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# Linking Spatial and temporal dynamic of bacterioplankton communities with ecological strategies across a coastal frontal area

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31

32

33 **Abstract**

34 Ocean frontal systems are widespread hydrological features defining the transition zone between  
35 distinct water masses. They are generally of high biological importance as they are often associated  
36 with locally enhanced primary production by phytoplankton. However, the composition of bacterial  
37 communities in the frontal zone remains poorly understood. In this study, we investigate how a coastal  
38 tidal front in Brittany (France) structures the free-living bacterioplankton communities in a spatio-  
39 temporal survey across four cruises, five stations and three depths. We used 16S rRNA gene surveys  
40 to compare bacterial community structures across 134 seawater samples and defined groups of co-  
41 varying taxa (modules) exhibiting coherent ecological patterns across space and time. We found that  
42 bacterial communities composition was strongly associated with the biogeochemical characteristics of  
43 the different water masses and that the front act as an ecological boundary for free-living bacteria.  
44 Seasonal variations in primary producers and their distribution in the water column appeared as the  
45 most salient parameters controlling heterotrophic bacteria which dominated the free-living community.  
46 Different dynamics of modules observed in this environment were strongly consistent with a  
47 partitioning of heterotrophic bacterioplankton in oligotroph and copiotroph ecological strategies.  
48 Oligotroph taxa, dominated by SAR11 Clade members, were relatively more abundant in low  
49 phytoplankton, high inorganic nutrients water masses, while copiotrophs and particularly opportunist  
50 taxa such as *Tenacibaculum sp* or *Pseudoalteromonas sp* reached their highest abundances during the  
51 more productive period. Overall, this study shows a remarkable coupling between bacterioplankton  
52 communities dynamics, trophic strategies, and seasonal cycles in a complex coastal environment.

53

## 54 1 Introduction

55 Bacteria dominate the marine environment in abundance, diversity and activity where they  
56 support critical roles in the functioning of marine ecosystems and oceanic biogeochemical cycles  
57 (Falkowski et al., Fenchel, and Delong 2008; Cotner and Biddanda 2002; Madsen 2011). In the coastal  
58 environment, they are closely associated with other planktonic organisms (e.g. viruses, phytoplankton  
59 and zooplankton) during the recycling of organic matter and inorganic nutrients through the so-called  
60 microbial loop (Azam and Malfatti 2007; Pomeroy et al. 2007). They form complex and highly  
61 dynamic assemblage (S. J. Giovannoni and Vergin 2012), with bacterioplankton diversity variations in  
62 space and time linked to changes in functional diversity (Galand et al. 2018). Therefore, understanding  
63 how the bacterioplankton composition varies in the environment remains one of the central question  
64 to elucidate so that we can better understand coastal ecosystem functioning (Fuhrman et al. 2015).

65 Phytoplankton development represents a major source of organic matter for heterotrophic  
66 bacteria in the water column. During their growth and upon bloom termination, algae release a complex  
67 bulk of dissolved organic matter that is almost only available for bacteria (Azam 1998; Fenchel and  
68 Jørgensen 1977). This organic matter processing requires diverse heterotrophic bacterioplankton  
69 among which one could find members of *Bacteroidetes (Flavobacteriaceae)*, *Roseobacter group*,  
70 *Gammaproteobacteria and Verrucomicrobia* (Buchan et al. 2014). These taxa contribute to the  
71 complexity of the marine ecosystem via different adaptive strategies, owing to the unequal access to  
72 their respective resource (Stocker 2012; Luo and Moran 2014). For instance, heterotrophs are generally  
73 distinguished as either oligotrophs or copiotrophs that compete at low and high nutrient concentrations  
74 respectively (Koch 2001; Giovannoni et al. 2014). They also present different degrees of ecological  
75 specialization, with generalist bacteria able to assimilate a broad variety of substrates, while specialists  
76 will compete for a narrow range of nutrients (Mou et al. 2008). Analysis of these ecological traits offer  
77 a simplified view of complex microbial communities and has gained interest for better understanding  
78 the dynamics of natural microbial communities and to gain insight into their role in the ecosystem  
79 (Krause et al. 2014; J. Raes et al. 2011; Haggerty and Dinsdale 2017).

80 Marine fronts are common mesoscale features in the ocean and are located at the transition  
81 between water masses of different physicochemical characteristics that actively shape the distribution  
82 of microbial organisms (phytoplankton, zooplankton and bacteria). Driven by currents and mixing,  
83 local nutrient input in the vicinity of the front generally enhances primary and secondary production,  
84 making the frontal zone an area of high biological importance (Olson and Backus 1985) and of critical  
85 influence on the microbial processing of organic matter (Baltar et al. 2015; Heinänen et al. 1995).  
86 However, the bacterial communities' composition involved in such dynamic systems remains to be  
87 investigated (Baltar et al. 2016).

88 The Ushant Front in the Iroise Sea (Brittany, France) is considered as a model for coastal tidal  
89 front (Le Fèvre 1986). Its position and characteristics are highly dynamic and influenced by  
90 atmospheric forcing and tidal currents which are strong in this area (Le et al. 2009). It occurs from May  
91 to October and leads to contrasting physicochemical environments with higher phytoplankton biomass  
92 at the frontal area (Le Fèvre et al. 1983). West of the front, stratification results in warmer oligotrophic  
93 surface waters and colder nutrient-rich deeper waters separated by a marked thermocline. East of the  
94 front, associated with highly variable conditions, permanently mixed coastal waters are characterized  
95 by an unlimited quantity of inorganic nutrients but with highly fluctuating conditions. These

96 contrasting water masses structure the distribution of primary producers with the dominance of small  
97 phytoplankton and dinoflagellates in surface stratified waters and diatoms in mixed waters (Birrien et  
98 al. 1991; GREPMA 1988; Videau 1987).

99 In this study, we investigated how such contrasting physicochemical and biological parameters will  
100 drive free-living bacterioplankton community structure. We then tested the hypothesis that the  
101 contrasted distribution of primary producers will select for bacteria with different adaptative strategies.  
102 To do so, using a network analysis we defined groups of co-varying bacterial OTUs that present the  
103 same dynamic across the samples, postulating that they may share the same ecological niches.

