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Temporal changes in the phytoplankton community along the French coast of the eastern English Channel and the southern Bight of the North Sea

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Abstract :

Long-term trends in the phytoplankton community along the French coast of the English Channel and southern Bight of the North Sea were studied and related to physico-chemical factors and large-climatic indices. Phytoplankton and hydrological data were acquired through three transects. Sampling took place between 1992 and 2011 as part of the French Phytoplankton Monitoring Network (REPHY) and the Regional Nutrients Monitoring Network (SRN). Trends in time-series were identified with dynamic linear models tailored to environmental monitoring data characteristics (e.g. irregular sampling frequency, missing data). Temporal and spatial patterns in the phytoplankton community were explained with a partial triadic analysis. Relationships between the phytoplankton community composition, environmental factors, and climatic indices were assessed using a redundancy analysis (RDA). The analyses revealed long-term changes in the community composition, characterized by a temporal structure that remained common to all transects. The abundance of some groups of taxa such as the one composed by Gymnodinium and Gyrodinium, as well as the group of Pseudo-nitzschia increased during the study period, whereas the abundance of other taxa as for example Guinardia and the group of Coscinodiscus and Stellarima globally decreased. More generally, the proportion of dinoflagellates relative to diatoms increased. Trends in environmental variables were also observed in most sites and related to decreases in nutrient concentrations and an increasing trend in salinity. The RDA indicated that the Atlantic Multidecadal Oscillation index and salinity were the main factors defining the temporal structure of the phytoplankton community. This suggests that variations observed in the phytoplankton community are linked to hydro-climatic changes in the coastal environment.

Keywords: climatic indicators, dynamic linear model, environmental drivers, multivariate analysis, phytoplankton

1. Introduction

In a context of climate change and concerns about anthropogenic pressures over coastal ecosystems, understanding the response of biological communities has become a necessity. In marine ecosystems, the phytoplankton is at the base of marine foodwebs, and its specific composition determines the pathways and efficiencies of the energy transfer through the upper trophic levels (<u>Cloern and Dufford, 2005; Dickman *et al.*, 2008</u>). The relative species abundance and the community composition contribute to the characterization of phytoplankton assemblage structure. This structure is governed by various environmental factors, allelopathic interactions and grazing, affecting cellular processes, and determining the population growth, community composition, and algal succession (e.g. <u>Tilman *et al.*</u>, 1982; <u>Marshall *et al.*, 2006</u>). These factors lead to recurrent patterns of phytoplankton composition observed at different temporal scales. However, these community properties may vary significantly with climate change and water-quality trends in coastal areas.

Long-term variations in phytoplankton communities have been recently observed and related to hydroclimatic changes, affecting biomass, community composition, and phenology (<u>Edwards and Richardson,</u> <u>2004</u>; <u>Suikkanen *et al.*, 2007</u>). These modifications in phytoplankton communities are often studied at different taxonomic scales. Thereby, recent works show a modification of the relative proportion between diatoms and dinoflagellates (<u>Klais *et al.*</u>, 2011; <u>Zhai *et al.*</u>, 2013</u>) and significant variations in the biomass of some taxonomic groups such as chrysophytes and chlorophytes (Suikkanen <u>et</u>

76 al., 2007). Functional approaches have also been used to assess temporal changes in 77 phytoplankton communities related to physical factors and climate (David et al., 2012). On 78 the French coast of the eastern English Channel and the southern Bight of the North Sea, 79 some studies highlighted the link between large-scale climatic features, local 80 meteorological and hydrographical conditions, modulating phytoplankton dynamics 81 (Breton et al., 2006, Goberville et al., 2010). These studies have proved the importance of 82 monitoring networks to assess long temporal changes in phytoplankton communities. 83 The Phytoplankton Monitoring Network (REPHY) is implemented and managed by 84 the French Research Institute for the Exploitation of the Sea (IFREMER). This program 85 aims to describe spatial and temporal dynamics of phytoplankton populations and to detect 86 and monitor phycotoxin-producing species involved in toxic shellfish outbreaks and 87 affecting human health (Amzil et al., 2001; David et al., 2012; Gailhard et al., 2002). 88 Previous works with these data have greatly improved our knowledge on spatial patterns of 89 microphytoplankton and highlighted the influence of hydrodynamic features in the 90 geographical structure of microalgal populations (David et al., 2012, Gailhard et al., 2002). 91 Seasonal variability and temporal succession patterns have also been characterized through 92 the analyses of phytoplankton events, defined by the phases of sudden growth, reached 93 maxima and decline of abundance (Beliaeff et al., 2001). In addition, Guarini et al. (2004) 94 showed the importance of the interaction between benthic and planktonic communities 95 through assessing the contribution of the microphytobenthos to the total phytoplankton 96 abundance. Regarding the eastern English Channel, Lefebvre et al. (2011) analyzed the 97 long-term variability of the major phytoplankton groups and gave a first insight of 98 establishing a link between environmental conditions and phytoplankton. However, even 99 though several studies have been conducted on the REPHY data sets, temporal tendencies 100 have been only partially addressed, either for the major phytoplankton components or the

specific community composition. At present, data on phytoplankton have been collected
over 20 years, and completed by a regional network which targets relations between
phytoplankton and nutrients.

