abundant in coastal lagoon sediments.

562 Beta diversity

563 McMurdie and Holmes (2014) recommends against rarefying datasets due to the risk of
564 removing true, rare ASVs. However, in our case, where we were unable to completely remove
565 tag-jumping contaminants, this pre-process could help alleviate some of the noise caused by
566 contaminants. Nevertheless, the CAP and PERMANOVA results on both the rarefied and

567 standardized (eDNA index) dataset are mostly in agreement, and show greater differences in
568 assemblage composition for the 16S primer than for the 12S primer, especially between the
569 filtration and sediment protocols (NF vs. Sed; PF vs. Sed). Although results from the 12S primer
570 were highly affected by the tag-jump contaminants in the dataset standardized using the eDNA
571 index (Fig. S5A), the CAP analysis was able to capture the overrepresentation of tidewater goby
572 in the filtration protocols compared to the sediment, a signal that was lost in the rarefied data
573 (Fig. 9A).

574 For the rarefied 16S primer dataset, most species that were over- and underrepresented in
575 the CAP analysis were also captured by the differential abundance analyses, such as *Guillardia*
576 *theta* (Geminigeraceae), which was overrepresented in the PF protocol compared to the NF
577 protocol. In addition, the species of foraminifera, *Elphidium williamsoni* (Elphidiidae) and the
578 archaea *Candidatus Nitrosopelagicus brevis* (Thaumarchaeota) were found to be overrepresented
579 in the sediment compared to the filtration protocols. The CAP results on the dataset standardized
580 using the eDNA index (Fig. S5B) showed different species as underrepresented in the sediment
581 compared to the filtration protocols, but those also showed up as significantly different in the
582 differential abundance analyses, with the exception of one, *Verminephrobacter aporrectodeae*.

583 Interestingly, the CAP analysis was also able to capture the underrepresentation of
584 tidewater gobies (*E. newberryi*) in the sediment protocol on the 16S primer when compared to
585 the NF protocol (Fig. 9B). The 16S ribosomal RNA region have been used for fish species
586 detection, and is not uncommon that different primer sets are taxonomically congruent (Shu,
587 Lidwig & Peng, 2020; Alberdi et al., 2018). This reinforces the idea discussed earlier
588 ('Bioinformatics and data pre-processing' section) that fish eDNA, at least in this environment,
589 is less concentrated in the sediment than in the water column, which contradicts other findings

590 from the literature (Perkins et al., 2014; Turner et al., 2015). But it is worth noting that this
591 underrepresentation of fish eDNA in the sediment was found to be of small effect for the 12S
592 primer, though, and there could be some bias related to how these two genes behave and degrade
593 differently in the environment for the fish fauna.

594 Lessons Learned

595 Here is a list of recommendations and best practices for eDNA sampling and analysis in coastal
596 environments that we have learned throughout this work and believe will be useful for others
597 working in similar environments with turbid water and highly heterogeneous sediment/soil:

- 598 1. Filtered water samples had an overall higher number of reads compared to sediment for
599 both primer sets. Therefore, we recommend the use of this protocol as it will increase
600 chances of species detection;
- 601 2. If using sediment samples, we recommend increasing the number of replicates and
602 mixing larger volumes before processing for DNA extractions (as in Taberlet,
603 Prud'Homme, et al., 2012);
- 604 3. Pre-freezing water samples prior to filtration can be an effective storage solution, at least
605 for Eukaryotes, and when used in combination with cellulose nitrate filters of up to 3 μm
606 pore size filters;
- 607 4. The use of dual-indexing and positive controls during library preparation will help
608 minimize and address cross-contamination from tag-jumping, as is now widely
609 recognized in many best-practice protocols (e.g. Deiner et al., 2017; Goldberg et al.,
610 2016);
- 611 5. Although rarefying the dataset is not recommended (McMurdie & Holmes, 2014), we
612 recognize that it can aid in reducing the noise of contaminants from your dataset, as long

613 as they are rare. Otherwise, the use of eDNA index (Kelly et al., 2019) can be an
614 alternative to standardize your dataset.

615 Conclusions

616 In this work, we assessed environmental DNA protocols for use in coastal lagoons, a highly
617 dynamic habitat at the intersection of terrestrial, freshwater and marine environments. Pre-
618 freezing water combined with the use of larger pore size filters (at least up to 3 μm) can be a
619 viable alternative for storage and processing of turbid water samples, at least in the case of fish
620 communities (12S, MiFish). However, the use of sediment samples as an alternative to
621 processing water samples should be done with caution, and at minimum the number of biological
622 replicates and/or volume sampled should be increased. Also, while sediment samples were able
623 to recover eDNA from organisms commonly found in the water column, such as the tidewater
624 goby, this was achieved during a period of relatively long lagoon closure, when there was no
625 recent scouring of sediments to the ocean.

626 While we expect these guidelines to be helpful in the development of strategies to use
627 eDNA as a monitoring resource in similar environments, protocol testing is still strongly advised
628 whenever possible, especially when working in a new system. Much work is necessary to
629 understand the full potential eDNA brings for the conservation and restoration of endangered
630 species and habitats.

631

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643

644 References

- 645 Alberdi, A., Aizpurua, O., Gilbert, M. T. P., & Bohmann, K. (2018). Scrutinizing key steps for
646 reliable metabarcoding of environmental samples. *Methods in Ecology and Evolution*,
647 9(1), 134–147. <https://doi.org/10.1111/2041-210X.12849>
- 648 Ballard, J., Pezda, J., Spencer, D., & Plantinga, A. (n.d.). *An Economic Valuation of Southern*
649 *California Coastal Wetlands*. [http://scwrp.org/wp-](http://scwrp.org/wp-content/uploads/2017/06/SoCalWetlands_FinalReport.pdf)
650 [content/uploads/2017/06/SoCalWetlands_FinalReport.pdf](http://scwrp.org/wp-content/uploads/2017/06/SoCalWetlands_FinalReport.pdf)
- 651 Bischoff, V., Zucker, F., & Moraru, C. (2021). Marine Bacteriophages. In D. H. Bamford & M.
652 Zuckerman (Eds.), *Encyclopedia of Virology (Fourth Edition)* (pp. 322–341). Academic
653 Press. <https://doi.org/10.1016/B978-0-12-809633-8.20988-6>
- 654 Bohmann, K., Chua, P., Holman, L. E., & Lynggaard, C. (2021). DNAqua-Net conference unites
655 participants from around the world with the quest to standardize and implement DNA-
656 based aquatic biomonitoring. *Environmental DNA*, 3(5), 884–888.
657 <https://doi.org/10.1002/edn3.207>
- 658 Bowman, J. P. (2014). The Family Cryomorphaceae. In E. Rosenberg, E. F. DeLong, S. Lory, E.
659 Stackebrandt, & F. Thompson (Eds.), *The Prokaryotes* (pp. 539–550). Springer Berlin
660 Heidelberg. https://doi.org/10.1007/978-3-642-38954-2_135
- 661 Buxton, A. S., Groombridge, J. J., & Griffiths, R. A. (2017). Is the detection of aquatic
662 environmental DNA influenced by substrate type? *PLOS ONE*, 12(8), e0183371.
663 <https://doi.org/10.1371/journal.pone.0183371>
- 664 Callahan, B. J., McMurdie, P. J., Rosen, M. J., Han, A. W., Johnson, A. J. A., & Holmes, S. P.
665 (2016). DADA2: High-resolution sample inference from Illumina amplicon data. *Nature*
666 *Methods*, 13(7), 581–583. <https://doi.org/10.1038/nmeth.3869>

- 667 Caporaso, J. G., Lauber, C. L., Walters, W. A., Berg-Lyons, D., Huntley, J., Fierer, N., Owens,
668 S. M., Betley, J., Fraser, L., Bauer, M., Gormley, N., Gilbert, J. A., Smith, G., & Knight,
669 R. (2012). Ultra-high-throughput microbial community analysis on the Illumina HiSeq
670 and MiSeq platforms. *The ISME Journal*, 6(8), 1621–1624.
671 <https://doi.org/10.1038/ismej.2012.8>
- 672 Caroe, C., & Bohmann, K. (2020). Tagsteady: A metabarcoding library preparation protocol to
673 avoid false assignment of sequences to samples. *BioRxiv*.
- 674 Chen, H. (2018). *VennDiagram: Generate High-Resolution Venn and Euler Plots* (1.6.20).
675 <https://CRAN.R-project.org/package=VennDiagram>
- 676 Curd, E., Gomer, J., Kandlikar, G., Gold, Z., Ogden, M., & Shi, B. (2018). *The Anacapa Toolkit*.
677 <https://github.com/limey-bean/Anacapa>
- 678 Curd, E., Kandlikar, G., & Gomer, J. (2018). *CRUX: Creating Reference libraries Using*
679 *eXisting tools*. [https://github.com/limey-bean/CRUX_Creating-Reference-libraries-](https://github.com/limey-bean/CRUX_Creating-Reference-libraries-Using-eXisting-tools)
680 [Using-eXisting-tools](https://github.com/limey-bean/CRUX_Creating-Reference-libraries-Using-eXisting-tools)
- 681 Deiner, K., Bik, H. M., Mächler, E., Seymour, M., Lacoursière-Roussel, A., Altermatt, F., Creer,
682 S., Bista, I., Lodge, D. M., & De Vere, N. (2017). Environmental DNA metabarcoding:
683 Transforming how we survey animal and plant communities. *Molecular Ecology*, 26(21),
684 5872–5895.
- 685 Deiner, K., Walser, J.-C., Mächler, E., & Altermatt, F. (2015). Choice of capture and extraction
686 methods affect detection of freshwater biodiversity from environmental DNA. *Biological*
687 *Conservation*, 183, 53–63.
- 688 Dejean, T., Valentini, A., Miquel, C., Taberlet, P., Bellemain, E., & Miaud, C. (2012). Improved
689 detection of an alien invasive species through environmental DNA barcoding: The

- 690 example of the American bullfrog *Lithobates catesbeianus*: Alien invasive species
691 detection using eDNA. *Journal of Applied Ecology*, 49(4), 953–959.
692 <https://doi.org/10.1111/j.1365-2664.2012.02171.x>
- 693 Dell’Anno, A., & Corinaldesi, C. (2004). Degradation and Turnover of Extracellular DNA in
694 Marine Sediments: Ecological and Methodological Considerations Degradation and
695 Turnover of Extracellular DNA in Marine Sediments: Ecological and Methodological
696 Considerations. *Applied and Environmental Microbiology*, 70(7), 4384–4386.
697 <https://doi.org/10.1128/AEM.70.7.4384>
- 698 Doi, H., Uchii, K., Matsushashi, S., Takahara, T., Yamanaka, H., & Minamoto, T. (2017).
699 Isopropanol precipitation method for collecting fish environmental DNA. *Limnology and*
700 *Oceanography: Methods*, 15(2), 212–218. <https://doi.org/10.1002/lom3.10161>
- 701 Earl, D. A., Louie, K. D., Bardeleben, C., Swift, C. C., & Jacobs, D. K. (2010). Rangewide
702 microsatellite phylogeography of the endangered tidewater goby, a genetically
703 subdivided coastal fish with limited marine dispersal. *Conservation Genetics*, 11, 103–
704 104. <https://doi.org/10.1007/s10592-009-0008-9>
- 705 Esling, P., Lejzerowicz, F., & Pawlowski, J. (2015). Accurate multiplexing and filtering for high-
706 throughput amplicon-sequencing. *Nucleic Acids Research*, 43(5), 2513–2524.
707 <https://doi.org/10.1093/nar/gkv107>
- 708 *FASTX-Toolkit*. (n.d.). Retrieved January 11, 2018, from http://hannonlab.cshl.edu/fastx_toolkit/
- 709 Fernandes, A. D., Macklaim, J. M., Linn, T. G., Reid, G., & Gloor, G. B. (2013). ANOVA-Like
710 Differential Expression (ALDEx) Analysis for Mixed Population RNA-Seq. *PLoS ONE*,
711 8(7), e67019. <https://doi.org/10.1371/journal.pone.0067019>

