
Evaluation of FluoroProbe® performance for the phytoplankton-based assessment of the ecological status of Mediterranean coastal lagoons

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Abstract :

The European Water Framework Directive and several other legislations worldwide have selected phytoplankton for monitoring the ecological status of surface waters. This assessment is a complicated task in coastal lagoons due to their intrinsic variability, prompting moves to use real-time measurements. Here, we tested the ability of the submersible spectrofluorometer FluoroProbe® to accurately estimate the phytoplankton biomass and to efficiently discriminate spectral groups in Mediterranean coastal lagoons, by using sub-surface water samples (n = 107) collected at Biguglia lagoon (Corsica) in different environmental situations (salinity and trophic state) from March 2012 to December 2014. We compared the estimates of biomass and phytoplankton group composition obtained with the FluoroProbe® (in situ and lab measurements) with the spectrofluorimetrically measured biomass and HPLC-derived quantifications of pigment concentrations. FluoroProbe® provided good estimates of the total phytoplankton biomass (particularly, the lab measurements). The FluoroProbe® data were significantly correlated with the HPLC results, except for the in situ measurements of very weak concentrations of blue-green and red algae. Our findings indicate that factory-calibrated FluoroProbe® is an efficient and easy-to-use real-time phytoplankton monitoring tool in coastal lagoons, especially as an early warning system for the detection of potentially harmful algal blooms. Practical instructions dedicated to non-specialist field operators are provided. A simple and efficient method for discarding in situ measurement outliers is also proposed.

Keywords : Phytoplankton, Monitoring, Mediterranean coastal lagoons, FluoroProbe®, HPLC

37 **Introduction**

38 Mediterranean coastal lagoons are increasingly exposed to nutrient enrichment, mainly driven by urbanization,
39 tourism and agricultural activities (Justic et al. 1995; Flo et al. 2011). Nutrient enrichment disturbs the ecosystem
40 metabolism and the structure of the native aquatic communities (Pasqualini et al. 2017), intensifies the eutrophication
41 of coastal waters (Nixon 1995; Glibert 2017), and favors the occurrence of harmful algal blooms (Collos et al. 2004;
42 Spatharis et al. 2007; Smayda 2008; Heisler et al. 2008; Cecchi et al. 2016; Glibert and Burford 2017). In response to
43 the degradation of water resources, the European Union Water Framework Directive 2000/60/EC (EU-WFD) was
44 put in place with the aim of maintaining and improving water quality. According to the EU-WFD, the phytoplankton
45 metrics required for defining and classifying the ecological status of transitional waters are: biomass (chlorophyll *a*),
46 community structure (composition and species abundances), and algal bloom frequency and intensity.
47 Chlorophyll *a* concentration is an integrative measure of the phytoplankton community responses to nutrient
48 enrichment (Giovanardi et al. 2018), and is generally quantified using a spectrofluorometer (e.g., Neveux and
49 Lantoiné 1993). However, with such deferred lab-methods, the diagnosis is not immediately available. Moreover,
50 they are intrusive (use of solvent) and labor-intensive. The eutrophication-driven increase in chlorophyll *a* is always
51 accompanied by changes in the phytoplankton community structure (Bec et al. 2011). The Utermöhl method
52 (Utermöhl 1958; CEN EN 15204 2006), which is based on traditional cell counts with an inverted microscope, is the
53 only method that allows the taxonomical resolution at the species level for the largest species. This method remains
54 widely used for the routine monitoring of potentially toxic taxa. However, taxonomic identification requires
55 experienced analysts (Vuorio et al. 2007), is time-consuming, and is debated for cells smaller than ~ 10 µm. Pigment
56 analysis by High-Performance Liquid Chromatography (HPLC) is an alternative option that provides a highly
57 reproducible and relatively rapid bulk estimate of the major phytoplankton groups present in a sample (Wright and
58 Jeffrey 1997; Schlüter et al. 2014; Leruste et al. 2016) by detecting diagnostic marker pigments, known as accessory
59 pigments (Vidussi et al. 2001; Johnsen and Sakshaug 2007; Marty et al. 2008; Bel-Hassen et al. 2009). All the
60 phytoplankton cell sizes of a sample are considered, including picophytoplankton and mesoplankton (Leruste et al.
61 2015). However, the lack of pigment specificity for some groups led to couple HPLC with optical microscopy to
62 identify dominant and sub-dominant species (Havskum et al. 2004). Flow cytometry was first used to analyze the
63 abundant picophytoplankton cells (Marie et al. 1997; Jacquet et al. 1998; Bec et al. 2005), and technical
64 developments have extended the cell size range up to hundreds of microns. However, this method does not provide

65 detailed insights on the sample taxonomic composition. FlowCAM theoretically allows combining flow cytometry
66 capabilities and microscopy accuracy, but significant methodological improvements are still required (See et al.
67 2005; Álvarez et al. 2011, 2014; Romero-Martínez et al. 2017). Recently, molecular techniques (metabarcoding)
68 have been developed for inventorying taxa in environmental samples, and they can efficiently identify rare, unknown
69 or hidden microorganisms (Grzebyk et al. 2017). However, even promising, such approaches are not yet adapted to
70 frequent and rapid monitoring because they are still in development and lack standardized procedures, they remain
71 time-consuming and expensive, and will require inter-calibration efforts with classical methods. Moreover, they
72 remain very sensitive to laboratory protocols and statistical analysis strategies (Eckford-Soper et al. 2018), and
73 require better supplemented DNA reference libraries (Rivera et al. 2018; Hering et al. 2018).

74 Beutler et al. (2002) designed a multi-wavelength probe called FluoroProbe[®] (BBE-Moldaenke, GmbH). This is a
75 submersible spectrofluorometer that can assess chlorophyll biomass and distinguish the main phytoplankton classes
76 by using the auto-fluorescence properties of pigment-containing micro-organisms. FluoroProbe[®] contains five light-
77 emitting diodes (450, 525, 570, 590 and 610 nm) for excitation of the accessory pigments associated with the
78 photosystem-II antenna system of phytoplankton, and one 370 nm diode for excitation and subsequent subtraction of
79 the fluorescence from Chromophoric Dissolved Organic Matter (CDOM). This technology is now widely applied in
80 very different environments by scientists and freshwater resource managers for lab and *in situ* studies of
81 phytoplankton communities (Rolland et al. 2010; Catherine et al. 2012, 2016; Švanys et al. 2014; Patidar et al. 2015;
82 Maloufi et al. 2016; Morgan-Kiss et al. 2016; Poxleitner et al. 2016; Silva et al. 2016; Wang et al. 2016; Blottière et
83 al. 2017; Cyr, 2017; Karpowicz and Ejsmont-Karabin, 2017; Giling et al. 2017; Teufel et al. 2017). FluoroProbe[®]
84 main advantage is the immediacy of the performed measurements, although it does not give the same level of
85 taxonomic precision as microscopy-based methods, and may significantly underestimate the biomass when
86 phytoplankton communities are very dense (i.e., chlorophyll *a* >250 µg.L⁻¹, Wang et al. 2016), or in samples with
87 very low biomasses (Bradie et al. 2018). Moreover, the relationships between fluorescence, chlorophyll *a* value and
88 the proportion of the different phytoplankton groups could be disturbed by strong variability linked to the
89 phytoplankton community structure, environmental conditions, physiological status, quenching, and irradiance
90 history of individual algal cells, making the conversion to phytoplankton biomass sometimes imprecise (Lawrenz et
91 al. 2010; McIntyre et al. 2010; Catherine et al. 2012; Escoffier et al. 2015; Blottière et al. 2017). The quality of the
92 provided estimates is also strongly dependent on the instrument calibration (factory or home-made settings), which is

93 generally performed using laboratory-grown phytoplankton strains that might not exactly reflect the environmental
94 conditions of natural samples (Harrison et al. 2016; Silva et al. 2016). Nevertheless, McIntyre et al. (2010) reported
95 that the relationships between fluorescence intensity and chlorophyll *a* are robust when the fluorescence variation
96 range is large enough, and when there is qualitative agreement between the phytoplankton classes defined by their
97 spectral fluorescence signatures and by other analytical methods.

98 Few studies have used FluoroProbe[®] measurements for the analysis/monitoring of coastal and transitional waters.
99 See et al. (2005) found that in the Gulf of Mexico, diatom and dinoflagellate contributions to the whole
100 phytoplankton biomass were comparable when using FluoroProbe[®] and HPLC. Similar results were reported by Liu
101 et al. (2012) by comparing FluoroProbe[®] estimates, HPLC results, and microscope observations in the Yellow Sea,
102 except when the biomass was very low (i.e., <1 µg.L⁻¹). van Beusekom et al. (2009) concluded that by combining the
103 FluoroProbe[®] technology and microscopy, the small-scale plankton dynamics (biomass and taxonomic spectrum)
104 could be fully captured in the Baltic Sea. Richardson et al. (2010) showed that in United States estuaries, the
105 estimates of biomass and taxonomic structure of phytoplankton obtained using the Algae Online Analyzer (a spectral
106 fluorometer, like FluoroProbe[®]) were consistent with those derived from HPLC-measured marker pigments. Houliez
107 et al. (2012) confirmed that FluoroProbe[®] and flow cytometry allowed obtaining similar *Phaeocystis globosa* annual
108 dynamic estimates in the coastal waters of the eastern English Channel. More recently, Gordon et al. (2016)
109 efficiently documented the temporal changes of phytoplankton assemblages associated with a shift from oligohaline
110 to hyperhaline conditions in the St Lucia Lake (South Africa) using a FluoroProbe[®]. However, to our knowledge, no
111 study has investigated FluoroProbe[®] performance for assessing the long-term phytoplankton dynamics in euryhaline
112 Mediterranean coastal lagoons, which often are characterized by very variable trophic statuses.

