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Defining the sediment microbiome of the Indian River Lagoon, FL, USA, an Estuary of National Significance --Manuscript Draft--

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Abstract:	The Indian River Lagoon, located on the east coast of Florida, USA, is an Estuary of National Significance and an important economic and ecological resource. The Indian River Lagoon faces several environmental pressures, including freshwater discharges through the St. Lucie Estuary; accumulation of a anoxic, fine-grained, organic-rich sediment; and metal contamination from agriculture and marinas. Although the Indian River Lagoon has been well-studied, little is known about its microbial communities; thus, a two-year 16S amplicon sequencing study was conducted to assess the spatiotemporal changes of the sediment microbiome. In general, the Indian River Lagoon exhibited a microbiome that was consistent with other estuarine studies. Statistically different microbiomes were found between the Indian River Lagoon and St. Lucie Estuary due to changes in porewater salinity causing microbes that require salts for growth to be higher in the Indian River Lagoon. The St. Lucie Estuary exhibited more obvious microbial seasonality, such as higher Betaproteobacteriales, a freshwater associated organism, in wet season and higher Flavobacteriales in dry season samples. Distance-based linear models revealed these microbiomes were more affected by changes in total organic matter and copper than changes in temperature. Anaerobic organisms, such as Campylobacteriales, were more associated with high total organic matter and copper samples while aerobic organisms, such as Nitrosopumilales, were more associated with low total organic matter and copper samples. This initial study fills the knowledge gap on the Indian River Lagoon microbiome and serves as an important baseline for possible future changes due to human impacts or environmental changes.
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1 Defining the sediment **microbiome** of the Indian River Lagoon, FL, USA, an Estuary of National
2 Significance

3

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14

15 **Abstract**

16 The Indian River Lagoon, located on the east coast of Florida, USA, is an Estuary of
17 National Significance and an important economic and ecological resource. The Indian River
18 Lagoon faces several environmental pressures, including freshwater discharges through the St.
19 Lucie Estuary; accumulation of a anoxic, fine-grained, organic-rich sediment; and metal
20 contamination from agriculture and marinas. Although the Indian River Lagoon has been well-
21 studied, little is known about its microbial communities; thus, a two-year 16S amplicon
22 sequencing study was conducted to assess the spatiotemporal changes of the sediment
23 microbiome. In general, the Indian River Lagoon exhibited a microbiome that was consistent
24 with other estuarine studies. Statistically different microbiomes were found between the Indian
25 River Lagoon and St. Lucie Estuary due to changes in porewater salinity causing microbes that
26 require salts for growth to be higher in the Indian River Lagoon. The St. Lucie Estuary exhibited
27 more obvious microbial seasonality, such as higher Betaproteobacteriales, a freshwater
28 associated organism, in wet season and higher Flavobacteriales in dry season samples. Distance-
29 based linear models revealed these microbiomes were more affected by changes in total organic
30 matter and copper than changes in temperature. Anaerobic organisms, such as
31 Campylobacterales, were more associated with high total organic matter and copper samples
32 while aerobic organisms, such as Nitrosopumilales, were more associated with low total organic
33 matter and copper samples. This initial study fills the knowledge gap on the Indian River Lagoon
34 microbiome and serves as an important baseline for possible future changes due to human
35 impacts or environmental changes.

36

37 **Introduction**

38 The Indian River Lagoon (IRL) is an Estuary of National Significance located on
39 Florida's east coast (USA) [1]. The lagoon has a total estimated annual economic value of \$7.6
40 billion [2]. It is connected, at its southern end, to the St. Lucie Estuary (SLE), another important
41 resource for the area [3]. The IRL has a high biodiversity because it is located at the border
42 between temperate and sub-tropical regions, allowing it to have plant and animal species from
43 both climates [4]. The IRL faces similar environmental issues to other estuaries, including
44 freshwater inputs, eutrophication, organic matter, and metal contamination [1,5].

45 Freshwater is introduced into the IRL via runoff from local waterways and discharges
46 from Lake Okeechobee, which are diverted into the SLE through the C-44 canal during periodic
47 releases based upon the Lake's water level [3]. This introduction of freshwater and its associated
48 contaminants causes problems for the ecosystem [3] and also introduces dissolved organic
49 material and plant matter that settles into the sediment to become the fine grained, highly organic
50 sediment known as "IRL muck" [6]. "IRL muck" (hereinafter referred to as muck) is defined as
51 sediment that has at least 75% water content, and the remaining solids fraction has at least 60%
52 fines and 10% total organic matter (TOM) [7]. Muck can lead to various negative ecological
53 impacts including nutrient flux in the water column triggering algal blooms and turbidity which
54 damages seagrasses by blocking sunlight [8]. About 10% of the IRL is covered in muck ranging
55 in depths from centimeters to several meters [1,7,9]. Anoxia is associated with muck and can
56 also be caused by freshwater discharges carrying contaminants from agriculture and urban
57 development [7,10,11]. A shift to an anoxic state alters the most energetically favorable terminal
58 electron acceptors for microbes, altering their population, diversity, and functions [10,11]. A
59 study of Chesapeake Bay (MD, USA), compared the microbiomes of anoxic, organic-rich, silty-

60 clay sediments to organic-poor, sandy sediments and found major differences due to the former
61 having microbial members that contribute to the high sulfide and methanogenic conditions in the
62 area [12].

63 Other pressures on the IRL include contaminants such as trace metals which can be
64 transported in discharges and runoff as part of metal-dissolved organic matter complexes that
65 precipitate onto the sediment once this freshwater meets the brackish water of the IRL [6,7,13].
66 A survey of the northern IRL found several sites with metals above normal levels, while a survey
67 in the SLE found a large accumulation of phosphorus and Cu, the latter likely due to Cu-
68 containing fungicides or cuprous oxide anti-fouling paints used in marinas [14–17]. The
69 interaction of microbes with heavy metals affects their chemical forms and therefore their
70 solubility, mobility, bioavailability, and toxicity [18]. In turn, prokaryotic assemblages can be
71 altered by the presence of heavy metals, which can lead to a decrease in microbial diversity and
72 functional redundancy [5,19–22]. A recent study in Chile, found that there was a significant
73 decrease in the abundance of bacteria in copper contaminated sites, while the abundance of
74 archaea was similar to a less contaminated site, likely due to copper resistance mechanisms [23].

75 The true extent of the IRL’s biodiversity cannot be understood without information on its
76 microbial communities [24]. Sediment microbes, especially in estuaries, face a wide range of
77 physicochemical gradients that can cause shifts in the microbial taxonomy as well as microbial
78 functional capabilities [13,25,26]. This study was carried out to provide the first data on the
79 microbial communities present in the IRL and to explore potential diversity changes due to
80 differences between: the IRL and SLE; samples with zero muck characteristics (0MC) and three
81 muck characteristics (muck or 3MC); and non-Cu contaminated and Cu-contaminated, high
82 TOM samples.

83

84 **Materials and methods**

85 **Site selection**

86 Sites were chosen based upon either being adjacent to continuous water quality monitoring
87 stations [27,28] or known to be muck [Manatee Pocket (MP) and Harbor Branch Channel (HB)]
88 or sandy (Jupiter Narrows and Hobe Sound) sites (Fig 1 and S1 Table). During the second year
89 of sampling, two additional marina sites [Harbortown Marina (HT) and Vero Beach Marina] and
90 two nearby less impacted sites (Barber Bridge and Round Island) were added. Using information
91 from NOWData (National Weather Service), the average monthly temperature and average
92 monthly rain sum from years 1990-2018 was obtained for the Melbourne Area, Fort Pierce Area,
93 and Stuart 4 E Stations [29]. Streamflow data was taken from DBHYDRO (South Florida Water
94 Management District) and from the United States Geological Services website [30,31].

95

96 **Fig 1. Sampling area map.** The sampling area with stars indicating the site locations, triangles
97 the location of the National Weather Service (NWS) [29] monitoring stations in Melbourne
98 (MB), Fort Pierce (FP), and Stuart (ST), and circles the inflow gauge locations [United States
99 Geological Service (USGS) [31] in yellow and the South Florida Water Management District
100 (SFWMD) [30] in red]; the streams or canals associated with these gauges are denoted by blue
101 lines. In Maps A and B black stars are the North Indian River Lagoon (IRL) sites: Merritt Island
102 Causeway (MI), Melbourne Causeway (MC) and the Sebastian Inlet (SI). Blue stars (maps A and
103 C) are the North Central IRL sites: Vero Beach Marina (VM), Barber Bridge (BB), Vero Beach
104 (VB), and Round Island (RI). Red stars (maps A and C) are the South Central IRL sites: Harbor
105 Branch Channel (HB), Linkport (LP), Fort Pierce (FP), and Harbortown Marina (HM). Green

106 stars (maps A and D) are the South IRL sites: Jensen Beach (JB), Manatee Pocket (MP), Jupiter
107 Narrows (JN) and Hobe Sound (HS). Maroon stars (maps A and D) are the St. Lucie Estuary
108 (SLE) sites: North Fork (NF), South Fork (SF), Middle Estuary (ME), and South Fork 2 (ST).
109 GPS coordinates for all study sites and environmental monitoring stations are located in S1
110 Table.

111

112 **Sample collection**

113 A total of 204 sediment samples were taken during four sampling periods over a two-year period,
114 with two sets of samples during the wet (May-Oct) and dry seasons (Nov-Apr). The samples
115 were taken during the following months: Aug-Sep 2016 (W16), Mar-Apr 2017 (D17), Oct-Nov
116 2017(W17), and Apr 2018 (D18). Samples were collected with published methods [7,21]. A
117 Wildco® Ekman stainless steel bottom grab sampler was deployed from a boat to collect
118 sediment in triplicate. The first replicate's sediment temperature was determined with a
119 thermometer. The top 2 cm of each replicate was subsampled with an ethanol-sterilized plastic
120 spoon. Three sub-samples of each replicate were taken to assess the microbial community (A),
121 dry sediment characteristics (B), and wet sediment characteristics (C). They were sealed and
122 placed on ice until returned to the lab. Subsamples A and B were collected in sterile 50-mL
123 Falcon tubes and 75-mL polystyrene snap-cap vials, respectively, and stored at -20° C. The
124 remnants of the top 2 cm of the sediment were placed in a double-bagged Ziploc freezer bag as
125 Subsample C and stored at 4° C. When ready for analysis, subsample B was thawed and dried for
126 48 hours at 60 °C on pre-weighed, acid-washed, glass petri dishes. The differences in pre- vs
127 post-drying weights were used to determine water content. The dried sediment was broken up
128 with an acid-washed mortar and pestle and sieved to remove the fraction above 2 mm (coarse)

129 from the sample. The remaining sand (2 mm-0.063 mm) and fines (<0.063 mm) fractions were
130 kept for further analysis. All collection plasticware was soaked in 5% HNO₃ for a minimum of
131 24 hours and then rinsed three times in 18.2 MΩ deionized water.