- 712 Ficetola, G. F., Miaud, C., Pompanon, F., & Taberlet, P. (2008). Species detection using
713 environmental DNA from water samples. *Biology Letters*, 4(4), 423–425.
714 <https://doi.org/10.1098/rsbl.2008.0118>
- 715 Gao, X., Lin, H., Revanna, K., & Dong, Q. (2017). A Bayesian taxonomic classification method
716 for 16S rRNA gene sequences with improved species-level accuracy. *BMC*
717 *Bioinformatics*, 18(1), 247. <https://doi.org/10.1186/s12859-017-1670-4>
- 718 Gbif.Org. (2022). *Occurrence Download* (p. 170487) [Darwin Core Archive]. The Global
719 Biodiversity Information Facility. <https://doi.org/10.15468/DL.HTJ3HT>
- 720 Goldberg, C. S., Turner, C. R., Deiner, K., Klymus, K. E., Thomsen, P. F., Murphy, M. A.,
721 Spear, S. F., McKee, A., Oyler-McCance, S. J., & Cornman, R. S. (2016). Critical
722 considerations for the application of environmental DNA methods to detect aquatic
723 species. *Methods in Ecology and Evolution*, 7(11), 1299–1307.
- 724 Harper, L. R., Buxton, A. S., Rees, H. C., Bruce, K., Brys, R., Halfmaerten, D., Read, D. S.,
725 Watson, H. V., Sayer, C. D., & Jones, E. P. (2019). Prospects and challenges of
726 environmental DNA (eDNA) monitoring in freshwater ponds. *Hydrobiologia*, 826(1),
727 25–41.
- 728 Hinlo, R., Gleeson, D., Lintermans, M., & Furlan, E. (2017). Methods to maximise recovery of
729 environmental DNA from water samples. *PloS One*, 12(6), e0179251.
- 730 Jacobs, D. K., Stein, E. D., & Longcore, T. (2011). Classification of California Estuaries Based
731 on Natural Closure Patterns: Templates for Restoration and Management Management.
732 *Technical Report, August*, 1–72.
- 733 Kandlikar, G. (2020). *ranacapa: Utility Functions and “shiny” App for Simple Environmental*
734 *DNA Visualizations and Analyses* (0.1.0). <https://github.com/gauravsk/ranacapa>

- 735 Kelly, R. P., Shelton, A. O., & Gallego, R. (2019). Understanding PCR Processes to Draw
736 Meaningful Conclusions from Environmental DNA Studies. *Scientific Reports*, *9*(1),
737 12133. <https://doi.org/10.1038/s41598-019-48546-x>
- 738 Kircher, M., Sawyer, S., & Meyer, M. (2012). Double indexing overcomes inaccuracies in
739 multiplex sequencing on the Illumina platform. *Nucleic Acids Research*, *40*(1), e3–e3.
740 <https://doi.org/10.1093/nar/gkr771>
- 741 Kumar, G., Farrell, E., Reaume, A. M., Eble, J. A., & Gaither, M. R. (2022). One size does not
742 fit all: Tuning eDNA protocols for high- and low-turbidity water sampling.
743 *Environmental DNA*, *4*(1), 167–180. <https://doi.org/10.1002/edn3.235>
- 744 Kwambana, B. A., Mohammed, N. I., Jeffries, D., Barer, M., Adegbola, R. A., & Antonio, M.
745 (2011). Differential effects of frozen storage on the molecular detection of bacterial taxa
746 that inhabit the nasopharynx. *BMC Clinical Pathology*, *11*(1), 2.
747 <https://doi.org/10.1186/1472-6890-11-2>
- 748 Langmead, B., & Salzberg, S. L. (2012). Fast gapped-read alignment with Bowtie 2. *Nature*
749 *Methods*, *9*(4), 357–359. <https://doi.org/10.1038/nmeth.1923>
- 750 Laramie, M. B., Pilliod, D. S., Goldberg, C. S., & Strickler, K. M. (2015). Environmental DNA
751 sampling protocol—Filtering water to capture DNA from aquatic organisms. *U.S*
752 *Geological Survey Techniques and Methods, Book 2*(Chapter A13), 15 p.
753 <https://doi.org/10.3133/TM2A13>
- 754 Larsson, A. J. M., Stanley, G., Sinha, R., Weissman, I. L., & Sandberg, R. (2018).
755 Computational correction of index switching in multiplexed sequencing libraries. *Nature*
756 *Methods*, *15*(5), 305–307. <https://doi.org/10.1038/nmeth.4666>

- 757 Leray, M., Yang, J. Y., Meyer, C. P., Mills, S. C., Agudelo, N., Ranwez, V., Boehm, J. T., &
758 Machida, R. J. (2013). A new versatile primer set targeting a short fragment of the
759 mitochondrial COI region for metabarcoding metazoan diversity: Application for
760 characterizing coral reef fish gut contents. *Frontiers in Zoology*, *10*(1), 34.
761 <https://doi.org/10.1186/1742-9994-10-34>
- 762 Levy-Booth, D. J., Campbell, R. G., Gulden, R. H., Hart, M. M., Powell, J. R., Klironomos, J.
763 N., Pauls, K. P., Swanton, C. J., Trevors, J. T., & Dunfield, K. E. (2007). Cycling of
764 extracellular DNA in the soil environment. *Soil Biology and Biochemistry*, *39*(12), 2977–
765 2991. <https://doi.org/10.1016/j.soilbio.2007.06.020>
- 766 Li, J., Lawson Handley, L.-J., Read, D. S., & Hänfling, B. (2018). The effect of filtration method
767 on the efficiency of environmental DNA capture and quantification via metabarcoding.
768 *Molecular Ecology Resources*, *18*(5), 1102–1114.
- 769 Liang, Z., & Keeley, A. (2013). Filtration Recovery of Extracellular DNA from Environmental
770 Water Samples. *Environmental Science & Technology*, *47*(16), 9324–9331.
771 <https://doi.org/10.1021/es401342b>
- 772 Love, M. I., Huber, W., & Anders, S. (2014). Moderated estimation of fold change and
773 dispersion for RNA-seq data with DESeq2. *Genome Biology*, *15*(12), 550.
774 <https://doi.org/10.1186/s13059-014-0550-8>
- 775 Majaneva, M., Diserud, O. H., Eagle, S. H., Boström, E., Hajibabaei, M., & Ekrem, T. (2018).
776 Environmental DNA filtration techniques affect recovered biodiversity. *Scientific*
777 *Reports*, *8*(1), 1–11.
- 778 Martin, M. (2013). Cutadapt removes adapter sequences from high-throughput sequencing reads.
779 *EMBnet.Journal*, *17*(1), 10–12.

- 780 McMurdie, P. J., & Holmes, S. (2013). phyloseq: An R Package for Reproducible Interactive
781 Analysis and Graphics of Microbiome Census Data. *PLoS ONE*, 8(4), e61217.
782 <https://doi.org/10.1371/journal.pone.0061217>
- 783 McMurdie, P. J., & Holmes, S. (2014). Waste Not, Want Not: Why Rarefying Microbiome Data
784 Is Inadmissible. *PLoS Computational Biology*, 10(4), e1003531.
785 <https://doi.org/10.1371/journal.pcbi.1003531>
- 786 Miya, M., Sato, Y., Fukunaga, T., Sado, T., Poulsen, J. Y., Sato, K., Minamoto, T., Yamamoto,
787 S., Yamanaka, H., Araki, H., Kondoh, M., & Iwasaki, W. (2015). MiFish, a set of
788 universal PCR primers for metabarcoding environmental DNA from fishes: Detection of
789 more than 230 subtropical marine species. *Royal Society Open Science*, 2(7).
790 <http://rsos.royalsocietypublishing.org/content/2/7/150088.abstract>
- 791 Nagarajan, R. P., Bedwell, M., Holmes, A. E., Sanches, T., Acuña, S., Baerwald, M., Barnes, M.
792 A., Blankenship, S., Connon, R. E., Deiner, K., Gille, D., Goldberg, C. S., Hunter, M. E.,
793 Jerde, C. L., Luikart, G., Meyer, R. S., Watts, A., & Schreier, A. (2022). Environmental
794 DNA Methods for Ecological Monitoring and Biodiversity Assessment in Estuaries.
795 *Estuaries and Coasts*, 45(7), 2254–2273. <https://doi.org/10.1007/s12237-022-01080-y>
- 796 Oksanen, J., Blanchet, F. G., Friendly, M., Kindt, R., Legendre, P., McGlenn, D., Minchin, P. R.,
797 O'Hara, R. B., Simpson, G. L., Solymos, P., Stevens, M. H. H., Szoecs, E., & Wagner, H.
798 (2019). *vegan: Community Ecology Package* (2.5-6). [https://CRAN.R-](https://CRAN.R-project.org/package=vegan)
799 [project.org/package=vegan](https://CRAN.R-project.org/package=vegan)
- 800 Pawlowski, J., Bruce, K., Panksep, K., Aguirre, F. I., Amalfitano, S., Apothéloz-Perret-Gentil,
801 L., Baussant, T., Bouchez, A., Carugati, L., Cermakova, K., Cordier, T., Corinaldesi, C.,
802 Costa, F. O., Danovaro, R., Dell'Anno, A., Duarte, S., Eisendle, U., Ferrari, B. J. D.,

- 803 Frontalini, F., ... Fazi, S. (2022). Environmental DNA metabarcoding for benthic
804 monitoring: A review of sediment sampling and DNA extraction methods. *Science of The*
805 *Total Environment*, 818, 151783. <https://doi.org/10.1016/j.scitotenv.2021.151783>
- 806 Perkins, T. L., Clements, K., Baas, J. H., Jago, C. F., Jones, D. L., Malham, S. K., & McDonald,
807 J. E. (2014). Sediment Composition Influences Spatial Variation in the Abundance of
808 Human Pathogen Indicator Bacteria within an Estuarine Environment. *PLoS ONE*, 9(11),
809 e112951. <https://doi.org/10.1371/journal.pone.0112951>
- 810 Pietramellara, G., Ascher, J., Borgogni, F., Ceccherini, M. T., Guerri, G., & Nannipieri, P.
811 (2009). Extracellular DNA in soil and sediment: Fate and ecological relevance. *Biology*
812 *and Fertility of Soils*, 45(3), 219–235. <https://doi.org/10.1007/s00374-008-0345-8>
- 813 Pilliod, D. S., Goldberg, C. S., Arkle, R. S., Waits, L. P., & Richardson, J. (2013). Estimating
814 occupancy and abundance of stream amphibians using environmental DNA from filtered
815 water samples. *Canadian Journal of Fisheries and Aquatic Sciences*, 70(8), 1123–1130.
816 <https://doi.org/10.1139/cjfas-2013-0047>
- 817 Port, J. A., O'Donnell, J. L., Romero-Maraccini, O. C., Leary, P. R., Litvin, S. Y., Nickols, K. J.,
818 Yamahara, K. M., & Kelly, R. P. (2016). Assessing vertebrate biodiversity in a kelp
819 forest ecosystem using environmental DNA. *Molecular Ecology*, 25(2), 527–541.
820 <https://doi.org/10.1111/mec.13481>
- 821 R Core Team. (2018). *R: A language and environment for statistical computing*. R Foundation
822 for Statistical Computing. <https://www.R-project.org/>
- 823 Rees, H. C., Maddison, B. C., Middleditch, D. J., Patmore, J. R. M., & Gough, K. C. (2014).
824 REVIEW: The detection of aquatic animal species using environmental DNA – a review

