



Norwegian Sea net community production estimated from O₂ and prototype CO₂ optode measurements on a Seaglider

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Abstract. We report on a pilot study using a CO₂ optode deployed on a Seaglider in the Norwegian Sea from March to October 2014. The optode measurements required drift and lag correction and in situ calibration using discrete water samples collected in the vicinity. We found that the optode signal correlated better with the concentration of CO₂, $c(\text{CO}_2)$, than with its partial pressure, $p(\text{CO}_2)$. Using the calibrated $c(\text{CO}_2)$ and a regional parameterisation of total alkalinity (A_T) as a function of temperature and salinity, we calculated total dissolved inorganic carbon content, $c(\text{DIC})$, which had a standard deviation of $11 \mu\text{mol kg}^{-1}$ compared with in situ measurements. The glider was also equipped with an oxygen (O₂) optode. The O₂ optode was drift corrected and calibrated using a $c(\text{O}_2)$ climatology for deep samples. The calibrated data enabled the calculation of DIC- and O₂-based net community production, $N(\text{DIC})$ and $N(\text{O}_2)$. To derive N , DIC and O₂ inventory changes over time were combined with estimates of air–sea gas exchange, diapycnal mixing and entrainment of deeper waters. Glider-based observations captured two periods of increased Chl a inventory in late spring (May) and a second one in summer (June). For the May period, we found $N(\text{DIC}) = (21 \pm 5) \text{ mmol m}^{-2} \text{ d}^{-1}$, $N(\text{O}_2) = (94 \pm 16) \text{ mmol m}^{-2} \text{ d}^{-1}$ and an (uncalibrated) Chl a peak concentration of $c_{\text{raw}}(\text{Chl } a) = 3 \text{ mg m}^{-3}$. During the June period, $c_{\text{raw}}(\text{Chl } a)$ increased to a summer maximum of

4 mg m^{-3} , associated with $N(\text{DIC}) = (85 \pm 5) \text{ mmol m}^{-2} \text{ d}^{-1}$ and $N(\text{O}_2) = (126 \pm 25) \text{ mmol m}^{-2} \text{ d}^{-1}$. The high-resolution dataset allowed for quantification of the changes in N before, during and after the periods of increased Chl a inventory. After the May period, the remineralisation of the material produced during the period of increased Chl a inventory decreased $N(\text{DIC})$ to $(-3 \pm 5) \text{ mmol m}^{-2} \text{ d}^{-1}$ and $N(\text{O}_2)$ to $(0 \pm 2) \text{ mmol m}^{-2} \text{ d}^{-1}$. The survey area was a source of O₂ and a sink of CO₂ for most of the summer. The deployment captured two different surface waters influenced by the Norwegian Atlantic Current (NwAC) and the Norwegian Coastal Current (NCC). The NCC was characterised by lower $c(\text{O}_2)$ and $c(\text{DIC})$ than the NwAC, as well as lower $N(\text{O}_2)$ and $c_{\text{raw}}(\text{Chl } a)$ but higher $N(\text{DIC})$. Our results show the potential of glider data to simultaneously capture time- and depth-resolved variability in DIC and O₂ concentrations.

1 Introduction

Climate models project an increase in the atmospheric CO₂ mole fraction driven by anthropogenic emissions from a pre-industrial value of $280 \mu\text{mol mol}^{-1}$ (Neftel et al., 1982) to 538–936 $\mu\text{mol mol}^{-1}$ by 2100 (Pachauri and Reisinger, 2007). The ocean is known to be a major CO₂ sink (Sabine et al., 2004; Le Quééré et al., 2009; Sutton et al., 2014); in fact, it

has taken up approximately 25 % of this anthropogenic CO₂ with a rate of $(2.5 \pm 0.6) \text{ Gt a}^{-1}$ (in C equivalents) (Friedlingstein et al., 2019). This uptake alters the carbonate system of seawater and is causing a decrease in seawater pH, a process known as ocean acidification (Gattuso and Hansson, 2011). The processes affecting the marine carbonate system include air–sea gas exchange, photosynthesis and respiration, advection and vertical mixing, and CaCO₃ formation and dissolution. For that reason, it is important to develop precise, accurate and cost-effective tools to observe CO₂ trends, variability and related processes in the ocean. Provided that suitable sensors are available, autonomous ocean glider measurements may help resolve these processes.

To quantify the marine carbonate system, four variables are commonly measured: total dissolved inorganic carbon concentration, $c(\text{DIC})$; total alkalinity, A_T ; the fugacity of CO₂, $f(\text{CO}_2)$; and pH. At thermodynamic equilibrium, knowledge of two of the four variables is sufficient to calculate the other two. Marine carbonate system variables are primarily measured on research ships, commercial ships of opportunity, moorings, buoys and floats (Hardman-Mountford et al., 2008; Monteiro et al., 2009; Takahashi et al., 2009; Olsen et al., 2016; Bushinsky et al., 2019). Moorings equipped with submersible sensors often provide limited vertical and horizontal but good long-term temporal resolution (Hemsley, 2003). In contrast, ship-based surveys have higher vertical and spatial resolution than moorings but limited repetition frequency because of the expense of ship operations. Ocean gliders have the potential to replace some ship surveys because they are much cheaper to operate and will increase our coastal and regional observational capacity. However, the slow glider speed of $1\text{--}2 \text{ km h}^{-1}$ only allows a smaller spatial coverage than ship surveys, and the sensors require careful calibration to match the quality of data provided by ship-based sampling.

Carbonate system sensors suitable for autonomous deployment have been developed in the past decades, in particular pH sensors (Seidel et al., 2008; Martz et al., 2010; Rérolle et al., 2013) and $p(\text{CO}_2)$ sensors (Atamanchuk et al., 2014; Bittig et al., 2012; Degrandpre, 1993; Goyet et al., 1992; Körtzinger et al., 1996). One of these sensors is the CO₂ optode (Atamanchuk et al., 2014), which has been successfully deployed to monitor an artificial CO₂ leak on the Scottish west coast (Atamanchuk et al., 2015b), on a cabled underwater observatory (Atamanchuk et al., 2015a), to measure lake metabolism (Peeters et al., 2016), for fish transportation (Thomas et al., 2017) and on a moored profiler (Chu et al., 2020).

$c(\text{DIC})$ and $c(\text{O}_2)$ measurements can be used to calculate net community production (NCP), which is defined as the difference between gross primary production (GPP) and community respiration (CR). At steady state, NCP is equal to the rate of organic carbon export and transfer from the surface into the mesopelagic and deep waters (Lockwood et al., 2012). NCP is derived by vertical integration to a spe-

cific depth that is commonly defined relative to the mixed layer depth (z_{mix}) or the bottom of the euphotic zone (Plant et al., 2016). A system is defined as autotrophic when GPP is larger than CR (i.e. NCP is positive) and as heterotrophic when CR is larger than GPP (i.e. NCP is negative) (Ducklow and Doney, 2013).

