

# Distributions of phytoplankton carbohydrate, protein and lipid in the world oceans from satellite ocean colour

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#### 13

#### Abstract

Energy value of phytoplankton regulates the growth of higher trophic species, affect-14 ing the tropic balance and sustainability of marine food webs. Therefore, developing 15 our capability to estimate and monitor, on a global scale, the concentrations of macro-16 molecules that determine phytoplankton energy value, would be invaluable. Reported 17 here are the first estimates of carbohydrate, protein, lipid, and overall energy value of 18 phytoplankton in the world oceans, using ocean-colour data from satellites. The esti-19 mates are based on a novel bio-optical method that utilises satellite-derived bio-optical 20 fingerprints of living phytoplankton combined with allometric relationships between phy-21 toplankton cells and cellular macromolecular contents. The annually-averaged phyto-22 plankton energy value, per cubic meter of sub-surface ocean, varied from less than 0.1 23 kJ in subtropical gyres, to 0.5–1.0 kJ in parts of the equatorial, northern and south-24 ern latitudes, and rising to more than 10 kJ in certain coastal and optically complex 25 waters. The annually-averaged global stocks of carbohydrate, protein and lipid were 26 0.044, 0.17 and 0.108 gigatonnes, respectively, with monthly stocks highest in September 27 and lowest in June, over 1997-2013. The fractional contributions of phytoplankton size 28 classes e.g., picoplankton, nanoplankton and microplankton to surface concentrations 29 and global stocks of macromolecules varied considerably across marine biomes classified 30 as Longhurst provinces. Among these provinces, the highest annually-averaged surface 31 concentrations of carbohydrate, protein, and lipid were in North-East Atlantic Coastal 32 Shelves, whereas, the lowest concentration of carbohydrate or lipid were in North At-33 lantic Tropical Gyral, and that of protein was in North Pacific Subtropical Gyre West. 34 The regional accuracy of the estimates and their sensitivity to satellite inputs are quanti-35 fied from the bio-optical model, which show promise for possible operational monitoring 36 of phytoplankton energy value from satellite ocean colour. Adequate in situ measure-37 ments of macromolecules and improved retrievals of inherent optical properties from 38 high-resolution satellite images, would be required to validate these estimates at local 39 sites, and to further improve their accuracy in the world oceans. 40

#### 41 Keywords

<sup>42</sup> Phytoplankton size spectra; ocean colour; carbohydrate; protein; lipid; energy content.

## <sup>43</sup> 1 Introduction

The autotrophic phytoplankton species in the upper ocean, constituting less than 1% of the 44 entire photosynthetic biomass on the globe, are responsible not only for  $\sim 50\%$  of the global an-45 nual carbon-fixation (Falkowski, 2012; Field et al., 1998), but also for providing life-support to 46 marine food-webs through its trophic connections. In addition to their biomass and species 47 composition, the cellular macromolecular contents and energy value of phytoplankton can 48 strongly impact the trophic balance within a marine ecosystem, e.g., by directly impacting 49 the developmental stages of grazers, and influencing the trophic-energy flow affecting the 50 production of higher trophic species (Breteler et al., 2005: Jónasdóttir, 1994; Litzow et al., 51 2006; Shin et al., 2003). The stoichiometric ratio, i.e., the relative elemental composition of 52 carbon, nitrogen and phosphorous in phytoplankton is known to vary with phytoplankton 53 assemblages across resource gradients in marine biomes (Geider and La Roche, 2002; Martiny 54 et al., 2013). Stoichiometric variations alter the nutritional quality of phytoplankton as food 55 to the grazers (Goldman and Caron, 1985; Sterner and Elser, 2002); and variations in nutri-56 ent bound or energy value of phytoplankton affect the stability and oscillatory dynamics of 57 producer-grazer interactions (e.g., Roy et al., 2005; Roy and Chattopadhyay, 2007a, b). It is, 58 therefore, imperative to monitor the variations in cellular macromolecular contents of marine 59 phytoplankton, on local, regional and global scales. In this context, possibilities of having 60 satellite-based estimates would be invaluable, given that in situ observations are often infre-61 quent, and inadequate for monitoring over large spatial scales. Moreover, conducting in situ 62 measurements of the macromolecular contents of phytoplankton in the global ocean, would 63 be extremely time consuming and considerably expensive. 64

Over the last two decades, several satellite-based methods have been developed to extend 65 our capabilities from routinely estimating chlorophyll concentration, to distinguishing phyto-66 plankton functional types (PFTs), in terms of the proportions of chlorophyll either in major 67 taxonomic groups, or in phytoplankton size classes (PSCs) (for more details, see, IOCCG, 68 2014; Mouw et al., 2017). Some progress has also been made to estimate phytoplankton car-69 bon (Behrenfeld et al., 2005; Kostadinov et al., 2016; Roy et al., 2017; Sathvendranath et al., 70 2009), and carbon-based classification of PSCs (Kostadinov et al., 2016; Roy et al., 2017), 71 from ocean colour. However, strong variations in phytoplankton cellular carbon and carbon-72 based macromolecules, with taxa, cell morphology, and environmental conditions such as 73 ambient light and available nutrient (Hitchcock, 1982; Marañón, 2008; Marañón et al., 2013; 74 Menden-Deuer and Lessard, 2000; Strathmann, 1967), impose additional layers of difficul-75 ties in converting the satellite-derived estimates of chlorophyll or carbon-to-macromolecular 76 concentrations. Certain phytoplankton macromolecules, such as cellular fatty acids, strongly 77 vary (e.g., between 1% and 85%, Chisti, 2007), not only among algal groups and species, but 78 also within a specific algal group, e.g., diatoms under different culture conditions (Opute, 79 1974). In addition to laboratory cultures, essential fatty acids in phytoplankton have also 80 been reported to vary with oceanographic conditions, such as sea-surface temperature and 81 chlorophyll-a, on regional scales (e.g., Budge et al., 2014; Pethybridge et al., 2015). However, 82 progress is yet to be made to estimate the variations in total phytoplankton lipid from satellite 83

data, on a global scale. Moreover, no method exists yet to estimate from satellite, either on a
regional or global scale, the spatiotemporal variations of other essential phytoplankton macromolecules, such as carbohydrate or protein. Given that the proportional contributions of these
macromolecules determine the energy value of phytoplankton, it would be useful to develop
an advanced ocean-colour-based method for estimating the macromolecular concentrations in
the ocean waters.

In this paper, the cellular macromolecular contents of marine phytoplankton, in partic-90 ular, the concentrations of carbohydrate, protein and lipid are estimated on a global scale, 91 for the first time, based on ocean-colour data from satellite remote-sensing. To do so, a 92 novel method is derived that utilises light-absorption coefficients of phytoplankton  $(a_{nh})$  - an 93 inherent optical property (IOP) retrievable from ocean colour (e.g., IOCCG, 2006), coupled 94 with allometric relationships between phytoplankton cells and their cellular macromolecular 95 contents, reported in the literature (Hitchcock, 1982; Marañón, 2008; Marañón et al., 2013, 96 2007; Menden-Deuer and Lessard, 2000; Moal et al., 1987; Peters, 1983; Strathmann, 1967). 97 The method builds on a semi-analytical algorithm for retrieving the exponent of the phyto-98 plankton size spectrum from satellite ocean colour, developed recently by Roy et al. (2013, 99 2011). The concentrations of the total macromolecular contents are further partitioned ac-100 cording to their contributions in three bulk PSCs, namely, picoplankton, nanoplankton and 101 microplankton. The estimates are obtained over the global ocean, and for different marine 102 biomes represented by Longhurst oceanographic provinces (Longhurst, 1995, 1998). Further, 103 insights on the estimation uncertainties are provided through detailed sensitivity analyses, 104 highlighting the possibilities of further improvements of the estimates, with the expectation 105 that the input satellite data would further improve, as the satellite era enters into higher 106 temporal and spatial resolution. 107

## $_{108}$ 2 Methodology

#### 109 2.1 Satellite validation

Global 4-km, level-3 mapped chlorophyll concentration, remote-sensing reflectance and the 110 IOPs were obtained from the European Space Agency's Ocean Colour Climate Change 111 Initiative (OC-CCI) project (freely available on http://www.esa-oceancolour-cci.org). 112 The OC-CCI data were produced by merging ocean-colour data from three satellite sen-113 sors: NASA-SeaWiFS, NASA-MODIS-Aqua and ESA-MERIS; further details on OC-CCI, 114 including data processing, temporal consistency of the data products and details of the algo-115 rithms used, can be found in Brewin et al. (2015); Müller et al. (2015). Monthly climatologies 116 of the mixed-layer depth were obtained on  $0.5^{\circ} \times 0.5^{\circ}$  spatial grid from Monthly Isopyc-117 nal & Mixed-layer Ocean Climatology (MIMOC, Schmidtko et al., 2013, available freely on 118 http://www.pmel.noaa.gov/mimoc/). To obtain depth-integrated estimates of the satellite-119 derived products from OC-CCI, the mixed-layer depths were remapped onto OC-CCI 4-km 120 grids using nearest-neighbour interpolation by implementing MATLAB2015b interpolation 121

routine (similar to previous studies, e.g., Roy et al., 2017).

