

# A multi-method autonomous assessment of primary productivity and export efficiency in the springtime North Atlantic

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## Abstract

15 Fixation of organic carbon by phytoplankton is the foundation of nearly all open-ocean ecosystems and a critical part of the global carbon cycle. But quantification and validation of ocean primary productivity at large scale remains a major challenge, due to limited coverage of ship-based measurements and the difficulty of validating diverse measurement techniques. Accurate primary productivity measurements from autonomous platforms would be highly desirable, due to much greater potential coverage. In pursuit of this goal we estimate gross primary productivity over two months in the springtime North Atlantic from  
20 an autonomous Lagrangian float using diel cycles of particulate organic carbon derived from optical beam attenuation. We test method precision and accuracy by comparison against entirely independent estimates from a locally parameterized model based on chlorophyll *a* and light measurements from the same float. During nutrient replete conditions (80% of the study period), we obtain strong relative agreement between the independent methods across an order of magnitude of productivities ( $r^2=0.97$ ), with slight under-estimation by the diel cycles method ( $-19\pm 5\%$ ). At the end of the diatom bloom, this relative  
25 difference increases to  $-58\%$  for a six-day period, likely a response to  $\text{SiO}_4$  limitation, which is not included in the model. In addition, we estimate gross oxygen productivity from  $\text{O}_2$  diel cycles and find strong correlation with diel cycles-based gross primary productivity over the entire deployment, providing further qualitative support to both methods. Finally, simultaneous estimates of net community productivity, carbon export and particle size suggest that bloom growth is halted by a combination of reduced productivity due to  $\text{SiO}_4$  limitation and increased export efficiency due to rapid aggregation. After the diatom  
30 bloom, high chlorophyll *a* normalized productivity indicates that low net growth during this period is due to increased heterotrophic respiration and not nutrient limitation. These findings represent a significant advance in the accuracy and completeness of upper ocean carbon cycle measurements from an autonomous platform.

## 1 Introduction

Measurement of ocean primary productivity (PP), the origin of nearly all organic carbon available to marine organisms, is essential to the study of marine ecosystems and predicting how they might respond to human activities. Because human influences such as climate change and fishing have global impact, improvements in global, mechanistic understanding of both the drivers of PP and its effects on ecosystems and their services should be of great value. However, progress is limited by the difficulty of measuring PP, which traditionally involves incubation experiments and/or radio- or stable isotope analysis, requiring cost, expertise, and ship availability. Understanding is further limited by the difficulty in validating PP, as each method has potential sources of bias, but generally no two methods measure the exact same quantity at the same temporal scale. Therefore, it is often unclear whether discrepancies between independent measurements are caused by biases or real differences. Satellite PP algorithms and global models can achieve the desired coverage, but these products still must be validated, ideally using an in situ dataset of confirmed accuracy that spans many years, in all seasons and in all oceans. Autonomous platforms can achieve such in situ coverage at a fraction of the cost of ship-based sampling, so the ability to estimate PP from an autonomous platform and validate these estimates using independent methods is highly desirable, both for directly enhancing understanding of ocean ecosystems and validating the models and satellite products that can approach true continuous global coverage.

Methods for estimating PP from diel cycles in particulate beam attenuation  $c_p$  (Siegel et al. 1989; Claustre et al. 1999; Cullen et al. 1992; Kinkade et al. 1999; J. Marra 2002; Dall’Olmo et al. 2011; Gernez et al., 2011; Omand et al., 2017; White et al. 2017) or  $O_2$  (Caffrey 2003; Hamme et al. 2012; Nicholson et al. 2015) are suited for application to autonomous platforms, many of which already carry  $O_2$  sensors and/or transmissometers. These methods rely on the light-dependence of PP, which causes a diel cycle in  $O_2$  and in  $c_p$  (due to its correlation with particulate organic carbon, POC). However, other factors such as zooplankton vertical migrations, mixing events,  $O_2$  air-sea flux, and POC/ $c_p$  ratios may have diel cycles that introduce bias in these PP estimates, so they cannot be relied upon without validation. Comparisons so far between diel cycles and independent PP estimates have been encouraging, generally agreeing within a factor of two to three (Cullen et al. 1992; Walsh et al. 1995; Kinkade et al. 1999; Hamme et al. 2012; Nicholson et al. 2015), but the independent estimates have not been of the same quantity at the same temporal scale, so these comparisons do not provide strong constraints on the accuracy of this method.

In this study we take three significant steps towards the goal of enhancing our understanding of ocean ecosystems by increasing coverage of accurate in situ PP estimates using autonomous platforms. First, we use diel cycles in measurements of  $c_p$  and  $O_2$  to simultaneously estimate two related quantities, gross primary productivity of particulate organic carbon (GPP) and gross oxygen productivity (GOP), in the surface mixed layer over a two-month period from an autonomous Lagrangian float. To our knowledge, this is the first time that  $c_p$ -based GPP and GOP have been simultaneously calculated using diel cycles from

any platform, let alone autonomously. Second, we compare our diel cycles-based GPP estimates with entirely independent estimates of the same quantity at the same spatial and temporal scale across a wide dynamic range of productivities. Again, to our knowledge, this represents the most rigorous validation of the diel cycles method to date. Third, we apply our mixed layer PP estimates, in conjunction with mixed layer O<sub>2</sub>, NO<sub>3</sub>, and POC budgets, to better understand how PP, heterotrophic respiration, and sinking flux all interact to regulate mixed-layer biomass in our study system: the spring diatom bloom in the Iceland Basin.

## 2 Methods

### 2.1 Study area and Platforms

The data presented here were collected by an autonomous Lagrangian mixed-layer float, two ships, and three autonomous Seagliders during the North Atlantic Bloom 2008 (NAB08) project. All data used here are available online at <http://www.bco-dmo.org/project/2098>. The float was deployed on April 4 in the Iceland Basin at 59°N, 20.5°W, near the 60°N site of the 1989 Joint Global Ocean Flux Study (JGOFS). The NAB08 project centered around the float, which was designed to drift in the surface mixed layer, mimicking the movement of plankton, except for daily profiles to 250 m (D'Asaro 2003). The float gathered data for two months, drifting northwest towards the Reykjanes Ridge, and ceased collecting data on May 25 at 61.8°N, 26.7°W (Fig. 1; black line) and was recovered on June 3. The timing of the daily float profiles was irregular until April 14, after which the float profiled each day between 15:00 GMT and dusk. The float carried an array of sensors, including two SBE-43-CTs for temperature and salinity, a WET Labs C-Star transmissometer for particulate organic carbon (POC), via particulate beam attenuation  $c_p$ , a WET Labs FLNTU (fluorescence and turbidity meter) for chlorophyll a fluorescence and POC, via particulate optical backscattering  $b_{bp}$ , a Seabird SBE-43 and an Aanderaa optode for oxygen, an ISUS (In Situ Ultraviolet Spectrophotometer) for NO<sub>3</sub>, a LICOR LI-192SA for planar photosynthetically active radiation (PAR). See Table 1 for a list of abbreviations used in more than one subsection. Three cruises provided calibration data for the float's sensors as well as more detailed biological and chemical measurements: a deployment cruise by the R.V. *Bjarni Saemundsson* (April 3-5), a process cruise by the R.V. *Knorr* (May 2-21) and a float "rescue" cruise by the R.V. *Bjarni Saemundsson* (June 4-5). The ships collected both in situ measurements and discrete water samples via an overboard CTD package, which profiled to 600 m. Both ships carried the same array of in situ sensors as the float, minus the ISUS NO<sub>3</sub> sensor and the Aanderaa optode. In addition, the R.V. *Knorr* carried a second CTD and an above-water PAR sensor. Unlike the float, both of the ship's PAR sensors measured scalar PAR. The Seagliders were deployed together with the float and piloted to follow it throughout the experiment. Over the deployment, the distance between the float and individual gliders ranged from 175 km to < 1 km. However, at least one glider was within 50 km of the float for almost the entire deployment, and starting on May 6, all gliders remained within 50 km. Seagliders carried an array of sensors, but here we only discuss Seaglider estimates of sinking flux, derived in Briggs et al. (2011) using spikes caused by large particles in  $b_{bp}$ , measured by a WET Labs BB2F.

## 2.2 Discrete sampling

Discrete samples from all three cruises were analyzed at depths ranging from near surface (3-5 m) to 600 m for particulate organic carbon (POC; n=343), chlorophyll *a* (*Chl*; n=935), SiO<sub>4</sub> and NO<sub>3</sub> (n=1001), and phytoplankton pigments (n=80). Detailed methodology for these analyses can be found in the following technical report: [http://data.bco-dmo.org/NAB08/Laboratory\\_analysis\\_report-NAB08.pdf](http://data.bco-dmo.org/NAB08/Laboratory_analysis_report-NAB08.pdf). Briefly, *Chl* samples were filtered onto GFF 0.7 μm filters and analyzed onboard using a Turner Designs Model 10-AU fluorometer. Following JGOFS protocols, POC samples were filtered onto pre-combusted GFF 0.7 μm filters, sealed in foil packets and stored at -20 °C until analysis onshore using a Perkin Elmer 2400 CHN analyzer. For nutrients, 60 mL samples were immediately frozen and stored at -20°C until analysis onshore using a Lachat Quickchem 8000 Flow Injection Analysis System. In addition, discrete samples on the May process cruise were analyzed for dissolved oxygen concentration via Winkler titrations (n=131) and for bacterial counts and phytoplankton community composition using a FACScan flow cytometer and a FlowCAM automated microscopic imager. Phytoplankton particles were divided into several groups based on optical properties, size, and morphology as described in Cetinić et al. (2012), with more detailed methods in a technical report accompanying the dataset: [http://data.bco-dmo.org/NAB08/Phytoplankton\\_Carbon-NAB08.pdf](http://data.bco-dmo.org/NAB08/Phytoplankton_Carbon-NAB08.pdf).

## 15 2.3 Calibration of in situ sensors

The ship's profiler was held at constant depth for 60 s prior to closing each bottle to capture a water sample. In situ measurements from the 30 s prior to bottle closing were averaged to obtain a single value for matchups with discrete samples. Ship in situ sensors were calibrated via linear regression against discrete measurements. Float in situ sensors were calibrated using data from ten calibration casts, in which the ship was brought to the float's location and both ship and float profiled simultaneously. Float NO<sub>3</sub> and oxygen sensors were calibrated directly against the discrete measurements taken during the calibration casts. All other float sensors were calibrated against the matching ship in situ sensors, in order to maximize the number of matchups. Individual calibration details for each float sensor are listed below.

### 2.3.1 Temperature and Salinity

The duplicate temperature (T) and salinity (S) sensors aboard the ship's profiler during the May process cruise agreed closely (median S difference  $\leq 0.0018$  and a median T difference  $\leq 0.0006$  °C for each of 134 profiles). The ship TS sensors were therefore used as standards, after de-spiking and averaging (more details at [http://data.bco-dmo.org/NAB08/Ship\\_TS\\_despiking-NAB08.pdf](http://data.bco-dmo.org/NAB08/Ship_TS_despiking-NAB08.pdf)). Duplicate T sensors aboard the float also agreed closely and were therefore combined into a single record without adjusting to match the ship. After reconciliation of duplicate S measurements on each platform, a small mismatch between float and ship salinity was identified from the calibration casts and corrected by subtracting 0.0075 from the float S (more details at [http://data.bco-dmo.org/NAB08/CTD\\_float\\_Calibration-NAB08.pdf](http://data.bco-dmo.org/NAB08/CTD_float_Calibration-NAB08.pdf)).