NCP can be quantified using bottle incubations or in situ biogeochemical budgets (Sharples et al., 2006; Quay, et al., 2012; Seguro et al., 2019). Bottle incubations involve measuring production and respiration in vitro under dark and light conditions. Biogeochemical budgets combine O₂ and DIC inventory changes with estimates of air–sea gas exchange, entrainment, advection and vertical mixing (Neuer et al., 2007; Alkire et al., 2014; Binetti et al., 2020).

The Norwegian Sea is a complex environment due to the interaction between the Norwegian Atlantic Current (NwAC) entering from the south-west, Arctic Water coming from the north and the Norwegian Coastal Current (NCC) flowing along the Norwegian coast (Nilsen and Falck, 2006). In particular, Atlantic Water enters the Norwegian Sea through the Faroe–Shetland Channel and Iceland–Faroe Ridge (Hansen and Østerhus, 2000) with salinity S between 35.1 and 35.3 and temperatures (θ) warmer than 6 °C (Swift, 1986). The NCC water differs from the NwAC with a surface $S < 35$ (Saetre and Ljoen, 1972) and a seasonal θ signal (Nilsen and Falck, 2006).

Biological production in the Norwegian Sea varies during the year and five different periods can be discerned (Rey, 2004): (1) winter with the smallest productivity and phytoplankton biomass; (2) a pre-bloom period; (3) the spring bloom when productivity increases and phytoplankton biomass reaches the annual maximum; (4) a post-bloom period with productivity mostly based on regenerated nutrients; (5) autumn with smaller blooms than in summer. Previous estimates of the DIC-based net community production ($N(\text{DIC})$) were based on discrete $c(\text{DIC})$ samples (Falck and Anderson, 2005) or calculated from $c(\text{O}_2)$ measurements and converted to C equivalents assuming Redfield stoichiometry of production or respiration (Falck and Gade, 1999; Skjelvan et al., 2001; Kivimäe, 2007). Glider measurements have been used to estimate NCP in other ocean regions (Nicholson et al., 2008; Alkire et al., 2014; Haskell et al., 2019; Binetti et al., 2020); however, as far as we know, this is the first study of net community production in the Norwegian Sea using a high-resolution glider dataset ($> 10^6$ data points; 40 s time resolution) and the first anywhere estimating NCP from a glider-mounted sensor directly measuring the marine carbonate system.

Table 1. Average sampling interval of Sea-Bird CTD, Aanderaa 4330F oxygen optode, Aanderaa 4797 CO₂ optode and a combined backscatter/chlorophyll *a* fluorescence sensor (Wetlabs Eco Puck BB2FLVMT) in the top 100, from 100 to 500 and from 500 to 1000 m.

Depth/m	<i>t</i> (CTD)/s	<i>t</i> (O ₂)/s	<i>t</i> (CO ₂)/s	<i>t</i> (Chl <i>a</i>)/s
0–100 m	24	49	106	62
100–500 m	31	153	233	–
500–1000 m	42	378	381	–

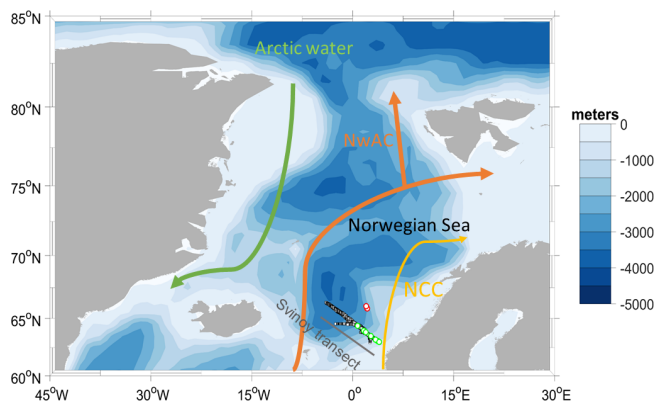


Figure 1. Map of the glider deployment and the main currents. The black dots are the glider dives; the green and the red dots are the water samples collected along the glider section and at Ocean Weather Station M (OWSM), respectively. The three main water masses (Skjelvan et al., 2008) are the Norwegian Coastal Current (yellow), the Norwegian Atlantic Current (NwAC, orange) and Arctic Water (green).

2 Materials and methods

2.1 Glider sampling

Kongsberg Seaglider 564 was deployed in the Norwegian Sea on 16 March 2014 at 63.00° N, 3.86° E, and recovered on 30 October 2014 at 62.99° N, 3.89° E. The Seaglider was equipped with a prototype Aanderaa 4797 CO₂ optode, an Aanderaa 4330F oxygen optode (Tengberg et al., 2006), a Sea-Bird conductivity–temperature–depth profiler (CTD) and a combined backscatter–chlorophyll *a* fluorescence sensor (Wetlabs Eco Puck BB2FLVMT). The mean sampling intervals for each sensor varied with depth (Table 1).

The deployment followed the Svinøy trench from the open sea towards the Norwegian coast. The glider covered a 536 km long transect eight times (four times in each direction) for a total of 703 dives (Fig. 1).

2.2 Discrete sampling

During the glider deployment, 70 discrete water samples from various depths (5, 10, 20, 30, 50, 100, 300, 500 and