A sufficiently large global *in situ* dataset on phytoplankton carbohydrate, protein and lipid that would ideally be required to validate the satellite-based estimates was unavailable. The historical *in situ* measurements on carbohydrate, protein and lipid, which were already compiled by Finkel et al. (2016*a*), did not cover the period over which satellite data (e.g., OC-CCI v2) were available (i.e., September 1997 onwards). These constraints on hindered satellite validation exercise in different oceanographic conditions.

Whilst direct measurements on carbohydrate, protein and lipid were unavailable, large 129 datasets on *in situ* phytoplankton abundance were available, e.g., those compiled in ma-130 rine biodiversity database (Sal et al., 2013), which included phytoplankton cell counts from 131 samples collected in different oceanographic cruises between 1992 and 2002, partly covering 132 the satellite period. Owing to the constraints on direct measurements, a validation exer-133 cise was attempted by converting the *in situ* data on phytoplankton abundance (Sal et al., 134 2013) into estimates of phytoplankton macromolecular concentrations, using allometric re-135 lationships from the literature (Finkel et al., 2016a). To do so, a subset of phytoplankton 136 abundance data (Sal et al., 2013) that overlapped with the OC-CCI v2 temporal coverage 137 (September 1997 - December 2013) were considered, and the concentrations of phytoplankton 138 carbohydrate, protein and lipid were computed using the information on phytoplankton cell 139 size (reported in Sal et al., 2013) and the corresponding allometric relationships (reported in 140 Finkel et al., 2016a). This subset included 250 samples collected from 1997 to 2002, across 141 various oceanographic regions (see Section 3.2, for the geographic locations); and consisted 142 of 943 species of diatom, dinoflagellate and coccolithophores with equivalent-spherical diam-143 eter ranging from  $1.34\,\mu\mathrm{m}$  to  $50\,\mu\mathrm{m}$  (to be consistent with the size range of microplankton 144 assumed within the algorithm, only the species with diameter  $<50 \ \mu m$  were considered). This 145 cell-diameter range covered nanoplankton, microplankton and a part of picoplankton. To be 146 consistent with previous studies (Roy et al., 2013, 2017), the diameter ranges of the three 147 phytoplankton size classes used in the model were picoplankton:  $0.2-2 \,\mu m$ , nanoplankton: 148  $2-20\,\mu\mathrm{m}$ , and microplankton:  $20-50\,\mu\mathrm{m}$ . Satellite matched-up chlorophyll concentrations and 149 IOPs were retrieved from OC-CCI data archive. Given that the sampling times were mostly 150 within the early years of SeaWiFS coverage (and SeaWiFS was the only contributing ocean-151 colour sensor over 1997-2002), a large number of gaps in satellite data were identified. To 152 maximise the number of validation data points, match-ups from composite satellite images 153 on daily (n = 39) and monthly (n = 249) scales were used. 154

The global annual stocks of phytoplankton carbohydrate, protein and lipid within the oceanic mixed layer were computed from the estimated surface concentrations, grid-by-grid, using the available mixed-layer depth values obtained from MIMOC (no specific depth profiles of the macromolecular concentrations were known from either *in situ* or remote sensing, on a global scale).

### <sup>160</sup> 2.2 Relating the size spectrum of phytoplankton to its cellular <sup>161</sup> macromolecular concentrations

Studies have shown that phytoplankton cell size strongly determines its cellular concen-162 trations of chlorophyll, carbon and carbon-based macromolecules through allometric rela-163 tionships (Hitchcock, 1982; Marañón, 2008; Marañón et al., 2013, 2007; Menden-Deuer and 164 Lessard, 2000; Moal et al., 1987; Peters, 1983; Strathmann, 1967). The allometric relation-165 ship between the cellular concentration of a macromolecule  $([M]_{cell}, \text{ expressed in the units})$ 166 of pg cell<sup>-1</sup>) and the volume of a phytoplankton cell  $(V_{cell}, \text{ in } \mu \text{m}^3)$  can be described by the 167 canonical equation:  $[M]_{cell} = a_M V_{cell}^{b_M}$ ; where, M stands for the macromolecule that can be 168 carbohydrate, protein or lipid, and  $a_M$ ,  $b_M$  are the allometric parameters with magnitudes 169 specific to a macromolecule M. For a given macromolecule,  $a_M$  and  $b_M$  would remain constant 170 across the size spectrum of phytoplankton cells. Assuming that the particle size distribution 171 of phytoplankton cells follows the power law (McCave, 1984; Reynolds et al., 2010; Sheldon 172 et al., 1972), the number of phytoplankton cells with equivalent spherical diameter D per 173 unit volume of seawater can expressed as:  $N(D) = k D^{-\xi}$ , with  $\xi$  as the exponent of the 174 phytoplankton size spectrum, and k as a constant related to the abundance of the total popu-175 lation. Following Roy et al. (2013), the concentration of phytoplankton chlorophyll-a ( $B_{total}$ ) 176 mg Chl m<sup>-3</sup>) within the cell-diameter range  $[D_{min}, D_{max}]$  can be expressed as a product of the 177 number of phytoplankton cells within that size class, the volume of each cell  $(\pi D^3/6)$ , and 178 the intracellular concentration of chlorophyll-a  $c_i$  (in mg m<sup>-3</sup>, parameterised as  $c_i = c_0 D^{-m}$ 179 with the magnitudes of  $c_0 = 3.9 \times 10^6$ , and m = 0.06 by Roy et al., 2011 using the *in situ* 180 measurements of Marañón et al., 2007), as follows: 181

$$B_{total} = \int_{D_{min}}^{D_{max}} \left[ \left( \frac{\pi}{6} D^3 \right) (c_0 D^{-m}) (k D^{-\xi}) \right] dD = \left( \frac{\pi}{6} k c_0 \right) \frac{D_{max}^{4-\xi-m} - D_{min}^{4-\xi-m}}{4-\xi-m}.$$
 (1)

Similarly, the total concentration of the macromolecule M (in mg m<sup>-3</sup>) due to all phytoplankton cells within a diameter range  $[D_{min}, D_{max}]$  can be expressed as a product of the number of cells and the cellular concentration  $[M]_{cell}$ :

$$[M]_{total} = \int_{D_{min}}^{D_{max}} [N(D) \times [M]_{cell}] dD = \int_{D_{min}}^{D_{max}} \left(kD^{-\xi}\right) \left[10^{-9} a_M \left(10^{18} \frac{\pi}{6} D^3\right)^{b_M}\right] dD,$$
  
$$= 10^{-9} k a_M \left(10^{18} \frac{\pi}{6}\right)^{b_M} \left(\frac{D_{max}^{3b_M-\xi+1} - D_{min}^{3b_M-\xi+1}}{3b_M - \xi + 1}\right); \qquad (2)$$

with the condition that  $[M]_{total} \rightarrow [10^{-9} k a_M (10^{18} \pi/6)^{b_M} \log_e (D_{max}/D_{min})]$ , when  $\xi \rightarrow (3b_M + 1)$ , applied to avoid division by zero. The factors  $10^{-9}$  and  $10^{18}$  are associated with the conversions of units from pg to mg, and m<sup>3</sup> to  $\mu$ m<sup>3</sup> respectively. Using Eqs. (1) and (2), the ratio of the macromolecular concentration to the chlorophyll concentration ( $\chi_M$ ) can be expressed as:

$$\chi_M = \frac{[M]_{total}}{B_{total}} = \frac{10^{-9} a_M \left(10^{18} \pi/6\right)^{b_M}}{(\pi/6) c_0} \left(\frac{D_{max}^{3b_M - \xi + 1} - D_{min}^{3b_M - \xi + 1}}{D_{max}^{4 - \xi - m} - D_{min}^{4 - \xi - m}}\right) \left(\frac{4 - \xi - m}{3b_M - \xi + 1}\right).$$
(3)