132 **Metal analysis**

133 Triplicate samples at each site were analyzed for heavy metals. Acid digests were prepared by
134 modification of methods described in three studies [7,32,33]. Briefly, 1 g (+/- 0.0003g) of dried
135 and sieved Subsample B was digested in 1 M HCl for one hour at 30° C with shaking at 150 rpm.
136 Digests were filtered with DigiTubes® and DigiFilters® (0.45 μm) (SCP Sciences, Champlain,
137 NY). Cu and Fe were measured with a four-point calibration curve on a Perkin Elmer 4000
138 atomic absorption spectrometer (Perkin Elmer, Waltham, MA). The calibration curve was
139 rechecked every ten to twelve samples to account for absorbance drift. Reagent samples (2%
140 HNO₃, 1 M HCl, 18.2 MΩ deionized water) and a method control sample were analyzed to
141 check for contamination.

142 **Sediment physical characteristics**

143 A modified procedure was used to determine sediment characteristics [7]. One gram of dried and
144 sieved Subsample B sediment was heated in a 550° C muffle furnace for four hours to burn off
145 the organic matter. The sediment weight loss was calculated and reported as percent TOM. Grain
146 size was determined by wet sieving 10-30 g of Sample C and drying to constant weight. The
147 gravel, sand, and fines percentages of the total dry weight were determined. PWS was
148 determined by centrifuging 20 g of Sample C and measuring the salinity of the resulting liquid
149 with a portable refractometer.

150

151 **Sequence analysis**

152 DNA was extracted from 0.25-0.3 g of sediment with the Qiagen PowerSoil DNA Isolation Kits
153 (Hilden, Germany), and its quality checked with a Nanodrop 2000 (Oxford Technologies,
154 Oxford, UK). Samples were sent to Research and Testing (Lubbock, TX, USA) for MiSeq 16S
155 sequencing to amplify the bacterial/archaeal 16S V4 region with the modified primers used by
156 the Earth Microbiome Project of 515F (GTGYCAGCMGCCGCGGTAA) and 806R
157 (GGACTACNVGGGTWTCTAAT) [34,35]. The raw sequences were trimmed to remove the
158 primers and quality-filtered with the FastX and TrimGalore programs respectively [36,37].
159 Quantitative Insights Into Microbial Ecology 2's (QIIME2) [38] standard 16S workflow was
160 used for analysis, and a Snakemake file was used for the orchestration for reproducibility
161 [39,40]. Sequences were joined with VSEARCH [41]. Next, they were denoised with Deblur
162 [42] run with default parameters, with the exceptions of the minimum reads parameter set to 0 to
163 account for metadata categories with smaller sample sizes and trim length set to 232 bases.
164 Amplicon Sequence Variants (ASVs) were annotated with a Naïve-Bayes classifier based on the
165 scikit-learn system and the SILVA database [43,44] (version 132). Mitochondrial, chloroplast
166 and unassigned sequences were filtered from the samples. The ASVs were aligned with MAFFT
167 [45] and then masked [38,46] to make a phylogenetic tree with FASTTREE [47] that was then
168 midpoint-rooted. Raw sequences have been uploaded into the National Center for Biotechnology
169 Information Sequence Read Archive (PRJNA594146) [48].

170 **Statistical analysis**

171 RStudio [49,50] (R Version 3.6.1) was used for data manipulation, visualization, generation of
172 alpha diversity statistics (Shannon), and data manipulation. Analyses were run with the following
173 library versions: phyloseq (1.28.0) [51], vegan (2.5-5) [52], ggplot2 (2_3.2.0) [53], reshape

174 (0.8.8) [54], tidyverse (1.2.1) [55], and FSA (0.8.25) [56]. ASVs that did not have at least ten
175 sequences associated with them across all samples were removed [57,58].

176 PRIMER7/PERMANOVA+ was also used to analyze the data [59–61]. The
177 environmental data was checked for highly colinear variables, greater than 0.70 [62], by
178 generating draftsman plots. This showed that TOM was positively colinear with water content,
179 percent fines, and Fe, and negatively colinear with percent sand. This allowed TOM to represent
180 all these variables in future analyses. The remaining environmental variables were normalized.
181 The biological data was square root transformed, then used to make a Bray-Curtis dissimilarity
182 matrix to create principal coordinates of analyses. Distance-based linear models were made with
183 a stepwise selection procedure, an AICc (An Information Criterion) selection criteria, 9 999
184 permutations, marginal tests, and a distance-based redundancy analysis plot [21]. Overall and
185 pair-wise permutational analysis of variance were conducted with 9 999 permutations, the
186 unrestricted method, Type III Sum of Squares, and Monte Carlo p-values [5,63]. Overall
187 statistical significance of environmental data and alpha diversity metrics were determined with
188 Kruskal-Wallis or Mann Whitney U tests for categories with greater than two or just two
189 subcategories, respectively [64,65]. Pairwise testing was conducted with the Dunn method [66].
190 All reported p-values were considered statistically significant if less than 0.05 after multiple
191 testing correction with the Benjamini-Hochberg (BH) method [67]. The Snakemake file used for
192 QIIME2 analysis and any subsequent scripts used in statistical analysis can be found on Github
193 (https://github.com/djbradshaw2/General_16S_Amplicon_Sequencing_Analysis) [68]. Measured
194 environmental data and metadata can be found in S2 and S3 Tables, respectively.

195

196

197 **Results**

198 **Weather and streamflow discharges**

199 Measured air temperatures were higher during each of the sampling periods than historical
200 temperatures (1990-2018), with W16 being the hottest, followed by W17, D18, and D17 (S1
201 Fig). Every sampling period besides D17, which was drier than usual, was wetter than usual
202 especially W17, for which rainfall more than doubled. Streamflow discharges matched this data
203 with the highest streamflow occurring during W17, especially at the C44 canal leading to the
204 South Fork 2 site (S4 Table).

205 **Porewater salinity and sediment temperature**

206 Porewater salinity (PWS) and sediment temperature were measured to assess changes between
207 sampling periods (Fig 2). Dunn testing indicated that IRL W16 and W17 sampling periods were
208 significantly different (BH p-values < 0.05) from each other as well as both the D17 and D18
209 periods, although these two were significantly similar (BH p-value = 0.16) (Fig 2A and S5
210 Table). In the SLE, W16 and W17 were significantly similar to one another (0.65) but different
211 from D17 and D18, which were also statistically similar to one another (0.23). The highest mean
212 sediment temperature occurred during the W16, and Dunn testing revealed that all sampling
213 periods were significantly different from one another except for the D17 and W17 temperatures
214 for both the IRL (0.60) and SLE (0.052).

215

216 **Fig 2. Porewater salinity and sediment temperature patterns.** Porewater salinity (PWS) (left)
217 and sediment temperature (right) by Estuary by Sampling Period (A) and by Location (B) for the
218 Indian River Lagoon (IRL) and St. Lucie Estuary (SLE). Bars denote largest and smallest values

219 within 1.5*the interquartile range, middle line is the median, ends of boxes are the first and third
220 quartiles. The letters on top of each boxplot denote the results from the pairwise Dunn test with
221 different letters denoting statistical significance (Benjamini-Hochberg adjusted p values < 0.05).
222 In A the letters show how each of the sampling periods were different within each estuary but do
223 not denote inter-estuary comparisons.

224

225 PWS generally increased towards the southern IRL while the SLE had the highest
226 interquartile range, but the lowest mean (Fig 2B). All sections of the IRL were significantly
227 different from the SLE, while only the North IRL sites were found to be statistically lower than
228 the South IRL sites (BH p-value = 0.045) (Fig 2B and S5 Table). Sediment temperature did not
229 vary greatly across locations, ranging from the highest mean of 26.5 °C (South) to the lowest of
230 24.4 °C (South Central). Each of the Location subcategories were not statistically different from
231 one another, except for the South Central IRL being significantly lower than the South
232 (0.00037).

233 **Muck and copper**

234 The sites that, on average, exceeded three muck characteristics were Middle Estuary and South
235 Fork, while those that exceeded at least one of the thresholds were HB, HT, Melbourne
236 Causeway, and MP (Fig 3). None of the other 13 sites exceeded the thresholds on average. Out
237 of the 204 samples, 40 were considered muck since their sediment characteristics exceeded three
238 thresholds (3MC), 14 only exceeded two thresholds, 10 samples exceeded one, and 140 exceeded
239 none (0MC) (S3 Table). 3MC samples (water content =81%, TOM = 24%, and silt/clay
240 percentage = 81%) had 2.4x higher water content, 7.2x higher TOM, and 9.6x higher silt/clay
241 percentages on average than 0MC samples (3.3%, 34%, 8.4%)

242

243 **Fig 3. Muck characteristics by Site.** Bar graph, with error bars denoting standard error,
244 summarizing the average muck characteristics associated with each site. Black bars represent
245 water content. The black dotted line denotes the percentage water content (75% [7]) that a site, if
246 it exceeded all three thresholds, could be considered muck. Percent silt/clay is represented by the
247 dark gray bars with the dark gray dotted line representing the 60% [7] threshold. Total organic
248 matter is represented by the light gray bars with the light gray dotted line representing the 10%
249 [7] threshold.

250

251 A sample was considered to have high TOM if it exceeded 10% and high Cu if it
252 exceeded 65 $\mu\text{g/g}$ [7,69]. Most samples that exceed at least one of the muck characteristics also
253 had high TOM (62/64) (Fig 4 and S3 Table). The sites that had samples with both high TOM and
254 high Cu (HiHi) included HB, MP, and HT, whereas the sites with samples that had high TOM
255 but low Cu (HiLo) included Middle Estuary, South Fork, Melbourne Causeway, HT, Linkport,
256 and South Fork 2. Only two samples, both from HB, were classified as having low TOM and
257 high Cu (LoHi). The remaining 140 samples had TOM and Cu values below the thresholds
258 (LoLo). HiHi samples (average Cu = 109 $\mu\text{g/g}$) had 3.6x and 23x more Cu than HiLo samples
259 (30 $\mu\text{g/g}$) and LoLo samples (4.7 $\mu\text{g/g}$), respectively.