1000 m) were collected on five different cruises on the R/V *Haakon Mosby* along the southern half of the glider transect on 18 March, 5 May, 6 and 14 June, and 30 October 2014. Samples for *c*(DIC) and *A*_T were collected from 10 L Niskin bottles following standard operational procedure (SOP) 1 of Dickson et al. (2007). The *c*(DIC) and *A*_T samples were preserved with saturated HgCl₂ solution (final HgCl₂ concentration: 15 mg dm⁻³) and analysed within 14 d after the collection. Nutrient samples from the same Niskin bottles were preserved with chloroform (Hagebo and Rey, 1984). *c*(DIC) and *A*_T were analysed on shore according to SOPs 2 and 3b (Dickson et al., 2007) using a VINDTA 3D (Marianda) with a CM5011 coulometer (UIC instruments) and a VINDTA 3S (Marianda), respectively. The precision of the samples' *c*(DIC) and *A*_T values was 1 μmol kg⁻¹ for both, based on duplicate samples and running Certified Reference Material (CRM) batch numbers 118 and 138 provided by professor A. Dickson, Scripps Institution of Oceanography, San Diego, USA (Dickson et al., 2003). Nutrients were analysed on shore using an Alpkem AutoAnalyzer. In addition, 43 water samples were collected at Ocean Weather Station M (OWSM) on five different cruises: 22 March on R/V *Haakon Mosby*, 9 May on R/V *G.O. Sars*, 14 June on R/V *Haakon Mosby*, 2 August and 13 November 2014 on R/V *Johan Hjort* from 10, 30, 50, 100, 200, 500, 800 and 1000 m depths. The OWSM samples were preserved and analysed for *A*_T and *c*(DIC) as the Svinøy samples. No phosphate and silicate samples were collected at OWSM. Temperature (*θ*) and salinity (*S*) profiles were measured at each station using a Sea-Bird 911 plus CTD. pH and *f*(CO₂) were calculated using the MATLAB toolbox CO2SYS (Van Heuven et al., 2011), with the following constants: *K*₁ and *K*₂ carbonic acid dissociation constants of Lueker et al. (2000), *K*(HSO₄⁻/SO₄²⁻) bisulfate dissociation constant of Dickson (1990) and borate to chlorinity ratio of Lee et al. (2010). The precision of *A*_T and *c*(DIC) led to an uncertainty in the calculated *c*(CO₂) of 0.28 μmol kg⁻¹. For the OWSM calculations, we used nutrient concentrations from the Svinøy section at a time as close as possible to the OWSM sampling as input. In the case of the glider, we derived a parameterisation for phosphate and silicate concentration as a function of sample depth and time. This parameterisation had an uncertainty of 1.3 and 0.13 μmol kg⁻¹ and a *R*² of 0.6 and 0.4, for silicate and phosphate concentrations, respectively. The uncertainty was calculated as the root mean square difference between measured and parameterised concentrations. This nutrient concentration uncertainty contributed an uncertainty of 0.04 μmol kg⁻¹ in the calculation of *c*(CO₂), which is negligible and smaller than the uncertainty caused by *A*_T and *c*(DIC).

2.3 Oxygen optode calibration

The last oxygen optode calibration before the deployment was performed in 2012 as a two-point calibration at 9.91 °C

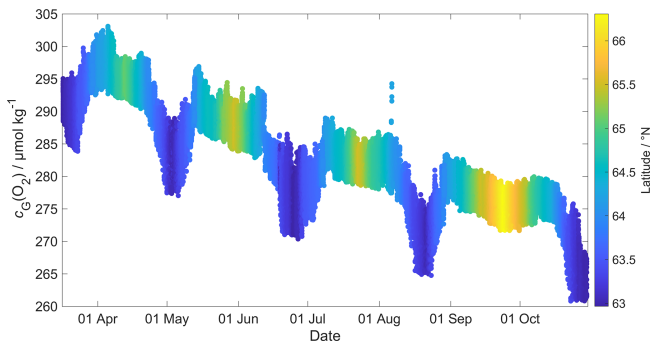


Figure 2. Glider oxygen concentration, $c_G(\text{O}_2)$, for $\sigma_0 > 1028 \text{ kg m}^{-3}$ coloured by latitude.

in air-saturated water and at 20.37 °C in anoxic Na₂SO₃ solution. Oxygen optodes are known to be affected by drift (Bittig and Körtzinger, 2015), which is even worse for the fast-response foils used in the 4330F optode for glider deployments. It has been suggested that it is necessary to calibrate and drift correct the optode using discrete samples or in-air measurements (Nicholson and Feen, 2017). Unfortunately, no discrete samples were collected at glider deployment or recovery.

To overcome this problem, we used archived data to correct for oxygen optode drift. These archived concentration data (designated $c_C(\text{O}_2)$) were collected at OWSM between 2001 and 2007 (downloaded from ICES database) and in the glider deployment region between 2000 and 2018 (extracted from GLODAPv2; Olsen et al., 2016). To apply the correction, we used the oxygen samples corresponding to a potential density $\sigma_0 > 1028 \text{ kg m}^{-3}$ (corresponding to depths between 427 and 1000 m), because waters of these potential densities were always well below the mixed layer and therefore subject to limited seasonal and interannual variability, as evidenced by the salinity S and potential temperature θ of these samples: S varied from 34.88 to 34.96, with a mean of 34.90 ± 0.01 ; θ varied from 0.45 to -0.76 °C, with a mean of (-0.15 ± 0.36) °C.

Figure 2 shows that the glider oxygen concentration ($c_G(\text{O}_2)$) corresponding to $\sigma_0 > 1028 \text{ kg m}^{-3}$ was characterised by two different water masses separated at a latitude of about 64° N. We used the samples collected north of 64° N to derive the glider optode correction because this reflects the largest area covered by the glider. We did not use the southern region because the archived samples from there covered only 5 d. For each day of the year with archived samples, we calculated the median concentration of the glider and the archived samples. Figure 3 shows a plot of the ratio between $c_C(\text{O}_2)/c_G(\text{O}_2)$ against the day of the year and a linear fit, which is used to calibrate $c_G(\text{O}_2)$ and correct for drift.

No lag correction was applied because the O₂ optode had a fast response foil and showed no detectable lag (< 10 s), based on a comparison between descent and ascent profiles.

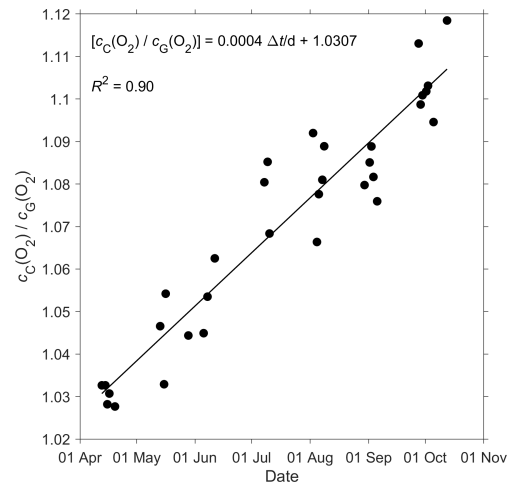


Figure 3. A linear fit of the ratio between the daily median of the discrete oxygen samples ($c_C(\text{O}_2)$) and glider oxygen data ($c_G(\text{O}_2)$) for $\sigma_0 > 1028 \text{ kg m}^{-3}$ was used to derive the $c_G(\text{O}_2)$ drift and initial offset at deployment. The time difference Δt is calculated with respect to the deployment day on 16 March.