Note that the expression of macromolecule-to-chlorophyll ratio  $\chi_M$  in Eq. (3) does not depend on the parameter k appearing in Eqs. (1) and (2). So, once  $\chi_M$  is computed,  $M_{total}$  can be computed from the observed value of  $B_{total}$  as:

$$M_{total} = \chi_M B_{total},\tag{4}$$

provided that  $\xi$ ,  $a_M$  and  $b_M$  of the population are known (see Sections 2.3, 2.4).

#### <sup>194</sup> 2.3 Size-partitioned cellular contents of phytoplankton

Assuming that the total biomass of phytoplankton is a sum of the biomasses of n nonoverlapping PSCs defined by cell-diameter ranges  $[D_i, D_j]$  with  $0 \le i < j \le n$ ,  $[M]_{total} = \sum [M]_{ij}$ , where  $[M]_{ij}$  denote the macromolecular concentration within the size class [i, j]. It follows from Eq. (4), that  $[M]_{ij} = \chi_{Mij} B_{ij}$ , with  $\chi_{Mij}$  and  $B_{ij}$ , respectively, are the macromolecule-to-chlorophyll ratio and the concentration of chlorophyll  $B_{ij}$  in the size class  $[D_i, D_j]$ , where  $\chi_{Mij}$  follows directly from using Eq. (3):

$$\chi_{M\,ij} = \frac{10^{-9} \, a_M \, \left(10^{18} \, \pi/6\right)^{b_M}}{\left(\pi/6\right) c_0} \left[\frac{D_j^{3b_M-\xi+1} - D_i^{3b_M-\xi+1}}{D_j^{4-\xi-m} - D_i^{4-\xi-m}}\right] \left[\frac{4-\xi-m}{3b_M-\xi+1}\right],\tag{5}$$

and the expression of  $B_{ij}$  is taken from Roy et al. (2013), so that,

$$[M]_{ij} = \chi_{M\,ij} \, B_{ij} = \chi_{M\,ij} \, \left( \frac{D_j^{4-\xi-m} - D_i^{4-\xi-m}}{D_{max}^{4-\xi-m} - D_{min}^{4-\xi-m}} \right) \, B_{total}; \tag{6}$$

202 and therefore,

$$[M]_{total} = \sum_{i=0, j=i+1}^{i=n-1, j=n} [M]_{ij} = \frac{B_{total}}{D_{max}^{4-\xi-m} - D_{min}^{4-\xi-m}} \sum_{i=0, j=i+1}^{i=n-1, j=n} \left[ \chi_{Mij} \left( D_j^{4-\xi-m} - D_i^{4-\xi-m} \right) \right] (7)$$

<sup>203</sup> Also, the fraction of  $[M]_{ij}$  to  $[M]_{total}$  can be computed as:

$$F_{M,ij} = \frac{[M]_{ij}}{[M]_{total}} = \frac{\chi_{M\,ij}, \left(D_j^{4-\xi-m} - D_i^{4-\xi-m}\right)}{\sum_{i=0,\,j=i+1}^{i=n-1,\,j=n} \left[\chi_{M\,ij} \left(D_j^{4-\xi-m} - D_i^{4-\xi-m}\right)\right]}.$$
(8)

Using the equations derived above, the concentrations of carbohydrate, protein and lipid can be partitioned into any number of PSCs. However, for the sake of discussion, in this study, the estimates are obtained for three major PSCs, namely, picoplankton, nanoplankton and microplankton, with cell-diameter bounds  $[D_0, D_1]$ ,  $[D_1, D_2]$  and  $[D_2, D_3]$ , respectively, where  $D_0 = 0.25 \ \mu m$ ,  $D_1 = 2 \ \mu m$ ,  $D_2 = 20 \ \mu m$ , and  $D_3 = 50 \ \mu m$  based on previous studies (Roy et al., 2013; Sieburth et al., 1978; Vidussi et al., 2001).

## 210 2.4 Allometric parameters $a_M$ and $b_M$ from the literature, and re-211 trieval of $\xi$ from satellite data

The allometric parameters  $a_M$  and  $b_M$  corresponding to phytoplankton species are reported 212 in several studies e.g., Finkel et al. (2016a); Hitchcock (1982); Menden-Deuer and Lessard 213 (2000); Moal et al. (1987). More recently, Finkel et al. (2016a) compiled a large database of 214 macromolecular concentrations in various eukaryotic microalgae from 53 published studies, 215 covering various taxonomic groups, culture conditions and growth phases; and reported the 216 allometric relationships between cell volume and concentrations of carbohydrate, protein, 217 and lipid in phytoplankton. In the current study,  $a_M$  and  $b_M$  are fixed based on Finkel et al. 218 (2016a) (see, their Table-II), and their reported values along with the confidence intervals are 219 used for estimating the macromolecular concentrations and performing uncertainty analyses 220 described in Section 2.5). 221

The exponent of the phytoplankton size spectrum  $\xi$  is retrieved from the specificabsorption coefficient of phytoplankton at 676 nm using a semi-analytical ocean-colour algorithm developed by Roy et al. (2013). For completeness, the major steps of this methodology are described in the Supplementary Materials, without fully reproducing it from Roy et al. (2013). However, for further details on the parameterisation and optimization steps related the retrieval of  $\xi$ , readers are referred to Roy et al. (2013, 2011).

#### 228 2.5 Uncertainties and biases

Although the method described above is founded on theories of light-absorption properties 229 and cellular allometric relationships of phytoplankton, the estimates need to be validated 230 against direct *in situ* measurements, which are currently unavailable. This limitation raises 231 the possibility of bias and uncertainties in satellite products at each pixel, leading to biased 232 estimates of the macromolecules on a global scale. The inaccuracy of the estimates may arise 233 from several sources, the most prominent of which is the uncertainties associated with the 234 satellite products used as inputs to the model, e.g., chlorophyll-a and absorption coefficients 235 of phytoplankton. The uncertainties in chlorophyll-a retrievals for optically complex (Case 236 II) waters are considerably large, when compared within those for the open oceans (Case 237 I waters), mainly due to the limitations of the empirical chlorophyll algorithms used (e.g., 238 IOCCG, 2000). The absorption coefficients of phytoplankton, on the other hand, being an 230 IOP are retrieved generally by semi-analytical algorithms, the performance of which also vary 240 for optically complex waters (e.g., IOCCG, 2006). 241

In the coastal oceans and optically complex waters, the retrievals are affected due to the presence of high concentration of coloured-dissolved organic matters (CDOM), sediments, other suspended materials and water constituents that interfere with light penetration and reflectance (IOCCG, 2000). Uncertainties in remote sensing retrievals can further be attributed to clouds, ice covers, solar zenith angles, sun glint, atmospheric dusts and aerosols (e.g., IOCCG, 2000; Maritorena et al., 2010). Thus, the satellite-derived estimates of carbohydrate protein and lipid presented on global maps (in the result section) comes with uncertainty and bias, an accurate estimation of which would be possible only when adequate *in situ* measurements on these quantities become available.