### 2.3.2 Oxygen

Comparison between the SBE-43 and optode oxygen sensors aboard the float revealed differing sources of bias in each sensor. Bias in SBE-43 oxygen was introduced by changes in pumping rate during different modes of float operation and by wave action near the surface. Bias in optode oxygen arose from its factory calibration, T and pressure effects, and a slower time response. After reconciliation of the two sensors to reduce these biases, the SBE-43 oxygen was brought in line with the discrete oxygen samples on the best six calibration casts by subtracting a constant offset of  $0.9 \mu\text{Mol kg}^{-1}$ . We conclude that the accuracy of the corrected in situ oxygen estimates is better than  $2 \mu\text{Mol kg}^{-1}$ , based on agreement with discrete samples (Winkler titrations). More details of the float's oxygen calibration can be found at [http://data.bco-dmo.org/NAB08/Oxygen\\_Calibration-NAB08.pdf](http://data.bco-dmo.org/NAB08/Oxygen_Calibration-NAB08.pdf).

### 10 2.3.3 POC from optical beam attenuation

Previous work has shown that measurements of light scattering by particles, including beam attenuation  $c_p$  and particulate backscattering  $b_{bp}$  correlate strongly with POC in the open ocean (Cetinic et al. 2012 and references therein). Calibration of our  $c_p$  and  $b_{bp}$  measurements and conversion to POC estimates are described in the next two subsections. Raw output from the float optical beam transmissometer was aligned with raw ship transmissometer output using matchups from eight of the calibration casts. Agreement was very good ( $r^2 = 0.99$ ), showing no evidence of sensor drift. Intercalibrated raw transmissometer output was converted to particulate optical beam attenuation  $c_p$  using the mean of factory calibrations performed on the ship's transmissometer before and after deployment. More details can be found at [http://data.bco-dmo.org/NAB08/C-Star\\_Calibration-NAB08.pdf](http://data.bco-dmo.org/NAB08/C-Star_Calibration-NAB08.pdf). We estimated  $c_p$ -derived POC ( $\text{POC}_{c_p}$ ) following Cetinic et al. (2012), but with a time-dependent adjustment in  $\text{POC}/c_p$  ratio to account for community changes. After subtracting the  $\text{POC}/c_p$  regression offset of  $0.015 \text{ m}^{-1}$  (Cetinic et al. 2012) from our  $c_p$  measurements, we computed the  $\text{POC}/c_p$  ratio for all ship POC and  $c_p$  samples where  $c_p > 0.2 \text{ m}^{-1}$  in the upper 30 m during the May process cruise (Fig. 2; gray points). Samples whose T, S,  $c_p$ , and  $b_{bp}$  matched the float ML measurements within  $0.25^\circ\text{C}$ , salinity of 0.01,  $0.1 \text{ m}^{-1}$ , and  $0.001 \text{ m}^{-1}$ , respectively are shown as black circles (Fig. 2). Three inflection points were fit by eye at 370, 310 and  $450 \text{ mg m}^{-2}$  on May 6, 11, and 15, respectively. A continuous estimate of  $\text{POC}/c_p$  at the float patch was obtained by interpolating between these points and assuming constant  $\text{POC}/c_p$  before May 6 and after May 15 (Fig. 2, red line). This continuous estimate of  $\text{POC}/c_p$  was multiplied by float  $c_p$  (minus offset of  $0.015 \text{ m}^{-1}$ ) to yield a  $c_p$ -based float POC estimate ( $\text{POC}_{c_p}$ ).

### 2.3.4 POC from optical backscattering

An average of pre- and post-deployment calibrations were used to convert raw backscattering output from both the float and the ship to the volume scattering function at the angle ( $140^\circ$ ) and wavelength (700 nm) of the sensors. The volume scattering function of seawater was then calculated following Zhang et al. (2009) and subtracted to yield scattering due to particles. The result was multiplied by  $2\pi\chi$  to yield the particulate backscattering coefficient  $b_{bp}$ , where the angle-dependent scale factor  $\chi$  is

1.132 for the FLNTU scattering sensors used in this study (M. Twardowski, pers. comm.). Float  $b_{bp}$  was aligned with ship  $b_{bp}$  using matchups from eight calibration casts ( $r^2 = 0.96$ ). More details can be found at [http://data.bco-dmo.org/NAB08/Backscatter\\_Calibration-NAB08.pdf](http://data.bco-dmo.org/NAB08/Backscatter_Calibration-NAB08.pdf). Glider  $b_{bp}$  was calibrated against the ship FLNTU in a similar fashion to the float (Briggs et al. 2011). We estimated  $b_{bp}$ -derived POC ( $POC_{bbp}$ ) following Cetinic et al. (2012) via the equation  
 5  $POC_{bbp} [mg\ C\ m^{-3}] = 37500\ b_{bp} [m^{-1}] - 14$ , derived from a linear regression between co-located measurements of POC and  $b_{bp}$  within the mixed layer from the May process cruise.

### 2.3.5 Chlorophyll *a*

Raw chlorophyll fluorometer output from the ship was converted to an initial *Chl* estimate  $Chl_{factory}$  using the factory calibrated scale factor and a dark offset derived from the minimum of all per-cast deep values (defined as the median between 550-580  
 10 m). An empirical fit between  $Chl_{factory}$ , T, PAR, and ship discrete *Chl* measurements was used to derive an in-situ *Chl* product (Eq. 1), which was strongly correlated with discrete *Chl* ( $r^2 = 0.87$ ). Float  $Chl_{factory}$  was aligned with ship

$$Chl = Chl_{factory} * \frac{2.1 * 10^{(T-9.2)*0.8+0.44}}{10^{(T-9.2)*0.8+1}} * \frac{(\log_{10}(PAR)*0.05+1.02)*\tanh\left(\frac{PAR}{95}\right)+0.55}{0.55*\frac{PAR}{95}}, \quad (1)$$

$Chl_{factory}$  using the matchups from eight calibration casts ( $r^2 = 0.95$ ), allowing calculation of *Chl* via Eq. (1), at the float as well. The FLNTU sensor was located at the bottom of the float, facing down, so *Chl* data was removed whenever the float was  
 15 moving upward at  $>1.7\ cm\ s^{-1}$ , due to possible entrainment of deeper water. *Chl* measurements where  $PAR > 75$  were also removed to eliminate non-photochemical quenching. In order to obtain a continuous, depth-resolved record of *Chl* for calculation of primary productivity, the remaining *Chl* estimates, from both mixed-layer mode and profiles, were filtered using a 5-point running median, averaged in one-hour, one-meter bins, and then interpolated in depth and time via triangulation-based 2-D linear interpolation, with distance calculated as  $\sqrt{(dz[m]/30[m])^2 + dt[days]^2}$  (i.e. a 30 m vertical interval and  
 20 a 1 day time interval were considered equidistant).

### 2.3.6 Nitrate

A post-deployment laboratory calibration, including temperature and salinity corrections, was used to obtain initial  $NO_3$  estimates from the float's ISUS  $NO_3$  sensor. An additional scale factor of 1.15 and offset of  $+2.6\ \mu M$  were required to bring these initial estimates in line with discrete samples taken during calibration casts. More details can be found in Alkire et al.  
 25 (2012) and at [http://data.bco-dmo.org/NAB08/ISUS\\_Nitrate\\_Calibration-NAB08.pdf](http://data.bco-dmo.org/NAB08/ISUS_Nitrate_Calibration-NAB08.pdf).

### 2.3.7 Silicate

$SiO_4$  was not measured by the float, but discrete shipboard  $SiO_4$  measurements from the top 15 m were considered to represent mixed-layer  $SiO_4$  at the float location if the corresponding temperature, salinity and  $NO_3$  measurements matched concurrent float ML measurements to within  $0.25^\circ C$ , salinity of 0.01, and  $0.8\ mmol\ m^{-3}$  respectively.

### 2.3.8 PAR

The factory calibration of the float PAR sensor was used “as is.”

### 2.4 Mixed layer depth

Mixed layer depth (MLD) was calculated at hourly intervals from float potential density anomaly estimates via the following steps: 1) Smooth density timeseries using a 5-point running median. 2) Average density into one-hour, one-meter bins. 3) Fill in the gaps with 2D linear interpolation, such that a 30 m vertical interval and a one-day time interval are considered equidistant. 4) For each hour, find the minimum potential density anomaly. 5) The MLD for each hour is defined as the shallowest depth where the potential density anomaly exceeds this minimum by  $\geq 0.01 \text{ kg m}^{-3}$ . The MLD was calculated twice, once excluding data when downward velocity exceeded  $1 \text{ m min}^{-1}$  and once excluding upward velocity exceeding  $1 \text{ m min}^{-1}$ . We use the average of these two estimates as the final MLD estimate, reducing the influence of single active profiles, which could differ from mean conditions due to entrainment or internal waves. When the float was close to neutral buoyancy, this MLD(t) estimate followed the lower limit of the vertical movement of the float during its ML mode. However, during periods of positive buoyancy, MLD(t) occasionally exceeded the maximum depth of the float during its ML mode (Fig. 3). Hourly MLD depth was also calculated using the same criteria from the output of the Bagniewski et al. (2011) data-assimilation model (see red line in Fig. 3) to permit testing of diel cycles methods within the model itself.

### 2.5 $K_{PAR}$

#### 2.5.1 Instantaneous $K_{PAR}$ estimates

The diffuse attenuation coefficient of PAR  $K_{PAR}$  was calculated from each pair of consecutive PAR measurements made at times  $t_1$  and  $t_2$  via Eq. (2), where  $z$  is depth,  $\bar{z}$  is the mean of  $z(t_2)$  and  $z(t_1)$ , and  $\bar{t}$  is the mean of  $t_2$  and  $t_1$ .

$$K_{PAR(measured)}(\bar{z}, \bar{t}) = \frac{\ln(PAR(t_1)) - \ln(PAR(t_2))}{z(t_2) - z(t_1)} \quad (2)$$

#### 2.5.2 $K_{PAR}$ fit method

The uncertainty of individual  $K_{PAR(measured)}$  estimates was high and depended strongly on  $dz$ , which ranged from 0.2 – 30 m with a mean of 1.3 m. These 14000  $K_{PAR(measured)}$  estimates were therefore fit to  $Chl$  and  $z$  using a non-linear least-squares multiple regression weighted by  $dz$  to obtain Eq. (3):

$$K_{PAR(modeled)}(Chl, z) = 0.064 * Chl^{0.51} + 0.20 * \max(z, 2.5)^{-0.63} + 0.0031. \quad (3)$$

In order to evaluate the performance of this fit,  $K_{PAR(measured)}$  precision was increased by eliminating estimates with  $dz < 2$  m and combining the remaining estimates into 21-point medians, yielding a total of 118 independent in-situ  $K_{PAR}$  estimates. A type-II linear regression of these estimates against 21-point medians of  $K_{PAR}$  estimated via Eq. (3) yielded an  $r^2$  of 0.85, a root mean

square error of  $0.014 \text{ m}^{-1}$ , and a mean bias of  $-0.004 \text{ m}^{-1}$ . The residual error was not significantly correlated with depth, time, solar zenith angle or the ratio of in situ *Chl* to  $b_{bp}$ , a proxy for plankton community in this system (Cetinić et al. 2015).