2.4 CO₂ optode measurement principle

The CO₂ optode consists of an optical and a temperature sensor incorporated into a pressure housing. The optical sensor has a sensing foil comprising two fluorescence indicators (luminophores), one of which is sensitive to pH changes, and the other is not and thus used as a reference. The excitation and emission spectra of the two fluorescence indicators overlap, but the reference indicator has a longer fluorescence lifetime than the pH indicator. These two fluorescence lifetimes are combined using an approach known as dual lifetime referencing (DLR) (Klimant et al., 2001; von Bültzingslöwen et al., 2002). From the phase shift (φ), the partial pressure of CO₂, $p(\text{CO}_2)$, is parameterised as an eight-degree polynomial (Atamanchuk et al., 2014):

$$\log[p(\text{CO}_2)/\mu\text{atm}] = C_0 + C_1\varphi + \dots + C_8\varphi^8, \quad (1)$$

where C_0 to C_8 are temperature-dependent coefficients.

The partial pressure of CO₂ is linked to the CO₂ concentration, $c(\text{CO}_2)$ and the fugacity of CO₂, $f(\text{CO}_2)$, via the following relationship:

$$\begin{aligned} c(\text{CO}_2) &= p(\text{CO}_2)/[1 - p(\text{H}_2\text{O})/p]F(\text{CO}_2) \\ &= K_0(\text{CO}_2)f(\text{CO}_2), \end{aligned} \quad (2)$$

where $F(\text{CO}_2)$ is the solubility function (Weiss and Price, 1980), $p(\text{H}_2\text{O})$ is the water vapour pressure, p is the total gas tension (assumed to be near 1 atm), and $K_0(\text{CO}_2)$ is the solubility coefficient. F and K_0 vary according to temperature and salinity.

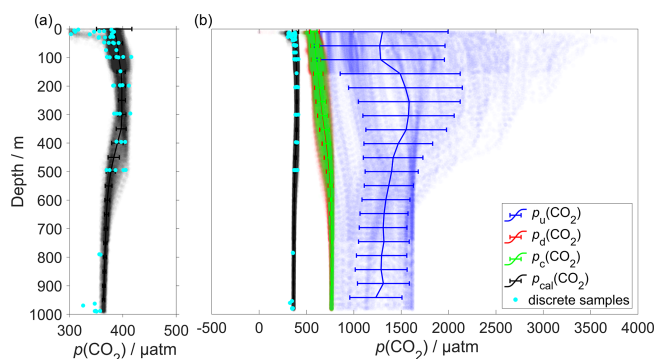


Figure 4. Panel (a) shows the calibrated $p(\text{CO}_2)$ ($p_{\text{cal}}(\text{CO}_2)$) in black and the discrete samples in azure. (b) Plot of $p(\text{CO}_2)$ versus depth where the continuous vertical lines are the mean every 50 m and the error bars represent the standard deviation. Blue colour shows $p_u(\text{CO}_2)$ without any correction; red shows $p_d(\text{CO}_2)$ corrected for drift; green represents $p_c(\text{CO}_2)$ corrected for drift and lag; black shows $p_{\text{cal}}(\text{CO}_2)$ calibrated against water samples (azure dots) collected during the deployment (Sect. 2.5). $p_{\text{cal}}(\text{CO}_2)$ had a mean standard deviation of 22 μatm and a mean bias of 1.8 μatm compared with the discrete samples.

2.5 CO₂ optode lag and drift correction and calibration

The CO₂ optode was fully functional between dives 31 (on 21 March 2014) and 400 (on 24 July 2014). After dive 400, the CO₂ optode stopped sampling in the top 150 m. Figure 4 shows the outcome of each calibration step: (0) uncalibrated optode output (blue dots), (1) drift correction (red dots), (2) lag correction (green dots) and (3) calibration using discrete water samples (black dots).

In order to correct for the drift occurring during the glider mission, we selected the CO₂ optode measurements in water with $\sigma_0 > 1028 \text{ kg m}^{-3}$ (just as for O₂; Sect. 2.3). We calculated the median of the raw optode phase shift data (“Cal-Phase” φ_{cal}) for each Seaglider dive. Then, we calculated a drift coefficient (m_i) as the ratio between the median φ_{cal} for a given dive divided by the median φ_{cal} of dive 31. Drift-corrected $\varphi_{\text{cal,d}}$ values were calculated by dividing the raw φ_{cal} by the specific m_i for each dive.

The CO₂ optode was also affected by lag (Atamanchuk et al., 2014) caused by the slow response of the optode to ambient $c(\text{CO}_2)$ changes in time and depth. The lag created a discrepancy between the depth profiles obtained during glider ascents and descents. To correct for this lag, we applied the method of Miloshevich (2004), which was previously used by Fiedler et al. (2013) and Atamanchuk et al. (2015b) to correct the lag of the Contros HydroC CO₂ sensor (Fiedler et al., 2013; Saderne et al., 2013). This CO₂ sensor has a different measurement principle (infrared absorption) than the CO₂ optode, but both rely on the diffusion of CO₂ through a gas-permeable membrane.

To apply the lag correction, the sampling interval (Δt) needs to be sufficiently small compared to the sensor re-

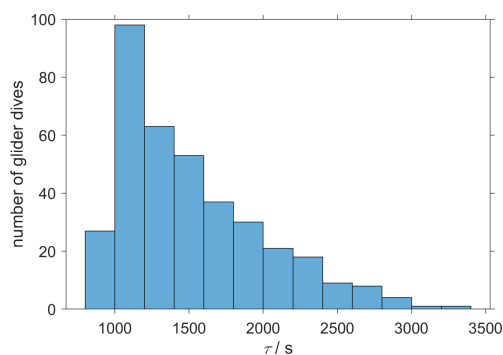


Figure 5. The histogram shows the distribution of the τ calculated from glider dives 31 to 400 to correct the CO₂ optode drift using the algorithm of Miloshevich (2004).

sponse time (τ) and the ambient variability (Miloshevich, 2004). Before the lag correction, $\varphi_{\text{cal,d}}$ was smoothed to remove any outliers and “kinks” in the profile using the Matlab function rLOWESS. The smoothing function applies a local regression every nine points using a weighted robust linear least-squares fit. Subsequently, τ was determined such that the following lag-correction equation (Miloshevich, 2004) minimised the $\varphi_{\text{cal,d}}$ difference between each glider ascent and the following descent:

$$p_c(\text{CO}_2, t_1) = \frac{p_d(\text{CO}_2, t_1) - p_d(\text{CO}_2, t_0)e^{-\Delta t/\tau}}{1 - e^{-\Delta t/\tau}}, \quad (3)$$

where $p_d(\text{CO}_2, t_0)$ is the drift-corrected value measured by the optode at time t_0 , $p_d(\text{CO}_2, t_1)$ is the measured value at time t_1 , Δt is the time between t_0 and t_1 , τ is the response time, and $p_c(\text{CO}_2, t_1)$ is the lag-corrected value at t_1 . We calculated a τ value for each glider dive and used the median of τ (1384 s, 25th quartile: 1101 s; 75th quartile: 1799 s) (Fig. 5), which was larger than Δt (258 s) and therefore met the requirement to apply the Miloshevich (2004) method. To apply the lag correction, the glider needs to sample same water mass during the ascent and descent. The difference between the ascent and descent was minimal: it was $(0.13 \pm 0.33)^\circ\text{C}$ for θ and 0.02 ± 0.04 for S . This lag correction reduced the average difference between glider ascent and descent from (71 ± 30) to $(21 \pm 26) \mu\text{atm}$.