Nevertheless, to understand and quantify the overall uncertainty levels in the satellite-251 derived estimates, a model sensitivity analysis was carried out. Theoretically, accurate esti-252 mations of the macromolecular concentrations in phytoplankton based on the above method 253 would depend on the allometric parameters  $(a_M \text{ and } b_M)$  and the estimates of  $\xi$ . The re-254 trieval of  $\xi$  further depends on satellite-derived estimates of chlorophyll-a and  $a_{ph}$ . Using 255 Eqs. (1-3), the relative sensitivities of the estimates of  $M_{total}$ , i.e.,  $\frac{\Delta M_{total}}{M_{total}}$ , can be computed 256 as a combined function of  $\frac{\Delta\xi}{\xi}$ ,  $\frac{\Delta a_M}{a_M}$ , and  $\frac{\Delta b_M}{b_M}$ . Following Roy et al. (2013), where  $\frac{\Delta\xi}{\xi}$  are reported pixel-by-pixel in the global ocean, a maximum overall  $\frac{\Delta\xi}{\xi}$  in the range 0–25% is 257 258 considered. For  $\frac{\Delta a_M}{a_M}$  and  $\frac{\Delta b_M}{b_M}$ , the half of the 95% spread with respect to the mean levels 259 reported by Finkel et al. (2016*a*) are considered. The resultant  $\frac{\Delta M_{total}}{M_{total}}$  are then computed pixel-by-pixel, as percentages of the default estimates. So, without the availability of ade-260 261 quate in situ measurements, the uncertainties discussed in the following sections should be 262 interpreted as model-based uncertainties, and not as those based on the *in situ* observations. 263

## <sup>264</sup> 3 Results and discussion

#### <sup>265</sup> 3.1 Macromolecular concentrations across phytoplankton size range

The ratios of carbohydrate-to-chlorophyll ( $\chi_{carbo}$ ), protein-to-chlorophyll ( $\chi_{prot}$ ) and lipid-tochlorophyll ( $\chi_{lipid}$ ) increase with  $\xi$  within the ranges given by [5.0, 9.5], [7.1, 48.9] and [3.1, 32], respectively (Fig. 1a). For any given value of  $\xi$ ,  $\chi_{prot}$  is higher than  $\chi_{carbo}$  and  $\chi_{lipid}$ . For low values of  $\xi$ ,  $\chi_{lipid}$  is lower than  $\chi_{carbo}$ , but it increases more rapidly with the assemblages of small phytoplankton cells, and so, for high values of  $\xi$ ,  $\chi_{lipid}$  is significantly higher than  $\chi_{carbo}$  (Fig. 1a).

The proportions of carbohydrate, protein and lipid increase with  $\xi$  in picoplankton 272 (Fig. 1b), and decrease with  $\xi$  in microplankton (Fig. 1d), but are unimodal in nanoplank-273 ton having magnitudes typically less than 50% with highest values in the middle rage of  $\xi$ 274 (Fig. 1c). At any given level of  $\xi$ , the proportion of lipid in picoplankton is higher than that of 275 carbohydrate or protein (with carbohydrate < protein < lipid) (Fig. 1b); but in microplank-276 ton the order is reversed to carbohydrate > protein > lipid (Fig. 1d). For nanoplankton these 277 proportions alter from carbohydrate < protein < lipid at the lower end of  $\xi$  to carbohydrate >278 protein > lipid at the higher end of  $\xi$  (Fig. 1c). These results show strong dependencies of phy-279 toplankton size structure on the available macromolecular concentrations with implications 280 on their stocks in mixed populations of phytoplankton. 281

For carbohydrate estimates, the relative uncertainties would be <30% for  $3.25 < \xi < 5$ (typically representing small-cell dominated populations), but would increase up to 60% at the lower end of  $\xi$  (typically representing large-cell dominated populations) (Fig. 1e, Table 1).

For protein estimates (Fig. 1f), the relative uncertainties would be <40% across the range of 285  $\xi$  provided that the relative uncertainty in  $\xi$  is <10%. If the relative uncertainties in  $\xi$  are 286 >15%, the uncertainties in protein would increase to >60% typically for  $3.25 < \xi < 4.5$ , but 287 would generally remain within <40% for populations dominated by either large or small cells 288 (i.e., at the low and high ends of  $\xi$ , see Table 1 for more details). For lipid estimates, the 289 relative uncertainties would be similar to those for protein: <40% for the low and high ends 290 of  $\xi$ , but >60% for the mid-range of  $\xi$ , if the uncertainties in  $\xi$  is >15% (Fig. 1g). Further 291 details on these uncertainty estimates for various combinations of uncertainties in  $\xi$  estimates 292 (based on Fig. 1e-g) are summarised in Table 1, and the propagations of the uncertainties in 293 the global ocean are discussed in Sections 3.7. 294

#### <sup>295</sup> 3.2 Comparison with estimates based on *in situ* abundance data

The matched-up in situ data were from specific cruises (see, Fig. 2a) with moderate sample 296 size having non-normal distribution; therefore, non-parametric statistics were implemented, 297 in particular, Spearman's correlation instead of Pearson's, and other non-parametric matrices 298 following Werdell et al. (2009). The *in situ* and satellite-based estimates generally follow 299 the 1 : 1 line, but with some level of spread around it (Fig. 2b-d, Supplementary Fig. S1), 300 with significant correlations (Spearman's  $\rho$ ) between them on linear scale, for carbohydrate: 301  $\rho = 0.25$ , p < 0.001; protein:  $\rho = 0.24$ , p < 0.001; and lipid:  $\rho = 0.23$ , p < 0.001 (Fig. 2b-d). 302 The root mean squared error (RMSE) and bias of the estimates vary for carbohydrate (RMSE) 303 10.20, bias  $-7.28 \text{ mg m}^{-3}$ ), protein (RMSE 21.55, bias  $-10.93 \text{ mg m}^{-3}$ ) and lipid (RMSE 9.77, 304 bias  $-4.87 \text{ mg m}^{-3}$ ). As expected, the RMSE and bias for daily match-ups, turn out to be 305 lower than those for monthly match-ups (see, Supplementary Table S1); but in both cases 306 their magnitudes are within a reasonable range, when compared with those for other derived 307 products, such as phytoplankton carbon (Kostadinov et al., 2016; Roy et al., 2017). 308

Following Werdell et al. (2009), three further metrices are computed for comparing the 309 estimates with daily (monthly) match-ups: the median satellite-to-in-situ-ratio (median ra-310 tio, found to be 0.51(0.71), 0.59(0.73), and 0.59(0.73) respectively), the median of the 311 relative-percent difference (median RPD, found to be -49.41 (-29.36), -40.86 (-27.38) and 312 -41.11 (-26.75), respectively), and the semi-interquartile percent differences (SIQ-PD, found 313 to be -48.50 (-67.65), -50.66 (-65.08) and -51.36 (-63.82), respectively) (see, Supplemen-314 tary Table S1). The median RPDs and SIQ-PDs are lowest for lipid estimates, followed by 315 those for protein and carbohydrate (Supplementary Table S1). The median ratios are < 1, 316 suggesting that the algorithm would generally underestimate the macromolecular concentra-317 tions (Fig. 2e). Also, the algorithm seems to produce relatively less natural variability of the 318 macromolecular concentrations, in comparison with those estimated from the *in situ* abun-319 dance data (Fig. 2e). However, it is worth mentioning that the median ratio, median RDP, 320 SIR-PD for SeaWiFS chlorophyll were reported (Werdell et al., 2009) to be in the ranges 321 [1.7, 81.5], [-34.7, 122.3], and [0.88, 1.69], respectively. Therefore, in terms of these metri-322 ces, the accuracy of the current estimates of the macromolecular concentrations are generally 323 comparable with that reported for SeaWiFS chlorophyll. 324

Nevertheless, these comparisons would be affected by several layers of uncertainties associated with the *in situ* and satellite estimates. For example, prominent natural variability of cell size of the 943 phytoplankton species would alter the *in situ* estimates of carbohydrate, protein and lipid, which were not possible to include in the *in situ* calculations; and the uncertainties in satellite inputs (chlorophyll, IOPs) would also affect the satellite retrievals of  $\xi$ (also see, Section 3.7).