113 Here, we tested FluoroProbe[®] ability to estimate the phytoplankton biomass and to discriminate the phytoplankton
114 spectral groups, by using sub-surface water samples (N=107) collected in the Biguglia lagoon (Corsica) in very
115 different environmental situations (salinity and trophic state) over 3 years. For this purpose, we compared
116 FluoroProbe[®] estimates of total biomass with the spectrofluorimetrically measured chlorophyll *a* values, and the
117 FluoroProbe[®] *in situ* and lab estimates of the taxonomic structure with HPLC-based identification of marker
118 pigments of the studied phytoplankton communities. Microscope counts were available only for a small fraction of
119 the samples (see Garrido et al. 2016), and were not included in the study. Moreover, the main end-users of our results
120 (i.e., managers of Mediterranean coastal lagoons) rarely have the required taxonomical expertise. Indeed, this study

121 wants to provide a simple, quick and efficient tool for non-specialists who need to characterize and monitor the
122 ecological status of the water bodies they supervise.

123

124 **Materials and methods**

125 Sampling site and sample preparation

126 Samples were collected in the Biguglia lagoon (Corsica, France, Mediterranean Sea; [Fig. 1](#)). This shallow coastal
127 lagoon is a confined ecosystem that is disturbed by growing eutrophication and that exhibits a wide range of
128 environmental conditions and trophic states documented since the 1980s ([Bec et al. 2011](#); [Garrido et al. 2013, 2016](#);
129 [Pasqualini et al. 2017](#)). From 2000, a very sharp drop of salinity has been observed in the entire lagoon, associated
130 with a gradual closing of the natural grau (sea side inlet) and a simultaneous increase (controlled or not) in the
131 freshwater input. This hydrological context affected the water quality: accumulation of nutrients supplied by run-off
132 and reduced possibility of dilution by seawater. This situation alerted the public authorities in charge of the
133 management and protection of this ecosystem, and remediation efforts have been undertaken to improve the water
134 quality of the Biguglia lagoon, mainly through the management of water fluxes: artificial openings of the grau in the
135 North, and increased freshwater entrance in the South. The effects of such hydrological manipulations have been
136 monitored through a long-term survey of the phytoplankton communities.

137 Grab samples were collected regularly at six different sampling points ([Fig. 1](#)), from March 2012 to December 2014.

138 All sampling points were deep enough for the *in situ* use of the FluoroProbe[®]. Water samples (n = 107) were
139 collected in a Niskin-like bottle and immediately distributed in opaque 1 L flasks. Two samples were systematically
140 collected at each location, one \approx 50 cm below the surface and a second \approx 50 cm above the sediment (in order to avoid
141 contamination by bottom resuspension), with the aim of tracking the contrasting phytoplankton communities that
142 may inhabit the two horizons (see for example Table 3 in [Garrido et al. 2013](#) and [Fig. 7](#) in [Garrido et al. 2016](#)).

143 Sample preparation and analyses were performed within two hours after sample collection. For each sample, salinity,
144 temperature, dissolved oxygen, and turbidity were measured *in situ* with a portable multi-parameter meter (YSI
145 Environmental Monitoring Systems). The concentration of dissolved inorganic nutrients was measured in aliquots
146 filtered through Whatman GF/F filters (0.7 μ m nominal pore size) using the methods described by [Holmes et al.](#)
147 ([1999](#)), [Raimbault et al. \(1999\)](#) and [Aminot and K erouel \(2007\)](#).

148

149 FluoroProbe® measurements

150 Phytoplankton biomass ($\mu\text{g eq. Chl } a.L^{-1}$) was assessed *in situ* and at lab using a FluoroProbe®. This device can
151 differentiate between “spectral” groups of phytoplankton on the basis of their relative chlorophyll *a* fluorescence
152 intensity at 680 nm (due to the photosystem-II core pigments; [Beutler et al. 2002](#); [Leboulanger et al. 2002](#); [Gregor
153 and Maršálek, 2004](#)). The instrument segregates phytoplankton in xanthophyll-containing brown algae
154 (dinoflagellates, diatoms and chrysophyceae) (525 nm), green algae (chlorophyceae, euglenophyceae and
155 prasinophyceae) rich in chlorophyll *a* and *b* (450 nm), blue-green algae (phycocyanin-rich cyanobacteria) (590 and
156 610 nm) and red algae (phycoerythrin-rich cyanobacteria and cryptophyceae) (570 nm; [Beutler et al, 2002](#)). The
157 system detects a set of characteristic fingerprints that are used by dedicated software (version 2.2.6) to calculate the
158 relative contributions of each algal class to the global phytoplankton biomass ([Beutler et al. 2002](#)). The UV LED
159 (370 nm) that was added to the latest generation of devices is used to measure the fluorescence of CDOM, and their
160 eventual influence is corrected by the software. Given the very contrasting phytoplankton communities potentially
161 associated to the different environmental conditions of our dataset, the FluoroProbe® was used with the factory
162 calibration. The factory fingerprints were obtained by the manufacturer from repeated fluorescence measurements
163 using 35 laboratory cultures that involved several species per spectral algal group (green: *Scenedesmus* sp.,
164 *Chlamydomonas* sp., *Monoraphidium* sp., *Chlorella* sp., *Miractinium* sp.; blue: *Microcystis* sp., *Synechococcus* sp.,
165 *Aphanizomenon* sp., *Anabaena* sp.; brown: *Cyclotella* sp., *Nitzschia* sp., *Synedra* sp., *Ceratium* sp., *Peridinium* sp.;
166 mixed: *Cryptomonas* sp.). The associated Chl *a* amounts were determined by HPLC. Samples were previously
167 adapted to the measuring light for 90 s and fluorescence measurements were performed at 20°C. For more details,
168 see [Beutler et al. \(2002\)](#). This corresponds also to the simplest way of using this tool by non-specialists.

169 The *in situ* measurements using the FluoroProbe® were performed at 50 cm below the surface and at 50 cm above the
170 sediment. In the laboratory, FluoroProbe® measurements were done using 25 mL fractions of each sample placed in
171 the original optical glass cuvette (25×25×70 mm) fitted into the device. Samples were systematically dark-adapted
172 for 20min before measurements ([Beutler et al. 2002](#)). The total chlorophyll *a* concentration and the equivalent
173 concentrations of chlorophyll *a* for each phytoplankton class identified on the basis of the fluorescence signals were
174 calculated with the FluoroProbe® software.

175

176 Spectrofluorimetric quantifications

177 Chlorophyll *a* was quantified using 100 ml of each sample filtered through Whatman GF/F filters (25 mm diameter;
178 0.7 µm nominal pore size) under gentle vacuum, and immediately stored at -20°C after filtration. The used filters
179 were ground in 90% acetone and extracted in the dark at 4°C for 24h. Pigment content was then measured by
180 spectrofluorimetry (Perkin–Elmer LS50b) and calculated according to [Neveux and Lantoiné \(1993\)](#).

181

182 HPLC analysis

183 Pigments were analyzed by HPLC as described in [Leruste et al. \(2015, 2016\)](#). In the laboratory, 1 L of each sample
184 was filtered through Whatman GF/F filters (47 mm diameter) under low-vacuum pressure (<100 mm Hg) and
185 immediately stored at -20°C, or at -80°C when possible. Pigments were extracted with 2.5 ml of 100% methanol in
186 the dark at 4°C for 5min. Samples were then sonicated 5 times for 10s (20 Watts) spaced by 10s in ice to avoid
187 excessive heating of the extracts. After 10min in the dark at 4°C, extracts were filtered on cellulose acetate filters
188 (0.45 µm pore size) to remove cell debris. An aliquot of 600 mL of each sample was diluted with 150 mL of Milli-Q
189 water, and then 150 mL of this dilution was injected in the HPLC system (Waters HPLC Alliance D600). Pigment
190 extracts were analyzed using the method described by [Wright and Jeffrey \(1997\)](#) with a flow rate of 1 ml.min⁻¹ and a
191 run duration of 29min. The HPLC system was calibrated with external standards (DHI Water and Environment,
192 Hørsholm, Denmark). Chromatograms were extracted at 440 nm, and pigments were identified by comparison with a
193 spectral library established from the pigment standard database and by checking the elution order and absorption
194 spectra ([Roy et al. 2011](#)), using the software Empower Pro 3. Each peak was checked and the baseline readjusted to
195 minimize errors due to noise. Pigments were then quantified by comparing the peak area with the standard
196 calibration curves (mg.L⁻¹).