The CO₂ optode output was calibrated using the discrete samples collected throughout the mission. Using the discrete sample time and potential density σ_0 , we selected the closest CO₂ optode output. A linear regression between optode output and $c(\text{CO}_2)$ from the discrete samples ($c_{\text{WS}}(\text{CO}_2)$) was used to calibrate the optode output $p_c(\text{CO}_2)$ in terms of $c(\text{CO}_2)$. $c(\text{CO}_2)$ had a better correlation than $p(\text{CO}_2)$ ($R^2 = 0.77$ vs. $R^2 = 0.02$).

Plotting the regression residuals ($c_r(\text{CO}_2)$, calculated as the difference between $c_{\text{WS}}(\text{CO}_2)$ and the value predicted by the regression) revealed a quadratic relation between the regression residuals and water temperature (θ). We have therefore included θ and θ^2 in the optode calibration (Fig. 6a).

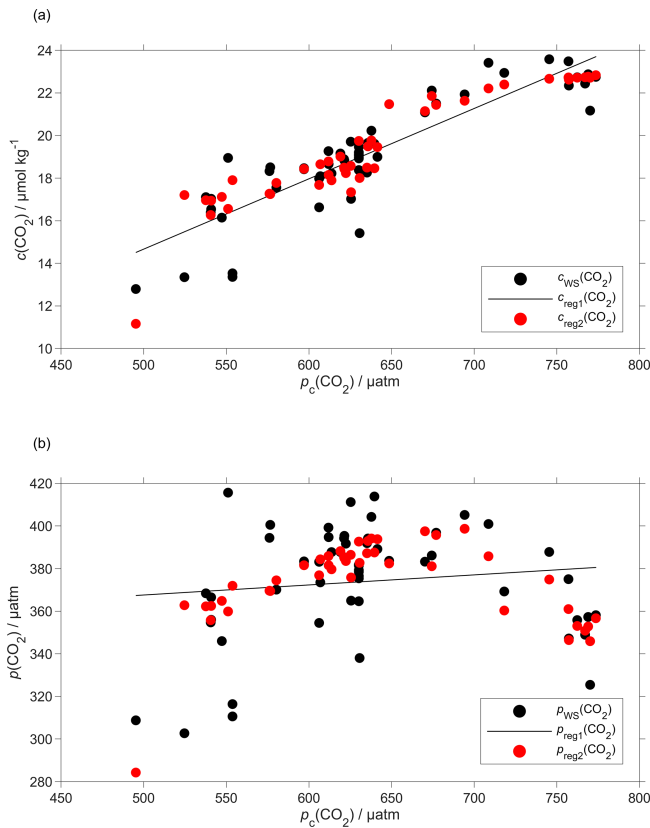


Figure 6. Regression (black lines, reg1) of the CO₂ optode output $p_c(\text{CO}_2)$ against (a) co-located concentration $c_{\text{WS}}(\text{CO}_2)$ that has an uncertainty of $0.28 \mu\text{mol kg}^{-1}$ (b) and partial pressure $p_{\text{WS}}(\text{CO}_2)$ of CO₂ in discrete water samples (black dots). Also shown are the values predicted by including θ and θ^2 in the regression used for optode calibration (red dots, reg2). The regression equations are (a) reg1: $c_{\text{WS}}(\text{CO}_2)/(\mu\text{mol kg}^{-1}) = (0.033 \pm 0.003)p_c(\text{CO}_2)/\mu\text{atm} - 1.8 \pm 1.6$ ($R^2 = 0.77$); reg2: $c_{\text{WS}}(\text{CO}_2)/(\mu\text{mol kg}^{-1}) = (0.12 \pm 0.14)\theta/^\circ\text{C} - (0.071 \pm 0.011)(\theta/^\circ\text{C})^2 + (0.0094 \pm 0.0048)p_c(\text{CO}_2)/\mu\text{atm} + 16 \pm 4$ ($R^2 = 0.90$). (b) reg1: $p_{\text{WS}}(\text{CO}_2)/\mu\text{atm} = (0.05 \pm 0.05)p_c(\text{CO}_2)/\mu\text{atm} + 344 \pm 33$ ($R^2 = 0.02$); reg2: $p_{\text{WS}}(\text{CO}_2)/\mu\text{atm} = (21 \pm 3)\theta/^\circ\text{C} - (1.9 \pm 0.2)(\theta/^\circ\text{C})^2 + (0.2 \pm 0.1)p_c(\text{CO}_2)/\mu\text{atm} + 209 \pm 76$ ($R^2 = 0.60$).

This second calibration increased the correlation coefficient R^2 from 0.77 to 0.90 and decreased the standard deviation of the regression residuals from 1.3 to $0.8 \mu\text{mol kg}^{-1}$. Even with the explicit inclusion of temperature in the calibration, the CO₂ optode response remained more closely related to $c(\text{CO}_2)$ than $p(\text{CO}_2)$ (Fig. 6b).

2.6 Regional algorithm to estimate A_T

To calculate $c(\text{DIC})$, we used two variables: (1) glider $c(\text{CO}_2)$ derived as described in Sect. 2.5 and (2) A_T derived using a regional algorithm based on S and θ depths of less than 1000 m. The algorithm followed the approach of Lee et al. (2006) and was derived using 663 water samples collected at OWSM from 2004 to 2014 and GLODAPv2 (Olsen

et al., 2016) data from the year 2000 in the deployment region. Discrete samples with $S < 33$ were removed because these values were lower than the minimum S measured by the glider. The derived A_T parameterisation is

$$A_{T,\text{reg}}/(\mu\text{mol kg}^{-1}) = 2317.03 \pm 12 + 33.12 \pm 6.21(S - 35) + 7.94 \pm 11.38(S - 35)^2 + 0.96 \pm 1.79(\theta/^\circ\text{C} - 20) + 0.01 \pm 0.06(\theta/^\circ\text{C} - 20)^2. \quad (4)$$

The parameterisation has an uncertainty of $8.2 \mu\text{mol kg}^{-1}$ calculated as the standard deviation of the residual difference between actual and parameterised A_T .