104

## 105 **Materials and Methods**

106

### 107 1. Study site and sampling design

108

109 For this study, we completed four east-west transects of about 60 km across the Iroise sea in September  
110 2014 (the 9th, 10th and 11th), March 2015 (the 10th, 11th and 12th), July 2015 (the 1st, 2nd and 3rd)  
111 and September 2015 (the 8th, 9th and 10th) aboard the R/V Albert Lucas. Station positions remained  
112 identical across cruises and were designed to span the front (Fig. 1A). Station **5** ( $48^{\circ}25\ N$ ,  $5^{\circ}30\ W$ ) is  
113 the most offshore station and is characterized by a strong stratification in summer. Station **4** ( $48^{\circ}25\ N$ ,  
114  $5^{\circ}20\ W$ ) and Station **3** ( $48^{\circ}20\ N$ ,  $5^{\circ}10\ W$ ) are closer the front. Station **2** ( $48^{\circ}15\ N$ ,  $5^{\circ}00\ W$ ) is present  
115 in the mixed area and Station **1** ( $48^{\circ}16\ N$ ,  $4^{\circ}45\ W$ ) is the most coastal station, near the outflow of the  
116 Bay of Brest. At each station, we obtained CTD profiles to assess the physical characteristics of the  
117 water column and establish the depth designed to capture the important biological and chemical  
118 features of the water column at the surface, bottom and Deep Chlorophyll Maximum (DCM).

119

120

### 121 2. Nutrients, phytoplankton counts and pigment analyses

122

123 Seawater was sampled at each depth for nutrients, biogenic silica (BSi), chlorophyll a (Chl a),  
124 particulate organic carbon and nitrogen (POC/PON) concentrations, and microscopic phytoplankton  
125 cell counts and identification. Dissolved inorganic phosphate (DIP) and silicate (DSi) concentrations  
126 were determined from seawater filtered on Nucleopore membrane filters (47 mm) and dissolved  
127 inorganic nitrogen (DIN) on Whatman GF/F filters (25 mm). Filters for DIN and DIP were then frozen,  
128 whereas filters for DSi were kept at 4 °C in the dark. Concentrations were later determined in the  
129 laboratory by colorimetric methods using segmented flow analysis (Auto-analyser AA3HR Seal-  
130 Analytical) (Aminot and K  rouel 2007). BSi was determined from particulate matter collected by  
131 filtration of 1 L of seawater through 0.6 µm polycarbonate membrane filter. The analysis was  
132 performed using the alkaline digestion method (Ragueneau and Tr  guer 1994). Total Chl a and  
133 phaeopigments (Pheo) were determined in particulate organic matter collected on 25 mm Whatman  
134 GF/F filters. The filters were frozen (-20°C) and analyzed later by a fluorometric acidification

135 procedure in 90% acetone extracts (Holm-Hansen et al. 1965). Particulate material for POC/PON  
136 measurements was recovered on pre-combusted (450 °C, 4 h) Whatman GF/F filters. Samples were  
137 then analyzed by combustion method (Strickland and Parsons 1972), using a CHN elemental analyzer  
138 (Thermo Fischer Flash EA 1112). Phytoplankton samples were fixed with Lugol's solution and cell  
139 counts were carried out using the Lund et al. (1958) method (Lund et al. 1958).

140

141

### 142 3. Bacterioplankton community sampling

143

144 Seawater for bacterial diversity analysis was collected at different depths using Niskin bottles and  
145 directly poured in 5 L sterile carboys previously rinsed three times with the sample. For each depth,  
146 three biological replicates samplings were done, consisting of three different Niskin deployments,  
147 except Station 4 in September 2014. Filtrations started in the on-shore laboratory 3 to 4 hours after  
148 sampling. Water samples were size-fractionated using three in-line filters of different porosity: 10 µm,  
149 3 µm (PC membrane filters, Millipore®) and 0.22 µm (Sterivex® filters, with PES membrane). In this  
150 study, we focused only on the 0.22-3 µm free-living fraction. All filters were frozen in liquid nitrogen  
151 before storage at -80 °C. In total, 2 to 5 liters of seawater were filtered each time, depending on filter  
152 saturation.

153

154

### 155 4. DNA extraction and sequencing

156

157 Half-filters were directly placed in Matrix B® tubes (filled with 0.1 mm silicate beads, MP  
158 Biomedicals) with lysis buffer (Tris pH 7.0, EDTA pH 8.0 and NaCl), SDS 10% and Sarkosyl 10%  
159 and subjected to physical lysis for 5 minutes on a vortex plate. The liquid fraction was then collected  
160 in order to perform a phenol-chloroform extraction using PCI (Phenol, Chlorophorm Isoamyl alcohol  
161 with a ratio of 25:24:1) and a precipitation step with isopropanol and 5 M sodium acetate. DNA was  
162 resuspended in 100 µL of sterile water.

163 Bacterial diversity was assessed by targeting the v4-v5 hypervariable regions of 16S rDNA with the  
164 primers 518F (CCAGCAGCYGCGGTAAN) / 926R (CCGTCAATTCNTTTRAGT-  
165 CCGTCAATTTCTTTGAGT - CCGTCTATTCCTTTGANT) (Nelson et al. 2014). PCR products  
166 were purified using Ampure XP® kit and DNA quantity was measured using Picogreen® staining and  
167 a plate fluorescence reader (TECAN® infinite M200 Pro). Each sample was diluted to the same  
168 concentration and pooled before sequencing on an Illumina MiSeq sequencer at the Marine Biological  
169 Laboratory (Woods Hole MA, USA).

170

171

### 172 5. Bioinformatics analysis

173

174 We obtained 22,523,398 raw reads, with a range from 31,583 to 2,001,687 reads per sample. Reads  
175 were merged and quality-filtered according to recommendations in Minoche et al. (Minoche et al.,

176 2011) using the Illumina-utils scripts (Eren et al. 2013). Those steps removed 17 % of all the sequences.  
177 We used the Swarm algorithm (Mahé et al. 2014) to cluster the 18,876,655 remaining sequences into  
178 1,611,447 operational taxonomic unit OTUs. Chimera detection was done using vsearch *de novo*  
179 (Rognes et al. 2016) resulting in 281,825 filtered OTUs. Taxonomic annotation for each OTU was  
180 done with the SILVA database v123 (Quast et al. 2012) using the mothur classify.seqs command  
181 (Schloss et al. 2009). 19,516 OTUs (1,247,339 sequences) were affiliated with non-bacterial taxa (e.g.  
182 archaea, chloroplasts, mitochondria, eukaryota) and removed from the dataset. As a result, we obtained  
183 262,308 OTUs for 134 samples, with a large number of singletons (205,292 OTUs), that were kept  
184 along the statistical analysis.