197 To establish the correlations between HPLC quantifications and FP results, a series of marker pigments were used:

198 (i) peridinin for dinoflagellates and chrysophyceae, and fucoxanthin for diatoms and chrysophyceae, (ii) chlorophyll
199 *b*, prasinoxanthin, lutein, violaxanthin and neoxanthin for chlorophyceae, euglenophyceae and prasinophyceae, (iii)
200 zeaxanthin for cyanobacteria, and (iv) alloxanthin for cryptophyceae ([Vidussi et al. 2001](#); [Marty et al. 2008](#); [Bel-](#)
201 [Hassen et al. 2009](#)).

202

203 Statistical analyses

204 Linear regressions were performed to assess the relationship between the phytoplankton biomass values deduced
205 from the *in situ* and lab FluoroProbe® measurements, and those quantified using a spectrofluorometer and HPLC
206 analyses. The Chi-2 tests and Kruskal-Wallis one-way analyses of variance on ranks were used to compare outliers
207 and ‘regular’ data. Bravais-Pearson correlation coefficients, regression slopes and the associated hypothesis tests
208 were then carried out to detect (i) correlation and regression slope coefficients different from 0, which indicated a
209 significant linear correlation between measurements, and (ii) regression slope coefficients different from 1, which
210 indicated a lack of quantitative agreement between measurements. Cases where, for example, the FluoroProbe®
211 measurements exceeded the spectrofluorometer or HPLC measurements are illustrated by the location of most
212 observations below the 1:1 regression line. All statistical analyses were done with the R software (R Core Team,
213 2015). The probability threshold value was fixed at $p = 0.01$.

214

215

216 **Results**

217 *In situ* and lab phytoplankton biomass measurements with the FluoroProbe®

218 The dataset encompassed very different environmental situations (Table 1) that are fully representative of conditions
219 encountered in euryhaline Mediterranean lagoons throughout the year (Souchu et al. 2010; Bec et al. 2011; Leruste et
220 al. 2016; Le Fur et al. 2018;), although hypertrophic conditions ($\text{Chl } a > 100 \mu\text{g.L}^{-1}$), which sometimes occur in such
221 systems during summer (Derolez et al. 2019) were not observed during this survey.

222 The *in situ* and lab FluoroProbe® values of 15 of the initial 107 water samples were considered *a posteriori* as
223 outliers (Fig. 2). These 15 outliers did not reveal any specific pattern or peculiarity when compared with ‘regular’
224 samples. Samples were collected from 2012 to 2014, and the FluoroProbe® measurement distribution of outliers over
225 time was not different compared with that of ‘regular’ data ($\text{Chi}2 = 2.078$; $p = 0.149$). Moreover, their presence could
226 not be explained by a seasonal effect, when considering the calendar ($\text{Chi}2 = 0.304$; $p = 0.582$) and the hydrological
227 seasons (Cecchi et al. 2016) ($\text{Chi}2 = 1.109$; $p = 0.292$). Similarly, the sampling site ($\text{Chi}2 = 2.059$; $p = 0.151$) and
228 sampling depth ($\text{Chi}2 = 0.327$; $p = 0.567$) also did not influence the occurrence of outliers. Finally, comparison of
229 the environmental conditions (temperature, salinity, turbidity, nutrient concentrations) and of the phytoplankton
230 biomass and the contribution of the four different spectral groups that constituted this biomass did not highlight any
231 significant difference between outliers and ‘regular’ samples ($p > 0.05$; Kruskal-Wallis one-way ANOVAs on ranks).

232 This suggests that there is a stochastic, but real risk of misleading FluoroProbe[®] measurements (with a frequency of
233 3 measurements in 20). It is not possible for a non-specialist to immediately identify outliers, whereas they are easily
234 graphically identifiable *a posteriori*. Therefore, we defined an empirical, but simple and efficient method for
235 discarding such data. After performing the *in situ* and lab FluoroProbe[®] measurements of one sample, if the
236 difference between the obtained values is higher than 50% of the *in situ* measurement, then the data should be
237 discarded.

238 After outlier elimination, the *in situ* and lab FluoroProbe[®] measurements of chlorophyll *a* were closely correlated
239 (Bravais-Pearson's rho = 0.861; regression slope = 0.875; p < 0.0001 in both cases, N = 92; Fig. 2), and the slope
240 coefficient was not statistically different from 1 (p = 0.023). The relationships between the chlorophyll *a*
241 concentrations provided by lab FluoroProbe[®] measurements and the spectrofluorimetric analyses were highly
242 significant (Fig. 3A). The Bravais-Pearson's rho (0.863) and regression slope (0.913) coefficients were significantly
243 different from 0 (p < 0.0001 in both cases), while the regression slope coefficient was not different from 1
244 (p = 0.124). Similar results were obtained for the *in situ* FluoroProbe[®] measurements (rho = 0.739), although a few
245 points were more dispersed, as indicated by the lower value of the regression slope (0.793; Fig. 3B), with a
246 coefficient that was significantly different from 1 (p = 0.008). Overall, the *in situ* FluoroProbe[®] measurements tended
247 to slightly, but systematically overestimate the concentrations of phytoplankton biomass.

248 In agreement, in the 15 outliers, the lab FluoroProbe[®] measurements were closely correlated with the
249 spectrofluorimetric and HPLC quantifications of chlorophyll *a* concentration (Spearman rank order correlation
250 coefficients: 0.807 and 0.868, respectively; p < 0.0001 for both), but not the *in situ* FluoroProbe[®] measurements
251 (Spearman rank order correlation coefficients: 0.404 and 0.611, p = 0.131 and 0.015, respectively).

252

253 Composition of phytoplankton assemblages

254 To evaluate FluoroProbe[®] ability to describe the composition of phytoplankton assemblages, we compared the
255 relationships between the biomass of each of the four discriminated groups calculated by FluoroProbe[®] and the
256 HPLC-measured concentrations of the accessory pigments specific to each group. For brown algae, the correlations
257 between HPLC (peridinin and fucoxanthin) and FluoroProbe[®] data (diatoms, dinoflagellates and chrysophyceae)
258 were significant (Fig. 4A & 4B). Both Bravais-Pearson's correlation coefficients (0.774 and 0.700 for lab and *in situ*
259 FluoroProbe[®] measurements, respectively) and linear correlation coefficients (0.599 and 0.490 for lab and *in situ*

260 measurements, respectively) were higher for lab measurements. The slope coefficients were different from 0 and also
261 from 1 ($p < 0.0001$ in both cases). Both approaches correctly described the main variations of brown algae
262 contribution to the whole biomass, although *in situ* FluoroProbe[®] data appeared more scattered than the lab values.
263 We observed exactly the same trends after excluding the dot corresponding to the highest biomass value (Bravais-
264 Pearson's correlation coefficients: 0.692 and 0.667 for lab and *in situ* FluoroProbe[®] measurements, respectively;
265 $p < 0.0001$ for both; linear correlation coefficients: 0.479 and 0.445 for lab and *in situ* measurements, respectively;
266 $p < 0.0001$ for both). For green algae also the correlations between HPLC (chlorophyll *b*, prasinoxanthin, lutein,
267 violaxanthin and neoxanthin) and FluoroProbe[®] data (chlorophyceae, euglenophyceae and prasinophyceae) were
268 very significant (Fig. 4C & 4D) (Bravais-Pearson's rho = 0.554 and 0.558, and linear regression coefficients = 0.307
269 and 0.311 for lab and *in situ* measurements, respectively), indicating that the two methods provided equivalent
270 information. The slope coefficients were different from 0 and also from 1 ($p < 0.0001$ for both). For red algae, the
271 correlations between HPLC (alloxanthin) and FluoroProbe[®] measurements (cryptophyceae) were again significant,
272 with stronger relationships for the lab FluoroProbe[®] estimates (Bravais-Pearson's rho = 0.533 and 0.489, and linear
273 regression coefficients = 0.285 and 0.240, for lab and *in situ* measurements, respectively; $p < 0.0001$ in all cases)
274 (Fig. 4E & 4F). The slope coefficients were statistically different from 0 and also from 1 ($p < 0.0001$ in both cases).
275 However, the exclusion of the most elevated biomass value from the dataset considerably reduced the strength and
276 significance of the correlations for lab measurements (Bravais-Pearson's correlation coefficient = 0.287, and linear
277 regression coefficient = 0.082) that, nevertheless, remained significant ($p = 0.006$ for both). This was no longer the
278 case for the *in situ* FluoroProbe[®] data (Bravais-Pearson's coefficient = 0.219, and linear regression coefficient =
279 0.048, $p = 0.037$ for both). The dispersion of data associated with the smallest biomasses (i.e., $< 1 \mu\text{g eq. Chl } a.L^{-1}$)
280 could be largely responsible of this loss of significance. For blue-green algae, the correlations were significant only
281 between HPLC (zeaxanthin) and lab FluoroProbe[®] measurements (cyanobacteria) (Bravais-Pearson's
282 coefficient = 0.423, and linear regression coefficient = 0.178; $p < 0.0001$ in both cases) (Fig. 4G), but not *in situ*
283 measurements (Bravais-Pearson's correlation = 0.164, $p = 0.119$, and linear regression coefficients = 0.028,
284 $p = 0.113$) (Fig. 4H). The points were largely dispersed relative to the regression line. Moreover, the concentrations
285 were very low (median concentrations: $0.05 \mu\text{g}.L^{-1}$ for HPLC-quantified zeaxanthin, and $0.06 \mu\text{g eq. Chl } a.L^{-1}$ for *in*
286 *situ* FluoroProbe[®]-assessed cyanobacteria), particularly when compared with those of the other taxonomic groups
287 (see concentration distribution for the different groups in Fig. 5), and were close to the detection limits of the