Besides these ecological results, it should be noted that the analyses of these time series requires methods able to identify structures in space and time with data that exhibit a very large variability. Several methods have been used with data coming from monitoring programs, specially multivariate and time-series analyses techniques (*e.g.* David *et al.*, 2012, Goberville *et al.*, 2010, Wasmund *et al.*, 2011).

109 The main objective of this paper is to characterize and quantify long temporal 110 changes in the phytoplankton community structure at three sites in the French coast of the 111 eastern English Channel and the southern Bight of the North Sea, and to explain these 112 changes in relation to physico-chemical factors and large-scale climatic indices. We 113 examine if variations of the phytoplankton community can be observed at different scales, 114 from specific community composition, to groups such as diatoms and dinoflagellates. A 115 statistical framework is developed to deal with common issues in the analysis of long-term 116 monitoring time series and depict the spatial structure of temporal patterns. 117

118 Material and Methods

119

120 Study area

121 The study of the phytoplankton community was conducted in three coast-to-offshore

122 transects, along the French coast: Bay of Somme (SOM), Boulogne (BL), and Dunkerque

123 (DK) (Figure 1). Each transect consists of three to four sampling stations. Boulogne and

124 the Bay of Somme are located in the eastern English Channel, and are characterized by a

125 coastal flow system often separated from the open sea by a frontal area (Brylinski *et al.*,

126 1996). The sampling area of Dunkerque, located in the southern Bight of the North Sea, is

127 characterized by shallow and well-mixed waters (Lefebvre et al., 2011). In addition, the

sampling zone is influenced by a macrotidal regime and by freshwater inputs coming from

- 129 the Seine, Somme and other tributaries (such as Canche, Authie and Liane rivers).
- 130

131 Data sets

132 Data was collected as part of the REPHY and the Regional Nutrients Monitoring Network 133 (SRN). Water samples were acquired from fortnightly to a monthly frequency from 1992 134 to 2011 between 0 and 1 meter, along with physical measurements and completed with 135 chemical analyses. Finally, the environmental variables included in this study were water temperature (°C), salinity (psu), dissolved inorganic nitrogen (DIN = $NH_4^+ + NO_3^- + NO_2^-$, 136 μ mol l⁻¹), silicate (μ mol l⁻¹) and phosphates (μ mol l⁻¹). For quantitative phytoplankton 137 138 analyses, samples were fixed with Lugol's solution and counted according to the Utermöhl 139 method (Utermöhl, 1958). Organisms were identified to the lowest possible taxonomic 140 level. Taxa that are difficult to discriminate with optical microscopy were grouped. This 141 was the case of various groups representing a set of multiple species or even genus (e.g. 142 Pseudo-nitzschia "seriata complex", Gymnodinium-Gyrodinium). These groups correspond 143 to the taxonomic units extracted from the database. In addition, within sample observation 144 conditions, phytoplankton experts identified and counted organisms whose size is greater 145 than 20 µm but also species whose size is smaller but have a chain structure or form a 146 colony. Smaller species were also counted if they are potentially toxic/noxious (e.g. 147 Chrysochromulina, Phaeocystis). Further details about sampling and processing of 148 phytoplankton and physico-chemical parameters are available in the literature (Aminot and 149 Kérouel, 2004, 2007; Belin and Neaud-Masson, 2012; Grossel, 2006; Lefebvre et al., 150 2011).

151	Two large-scale climatic indices were used: the winter North Atlantic Oscillation
152	index (NAO _w , Hurrell, 1995) and the Atlantic Multidecadal Oscillation (AMO, Kerr,
153	2000). The NAO index can be defined as the difference in atmospheric pressure at sea
154	level between the high pressure centre near the Azores and the low pressure centre near
155	Iceland (Rodwell et al., 1999). This oscillation has been related with factors such as
156	temperature, precipitation, wind, storms and mixed-layer depth (Hurrell and Deser, 2009;
157	Trigo et al., 2002). The winter NAO index (December-March) was used for its more
158	pronounced correlation with temperature, wind and precipitation, and because its influence
159	persists in subsequent months (Ogi et al., 2003; Stenseth et al., 2003). This index was
160	obtained from the website http://climatedataguide.ucar.edu/. The AMO is a pronounced
161	signal of climate variability in the North Atlantic sea surface temperature (Dijkstra et al.,
162	2006). The AMO index was downloaded from
163	http://www.esrl.noaa.gov/psd/data/timeseries/AMO/.
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 166 167 168 169 170 171 172 173 174 	Preliminary processing of phytoplankton data Despite the use of taxonomic units as described previously, some confusions due to changes in phytoplankton taxonomists or any other expertise variation during the period may arise. These changes may lead to heterogeneous data regarding taxonomic classification and hence to a misinterpretation of phytoplankton time series. To homogenize the data, taxonomic units were grouped into the lowest common taxonomic level. For example, <i>Coscinodiscus, Coscinodiscus radiatus- marginatus, Coscinodiscus-</i> <i>Stellarima</i> are three different labels available in the database and were finally replaced by <i>Coscinodiscus-Stellarima</i> . In total, 91 of such taxonomic units were identified over the selected period. Furthermore, discussions with observers and phytoplankton experts were

To normalize data and reduce the effect of dominant taxa, abundances were log_{10} (*x* + 1)-transformed. In addition, a simple phytoplankton community index was defined as the proportion of dinoflagellates relative to diatoms. This proportion was calculated as the abundance of total dinoflagellates divided by the sum of total abundance of diatoms and dinoflagellates.