To test this parameterisation, we compared the predicted $A_{T,\text{reg}}$ values with discrete measurements ($A_{T,\text{WS}}$) collected close in terms of time, potential density (σ_0) and distance to the glider transect ($n = 60$). These discrete samples and the glider had mean temperature and salinity differences of $(0.17 \pm 0.68)^\circ\text{C}$ and 0.03 ± 0.013 , respectively. The mean difference between $A_{T,\text{WS}}$ and $A_{T,\text{reg}}$ was $(2.1 \pm 6.5) \mu\text{mol kg}^{-1}$.

This A_T parameterisation was used in CO₂SYST (Van Heuven et al., 2011) to calculate $c(\text{DIC})$ from $A_{T,\text{reg}}$ and the calibrated $c(\text{CO}_2)$, $c_{\text{G,cal}}(\text{CO}_2)$. These calculated $c_{\text{G,cal}}(\text{DIC})$ values were compared with $c_{\text{WS}}(\text{DIC})$ of the same set of discrete samples used to calibrate $c_{\text{G,cal}}(\text{CO}_2)$, the only difference being that instead of the actual total alkalinity of the water sample ($A_{T,\text{WS}}$), we used $A_{T,\text{reg}}$. The mean difference between $c_{\text{G,cal}}(\text{DIC})$ and $c_{\text{WS}}(\text{DIC})$ was $(3 \pm 11) \mu\text{mol kg}^{-1}$, with the non-zero bias and the standard deviation due to the uncertainties in the $A_{T,\text{reg}}$ parameterisation and the $c_{\text{G,cal}}(\text{CO}_2)$ calibration.

2.7 Quality control of other measurement variables

The thermal lag of the glider conductivity sensor was corrected using the method of Gourcuff (2014). Single-point outliers in conductivity were removed and replaced by linear interpolation. The glider CTD salinity was affected by presumed particulate matter stuck in the conductivity cell (Medeot et al., 2011) during dives 147, 234, 244, 251, 272, 279, 303, 320 and 397, and sensor malfunction caused a poor match between glider ascent and descent during dives 214, 215, 235 and 243. These dives were removed from the subsequent analysis.

Glider-reported chlorophyll concentrations, $c_{\text{raw}}(\text{Chl } a)$, were computed using the factory coefficients. $c_{\text{raw}}(\text{Chl } a)$ was affected by photochemical quenching during the daytime dives. To correct for quenching, we used the method of Hemsley et al. (2015) based on the nighttime relationship between fluorescence and optical backscatter. This relationship was established in the top 60 m and the nighttime values were selected between sunset and sunrise. We calculated a linear fit between $c_{\text{raw}}(\text{Chl } a)$ measured at night, $c_{\text{N}}(\text{Chl } a)$, and the backscatter signal measured at night (b_{N}).

The slope and the intercept were then used to derive corrected daytime $c_D(\text{Chl } a)$. The glider-reported chlorophyll concentration has not been calibrated against in situ samples and is not expected to be accurate, even after correction for quenching. However, it should give an indication of the depth of the deep chlorophyll concentration maximum (z_{DCM}) and the direction of chlorophyll concentration change (up/down). The 8 d means of $c_{\text{raw}}(\text{Chl } a)$ were compared with satellite 8 d composite chlorophyll concentration (Fig. 7) from Ocean Colour CCI (<https://esa-oceancolour-cci.org/>, last access: 7 May 2020) and gave a mean difference of $(0.12 \pm 0.08) \text{ mg m}^{-3}$.

2.8 Calculation of oxygen-based net community production $N(\text{O}_2)$

Calculating net community production N from glider data is challenging because the glider continuously moves through different water masses. For that reason, we subdivided the transect by binning the data into 0.1° latitude intervals to derive O₂ concentration changes every two transects. The changes were calculated between transects in the same direction of glider travel (e.g. transects 1 and 3, both in the N–S direction) to have approximately the same time difference (40–58 d) at every latitude. If instead we had used two consecutive transects, this would lead to a highly variable time difference of near 0 to about 50 d along the transect.

We calculated $N(\text{O}_2)$ (in $\text{mmol m}^{-2} \text{d}^{-1}$) from the oxygen inventory changes ($\Delta I(\text{O}_2)/\Delta t$) corrected for air–sea exchange $\Phi(\text{O}_2)$, normalised to z_{mix} when z_{mix} was deeper than the integration depth of z_{lim} , entrainment $E(\text{O}_2)$ and diapycnal eddy diffusion $F_v(\text{O}_2)$:

$$N(\text{O}_2) = \frac{\Delta I(\text{O}_2)}{\Delta t} + \Phi(\text{O}_2) \frac{\min(z_{\text{lim}}, z_{\text{mix}})}{z_{\text{mix}}} - E(\text{O}_2) - F_v(\text{O}_2). \quad (5)$$

The inventory changes were calculated as the difference between two transects of the integrated oxygen concentration $C(\text{O}_2)$. $C(\text{O}_2)$ (in mmol m^{-3}) was derived from the oxygen content $c(\text{O}_2)$ (in $\mu\text{mol kg}^{-1}$) by multiplication with the water density (about 1027 kg m^{-3} , but we used the actual values). A default integration depth of 45 m was chosen to capture the deepest extent of the deep chlorophyll maximum (z_{DCM}) found during the deployment, which likely represents the extent of the euphotic zone.

The inventory changes for every latitude bin were calculated using the following equation:

$$\frac{\Delta I(\text{O}_2)}{\Delta t} = \frac{\int_0^{45 \text{ m}} C_{n+1}(\text{O}_2, z) dz - \int_0^{45 \text{ m}} C_n(\text{O}_2, z) dz}{t_{n+1} - t_n}, \quad (6)$$

where n is the transect number, t is the day of the year, and $C(\text{O}_2, z)$ is the vertical O₂ concentration profile.

The air–sea flux of oxygen, $\Phi(\text{O}_2)$, was calculated for each glider dive using the median $C(\text{O}_2)$, θ and S in the top 10 m.