185

186

## 187 6. Statistical analysis

188

189 Libraries were normalized for read number using DESeq package in R (Anders and Huber 2010). A  
190 visualization of the bacterial community structure similarity was assessed with an NMDS based on  
191 Bray-Curtis dissimilarities using vegan R package (Oksanen et al. 2007). Influence of depth, station  
192 and sampling time on bacterial communities was investigated using a permutational multivariate  
193 analysis of variance (PERMANOVA) based on Bray-Curtis dissimilarities and 999 permutations. We  
194 used principal coordinate analysis (PCA) ordination to characterize water samples environmental  
195 parameters. Finally, we examined the presence of biomarker OTUs (i.e. OTUs with a significant higher  
196 relative abundance in a given condition) of the different stations within each cruise using the LefSe  
197 software (Segata et al. 2011).

198

## 199 7. Network analysis

200

201 Network analysis was done based on a truncated matrix only containing the most abundant OTUs with  
202 more than 50 sequences in at least three samples. The filtered matrix contained 681 OTUs that  
203 accounted for 79% of all the sequences. This filtering step was necessary to avoid false positive  
204 correlations in the network analysis. Our module detection analysis followed part of the pipeline of  
205 Chafee *et al.* (Chafee et al. 2018). First, the application of SPIEC-EASI computes pairwise co-variance  
206 based correlation between OTUs in order to address the sparsity and composition issues inherent to  
207 microbial abundance data (Kurtz et al. 2015). The possible interaction was then inferred using the  
208 glasso probabilistic inference model with a lambda.min.ratio of 0.01. Based on this model, only the  
209 significant covariance values were extracted and transformed into a correlation matrix using the  
210 cov2cor() function.

211 We then delineated network modules as groups of highly interconnected OTUs that presented very  
212 close variations in relative abundances across the studied samples. Those modules were defined using  
213 the Louvain algorithm (Blondel et al. 2008). We used Gephi (Bastian et al. 2009) and Force Atlas 2  
214 layout algorithm to generate a visualization of the main correlations (>0.3). Module eigengenes (ME,  
215 the first principal component, considered a representative of the OTUs distribution of a given module)  
216 were calculated based on the relative abundance matrix using WGCNA function moduleEigengenes()  
217 (Langfelder and Horvath 2008). Those ME were used to calculate Pearson correlations between the

218 different modules and the environmental variables, using the WGCNA commands `moduleTraitCor()`,  
219 `moduleTraitPvalue()`.

220 We used the methods proposed by Newton et al. to infer the trophic strategy of OTUs within each  
221 modules (Newton and Shade 2016). In that study, they differentiated opportunist taxa with high  
222 abundance variability in space and time, and marathoners with low abundance variability. They  
223 distinguished these using the Coefficient of Variation (CV) of OTU dynamic combined with their  
224 abundance and prevalence in the samples. Based on their method, we defined high CV opportunist  
225 OTUs (or HCV) as OTUs above 5% upper boundary of the linear modeling 95% confidence interval  
226 in a CV plot against OTU occurrence. Conversely, we defined low CV marathoners OTUs (or LCV)  
227 as OTUs below the 5% lower boundary of the linear modeling. In our analysis, OTU occurrence was  
228 defined by a relative abundance  $> 0.05\%$ .

229  
230 All the scripts used for this analysis are available at: <https://loimai.github.io/BBobs/>

231

## 232 **Results**

### 233 1. Environmental settings

234 This study was carried out between September 2014 and September 2015 in the Iroise Sea, off  
235 Brittany (Northeast Atlantic), in the vicinity of the Ushant island. The Ushant front position and  
236 characteristics were estimated using Satellite Surface Temperature (SST) maps (Fig. 1) and CTD data  
237 collected at the dates of sampling (suppl. data 1). The March 2015 sampling took place before the onset  
238 of the Ushant front and presented a homogeneous temperature around 10 °C across all the stations.  
239 High nutrient concentrations at all stations ( $\text{Si(OH)}_4$ : 1.82 to 4.38  $\mu\text{M}$ , nitrates: 5.88 to 12.13  $\mu\text{M}$ ,  
240 suppl. data 2), low surface Chl a ( $<1 \mu\text{g.L}^{-1}$  except for Station 1, suppl. data 3) and overall low  
241 phytoplankton cells counts observed during this cruise (suppl. data 2) indicated that sampling occurred  
242 prior to the development of the phytoplankton spring blooms. In summer (September 2014, July 2015  
243 and September 2015), SST maps showed a sharp transition between coastal and offshore temperatures  
244 confirming the presence of a frontal system. The different observed stratification regimes (suppl. data  
245 1) coincided with distinct physicochemical patterns and phytoplanktonic patterns across seasons (Fig.  
246 1B, suppl. data 2). In late summer (September samples), offshore deep waters shared close  
247 characteristics with winter waters (March sample, Temperature: 11.7 to 12.3 °C), with few  
248 phytoplankton cells, similar concentration of  $\text{Si(OH)}_4$  (1.61 to 3  $\mu\text{M}$ ) and nitrates (3.99 to 6.41  $\mu\text{M}$ ).  
249 Conversely, surface waters presented high temperatures (14.7 to 18.2 °C), a nutrient depletion  
250 ( $\text{Si(OH)}_4$ : 0.02 to 1.65  $\mu\text{M}$ ,  $\text{NO}_2+\text{NO}_3$ : 0.00 to 0.09  $\mu\text{M}$ ) and high phytoplankton cell counts. In early  
251 summer (July), inorganic nutrient concentrations were overall lower than in September, probably  
252 consumed by the spring bloom coinciding with the onset of the front around May-June. A significant  
253 bloom occurred at stations 2 and 3, on the 27th of June, five days before the sampling period as seen  
254 in the satellite surface Chl a observations (suppl. Data 3 B.).

255



## 256 2. Bacterial community dynamics in the Iroise Sea

257 In the Iroise Sea, free-living bacterial communities structure presented clear time and spatial  
258 patterns as shown in the NMDS ordination plot (Fig. 1C, for each cruise separately see suppl. Data 4).  
259 Firstly, community structure displayed strong seasonality, with sample primarily grouping by sampling  
260 cruises. We also observed similar communities in stratified waters of September 2014 and 2015  
261 samples (Fig. 1C), suggesting that these seasonal pattern were recurring .