825 of eDNA as a survey tool in ecology. *Journal of Applied Ecology*, 51(5), 1450–1459.
826 <https://doi.org/10.1111/1365-2664.12306>

827 Robson, H. L. A., Noble, T. H., Saunders, R. J., Robson, S. K. A., Burrows, D. W., & Jerry, D.
828 R. (2016). Fine-tuning for the tropics: Application of eDNA technology for invasive fish
829 detection in tropical freshwater ecosystems. *Molecular Ecology Resources*, 16(4), 922–
830 932. <https://doi.org/10.1111/1755-0998.12505>

831 RStudio Team. (2020). *RStudio: Integrated Development for R*. RStudio, PBC.
832 <http://www.rstudio.com/>

833 Sales, N. G., Wangenstein, O. S., Carvalho, D. C., & Mariani, S. (2019). Influence of
834 preservation methods, sample medium and sampling time on eDNA recovery in a
835 neotropical river. *Environmental DNA*, 1(2), edn3.14. <https://doi.org/10.1002/edn3.14>

836 Santoro, A. E., Dupont, C. L., Richter, R. A., Craig, M. T., Carini, P., McIlvin, M. R., Yang, Y.,
837 Orsi, W. D., Moran, D. M., & Saito, M. A. (2015). Genomic and proteomic
838 characterization of “*Candidatus Nitrosopelagicus brevis*”: An ammonia-oxidizing
839 archaeon from the open ocean. *Proceedings of the National Academy of Sciences*, 112(4),
840 1173–1178. <https://doi.org/10.1073/pnas.1416223112>

841 Sard, N. M., Herbst, S. J., Nathan, L., Uhrig, G., Kanefsky, J., Robinson, J. D., & Scribner, K. T.
842 (2019). Comparison of fish detections, community diversity, and relative abundance
843 using environmental DNA metabarcoding and traditional gears. *Environmental DNA*,
844 1(4), 368–384. <https://doi.org/10.1002/edn3.38>

845 Schaarschmidt, F., & Gerhard, D. (2019). *PairwiseCI: Confidence Intervals for Two Sample*
846 *Comparisons* (0.1-27). <https://CRAN.R-project.org/package=pairwiseCI>

- 847 Schnell, I. B., Bohmann, K., & Gilbert, M. T. P. (2015). Tag jumps illuminated—Reducing
848 sequence-to-sample misidentifications in metabarcoding studies. *Molecular Ecology*
849 *Resources*, 15(6), 1289–1303. <https://doi.org/10.1111/1755-0998.12402>
- 850 SCWRP. (2018). *Wetlands on the Edge: The Future of Southern California’s Wetlands:*
851 *Regional Strategy 2018* (p. 142). California State Coastal Conservancy.
852 scwrp.databasin.org
- 853 Sekar, R., Kaczmarczyk, L. T., & Richardson, L. L. (2009). Effect of Freezing on PCR
854 Amplification of 16S rRNA Genes from Microbes Associated with Black Band Disease
855 of Corals. *Applied and Environmental Microbiology*, 75(8), 2581–2584.
856 <https://doi.org/10.1128/AEM.01500-08>
- 857 Shaffer, H. B., Fellers, G. M., Randal Voss, S., Oliver, J. C., & Pauly, G. B. (2004). Species
858 boundaries, phylogeography and conservation genetics of the red-legged frog (*Rana*
859 *aurora/draytonii*) complex. *Molecular Ecology*, 13(9), 2667–2677.
860 <https://doi.org/10.1111/j.1365-294X.2004.02285.x>
- 861 Shirazi, S., Meyer, R. S., & Shapiro, B. (2021). Revisiting the effect of PCR replication and
862 sequencing depth on biodiversity metrics in environmental DNA metabarcoding. *Ecology*
863 *and Evolution*, 11(22), 15766–15779. <https://doi.org/10.1002/ece3.8239>
- 864 Smart, A. S., Weeks, A. R., Rooyen, A. R., Moore, A., McCarthy, M. A., & Tingley, R. (2016).
865 Assessing the cost-efficiency of environmental DNA sampling. *Methods in Ecology and*
866 *Evolution*, 7(11), 1291–1298. <https://doi.org/10.1111/2041-210X.12598>
- 867 Stein, E. D., Cayce, K., Salomon, M., Bram, D. L., De Mello, D., Grossinger, R., & Dark, S.
868 (2014). *Wetlands of the Southern California Coast: Historical Extent and Change Over*
869 *Time* (SFEI Report 720; SCCWRP Technical Report 826; p. 58). Southern California

- 870 Coastal Water Research Project and San Francisco Estuary Institute.
871 <https://www.caltsheets.org/socal/download.html>
- 872 Suomalainen, L.-R., Reunanen, H., Ijäs, R., Valtonen, E. T., & Tirola, M. (2006). Freezing
873 Induces Biased Results in the Molecular Detection of *Flavobacterium columnare*.
874 *Applied and Environmental Microbiology*, 72(2), 1702–1704.
875 <https://doi.org/10.1128/AEM.72.2.1702-1704.2006>
- 876 Swift, C. C., Haglund, T. R., Ruiz, M., & Fisher, R. N. (1993). The Status and Distribution of the
877 Freshwater Fishes of Southern California. *Bulletin of the Southern California Academy of*
878 *Sciences*, 92(3), 101–167.
- 879 Swift, C. C., Spies, B., Ellingson, R. A., & Jacobs, D. K. (2016). A New Species of the Bay
880 Goby Genus *Eucyclogobius*, Endemic to Southern California: Evolution, Conservation,
881 and Decline. *PloS One*, 11(7), e0158543. <https://doi.org/10.1371/journal.pone.0158543>
- 882 Taberlet, P., Coissac, E., Pompanon, F., Brochmann, C., & Willerslev, E. (2012). Towards next-
883 generation biodiversity assessment using DNA metabarcoding: NEXT-GENERATION
884 DNA METABARCODING. *Molecular Ecology*, 21(8), 2045–2050.
885 <https://doi.org/10.1111/j.1365-294X.2012.05470.x>
- 886 Taberlet, P., Prud'Homme, S. M., Campione, E., Roy, J., Miquel, C., Shehzad, W., Gielly, L.,
887 Rioux, D., Choler, P., Clément, J.-C., Melodelima, C., Pompanon, F., & Coissac, E.
888 (2012). Soil sampling and isolation of extracellular DNA from large amount of starting
889 material suitable for metabarcoding studies: EXTRACTION OF EXTRACELLULAR
890 DNA FROM SOIL. *Molecular Ecology*, 21(8), 1816–1820.
891 <https://doi.org/10.1111/j.1365-294X.2011.05317.x>

- 892 Takahara, T., Minamoto, T., & Doi, H. (2015). Effects of sample processing on the detection rate
893 of environmental DNA from the Common Carp (*Cyprinus carpio*). *Biological*
894 *Conservation*, *183*, 64–69. <https://doi.org/10.1016/j.biocon.2014.11.014>
- 895 Thomsen, P. F., & Willerslev, E. (2015). Environmental DNA – An emerging tool in
896 conservation for monitoring past and present biodiversity. *Biological Conservation*, *183*,
897 4–18. <https://doi.org/10.1016/j.biocon.2014.11.019>
- 898 Torti, A., Lever, M. A., & Jørgensen, B. B. (2015). Origin, dynamics, and implications of
899 extracellular DNA pools in marine sediments. *Marine Genomics*, *24*, 185–196.
900 <https://doi.org/10.1016/j.margen.2015.08.007>
- 901 Tsuji, S., Takahara, T., Doi, H., Shibata, N., & Yamanaka, H. (2019). The detection of aquatic
902 macroorganisms using environmental DNA analysis—A review of methods for
903 collection, extraction, and detection. *Environmental DNA*, *1*(2), 99–108.
904 <https://doi.org/10.1002/edn3.21>
- 905 Turner, C. R., Barnes, M. A., Xu, C. C. Y., Jones, S. E., Jerde, C. L., & Lodge, D. M. (2014).
906 Particle size distribution and optimal capture of aqueous microbial eDNA. *Methods in*
907 *Ecology and Evolution*, *5*(7), 676–684. <https://doi.org/10.1111/2041-210X.12206>
- 908 Turner, C. R., Uy, K. L., & Everhart, R. C. (2015). Fish environmental DNA is more
909 concentrated in aquatic sediments than surface water. *Biological Conservation*, *183*, 93–
910 102. <https://doi.org/10.1016/j.biocon.2014.11.017>
- 911 van der Loos, L. M., & Nijland, R. (2021). Biases in bulk: DNA metabarcoding of marine
912 communities and the methodology involved. *Molecular Ecology*, *30*(13), 3270–3288.
913 <https://doi.org/10.1111/mec.15592>

- 914 WeatherSpark.com. (n.d.). *Historical Weather Summer 2018 at Point Mugu Naval Air Warfare*
915 *Center*. WeatherSpark.Com. Retrieved June 8, 2022, from
916 [https://weatherspark.com/h/s/145310/2018/1/Historical-Weather-Summer-2018-at-Point-](https://weatherspark.com/h/s/145310/2018/1/Historical-Weather-Summer-2018-at-Point-Mugu-Naval-Air-Warfare-Center;-California;-United-States#Figures-Rainfall)
917 [Mugu-Naval-Air-Warfare-Center;-California;-United-States#Figures-Rainfall](https://weatherspark.com/h/s/145310/2018/1/Historical-Weather-Summer-2018-at-Point-Mugu-Naval-Air-Warfare-Center;-California;-United-States#Figures-Rainfall)
- 918 Williams, K. E., Huyvaert, K. P., & Piaggio, A. J. (2017). Clearing muddied waters: Capture of
919 environmental DNA from turbid waters. *PLOS ONE*, 12(7), e0179282.
920 <https://doi.org/10.1371/journal.pone.0179282>
- 921 Zinger, L., Lionnet, C., Benoiston, A.-S., Donald, J., Mercier, C., & Boyer, F. (2020). *metabarR:*
922 *An R package for the evaluation and improvement of DNA metabarcoding data quality*
923 [Preprint]. Bioinformatics. <https://doi.org/10.1101/2020.08.28.271817>
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926 Tables

927 Table 1: Detailed information of the primers used.

Primer	Targets	Forward Primer	Reverse Primer	Reference
12S	Fish	GTCGGTAAAACCTCGTG CCAGC	CATAGTGGGGTATCT AATCCCAGTTTG	Miya et al., 2015
16S	Bacteria and archaea	GTGYCAGCMGCCGCGG TAA	GGACTACNVGGGTWT CTAAT	Caporaso et al., 2012 (F: 515F and R: 806R)
COI	Animals	GGWACWGGWTGAACW GTWTAYCCYCC	TANACYTCnGGRTGN CCRAARAYCA	Leray et al., 2013

928

929 Table 2: Confidence interval estimates and upper/lower limits of species rarefaction curves (Fig.
930 4). NF: no freezing; PF: pre-freezing; Sed: sediment.