181

182 Data analyses

183 Univariate and mutlivariate analyses were performed to assess changes in phytoplankton

184 community, taking into account different scales of variability either taxonomic, temporal

185 or spatial. The Figure 2 shows the pathways of statistical analyses applied in this study. All

analyses and graphical representations were performed with the R software (R Core Team,

187 2012).

188

189 Univariate analyses of phytoplankton and environmental data

190 In order to characterize seasonal patterns of the main taxonomic groups, monthly median

abundances, less sensitive to extreme values, were calculated over 1992–2011.

192 Bacillariophyceae, Dinoflagellata, and Prymnesiophyceae were previously identified as the

193 most common classes in this zone (Lefebvre *et al.*, 2011).

194 Dynamic Linear Models (DLM, West and Harrison, 1997) were used to study long-

195 term trends in phytoplankton and environmental variables (Figure 2). DLM were

196 previously used in other studies for analysing phytoplankton time series (Cottingham and

197 Carpenter, 1998; Soudant et al., 1997). They are indeed time series models, *i.e.* they take

198 into account the temporal structure. Moreover, the parameters of the model can evolve

199 with time, hence the "dynamic" designation. This class of models was chosen because they

200 are well tailored to environmental monitoring data characteristics (*e.g.* irregular sampling

201 frequency, missing data, outliers). Furthermore, as for all time series models, exogenous 202 information can be taken into account using an "intervention", that is, modifying some 203 model parameters. This was particularly useful for incorporating information such as 204 changes in measurement methods applied to physico-chemical variables. All these changes 205 in measurement methods were analysed. For example, from 1992 to 2011, two main 206 methods have been used for the measurement of salinity: Knudsen titration and 207 conductivity sensors. In such cases, using the intervention allows to correct the time series 208 and thus, ensures comparability and continuity of measurements through the time series. 209 To extract trends in time series, a common model structure was used for all 210 variables. The first component is a second order polynomial trend: it remains simple and 211 allows to capture complex evolutions. The second component is a 12-month seasonal 212 factor. This was chosen since sampling frequency was mainly monthly. Furthermore, this 213 temporal unit is fine enough to capture different annual periodic patterns. Finally, when 214 more than one observation was available within the month, the median was used to 215 aggregate the data. Outliers were detected by examination of standardized residuals. Under 216 the assumption of a normal distribution of standardized residuals, it is expected that about 217 95% of them are within the \pm 1.96 interval. Values standing outside this interval were 218 considered as outliers. They were not removed but treated using interventions. Model 219 checking was assessed by residual diagnosis. Serial autocorrelation and normality were 220 tested by plotting the autocorrelation function of the standardized residuals as well as their 221 normal OO-plot. Deviations from normality were rare. In the case where serial 222 autocorrelation was present, an AR(1) term was added to the model. The results showed 223 that adding this term does not affect trends. Thus, models have been kept in the original 224 structure as described above.

DLM were first used to study long-term trends in diatoms and dinoflagellates functional groups, as well as trends of some taxonomic units, selected through multivariate techniques (explained below). Physico-chemical trends were identified with DLM, and the annual mean trend was used for multivariate analyses. Time series analyses with DLM were conducted with the *dlm* package for R (Petris, 2010).

230

231 Multivariate analyses of phytoplankton community composition

232 For multivariate analyses, the genus was used as the smallest taxonomic level. Aggregating 233 the data at this level allows to capture most of the information considering that little 234 information is lost using the genus level rather than the species one (Cottingham and 235 Carpenter, 1998). In order to focus on the main changes in phytoplankton community, rare 236 taxa were excluded from the analyses, based on the occurrence frequency over 1992–2011. 237 A taxa was discarded when its occurrence frequency was less or equal to three years in all 238 sites. This resulted in a loss of less than one percent of the total abundance for each site, 239 which represents 29 discarded taxonomic units. Sixty-two taxonomic units remained for 240 mutivariate analyses.

241 Spatial and temporal characteristics of the phytoplankton data led to obtain a three-242 dimensional table (taxonomic units; sites; years). Partial Triadic Analysis (PTA), used with 243 such multi-table data, was computed to elucidate temporal patterns of the phytoplankton 244 community composition (Figure 2, package ade4 for R, Dray et al., 2007). This method 245 allows simultaneous principal component analyses of a set of tables by searching for an 246 average temporal structure called compromise, and studying the spatial stability of this 247 structure (Blanc and Beaudou, 1998). In addition to the analysis performed at genus level, 248 the PTA was also carried out at upper levels (*e.g.* family, order, class).

249

250 Relationships between phytoplankton, environmental factors and climatic indices

251 A Redundancy Analysis (RDA) was conducted with data from all sampling stations in

252 order to study the relationships between patterns in the phytoplankton community

253 composition, environmental variables and climatic indices (Figure 2, package vegan for R,

254 Oksanen et al., 2012). The significance of the analysis and all variables was assessed using

255 permutation tests (499 permutations).