262 Besides, we also observed an important spatial influence, as within each sampling cruise,  
263 communities were markedly different between stations. In winter samples, communities were highly  
264 similar throughout the water column (Depth was non-significant, Permanova  $\text{Pr}( > f ) = 0.225$ ), but  
265 presented a coastal to offshore gradient (Station was significant, Permanova  $\text{Pr}( > f ) = 0.001$ , suppl. Data  
266 4) that followed a gradient of salinity, temperature and inorganic nutrient (sup. data 2). Conversely, in  
267 September, when the front was the strongest, free-living bacterial communities were much more  
268 heterogeneous with clear patterns associated to the set-up of the front : stratification resulted in highly  
269 diverging communities with depth (Depth  $\text{Pr}( > f ) = 0.004$ ), while this pattern was not seen for the mixed  
270 and coastal samples (Depth was not significant,  $\text{Pr}( > f ) = 0.222$  and  $0.434$ , suppl. Data 4). In July, the  
271 patterns were mostly associated with depth for all the stations, except station 1 (suppl. Data 4).

272 Overall, bacterial communities structure followed the different water masses, mirroring environmental  
273 physicochemical variations represented in the PCA plot of environmental variables (Fig. 1B): deep  
274 water communities in the summer crews remained more similar to winter communities (Fig. 1C). In  
275 contrast, surface bacterial communities in stratified regimes and in most of the stations in July departed  
276 from this typical winter structure. In addition, for these summer cruises, the DCM samples either cluster  
277 with surface or deep samples. This could be explained by the difficulty to sample the DCM precisely,  
278 as its depth can vary between the CTD measurements and the biological sampling.

279 Using the LefSe algorithm (Segata et al. 2011), we found that some stations displayed specific  
280 biomarker OTUs within each sampling time (suppl. Data 5). Biomarkers such as OTU1 (*Amylibacter*)  
281 and OTU4 (*Planktomarina*) were found at the most near-shore station (Station 1) in September 2014  
282 and March 2015 and other biomarkers were found in surface stratified waters in early summer (OTU29  
283 *NS4 marine group*) and late summer (OTU10 *Synechococcus*) cruises, but no biomarker was identified  
284 for the mixed and frontal stations except in July 2015 at Station 2 with OTU35 (*Aliivibrio*) and OTU74  
285 (*Pseudoalteromonas*).

286

## 287 8. Description of modules

288 We further investigated groups of co-varying OTUs that could potentially share the same  
289 ecological niches in this partitioned marine environment via a co-occurrence network analysis. In the  
290 inferred network, we were able to identify 14 sub-networks (Fig. 2A and 2B) defining groups of OTUs  
291 (termed modules) with similar distribution patterns across the entire study. Two modules (5 and 4)  
292 were dominant in our dataset and respectively accounted for 32.7% and 20.9% of all sequences. Eight  
293 modules represented between 1.6% and 14.2%, and four were rarer with less than 1% abundance. The  
294 OTU taxonomy in each module is summarized in suppl. Data 6 and Fig. 4 presents the distribution of  
295 the dominant families among each module.

296 Since its eigengene could characterize each module, we investigated to which extent module  
297 distribution could be correlated (i.e. Pearson correlation) with environmental parameters (Fig. 2C).  
298 Correlation patterns partitioned modules into three significant sub-networks. The first one mostly  
299 comprises modules 5 and 13 representing 32.7 and 6.8% of the dataset respectively that correlated  
300 positively to inorganic nutrient concentrations and negatively to temperature and Particulate Organic  
301 Matter (POC, PON) values. Their relative abundance in the different samples showed that they were  
302 dominant in oligotrophic waters: together they ranged from 39% at Station 1 to 72% at Station 5 of the  
303 late winter communities, and 54% to 59% of the deep stratified waters in September (Fig. 3). SAR11  
304 Surface 1 (46.7% of the module sequences), *ZD0405 marine group* (12%) and SAR86 Clade (6.4%)  
305 dominate the main module (module 5). Module 13 showed a different diversity as being dominated by  
306 members of *Marinimicrobia* (15.8%), *SAR11 Deep 1* (12%), *SAR11 Surface 1* (8.9%) and  
307 *Salinisphaeraceae* (7.9%).

308 Modules 4, 8, 2 and 1 contributed to the second sub-network, representing 20.9, 14.2, 5.5 and  
309 5.4% of the dataset respectively. They presented clear inverse correlations compared with modules of  
310 the first group: they correlated positively with temperature, POM and Diatoms but negatively with  
311 inorganic nutrient concentrations. In relative abundance, they dominated the samples of July and in the  
312 surface and coastal stations. Their taxonomy was distinct from the first group; i.e. together they  
313 gathered the majority of *Flavobacteriaceae* (70%) and *Rhodobacteraceae* (88%). Also, module 9 was  
314 enhanced in July 2015 but was almost exclusively present in Station 2 and strongly correlated with  
315 ammonium (0.72,  $p < 0001$ ). Its taxonomy was also particular as dominated by *Vibrionaceae* (58%)  
316 and *Pseudoalteromonadaceae* (30%).

317 The third sub-network (modules 6 and 3) was less abundant and presented different correlations  
318 with environmental parameters. They respectively account for 1.6 and 2.6% of the dataset, were  
319 correlated with specific phytoplankton groups (dinoflagellates, nanophytoplankton) and were  
320 dominant in the surface and DCM of stratified waters in September where they made up to 34% of the  
321 community. Among them, module 3 was constituted almost exclusively of Cyanobacteria (94%)  
322 affiliated with the *Synechococcus* genus.

323 To examine whether the modules gather OTUs that exhibited different ecological behavior (i.e.  
324 marathoners or opportunist), we computed the coefficient of variation (CV) for all OTU used in the  
325 network analysis and plotted this metric against OTU occurrence as proposed in Newton et al. (Newton  
326 and Shade 2016). We then examined their module membership (suppl. Data 7 and Fig. 4B): modules  
327 showed a clear gradient ranging from relatively low CV (LCV) OTUs (modules 4, 11, 14, 5 and 13) to  
328 high CV (HCV) OTUs (modules 9, 10, 7 and 2).