# <sup>331</sup> 3.3 Phytoplankton carbohydrate, protein and lipid in the world <sup>332</sup> oceans

Strong spatial variability of the annually-averaged  $\chi_{carbo}$ ,  $\chi_{prot}$ ,  $\chi_{lipid}$ , carbohydrate, protein 333 and lipid are found over the world's oceanic biomes, for the period of study (Fig. 3). The 334 magnitude of  $\chi_{carbo}$  varies from <5 in the high-chlorophyll coastal waters and large parts of 335 the northern latitudes beyond 40 degree north (Fig. 3a,c), to >9 in the open oceans and Case I 336 waters (Fig. 3c). Similarly,  $\chi_{prot}$  (Fig. 3e) or  $\chi_{lipid}$  (Fig. 3g) vary, respectively, from <15 or <10 337 in the coastal waters and northern latitudes, to >45 or >30, respectively, in the open oceans 338 and Case I waters. These results generally reflect that the oceanographic regions dominated 339 by large and small phytoplankton are respectively represented by low and high values of  $\chi_{carba}$ 340  $\chi_{prot}$  or  $\chi_{lipid}$ . In the Atlantic and Pacific subtropical gyres, despite the high magnitudes of 341  $\chi_{carbo}, \chi_{prot}$  and  $\chi_{lipid}$ , the concentrations of carbohydrate, protein and lipid are typically low 342  $(< 0.5, 1.0 \text{ and } 1.0 \text{ mg m}^{-3}, \text{respectively})$ , and the spatial pattern is similar to the distribution 343 of low chlorophyll. Most of the coastal oceans and Case II waters are generally characterised 344 by higher than 5, 10 and 10 mg m<sup>-3</sup> of carbohydrate, protein and lipid, respectively, which in 345 places spike beyond 50, 100 and 100 mg m<sup>-3</sup>, respectively (Fig. 3d,f,h). It is noteworthy that 346 some of these very high values may be attributed to the uncertain or erroneous retrievals of 347 chlorophyll and other optical properties in the optically complex water (as also discussed in 348 Section 2.5). 349

Applying the macromolecular concentration-to-energy conversion factors, i.e, 4.2 kcal g<sup>-1</sup> for carbohydrate, 4.19 kcal g<sup>-1</sup> for protein, 9.5 kcal g<sup>-1</sup> for lipid (Finkel et al., 2016*a*; Hitchcock, 1982), the chemical-energy values of the surface-ocean phytoplankton can be computed (Fig. 3b,d,f,h). The annual average of the phytoplankton energy-value is generally less than 0.1 kJ per m<sup>-3</sup> of ocean water in the subtropical gyres, but goes up to 0.5–1.0 kJ per m<sup>-3</sup> in parts of the equatorial, northern and southern latitudes, and beyond 10 kJ per m<sup>-3</sup> in certain coastal and optically complex waters (Fig. 3b).

# <sup>357</sup> 3.4 Size-partitioned phytoplankton carbohydrate, protein and lipid <sup>358</sup> in the world oceans

Picoplankton contributions to carbohydrate (in the range  $[0.1, 1.0] \text{ mg m}^{-3}$ ), protein (in the range  $[1.0 \ 5.0] \text{ mg m}^{-3}$ ) or lipid (in the range  $[0.5, 3.0] \text{ mg m}^{-3}$ ) dominate over the contributions of nanoplankton and microplankton in the open oceans and equatorial gyres (Fig. 4a,d,g). In the northern latitudes beyond 40 degrees and in coastal waters, microplankton contributions to carbohydrate, protein and lipid are higher than those of picoplankton and nanoplankton, with approximate ranges [2.5, 10], [2.0 25] and [0.5, 5.0] mg m<sup>-3</sup>, respectively (Fig. 4c,f,i). Nanoplankton contributions are generally in the range [1, 3] mg m<sup>-3</sup> of carbohydrate, [1, 5] mg m<sup>-3</sup> of protein and [1, 5] mg m<sup>-3</sup> of lipid, respectively (Fig. 4b,e,h), except in the oligotrophic gyres, where all the concentrations reduce to less than 0.05 mg m<sup>-3</sup> (Fig. 4b,e,h).

#### 368 3.5 Macromolecular concentrations in Longhurst provinces

The geographical variations of carbohydrate, protein and lipid in the world oceans can be in-369 ferred from their regionally-binned concentrations in the Longhurst biogeographical provinces 370 (Longhurst, 1995, 1998). Given that the ocean-colour data from satellites are inadequate (and 371 may be more erroneous) in the polar regions over most of the year, the estimates from the po-372 lar provinces (6 out of 54 Longhurst provinces) are excluded from further discussion. For the 373 remaining 48 provinces, the spatial estimates of  $\chi_{carbo}$ ,  $\chi_{prot}$ ,  $\chi_{lipid}$  and the concentrations car-374 bohydrate, protein and lipid are computed from their corresponding annually-averaged global 375 maps (Fig. 5). These provinces include 14 Westerlies (NADR, GFST, NASW, MEDI, NASE, 376 PSAE, PSAW, KURO, NPPF, NPSW, TASM, SPSG, SSTC, SANT), 12 Trades (NATR, 377 WTRA, ETRA, SATL, CARB, MONS, ISSG, NPTG, PNEC, PEQD, WARM, ARCH) and 378 22 Coastal (NECS, CNRY, GUIN, GUIA, NWCS, BRAZ, FKLD, BENG, EAFR, REDS, 379 ARAB, INDE, INDW, AUSW, ALSK, CCAL, CAMR, CHIL, CHIN, SUND, AUSE, NEWZ) 380 provinces (full names of the provinces are given in Supplementary Table S2, and the descrip-381 tions in Longhurst, 1995, 1998). The Westerlies, Trades and Coastal provinces are shown in 382 Fig. 2a. 383

Spatial variability of the estimates in the Coastal provinces are found to be higher than 384 those in the Westerlies or Trades provinces (Fig. 5), with the lowest variability in the West-385 erlies provinces (Fig. 3-5), reflecting that coastal upwellings would strongly influence the dis-386 tribution of phytoplankton macromolecules (similar to chlorophyll distribution). The spatial 387 medians of  $\chi_{carbo}$ ,  $\chi_{prot}$  and  $\chi_{lipid}$  are lowest (5.69, 13.86 and 8.0, respectively) for the NWCS 388 (North-West Atlantic Coastal Shelves) province, and highest (8.95, 45.13, 29.56, respectively) 389 for the NPSW (North Pacific Subtropical Gyre West) province (Fig. 5a,b,c, and Supplemen-390 tary Table S2). The NECS (North-East Atlantic Coastal Shelves) province is characterised 391 by the highest surface concentrations (Fig. 5d,e,f) of the annually-averaged spatial medians 392 of carbohydrate (9.53 mg m<sup>-3</sup>), protein (25.2 mg m<sup>-3</sup>), and lipid (14.81 mg m<sup>-3</sup>). The low-393 est surface concentrations (spatial median) of carbohydrate  $(0.60 \text{ mg m}^{-3})$  and lipid  $(1.75 \text{ m}^{-3})$ 394  $mg m^{-3}$ ) are obtained in the NATR (North Atlantic Tropical Gyral) province (Fig. 5d,f, Ta-395 ble S2), whereas, the lowest concentrations of protein  $(2.11 \text{ mg m}^{-3})$  is obtained in the NPSW 396 (North Pacific Subtropical Gyre West) province (Fig. 5e Table S2), both of which are generally 397 populated by small picoplankton throughout the year. 398