288 instrument. Overall, the relationships were generally more robust with lab than with *in situ* FluoroProbe®
289 measurements, although FluoroProbe® tended to systematically overestimate phytoplankton biomass.
290 Finally, analysis of the HPLC data on the phytoplankton assemblage composition revealed an important
291 heterogeneity (Fig. 5). The contribution of cyanobacteria varied between 0 and 67%, but remained almost always
292 marginal (median value = 3%). Cyanobacteria were mostly associated with small phytoplankton biomasses, as
293 indicated by the inverse relationship between the percentage of cyanobacteria and the chlorophyll *a* concentration
294 (by spectrofluorometer) (Spearman rank order correlation coefficient = -0.336, $p = 0.0011$). The contribution of
295 diatoms and dinoflagellates varied from 0 to 100%, with a median value of 49.8%. Diatoms and dinoflagellates were
296 often associated with elevated chlorophyll *a* biomasses (Spearman rank order correlation coefficient = 0.262,
297 $p = 0.0117$). The contributions of chlorophyceae and cryptophyceae (median values: 26% and 17%, respectively)
298 varied independently of the chlorophyll *a* concentration (Spearman rank order correlation coefficients = -0.0812 and
299 -0.0556; $p = 0.441$ and 0.598 , respectively), and were not linked one to the other (Spearman rank order correlation
300 coefficients = 0.101, $p = 0.338$), or with the abundance of cyanobacteria (Spearman rank order correlation
301 coefficients = 0.0418 and 0.0574, $p = 0.692$ and 0.586 , respectively). Conversely, their contributions were strongly
302 and inversely correlated with the variations in diatom and dinoflagellate assemblages (Spearman rank order
303 correlation coefficients = -0.626 and -0.602 for chlorophyceae and cryptophyceae respectively, $p < 0.0001$ for both).

304

305 Discussion

306 Water quality monitoring in coastal ecosystems is crucial owing to the multitude and magnitude of environmental
307 threats. In France, it is also an obligation registered in the EU-WFD. Phytoplankton taxonomy-based tools are
308 already available, but they require important laboratory efforts and a solid expertise, particularly for the correct
309 identification of organisms. Alternative approaches that exploit the fluorescence properties of algae, such as the
310 spectrofluorometer Fluoroprobe®, are now routinely used in freshwater ecosystems (Catherine et al. 2016; Švanys et
311 al. 2014; Patidar et al. 2015; Maloufi et al. 2016; Morgan-Kiss et al. 2016; Poxleitner et al. 2016; Silva et al. 2016;
312 Wang et al. 2016; Blottière et al. 2017; Cyr, 2017; Karpowicz and Ejsmont-Karabin, 2017; Giling et al. 2017; Teufel
313 et al. 2017). However, the performances of this instrument have never been formally assessed in Mediterranean
314 coastal ecosystems, where both environmental conditions and the composition of phytoplankton assemblages are
315 hugely variable and different from those of freshwater systems. To evaluate Fluoroprobe® performance

316 (phytoplankton biomass quantification and discrimination of the main phytoplankton classes) in such conditions, we
317 used 107 samples collected at six sites in the coastal Biguglia lagoon between March 2012 and December 2014. Our
318 aim was to provide valid outcomes for users who are not specialists in phytoplankton (ecology and taxonomy), but
319 are involved in the regular monitoring and follow-up of Mediterranean coastal aquatic ecosystems.

320 One of the first problems associated with this *a posteriori* approach was the presence of 15 outliers among the 107
321 FluoroProbe[®] measurements. Beyond their graphical identification, which remains subjective, we defined an
322 empirical and very simple method (i.e., a difference between the *in situ* and lab measurements higher than 50% of
323 the *in situ*) to allow the identification of such outliers by field users and managers who are the main end-users
324 targeted by this study. This implies that lab measurements must also be available. We failed to find any element to
325 explain the presence of outliers. The lab, but not the *in situ* FluoroProbe[®] measurements were closely correlated with
326 the spectrofluorimetric and HPLC quantifications of chlorophyll *a* in the 15 outliers. This suggests that *in situ*
327 measurements are primarily responsible for the observed discordances. Indeed, although the samples used for
328 laboratory analyses (i.e., lab FluoroProbe[®] measurement and HPLC and spectrofluorimetric quantifications) were
329 theoretically collected at the same depth where the *in situ* measurement was performed, depth discrepancies cannot
330 be excluded. This could lead to measurement differences in the case of vertical heterogeneity in the phytoplankton
331 community repartition. Such heterogeneity was previously observed (but not during the survey) in the Biguglia
332 lagoon, particularly during periods of important water fluxes that may generate the development of a salt wedge
333 ([Garrido et al. 2016](#)), and/or near the marine outlet of the lagoon where important vertical gradients might induce
334 significant stratifications of water masses and their phytoplankton communities ([Garrido et al. 2013](#)). Coastal
335 lagoons are by definition transitional ecosystems where both marine and freshwater circulate and often create
336 complex hydrodynamic patterns even at short spatial and time scales. The associated patchiness of phytoplankton
337 structures in such ecosystems has been known for decades ([Platt and Denman, 1980](#); [Therriault and Platt, 1981](#)), and
338 remains a challenging issue when working on discrete samples. Moreover the existence of very fine vertical
339 microstructures, particularly in the vicinity of important vertical gradients, should also be considered ([Viličić et al.](#)
340 [1989](#)), with possible phytoplankton decay and mortality that may considerably modify their chlorophyll content and
341 thus their fluorescence properties. Finally, turbidity maxima regularly occur at the halocline, which may locally
342 disturb the photosynthetic machinery with important consequences on microalgae fluorescence. Within such
343 microlayers, although the composition of phytoplankton assemblages is not directly involved, adaptive physiological

344 strategies, involving particularly photosynthesis and nutrient uptake, allow phytoplankton to cope with contrasting
345 environments, but affect its auto-fluorescence properties. This has been extensively documented in near-surface
346 microlayers, which were not involved here (sub-surface sampling was systematically performed at about 50 cm
347 below the surface). Depending on the vertical organization of the environmental parameters, particularly in the case
348 of strong thermal stratification (i.e., during summer), phytoplankton cells could be confined in water masses with
349 poor or even deleterious conditions for their metabolism. For instance, variations in light exposure might lead to
350 important physiological adaptations. Alternatively, concomitantly, the pigment ratio (amount of pigments and
351 quantity of active chlorophyll per cell) could vary considerably (see the third edition of the reference book of Kirk,
352 2011) and explain the observed discrepancies. However, such ecophysiological mechanisms seem improbable here,
353 because neither the depth where outliers were sampled (i.e., potentially involving light inhibition of sub-surface
354 samples), nor the season of their collection (e.g., potentially corresponding to stratification periods during summer)
355 appeared to affect the outlier occurrence. Lastly, *in situ* measurements were performed without any sample
356 preparation, whereas lab measurements were systematically done after sample adaptation to the dark for 20 minutes,
357 as classically recommended (Beutler et al, 2002). Such difference might justify important modifications in the auto-
358 fluorescence properties of phytoplankton, primarily due to changes in the redox states of photosynthetic chain
359 intermediates and, subsequently, of their respective fluorescence fingerprints. This hypothesis cannot be tested
360 because direct *in situ* dark-adaptation is impossible with the FluoroProbe[®] device.

361 Beyond the large variability of environmental conditions regularly encountered in such ecosystem (Table 1), the
362 natural phytoplankton assemblages sampled and used for testing the FluoroProbe[®] performance revealed also an
363 important heterogeneity (Fig. 5). Therefore, we may assume that almost all possible types of transitional water were
364 encountered at least once during sampling and were incorporated in the analyses.

365 The strong relationships between chlorophyll *a* concentrations provided by the FluoroProbe[®] measurements and
366 spectrofluorometer analyses are in agreement with literature data on lakes and reservoirs (Gregor and Maršálek,
367 2004; Gregor et al. 2005; Rolland et al. 2010; Catherine et al. 2012), and on marine and transitional waters (See et al.
368 2005; MacIntyre et al. 2010; Richardson et al. 2010; Houliez et al. 2012, 2017; Ostrowska et al. 2015). Errors in
369 absolute concentration assessments may arise from various causes. Differences in abiotic environment (e.g., light,
370 nutrient concentrations, CDOM; Lawrenz et al. 2010; MacIntyre et al. 2010) and phytoplankton community
371 composition (especially cell size) can affect the fluorescence responses, resulting in variations in the fluorescence-to-