## 2.6 Depth-resolved PAR

In order to calculate PAR at all depths, PAR was extrapolated from a reference depth  $z_{ref}$  via Eq. (4):

$$5 \quad PAR_{extrapolated}(z) = PAR(z_{ref}) * \exp\left(\int_z^{z_{ref}} K_{PAR} dz\right), \quad (4)$$

using  $K_{PAR(modelled)}$  calculated via Eq. (3) from the float's continuous *Chl*. When the float was within the top 50 m,  $z_{ref}$  was the depth of the float and  $PAR(z_{ref})$  was the float's PAR measurement. The performance of this extrapolation was evaluated by comparing  $PAR_{extrapolated}(0-)$  (just below the surface) calculated via Eq. (4) with scalar  $PAR(0+)$  measured by the ship's underway system. For all measurements where the ship was within 1 km of the float, the float was in the top 50 m, and  $PAR(0+)$  was greater than  $1 \mu\text{mol m}^{-2} \text{ s}^{-1}$ ,  $PAR(0+)$  and  $PAR_{extrapolated}(0-)$  were highly correlated ( $r^2 = 0.96$  on a linear scale and  $r^2 = 0.99$  on a logarithmic scale). The geometric mean of the ratio of  $PAR_{extrapolated}(0-)$  to  $PAR(0+)$  was 0.92 and the geometric (multiplicative) standard deviation was a factor of 1.19. For several hours each afternoon, while the float profiled to 250 m, float PAR measurements were not available, so  $PAR(0-)$  was estimated using an empirical function of solar zenith angle and an empirical index of cloud cover. First, a double exponential was fit to 36000 PAR measurements obtained in the top 1 m over a range of solar zenith angle  $\theta$  from  $-6^\circ$  to  $90^\circ$  by a global network of 100 "Biogeochemical Argo" type profiling floats to obtain  $PAR_{modelled}(0-)$ , an estimate of  $PAR(0-)$  under mean cloud and atmospheric conditions:

$$15 \quad \log_{10}(PAR_{modeled}(\theta)) = 2.5 * \exp(0.0030 * \theta) - 1.7 * \exp(-0.10 * \theta), \quad (5)$$

To adjust for clouds,  $PAR_{extrapolated}(0-)$  from the Lagrangian float (via Eq. 4) was divided by corresponding estimates  $PAR_{modelled}(0-)$  from to obtain an index of sunniness, which was averaged into 15 min bins to remove noise from wave focusing. This sunniness index ranged from 0.1 to 3.6 over the entire float deployment. Sunniness index at time  $t$  was estimated using a  $\pm 1$ -day running mean of these sunniness index estimates, weighted by the inverse square of  $t-t_i$ , where  $t_i$  is the time of each measurement. This running mean sunniness index was then multiplied by  $PAR_{modelled}(0-)$  to obtain  $PAR_{adjusted}(0-)$ , which was used as  $PAR(z_{ref})$  in Eq. (4) during the afternoon gaps.

## 2.7 O<sub>2</sub> air-sea flux

25 O<sub>2</sub> air-sea flux was calculated following Alkire et al. (2012). Briefly, wind speeds were taken from the NCEP WW3 Global Reanalysis product, except during the May cruise, when ship wind measurements were used. O<sub>2</sub> saturation was calculated following García and Gordon (1992). Air-sea flux was calculated following (Wanninkhof 1992), modified to account for bubble injection following Woolf and Thorpe (1991). Hourly  $dO_2/dt$  in the ML due to air-sea flux was estimated by dividing hourly flux estimates by hourly MLD.



## 2.8 Primary Productivity estimates

### 2.8.1 Diel cycles of O<sub>2</sub> and POC

“Typical” diel cycles (minimum near dawn and maximum near dusk) were observed in mixed layer records of O<sub>2</sub> (Fig. 4), consistent with previous studies (Caffrey 2003; Hamme et al. 2012; Nicholson et al. 2015). We estimated mixed layer gross oxygen productivity (GOP) at half-day intervals from these diel cycles. To estimate morning GOP, ML O<sub>2</sub> concentrations were smoothed with a 3-point running median and a type I linear regression (O<sub>2</sub> vs time) was fit to data from dusk to dawn (Fig. 4a; solid black line). The regression fit was projected forward to provide an estimate of noontime mixed-layer O<sub>2</sub> in the absence of GOP. Measured noontime O<sub>2</sub> was calculated from a type I linear regression of O<sub>2</sub> data taken within 1 h of local noon. Morning mixed layer GOP was calculated as the difference between measured and projected concentration (Fig. 4; blue vertical bar) and divided by 0.5 d to convert to units of mmol m<sup>-3</sup> d<sup>-1</sup>. Afternoon GOP was calculated in a similar fashion, by subtracting noontime mixed-layer O<sub>2</sub> from the noontime extrapolation of a linear fit of the following night’s data. Similar diel cycles were observed in mixed layer POC<sub>cp</sub>, and the same method was used to calculate mixed layer gross primary productivity of POC (GPP<sub>cp</sub>) from these cycles (Fig. 4b). Diel cycles in POC<sub>bbp</sub> were less regular and usually out of phase with O<sub>2</sub> and POC<sub>cp</sub> cycles, but GPP<sub>bbp</sub> was calculated in the same way as GOP and GPP<sub>cp</sub> for comparison. Note that this diel cycles method assumes homogeneous mixing to a constant depth and that any gain or loss terms other than GOP (or GPP) are constant day-to-night over the period of a single calculation (~18 h). However, we find a clear diel cycle in MLD (Fig. 3), which amplifies the diel cycle in O<sub>2</sub> (and *c<sub>p</sub>* and *b<sub>bp</sub>*), causing PP calculated from diel cycles to exceed mean PP within the daily mean MLD. This is because night-time ML deepening enhances loss of ML concentration relative to daytime mixing losses. Analysis of the output of a coupled physical-biological model assimilating data from the Lagrangian float (Bagniewski et al. 2011), which accurately reproduced the diel cycle in mixing (Fig. 3, black line) shows that the mixing-amplified diel cycles of O<sub>2</sub> in the ML yield daily GOP estimates that correspond approximately to the mean GOP above the daily *minimum* MLD. Regression of diel GOP, calculated from ML O<sub>2</sub> timeseries output by the model, as a function of “true” model GOP, forced through zero, yields a slope ±95 % confidence interval of 0.91±0.12 and a RMSE of 0.12 mmol m<sup>-3</sup> d<sup>-1</sup>. We therefore interpret our daily GOP and GPP<sub>cp</sub> estimates as representing daily mean productivity between the surface and daily minimum MLD. Bias in GOP due to day-night differences in air-sea flux was also estimated using the difference between mean morning (or afternoon) dO<sub>2</sub>/dt due to air-sea flux and that of the previous (or next) nighttime. Mean bias was small (<5% of GOP), and linked primarily to the MLD diel cycle, so a separate correction was not deemed necessary. Other potential biases are discussed in sections 4.1.2 to 4.1.4.

### 2.8.2 <sup>14</sup>C incubations

During the April and May cruises, daily two-hour <sup>14</sup>C incubation experiments were conducted (n=28) to estimate photosynthetic parameters. Each day, a water sample was taken from the *Chl* maximum, as determined by in situ fluorescence, and duplicate 2 h <sup>14</sup>C incubations were carried out at 7 different PAR levels ranging from 0-400 μmol m<sup>-2</sup> s<sup>-1</sup>. The dark incubation <sup>14</sup>C activities were weakly but significantly correlated with *Chl* (type II linear regression; r<sup>2</sup> = 0.19; p<0.05; apparent

NPP =  $Chl * 0.036 \pm 0.026 \text{ mg C mg Chl}^{-1} \text{ h}^{-1} + 0.049 \pm 0.035 \text{ mg C m}^{-3} \text{ h}^{-1}$ ); dark activities were treated as sample-specific blanks and subtracted from the light incubation activities of the corresponding water sample. The resulting productivity estimates were interpreted as net primary productivity (NPP), based on findings that most phytoplankton do not respire old carbon when newly fixed carbon is available (John Marra and Barber 2004; Pei and Laws 2013). Note that if, contrary to our assumptions, phytoplankton did respire old carbon at all light levels during these incubations, then our calculations below overestimate NPP and underestimate phytoplankton respiration ( $R_\phi$ ), but GPP is unbiased. On the other hand, if old carbon is respired only in the low light incubations, then we under-estimate  $R_\phi$  and GPP, but little bias is introduced in NPP. Seven of the 350 individual NPP estimates (all from the April cruise) were judged to be positive outliers and were manually removed before further analysis. In ~60% of the incubation experiments, NPP decreased with increasing PAR for PAR > 200  $\mu\text{mol quanta m}^{-2}\text{s}^{-1}$ . We conclude that this apparent photoinhibition is likely not representative of most field conditions, because in situ measurements of  $Chl$ -normalized  $dO_2/dt$  showed no consistent relationship with PAR between PAR values of 100 and 1000  $\mu\text{mol m}^{-2}\text{s}^{-1}$ . We therefore removed values of NPP where PAR > 200  $\mu\text{mol m}^{-2}\text{s}^{-1}$  if they were lower than the second-highest NPP observed where PAR < 200  $\mu\text{mol m}^{-2}\text{s}^{-1}$  (56 of 110 high light points removed). Remaining NPP vs PAR data were fit to an empirical “PvE” model represented in Eqs. (6-8):

$$\lambda = PAR \frac{\alpha}{P_M} \varepsilon, \quad (6)$$

$$GPP = P_M \left( 1 + \frac{1}{\varepsilon} \sum_{i=0}^{\varepsilon-1} e^{-\lambda \frac{\lambda^i}{i!}} i - \sum_{i=0}^{\varepsilon-1} e^{-\lambda \frac{\lambda^i}{i!}} \right), \quad (7)$$

$$NPP = GPP - R_\phi, \quad (8)$$

based on four parameters: maximum GPP ( $P_m$ ), the initial slope of GPP/PAR ( $\alpha$ ),  $R_\phi$ , and an efficiency factor ( $\varepsilon$ ) representing “sharpness” of the transition between light-limited and light-saturated photosynthesis. This parameterization is based on a conceptual model of photosynthesis in which there is a rate-limiting step that can receive and “store” up to  $\varepsilon$  “packets” of energy at once at above the limiting rate without wasting any of these packets. We used a single  $\varepsilon$  value of six, which provided the best overall least squared fit across all incubation experiments. This  $\varepsilon$  value yields a NPP vs PAR relationship that is “sharper” than the commonly used “tanh” model (Harrison and Platt 1986) and more linear at low PAR, leading to smaller y-offset (smaller  $R_\phi$  estimate). See Fig. 5 for example fits. A power law was then fit between in situ  $Chl$  estimates from the ship’s profiling package (calculated via Eq. 1) and each of the three parameters obtained from each NPP vs PAR fit ( $P_M$ : Fig. 6a;  $\alpha$ : Fig. 6b; and  $R_\phi$ : Fig. 6c). Fits with  $P_M$  and  $\alpha$  used data from all cruises, but the fit with  $R_\phi$  included only data from the process cruise (Fig. 6c; solid circles), as signals were too low to constrain  $R_\phi$  in April and  $R_\phi$  appeared consistently higher during the June cruise, possibly due to higher temperature.