The size-partitioned estimates also vary considerably across the 48 Longhurst provinces (Table S3, also Fig. 4). The spatial medians of picoplankton carbohydrate, protein and lipid

are lowest (0.13, 1.11 and 0.83 mg m<sup>-3</sup>, respectively) in the MEDI (Mediterranean Sea, Black 401 Sea) province, and highest (1.87, 13.33 and 9.43 mg m<sup>-3</sup>, respectively) in the NECS (NE 402 Atlantic Coastal Shelves) province (Table S2, Fig. 4a,d,g). For nanoplankton, the median 403 concentrations vary from their lowest values (0.09, 0.13 and 0.06 mg m<sup>-3</sup>, respectively) in the 404 WARM (W. Pacific Warm Pool Trades) province, to their highest values (3.34, 7.36 and 3.66 405  $mg m^{-3}$ , respectively) in the CHIN (China Sea Coastal) province (Table S2, Fig. 4b,e,h). For 406 microplankton, the median concentrations of carbohydrate and protein vary from their lowest 407 values (0.01 mg m<sup>-3</sup> for both) in the WARM province, to their highest values (3.39 and 3.37) 408  $mg m^{-3}$ , respectively) in the CHIN province; but that for lipid is found to be highest (1.42) 409  $mg m^{-3}$ ) in the NECS (NE Atlantic Coastal Shelves) province, and lowest (0.01  $mg m^{-3}$ ) in 410 the WARM province (Table S2, Fig. 4c, f, i). Unsurprisingly, the province-wise distribution of 411 the three macomolecular concentrations show spatial patterns generally consistent with our 412 understanding of the biogeography of phytoplankton size structure. 413

### <sup>414</sup> 3.6 Global-ocean stocks of phytoplankton macromolecules

The annually-averaged global stocks are: 0.044 Gt of carbohydrate with monthly range [0.041, 0.05] Gt; 0.17 Gt of protein with monthly range [0.155, 0.18] Gt; and 0.108 Gt of lipid with monthly range [0.098, 0.121] Gt (Fig. 6, and Supplementary Table S4). The largest global stocks are obtained in the month of September, which generally matches with the time of phytoplankton bloom in large parts of the equatorial-southern hemisphere (Kostadinov et al., 2017). The smallest stocks are obtained in the month of June, generally after the termination of the spring blooms.

The percentages of the size-partitioned carbohydrate, protein and lipid stocks also vary 422 over the months of the years (Fig. 6). The stocks constitute the lowest percentage of picoplank-423 ton carbohydrate  $\sim 46\%$  (equivalent to 0.02 Gt, with monthly range of 43-53\%), compared with 424 the percentages of picoplankton protein  $\sim 78\%$  (equivalent to 0.133 Gt, with monthly range of 425 76-83%), and picoplankton lipid  $\sim 85\%$  (equivalent to 0.092 Gt, with monthly range of 83-88%) 426 (Supplementary Table S4). The stocks further constitute  $\sim 33\%$  of nanoplankton carbohydrate 427 (equivalent to 0.015 Gt, with monthly range of 32-36%), which is considerably higher that the 428 percentages of nanoplankton protein  $\sim 17\%$  (equivalent to 0.028 Gt, with monthly range of 429 14-18%), and nanoplankton lipid  $\sim 12\%$  (equivalent to 0.013 Gt, with monthly range of 10-430 13%). Similarly, the percentage of microplankton carbohydrate  $\sim 21\%$  (equivalent to 0.009 Gt, 431 with monthly range of 16-24%) is significantly higher than the percentages of microplankton 432 protein  $\sim 5\%$  (equivalent to 0.009 Gt, with monthly range of 3-7%) and microplankton lipid 433  $\sim 3\%$  (equivalent to 0.003 Gt, with monthly range of 2-4%). But clearly, for any given macro-434 molecular stock, the largest contribution comes from picoplankton and the smallest from 435 microplankton (Fig. 6). 436

<sup>437</sup> No previous estimates were available to compare with the stocks of carbohydrate, protein <sup>438</sup> and lipid reported here. However, the carbon-based macromolecular stocks could be viewed <sup>439</sup> in conjunction with the stocks of total phytoplankton biomass (in carbon units), which were

estimated previously from satellite remote sensing (e.g., Behrenfeld et al., 2005; Kostadinov 440 et al., 2016; Roy et al., 2017). For example, the anually-averaged stocks of the total phyto-441 plankton biomass varied between 0.2 GtC to 1.0 GtC depending on the estimation method 442 (e.g., Behrenfeld et al., 2005; Falkowski et al., 1998; Kostadinov et al., 2016; Roy et al., 2017; 443 Stramski et al., 2008; Taylor et al., 2012). The annually-averaged stocks of carbohydrate 444 protein and lipid and their sum total, estimated above, are within this range. Recent studies 445 (Finkel et al., 2016a, b) also suggested that under 'nutrient-sufficient, exponential growth con-446 ditions' the median composition of the dry weight of microalgae contains 15% carbohydrate 447 32.2% protein and 17.3% lipid. With respect to the most recent satellite-based estimates of 448 phytoplankton biomass (i.e.,  $\sim 0.3$  GtC, based on Kostadinov et al., 2016; Roy et al., 2017), 449 the percentages of the annually-averaged global stocks (which included both nutrient suffi-450 cient and oligotrophic waters) of carbohydrate, protein and lipid are  $\sim 15\%$ ,  $\sim 57\%$ ,  $\sim 36\%$ , 451 respectively. These preliminary results thus suggests that on a global scale, the relative pro-452 portions of carbohydrate in phytoplankton might be more robust than the proportions of 453 protein and lipid. However, direct *in situ* measurements would be required to further validate 454 these results. 455

### 456 3.7 Algorithm uncertainties on global map

The uncertainty propagation maps based on the sensitivity analysis suggest that the relative 457 uncertainties in lipid estimates would be higher than those in protein or carbohydrate for 458 most of the world's productive regions (Fig. 7); but in the less productive oligotrophic waters, 459 the relative uncertainties in all the estimates would be generally comparable. The relative 460 uncertainties in carbohydrate estimates would be within 30-45% in most of the upwelling and 461 productive regions and coastal waters, but would reduce to <15% in the subtropical gyres and 462 oligotrophic waters (Fig. 7a). Similar spatial pattern are obtained for the relative uncertainties 463 in protein and lipid estimates, although the magnitudes of the relative uncertainties would 464 be different. For protein and lipid the relative uncertainties would be <15% and <25%. 465 respectively, inside the gyres, and between 30 - 40% and 35 - 50%, respectively, in major 466 parts of the Northern hemisphere; and but would increase up to 60 - 64% and 65 - 80%. 467 respectively, in large parts of the southern ocean and around the overlapping regions of the 468 oligotrophic and eutrophic waters (Fig. 7b,c). 469

## 470 4 Concluding remarks

A71 Although a variety of satellite-based ocean-colour algorithms have already been developed 472 to retrieve chlorophyll-a and its contributions in PFTs and PSCs (e.g., review by Mouw 473 et al., 2017), and phytoplankton carbon (Behrenfeld et al., 2005; Kostadinov et al., 2016; 474 Roy et al., 2017; Sathyendranath et al., 2009), no methodology exists so far to estimate from 475 satellites, the concentrations of macromolecules that essentially determine the energy value 476 of phytoplankton. The bio-optical method presented here would be the first one to compute,

from satellite data, the concentrations of phytoplankton carbohydrate, protein and lipid, and 477 the resultant energy value of phytoplankton on a global scale. In this novel approach, the 478 satellite-derived bio-optical fingerprints of the living phytoplankton combined with allomet-479 ric relationships are used, which builds on the ocean-colour algorithms recently developed 480 for retrieving phytoplankton cell size, the exponent of the phytoplankton size spectra, phyto-481 plankton carbon and PSCs from satellite (Roy et al., 2013, 2011, 2017). Presented are the first 482 estimates of annually-averaged concentrations of carbohydrate, protein, lipid, and ratios of 483 chlorophyll-a to cellular macromolecular concentrations over the global oceans as well as those 484 for the Longhurst biogeochemical provinces, over the period 1997-2013. Although the current 485 estimates are based on the OC-CCI merged satellite products, by design, the methodology 486 would be equally applicable to ocean-colour data from any other satellite sensor. 487