Primer	Comparison	Estimate	Lower	Upper
	PF-NF	-1.43693	-3.77705	0.903185
12S	Sed-NF	-6.24818	-11.198	-1.29838
	Sed-PF	-4.81125	-9.76679	0.144296
	PF-NF	-0.2004	-0.29947	-0.10132
16S	Sed-NF	-0.54775	-0.73983	-0.35568
	Sed-PF	-0.34736	-0.54064	-0.15408

931

932

933 Table 3: PERMANOVA and pairwise PERMANOVA (rarefied dataset) between all three
 934 protocols: no-freezing, pre-freezing and sediment. P.adjusted is the adjusted p-value after FDR
 935 correction.

Primer	PERMANOVA		Pairwise PERMANOVA				
	R2	Pr(>F)	Comparison	F.Model	R2	p.value	p.adjusted
12S	0.3803	0.05	No-freezing vs Pre-freezing	2.07252	0.20576	0.169	0.295
			No-freezing vs Sediment	3.56051	0.41592	0.295	0.295
			Pre-freezing vs Sediment	2.25713	0.31102	0.229	0.295
16S	0.6922	0.001	No-freezing vs Pre-freezing	10.3356	0.56369	0.006	0.009
			No-freezing vs Sediment	12.1022	0.63355	0.007	0.009
			Pre-freezing vs Sediment	12.5474	0.6419	0.009	0.009

936

937 Table 4: Pairwise PERMANOVA and pairwise PERMANOVA (eDNA index dataset) between
 938 all three protocols: no-freezing, pre-freezing and sediment pre- and no freezing water prior to
 939 filtration and sediment samples. P.adjusted is the adjusted p-value after FDR correction.

Primer	PERMANOVA		Pairwise PERMANOVA				
	R2	Pr(>F)	Comparison	F.Model	R2	p.value	p.adjusted
12S	0.2113	0.055	No-freezing vs Pre-freezing	1.25387	0.1355	0.297	0.297
			No-freezing vs Sediment	1.76131	0.18044	0.112	0.168
			Pre-freezing vs Sediment	1.69357	0.17471	0.077	0.168
16S	0.4417	0.001	No-freezing vs Pre-freezing	1.47905	0.15603	0.005	0.0135
			No-freezing vs Sediment	5.96537	0.42715	0.017	0.017
			Pre-freezing vs Sediment	6.51459	0.44883	0.009	0.0135

940

941

942 Figure legends

943 Figure 1: Photo of Topanga lagoon taken on August 22nd, 2018, a few weeks after collection.
944 There was no record of precipitation for the previous three months and the lagoon was closed to
945 the ocean by a sandbar. There was also no sign of recent waves topping over the sandbar and
946 reaching the lagoon.

947 Figure 2: Flowchart of the study design. Water was taken as a single sample in a 18 L water jug
948 from the middle of the lagoon (Fig. 1). In the laboratory, the water was split in batches to follow
949 either the centrifugation or the filtration protocol. Sediment samples were taken in 2 μ L
950 cryotubes at the same location where water was taken. Subsamples of these were then pooled in
951 triplicates prior to extraction (total 0.25-0.3 g). The dotted box indicates which part of the study
952 design is addressed in the main paper, and which is addressed in the supplemental material. The
953 black 'X' indicates the stage at which DNA extraction was performed and sequenced. Figures
954 were obtained from Biorender.

955 Figure 3: Venn diagrams of A) 12S and B) 16S primers showing the number of species found at
956 and between each protocol. Sediment samples showed the highest number of unique species for
957 both primers, although for the 12S dataset, about 28% are the result of contamination from tag-
958 jumping.

959 Figure 4: Species rarefaction curves based on sequencing effort for each protocol. A) 12S
960 primer; B) 16S primer. With the exception of the water samples for the 12S primer, none of the
961 curves have reached a plateau, although we expect the high diversity seen for the 12S sediment
962 samples be due to contamination from tag-jumping.

963 Figure 5: Confidence interval (CI) for slopes of rarefaction curves (Fig. 3) for each pairwise
964 comparison of the different protocols. Only the comparison between pre- versus no freezing
965 water samples, and pre-freezing versus sediment samples for the 12S primer (A) have come out
966 non significant. The remaining comparisons showed significant differences between rarefaction
967 slopes.

968 Figure 6: Plots of log₂fold change of families of bacteria and archaea (16S primer) for the
969 pairwise comparison between A) no freezing versus sediment; and B) pre-freezing versus
970 sediment. Circles are colored by phylum. Species present above zero are overrepresented in the
971 pre- or no freezing protocol, and species below the zero threshold are overrepresented in the
972 sediments.

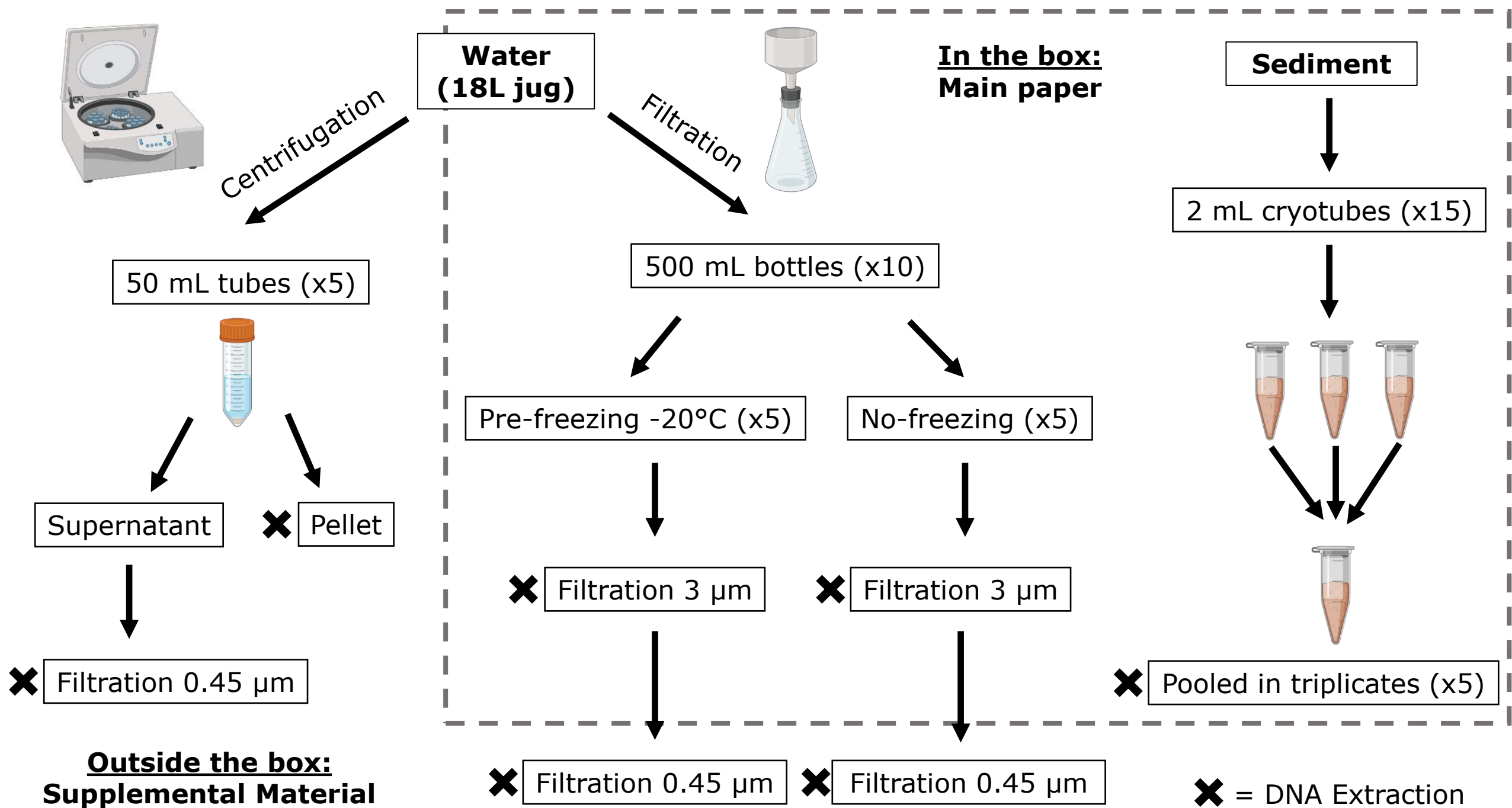
973 Figure 7: MW (left) and BA (right) plots for 12S primer dataset. In both plots, gray represents
974 features that are abundant, but not differentially abundant; black are rare, but not differentially
975 abundant. Diagonal dashed lines are shown for zero-intercept with slope of ± 1 .

976 Figure 8: MW (left) and BA (right) plots for 16S primer dataset. In both plots, gray represents
977 features that are abundant, but not differentially abundant; black are rare, but not differentially
978 abundant. Red represents features called as differentially abundant with $q < 0.1$. Diagonal dashed
979 lines are shown for zero-intercept with slope of ± 1 .

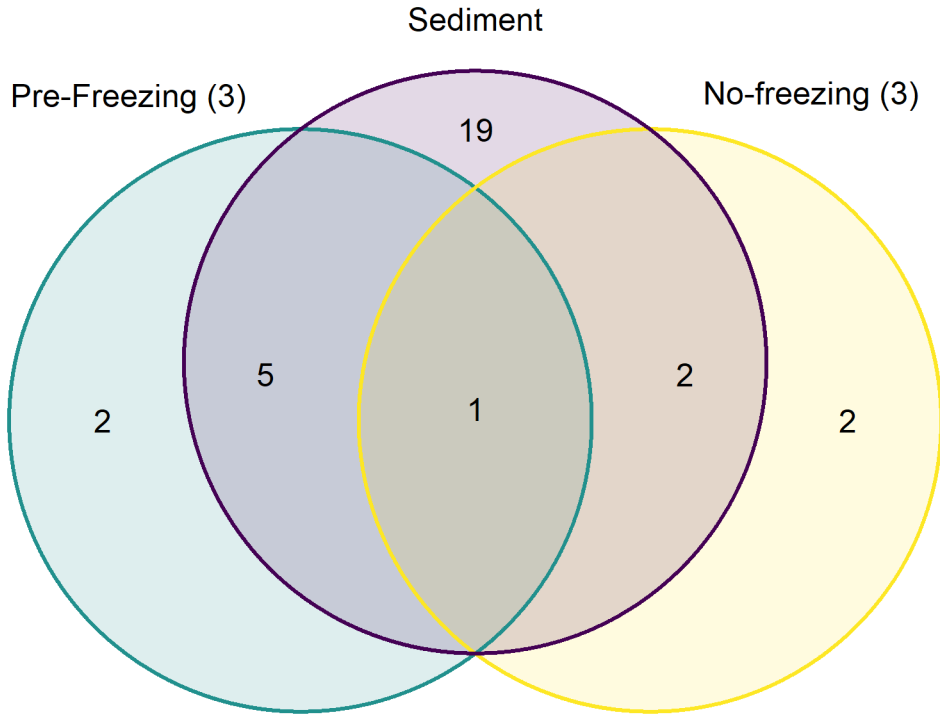
980 Figure 9: Constrained Analysis of Principal Coordinates (CAP) of rarefied A) 12S and B) 16S
981 primer datasets. Circles are colored by protocol. While there were no discernible differences in
982 assemblage composition for the 12S primer, in the 16S primer *Guillardia theta* is
983 overrepresented in the PF protocol in comparison to the NF protocol. *Eucyclogobius newberyii* is
984 underrepresented in the sediments compared to the filtration protocols, especially the NF
985 protocol.