260

261 **Fig 4. Total organic matter by copper.** Point graph showing the relationship between copper
262 (Cu) concentration ($\mu\text{g/g}$ sediment) and total organic matter (TOM) percentage. Each of the 204
263 samples are represented by a point. Color represents the site with dark blue representing Barber
264 Bridge, bright blue Fort Pierce, red Harbor Branch Channel, purple Harbortown Marina, yellow

265 Hobe Sound, pink Jensen Beach, maroon Jupiter Narrows, dark green Linkport, light blue
266 Manatee Pocket, gold Melbourne Causeway, tan Merritt Island Causeway, dark gray Middle
267 Estuary, light blue North Fork, bright green Round Island, light gray Sebastian Inlet, black South
268 Fork, dark purple South Fork 2, light green Vero Beach, and turquoise Vero Beach Marina.
269 Shape represents the number of muck characteristics with circles representing three muck
270 characteristics, triangles two, squares one, and pluses zero. The blue line at 10% [7] represents
271 the threshold that separates the low TOM (left) from the high TOM (right) sites whereas the red
272 line at 65 µg/g [69] separates the high Cu (above) from the low Cu (below) sites.

273

274 **General sequence information**

275 There were 110 575 ASVs associated with the samples in this study. Filtering, described above,
276 reduced the number of ASVs to 16 027. This filtering step also reduced the number of sequences
277 from 1 857 744 to 1 598 653 (13.9%). The overall microbiome had 63 phyla, 193 classes, 472
278 orders, 799 families, 1 315 genera, and 1 691 species.

279 **Alpha diversity**

280 The mean Shannon alpha diversity was 6.45; its distribution was significantly correlated (p-
281 values <2.2e-16) with observed ASVs (rho = 0.92), Fisher diversity (0.97), Simpson diversity
282 (0.69), and Chao1 (0.92) with Spearman correlation tests. There were significant differences
283 between Sites (BH p-value = 0.020), Estuary (0.034), Location (0.0083), Sampling Period (1.1e-
284 15), IRL-focused Sampling Period (3.76e-15), and SLE-focused Sampling Period (3.7e-07)
285 categories, but not by TOM/Cu (0.92), Muck (0.78), or Season (0.095) (S6 Table). Dunn analysis
286 did not reveal any significantly different site pairs. Both the IRL and SLE exhibited the same
287 patterns in terms of alpha diversity (Fig 5A). The D17 and D18 sampling periods were

288 statistically similar to one another (IRL BH p-value = 0.16, SLE = 0.93); but were statistically
289 dissimilar to the other two sampling periods, which were also significantly different from each
290 other. Dunn testing at the Location level revealed that the North sites were statistically lower
291 than the SLE, South, and South Central sites (Fig 5B).

292

293 **Fig 5. Shannon diversity patterns** Boxplots of Shannon diversity based by Estuary and
294 Sampling Period (A) and Location (B) categories. IRL stands for Indian River Lagoon and SLE
295 for St. Lucie Estuary. The letters on top of each boxplot denote the results from the pairwise
296 Dunn test with different letters denoting statistical significance (Benjamini-Hochberg adjusted p
297 values < 0.05). In A the letters show how each of the sampling periods were different within
298 each estuary but do not denote inter-estuary comparisons. Bars denote largest and smallest values
299 within 1.5 times the interquartile range, middle line is the median, ends of boxes are the first and
300 third quartiles.

301

302 **Microbial community makeup of estuaries**

303 The top three phyla in the IRL and SLE were the same: Proteobacteria, Bacteroidetes, and
304 Chloroflexi (S2A Fig). The percentage of Epsilonbacteraeota was 16x higher in the IRL (2.2%)
305 than the SLE (0.14%), whereas the percentage of Nitrospirae in the SLE (3.3%) was 8.7x more
306 than in the IRL (0.38%). Desulfobacterales, Flavobacteriales, Anaerolineales and
307 Steroidobacterales were four of the top five orders that overlapped between estuaries. (Fig 6A).
308 The most common order for the SLE was Betaproteobacteriales (7.9%), which was 18x higher
309 than the IRL (0.44%). The other top IRL order was Cellvibrionales (4.4%) which was 2.8x
310 higher than in the SLE (1.5%). The following orders also occurred at levels double or greater in

311 the IRL than in the SLE: *Pirellulales* (2.6x, IRL = 2.4%, SLE = 0.95%), *Campylobacteriales*
312 (16.0x, 2.2%, 0.14%), *B2M28* (9.6x, 1.9%, 0.20%), *Actinomarinales* (4.0x, 1.7%, 0.43%), and
313 *Thiotricales* (3.3x, 1.7%, 0.51%). The SLE had more pronounced differences between seasons
314 than the IRL, which had the same top five orders throughout all sampling periods. (S3 Fig) In the
315 SLE, *Betaproteobacteriales* was 2.4x higher in the wet seasons (11.1%) than in the dry seasons
316 (4.6%), whereas *Flavobacteriales* decreased about 4.5x between the dry (6.7%) and wet (1.5%)
317 seasons. The following also saw decreases of at least 2x: *Actinomarinales* (2.1x, dry = 0.57%,
318 wet = 0.28%), *Desulfuromonadales* (2.7x, 1.7%, 0.65%) *Verrucomicrobiales* (5.3x, 2.4%,
319 0.44%), *Rhodobacteriales* (4.5x, 1.7%, 0.38%), *Campylobacteriales* (2.8x, 0.20%, 0.072%).

320

321 **Fig 6. Microbial community patterns.** Stacked bar graphs showing the phylogenetic orders
322 with a mean prevalence greater than 1% across all samples associated with (A) Indian River
323 Lagoon (IRL) and St. Lucie Estuary (SLE); (B) three muck characteristics and zero muck
324 characteristics samples and (C) high total organic matter/high copper (HiHi) and high total
325 organic matter/low copper (HiLo) samples. TOM stands for total organic matter and Cu stands
326 for copper.

327

328 **Microbial community makeup of 3MC/0MC and HiHi/HiLo** 329 **samples**

330 Samples with three (3MC) and zero muck characteristics (0MC) shared four of their top five
331 orders including *Desulfobacteriales*, *Anaerolineales*, *Flavobacteriales*, and *Steroidobacteriales* (Fig
332 6B). *Bacteroidales* and *Cellvibrionales* ~~make~~ up the rest of the top five for 3MC and 0MC
333 samples respectively. Orders that were found to be at least twice as common in the 3MC samples

334 versus the OMC samples were: Betaproteobacteriales (3.5, 3MC = 4.8%, 0MC = 1.4%),
335 Campylobacterales (2.3, 1.9%, 0.81%), and Ectothiorhodospirales (2.4, 3.0%, 1.3%). OMC
336 samples had higher levels of certain orders including Actinomarinales (4.8x, 0.39%, 1.9%),
337 B2M28 (2.2, 0.80%, 1.8%), Nitrosopumilales (2.7, 0.76%, 2.1%), Pirellulales (5.4, 0.51%,
338 2.8%), Rhizobiales (2.2, 0.97%, 2.2%), and Synechococcales (2.0, 0.79%, 1.6%).

339 The top phyla for the high TOM and low Cu (HiLo) samples matched the order above for
340 3MC. High TOM and high Cu (HiHi) samples had the same top three phyla (Proteobacteria,
341 Bacteroidetes, Chloroflexi) while Epsilonbacteraeota and Crenarchaeota replaced Acidobacteria
342 and Planctomycetes. (S4A Fig). HiHi and HiLo shared three of the top five orders with the 3MC
343 and OMC samples: Desulfobacteriales, Flavobacteriales, and Anaerolineales (Fig 6C).

344 Completing the top five for HiHi was Campylobacterales and Bacteroidales; the former was 2.2x
345 more abundant in HiHi samples (5.2%) than in HiLo samples (2.3%). Betaproteobacteriales and
346 Steroidobacteriales completed the top five for HiLo, with the former being found 4.0x more in
347 HiLo samples (5.8%) than in HiHi samples (1.4%). Nitrosopumilales was 2.3x higher in the
348 HiLo (0.91%) than the HiHi samples (0.40%).

349 **Beta diversity**

350 Permutational analysis of variance results showed significant differences between the IRL and
351 SLE samples, and across samples among the Muck and TOM/Cu subcategories (Monte Carlo p-
352 values = 0.0001) (Table 1). 3MC and OMC samples were statistically different from one another
353 (0.0001), and from samples with two and one muck characteristics, which were statistically
354 similar to one another (0.76). HiHi samples were significantly dissimilar (0.0001) from LoLo
355 and HiLo samples but not from LoHi samples (0.32). LoLo samples were also not significantly
356 different than LoHi samples (0.072), but were from HiLo samples (0.0001). All Location

357 combinations were significantly different than one another (0.0001), along with most sampling
 358 periods (<0.05) except the SLE D17 and D18 sampling periods (0.29) (S7 Table).

359 **Table 1. Summarized permutational analysis of variance results.**

Overall Parameter Category^a	Pseudo-F	P(perms) ^c	P(MC) ^d
Pair-wise test category ^b	t statistic		
Estuary	42	0.0001	0.0001
TOM^e/Cu^f	7.2	0.0001	0.0001
Low TOM/Low Cu, High TOM/High Cu	2.7	0.0001	0.0001
Low TOM/Low Cu, Low TOM/High Cu	1.2	0.045	0.072
Low TOM/Low Cu, High TOM/Low Cu	3.7	0.0001	0.0001
High TOM/High Cu, Low TOM/High Cu	1.1	0.47	0.32
High TOM/High Cu, High TOM/Low Cu	2.8	0.0001	0.0001
Low TOM/High Cu, High TOM/Low Cu	1.4	0.034	0.061
Muck Characteristics	5.4	0.0001	0.0001
0, 2	1.9	0.0002	0.0001
0, 3	3.4	0.0001	0.0001
0, 1	1.6	0.0028	0.0022
2, 3	1.4	0.03	0.034
2, 1	0.82	0.88	0.76
3, 1	1.5	0.013	0.017
Estuary by Season	17	0.0001	0.0001
All pairwise analyses had P(MC) values equal to 0.0001 ^g			
Location	10	0.0001	0.0001
All pairwise analyses had P(MC) values equal to 0.0001 ^g			
Estuary by Sampling Period	10	0.0001	0.0001
SLE-D17, SLE-D18	1.1	0.25	0.29
All other pairwise analyses had P(MC) values less than 0.05 ^g			
Site	11	0.0001	0.0001
Barber Bridge, Vero Beach Marina	1	0.31	0.4
All other site pairwise analysis have P(MC) values below 0.05 ^g			

360 ^aBold text represents results from overall category and ^bregular text represents results from the
 361 pair-wise testing results.