372 chlorophyll *a* ratio (F^{Chl}) that is used to convert the fluorescence data into chlorophyll *a* units. In addition, as the
373 FluoroProbe[®] method measures the fluorescence emitted by physiologically active cells, the estimated values of
374 chlorophyll *a* concentration are influenced by the cell physiological status (active, senescent and/or lysed) that can
375 also affect their pigment content (MacIntyre et al. 2010). The correlations between FluoroProbe[®] measurements and
376 HPLC analyses were systematically strongly significant for brown, red and green algae (FluoroProbe[®]
377 overestimation), in line with the literature (Richardson et al. 2010; Rolland et al. 2010; Catherine et al. 2012), but not
378 for cyanobacteria. FluoroProbe[®] diagnostic efficiency appeared to be lower also for discriminating very weak
379 biomasses of cryptophytes. Diatoms and dinoflagellates are always well represented and often dominant in
380 Mediterranean lagoon waters (Ayadi et al. 2004; Bec et al. 2011; Carić et al. 2011; Garrido et al. 2016), as observed
381 here. Chlorophyceae and cryptophyceae tend to dominate phytoplankton communities in lagoon areas influenced by
382 freshwater inputs (Gregor and Maršálek, 2004; Catherine et al. 2012; Garrido et al. 2016). Cyanobacteria are
383 regularly sampled in Mediterranean lagoons, notably in the most degraded (towards hypertrophy) ecosystems (Bec et
384 al. 2005; Chomérat et al. 2007; Pulina et al. 2011). During the survey, the cyanobacteria biomass in the Biguglia
385 lagoon was always very small, close to the FluoroProbe[®] limit of detection value, although an impressive summer
386 bloom of cyanobacteria has already been reported in this ecosystem (up to 100 $\mu\text{g}\cdot\text{L}^{-1}$ in 2007, associated with the
387 massive development of the potentially toxic cyanobacteria *Anabaenopsis circularis*; Ifremer data). The annual
388 fluctuations of cyanobacteria abundance within Mediterranean coastal lagoons exhibit a strong seasonality that could
389 be related to water temperature, availability of nutrients and/or irradiance (Agawin et al. 1998; Bec et al. 2005;
390 Chomérat et al. 2007; Armi et al. 2010). The different causes (environmental and physiological) that may induce
391 interferences or excessive biases when assessing *in situ* cyanobacteria biomasses have been recently discussed by
392 Zamyadi et al. (2016), who concluded that '*bbe* [i.e. FluoroProbe[®]] has the smallest bias'. The environmental risks
393 and health hazards potentially associated with cyanobacteria are directly linked to the importance of their biomass
394 during proliferation events. The World Health Organization (WHO) indicated thresholds below which the risks
395 remain low to moderate (i.e., 50 $\mu\text{g Chl } a\cdot\text{L}^{-1}$ or 100,000 $\text{cells}\cdot\text{L}^{-1}$) (Chorus and Bartram, 1999). Within the context of
396 a monitoring network aware of such possible occurrences (as this is the case in the Biguglia lagoon), even relatively
397 imprecise biomass measurements are useful to alert about a possible change in cyanobacteria abundance, prompting
398 the collection of grab samples for microscopic analyses.

399 The discrepancies between FluoroProbe[®] assessments and HPLC analyses for some phytoplankton groups could
400 stem from the heterogeneity of such communities and the fact that the probe was factory-calibrated (species and
401 CDOM concentrations; see the Introduction part in [Twiss, 2011](#)). Correlations could be significantly improved by
402 calibrating in function of the representative species in the region under study ([Leboulanger et al. 2002](#); [Richardson et](#)
403 [al. 2010](#); [Houliez et al. 2012](#); [Kring et al. 2014](#)) and by factoring in CDOM concentrations ([Lawrenz et al. 2010](#)).
404 This would have been difficult here due to the heterogeneity of the many phytoplankton communities studied.
405 However, the ecological situation of the Biguglia lagoon has changed in recent years. The apparition and durable
406 maintenance of harmful dinoflagellates especially the species *Prorocentrum minimum* (Garrido et al. 2016) requires
407 special attention. In this new context, the re-calibration of the instrument with a reference specific of this species
408 would be perfectly justified in order to allow the effective monitoring of this noxious invader by managers. However,
409 one of the disadvantages of the FluoroProbe is that it is able to discriminate only four phytoplankton groups so that
410 the addition of a new fingerprint is only possible if one of the four default fingerprints is disabled. The selection of
411 this fingerprint is not obvious, and the added-value of this re-calibration has thus to be perfectly
412 assessed. Comparisons of the FluoroProbe[®] dataset against HPLC analyses may sometimes be skewed by the non-
413 specificity of some pigments for phytoplankton taxonomic compositions ([Richardson et al. 2010](#)). For example,
414 phycoerythrin-rich cyanobacteria are combined with red algae and this leads to weak correlations when comparing
415 results provided by these two methods ([Beulter et al. 2002](#); [Echenique-Subiabre et al. 2016](#)). On another note,
416 FluoroProbe[®] factory calibration may identify haptophytes as mixtures of brown algae and green algae (See et al.
417 2005; MacIntyre et al. 2010; Richardson et al. 2010; Houliez et al. 2012). However, haptophytes never constitute
418 significant populations within Mediterranean coastal lagoons (cf. Fig. 5 in [Leruste et al. 2016](#)), and noticeably in the
419 Biguglia lagoon ([Leruste et al. 2019](#)) where this potential bias remains thus negligible.

420 **Conclusions**

421 In Mediterranean coastal ecosystem, FluoroProbe[®] provided very good estimates of phytoplankton total biomass and
422 good indications of phytoplankton community composition, after the removal of outliers. Therefore, it can be used to
423 obtain information for water management in a timely manner. A clear advantage offered by this instrument is the
424 ability to differentiate phytoplankton assemblages into broad groups, thus making FluoroProbe[®] suitable for large-
425 scale surveys of lagoon environments using simply the factory calibrations. FluoroProbe[®] could be a useful
426 phytoplankton monitoring tool, especially as an early warning system for the detection of harmful algal blooms in

427 lagoon waters that constitute an increasing threat to Mediterranean transitional ecosystems (see Figure 1 in [Cecchi et](#)
428 [al. 2016](#)). FluoroProbe® could be used for the long-term monitoring of phytoplankton communities as required by the
429 EU-WFD, because it offers higher temporal resolution than the classical HPLC analysis, in real-time and at far lower
430 cost, with a rapid non-destructive measure of their relative abundance. This is a major advantage given the amount of
431 time and expertise required for the microscopy and HPLC analyses. The limited information obtained for some
432 groups (particularly, cyanobacteria) suggests that FluoroProbe® should be coupled with other methods to strengthen
433 the precision and quality of the observations. Specific calibrations will undoubtedly increase FluoroProbe®
434 diagnostic performances; however, we strongly recommended complementing FluoroProbe® measurements with less
435 frequent sample collections for microscopic analyses to accurately identify the phytoplankton taxa involved in the
436 spatial-temporal dynamics. Ultimately, users need to determine whether FluoroProbe® inherent error is acceptable
437 for their type of research or monitoring ([Kring et al. 2014](#)).

438 **Competing interests**

439 The authors declare that they have no conflict of interest.

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444

445 **Tables**

446 **Table 1** Environmental characteristics of the phytoplankton samples considered in the study (< *LoD*: below limit of
447 detection).

448

449 **Figure captions**

450 **Fig. 1** Situation map: the Biguglia lagoon and the sampling points (depths are indicated).

451 **Fig. 2** Relationship between lab and *in situ* FluoroProbe[®] measurements. Influential outliers (grey dots; n=15 out of
452 107 samples) were not included in the statistical analyses. The solid line corresponds to the 1:1 regression line and
453 the dashed line corresponds to the trend line from simple linear regression.

454 **Fig. 3** Relationship between spectrofluorimetric measurements and lab (A) and *in situ* (B) FluoroProbe[®] assessments
455 of the phytoplankton biomass. The solid line corresponds to the 1:1 regression line and the dashed line corresponds
456 to the trend line from simple linear regression.

457 **Fig. 4** Relationships between HPLC measurements of specific accessory pigments and lab and *in situ* FluoroProbe[®]
458 assessments of the biomass of the different phytoplankton groups (N = 92): xanthophyll-containing brown algae (A
459 & B); green algae rich in chlorophyll *a* and *b* (C & D); phycoerythrin-rich red algae (E & F) and phycocyanin-rich
460 blue-green algae (G & H) (N=92 samples). The scales of the X- and Y-axes are not the same for the different
461 phytoplankton groups. The solid line corresponds to the 1:1 regression line and the dashed line corresponds to the
462 trend line from simple linear regression.

463 **Fig. 5** HPLC-deduced composition of the phytoplankton assemblages of the 92 water samples considered in this
464 study (% of specific accessory pigments).

465

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713

714 **Table 1** Environmental characteristics of the phytoplankton samples considered in the study (< *LoD*: below limit of
715 detection).