### 2.8.3 Chl-based GPP and NPP

The relationships in Fig. 6 were used to estimate photosynthetic parameters  $P_M(t,z)$ ,  $\alpha(t,z)$ , and  $R_\phi(t,z)$  and their uncertainty intervals at the float location from  $Chl(t,z)$  (Section 2.3.5). We estimated gross primary productivity  $GPP_{Chl}(t,z)$  and net primary productivity  $NPP_{Chl}(t,z)$  via Eqs. (6-8) using the above photosynthetic parameters,  $PAR_{extrapolated}(t,z)$  (Section 2.6), and

$\varepsilon=6$  as input. Uncertainties were propagated from  $P_M(t,z)$ ,  $\alpha(t,z)$ , and  $R_\phi(t,z)$  using the conservative assumption that they covary (i.e. upper bound of  $NPP_{Chl}$  was derived from upper bounds of  $P_M$  and  $\alpha$  and lower bound of  $R_\phi$ ).

## 2.9 Area-weighted mean particle diameter

Area-weighted mean particle diameter  $D_{bbp}$  10-50 m depth bin was estimated following Briggs et al. (2013) via Eqs. (9-11):

$$5 \quad D_{bbp} = 2 \sqrt{\frac{\text{Var}[b_{bp}(t)]}{E[b_{bp}(t)]} \frac{V}{Q_{bb}} \frac{1}{\gamma(\tau)} \frac{1}{\pi}}, \quad (9)$$

$$\gamma(\tau) = \begin{cases} 1 - (3\tau)^{-1}, & \text{if } \tau \geq 1 \\ \tau - \tau^2/3, & \text{if } \tau \leq 1 \end{cases} \quad (10)$$

$$\tau = \left( \frac{t_{res}}{t_{samp}} \right), \quad (11)$$

where  $\text{Var}[b_{bp}(t)]$  is the variance in  $b_{bp}$  due to random distribution of particles in space,  $E[b_{bp}(t)]$  is mean  $b_{bp}$ ,  $V$  is sensor sample volume,  $Q_{bb}$  is the backscattering efficiency, and  $\gamma$  and  $\tau$  are functions of residence time in the sample volume  $t_{res}$  and sample integration time  $t_{samp}$ .  $\text{Var}[b_{bp}(t)]$  and  $E[b_{bp}(t)]$  were calculated once per profile (ascent or descent) using all data between 10-50 m. Prior to calculation of  $\text{Var}[b_{bp}(t)]$ , the  $b_{bp}$  timeseries was de-trended by subtracting a 7-point running median and large outliers (greater than five times the interquartile range) were removed before the variance was calculated on the residuals. A  $V$  of 0.62 ml was used (Briggs et al. 2013) and a  $Q_{bb}$  of 0.02 was assumed (based on empirical  $b_{bp}/c_p$  ratio of  $\sim 0.01$  and theoretical value of  $Q_c=2$  for diameter  $\gg$  wavelength; Bohren and Huffman, 1983). A  $t_{res}$  of 0.02 s was chosen based on a 6 mm path through the sample volume and a platform velocity of 30 cm s<sup>-1</sup> and  $t_{samp}$  was 1 s.

## 2.10 Sinking POC flux

POC<sub>bbp</sub> profiles from both gliders and the float were divided into a “small” particle baseline (7-point running minimum followed by running maximum) and a “large” particle “spike” signal (residuals above the baseline). This approach, developed by Briggs et al. (2011), is based on the finding that large, fast sinking particles, owing to their rarity and light-scattering characteristics, can create individual large spikes in mesopelagic  $b_{bp}$  clearly distinguishable from background concentrations (Briggs et al. 2011). Large particle POC<sub>bbp</sub> was multiplied by a bulk sinking speed of 75 m d<sup>-1</sup> to estimate large POC flux (Briggs et al. 2011). A broad plausible range of bulk sinking speeds  $5 \pm 5$  m d<sup>-1</sup> was used to estimate small POC sinking flux, which was added to large POC flux to yield total sinking POC flux. Sinking POC flux was bin averaged in 50 m vertical bins and either running 2-day bins (gliders) or longer discrete bins to match bloom stages (float).

### 3 Results

#### 3.1 Evolution of the spring bloom

From float deployment through April 17, MLD was variable (often >200 m but occasionally < 50 m; Fig. 3), mixed-layer nutrients were high ( $\text{NO}_3 \approx 12 \text{ mmol m}^{-3}$ ;  $\text{SiO}_4 \approx 4 \text{ mmol m}^{-3}$ ), biomass was low ( $\text{Chl} \approx 0.35 \text{ mg m}^{-3}$ ;  $\text{POC}_{\text{cp}} \approx 35 \text{ mg m}^{-3}$ ), and  $\text{O}_2$  was undersaturated by  $\sim 10 \text{ mmol m}^{-3}$  (Fig. 7). Mixed-layer biomass concentrations increased over the next month, peaking in mid-May. This broad increase was punctuated by several 1-2 day periods of decrease, most associated with clear mixed-layer deepening (Fig. 7).  $\text{SiO}_4$  was depleted to its lowest level on May 11,  $\text{Chl}$  concentration peaked on May 12, and  $\text{NO}_3$  depletion and  $\text{POC}_{\text{cp}}$  and  $\text{O}_2$  concentrations peaked on May 13. From bloom peak to May 16,  $\text{Chl}$  decreased dramatically (77%),  $\text{POC}_{\text{cp}}$  and  $\text{O}_2$  decreased moderately (by 9 and 13  $\text{mmol m}^{-3}$ , respectively), and  $\text{NO}_3$  and  $\text{SiO}_4$  concentrations recovered slightly (by 0.8 and 0.4  $\text{mmol m}^{-3}$ , respectively).

#### 3.2 Primary productivity estimates

All GPP and GOP estimates were averaged into 3-day bins to improve precision of the diel-cycles-based estimates. To first order,  $\text{GPP}_{\text{Chl}}$  followed  $\text{Chl}$ , low in early April (0.5-1.0  $\text{mmol m}^{-3} \text{ d}^{-1}$ ), peaking near 10  $\text{mmol m}^{-3} \text{ d}^{-1}$  between May 7 and 13, then decreasing to near 3  $\text{mmol m}^{-3} \text{ d}^{-1}$  or below after the bloom (Fig. 8). But increases in  $\text{GPP}_{\text{Chl}}$  led increases in  $\text{Chl}$  by 1-2 days during ML shoaling (and high growth) events on April 24-27 and May 6-8 (Fig. 8; pale vs. dark green) due to higher ML averaged PAR (not shown). For the entire “bloom growth” phase from early April through May 9,  $\text{GPP}_{\text{Chl}}$  was strongly correlated with both cycle-based estimates of both GOP (Fig. 8 and Fig. 9b; blue) and  $\text{GPP}_{\text{cp}}$  (Fig. 8 and Fig. 9c; blue). GOP was a factor of 2.1 higher than  $\text{GPP}_{\text{Chl}}$  on a molar basis, while  $\text{GPP}_{\text{cp}}$  was slightly lower (factor of 0.81).  $\text{GPP}_{\text{bbp}}$  was poorly correlated with  $\text{GPP}_{\text{Chl}}$  and significantly lower (by 60%; Fig. 9d). From noon May 10 to noon May 11, diel cycles could not be calculated, because the float was trapped at the surface, due to high stratification and slight positive buoyancy. At peak biomass (May 11-13), and the bloom decline (May 13-16), both  $\text{GOP}/\text{GPP}_{\text{Chl}}$  and  $\text{GPP}_{\text{cp}}/\text{GPP}_{\text{Chl}}$  were substantially lower than during bloom growth (Fig. 8, pink highlighted region, and Fig. 9; pink symbols). In the post-bloom period (May 16-24),  $\text{GOP}/\text{GPP}_{\text{Chl}}$  and  $\text{GPP}_{\text{cp}}/\text{GPP}_{\text{Chl}}$  increased again, similar to the bloom growth ratios (Fig. 8 and Fig. 9; red symbols). When all bloom phases are combined, best-fit ratios of GOP and  $\text{GPP}_{\text{cp}}$  to  $\text{GPP}_{\text{Chl}}$  are 1.7 and 0.6, respectively and correlations are considerably less strong ( $r^2$  of 0.67 and 0.49, respectively). However, the estimates of productivity from diel cycles (GOP and  $\text{GPP}_{\text{cp}}$ ) remained strongly correlated for the entire deployment. Over the entire study period, morning estimates of GOP and  $\text{GPP}_{\text{cp}}$  were not significantly different from the afternoon estimates, while morning  $\text{GPP}_{\text{bbp}}$  estimates were significantly lower than afternoon estimates (80% lower overall). However, morning-afternoon patterns appear to change starting on May 13, when the bloom decline starts (e.g. Fig. 4). From May 13-24, there is no significant difference between morning and afternoon  $\text{GPP}_{\text{bbp}}$ , but afternoon estimates of  $\text{GPP}_{\text{cp}}$  and GOP, were lower than morning estimates by 70% and 43%, respectively. These differences were near the threshold of statistical significance: mean afternoon-morning difference  $\pm 2$  standard errors was  $-2.3 \pm 2.3 \text{ mmol m}^{-3} \text{ d}^{-1}$  for  $\text{GPP}_{\text{cp}}$  and  $-3.0 \pm 2.5 \text{ mmol m}^{-3} \text{ d}^{-1}$  for GOP.

### 3.3 Depth integrated GPP, NPP, and NCP and carbon export

Alkire et al. (2012) estimated net community productivity (NCP) integrated within the top 50-60 m and carbon export from 50-60 m at the float location for four periods of stable stratification: the “early bloom” (April 23-27), “main bloom” (May 6-13), “decline” (May 13-14) and “post bloom” (May 20-24). We integrated  $GPP_{Chl}$  and  $NPP_{Chl}$  to the same depth and time ranges in order to assemble detailed organic carbon budgets for these periods (Fig. 10). Each budget term carries considerable uncertainty, but based on the central estimates, the partitioning of fixed carbon appeared to change substantially over the course of the bloom. Note that these NCP estimates include net production of dissolved organic carbon (DOC), while  $NPP_{Chl}$  excludes any photosynthetic DOC production.  $NPP_{Chl}$  and NCP estimates were similar during the early and main bloom, suggesting moderate-to low heterotrophic respiration. During the early bloom period, export was also low (~22-28% of  $GPP_{Chl}$ ), allowing rapid accumulation of biomass. During the main bloom,  $GPP_{Chl}$  nearly doubled as biomass increased, but a larger fraction (~50%) was exported, leaving ~25% to accumulate. During the bloom decline, apparent community respiration (defined as difference between  $GPP_{Chl}$  and NCP) was 156% of  $GPP_{Chl}$  and export was an additional 50-80%. In the post-bloom period, community respiration was again high (~100% of GPP), and export was much lower (0-15% of GPP). Our  $NPP_{Chl}$  estimates and  $b_{bp}$  “spike”-based sinking flux estimates provide a continuous, high-resolution picture of the link between productivity and export at 125 m for the entire study period (Fig. 11a). Float and glider-based POC export estimates agree broadly at this depth (red lines), suggesting that the higher-resolution glider timeseries are representative of the float patch as well. While export at 125 m is coupled with  $NPP_{Chl}$  (Fig. 11a), there is a rapid increase in export efficiency between May 3-6 from ~20% to 40%. Area-weighted mean particle diameter ( $D_{bfp}$ ) ranged from 90-150  $\mu m$  during April, peaked at 250  $\mu m$  on May 7-8 (Fig. 11b), coincident with peak biomass as measured by both  $Chl$  and  $POC_{bfp}$  from the gliders (not shown).  $D_{bfp}$  fell rapidly on May 9, coincident with a ML deepening event. Post-bloom  $D_{bfp}$  ranged from 150-190  $\mu m$  (Fig. 11b).