We followed the method of Woolf and Thorpe (1991) that includes the effect of bubble equilibrium supersaturation in the calculations:

$$\Phi(\text{O}_2) = k_w(\text{O}_2) \{ (C(\text{O}_2) - [1 + \Delta_{\text{bub}}(\text{O}_2)] C_{\text{sat}}(\text{O}_2)) \}, \quad (7)$$

where $k_w(\text{O}_2)$ is the gas transfer coefficient, $\Delta_{\text{bub}}(\text{O}_2)$ is the increase of equilibrium saturation due to bubble injection, and $C_{\text{sat}}(\text{O}_2)$ is the oxygen saturation. $C_{\text{sat}}(\text{O}_2)$ was calculated from S and θ using the solubility coefficients of Benson and Krause (1984), as fitted by Garcia and Gordon (1992). $\Delta_{\text{bub}}(\text{O}_2)$ was calculated from the following equation:

$$\Delta_{\text{bub}}(\text{O}_2) = 0.01 \left(\frac{U}{U_0} \right)^2, \quad (8)$$

where U is 10 m wind speed with 1 h resolution (ECMWF ERA5; <https://www.ecmwf.int/en/forecasts/datasets/reanalysis-datasets/era5>, last access: 1 February 2021) and U_0 represents the wind speed when the oxygen concentration is 1 % supersaturated and has a value of 9 m s^{-1} (Woolf and Thorpe, 1991). U has a spatial resolution of 0.25° latitude and 0.25° longitude and was interpolated to the glider position at the beginning of the dive.

The transfer velocity $k_w(\text{O}_2)$ was calculated based on Wanninkhof (2014):

$$\frac{k_w(\text{O}_2)}{\text{cm h}^{-1}} = 0.251 \left(\frac{Sc(\text{O}_2)}{660} \right)^{-0.5} \left(\frac{U}{\text{m s}^{-1}} \right)^2. \quad (9)$$

The Schmidt number, $Sc(\text{O}_2)$, was calculated using the parameterisation of Wanninkhof (2014). To account for wind speed variability, $k_w(\text{O}_2)$ applied to calculate $N(\text{O}_2)$ was a weighted mean. This value was calculated using the varying daily-mean wind speed U in the time interval between t_n and $t_{n+1}(\Delta t)$ (50 d) using a five-point-median z_{mix} (Sect. 3.2) (Reuer et al., 2007). The time interval is the same as the one used to calculate $\frac{\Delta I(\text{O}_2)}{\Delta t}$.

The entrainment flux, $E(\text{O}_2)$, was calculated as the oxygen flux when the mixed layer depth deepens in time and is greater than z_{lim} at time t_2 :

$$E(\text{O}_2) = \frac{I(\text{O}_2, t_1, z_{\text{mix}}(t_2)) \frac{z_{\text{lim}}}{z_{\text{mix}}(t_2)} - I(\text{O}_2, t_1, z_{\text{lim}})}{t_2 - t_1}, \quad (10)$$

where $t_2 - t_1$ represents the change in time, z_{mix} is the mixed layer depth, $I(\text{O}_2, t_1, z_{\text{mix}}(t_2))$ is the expected inventory that would result from a mixed layer deepening to $z_{\text{mix}}(t_2)$ between t_2 and t_1 , and $I(\text{O}_2, t_1, z_{\text{lim}})$ is the original inventory at t_1 .

The effect of diapycnal eddy diffusion (F_v) was calculated at z_{mix} when it was deeper than z_{lim} and at z_{lim} when z_{mix} was shallower than z_{lim} , using the following equation:

$$F_v(\text{O}_2) = K_z \frac{\partial C(\text{O}_2)}{\partial z}, \quad (11)$$

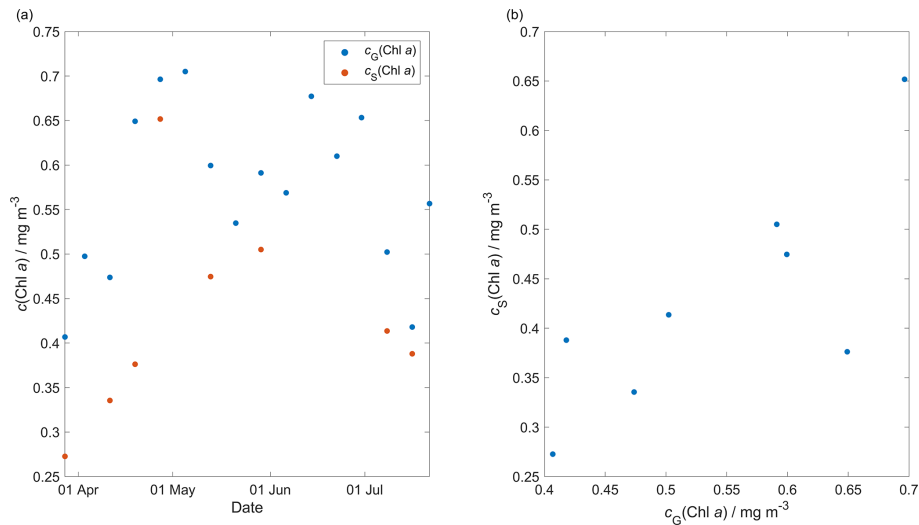


Figure 7. Comparison between the 8 d glider $c(\text{Chl } a)$ ($c_G(\text{Chl } a)$) mean and the 8 d satellite $c(\text{Chl } a)$ ($c_S(\text{Chl } a)$) download from Ocean Colour CCI (<https://esa-oceancolour-cci.org/>, last access: 7 May 2020) as a time series (a) and scatter plot (b).

for a vertical eddy diffusivity (K_z) of $10^{-5} \text{ m}^2 \text{ s}^{-1}$ (Naveira Garabato et al., 2004). The effect of $F_v(\text{O}_2)$ on $N(\text{O}_2)$ was negligible (Fig. A2b) with a median of $(-0.1 \pm 0.5) \text{ mmol m}^{-2} \text{ d}^{-1}$.