256

257 Results

258

From 1992 to 2011, Bacillariophyceae, Dinoflagellata, and Prymnesiophyceae were the

260 most abundant groups, averaging 81% of the total abundance. Among these groups,

261 diatoms and dinoflagellates represented over 84% of the total diversity. Principal diatom

taxa in terms of abundance were the genus Guinardia, Chaetoceros, Rhizosolenia, Pseudo-

263 nitzschia, Paralia, Skeletonema and Leptocylindrus. The group Gymnodinium-Gyrodinium

264 was the major dinoflagellate present over the period. *Phaeocystis* (Prymnesiophyceae) was

a major contributor to the phytoplankton total abundance between March and May. When

266 present, *Phaeocystis* median contribution varies from 74% to 90%, with the highest

267 concentrations generally observed in stations near the coast. The complete list of taxa

268 identified is given in the Supplementary Material (Table S1) with information on the

taxonomic groups they belong to, and highlighting those mainly discussed in this study.

270

271 Seasonal patterns in phytoplankton and environmental data

272 Although all sites within a same transect generally present the same patterns, three

273 sampling sites, each one belonging to a different transect were chosen as their results

274 present the most pronounced variations. Consequently, these sampling sites will be used

275 for graphical representations. Among the major phytoplankton groups, *Phaeocystis*

276 (Prymnesiophyceae) showed a clear seasonal pattern, with peaks that appear generally in

277 April (Figure 3). Diatoms were present throughout all the year with highest abundance

usually observed between June and July in Boulogne and the Bay of Somme (Figure 3). In

279 contrast, in Dunkerque, diatoms showed high abundances early in the year (March).

280 Seasonal patterns of diatoms, dinoflagellates and Prymnesiophyceae are showed for each

sampling station in Figure S1 in the Supplementary Material (SM).

282 Spring-summer diatoms generally included the genus *Guinardia*, *Chaetoceros*,

283 Pseudo-nitzschia and Rhizosolenia while Leptocylindrus was characteristic in summer. The

autumn-winter period was characterized by the presence of the diatom *Skeletonema*, the

285 groups Thalassionema-Thalassiothrix-Lioloma, Thalassiosira-Porosira, Coscinodiscus-

286 Stellarima and the genus Ditylum. Others groups present at this period of the year included

the small single-cell flagellate *Dictyocha*. Dinoflagellate abundances increased from the

beginning of the year peaking between July and September, then decreasing at the end of

the year (Figure 3). The dinoflagellates Gymnodinium-Gyrodinium, Protoperidinium-

290 *Peridinium*, and *Prorocentrum* are common genera during this period.

291 Seasonal patterns of environmental variables are illustrated for each sampling 292 station in Figure S2 (SM). Nutrients data reflected typical temperate seasonal patterns, 293 remaining almost similar between sites. DIN, phosphate and silicates reached maxima 294 during the winter and decreased over the spring and early summer to levels close to the 295 detection limits (Figure 4). A further increase in nutrients was observed in the late 296 summer-autumn period. During winter, the Bay of Somme showed the highest concentration of DIN and silicates, with average values between 22.7 μ mol 1⁻¹ and 44.3 297 μ mol l⁻¹ for DIN and between 9.1 μ mol l⁻¹ and 20.7 μ mol l⁻¹ for silicates. Lower 298 299 concentrations of DIN and silicates were found at Boulogne with ranges of 11–23.8 umol 1⁻

300	1 and 5.1–8.7 µmol l ⁻¹ , respectively. At Dunkerque the concentration of these nutrients
301	varied from 14.5 to 31.5 μ mol l ⁻¹ for DIN and from 7.7 to 12.9 μ mol l ⁻¹ for silicates.
302	The phosphates concentration varied from 0.73 to 1.1 μ mol l ⁻¹ at Boulogne and
303	reached comparable values at Dunkerque and the Bay of Somme (0.83–1.27 $\mu mol~l^{-1}$ and
304	0.82–1.76 μ mol l ⁻¹ respectively). Furthermore, a spatial heterogeneity of the nutrient
305	concentrations was observed in "coast-to-offshore" transects, with higher concentrations
306	near the coast. The temperature varied from an average minimum of 5.9°C between
307	January and February to a maximum of 18.9°C between July and August. The salinity
308	varied between 31.6 and 34.4 psu. In addition, the sampling station of Bif was
309	characterized by a lower salinity (minima 9.1 psu), related to a major influence of
310	freshwater inputs from the Somme river.
311	
312	Long-term trend in diatoms and dinoflagellates groups and environmental data based on
313	univariate analysis
314	The long-term trends in the diatoms and dinoflagellates abundance are showed in Figure 5.
315	Minor changes in the abundance of diatoms are observed at Dunkerque and in the Bay of
316	Somme while a decrease in the abundance is observed in Boulogne between 1992 and
317	1997, followed by an increase between 1998 and 2005. The abundance of dinoflagellates
318	between 1992 and 2011 showed more pronounced variations. From 2000 to 2007, the
319	mean level of the dinoflagellates logarithmic abundance increased by 60% (from 2.5 to 4
320	log ₁₀ cells l ⁻¹). Spatial differences in the trend of dinoflagellates are observed among
321	transects between 1992 and 2000, but abundance patterns showed minor variations within
322	sampling stations of the same transect. The abundance of dinoflagellates decreased at
323	Dunkerque and Boulogne between 1992 and 2000, while minor changes were observed at
324	the Bay of Somme. The phytoplankton community showed an increasing trend in the

proportion of dinoflagellates relative to diatoms, with values generally below 0.1 between
1992 and 2001 and frequently above this value after 2002 (Figure 5). In addition, increases
in dinoflagellates abundance were observed throughout all the seasonal cycle after 2001.
Regarding physico-chemical variables, dissolved inorganic nitrogen decreased in
Dunkerque from the beginning of the period (Figure 6 a) and after 2000 at Boulogne. The
phosphate concentration decreased at Dunkerque and the Bay of Somme at the start of
period with a minor increase after 2000s (Figure 6 b). A decrease was observed at