362 ^cP(perms) stands for the permutation p value, ^dP(MC) stands for Monte Carlo p value, ^eTOM
 363 stands for total organic matter, and ^fCu stands for copper. ^gThe full results can be seen in S7
 364 Table.

365 **Influence of environmental variables**

366 The environmental variable most statistically associated with microbial variation between the
367 samples was PWS (Pseudo-F = 20.396, proportion = 0.09171), followed by TOM (15.244,
368 0.064028), Cu (7.5017, 0.030522), and finally sediment temperature (5.5491, 0.022076) (Fig 7
369 and S8 Table). PWS generally decreased from the upper left corner to lower right, separating the
370 IRL sites from the SLE sites. TOM generally decreased from the lower left corner to the upper
371 right, separating the samples with muck characteristics from those with none. Cu generally
372 decreased from the top of the graph, where the HiHi and LoHi samples were found, to the
373 bottom, where the HiLo and LoLo samples were.

374

375 **Fig 7. Distance-based redundancy analysis.** Distance-based redundancy analysis of the
376 sediment samples with colors representing the amount of copper [high or > 65µg/g [69] (black),
377 low or < 65µg/g (gray)]. The shape shows the number of muck characteristics associated with a
378 sample with circles representing 1-3 characteristics and triangles representing no muck
379 characteristics. A filled shape is an Indian River Lagoon sample and a hollow shape represents a
380 St. Lucie Estuary sample. The results of the distance-based linear models are shown by the lines
381 and their associated environmental parameter and shown in S8 Table.

382

383 **Discussion**

384 **Environmental parameters and seasonality**

385 Sampling was scheduled to capture seasonality, but the extended impacts of Hurricane Irma
386 which made landfall in Florida, USA on September 10, 2017 shifted one of the sampling periods

387 from Aug-Sept 2017 to Oct-Nov 2017. PWS seasonality was more defined in the SLE than the
388 IRL since there was a statistically lower PWS in the SLE during both wet seasons compared with
389 the dry seasons (Fig 2A). This reflects the increased rainfall and consequent discharges from
390 Lake Okeechobee/C44 and other canals/streams during these times (S1 Fig and S4 Table). In
391 contrast, the PWS in the IRL was statistically different between the wet seasons (Fig 2A).
392 Although the sediment temperature was highest during W16, W17 was either statistically similar
393 to or lower than the two dry seasons in both the IRL and SLE (Fig 2A). In conclusion,
394 seasonality was a prominent environmental factor in the SLE in terms of PWS but not sediment
395 temperature; however, neither parameter showed distinguishable seasonality in the IRL. PWS
396 generally increased from North to South in the IRL, possibly due a greater number of oceanic
397 inlets in the south (Figs 1 and 2B) [4,70]. Owing to higher freshwater releases, the SLE had a
398 statistically lower PWS with a wide range (Fig 2B and S4 Table) [70].

399 **Muck accumulation**

400 Muck is formed by the bacterial degradation of organic matter at the transition between
401 freshwater and estuarine waters [6,7]. The sites that had at least one muck sample were located
402 near this transition in either the SLE (South Fork, Middle Estuary) or C25 (HT) (Figs 1 and 3).
403 Samples with at least one muck characteristic were also near areas that allowed accumulation of
404 organic matter due to restricted flow caused by the shape and bathymetry of the area (HB, MP)
405 or high residence time (MC) (Figs 1 and 3) [71,72]. Although South Fork 2 is located adjacent to
406 the C44 canal, it experiences periodic large volumes of high velocity flow that could prevent fine
407 particles and organic matter from settling (Fig 1).

408

409

410 **Copper contamination**

411 Copper can be found in the sediment near marinas due to the use of copper-antifouling paints,
412 which explains why HB, located in a channel historically used for large boats, HT, located by a
413 large active marina, and MP, located by a small active marina, were the only sites to have high
414 Cu (Figs 1 and 4) [69,73]. The site by the active Vero Beach Marina is not located in a flow-
415 restricted area, possibly allowing the current to move contaminants away from the area (Fig 1).

416 **Alpha diversity patterns**

417 Alpha diversity was highest in the warmest period (W16) and increased from north to south,
418 matching the pattern of higher diversity in warmer environments seen in other studies (Figs 5A
419 and 5B) [74]. The diversity drop associated with W17 samples may be related to the impact of
420 Hurricane Irma (Fig 5A) [75].

421 **Estuarine microbial community differences**

422 PWS has been identified as an important factor in microbiome variation in other studies [63].
423 The most common phylum associated with either estuary was Proteobacteria, which is consistent
424 with other estuarine studies (S2A Fig) [12,63,75,76]. Proteobacteria has members that are
425 capable of utilizing a wide variety of substrates, which allow them to occupy many different
426 environments [77]. The most common Proteobacteria orders in the IRL and SLE included
427 Desulfobacterales and Steroidobacterales, while the IRL had more Cellvibrionales, and the SLE
428 had more Betaproteobacterales (Fig 6A). Most members of Desulfobacterales, including its
429 high-percentage families Desulfobacteraceae and Desulfobulbaceae (S2C Fig), are sulfate-
430 reducing bacteria and have also been found in high percentages in other estuary studies [63,78].
431 The family Woeseiaceae (order Steroidobacterales) (S2C Fig) has members capable of

432 facultative sulfur- and hydrogen-based chemolithoautotrophy and is considered a core member
433 of marine sediments [79,80]. The family Halieaceae (order Cellvibrionales) (S2C Fig) is found in
434 coastal marine areas and is capable of aerobic photoheterotrophic growth [81]. *Halioglobus* (S2D
435 Fig) (Halieaceae) is capable of denitrification and requires NaCl for growth, which may be why
436 it was more prevalent in the IRL [82]. Betaproteobacteriales was shown in multiple studies to
437 increase in freshwater-influenced areas of estuarine lagoons, which is similar to this study where
438 Betaproteobacteriales increased in the SLE during the wet season samples (Fig 6A and S3 Fig)
439 [63,76]. Flavobacteriales (phylum Bacteroidetes) (Fig 6A) is also commonly abundant in other
440 estuary studies [75]. Most members of the family Flavobacteriaceae (S2C Fig) require NaCl or
441 seawater salts for growth, which explains the decrease in Flavobacteriales during the wet seasons
442 in the SLE (S3 Fig) [83]. Anaerolineaceae (phylum Chloroflexi, order Anaerolineales) (Fig 6A
443 and S2A Fig) is comprised of obligate anaerobes with most members capable of breaking down
444 proteinaceous carbon sources [84]. Behera et al. (2017), in another study into the effects of
445 freshwater on a brackish lagoon, found that the phylum Acidobacteria and classes
446 Gammaproteobacteria and Alphaproteobacteria were higher in more marine environments [63].
447 Our study instead found that there were more Acidobacteria in the more freshwater SLE along
448 with relatively equal amounts of Gamma- (0.61% more in IRL) and Alphaproteobacteria (0.13%
449 more in SLE) (S2A and S2B Figs).

450 Microbes that are highly abundant in a system are likely to be the actively-metabolizing
451 part of the community; although a portion of the rare microbes can be active, they are more
452 likely to be dormant or dead cells [85,86]. Since Desulfobacterales and Steroidobacterales had
453 high relative abundances, they were likely responsible for some of the sulfur cycling in the
454 lagoon [78,80]. Likewise, carbon cycling was likely affected by the photosynthetic

455 Cellvibrionales and members of Anaerolineales [82,84]. Nitrogen cycling also was affected by
456 denitrifying members of Cellvibrionales, like Halioglobus and nitrogen-fixing members of
457 Betaproteobacteriales [78,81].

458 **Microbiome shifts associated with copper and muck**

459 Alpha diversity did not decrease in stressed sediments (sediment with muck characteristics or
460 Cu-contamination) as seen in other studies exploring the effects of metals and clay/silt [5,19].
461 While a diversity decrease can be an indicator of impaired environmental health, organisms can
462 also become adapted to stressors with the largest drop in diversity associated with initial
463 exposure to contamination [5]. Thus, it is possible the community has had enough time to adapt
464 to contamination and for tolerant species to flourish [87]. There were significant differences
465 between the 0MC and 3MC microbiomes, which could be partially due to the smaller pore size
466 in muck affecting the ability of some microbes to flourish [88].

467 TOM has also been seen as an important environmental variable in other studies [89] or
468 studies that measured TOM-covarying parameters such as percent fines [5] or silt [21]. The top
469 phyla matched between the 0MC and IRL samples as well as the 3MC and SLE samples (S2A
470 and S4A Figs). This pattern likely occurred because 43.8% (21/48) of the samples in the SLE
471 were classified as 3MC samples whereas only 12.2% (19/156) of the IRL sites were classified as
472 3MC samples. This could help explain why there were higher percentages of
473 Betaproteobacteriales in the 3MC samples as well as the HiLo samples since there were also no
474 HiHi samples in the SLE samples.