716

717

	N	Mean	Min	Max
Temperature (°C)	137	18.5	6.9	30.1
Turbidity (NTU)	131	3.7	< <i>LoD</i>	35.4
Salinity	137	13.9	2.0	38.3
Dissolved Oxygen (%)	116	91.2	39.9	167.1
Ammonium (µM)	93	5.3	< <i>LoD</i>	48.2
Nitrate (µM)	93	8.1	< <i>LoD</i>	93.2
Phosphates (µM)	84	0.2	< <i>LoD</i>	1.3
Chl <i>a</i> (µg.L ⁻¹ , spectrofluorometer)	137	6.3	0.1	42.2

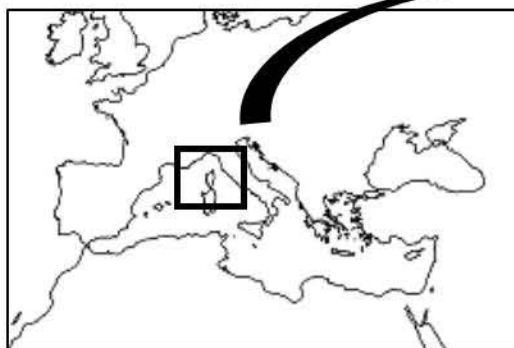
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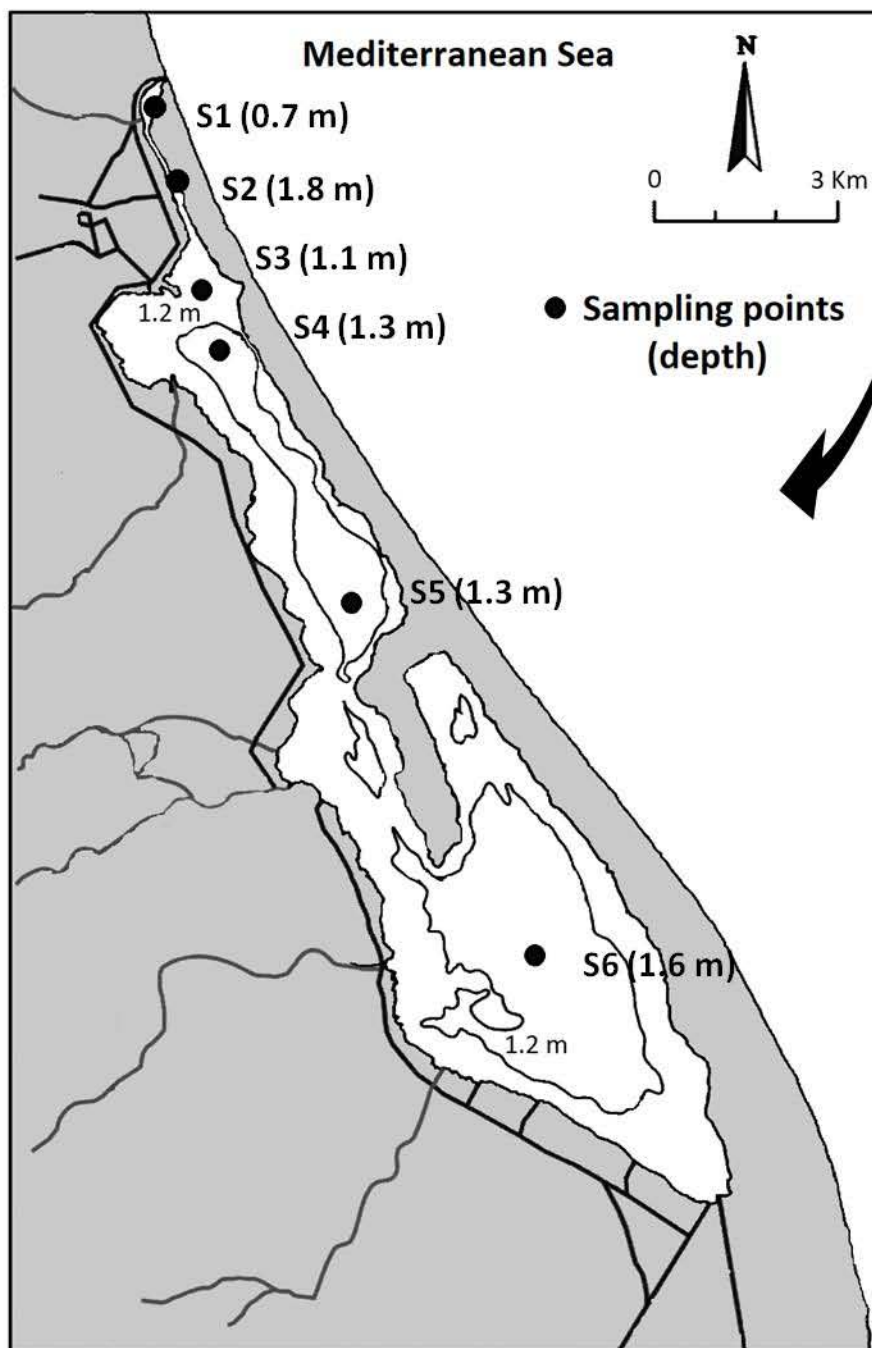