Boulogne over the period. No trend was observed for silicates at Dunkerque while it seems

to slightly increase at Boulogne after 2000s (Figure 6 c). At the Bay of Somme, DIN and

silicates seem to vary cyclically, with periods of higher concentrations (1994–1995, 2000–

2001) followed by years with lower concentrations (1996–1998, after 2002). Time series of
salinity at Boulogne and the Bay of Somme showed large increases during the period

337 (Figure 6 d).

338

339 Long-term changes in the phytoplankton community based on multivariate analysis 340 As suggested by the PTA results, the community structure changed between 1992 and 341 2011 (Figure 7). The first two axes account for 58.7% of the total variability. PTA objects, 342 corresponding to (site; year) pairs, were projected on the first and second axis of the 343 compromise (Figure 7 a). Each dotted line represents a given sampling station. The first 344 axis globally distinguishes two periods: 1992-2001 and 2002-2011. In the latter period, 345 the second axis splits the period into two groups: 2002–2007 and 2008–2011. This 346 temporal trajectory was generally seen at all the sampling stations. Such result highlights a 347 temporal structure of the phytoplankton community composition and a trend in taxa 348 abundances that change largely during the sampling period while remaining almost 349 constant in space. Among diatoms, *Melosira*, and *Stephanopyxis* were usually common in

- as a large start of the second second
- 351 Amphidinium, Alexandrium, and Polykrikos were characteristic after 2002 (Figure 7 b).
- 352 Finally, Heterocapsa (dinoflagellate), Torodinium (dinoflagellate), and Eutreptiella
- 353 (euglenid) were mostly common over the third period.
- 354 Changes in the abundance of some taxonomic units were observed over the period
- 355 (Figure 8 and Figure S3 in SM). The diatom *Guinardia* generally decreased at the Bay of
- 356 Somme and Dunkerque while no pronounced variation was detected at Boulogne. Other
- 357 taxa, mainly diatoms, showed a decline in their abundances (e.g. Coscinodiscus-Stellarima,
- 358 Eucampia-Climacodium, Ceratulina) while the abundance of Pseudo-nitzschia increased
- during the period of study. Among dinoflagellates, the group of Gymnodinium-Gyrodinium
- 360 showed a strong increase in abundance, which has more than doubled (in logarithmic
- 361 scale) between 2002 and 2007. *Phaeocystis* showed no trend in the Bay of Somme, while a
- 362 minor increase was observed at Dunkerque and Boulogne.
- 363 The analyses of taxonomic groups from family to class indicate a similar temporal
- 364 structure. Nevertheless, with the increase in the level of aggregation, presence/absence
- 365 patterns of specific taxonomic units are more difficult to detect while changes in
- abundances tend to become more apparent.
- 367

368 Relationships between phytoplankton community, environmental factors and climatic

- 369 indices
- 370 The redundancy analysis accounted for 21.8% of the total variance of the data (all
- 371 canonical axes). Of this, 68.8% was explained by the first two axes. These canonical axes
- and the seven variables were significant (p < 0.01). On the first axis, the AMO index,
- 373 salinity, NAO_w index, and temperature were the main explanatory variables (Figure 9 a).
- 374 The first axis describes a temporal structure, with two different periods, before and after
 - 15

2001 (Figure 9 a, c). The second axis is mainly defined by silicates, dissolved inorganic
nitrogen, but also salinity (Figure 9 a). It shows a spatial pattern between transects.
Sampling stations in the Bay of Somme are characterized by high concentrations of
silicates and dissolved inorganic nitrogen, and a low salinity (Figure 9 a). Temporal
patterns were observed in transects of Dunkerque and Boulogne in relation to a decreasing
trend in nutrients concentrations (mainly dissolved inorganic nitrogen) over 1992–2011
(Axis 2–RDA, Figure 9 a).

382 The distribution of phytoplankton taxonomic units along the first gradient was 383 positively related to the salinity, AMO index and temperature, and negatively to the NAO_w 384 index and phosphates concentration. The AMO index, salinity and temperature vectors 385 pointed to the second period (after 2001) when dinoflagellates increased (Figure 9 a, b). 386 Diatoms were mostly grouped in the lower-left quadrant and thus positively linked to 387 silicates, DIN and NAOwindex. However, some diatoms such as Pseudo-nitzschia showed 388 their highest abundance during the second period (positive score on the first axis), 389 suggesting that these diatoms were favoured under an increased salinity and positively 390 associated with the AMO index and temperature. Other genera show low scores in both 391 axes, thus suggest its presence along all transects and over mostly the whole period of 392 study (e.g. Rhizosolenia, Skeletonema).