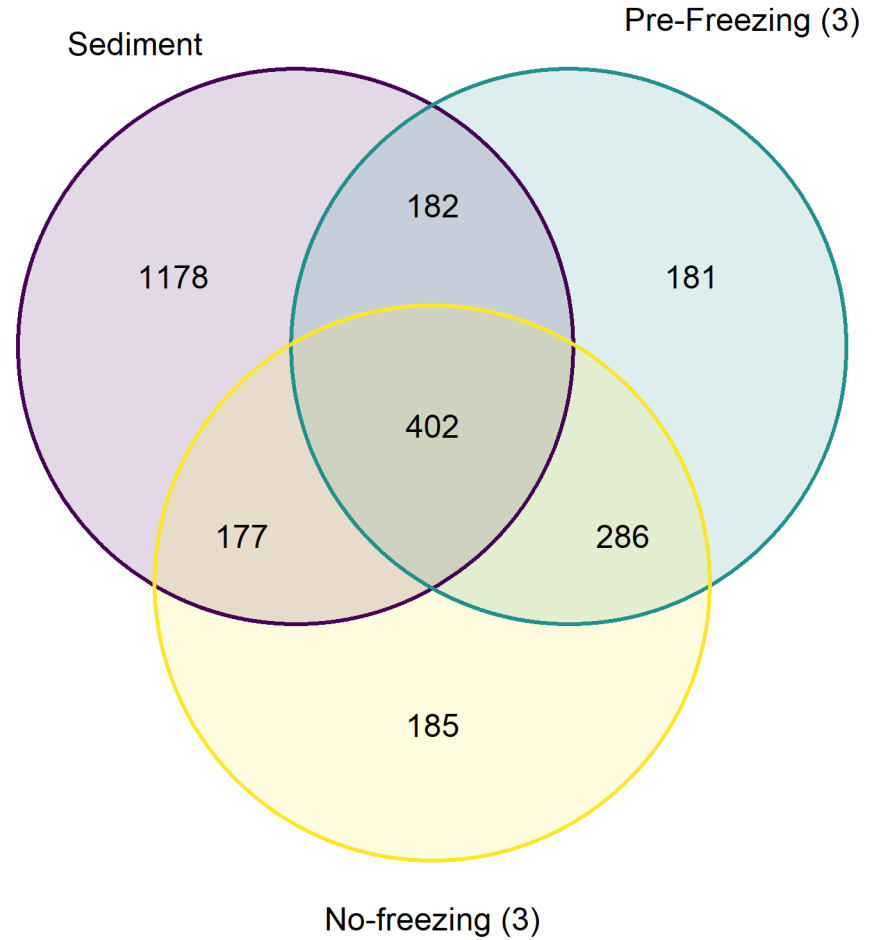
bioRxiv preprint doi: <https://doi.org/10.1101/2022.06.17.495388>; this version posted March 17, 2023. The copyright holder for this preprint (which was not certified by peer review) is the author/funder, who has granted bioRxiv a license to display the preprint in perpetuity. It is made available under aCC-BY-NC 4.0 International license.