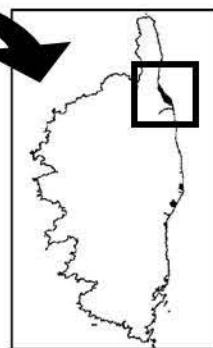
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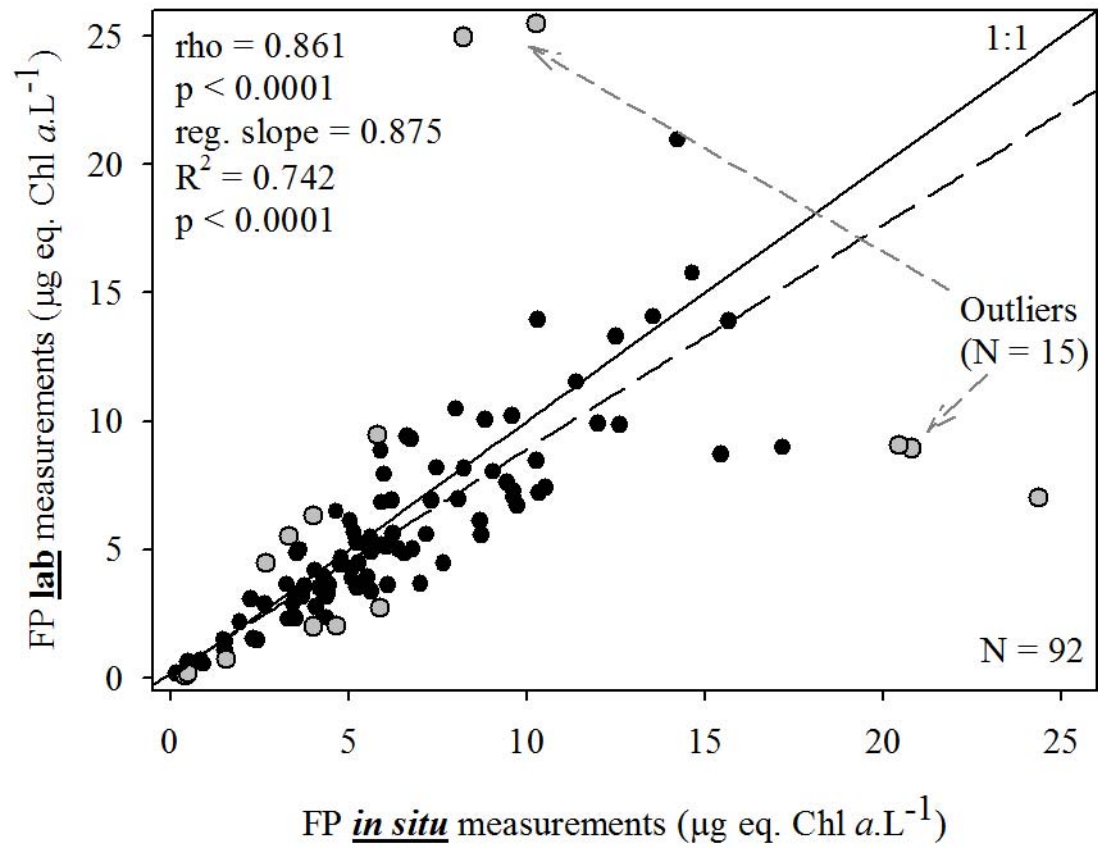
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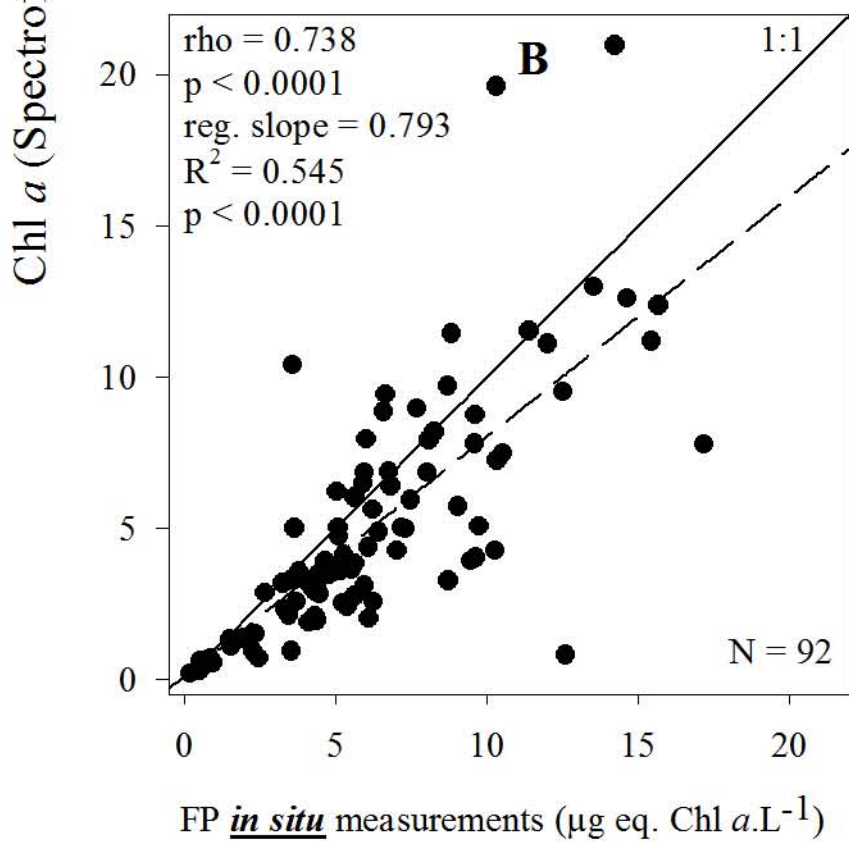
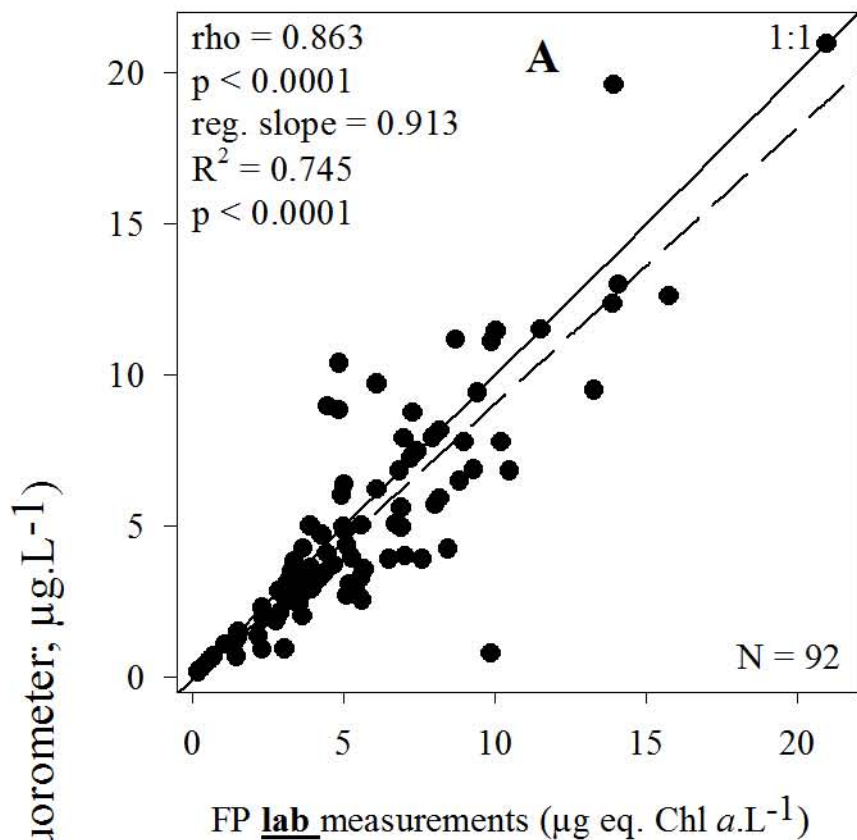
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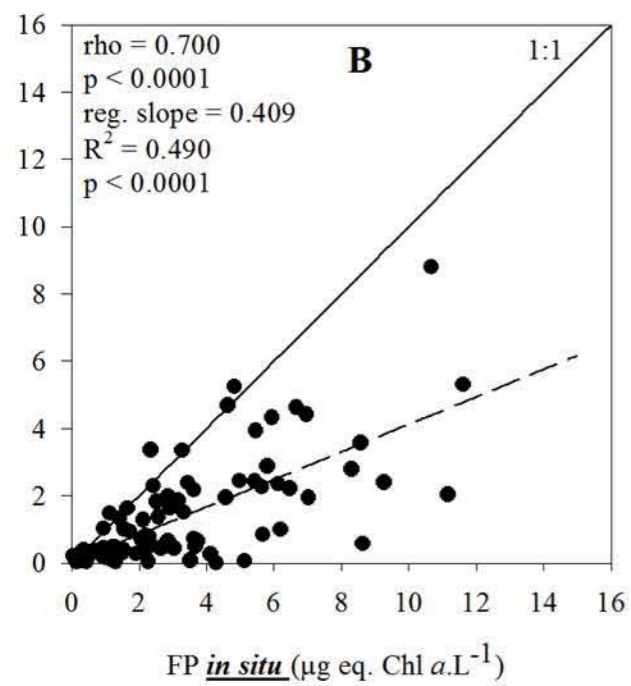
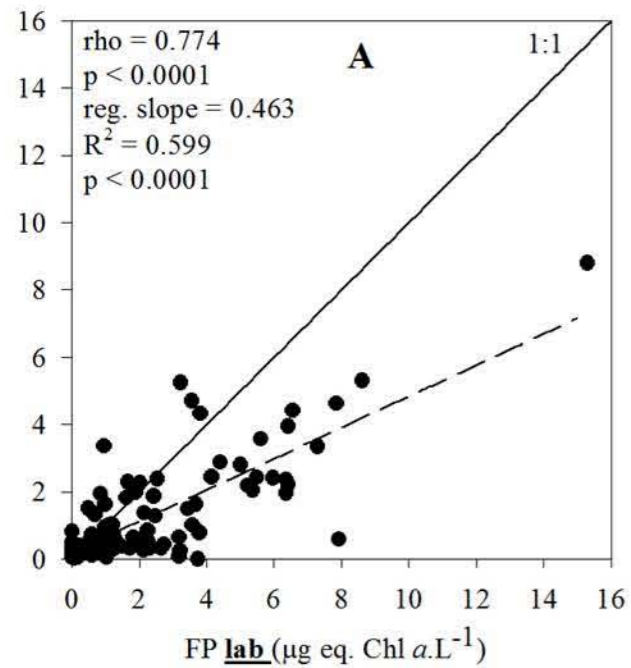
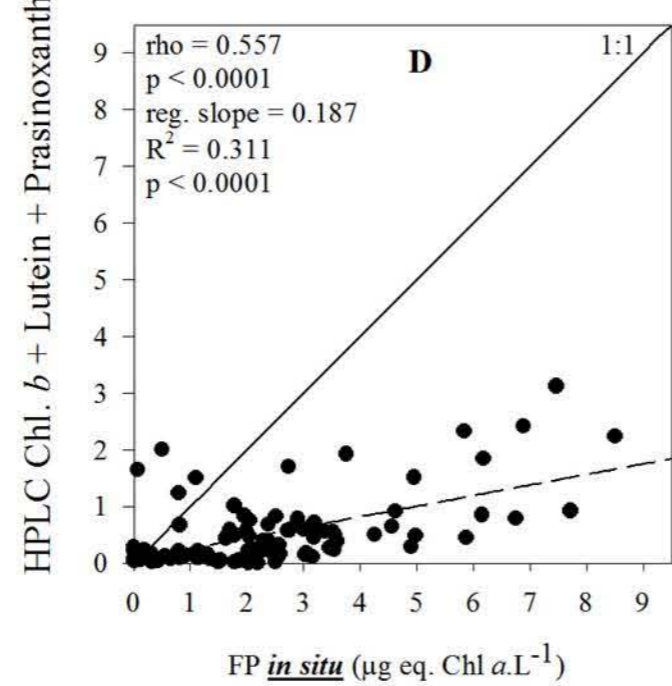