## 4 Discussion

### 4.1 Accuracy of PP estimates

The combination of three estimates of primary productivity and one estimate of community productivity, all from the same platform at comparable temporal and horizontal scales, provides a unique opportunity to evaluate the accuracy of all methods. Each of our PP methods is discussed in turn in sections 4.1.1-4.1.4.

#### 4.1.1 $GPP_{Chl}$

$GPP_{Chl}$  and  $GPP_{cp}$  are estimates of the same quantity, obtained independently.  $GPP_{Chl}$  is derived from PAR and  $Chl$  estimates using robust local parameterizations obtained from  $^{14}C$  incubations.  $GPP_{cp}$  is derived entirely from  $c_p$  measurements, converted to POC using another robust, local empirical relationship. The averaging depth (daily minimum MLD) for  $GPP_{Chl}$  was chosen to match the diel cycles method based on results of a model tuned to match local conditions (Bagniewski et al. 2011). In this

context, the combination of strong correlation and absolute agreement between  $GPP_{Chl}$  and  $GPP_{cp}$  (Fig. 9c; within 19%) provides confidence in both methods during the bloom growth and post bloom periods. The  $GOP/GPP_{Chl}$  slope of 2.1 (Fig. 9b) is at the upper end of the expected range, providing additional first-order support for  $GPP_{Chl}$  accuracy. Neither GPP method includes DOC production, so the range of expected photosynthetic quotients (~1-1.45; Laws 1991; Robertson et al. 1993), combined with the fraction of GPP released as DOC in marine/estuarine environments (2-50%; Baines and Pace, 1991) imply a possible  $GOP/GPP$  range of 1-2.9. During the main bloom observed by the float in this study, Alkire et al. (2012) estimate that DOC accounts for 22-40% of NCP in the mixed layer during the main bloom. If these estimates apply to GPP as well, our expected  $GOP/GPP$  range narrows to 1.3-2.4 (Fig. 9a,b; gray dashed lines). Thus, our  $GOP$  estimates suggest either that both the photosynthetic quotient and phytoplankton DOC production are high during bloom growth (and GPP is accurate) or that both GPP estimates are biased low. As mentioned in section 2.8.2, a negative bias in  $GPP_{model}$  could be explained if phytoplankton respired substantial old, unlabelled carbon in our low light incubations, but not in the high light incubations. In this case a separate explanation (see next section) is needed for the high  $GOP/GPP_{cp}$  slope of 2.6 (Fig. 9a).

During bloom peak and decline, the strong discrepancies between  $GPP_{cp}$  and  $GPP_{Chl}$  imply either an underestimate by  $GPP_{cp}$  (discussed in the next section) or an over-estimate by  $GPP_{Chl}$ . If diatoms reduce GPP in response to sustained  $SiO_4$  limitation then we expect  $GPP_{Chl}$  over-estimation at peak biomass, given that  $GPP_{Chl}$  is only a function of *Chl* and PAR, without any nutrient limitation term. Twelve mixed-layer  $SiO_4$  samples were collected in the vicinity of the float on May 11-13, and the mean and maximum measured concentrations were 0.3, and 0.6  $mmol\ m^{-3}$ , respectively, suggesting that diatom growth was most likely severely limited (Fig. 7a). This does not necessarily imply that diatom carbon fixation rates were reduced, but previous studies have indeed observed a large and reversible reduction in apparent diatom photosynthetic efficiency under multi-day  $SiO_4$  limitation (Lippemeier, Hartig, and Colijn 1999; Lippemeier et al. 2001). Both FlowCAM microscopy and HPLC pigments indicate that diatoms accounted for  $\geq 50\%$  of phytoplankton biomass at bloom peak (Cetinić et al. 2015), so we expect a substantial reduction in bulk phytoplankton growth (and likely GPP) under these conditions. This expectation, combined with the observed reduction in both  $GPP_{cp}$  and  $GOP$  at bloom peak, leads us to conclude that  $GPP_{Chl}$  is most likely over-estimated at bloom peak. This conclusion agrees with the coupled physical-biological model of Bagniewski et al. (2011), which assimilated float biogeochemical measurements and achieved optimal fit when diatom GPP was limited by  $SiO_4$  with a half-saturation constant of 1  $\mu mol\ m^{-3}$ . GPP inferred from this model closely matches our observed  $GPP_{cp}$  during  $SiO_4$  limitation (Fig. 8, gray line vs black circles), even though Bagniewski et al. (2011) assimilated daily binned data, removing any diel cycle information. On the other hand, three  $^{14}C$  incubations were conducted between May 10-14 using water with  $SiO_4 < 0.5\ mmol\ m^{-3}$ , although not at the float location, and there was not a substantial reduction in measured  $P_M$ . These samples may not be representative of the water sampled by the float, despite similar *Chl* and  $SiO_4$  concentrations, or it is possible that a bottle effect enhanced GPP. But we cannot rule out the alternative hypothesis that the Si-limited community continued to fix carbon at a constant rate, and that  $GPP_{cp}$  and  $GOP$  estimates were reduced for another reason (discussed in next sections).

Apart from Si limitation, possible explanations for  $GPP_{Chl}$  over-estimation include over-estimation of *Chl* due to increased fluorescence, underestimation of the MLD, or photoinhibition. Again, *Chl* was calculated the same way for the floats and ship, so if the in-situ fluorometric method over-estimated *Chl* at bloom peak, we would expect to see a deviation from the observed relationships between photosynthetic parameters and *Chl* (Fig. 6). So this explanation, while plausible given the high  $Chl/b_{bp}$  ratio at bloom peak (Cetinić et al. 2015), also requires that none of our low  $SiO_4$  bottle samples were representative of the bloom peak at the float location. The density-based MLD estimates appear quite robust during this period, consistently shallow and stable at 10-20 m and matched by the vertical motion of the float. And the daytime increases in  $O_2$  (Fig. 4) and  $c_p$  on May 11 and 12 show no sign of photoinhibition, despite a peak hourly-averaged PAR of  $>750 \mu mol m^{-2} s^{-1}$ ; increases are smooth throughout the morning and appear to continue at the same rate in the afternoon (Fig. 4). This result supports our decision to exclude photoinhibited bottle incubation data from our productivity vs. PAR fits, and we suggest that photoinhibition terms from bottle incubations should be applied with caution, if at all, in future studies. In this system, photoinhibition is likely reduced during deeper mixing events due to the shorter time phytoplankton is exposed to high light. During the stratified conditions, reduction in photoinhibition can be attributed to photoadaptation.

$SiO_4$  limitation may also explain some of the discrepancy between  $GPP_{cp}$  and  $GPP_{Chl}$  during the bloom decline (May 13-16), at least when afternoon estimates are excluded (see Fig. 9b,c; open pink circles). Lower afternoon  $GPP_{cp}$  and GOP, combined with very shallow ( $<5$  m) MLDs at noon on May 13 and 15, also raise the possibility of significant photodamage inhibiting afternoon productivity. Mean ML PAR exceeded  $500 \mu mol m^{-2} s^{-1}$  for at least two hours on both days. However, the negative afternoon  $GPP_{cp}$  estimates at this time suggest a bias in the diel cycles method as well (see next section).

#### 4.1.2 $GPP_{cp}$

Potential sources of bias unique to  $GPP_{cp}$  include a diel cycle in grazing (e.g. due to diel migration of zooplankton), a diel cycle in export loss, or a diel cycle in the POC/ $c_p$  ratio. The tight correlations between  $GPP_{cp}$  and GOP throughout the entire study period ( $r^2=0.95$ ; Fig. 9a) and between  $GPP_{cp}$  and  $GPP_{Chl}$  during bloom growth ( $r^2=0.96$ ; Fig. 9b) provide encouraging support for  $GPP_{cp}$  as a measure of *relative* primary productivity at the very least. Furthermore, the quantitative agreement between  $GPP_{cp}$  and  $GPP_{Chl}$  during both bloom growth and post-bloom (Fig. 9c; slope:  $0.82 \pm 0.06$ ) is very close to our expected slope of 0.93 from model results (reanalysis of Bagniewski et al. 2011), suggesting that  $GPP_{cp}$  accuracy is comparable to other methods across most of the conditions encountered. These findings agree closely with those of White et al. (2017), who find a  $GPP_{cp}/NPP$  ratio of 1.1 across a factor of three dynamic range of productivities in the subtropical North Pacific, suggesting that  $GPP_{cp}$  is either accurate, or slightly under-estimates true GPP. Taken together, our results are highly encouraging regarding the widespread applicability and accuracy of the  $GPP_{cp}$  method. However, it should be noted that our results still may not apply to certain other systems, where different phytoplankton size and/or timing of cell division could alter the diel POC/ $c_p$  relationship (Dall'Olmo, et al. 2011).

If the  $\text{SiO}_4$  limitation hypothesis is correct, then  $\text{GPP}_{\text{cp}}$  during the bloom peak and morning  $\text{GPP}_{\text{cp}}$  during the bloom decline may be accurate as well. On the other hand, if  $\text{GPP}_{\text{chl}}$  is accurate during this time, then  $\text{GPP}_{\text{cp}}$  is biased low by  $\sim 50\%$  at this time. It is unclear what might cause such a low bias, especially at bloom peak. Between the afternoon of May 11 and the morning of May 13, there is no anomaly in the diel cycles of  $\text{POC}_{\text{cp}}$  or  $\text{O}_2$  (Fig. 10) indicative of daytime mixing, advection, or possible photoinhibition, and there is no change in the relationship between  $\text{GPP}_{\text{cp}}$  and  $\text{GOP}$  (Fig. 9a; rightmost pink symbol). Without grazing data, we cannot rule out enhanced daytime grazing as a possible explanation, although grazing is generally expected to be higher at night. Alternatively, particularly high photo-oxidation could potentially dampen  $\text{O}_2$  diel cycles during this period and perhaps  $c_p$  diel cycles as well. This hypothesis is supported by laboratory measurements of diatom productivity under nutrient limitation (Spilling et al. 2015), although again we would need to explain why reduced  $P_M$  was not observed in our bottle incubations. On the other hand, the afternoon  $\text{GPP}_{\text{cp}}$  estimates during the bloom decline period show a clear example of negative bias in the diel cycles method. On May 13, 14, and 15 (bloom decline), rates of net  $\text{POC}_{\text{cp}}$  (and  $\text{O}_2$ ) accumulation are positive or near zero in the morning but negative each afternoon (e.g. Fig. 4; May 13). One plausible explanation is horizontal advection of the float relative to the ML during its afternoon profile, causing it to resurface in water with lower biomass. During this period, comparison with ship, autonomous glider, and satellite measurements (Alkire et al. 2012) shows that the float was at the edge of a high biomass (and  $\text{O}_2$ ) patch, so advection during this time would most likely cause loss in  $\text{POC}_{\text{cp}}$  and  $\text{O}_2$ . Note that the afternoon reductions in  $\text{GPP}_{\text{cp}}$  during bloom decline are greater than the afternoon reductions in  $\text{GOP}$  (e.g. Fig. 4). This result is possible with the horizontal advection/mixing hypothesis alone, but high export, combined with a shallow afternoon MLD may also play a role. Shallow MLD enhances the loss of ML concentration for a given export rate, and night-time mixing can re-entrain some of this export, reducing the ML  $\text{POC}$  diel cycle relative to the  $\text{O}_2$  diel cycle.