329

## 330 Discussion

331

332 *The spatial and temporal dynamics of bacterial communities associated with a coastal tidal front*

333

334 In this study, we investigated the spatiotemporal variations of free-living bacterial community  
335 composition over one year across contrasting coastal water masses characterized by a seasonal tidal  
336 front structure. For each cruise (season) we observed a spatial influence on bacterioplankton  
337 community composition. However, in the studied area, winter communities were comparatively much  
338 more homogeneous with only an influence from coast to the offshore. At this time of the year, the river  
339 debit were at the highest and this is known to influence free-living bacterial communities in coastal  
340 area (Tréguer et al. 2014; Pizzetti et al. 2016). In summer, bacterial communities diverged according  
341 to the station and depth, highlighting the importance of the onset of a tidal front and open ocean  
342 stratification for bacterioplankton community structure. This was mainly due to the sharp divergence  
343 of summer surface samples, as deep summer samples remained closer to winter ones (Fig. 1C). The  
344 existence of two contrasting types of bacterioplankton dynamics (winter and deep vs. summer surfaces)  
345 reflects similar patterns in sample ordination based on biogeochemical parameters and can be  
346 associated to phytoplankton development, temperature increase or lowering in inorganic nutrient  
347 concentrations, which are key parameters influencing bacterial communities dynamic in marine coastal  
348 environment (Fuhrman et al. 2015 ; Giovannoni and Vergin 2012).

349 These observations of a frontal zone as a sharp ecological transition for bacterioplankton  
350 among different water masses are consistent with reports in stratified waters (Cram et al. 2015;  
351 Ghiglione et al. 2008) and from other frontal areas (Baltar et al. 2016 ) confirming their role as  
352 ecological boundaries in the oceans (Raes et al. 2018). Interestingly, the frontal area itself around  
353 station 3-4 did not exhibit a specific bacterial composition (Fig. 2C and Fig 4), but could rather result  
354 from mixing between adjacent communities, as already suggested for the phytoplankton and  
355 zooplankton (Sournia 1994) and in another marine boundaries such as a shelf break (Zorz et al. 2019).  
356 We could not identify any biomarker OTUs associated with frontal stations. Hence, bacterioplankton  
357 communities associated with enhanced productivity at the Ushant front (Videau 1987) could result  
358 from an increase in resident plankton density (Franks 1992) rather than the emergence of distinct  
359 communities.

360 In addition, we observed a remarkable seasonal recurrence between the two September cruises,  
361 within each station and depth. This pattern was comparatively more accentuated toward the open ocean  
362 (station 4 and 5) than near shore (stations 1 to 3). This temporal dynamic is coherent with previous  
363 reports of seasonal patterns in single stations as shown in the North Sea (Chafee et al. 2018) or the  
364 English Channel (Gilbert et al. 2012) time series. Our results suggest that despite a substantial  
365 heterogeneity in our system, such seasonal recurrence patterns also extend to stations along  
366 geographical and bathyal gradients.

367

368

369 *Distinct bacterioplankton trophic strategies associated with tidal front temporal and spatial dynamics*

370 Using network analysis on samples exhibiting strong temporal and spatial variations, we could  
371 define modules of ecologically coherent OTUs. As suggested elsewhere, such modules can be

372 considered as OTUs potentially sharing the same environmental niches with coherent ecological  
373 strategies (Eiler et al. 2012). In association with previously characterized dominant taxa in each  
374 module, we aim at identifying higher order bacterioplankton community organization and revealing  
375 ecological drivers of community dynamics in the frontal zone.

376 Our study gives evidence of module partitioning into two major heterotrophs-dominated sub-  
377 networks, exhibiting distinct inverse covariance patterns (Fig. 2A and 2B). Their correlations with  
378 environmental variables also clearly separated conditions between water masses with active primary  
379 producers (low nutrients and high POM values) that were present in summer surface samples, and  
380 conditions with overall low development of phytoplankton typical of winter and deep samples.  
381 Development and decay of phytoplankton lead to the release of dissolved organic matter, which is  
382 almost only available for heterotrophic bacteria (Azam and Malfatti 2007). Thus, these two different  
383 conditions likely selected for copiotrophic or oligotrophic heterotrophic free-living bacteria adapted to  
384 different concentrations of organic matter in the environment (Giovannoni et al. 2014). The taxonomic  
385 affiliation was highly coherent with these observations. Modules 5 and 13 are dominated by SAR11  
386 clades, which typically thrive in basal concentrations of organic matter (Morris et al. 2002) or members  
387 of taxa typical of the sub-euphotic zones such as *Marinimicrobia*, or SAR11 Deep 1 (Agogue et al.  
388 2011). Conversely, modules 4, 8, 2, 1 and 9 were dominated by members of the *Rhodobacteraceae* and  
389 *Flavobacteriaceae*, typically associated with phytoplankton-derived organic matter (Buchan et al.  
390 2014; Teeling et al. 2012). Thus, the availability of phytoplankton-derived organic matter will drive  
391 heterotroph dynamics.

392 The last sub-network with module 3, dominated by *Synechococcus*, highlights the contribution  
393 of phototrophic bacteria. These cyanobacteria became strongly dominant in the surface of well  
394 stratified-waters in late summer, which is already known to favor these small phytoplanktonic cells  
395 (Taylor and Joint 1990; Cadier et al. 2017).

396 Interestingly, known chemoautotrophs involved in the nitrification such as *Nitrospinaea* did  
397 not form a separate module and were included in the oligotrophic module 13. Nitrification in the water  
398 column is typically found in the sub-euphotic zone as nitrite-oxidizing bacteria could be light-sensitive  
399 (Lomas and Lipschultz 2006) and can be outcompeted by phytoplankton for the uptake of nitrite (Smith  
400 et al. 2014; Wan et al. 2018). This fact could explain why they exhibit a dynamic similar to oligotrophic  
401 bacteria.

402

403

#### 404 *Differential bacterioplankton responses to organic matter availability*

405

406 Using the OTU coefficients of variation to distinguish rather marathoners or opportunist taxa,  
407 we examined whether heterotrophic bacteria had different responses to organic matter availability in  
408 summer samples. Their dynamic and composition was highly coherent with a partitioning of the  
409 microbial loop between different taxa (Bryson et al. 2017). For instance, several modules presented a  
410 majority of HCV OTUs (Fig. 4B): *Tenacibaculum sp.* (a *Flavobacteriaceae* dominating Module 2) was  
411 highly abundant reaching up to 20.7% of the sequences in the 2015 July surface waters shortly after  
412 significant phytoplankton bloom. One single *Tenacibaculum* OTU dominated the surface of 4 stations  
413 over a 40km distance just a few days after a phytoplankton bloom. This dynamic of *Tenacibaculum*

414 genus is coherent with previous observations of a recurrent population increase after seasonal  
415 phytoplankton blooms in coastal water (Teeling et al. 2016) while being part of the rare biosphere  
416 otherwise (Alonso-Sáez, Díaz-Pérez, and Morán 2015). During the same period, module 9, dominated  
417 by *Vibrionales* (*Aliivibrio sp*) and *Alteromonadales* (*Pseudoalteromonas sp.*) class, was especially  
418 abundant at Station 2, reaching up to 27% of the sequences (Fig. 3). These taxa are known copiotroph  
419 bacteria that can sharply increase in abundance in response to high substrate loading (Tada et al. 2012).  
420 These results strongly support an opportunist lifestyle of bacteria in modules 2 and 9. When they are  
421 dominant, bacteria with such an opportunist strategy have a critical role as they can contribute to a  
422 substantial fraction of organic matter recycling (Pedler et al. 2014).