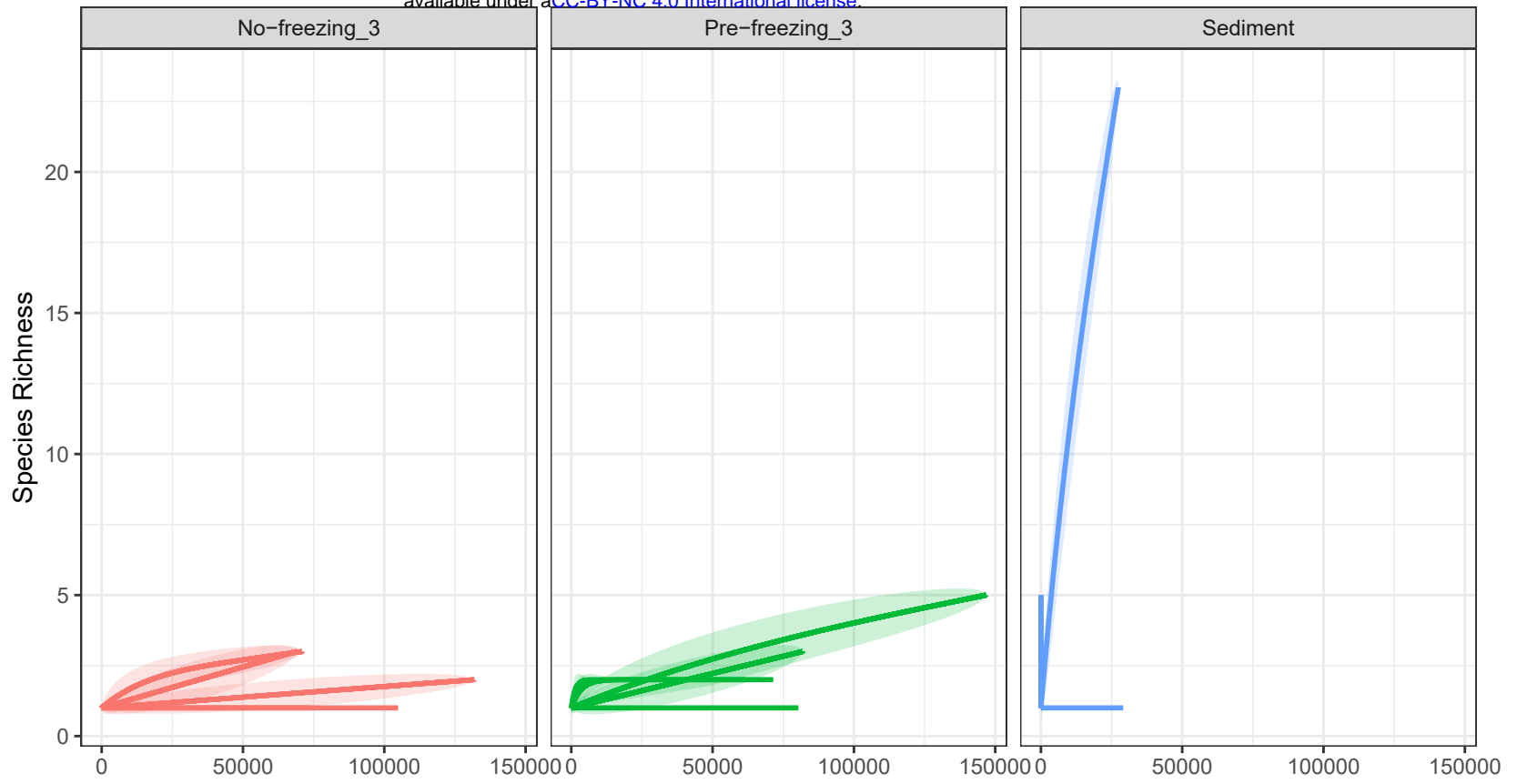


A) 12S

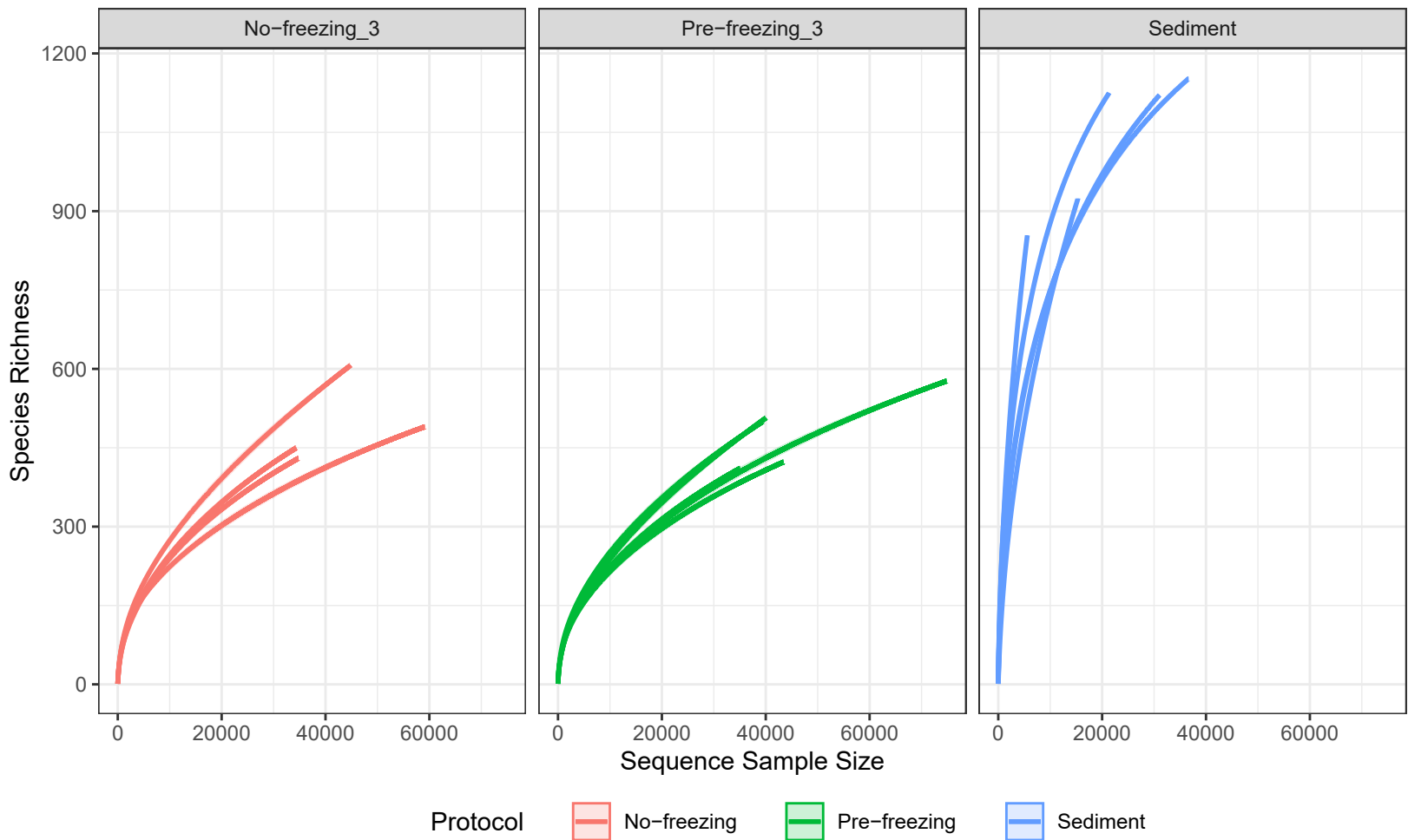


B) 16S

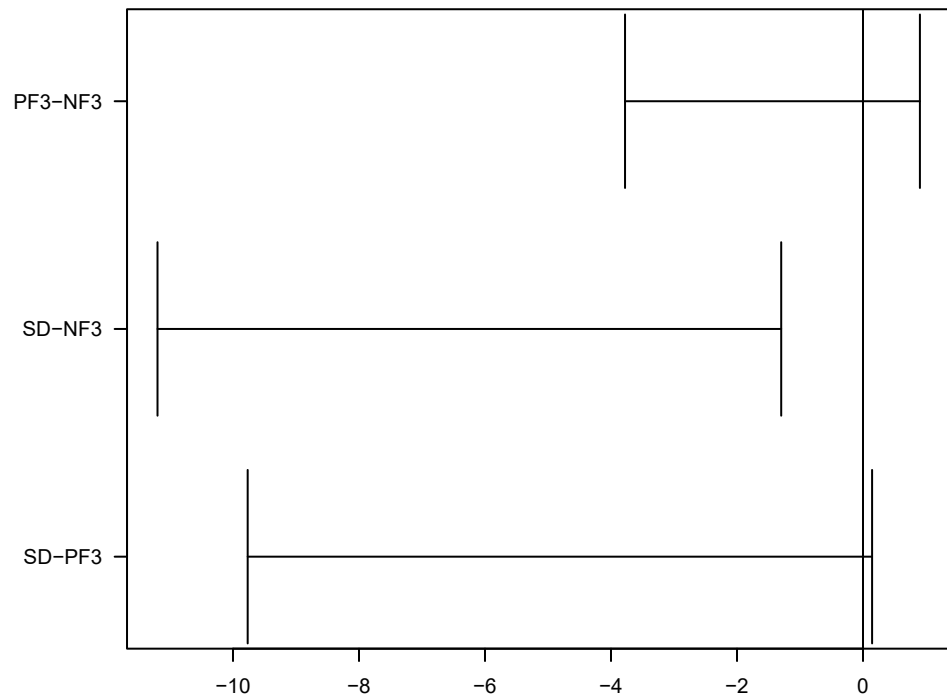




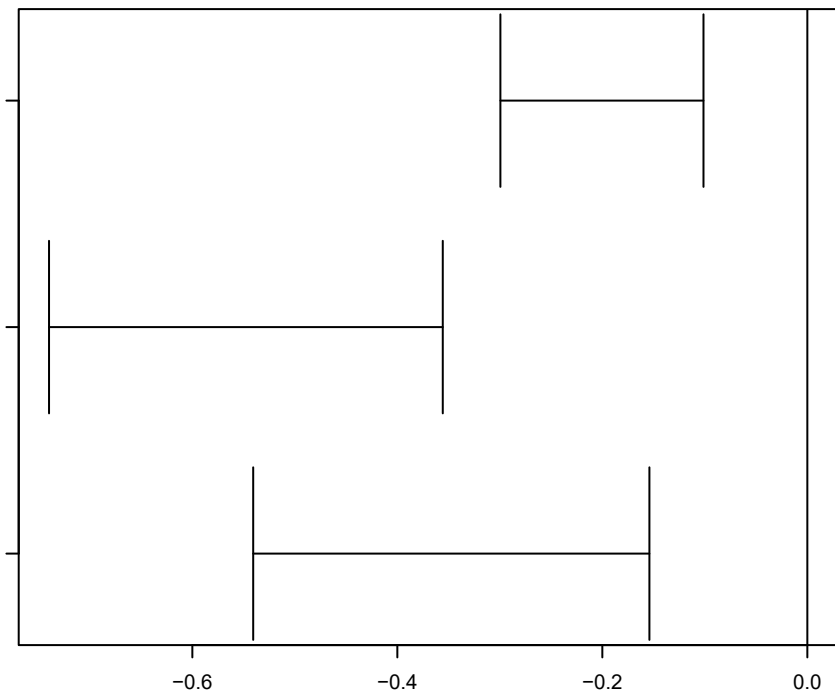
B) 16S



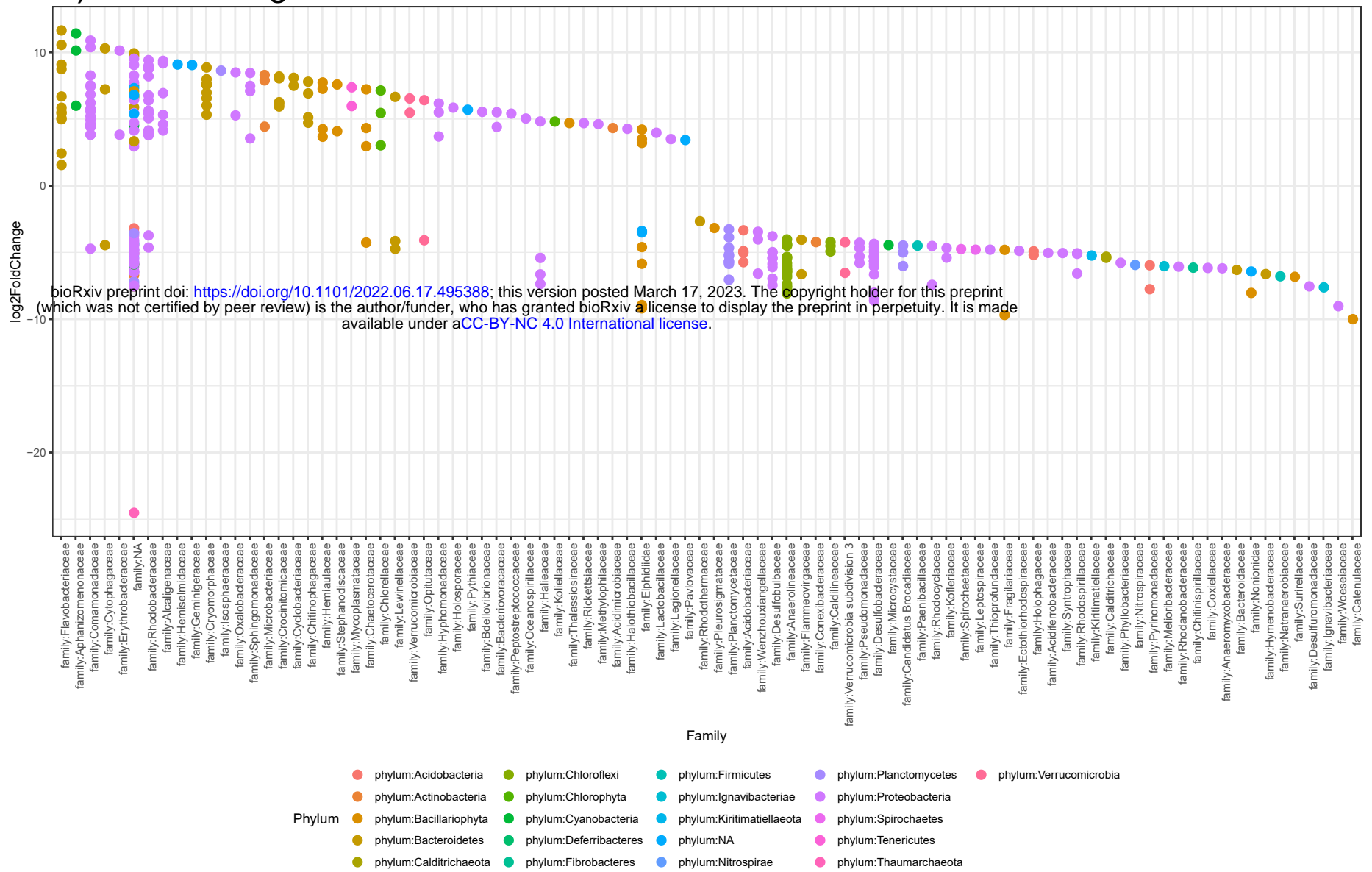
A) 12S



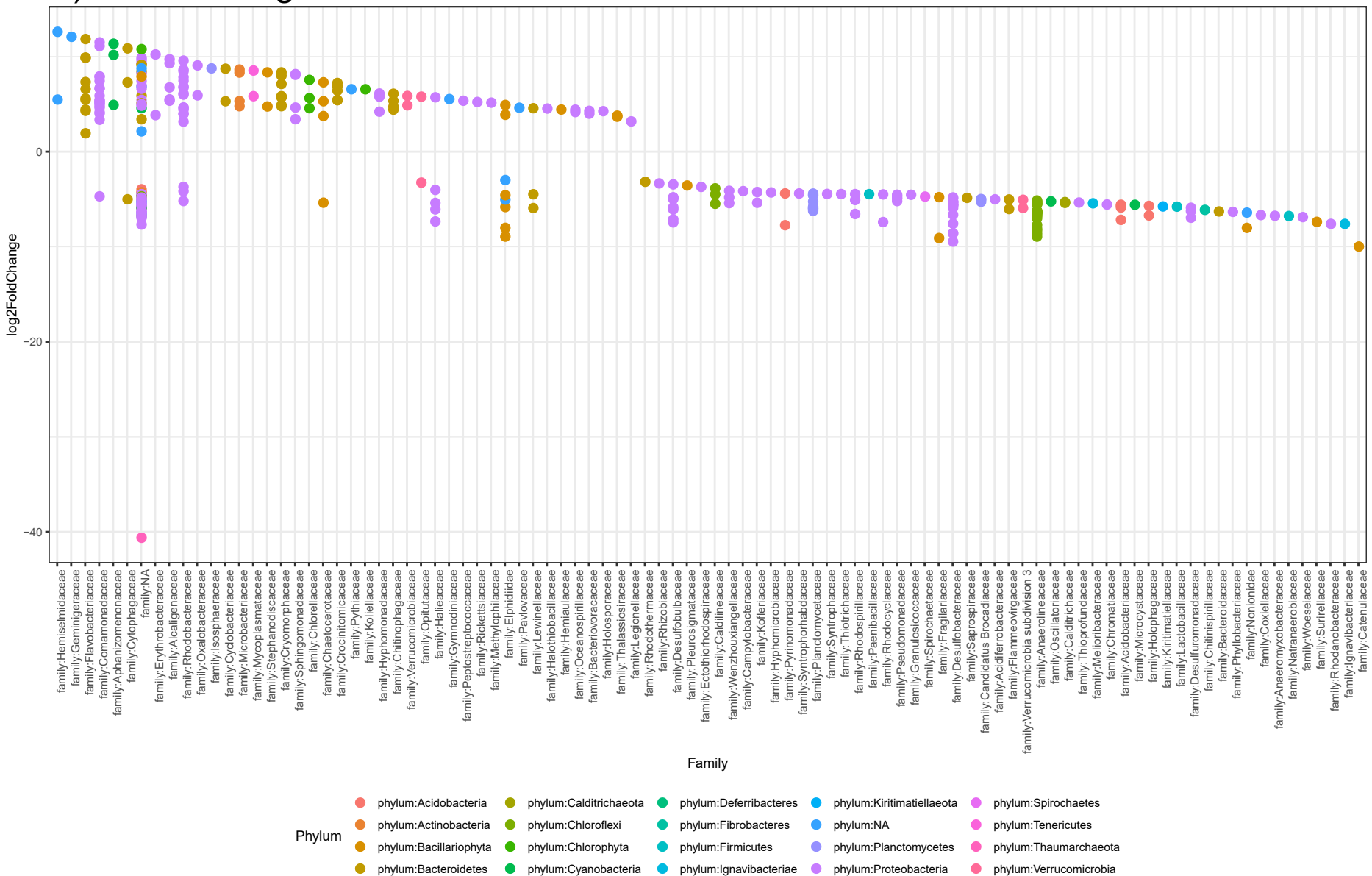
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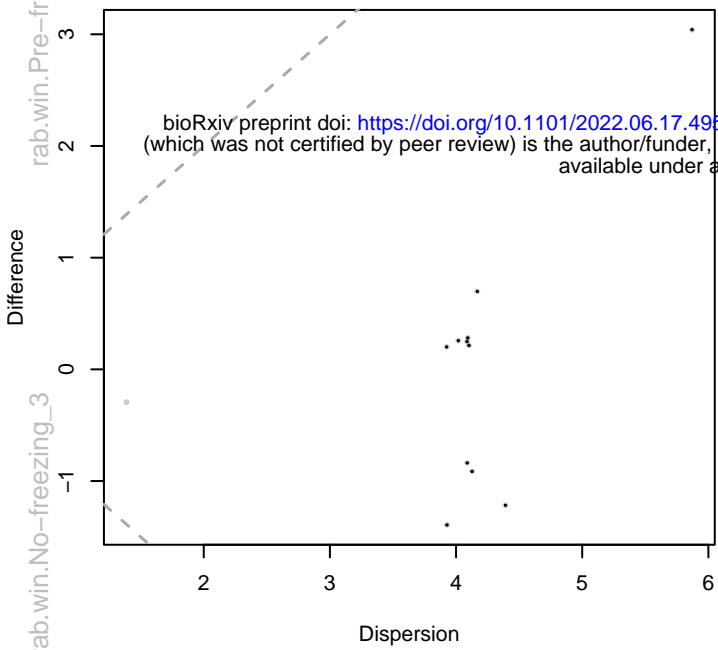
A) No-freezing Vs Sediment