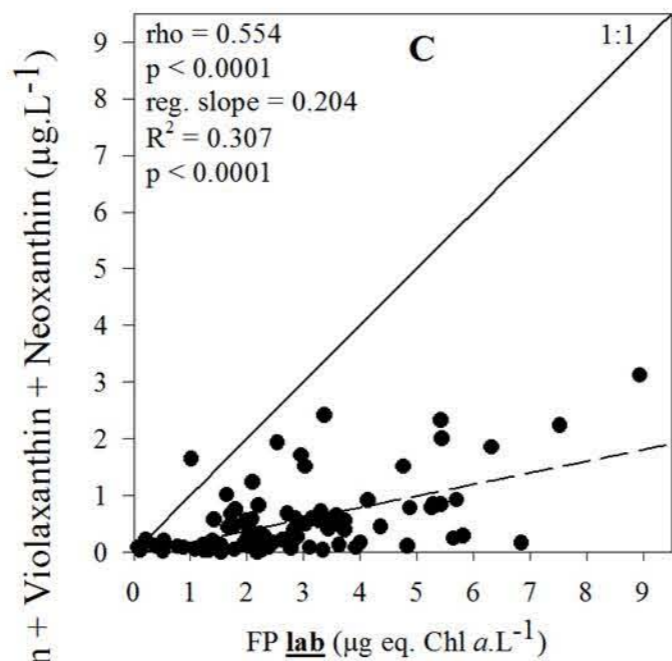
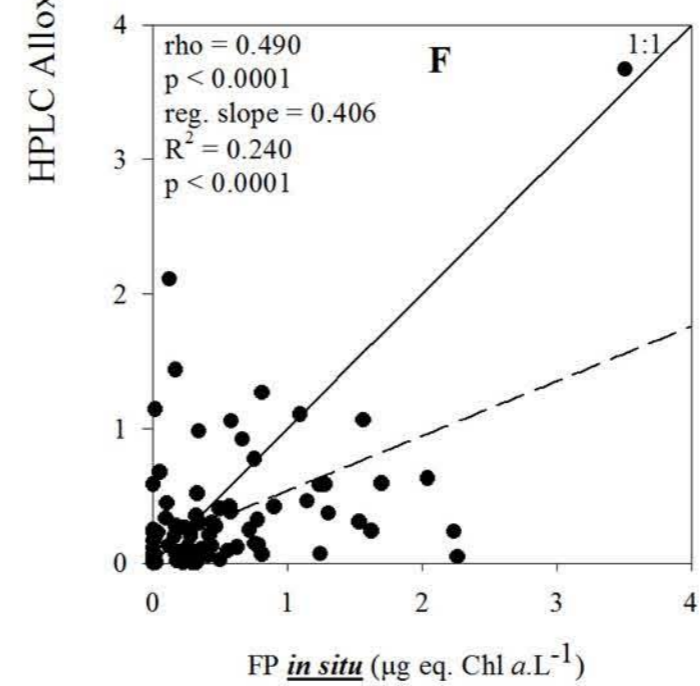
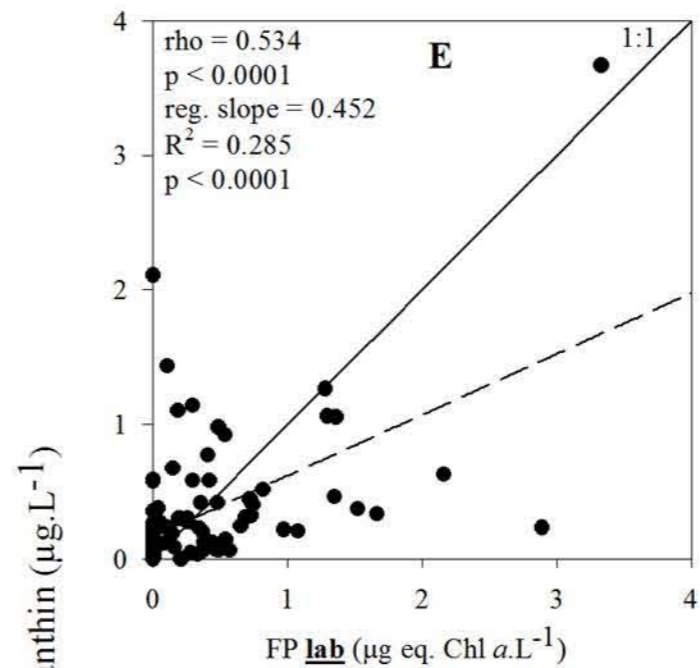
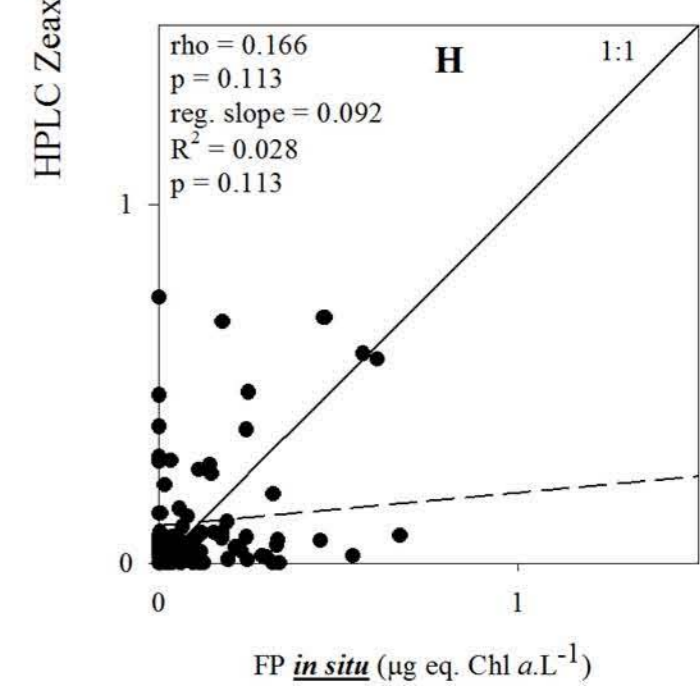
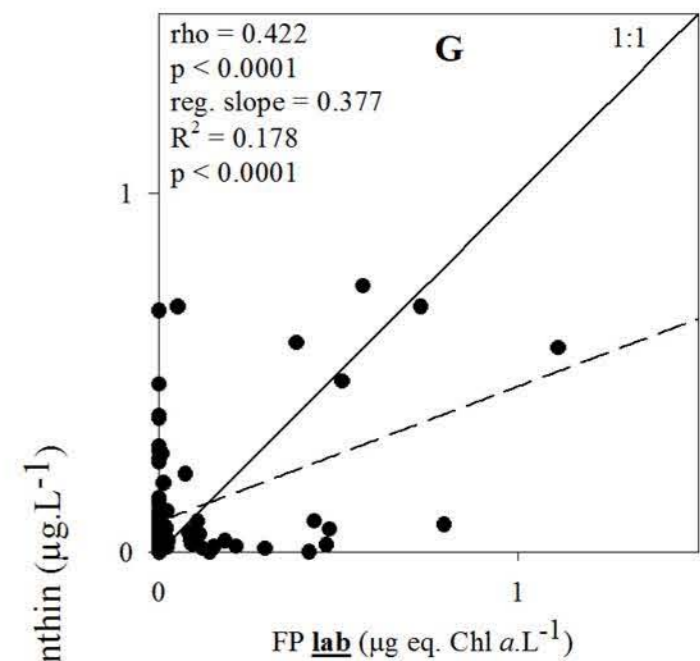
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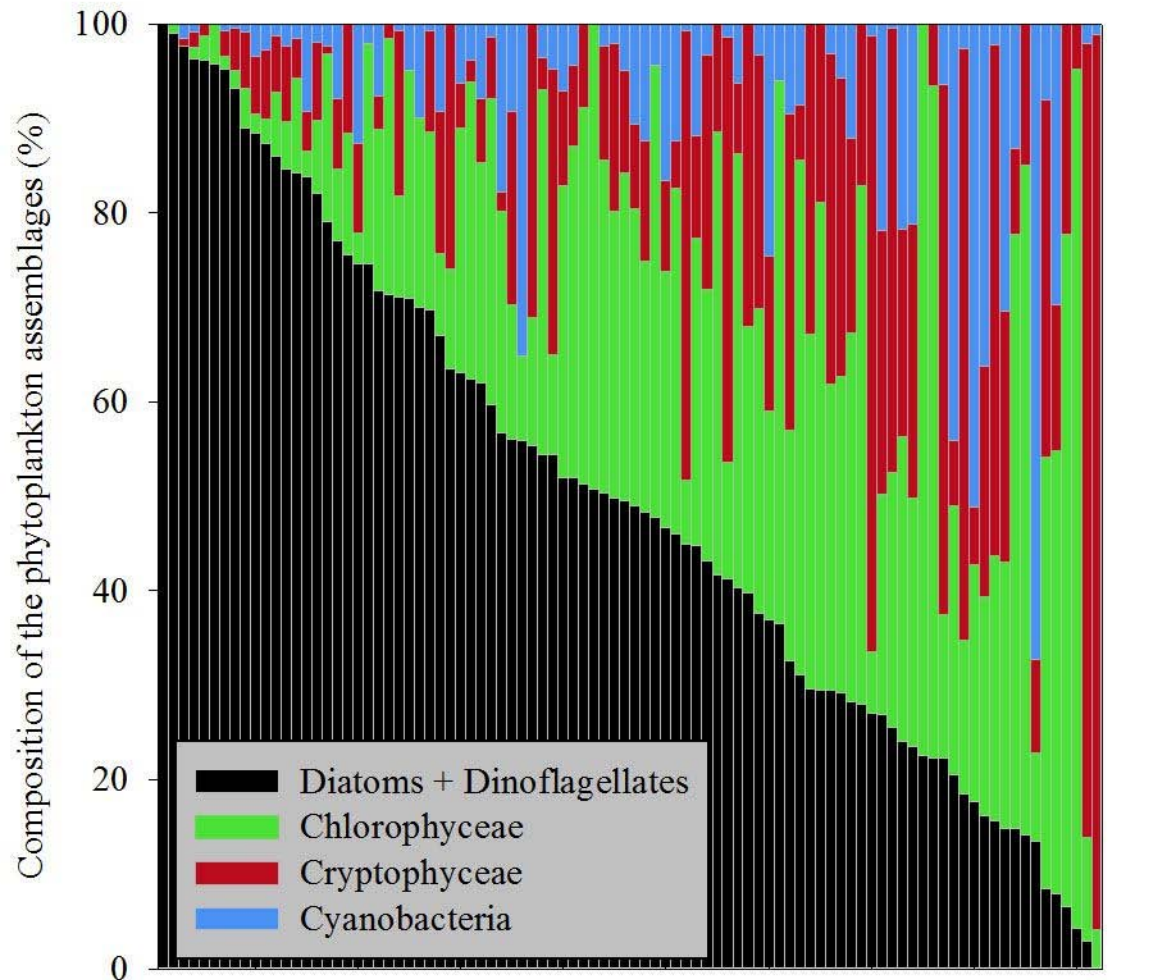




Phytoplankton biomass



Brown Algae**Green Algae****Red Algae****Blue-green Algae**



Ordination of the 92 phytoplankton assemblages considered in this study