475 In comparison to HiLo samples and 0MC samples, HiHi and 3MC samples had higher
476 percentages of Epsilonbacteraeota and Crenarchaeota (S4A and S5A Figs). A recent study
477 suggested that Archaea, such as the Crenarchaeota, have a greater resistance to copper

478 contamination due to their ability to sequester or pump out copper; this study also found
479 Crenarchaeota flourishing in their copper-contaminated samples [23]. Members of the order
480 Campylobacterales (phylum Epsilonbacteraeota) were also found at higher abundances in HiHi
481 and 3MC sediments; some of its members, including the genus *Sulfurovum*, have been found in
482 sulfide- and hydrocarbon-rich sediments similar to muck [90] as well as metal-contaminated
483 sediments [91,92]. (Figs 6B and 6A; S4D and S5D Figs). *Sulfurovum* is a mesophilic facultative
484 anaerobe, requires salts for chemolithoautotrophic growth with elemental sulfur or thiosulfate as
485 an electron donor, nitrate or oxygen as an electron acceptor, and CO₂ as its carbon source [93].
486 Conversely, 0MC and HiLo samples had higher abundances of the *Candidatus Nitrosopumilus*
487 genus and its associated higher taxonomic ranks *Nitrosopumilus* is similar to *Sulfurovum* in that
488 it uses CO₂ as its carbon source and is halophilic, but it grows chemolithoautotrophically by
489 conducting ammonia oxidation to nitrite and is aerobic. Other families which are typically
490 aerobic and were more abundant in the 0MC samples included Pirellulaceae and
491 Sandaracinaceae [94–96]. This shows that the microbiome differences between 3MC and 0MC
492 samples were at least partially due to the former being typically more anaerobic since increased
493 organic matter can lead to increased respiration and depletion of oxygen [7]. 3MC samples also
494 had lower abundances of Cyanobiaceae which could be due to the increased turbidity associated
495 with muck and its higher percentage of silt/clay [7]. Sediment microbial communities have been
496 shown in other studies to be greatly affected by carbon sources, electron acceptors, and amount
497 of oxygen in an area [12,89].

498 **Conclusions**

499 The most important variable causing shifts between the microbiomes was PWS, this was mainly
500 due to the influence of seasonal freshwater discharges into SLE causing microbiome differences

501 in comparison to the IRL. Other observed differences included increases in anaerobic organisms
502 in the higher TOM 3MC samples and aerobic organisms in the lower TOM 0MC samples.
503 Tracking changes in the differentially abundant microbes present in different sediment types will
504 allow management agencies to predict areas that are at risk of developing muck due to microbial
505 influences or becoming sufficiently copper-contaminated to cause biological harm. This study
506 provides the first NGS data on the microbial diversity of the IRL which will serve as an
507 important baseline for future studies to measure the impact of anthropogenic inputs and natural
508 disasters. This data can also be used by researchers in other estuarine areas to compare their
509 results to determine if their systems are facing similar shifts in the microbiomes due to similar
510 anthropogenic impacts.

511 Future studies should be performed with greater sequencing depth and higher sampling
512 frequency, which could allow more of the diversity and rarer taxa in the samples to be captured,
513 and shotgun metagenomics to identify functional differences between sites. Incorporating the
514 measurement of anoxia and biogeochemical cycles would help to further delineate which
515 environmental variables are causing shifts to the microbiomes between sediment types and
516 geographical locations.

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773 **Supporting information**

774 **S1 Fig. National Weather Service temperature and rain sum patterns.** National Weather
775 Service NOWData showing the four sampling periods (Aug-Sep 2016 (dark orange), Mar-Apr
776 2017a (dark green), Oct-Nov 2017b (light orange), Apr 2018 (light green)) as well as the historical
777 temperature (A) and rain sum (B) during those months for the years 1990-2018 (dark gray). Bars
778 denote standard error.

779 **S2 Fig. Other taxonomic levels by Estuary.** Stacked bar graphs showing the phyla (A), classes
780 (B), orders (C), families (D), and genera (E) that have a mean of greater than 1% across all samples.
781 These graphs show the differences between the two main basins of the study, (Indian River Lagoon
782 (IRL) or St. Lucie Estuary (SLE))

783 **S3 Fig. Estuary orders by Sampling Period.** Stacked bar graph showing the orders with a mean
784 greater than 1% across all samples grouped by estuary (Indian River Lagoon (IRL) or St. Lucie
785 Estuary (SLE)) and sampling period (Aug/Sept 2016, Mar/Apr 2017, Oct/Nov 2017, and Apr
786 2018).

787 **S4 Fig. Other taxonomic levels by Muck classification.** Stacked bar graphs showing the phyla
788 (A), classes (B), orders (C), families (D), and genera (E) that have a mean of greater than 1%
789 across all samples. These graphs show the differences between the samples with three and zero
790 muck characteristics.

791 **S5 Fig. Other taxonomic levels by Total Organic Matter/Copper classification.** Other
792 taxonomic levels by Total Organic Matter/Copper classification
793 Stacked bar graphs showing the phyla (A), classes (B), orders (C), families (D), and genera (E)
794 that have a mean of greater than 1% across all samples. These graphs show the differences between
795 the samples with high TOM and copper (HiHi) and high TOM and low copper (HiLo).

796 **S1 Table. GPS Coordinates.** ^aNWS stands for National Weather Service [29], ^bIRFWCD for
797 Indian River Farms Water Control District, ^cUSGS for United States Geological Service [31],
798 ^dSFWMD for South Florida Water Management District [30].

799 **S2 Table. Measured environmental variables per sample.** Sediment temperature was
800 determined using a thermometer, water content by weight loss during oven drying, total organic
801 matter by weight loss in a muffle furnace, grain size fractions (gravel, sand, and silt/clay) were
802 determined using wet sieving, and Cu and Fe were measured using an atomic adsorption
803 spectrometer. See manuscript for details.

804 **S3 Table. Metadata per sample.** Table containing information pertaining to each sample such as
805 when (Sampling Season, Season) and where (Site, Location, Estuary) it was taken. A sample was
806 considered to have a muck characteristic if it exceeded one of the muck thresholds: 10% for total
807 organic matter, 60% for silt/clay fraction, and 75 % for water content [7]. Additionally, a sample
808 was considered to have high copper if it exceeded 65 µg/g [69]. ^aIRL for Indian River Lagoon,
809 ^bSLE for St. Lucie Estuary, ^cTOM stands for total organic matter, ^dCu for copper, ^eLoLo for low
810 TOM/low Cu, ^fHiHi for high TOM/low Cu, ^gLoHi for low TOM/low Cu, and ^hHiLo stands for
811 high TOM/low Cu.

812 **S4 Table. Average monthly means for canal daily discharges (ft³/s).** Data was taken from the
813 *United States Geological Services online database [31] or the **South Florida Water
814 Management District's DBHYDRO online database [30]. ^aThe regional location each canal/stream
815 was found in, ^bdata from the month before and months during each sampling period and ^cthe entire
816 survey. ^dThe the average streamflow from all fourteen stream/canals during the months before and
817 during each sampling period. IRL stands for Indian River Lagoon.

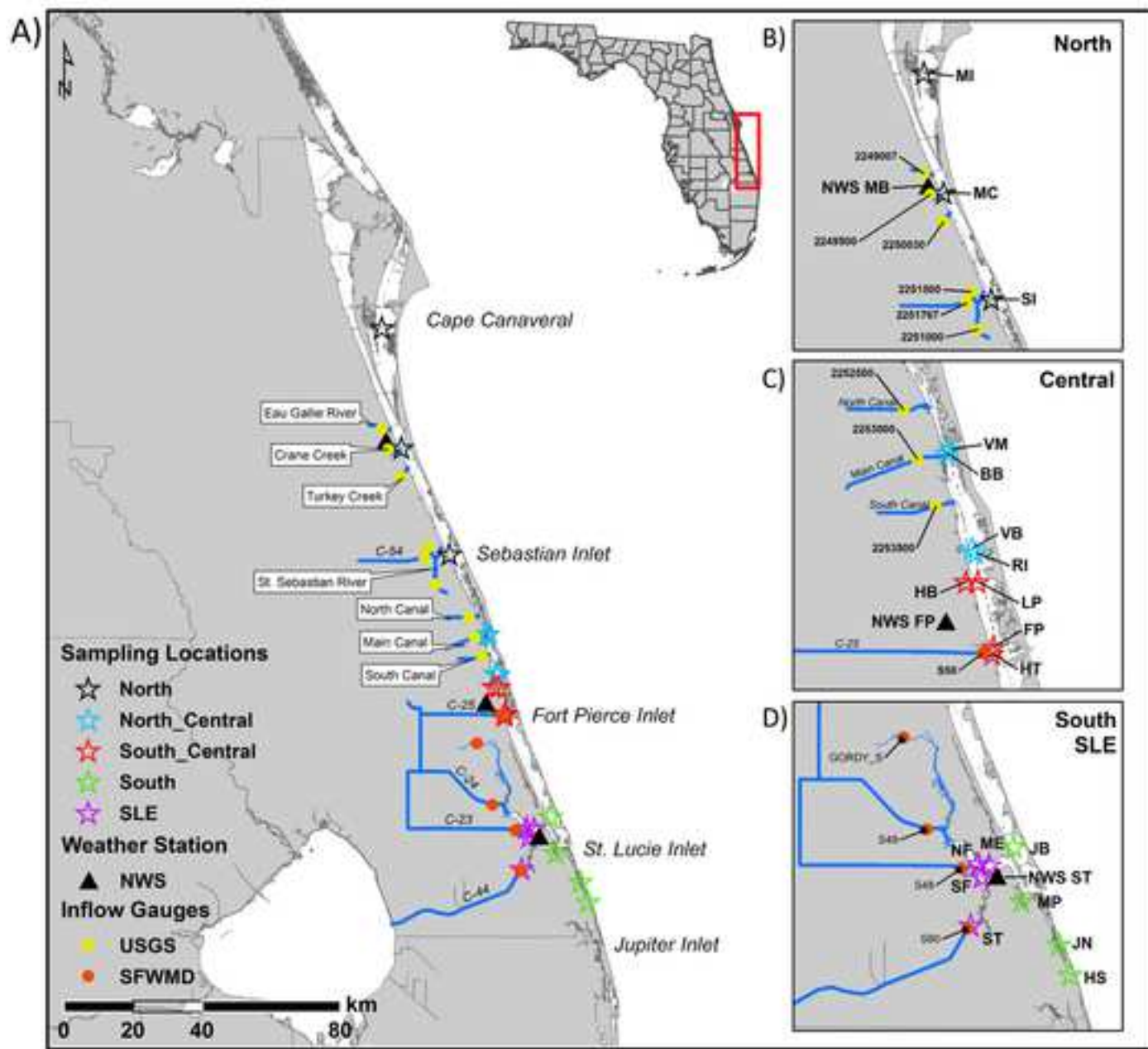
818 **S5 Table. Environmental parameters statistical analysis.** ^aBold text is associated with testing
819 the overall differences within a category with Kruskal-Wallis. ^bRegular text is associated with pair-
820 wise Dunn testing. ^cBH stands for Benjamini-Hochberg, ^dIRL stands for Indian River Lagoon and
821 ^eSLE stands for St. Lucie Estuary.