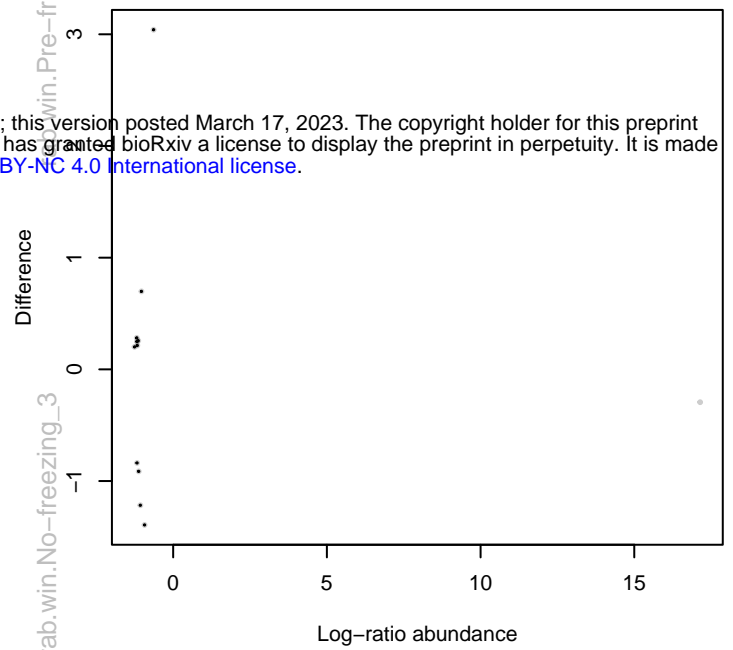
B) Pre-freezing Vs Sediment



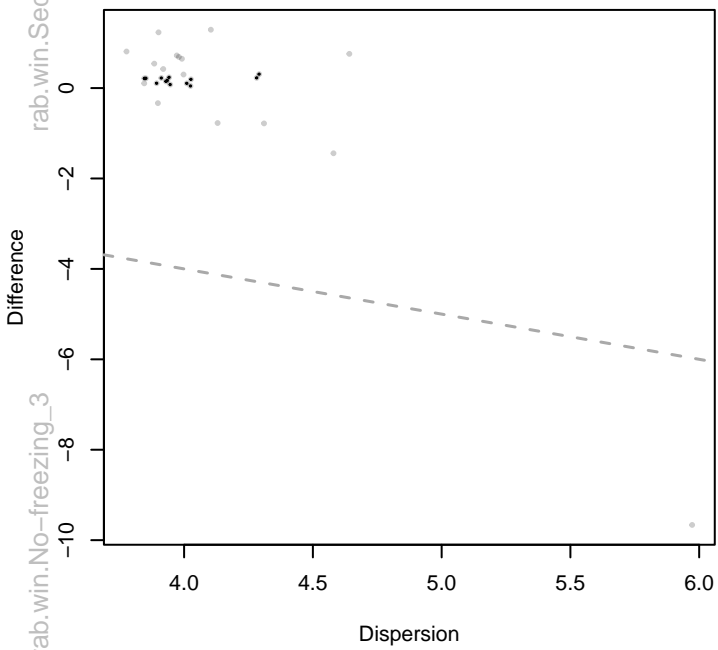
(Pre Vs No freezing) MW Plot



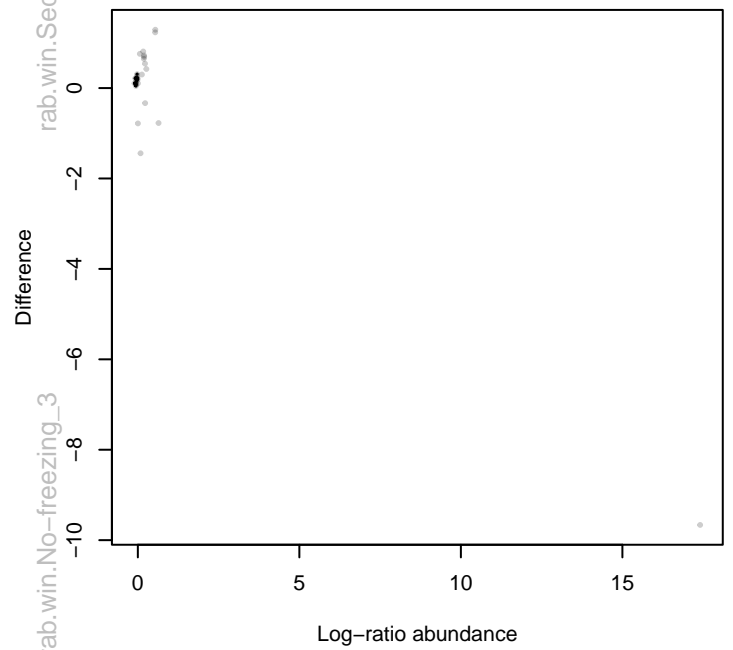
(Pre Vs No freezing) MA Plot



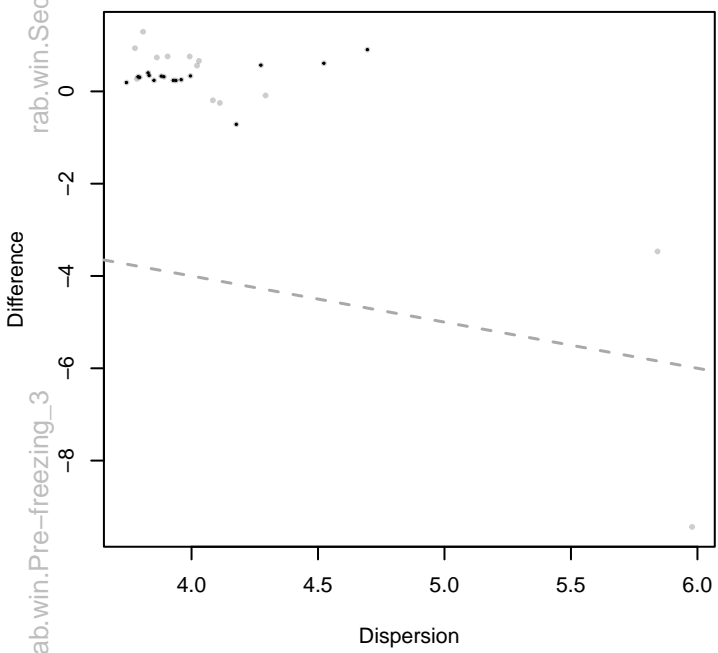
(No freezing Vs Sediment) MW Plot



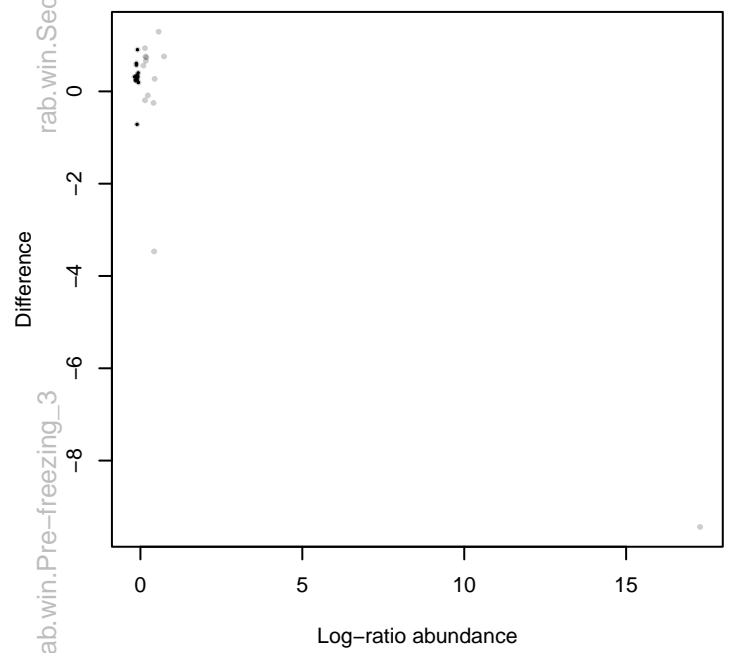
(No freezing Vs Sediment) MA Plot



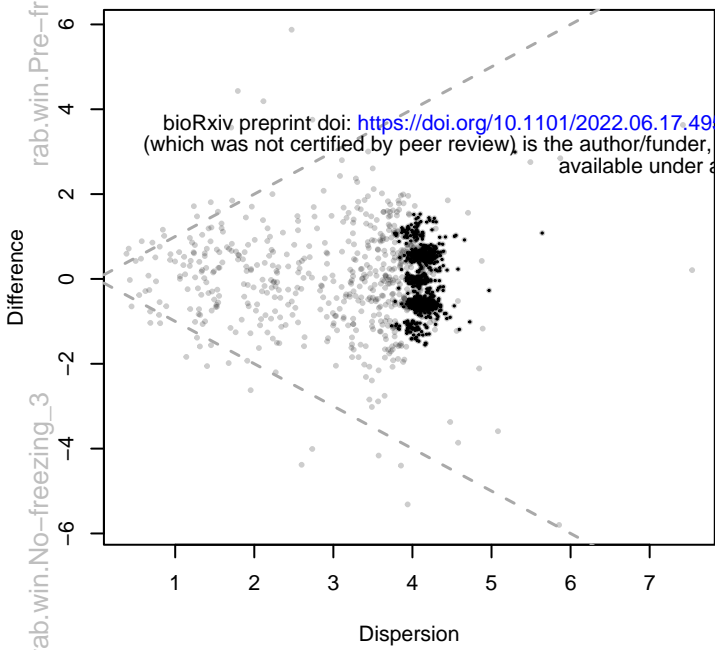
(Pre-freezing Vs Sediment) MW Plot



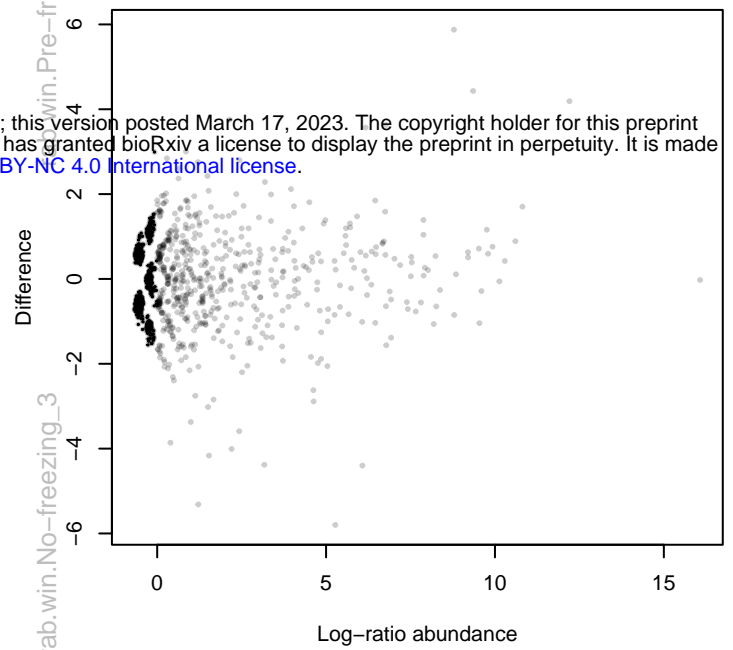