393

394 Discussion

395

396 Patterns of phytoplankton community changes

397 During the past 20 years, the structure of the phytoplankton community of the eastern

398 English Channel and the southern Bight of the North Sea was subject to changes. We

399 observed an increasing trend of dinoflagellates during the last decade while minor changes

400 were observed for diatoms between 1992 and 2011. The PTA allowed us to understand 401 these changes at a finer taxonomic scale. Some diatoms taxa globally showed a declining 402 trend in their abundances (e.g. Guinardia, Coscinodiscus-Stellarima). In contrast, the 403 abundance of *Pseudo-nitzschia* appears to increase over the 20 years of study. Diatoms like 404 Melosira and Stephanopyxis were mainly found before 2001. Some dinoflagellates such as 405 *Gymnodinium-Gyrodinium* showed a strong increasing trend in their abundances after 406 2001. Other genera such as Alexandrium, Amphidinium and Polykrikos were mostly found 407 in the last decade (2002-2011).

408 In our study, the PTA enabled to describe temporal patterns and also facilitated 409 comparisons between multiple sampling stations, determining if the long-term trend is 410 spatially constant. This analysis was previously used to study the temporal and spatial 411 structures of phytoplankton communities (David et al., 2012). Globally, three periods were 412 identified (1992–2001, 2002–2007 and 2008–2011), and such temporal pattern is globally 413 exhibited by all transects. In combination with RDA to further assess relationships between 414 phytoplankton communities, physical-chemical variables and climatic indices, we found a 415 clear separation between the periods before and after 2001, and some of the observed 416 trends in the phytoplankton community could be related with physico-chemical variations 417 and climatic indices. The time at which the changes in the phytoplankton community 418 structure occurred is consistent with a shift observed in the French coastal system after 419 2001, manifested through an increasing salinity and decreasing nutrients, impacting 420 chlorophyll a concentrations (Goberville et al., 2010). Similar temporal patterns were 421 observed in the functional and specific diversity along the French Atlantic coast, although 422 spatial patterns were most apparent, mainly related to a latitudinal gradient and 423 hydrodynamic characteristics (David et al., 2012). Moreover, during the last decades, 424 major modifications in the phytoplankton composition, leading to an increasing importance

of some groups such as dinoflagellates, have been documented in the central North
Atlantic Ocean, North Sea and Baltic Sea (Klais *et al.*, 2011; Leterme *et al.*, 2005; Zhai *et al.*, 2013). The rather low percentage of the variance explained by the RDA in this study
has also been observed in previous analyses of phytoplankton data (Suikkanen *et al.*, 2007,
Wasmund *et al.*, 2011). This has been related to environmental factors usually not
considered in the analyses, but having influence in phytoplankton populations, such as
allelopathic interactions, interspecific competition and grazing.

432 It should be noted, however, that changes in staff involved in the identification of 433 microalgal species, or any other expertise variation, may lead to a misinterpretation of 434 phytoplankton time series. These issues have been already questioned, especially with long 435 term phytoplankton data sets (Wasmund and Uhlig, 2003; Wiltshire and Dürselen, 2004). 436 Obviously, with more than 20 years of REPHY monitoring network along the French 437 coast, such changes did occur. To deal with these biases, other studies have suggested a 438 regrouping of taxonomic units (Gailhard et al., 2002, Guarini et al., 2004). In our study, 439 clustering taxonomic units has been carried out with morphologically close taxa. 440 Furthermore, discussions with observers directly involved in the identification and 441 counting, as well as phytoplankton experts, suggest that our trends are reliable. In addition, 442 the results of PTA performed with different taxonomic grouping, from family to class, 443 show a similar temporal structure, which supports our interpretation. 444 Previous studies have highlighted temporal changes in the phytoplankton 445 community structure at different scales. In the English Channel, Widdicombe et al. (2010) 446 identified major changes in the community such as the decrease of the average abundance 447 of diatoms and the increase of some dinoflagellates abundance. For the study area, changes 448 have also been observed in chlorophyll *a* concentrations, used as an indicator of 449 phytoplankton biomass trends (Goberville et al., 2010; Lefebvre et al., 2011). Other

450 approaches can also be suitable with available species data sets. Modifications in the

451 ecological niche of species have been detected using long-term species data (Gebühr *et al.*,

452 2009; Grüner *et al.*, 2011). Moreover, a functional approach can be particularly relevant to

453 understand prey-predator processes and assess factors driving the structure of

454 phytoplankton communities (David *et al.*, 2012).

455

456 Links between phytoplankton changes and environmental variables

457 Several dinoflagellates genera have shown an increasing trend over the period and such

458 trend seems to be linked to AMO index, salinity and water temperature. In our study, the

459 AMO index was one of the principal factors defining the temporal structure observed in the

460 phytoplankton community, suggesting a possible link between large-scale climate

461 variability and phytoplankton. Similar temporal patterns were globally observed in all

462 transects, suggesting that the climate forcing may be stronger than local prevailing

463 conditions. The link between phytoplankton community structure and climate forcing has

464 been previously discussed in literature (Breton *et al.*, 2006; Irigoien *et al.*, 2000; Leterme

465 *et al.*, 2005). The AMO index reflects changes in the sea surface temperature, a main factor

466 governing the changes in the coastal environment (Beaugrand, 2009) and may favour

467 dinoflagellates growth. Although the AMO seems to be more related with the

468 phytoplankton community structure, the NAO index was also positively associated with

some diatoms and negatively related to dinoflagellates, a relation previously observed in

470 the North Atlantic Ocean and the English Channel (e.g. Irigoien et al., 2000; Zhai et al.,

471 2013).