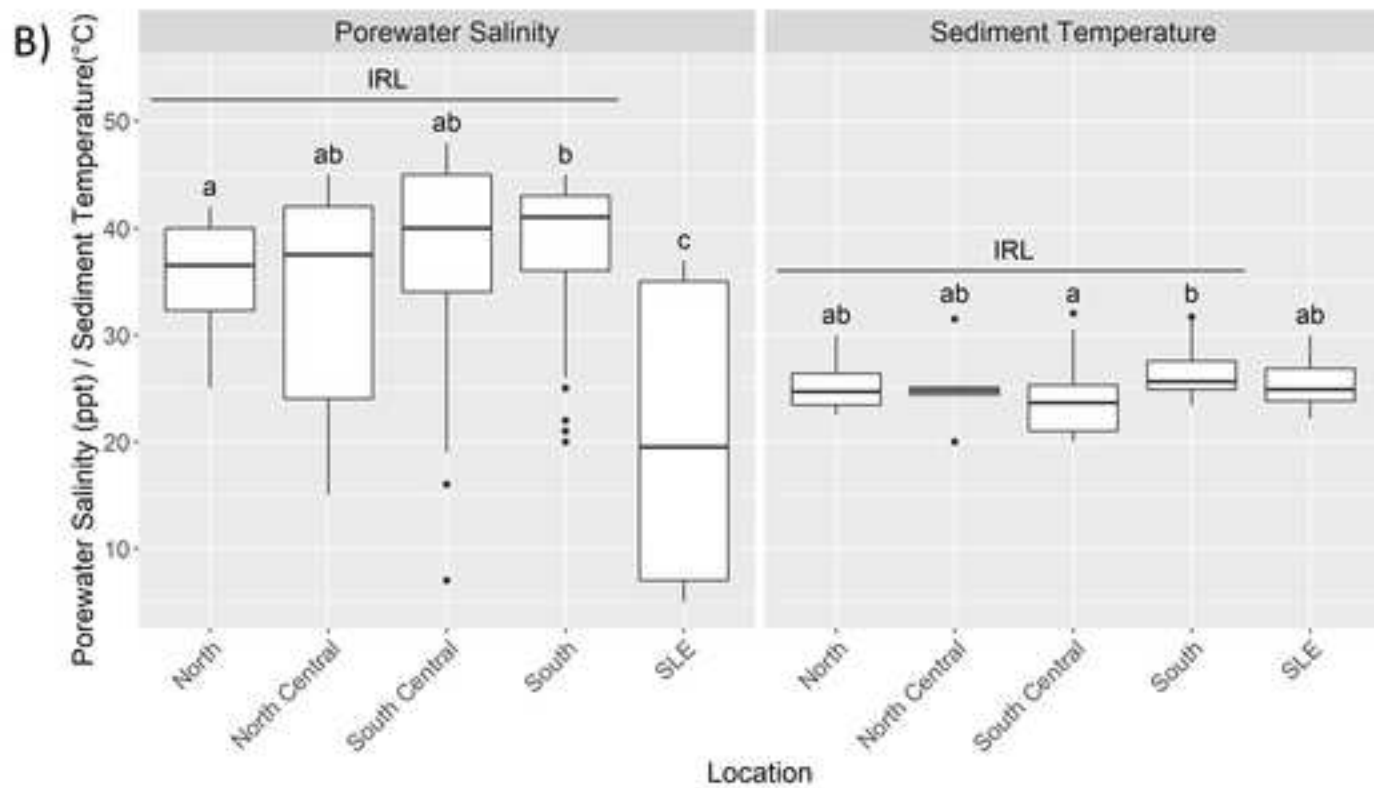
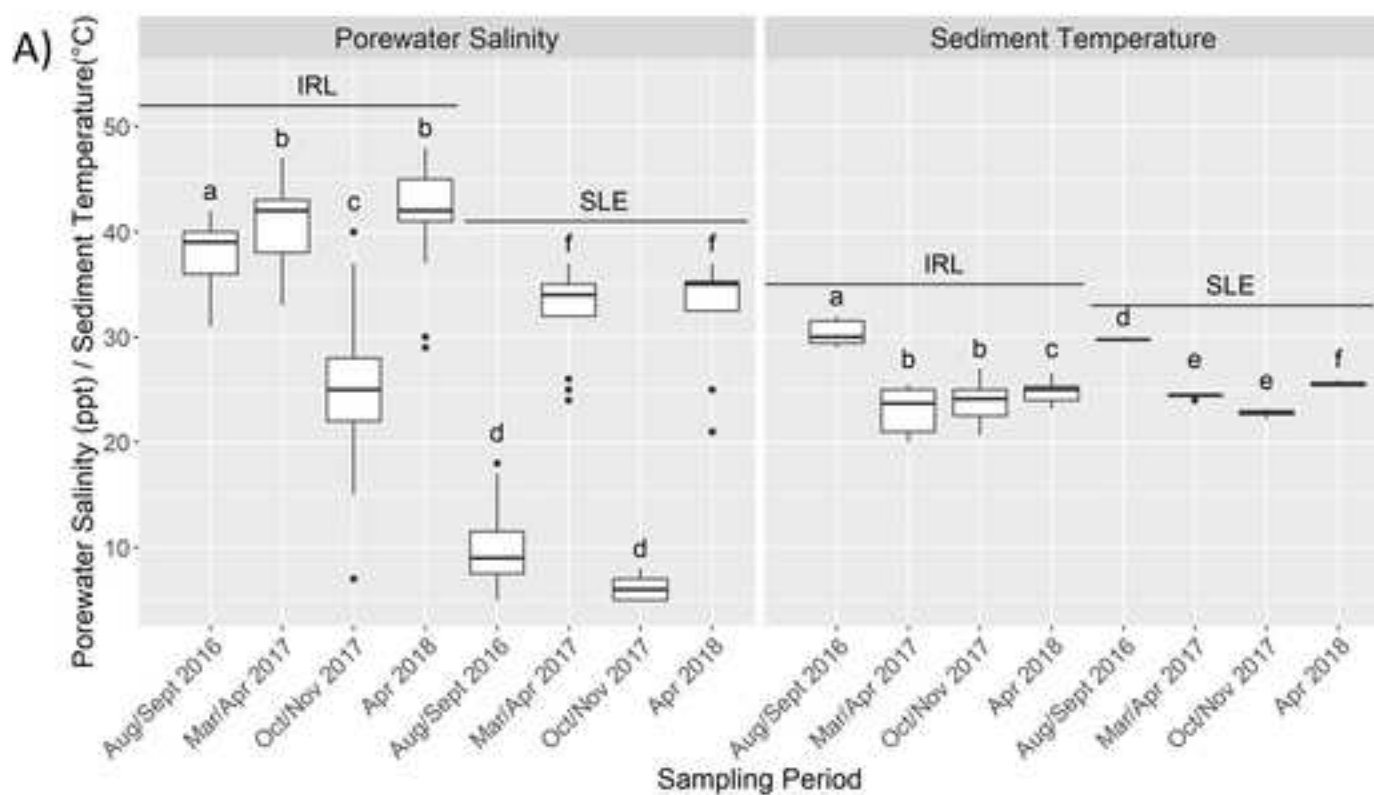
822 **S6 Table. Shannon diversity statistical analysis.** ^aBold text is associated with testing the overall
823 differences within a category with Kruskal-Wallis or Mann-Whitney use with a * indicating the
824 latter was used. ^bRegular text is associated with pair-wise Dunn testing. ^cBH stands for Benjami-
825 Hochberg, ^dTOM for total organic matter, ^eCu for copper, ^fIRL for Indian River Lagoon and ^gSLE
826 for St. Lucie Estuary.

827 **S7 Table. Full permutational analysis of variance results.** ^aBold text is associated with testing
828 the overall differences within a category and ^bregular text is associated with pair-wise testing.
829 ^cP(perms) stands for permutational p value, P(MC) for Monte-Carlo p value, ^eTOM for total
830 organic matter, ^fCu for copper, ^gIRL stands for Indian River Lagoon and ^hSLE for St. Lucie
831 Estuary.

832 **S8 Table. Distance-based linear model results.** ^aMarginal statistical tests are displayed for all
833 variables and ^bsequential statistical results are shown only if the variable was determined to
834 contribute a statistically significant amount of variation between microbial samples (p value <
835 0.05). ^cSS for sum of squares and ^dAICc stands for An Information Criterion.