(Pre-freezing Vs Sediment) MA Plot



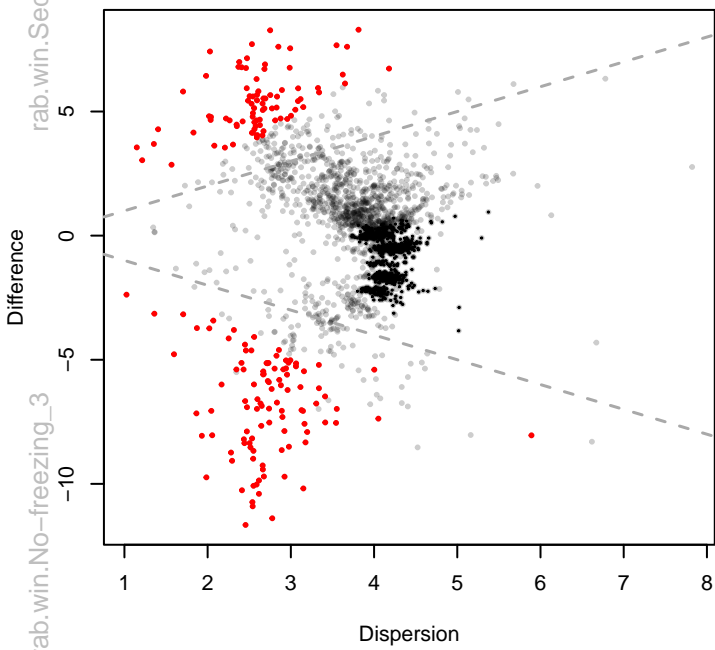
(Pre Vs No freezing) MW Plot



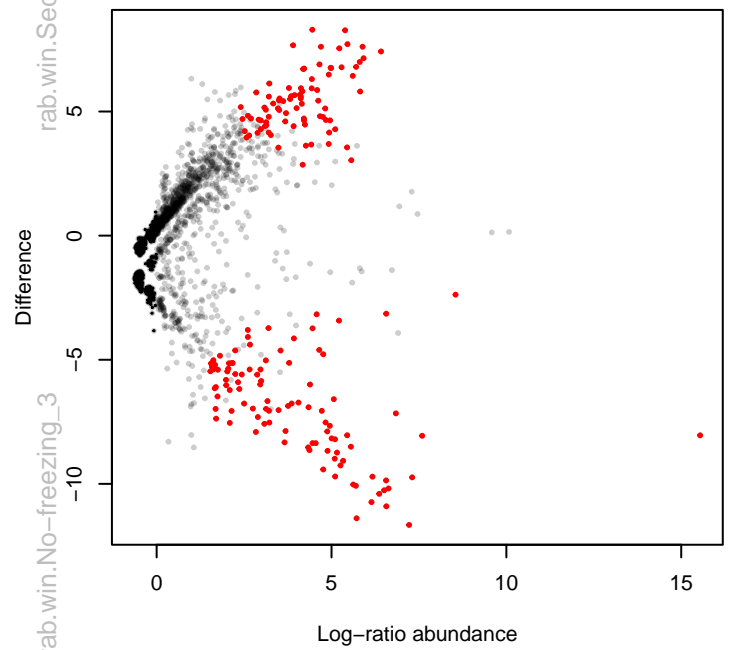
(Pre Vs No freezing) MA Plot



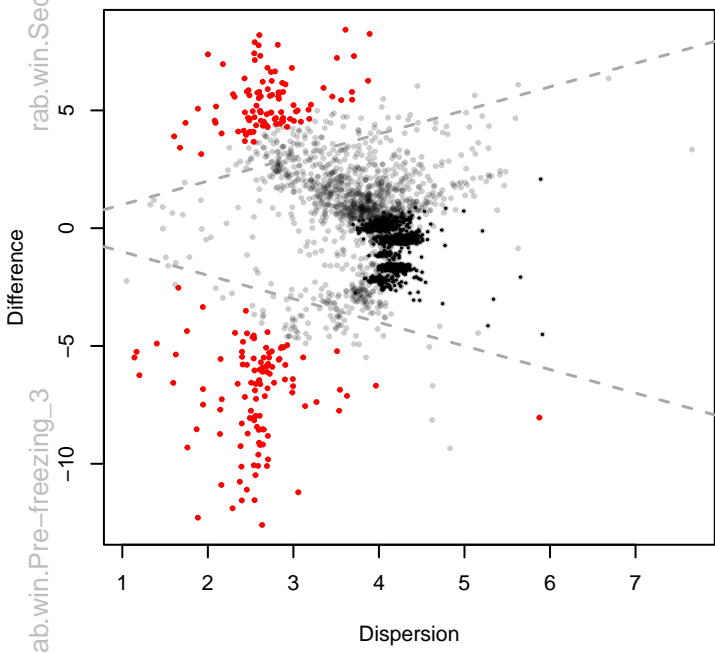
(No freezing Vs Sediment) MW Plot



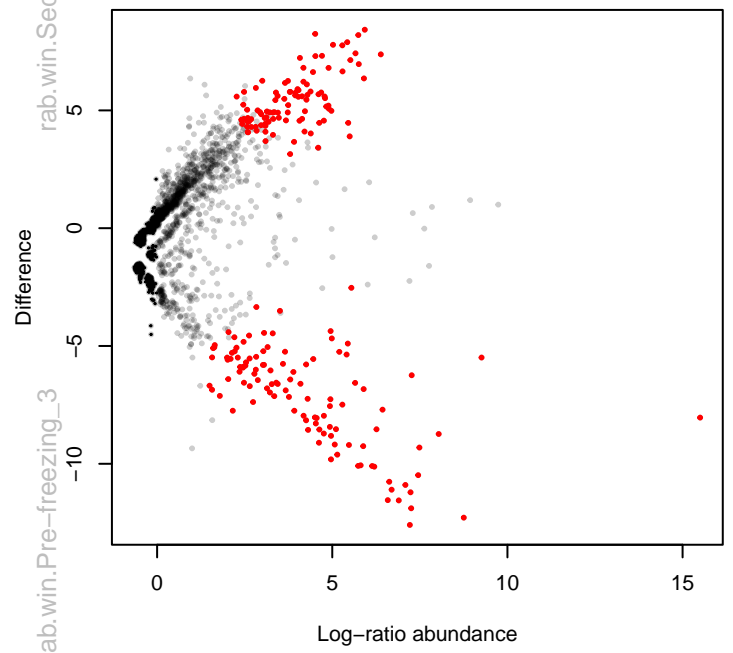
(No freezing Vs Sediment) MA Plot



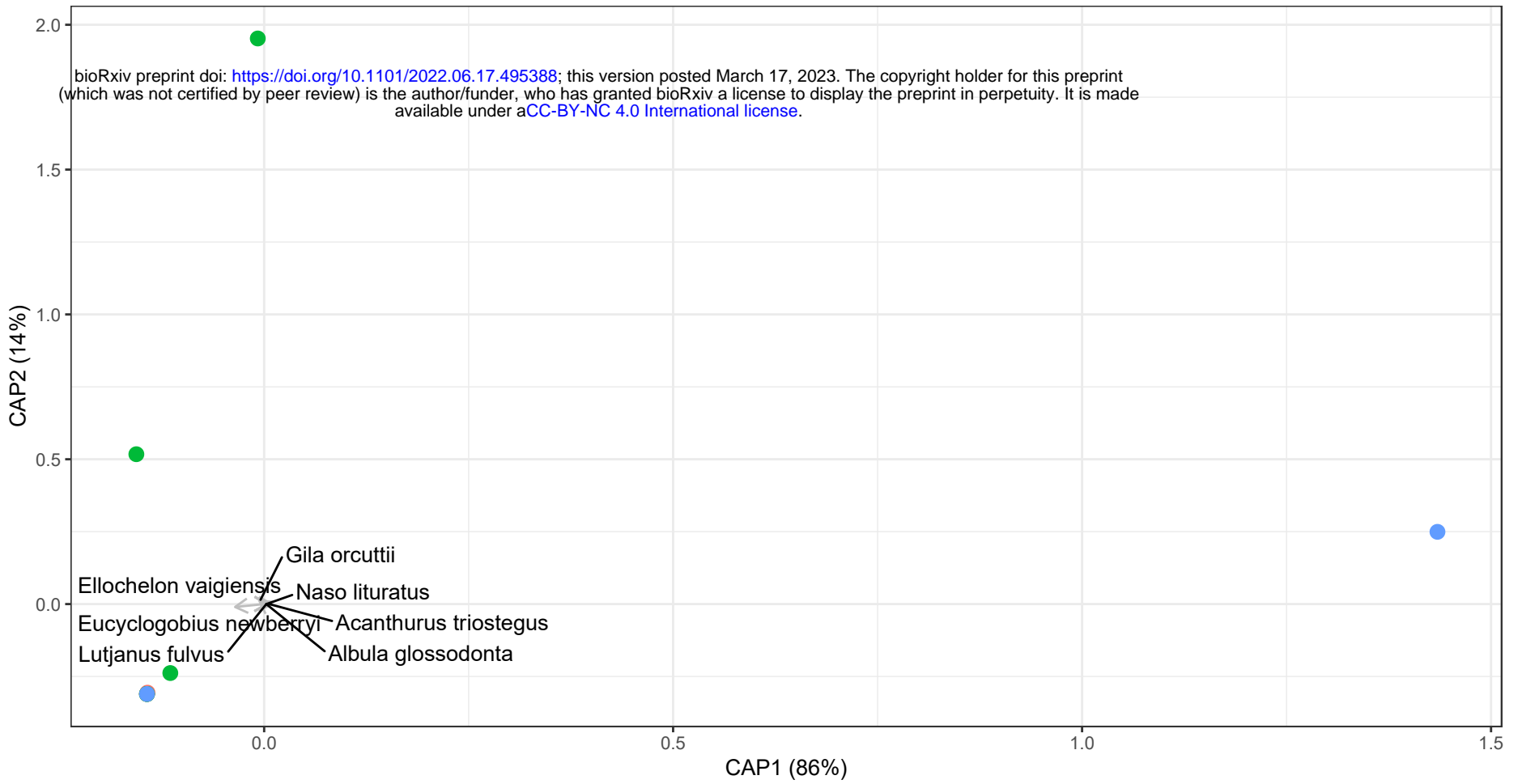
(Pre-freezing Vs Sediment) MW Plot



(Pre-freezing Vs Sediment) MA Plot



A) 12S



B) 16S