### 4.1.3 GOP

The tight fit between  $\text{GOP}$  and both  $\text{GPP}$  estimates over most of the study period provides important support for the  $\text{O}_2$  diel cycles method as a measure of relative primary production in this region. Again, because all estimates were independent and taken at the same scale, and because the two-month deployment allowed 14 independent matchups at 3-day timescale, spanning a wide range of productivities, this dataset represents the most extensive validation to date of the  $\text{O}_2$  diel cycles method as a measure of *relative* primary productivity. Additionally, the overall *accuracy* of our  $\text{GOP}$  estimates may be assessed indirectly through comparison with independent ship-based  $\text{GOP}$  estimates made during the May process cruise (Quay et al., 2012) and through comparison of our  $\text{GOP}/\text{GPP}$  and  $\text{GOP}/\text{NPP}$  ratio estimates with previous estimates from this region. Quay et al. (2012) estimated ML-integrated  $\text{GOP}$  using measurements of three oxygen isotopes:  $^{16}\text{O}$ ,  $^{17}\text{O}$ , and  $^{18}\text{O}$ , taken daily between May 3 and 21 during the process cruise. Mean  $\text{GOP}$  calculated by this method was  $245 \text{ mmol O}_2 \text{ m}^{-2} \text{ d}^{-1}$ . This method integrates over several weeks, so we interpret their estimate to correspond roughly to mean ML depth-integrated  $\text{GOP}$  between April 19 (2 weeks before the first sample) and May 21. For comparison, we multiply each half-day  $\text{GOP}$  estimate by MLD to obtain ML-integrated  $\text{GOP}$  and obtain an average from April 19 to May 21 of  $149 \text{ mmol m}^{-2} \text{ d}^{-1}$ , 40% lower than Quay et al. (2012)'s



estimate. However, our estimate integrates to the daily minimum MLD, and while the triple O<sub>2</sub> isotope method assumes constant MLD, we expect it to more closely approximate daily maximum MLD in the presence of diel MLD fluctuations, given its long integration time. Mean GPP<sub>chl</sub> during this period, integrated to the bottom of the daily *minimum* MLD is 30% lower than mean GPP<sub>chl</sub> integrated to the daily *maximum* MLD. If we assume the same relative difference for GOP, we obtain a revised ML-integrated GOP estimate of 213 mmol m<sup>-2</sup> d<sup>-1</sup>, 13% lower than Quay et al. (2012)'s estimate. Given the uncertainties associated with the GOP methods as well as the differing spatio-temporal scales, this result provides first-order support for the accuracy of both methods. Our findings reinforce those of Hamme et al. (2012), who, in the Southern Ocean in March/April, found that mean ML integrated GOP calculated via O<sub>2</sub>/Ar diel cycles (similar to our method) was 18% lower than GOP calculated via triple oxygen isotope method (similar to Quay et al., 2012).

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Bender et al. (1992) calculated a GOP/NPP ratio of 2.5 during the spring bloom in the Northeast Atlantic, using in situ <sup>18</sup>O incubations and 24h <sup>14</sup>C incubations. We calculate GOP/NPP<sub>chl</sub> as shown in Fig. 9b, but replace GPP<sub>chl</sub> with NPP<sub>chl</sub> and obtain a best-fit ratio and 95% confidence interval of 2.4±0.2 for the bloom growth period and 1.7±0.4 for the entire deployment. These fits appear to support the accuracy of both our NPP<sub>chl</sub> and GOP estimates during the bloom growth phase, consistent with our other findings. However, our GOP/GPP ratio estimates of 2.6±0.2 (Fig. 9a) and 2.1±0.2 (Fig. 9b) are near or above the high end of our expected range of 1.3-2.4 (see section 4.1.1). As discussed in previous sections, these ratios may be the result of high photosynthetic quotient and high DOC production, combined with a small negative bias in GPP<sub>cp</sub>. Our GOP estimates may also be too high, but we cannot think of a plausible mechanism that would cause a substantial over-estimate of diel-based GOP (but not of GPP<sub>cp</sub>). Regardless of the source of our high GOP/GPP ratios, they are also consistent with Hamme et al. (2012), who also estimated GPP from on-deck PvE incubations as well as GOP via O<sub>2</sub>/Ar diel cycles, providing a very close methodological comparison in a different environment (autumn, Southern Ocean). They obtain an even higher GOP/GPP ratio of 3.6. However, Hamme et al. (2012) assumed that 1-2h <sup>14</sup>C incubations represent GPP, while we assume that these same incubations represent NPP (when NPP>0). If our assumption is correct, then their method provides a quantity closer to daytime NPP than GPP. However, even in this case, assuming moderate daytime phytoplankton respiration rates (≤30% of GPP), GOP/GPP during their study was >2.5, in agreement with our estimates. It is also worth noting that related studies comparing diel cycles in O<sub>2</sub> and pCO<sub>2</sub> measurements (Johnson 2010; Merlivat, et al., 2015), both of which include the effects of DOC production, have found ratios of daytime oxygen production to carbon production that are within the expected range of 1-1.45. These results provide further support for diel cycles-based O<sub>2</sub> production and the hypothesis that DOC production may drive the high GOP/GPP observed in this and other studies (Bender et al. 1992; Hamme et al. 2012).

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In total, the available evidence provides first-order support for the accuracy of our diel cycles-based GOP estimates. Our findings build on important recent work in diverse environments showing that diel cycles in O<sub>2</sub>/Ar ratio yield ML GOP estimates that are consistent with independent GOP estimates (Hamme et al. 2012), and that diel cycles in O<sub>2</sub> measurements from autonomous gliders in the subtropical Pacific provide GOP estimates that are a reasonable multiple of independent NPP

results (Nicholson et al. 2015). Our results add a third ocean region (springtime North Atlantic) and a third platform (Lagrangian mixed-layer float), in addition to new comparisons with  $c_p$  and  $b_{bp}$  diel cycles.

#### 4.1.4 GPP<sub>bbp</sub>

Because diurnal variability in  $b_{bp}$  can be estimated from geostationary satellites (Neukermans et al. 2012), the ability to accurately estimate GPP from  $b_{bp}$  diel cycles would be extremely valuable. While ship-based measurements from NAB08 show that  $b_{bp}$  and  $c_p$  were equally well correlated with POC over the May cruise (Cetinić et al. 2012), the poor matchups we find between GPP<sub>chl</sub> and GPP<sub>bbp</sub> suggest that diel changes are present in POC/ $b_{bp}$  and can cause strong, consistent bias in GPP<sub>bbp</sub>. Our results agree with previous findings that while beam attenuation and forward scattering by phytoplankton increase immediately after they begin to photosynthesize,  $b_{bp}$  and side scattering often do not, both in the lab (Ackleson et al. 1993; Poulin et al., 2018) and in the ocean (Kheireddine and Antoine 2014). These results caution against the use of  $b_{bp}$  diel cycles to estimate GPP without further research. However, it is worth noting that our afternoon GPP<sub>bbp</sub> estimates are reasonably well correlated with GPP<sub>chl</sub> ( $r^2=0.63$ ,  $m = 0.75\pm 0.23$ ; data not shown) during the bloom growth period. If this result is found to be robust in other times and places, then a useful estimate of GPP from satellite (and other)  $b_{bp}$  timeseries may be possible. However, even if the  $b_{bp}$  diel cycle cannot be used to estimate GPP, it likely contains other useful information, especially in combination with  $c_p$  and/or O<sub>2</sub>. If robust relationships between plankton community and/or physiology and  $b_{bp}$  diel cycles can be established (and, ideally, understood mechanistically), then measurements of  $b_{bp}$  diel cycles may still provide valuable oceanographic information, whether from in situ platforms or satellite.

#### 4.2 Combined upper layer carbon budgets

Taken together with Alkire et al.'s (2012) NCP and carbon export estimates and our adaptation of Briggs et al.'s (2011) depth-resolved carbon fluxes, our productivity and bulk particle estimates provide a remarkably detailed, high-resolution picture of carbon flows over the entire spring bloom. From April 4-17, ML *Chl*, POC, and O<sub>2</sub> concentrations changed little, despite large fluctuations in MLD, while NO<sub>3</sub> increased slightly during deep-mixing, presumably due to entrainment, but was stable during shallow (<100 m) mixing. Consistent positive 125 m integrated NPP<sub>chl</sub> (Fig. 11a) was therefore likely balanced by heterotrophic respiration. From April 18 to May 7, ML shoaling events coincided with several pulses of high net growth in POC<sub>cp</sub> and *Chl* (Fig. 7), and the close match between NPP<sub>chl</sub> and NCP during these periods (Fig. 10) suggests minimal role of grazing in regulating this growth. From May 6-7, all four gliders observed a rapid, aggregation event (Fig. 11b) that triggered a dramatic pulse in carbon export, both from the float patch and the broader (~30 km) glider survey area (Fig. 11a; blue and red lines). This pulse sank through the mesopelagic at ~75 m d<sup>-1</sup> and was composed primarily of fragile aggregates containing live phytoplankton including *Chaetoceros sp.* resting spores (Martin et al. 2011; Briggs et al. 2011; Rynearson et al. 2013). This aggregate export was the largest loss term of surface POC during the “main bloom”, reducing biomass accumulation rate by ≥ 50% (Fig. 10). While SiO<sub>4</sub> limitation has been proposed as a cause of this rapid sinking event (Bagniewski et al. 2011), this aggregation commenced when SiO<sub>4</sub> concentrations were still >2 mmol m<sup>-3</sup> (Fig. 7a) and five days prior to the ~35%

reduction in GOP and  $GPP_{cp}$  that we attribute to  $SiO_4$  limitation (pink band in Fig. 11b). The exact cause of this rapid aggregation event is unknown, but likely involves a combination of moderately high particle concentration ( $POC_{cp} > 10 \text{ mmol m}^{-3}$ ), weakening of mixing (which could break fragile aggregates), and production of transparent exopolymer particles (P. Martin et al. 2011; Alkire et al. 2012). The combination of high export and reduced productivity at the end of the diatom bloom (May 12-14) appears to end the ML biomass accumulation. However, we conclude that the subsequent, sharp decline in ML Chl,  $POC_{cp}$ , and  $O_2$  from May 14-15 (Fig. 7) was probably not the result of a dramatic increase in heterotrophic respiration, as implied by the strong negative NCP estimate (Fig. 10) of Alkire et al. (2012). Our conclusion stems from the night-time ML  $O_2$  loss rates, which do not increase at all between the bloom peak the bloom decline (see Fig. 4a). Instead, the ML  $O_2$  decline appears to be caused by further GPP decreases, due to continued  $SiO_4$  limitation and a decline in *Chl* (Fig. 7b,d), likely enhanced by export of phytoplankton from the shallow ML. The  $O_2$  decline (and accelerating *Chl* decline) may have been enhanced by advection of the float relative to the thin surface ML during afternoon profiles (see section 4.1.2), or perhaps an additional, light-dependent process, such as photoinhibition or photorespiration (Spilling et al. 2015), nearly eliminated GPP during this time, but only in the afternoons. After the decline of the diatom bloom, the different productivity estimates again provide a consistent picture, this time of top-down control. GOP and  $GPP_{cp}$  again show no sign of nutrient limitation (Fig. 9b,c, red symbols), and  $NPP_{chl}$  is apparently balanced by heterotrophic respiration. Glider estimates of sinking POC export were low, but higher than early bloom export, despite similar NPP (Fig. 11a) and higher respiration. This result highlights the de-coupling between NCP and export on weekly-monthly timescales in this dynamic system and suggests that biomass and particle size are better predictors of sub-seasonal export dynamics. The changing export efficiencies that we observed ( $< 15\%$  through most of April, to  $\sim 57\%$  during the main bloom to  $\sim 33\%$  in the post-bloom period), provide a complex picture of “the spring bloom”, but still agree broadly with the export ratio of 45% calculated by Buesseler and Boyd (2009) in the North Atlantic spring bloom using JGOFS data, among the highest export efficiencies observed in the open ocean. However, unlike Buesseler and Boyd (2009), and in line with the conclusions of Martin et al. (1993), we see significant flux attenuation in the 100 m below the euphotic zone. For example, 35-48% of flux is lost between 60 m (Fig. 10) and 125 m (Fig. 11a) during the main bloom).