Recent studies based on either ocean-colour data (Behrenfeld et al., 2005; Kostadinov et al., 488 2016; Roy et al., 2017; Stramski et al., 2008) or Earth System models (e.g., CMIP5, Taylor 489 et al., 2012), have attempted to improve the estimates of the stocks of phytoplankton carbon, 490 and have narrowed down the estimation range of the annually-averaged stocks. But unclear is 491 how the total carbon stock partitions into the stocks of essential carbon-based macromolecules 492 in phytoplankton. For example, although the proportions of the macromolecules to dry weight 493 of phytoplankton are reported for ideal nutrient-rich conditions (Finkel et al., 2016 a, b), little 494 in known about those proportions in diverse oceanographic regions where growth conditions 495 deviate from ideal. This study independently estimates the annually-averaged stocks of the 496 three essential phytoplankton macromolecules, and finds that the sum total of these estimates 497 are well within the range of the reported stocks of total phytoplankton carbon. The estimates 498 would be potentially useful for understanding the cellular allocation of carbon to carbohydrate. 499 protein and lipid pools in phytoplankton, both spatially and over time, with implications for 500 trophic transfer models, and higher trophic or fisheries models. 501

The lack of adequate direct measurements on carbohydrate, protein and lipid overlap-502 ping the temporal coverage of the ocean-colour data have restricted rigorous validation of the 503 satellite-derived estimates. Therefore, new in situ measurements of phytoplankton macro-504 molecules across various oceanic conditions should be a priority, for increasing the reliability 505 and reducing the bias and uncertainties of the satellite-based estimates. Adequate direct 506 measurement would also allow computation of observation-based uncertainties such as RMSE 507 and bias, pixel-by-pixel, and providing those to the users. The sensitivity analyses carried 508 out here, with assumptions on fixed relative uncertainties of <30% for the input parameters 509 (following the requirement provided by Global Climate Observing System, GCOS, 2011), have 510 identified oceanographic regions where the estimates would be less (or more) sensitive to rela-511 tive uncertainties in satellite inputs. But, how the relative uncertainties may alter (reduce or 512 increase), due to regional variations of uncertainties in the input parameters, and how those 513 may impact the estimates of the global stocks, would require further investigations. The 514 sensitivity analyses however have shown promise that the estimation errors could reduce, as 515 the retrievals of satellite-based IOPs become more accurate. Finally, due to the constraints of 516 inadequate in situ validation data, and large uncertainties and biases in the optically complex 517 waters, arising from the presence of high concentration of coloured-dissolved organic matters, 518

sediments, clouds and ice, the current estimates may be less reliable in coastal waters and high latitudes, than those in open oceans. So, the applicability and reliability of the estimates to optically complex waters would also be subject to further investigations, possibly including improved satellite inputs, as the satellite era enters into higher temporal and spatial resolution.

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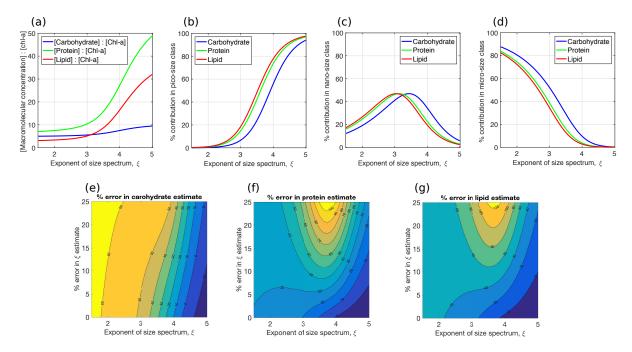
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## <sup>680</sup> Figure captions

Figure 1: (a) Carbohydrate-to-chlorophyll ( $\chi_{carbo}$ ), protein-to-chlorophyll ( $\chi_{prot}$ ) and lipid-to-chlorophyll ( $\chi_{lipid}$ ) ratios of the mixed-phytoplankton population derived (using Eq. 3) as functions of the exponent of the phytoplankton size spectrum ( $\xi$ ). (b)-(d) Size-partitioned carbohydrate, protein and lipid proportions in: (b) picoplankton, (c) nanoplankton and (d) microplankton, derived using Eq. (8). (e)-(g) Algorithm-based relative uncertainties in the estimates of: (e) carbohydrate, (f) protein and (g) lipid, quantified as a joint function of the relative uncertainties in  $\xi$ ,  $a_M$  and  $b_M$  (see, Section 2.5). The 95% confidence levels for the allometric parameters reported in Finkel et al. (2016*a*) are considered for computing the % uncertainties in the parameters with respect to their reported means, along with a range of 0-25% relative uncertainty in  $\xi$  (following Roy et al. (2013)).

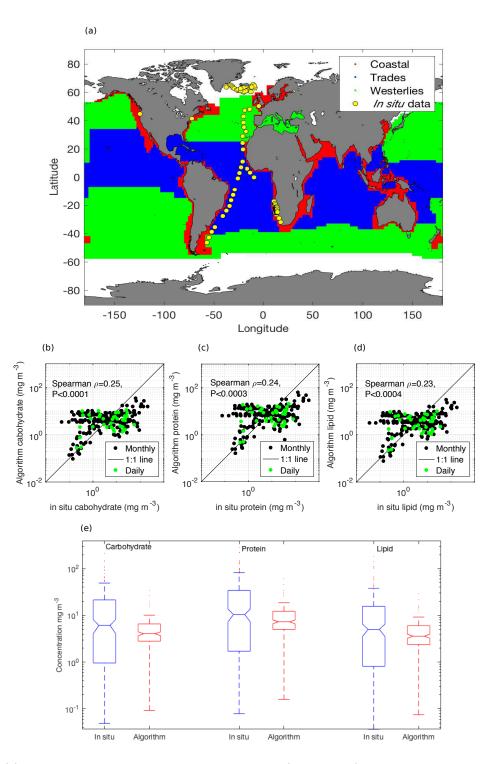


Figure 2: (a) Geographic locations of the *in situ* samples (yellow dots) used from the marine biodiversity database (Sal et al., 2013); this subset overlapped with the temporal coverage of satellite data, and were considered for computing phytoplankton carbohydrate, protein and lipid using species size and cell abundances, and by applying the allometric relationships reported in Finkel et al. (2016a). The Westerlies, Trades and Coastal Longhurst provinces are shown in different colours. (b)-(d) Satellite match-ups from daily (green dots) and monthly (black dots) images were considered for comparing the satellite-derived (b) carbohydrate, (c) protein and (d) lipid with the *in situ* estimates. (e) Box-plots comparing the estimates from *in situ* with satellite based on the current method.

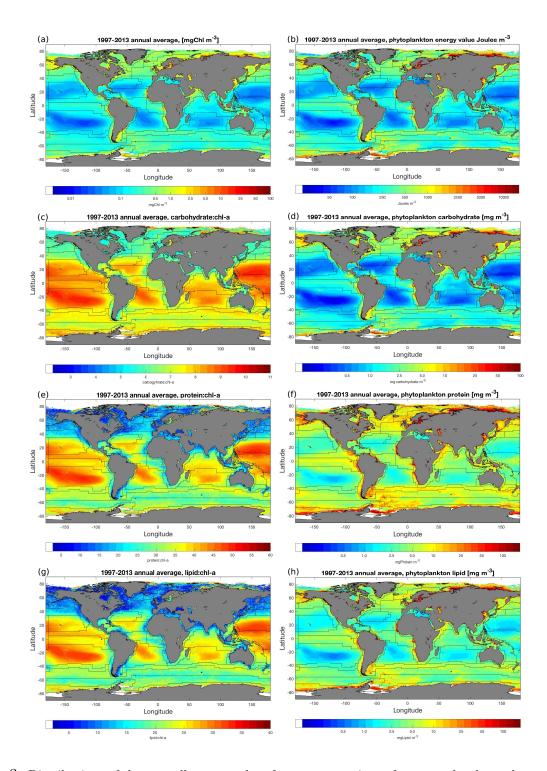


Figure 3: Distributions of the annually-averaged surface concentrations of macromolecules and energy value of phytoplankton over 1997-2013. Overlaid on the global maps are thin black lines representing the boundaries of the Longhurst biogeographical provinces (Longhurst, 1995, 1998). Annual averages of (a) surface chlorophyll in  $[mg m^{-3}]$ ; (b) chemical energy value of phytoplankton in  $[Joules m^{-3}]$  as a combinations of the estimated carbohydrate, protein, lipid; (c) carbohydrate to chlorophyll ratio (dimensionless); (d) concentration of carbohydrate in  $[mg m^{-3}]$ ; (e) protein to chlorophyll ratio (dimensionless); (f) concentration of protein in  $[mg m^{-3}]$ ; (g) lipid to chlorophyll ratio (dimensionless); (h) concentration of lipid in  $[mg m^{-3}]$ , computed based on the methodology described in Section 2.