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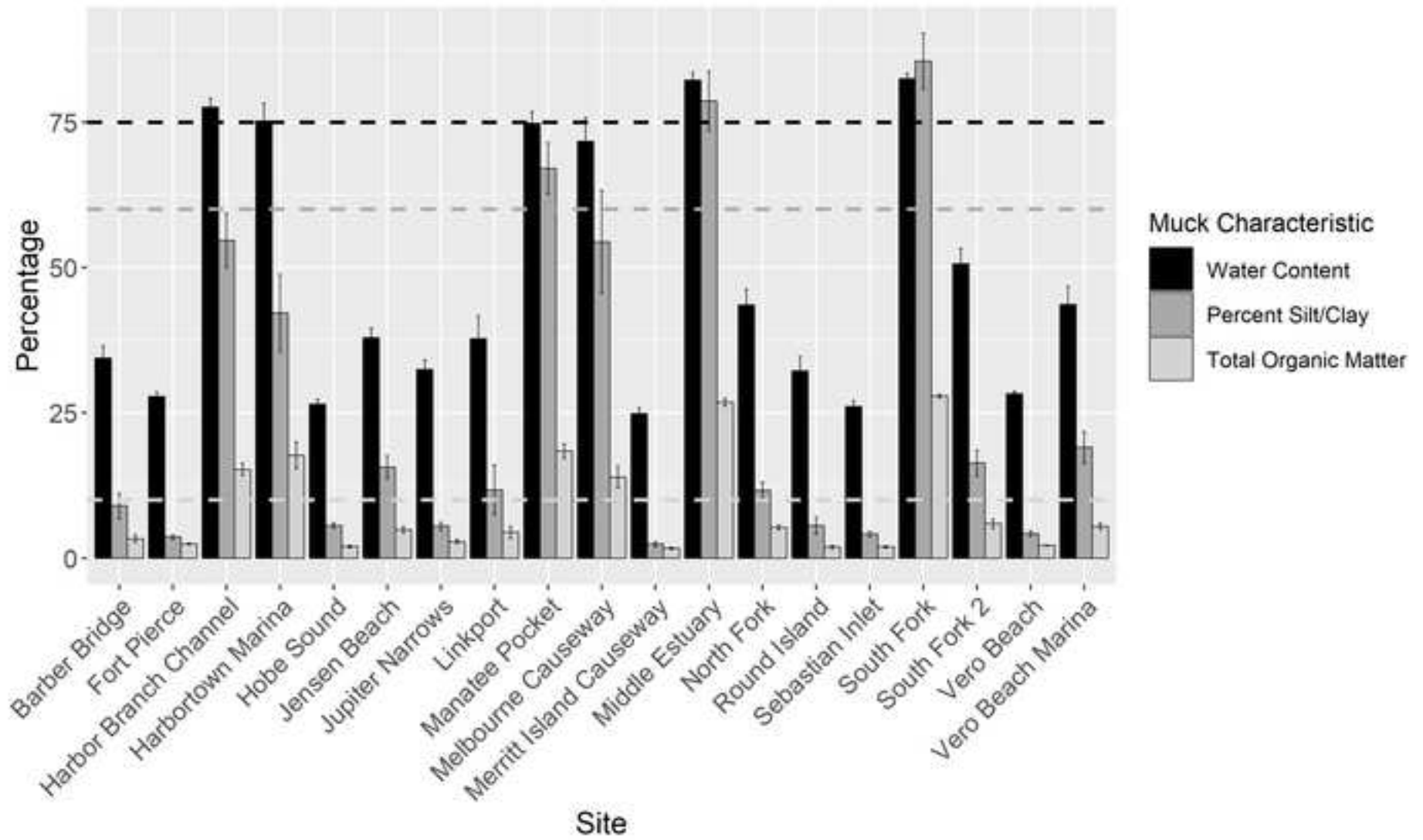
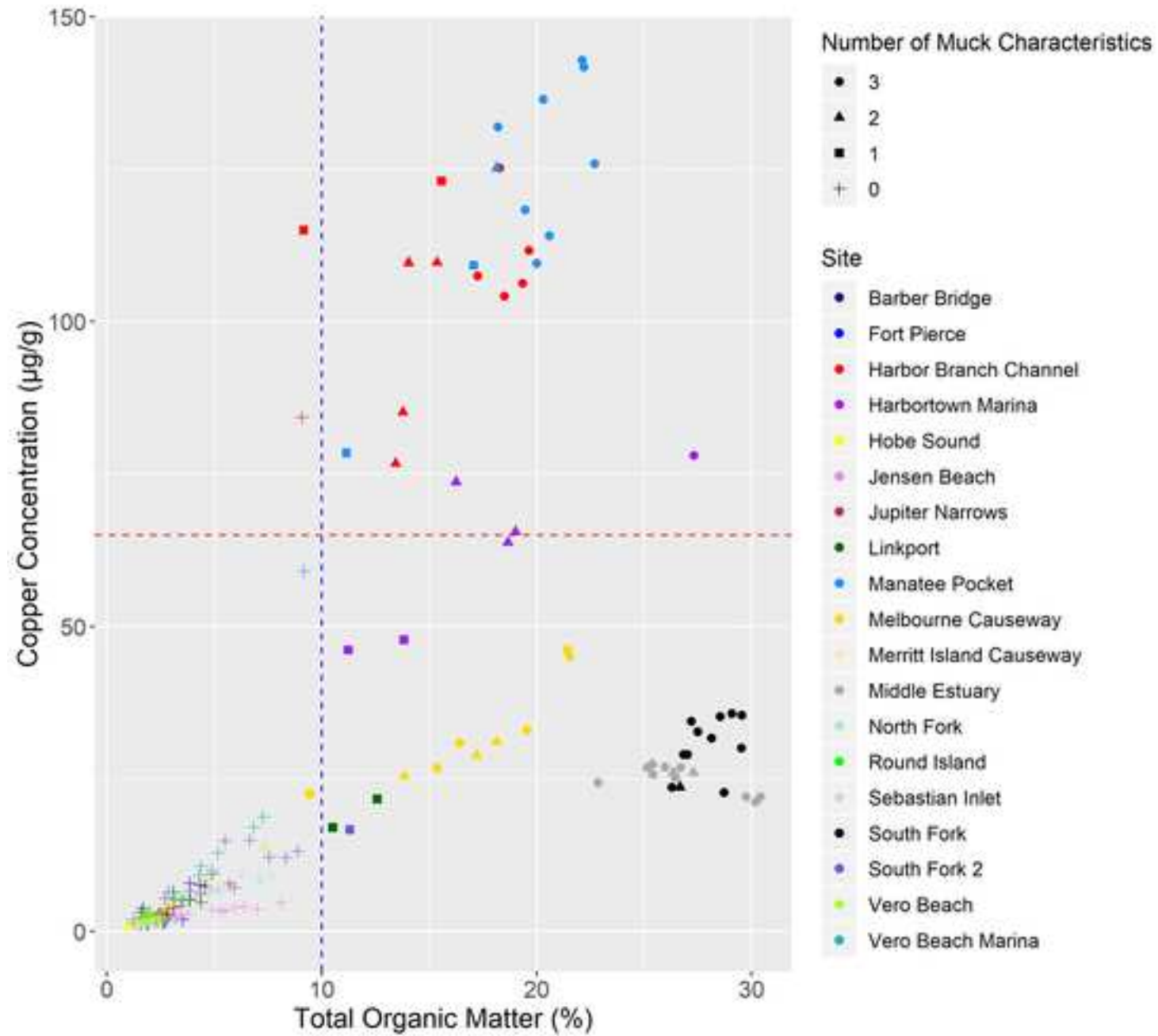
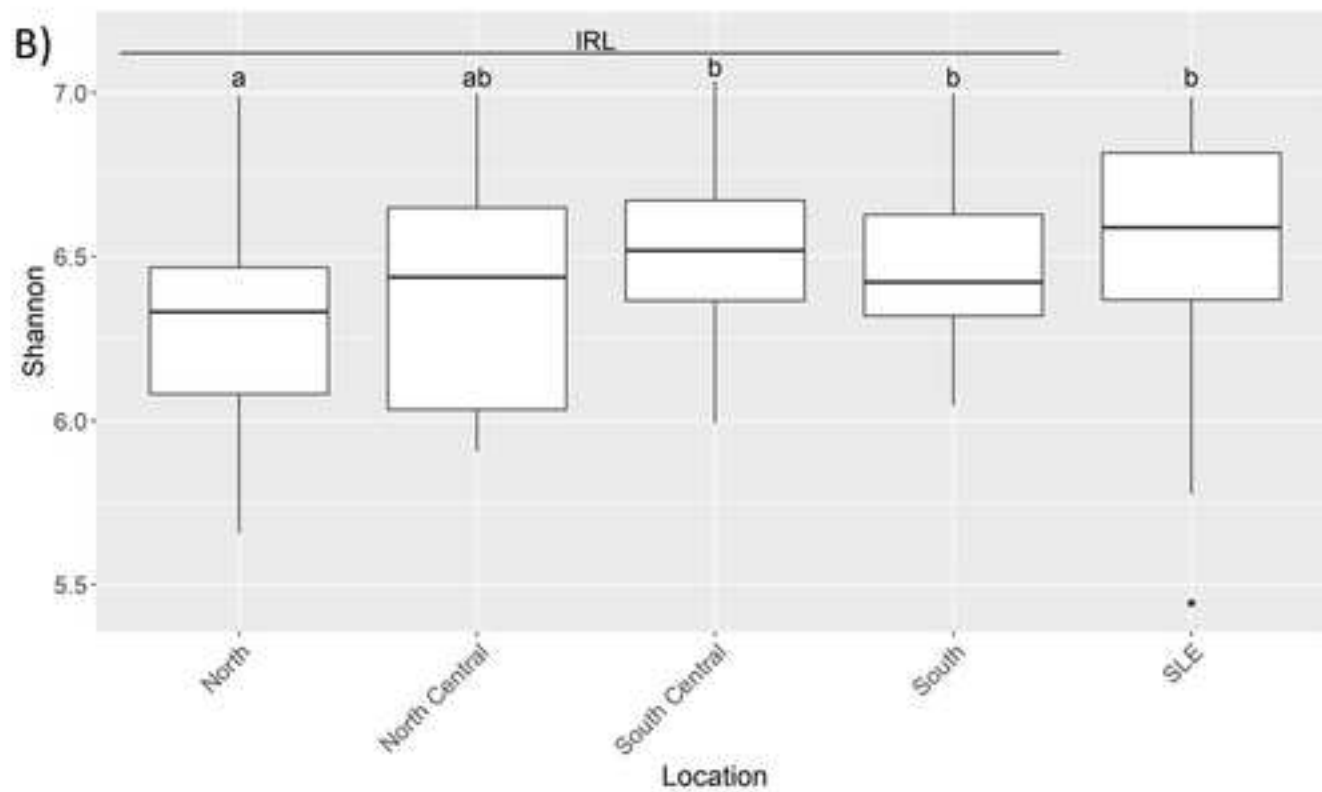