423 In contrast, module 4 was mostly composed of LCV OTUs with a cosmopolitan distribution  
424 across time, depth and geography. The dominant family in module 4, the *Rhodobacteraceae*, are known  
425 as generalist bacteria with a versatile substrate utilization (Moran et al. 2004; R. J. Newton et al. 2010).  
426 Some of the free-living *Rhodobacteraceae* grow on low molecular weight organic compounds and  
427 numerous studies pointed out that they compete for the same substrate as SAR11 Clade, but are more  
428 competitive at higher organic matter concentrations. They can be very successful in coastal waters, an  
429 example being a representative of the genus *Planktomarina* that is dominant in the North Sea (Voget  
430 et al. 2015). Other dominant taxa of this module, such as SAR86 Clade, *Methylophilaceae* or members  
431 of SAR116 Clade are widespread in marine environments and are also known to target small organic  
432 molecules (Dupont et al. 2012; Huggett et al. 2012; Oh et al. 2010).

433 Module 8 presented mixed CV values among its OTUs. Many members of this module, such  
434 as *Flavobacteriaceae* (Buchan et al. 2014), *Verrucomicrobiacea* (Martinez-Garcia et al. 2012) or  
435 *Cryomorphaceae* (Bowman 2014), can use complex algal-derived substrates and present a typical  
436 increase in abundance after phytoplankton blooms (Teeling et al. 2016). We interpret these  
437 observations to be a result of the relatively low sampling time resolution that probably fails to resolve  
438 rapid OTU variations of other opportunistic bacteria.

439 Overall, this study allowed us to delineate several groups of bacterioplankton trophic strategies and  
440 their variations in response to biogeochemical cycles in a highly dynamic coastal environment.  
441 Oligotrophs, typically represented by SAR11 OTUs, were present in all water masses. They  
442 outcompete other heterotrophic organisms in low labile organic matter environments such as winter  
443 and deep offshore water masses, while copiotrophs communities develop during the high productivity  
444 period. Within copiotrophs, several taxa show patterns of adaptation to rapidly changing conditions in  
445 their substrate availability: opportunist taxa can rapidly become highly dominant (represented by  
446 *Tenacibaculum* and *Alteromonadales*) after local events such as phytoplankton bloom, while more  
447 generalist taxa such as *Rhodobacteraceae* exhibit a ubiquitous distribution. It is likely that this  
448 dichotomy between oligotroph and copiotroph is an over-simplification in trophic strategies, as  
449 suggested by numerous studies (Bryson et al. 2017; Mayali et al. 2012). Heterotrophs could instead  
450 follow a gradient from oligotrophy to copiotrophy, and only finer analysis such as growth rate  
451 measurements or dissolved organic matter quantification could help to better describe the trophic  
452 strategies of heterotrophs (Kirchman 2016). However, these complex heterotrophic community  
453 dynamics observed in this study highlight the central role of free-living bacteria in organic matter  
454 cycling. Indeed, previous studies of the Iroise Sea nutrient cycling demonstrated that after the initial  
455 depletion of winter nutrients pool, phytoplankton growth strongly relies on nutrients recycled through

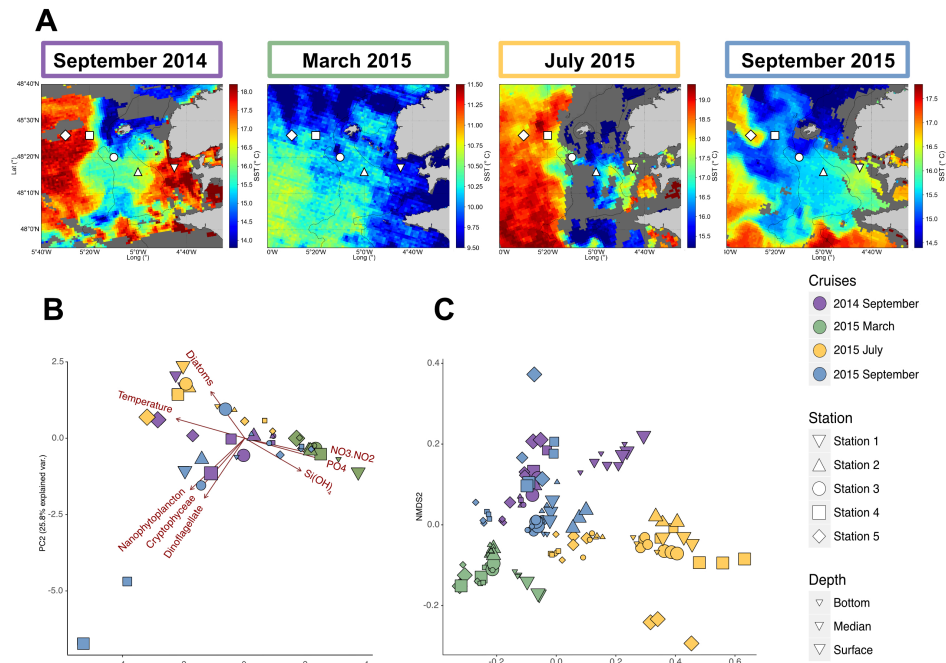
456 the microbial loop (LHelguen et al. 2005), through which they are remineralized several times in a  
457 seasonal cycle (Birrien et al. 1991). In this system, bacteria could be directly responsible for up to 25%  
458 of urea with the remaining originating from ciliates mainly, which also mostly feed on bacteria  
459 (LHelguen et al. 2005)