## 25 5 Conclusions

Our results, placed in the context of previous studies, provide strong support for the diel cycles method as a means to obtain estimates of GOP (from  $O_2$ ) and GPP (from  $c_p$ ) with reasonable accuracy relative to existing methods and enough precision on 3-day timescales to clearly resolve a spring diatom bloom. The range of biomass, mixing regimes and phytoplankton communities in this study, combined with previous results from the subtropics, suggest that these methods are not overly dependent on particular ocean conditions. Because the diel cycles method is well suited for autonomous platforms, it has the potential to greatly increase our coverage of in-situ productivity estimates, providing both direct knowledge of this critical biological rate and greatly enhanced validation datasets for satellite-derived and modelled productivity. Our results also support

the use of short-term  $^{14}\text{C}$  incubations to parameterize simple PvE models for application to autonomous measurements, at least in the absence of strong nutrient limitation. We find high GOP/GPP ratios of 2.1-2.6 through most of the study, suggesting high DOC production and/or a possible moderate under-estimation of GPP by both methods. Finally, combined high-resolution estimates of NPP, particle size and sinking flux during the North Atlantic spring bloom show strong coupling between the three, modulated by a dramatic increase in export efficiency at bloom peak, apparently due to rapid aggregation.

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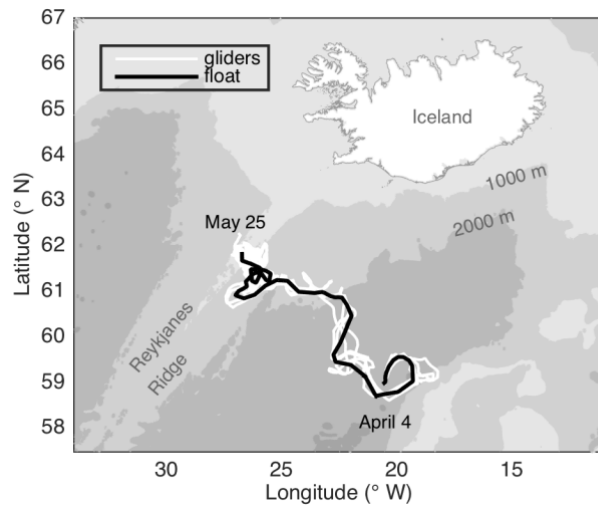
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Table 1. Abbreviations used in more than one subsection of the text

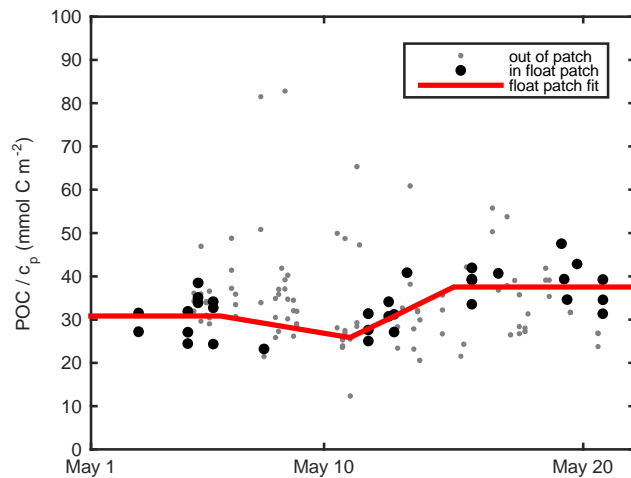
Abbreviation	Description
$b_{bp}$	particulate optical backscattering coefficient
$Chl$	Chlorophyll $a$ concentration
$c_p$	particulate optical beam attenuation coefficient
$D_{bbp}$	area-weighted mean particle diameter from optical backscattering
DOC	dissolved organic carbon concentration
GOP	Gross O <sub>2</sub> productivity from O <sub>2</sub> diel cycles
GPP	Gross primary productivity
GPP <sub>bbp</sub>	GPP from optical backscattering diel cycles
GPP <sub>chl</sub>	GPP from in situ chlorophyll and light measurements
GPP <sub>cp</sub>	GPP from optical beam attenuation diel cycles
JGOFS	Joint Global Ocean Flux Study
$K_{PAR}$	Diffuse attenuation coefficient of PAR
ML	Mixed layer
MLD	Mixed layer depth
NPP	Net primary productivity
NPP <sub>chl</sub>	NPP from in situ chlorophyll and light measurements
PAR	photosynthetically available radiation
$P_m$	Maximum GPP (light saturated)
POC	Particulate organic carbon concentration
POC <sub>bbp</sub>	POC from optical backscattering
POC <sub>cp</sub>	POC from optical beam attenuation
PP	Primary productivity
$R_\phi$	Phytoplankton respiration rate
$S$	Salinity
$T$	Temperature
$\alpha$	Initial slope GPP/PAR (light limited)
$\varepsilon$	coefficient representing "sharpness" of NPP vs PAR relationship



**Fig. 1. Study area with tracks of autonomous Lagrangian mixed-layer float and autonomous Seagliders.**

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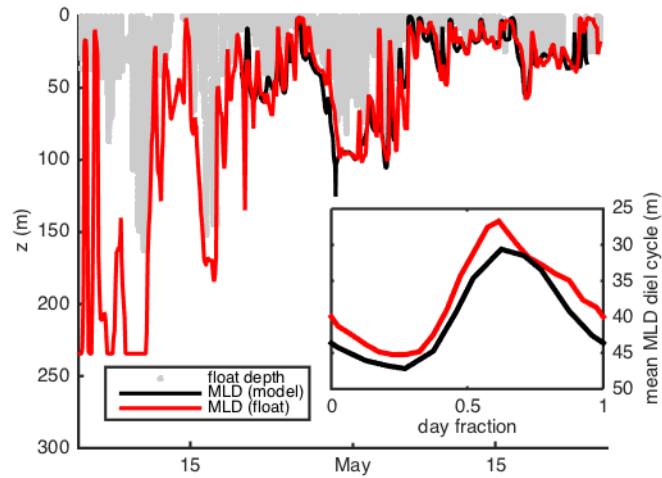
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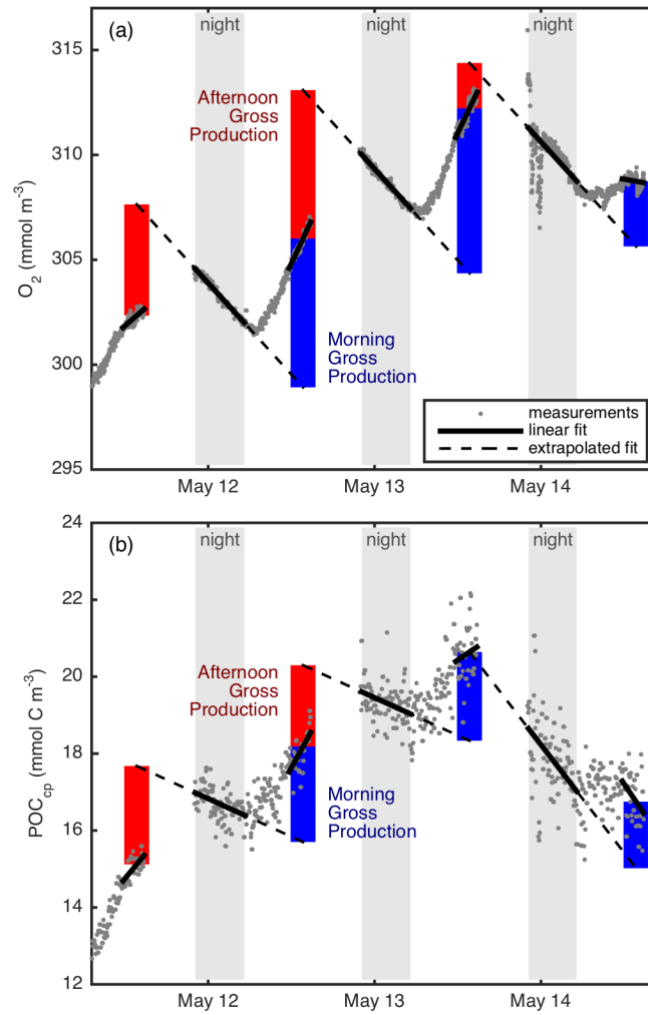
**Fig. 2.**  $POC/c_p$  from the May cruise in upper 30 m, and fit used to calculate  $POC_{cp}$ .

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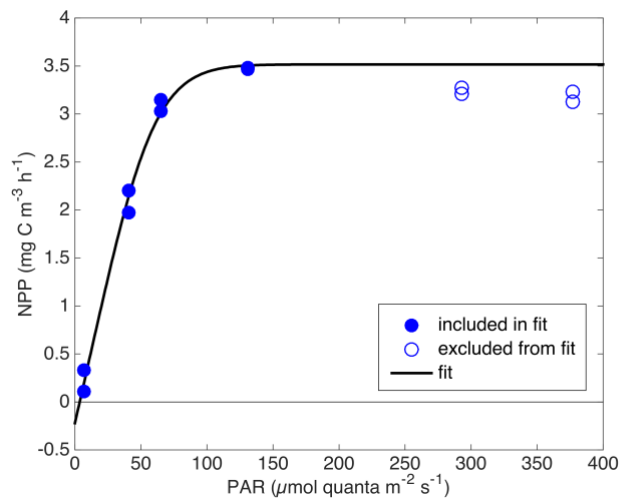
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5 **Fig. 3** Hourly mixed-layer depth estimates calculated directly from float density measurements and from the Bagniewski et al. (2011) data assimilation model (black line), along with the depth of the float in mixed-layer mode (red line). Inset shows mean MLD diel cycle over the entire duration of the model (April 21-May 24). All MLD estimates use a density threshold of  $0.01 \text{ kg m}^{-3}$  to better approximate active mixing on an hourly timescale.



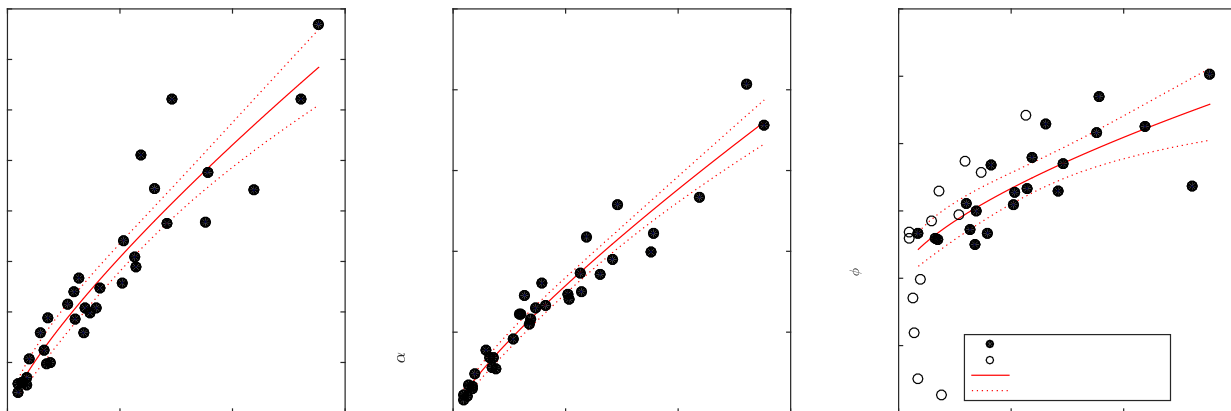
**Fig. 4.** Calculation of gross production of O<sub>2</sub> (a) and POC<sub>cp</sub> (b) in the ML from their diel cycles.