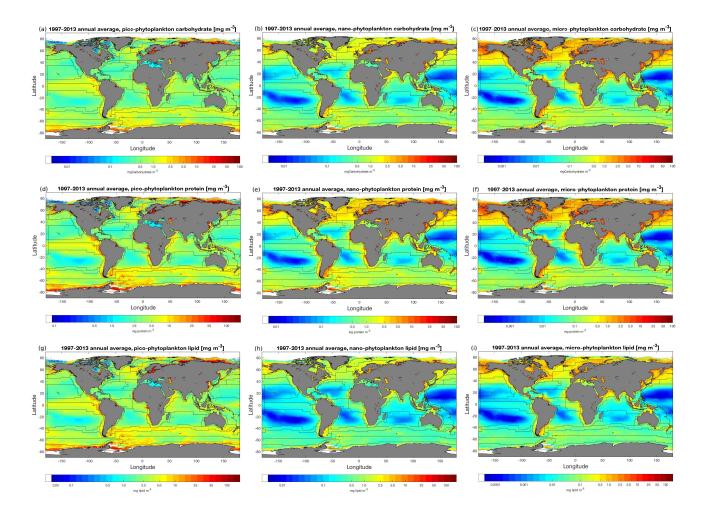


Figure 4: Annually-averaged surface macromolecular concentrations  $[mg m^{-3}]$  in picoplankton, nanoplankton, and microplankton over 1997-2013: (a) picoplankton carbohydrate, (b) nanoplankton carbohydrate, (c) microplankton carbohydrate; (d) picoplankton protein, (e) nanoplankton protein, (f) microplankton protein; and (g) picoplankton lipid, (h) nanoplankton lipid, (i) microplankton lipid, computed based on the methodology described in Section 2. Overlaid on the global maps are thin black lines representing the boundaries of the Longhurst biogeographical provinces (Longhurst, 1995, 1998).

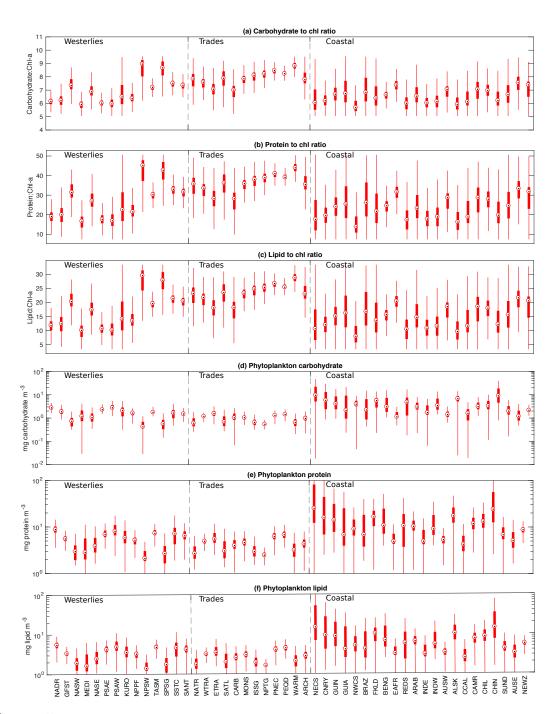


Figure 5: Annually-averaged surface macromolecular composition within Longhurst biogeographical provinces (Longhurst, 1995) computed over 1997-2013. Box plots with annual median (black dots), interquartile ranges (thick red bar), and ranges (thin whiskers) for (a) carbohydrate-to-chlorophyll ratio ( $\chi_{carbo}$ ), (b) protein-to-chlorophyll ratio ( $\chi_{prot}$ ), (c) lipid-to-chlorophyll ratio ( $\chi_{lipid}$ ), (d) carbohydrate (mg m<sup>-3</sup>), (e) protein (mg m<sup>-3</sup>), and (f) lipid (mg m<sup>-3</sup>), are shown for 48 Longhurst provinces. The provinces include 14 Westerlies (NADR, GFST, NASW, MEDI, NASE, PSAE, PSAW, KURO, NPPF, NPSW, TASM, SPSG, SSTC, SANT), 12 Trades (NATR, WTRA, ETRA, SATL, CARB, MONS, ISSG, NPTG, PNEC, PEQD, WARM, ARCH) and 22 Coastal (NECS, CNRY, GUIN, GUIA, NWCS, BRAZ, FKLD, BENG, EAFR, REDS, ARAB, INDE, INDW, AUSW, ALSK, CCAL, CAMR, CHIL, CHIN, SUND, AUSE, NEWZ) provinces. The provinces within Westerlies, Trades and Coastal are arranged from north to south as they appear in the Longhurst's original list. Descriptions of the provinces can be found in Longhurst (1995, 1998), and the full names of the provinces along with the plotted median values of the full averages are given in Table S1.

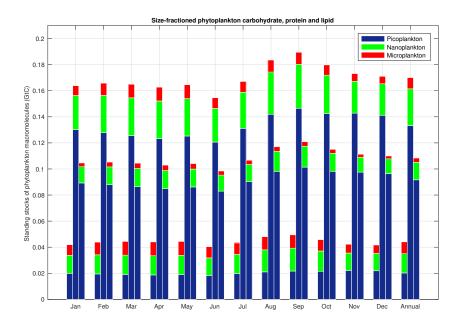


Figure 6: Annually-averaged macromolecular compositions for three phytoplankton size classes over 1997-2013. Grouped bars represent the monthly and annual stocks of the total (height of each bar) and size-partitioned (blue - picoplankton fraction, green - nanoplankton fraction, and red - microplankton fraction) estimates of carbohydrate (first bar in each group), protein (second bar in each group) and lipid (third bar in each group), computed from the surface concentrations through integrations over the mixed-layer depths. All concentrations are expressed in gigatonnes (Gt).

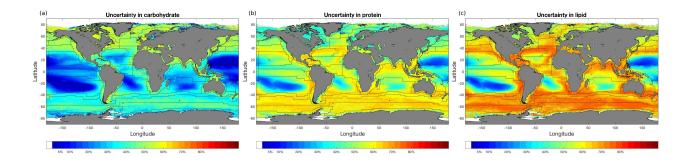


Figure 7: Algorithm uncertainty maps corresponding to the estimates of (a) phytoplankton carbohydrate, (b) phytoplankton protein and (c) phytoplankton lipid based on the sensitivity analysis in Section 2.5. Annually-averaged uncertainties in estimating the surface concentrations of carbohydrate, protein and lipid are shown for an overall relative uncertainty of 25% in  $\xi$  retrievals combined with 95% confidence intervals of the allometric parameters reported by Finkel et al. (2016*a*).

Table 1: Summary of overall uncertainties (mean with ranges) in carbohydrate, protein and lipid estimates as a function of uncertainties in  $\xi$  and allometric parameters, as shown in Fig. (1e,f,g)

	$\xi < 3.25$	$3.25 \le \xi \le 4.5$	$\xi > 4.5$	
	56% (41-62%)	25% (0-52%)	0.3%~(0-7%)	Carbohydrate
$\Delta \xi / \xi \le 15\%$	31%~(15-55%)	29% (0-66%)	7% (0 - 30%)	Protein
	44%~(19-70%)	32%~(0-77%)	5%~(0-30%)	Lipid
$\Delta \xi / \xi > 15\%$	59% (52–62%)	40% (8-60%)	4% (0-18%)	Carbohydrate
	42% (32–83%)	74% (31–108%)	29% (10-55%)	Protein
	58%~(46-105%)	$85\%~(31{-}126\%)$	27%~(7-59%)	Lipid