460

461 **Conclusion**

462 Here, using a complex 3D (vertical, horizontal, and temporal) survey of free-living bacterial diversity  
463 in correlation with the seasonal dynamics of a coastal tidal front, we have shown that such complex  
464 mesoscale features controls the dynamic of free-living bacterial communities. The Ushant tidal front  
465 acts as an ecological boundary for those communities. We illustrated the link between different water  
466 masses on this dynamic with for instance photosynthetic bacteria enriched in oligotrophic stratified  
467 waters in late summer. Using a network analysis, we were able to gain insight on ecologically coherent  
468 modules of free-living bacteria driven by the availability of organic matter produced by the  
469 phytoplankton. This has major implication for our understanding of the microbial loop in coastal  
470 systems. In the future, these observations should be confirmed with measurements and characterization  
471 of dissolved organic matter quantity and quality in the water column as well as direct measurement of  
472 growth rates in order to more precisely describe bacterial trophic strategies. Moreover, to better  
473 understand these systems, future studies should include the characterization of other key members of  
474 the planktonic communities such as viruses, ammonium oxidizing archaea, eukaryotic phytoplankton  
475 and grazers, as well as particle-attached bacteria that are central for organic matter and biogeochemical  
476 cycles.

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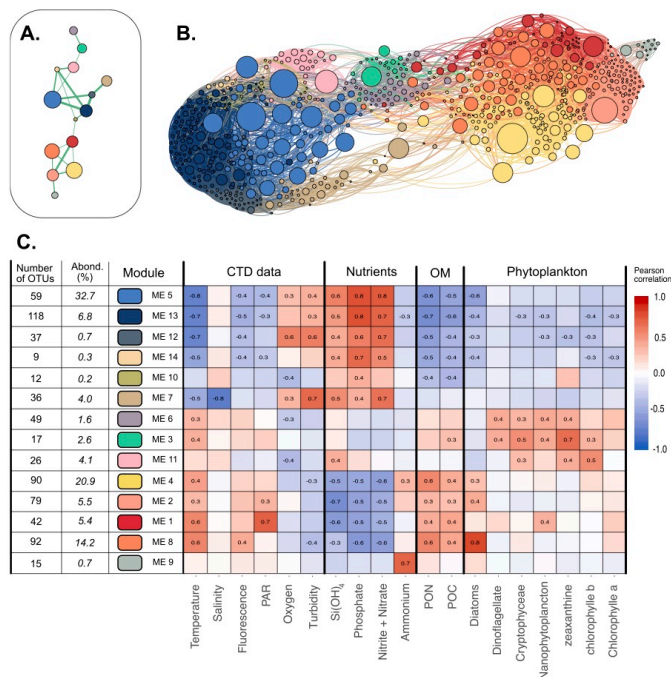


479

480 **Figure 1 | Environmental and bacterial communities characteristics of the Iroise Sea for the different**  
 481 **cruises.**

482 **A.** Map of the Satellite Sea Surface Temperatures (SST) in the Iroise Sea for each cruise (September 2014 the  
 483 8th, March 2015 the 6th, July 2015 the 2<sup>nd</sup> and September 2015 the 8th), highlighting in the summer cruises the  
 484 presence of the Ushant front that separates offshore warm stratified waters and coastal cooler mixed waters. The  
 485 shape corresponds to the different stations sampled, with at each time 2 or 3 depths. Surface waters are highly  
 486 dynamic and some eddies are conspicuous during September 2015, bringing cold water into surface stratified  
 487 waters. Thus SST map are not enough to define the frontal area and vertical profiles are needed to characterize  
 488 the different water masses (suppl. Data 1) **B.** Principal Component Analysis of the environmental characteristics  
 489 for the different samples (n=49) based on temperature, inorganic nutrients ( $Si(OH)_4$ ,  $NO_3+NO_2$ ,  $PO_4$ )  
 490 concentrations and microscopic count of the different phytoplanktonic groups (*Diatom*, *Nanophytoplankton*,  
 491 *Cryptophyceae* and *Dinoflagellate*) **C.** NMDS of the bacterial diversity based on Bray-Curtis dissimilarities  
 492 among the different cruises, stations and depth (stress=0.091). For bacterial diversity, biological triplicates were  
 493 done for each station and depth, explaining the higher number of samples (n=135).





494

495

**Figure 2 | Visualization of the different modules and their main characteristics**

496

**A.** Representation of the main connectivity between the different modules detected with the Louvain algorithm.

497

Each node represent a module, each edge represents the median of the positive correlations between two

498

modules. Only the strongest connections are shown.

499

**B.** Network visualisation of the dataset. Each node represents an OTU, while each edge represent a positive

500

correlation (>0.3) obtained from the covariance matrix calculated with SPIEC-EASI. The node size depends on

501

the abundance of one OTU (in number of sequences) in the entire dataset. The edge size depends on the value

502

of the correlation. Nodes colors represent the OTU affiliation to the different Louvain communities. Network

503

was represented using Gephi and Force Atlas layout algorithm. **C.** Presentation of the 14 modules detected.

504

Pearson correlations between each module eigengene and the different environmental data are presented in the

505

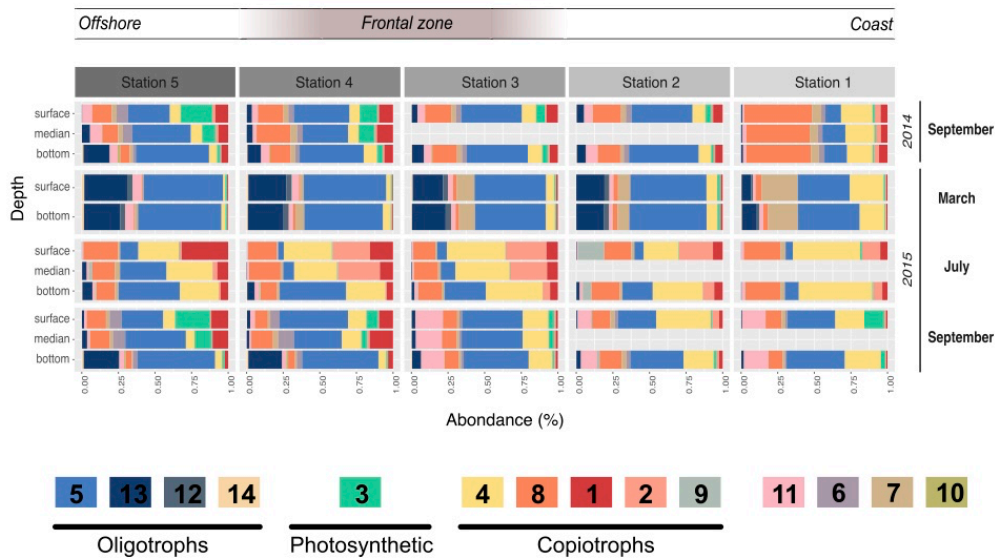
heatmap. OM : Organic Matter, PAR: Photosynthetically Available Radiation, PON : Particulate Organic

506

Nitrogen, POC: Particulate Organic Carbon. Only the significant correlations (p.value > 10<sup>-4</sup>) are shown. The

507

number of OTUs in each module and the relative abundance of each module (in % of all the reads) are detailed.