**Fig. 5. Example NPP vs PAR relationship from <sup>14</sup>C incubations, with best fit “PvE” curve.**

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**Fig. 6.** Photosynthetic parameters  $P_M$  (a),  $\alpha$  (b), and  $R_\phi$  (c) vs in situ  $Chl$  with least squares power law fits and 95% confidence intervals.  $R_\phi$  estimates from April and June cruises are excluded from fit (panel c, open circles).

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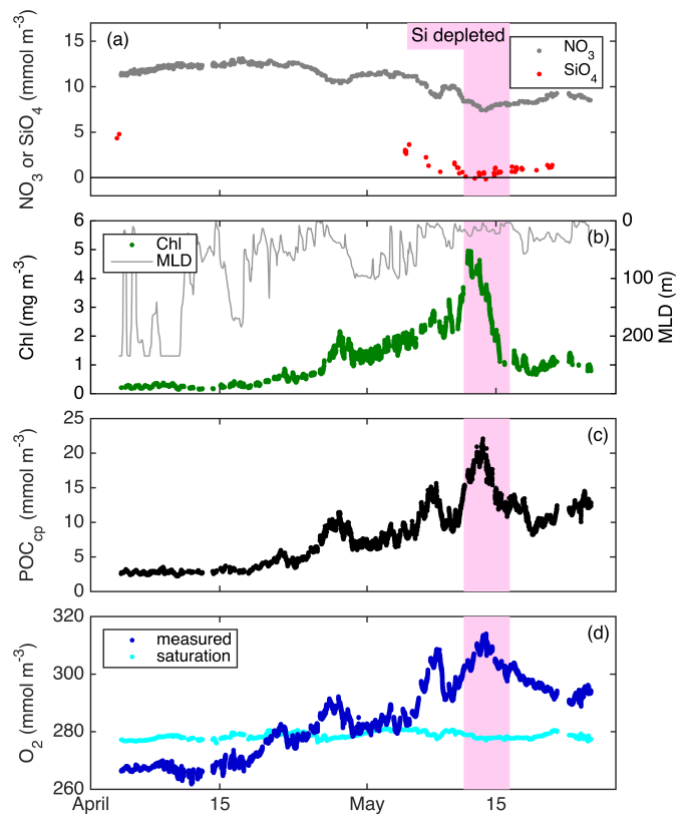
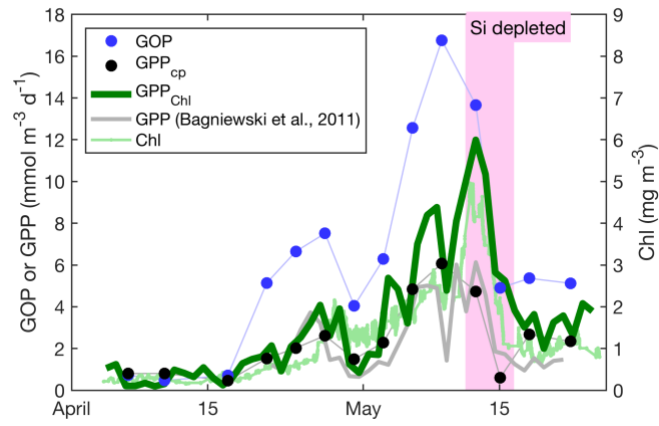
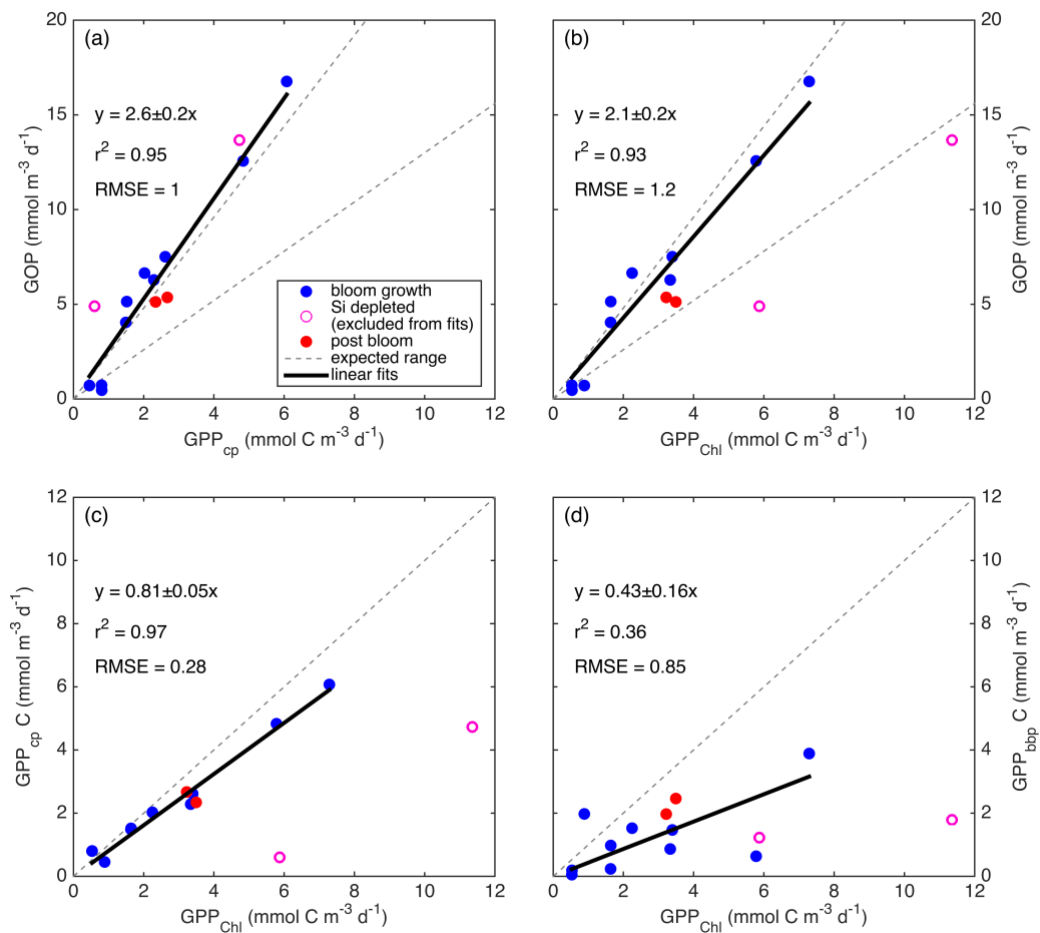


Fig. 7. Float patch mixed layer timeseries of  $\text{NO}_3^-$  (a),  $\text{SiO}_4$  (a), MLD (b),  $\text{Chl}$  (b),  $\text{POC}_{\text{cp}}$  (c),  $\text{O}_2$  (d), and the concentration of  $\text{O}_2$  saturation (d).

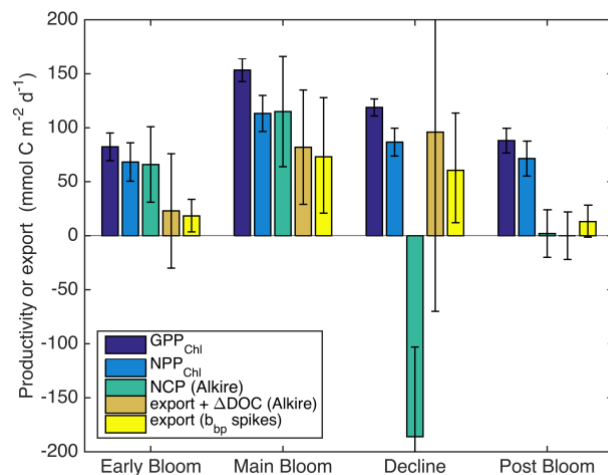




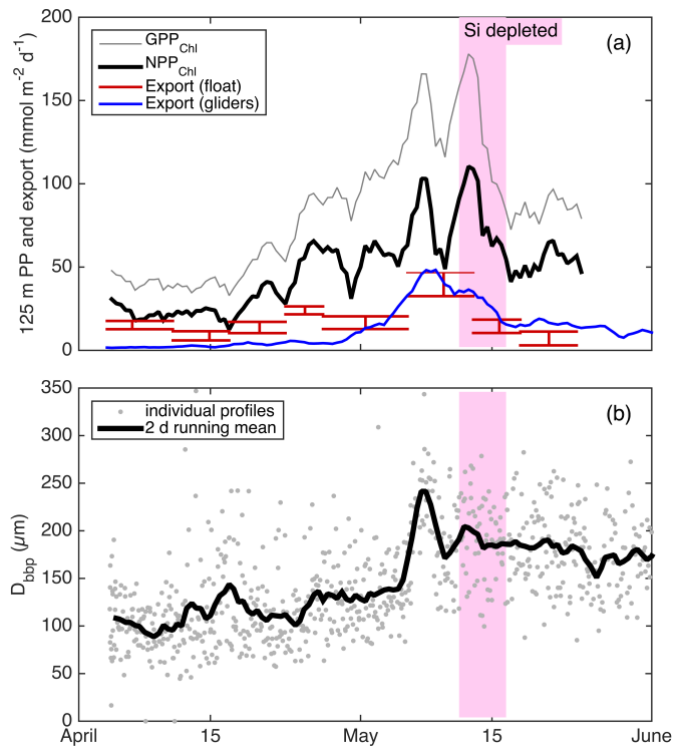
**Fig. 8** Primary productivity estimates within the daily minimum ML.  $GPP_{chl}$ ,  $GOP$ ,  $GPP_{cp}$  and  $GPP$  from Bagniewski et al. (2011), along with ML Chl. Diel cycles-based estimates are 3-day means; other productivity estimates are daily and Chl is continuous.



**Fig. 9. Relationships between primary productivity estimates: GOP vs  $GPP_{cp}$  (a), GOP vs  $GPP_{ChI}$  (b),  $GPP_{cp}$  vs  $GPP_{ChI}$  (c), and  $GPP_{bbp}$  vs  $GPP_{ChI}$  (d). Type I linear regressions are forced through the origin and include all data except the  $SiO_4$ -depleted period (pink circles). Expected range of GOP/GPP (a,b; dashed lines) assumes a photosynthetic quotient between 1-1.45 and 22-40% of fixed carbon released as DOC (see text).**



**Fig. 10.** Estimates of sources and sinks of organic carbon integrated over the top 60 m:  $GPP_{Chl}$  and  $NPP_{Chl}$  and sinking particle export (this study), as well as NCP and loss due to the sum of sinking particle export and net DOC production and sinking particle export only (Alkire et al. 2012). Bloom periods follow Alkire et al. (2012) and are defined in the text (Section 3.3).



**Fig. 11. (a) Continuous productivity and export from the autonomous float and gliders, to/from the top 125 m over the entire study period. Productivity and glider export are 2-day running means while float export is averaged over longer periods denoted by the width of the bars. Bar height denotes uncertainty bounds. (b) Near surface glider  $D_{bbp}$  estimates from 10-50 m.**