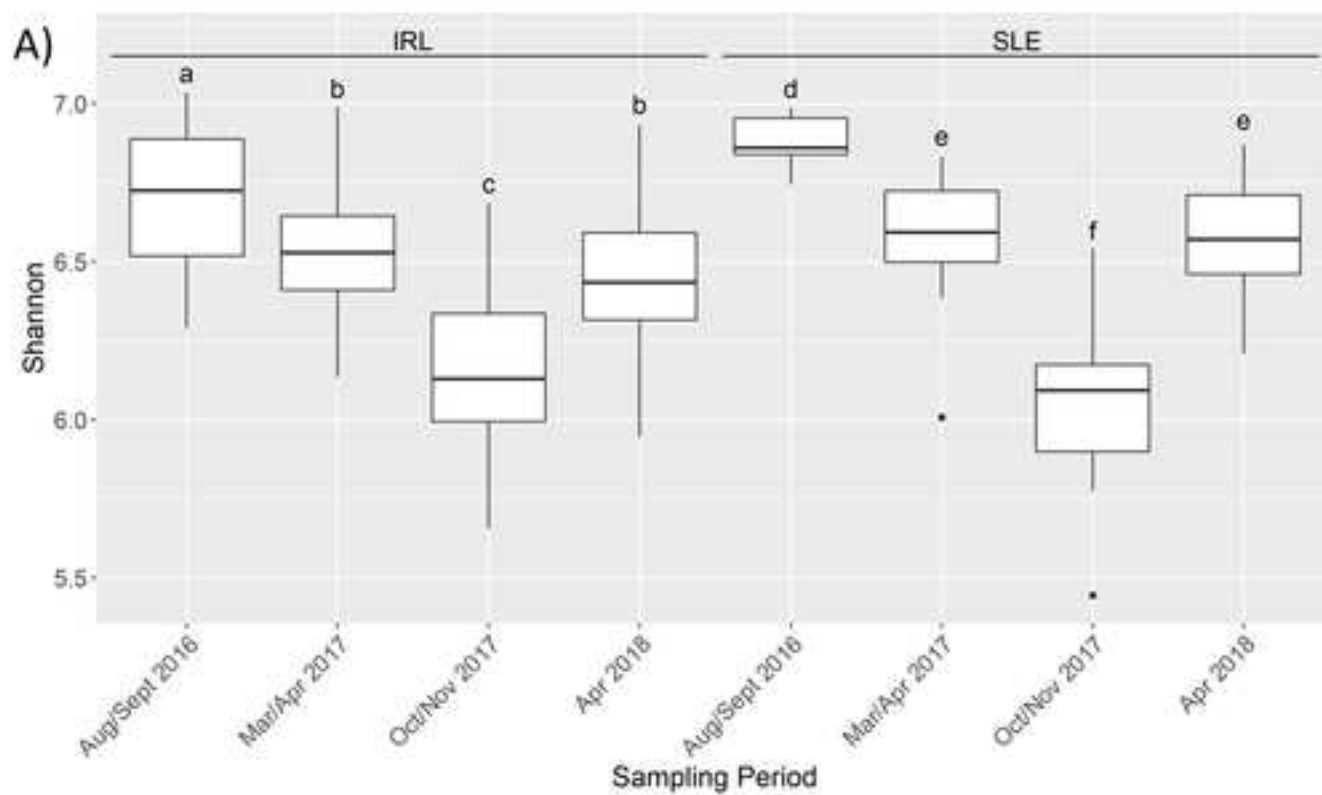
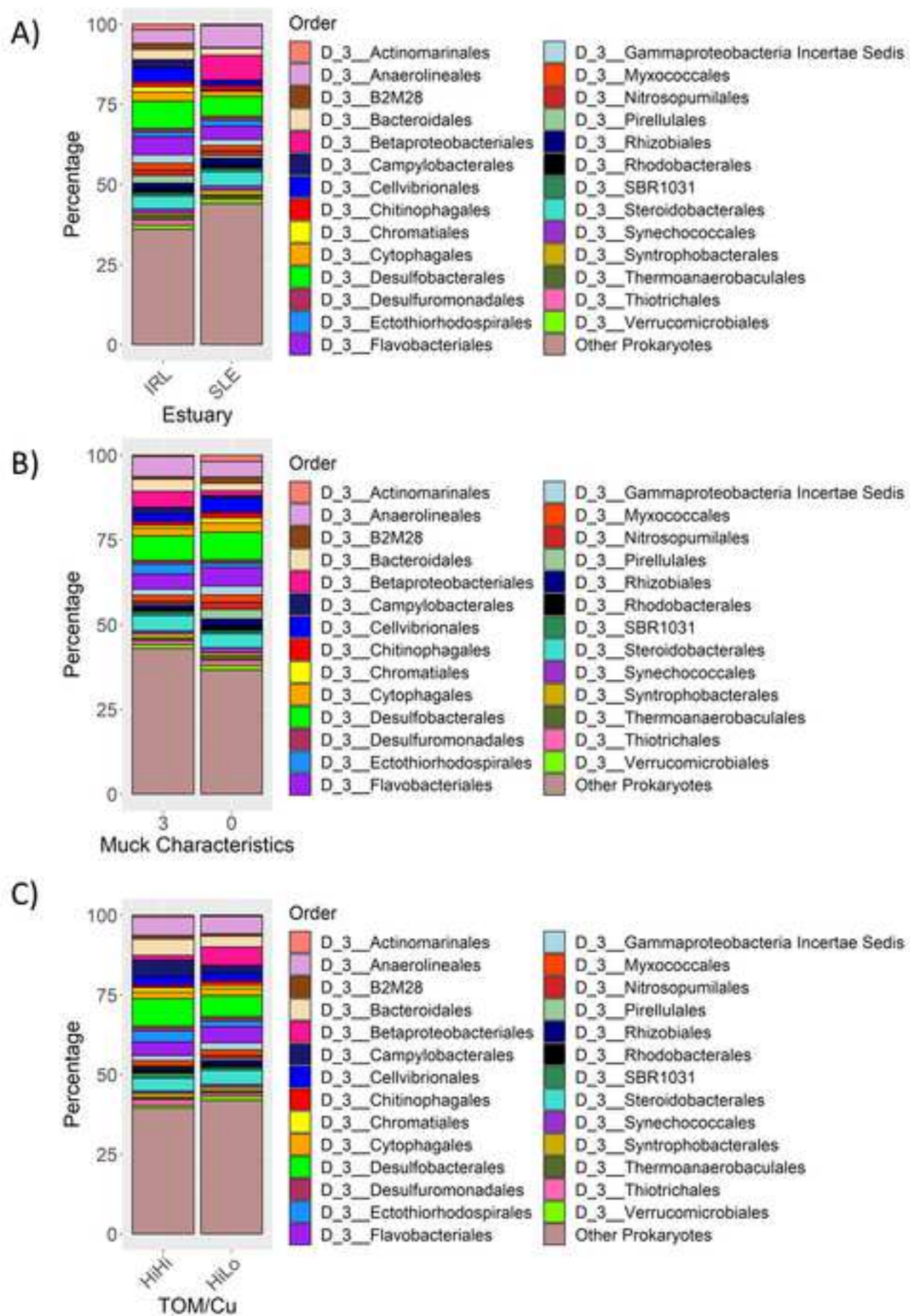
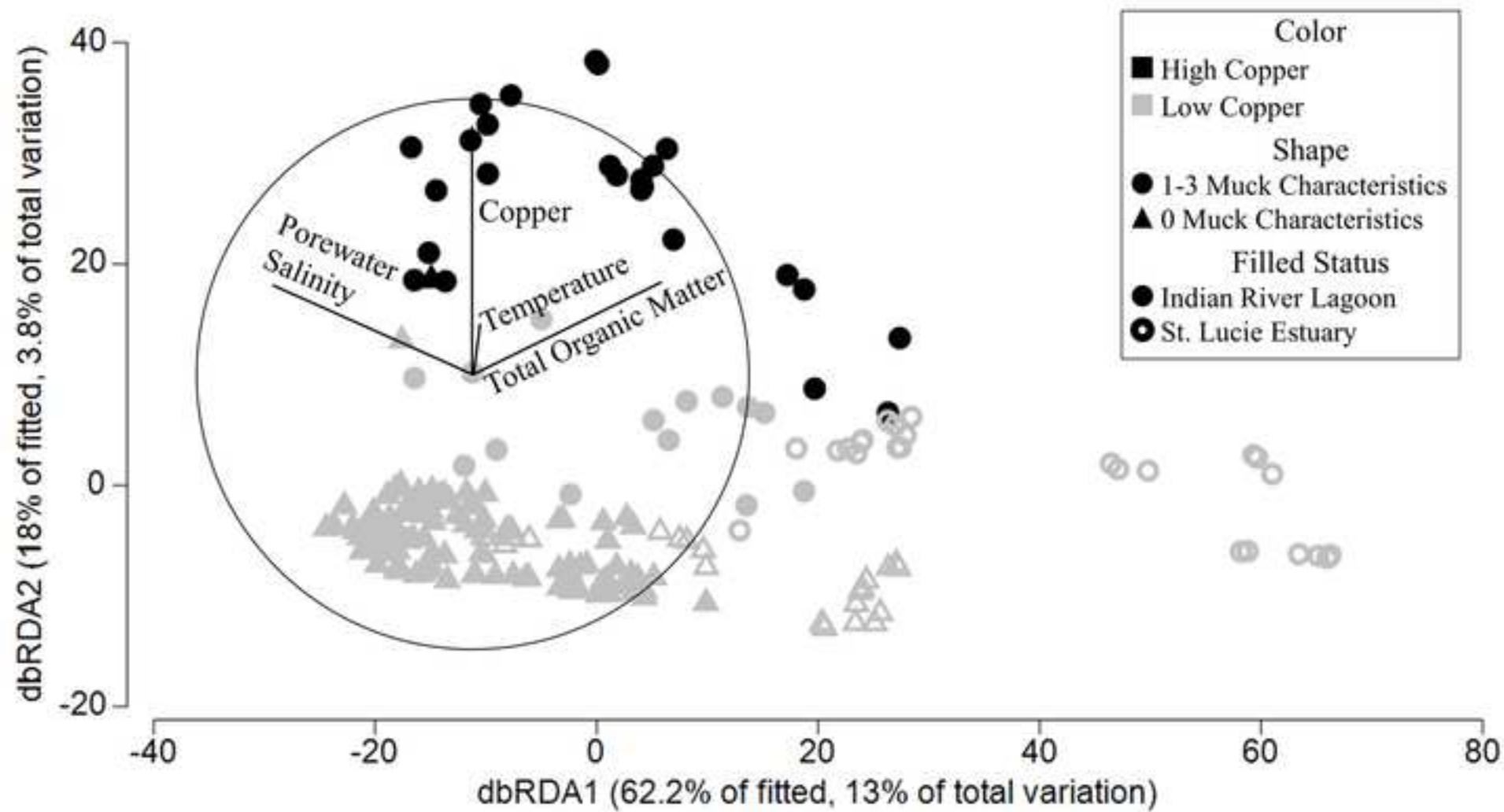


Fig 4























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