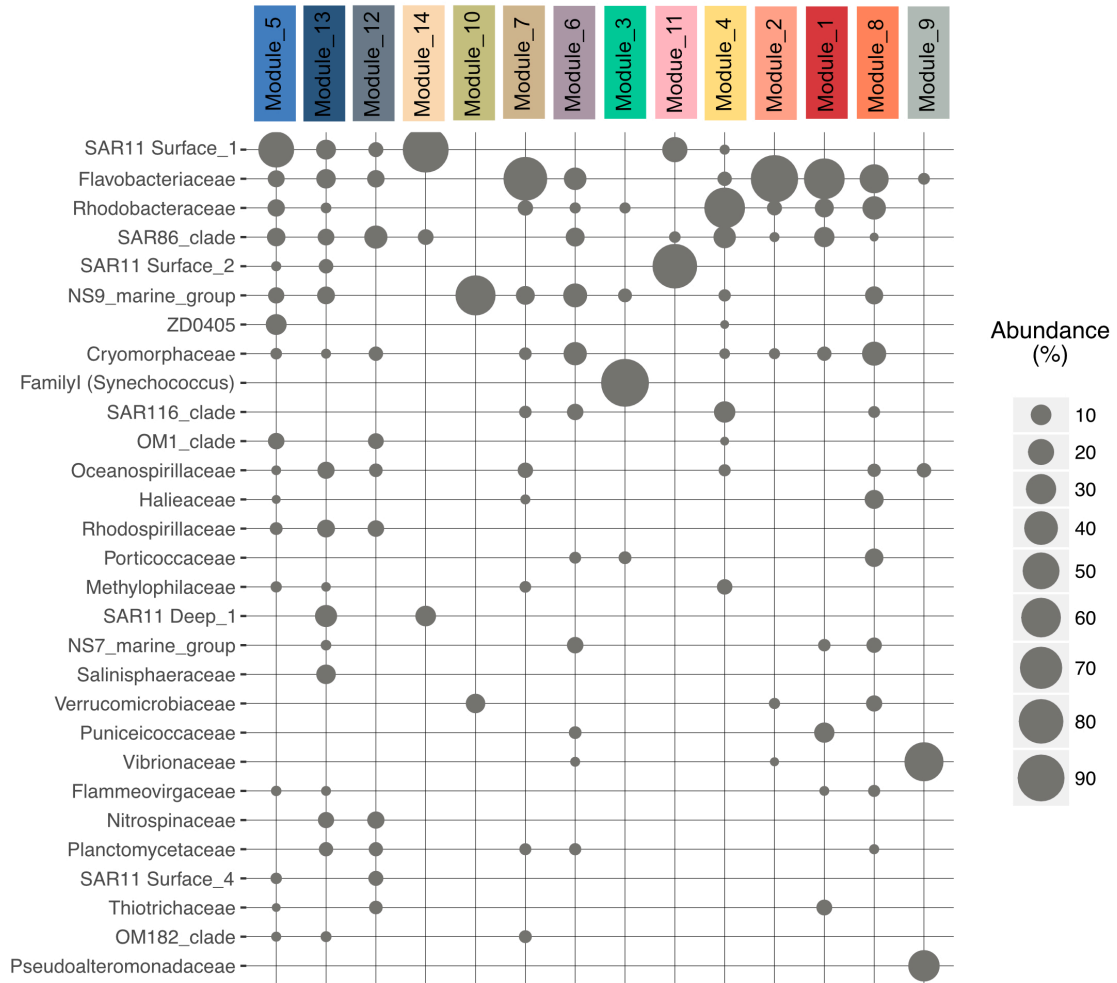


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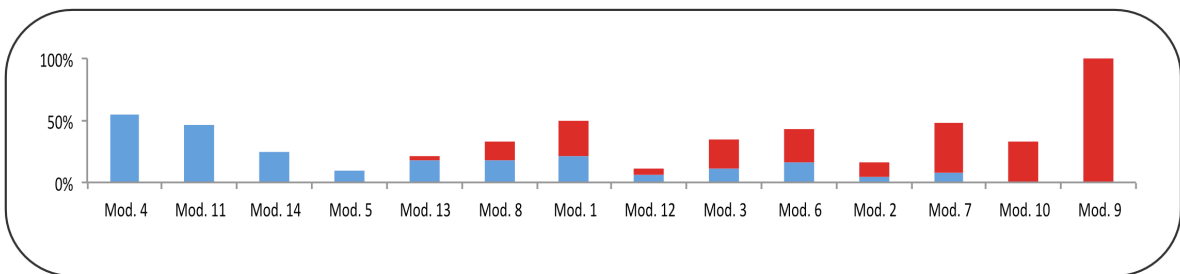
509 **Figure 3 | Module dynamics across the different campaigns, stations and depths**

510 Relative abundance of the different modules in each campaign, station and depth. The value shown resulted of  
 511 the mean of the triplicates done for each sampling. The legend includes the different hypotheses of the modules'  
 512 trophic strategies based on their correlations to the environmental parameters, their taxonomy and their dynamic.  
 513 Oligotrophic modules are likely to present heterotrophic bacteria adapted to low organic matter levels, while  
 514 copiotrophic bacteria are likely more competitive under higher organic matter levels. Finally the photosynthetic  
 515 module is almost only composed of Cyanobacteria.  
 516

A.



B.



518 **Figure 4 | Modules' taxonomy and Coefficient of Variation of their OTUs**

519 A. Relative abundance of the 30 main families in each module. B. Proportion of OTUs (in % of all the OTUs  
520 in a module) that present a High Coefficient of Variation (HCV) in red, or a Low Coefficient of Variation  
521 (LCV) in blue defined based on Newton method (Newton et al., 2016).

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