472 Regional climate conditions modify coastal environment, influencing physico473 chemical parameters (Goberville *et al.*, 2010). Along the French coast of the English
474 Channel, observed changes in environmental conditions, such as salinity increases and

475 decreasing nutrient concentrations were related with the sea surface temperature increase, 476 reduced rainfall and intensity of zonal winds (Goberville et al., 2010). Oceanic-477 atmospheric drivers such as AMO and NAO are likely to impact precipitation patterns and 478 may affect the river run-off as well as inputs on terrestrial-derived nutrients, salinity and 479 turbidity (e.g. Enfield et al., 2001; Harley et al., 2006, Trigo et al., 2004). In a study of the 480 three main Iberian river basins, Trigo et al., (2004) have shown an influence of the NAO 481 on precipitation patterns and river flow, but also highlighted the irregularity and the high 482 inter-decadal variability of this impact. The river flow patterns may also reflect human 483 impacts such as damming and channels. Concerning the Somme river, we found a weak 484 correlation between the NAO and the river flow (Pearson correlation coefficient r = 0.15, p 485 < 0.02) although some influence can be observed. These results may suggest more local 486 influencing conditions over the river flow. When analysing the influence of the Somme 487 river flow on the nutrient concentrations and salinity trends in the Bay of Somme, we 488 generally observed a significant (p < 0.05) and a high correlation coefficient (highest values $r_{DIN} = 0.89$, $r_{SIOH} = 0.77$, $r_{SALI} = -0.52$, r_{PO4} not significant). Both, NAO and river flow 489 490 seem to modulate separately the Bay of Somme hydrologic conditions. In addition, the 491 frequency of important flow events in the Somme river seems to decrease after 2004. In 2001, the maxima river flow was about 96 m³ s⁻¹, while from 2004 to 2011, the maxima 492 river flows ranges between 30 and 49 m³ s⁻¹ (Data: Artois–Picardie Water Agency). 493 494 Generally, our results suggest an increasing trend in salinity and a decreasing trend 495 in dissolved inorganic nitrogen and phosphates, while a certain variability in trends 496 between sampling sites has to be noted. At Dunkerque, DIN concentrations decreased 497 throughout the study period and during the 2000's in Boulogne. In the Bay of Somme, a 498 period of low concentrations was observed from 2002. Our analysis suggests a decreasing 499 trend in phosphates in Boulogne over the whole period while at Dunkerque and the Bay of

500 Somme it decreases at the start of 1990's with minor increases in 2000's. The decreasing 501 trend in phosphates has been already highlighted in SRN data series and attributed to a 502 result of phosphates-reduction practice since 1970s (Lefebvre et al., 2011). Similar results 503 were observed by Romero et al. (2013) while assessing nutrients export for a large number 504 of rivers in south-western Europe (including France). The authors highlighted a general 505 decrease in phosphorus, a general decrease in phytoplankton biomass during summer, as 506 well as an increase in the summer abundance of dinoflagellates over the past 20 years. It is 507 well-known that the nutrient availability is a major factor for algal growth (Tilman et al., 508 1982). In general, diatoms dominate under high-nutrient and turbulent conditions whereas 509 dinoflagellates prevail under periods of stratified conditions and low nutrient 510 concentrations (Margalef, 1978). Under low-nutrient conditions, the use of mixotrophy in 511 algal flagellates may represent an advantage, making the growth not strictly dependent on 512 the inorganic nutrients supply (Cloern and Dufford, 2005; Smayda, 2002). Considering the 513 heterogeneity of phytoplankton taxonomic classification levels in the REPHY database, we 514 were not able to associate species, mainly dinoflagellates, to mixotrophic and heterotrophic 515 functional groups. However, potential groups where mixotrophic species are present, like 516 Gymnodinium-Gyrodinium, showed an increasing trend in abundance.

517 Phytoplankton changes are usually not disconnected from other ecosystem changes 518 and interactions between ecosystem components may amplify or attenuate changes at the 519 ecosystem level. Multiple studies have recently highlighted the response of marine 520 ecosystems to climate change (e.g. Beaugrand, 2009; Harley et al., 2006; Hays et al., 521 2005). These changes can be observed at several levels of the ecosystem such as primary 522 producers, zooplankton and fish communities (Beaugrand, 2004; Beaugrand, 2009; 523 Hawkins *et al.*, 2003). Thus, they may alter community properties like diversity, 524 dominance, community composition and lead to modifications in the marine ecosystems

525	functioning (Beaugrand, 2004; Wernberg et al., 2011). However, the lack of long-term
526	data on other biological components makes difficult to separate ecosystem responses to
527	changes. There is a growing interest in analysing phytoplankton data sets in coastal areas,
528	establishing trends for phytoplankton as well as for associated variables such as inorganic
529	nutrients. These constitute the activities being addressed by the ICES Working Group on
530	Phytoplankton and Microbial Ecology (WGPME, Li et al., 2011).

532 Conclusion and future works

533

534 The analysis of the phytoplankton time series allowed the identification of important 535 temporal changes in the phytoplankton community structure between 1992 and 2011 along 536 the French coast of the eastern English Channel and the southern Bight of the North Sea. 537 We found some evidence of changes in the community composition as well as significant 538 long-term trends in the abundance of some taxonomic units that seem to reflect climate-539 driven changes in the coastal environment. Since phytoplankton plays a key role in the 540 pathways and efficiency of energy transfer through the upper trophic levels and 541 participates in the nutrients cycles, the observed long-term changes in the phytoplankton 542 community structure may have important ecological consequences (e.g. Edwards and 543 Richardson, 2004; Hays et al., 2005). However, further studies are necessary to assess the 544 modifications of the phytoplankton community and the factors forcing these changes. In 545 particular, the causal relationships between climate and environmental drivers on one hand, 546 and phytoplankton communities and species phenology on the other hand, need to be 547 described to assess whether changes correspond to a major shift in the ecosystem structure and functioning. Future works will therefore follow two lines of research: 1) apply the 548 549 same statistical framework to other REPHY monitoring sites, test for similar occurrences

550	of changes and relate changes to environmental drivers, 2) determine the ecological niche
551	of the most significant species and assess whether and how the phytoplankton phenology
552	responds to environmental changes.
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554	Supplementary material
555	
556	The following supplementary material is available at ICESJMS online version of the
557	manuscript:
558	Table S1 A list with assigned codes and taxonomic groups of phytoplankton taxa
559	identified.
560	Figure S1 Seasonal patterns of the three main phytoplankton groups at ten stations along
561	the French Coast of the English Channel and the southern Bight of the North Sea between
562	1992 and 2011.
563	Figure S2 Seasonal patterns of environmental variables at ten stations along the French
564	Coast of the English Channel and the southern Bight of the North Sea between 1992 and
565	2011.
566	Figure S3 Long-term trends in the abundance of six taxonomic units over the sampling
567	period at Dunkerque, Boulogne and the Bay of Somme.
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Figure 1. Location of the phytoplankton sampling stations in the Bay of Somme and

772 Boulogne transects (English Channel) and Dunkerque transect (southern Bight of the North

- 773 Sea).

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Figure 2. Pathways of statistical analyses followed in this study.





802 Figure 3. Median monthly abundance of the main phytoplankton taxonomic groups in the



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Figure 4. Box-and-whisker plot of the annual variation of physico-chemical measurements
between 1992 and 2011 at Dunkerque, Boulogne and the Bay of Somme. Horizontal line
denotes the median value for each month, box represents first and third quartiles and

824 whisker represents last value within 1.5 times the interquartile distance.





Figure 5. Long-term trends of diatoms (first column) and dinoflagellates (second column)
abundances, and proportion of dinoflagellates relative to diatoms (third column) between
1992 and 2011 in three sampling sites. Solid black lines and dotted lines represent trends
and 90% confidence interval respectively, as predicted by the Dynamic Linear Model.
Gray and dark dots correspond to deseasonalized observations and outliers respectively.

836 LD (detection limit).



Figure 6. Long-term trends of (a) DIN, (b) phosphates, (c) silicates concentration and (d)
salinity between 1992 and 2011 at Dunkerque (Site DK4), Boulogne (Site BL2) and Bay of
Somme (SOM1). Solid black lines and dotted lines represent trends and 90% confidence
interval respectively, as predicted by the Dynamic Linear Model. Gray and dark dots
correspond to deseasonalized observations and outliers respectively. *LD (detection limit)*.



Figure 7. Partial triadic analysis results conducted on the phytoplankton community data from ten sampling stations with the periods 1992–2001, 2002–2007 and 2008–2011. (a) First and second axis for (site; year) objects. Each dotted line represents a given sampling station. (b) First and second axis for taxonomic units. For readability, only taxonomic units with large coordinates in the first and second axis were represented, the other taxa were represented with black crosses. The plot displays 43.3% and 15.4% of the variance in the first and second axis respectively. For details in codes and taxonomic groups of phytoplankton units, see Table S1 in SM.



Figure 8. Long term trend in the abundance of four selected taxonomic units over the sampling period at the Bay of Somme, the transect showing the most pronounced variations. Solid black lines and dotted lines represent trends and 90% confidence interval respectively, as predicted by the Dynamic Linear Model. Gray and dark dots correspond to deseasonalized observations and outliers respectively. LD (detection limit). CoscStel Coscinodiscus-Stellarima, Guinard Guinardia, GymnGyro Gymnodinium-Gyrodinium, Pseunitz Pseudo-nitzschia.



887 Figure 9. Redundancy analysis of phytoplankton units between 1992 and 2011. (a) RDA 888 biplot (site; year) score (filled symbols), and environmental variables (box). The bottom-889 left scale corresponds to site-year objects and top-right scales to environmental variables, 890 respectively. Colours on filled symbols represent periods (light gray symbols for years 891 before 2001, dark gray symbols after 2001. (b) Details for phytoplankton genera (black 892 labels for dinoflagellates, gray labels for other taxa), genera which percentage of variance 893 explained in RDA is lower than 20% are indicated by black crosses. (c) Sites-temporal 894 trajectories represented for the first axis. Abbreviations of environmental variables: DIN

- 895 (dissolved inorganic nitrogen), SIOH (silicate), PO4 (phosphates), TEMP (temperature),
- 896 SALI (salinity), NAO_w (winter North Atlantic Oscillation index), AMO (Atlantic
- 897 Multidecadal Oscillation index). For details in codes and taxonomic groups of
- 898 phytoplankton units, see Table S1 in SM.
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- 900