2.9 Calculation of dissolved inorganic carbon-based net community production, $N(\text{DIC})$

$N(\text{DIC})$ was expressed in $\text{mmol m}^{-2} \text{ d}^{-1}$ and was calculated from DIC inventory changes ($\Delta I(\text{DIC})/\Delta t$), air–sea flux of CO₂, $\Phi(\text{CO}_2)$, entrainment $E(\text{DIC})$ and diapycnal diffusion $F_v(\text{DIC})$:

$$N(\text{DIC}) = -\frac{\Delta(\text{DIC})}{\Delta t} - \Phi(\text{CO}_2) \frac{\min(z_{\text{lim}}, z_{\text{mix}})}{z_{\text{mix}}} + E(\text{DIC}) + F_v(\text{DIC}). \quad (12)$$

Firstly, $\Phi(\text{CO}_2)$ was calculated using the 10 m wind speed with 1 h resolution downloaded from ECMWF ERA5. As for oxygen, we selected the closest wind speed data point at the beginning of each glider dive. We used the monthly mean atmospheric CO₂ dry mole fraction ($x(\text{CO}_2)$) downloaded from the Greenhouse Gases Reference Network Site (<https://www.esrl.noaa.gov/gmd/ccgg/ggrn.php>, last access: 28 February 2019) closest to the deployment at Mace Head, County Galway, Ireland (Dlugokencky et al., 2015). Using $x(\text{CO}_2)$, we calculated the air-saturation concentration $C_{\text{atm}}(\text{CO}_2)$:

$$C_{\text{atm}}(\text{CO}_2) = x(\text{CO}_2) p_{\text{baro}} F(\text{CO}_2), \quad (13)$$

where p_{baro} is the mean sea level pressure and $F(\text{CO}_2)$ is the CO₂ solubility function (in $\text{mol dm}^{-3} \text{ atm}^{-1}$) calculated from surface θ and S (Weiss and Price, 1980).

The seawater $C(\text{CO}_2)$ at the surface was calculated using the median in the top 10 m between the glider ascent and

descent of the following dive $C(\text{CO}_2)$. From this, $\Phi(\text{CO}_2)$ was calculated:

$$\Phi(\text{CO}_2) = k(\text{CO}_2)[C(\text{CO}_2) - C_{\text{atm}}(\text{CO}_2)]. \quad (14)$$

$k(\text{CO}_2)$ was calculated using the parameterisation of Wanninkhof (2014):

$$\frac{k_w(\text{CO}_2)}{\text{cm h}^{-1}} = 0.251 \left(\frac{Sc(\text{CO}_2)}{660} \right)^{-0.5} \left(\frac{U}{\text{ms}^{-1}} \right)^2. \quad (15)$$

$Sc(\text{CO}_2)$ is the dimensionless Schmidt number at the sea-water temperature (Wanninkhof, 2014). To account for wind speed variability, $k_w(\text{CO}_2)$ applied to calculate $N(\text{DIC})$ was a weighted mean based on the varying daily-mean wind speed U in the time interval between t_n and t_{n+1} (Δt) used to calculate $\frac{\Delta I(\text{DIC})}{\Delta t}$ and for 40–50 d to calculate $\Phi(\text{CO}_2)$ (Sect. 3.2) (Reuer et al., 2007).

The DIC inventory changes were calculated in the top 45 m with the following equation:

$$\frac{\Delta I(\text{DIC})}{\Delta t} = \frac{\int_0^{45 \text{ m}} C_{n+1}(\text{DIC}, z) dz - \int_0^{45 \text{ m}} C_n(\text{DIC}, z) dz}{t_{n+1} - t_n}. \quad (16)$$

Just as for $C(\text{O}_2)$, $C(\text{DIC})$ (in mmol m^{-3}) was derived from the DIC content $c(\text{DIC})$ (in $\mu\text{mol kg}^{-1}$) by multiplication with the water density (about 1027 kg m^{-3} , but we used the actual values).

The entrainment flux, $E(\text{DIC})$, was calculated as the DIC flux when the mixed layer depth deepens in time and is greater than z_{lim} at time t_2 :

$$E(\text{DIC}) = \frac{I(\text{DIC}, t_1, z_{\text{mix}}(t_2)) \frac{z_{\text{lim}}}{z_{\text{mix}}(t_2)} - I(\text{DIC}, t_1, z_{\text{lim}})}{t_2 - t_1}. \quad (17)$$

As for oxygen, the effect of diapycnal eddy diffusion (F_v) was calculated at z_{mix} when it was deeper than z_{lim} and at

z_{lim} when z_{mix} was shallower than z_{lim} , using the following equation:

$$F_v(\text{DIC}) = K_z \frac{\partial C(\text{DIC})}{\partial z}, \quad (18)$$

for a K_z of $10^{-5} \text{ m}^2 \text{ s}^{-1}$ (Naveira Garabato et al., 2004). The effect of $F_v(\text{DIC})$ was negligible (Fig. A2a) with a median of $(0.1 \pm 0.3) \text{ mmol m}^{-2} \text{ d}^{-1}$.

The contribution of horizontal advection to $N(\text{DIC})$ was considered minimal over the timescales where we calculated inventory changes because previous studies have shown that changes in $C(\text{DIC})$ during summer are mainly controlled by biology and air–sea interactions (Gislefoss et al., 1998). For that reason, previous studies that estimated N in the Norwegian Sea have also neglected advective fluxes (Falck and Anderson, 2005; Falck and Gade, 1999; Kivimäe, 2007; Skjelvan et al., 2001).

Uncertainties in $N(\text{DIC})$ and $N(\text{O}_2)$ were evaluated with a Monte Carlo approach. The uncertainties of the input variables are shown in Table 2; we repeated the analysis 1000 times. The total uncertainty in N was calculated as the standard deviation of the 1000 Monte Carlo simulations.

3 Results

The uncorrected $p(\text{CO}_2)$ values presented in Fig. 8 were analysed up to dive 400 (24 July 2014). For the following dives, the CO₂ optode stopped sampling in the first 150 m (Fig. 8d). Instead, the uncorrected temperature θ , salinity S , $c(\text{O}_2)$ and $c_{\text{raw}}(\text{Chl } a)$ were analysed for all the dives (30 October 2014). The raw optode $c(\text{O}_2)$ data were calibrated and drift corrected, and $c(\text{CO}_2)$ was drift corrected, lag corrected and recalibrated, then used to quantify the temporal and spatial changes in N and Φ together with the quenching corrected $c_{\text{raw}}(\text{Chl } a)$ to evaluate net community production changes.

3.1 O₂ and CO₂ optode calibration

The O₂ optode drift caused a continuous and unexpected decrease of the uncorrected $c_G(\text{O}_2)$ from 290 to $282 \mu\text{mol kg}^{-1}$ for $\sigma_0 > 1028 \text{ kg m}^{-3}$ (Fig. 8c). The ratio $c_C(\text{O}_2)/c_G(\text{O}_2)$ against day of the year used for the drift correction had a good correlation with time ($R^2 = 0.90$), showing a continuous increase of 0.0004 d^{-1} (Fig. 3), equivalent to a decrease in the measured glider O₂ concentration of $0.11 \mu\text{mol kg}^{-1} \text{ d}^{-1}$. It was possible to apply the correction because $c_C(\text{O}_2)$ had low temporal variability for the chosen potential density $\sigma_0 > 1028 \text{ kg m}^{-3}$. The $c_C(\text{O}_2)$ values from OWSM and GLODAPv2 had a mean of $(305 \pm 3) \mu\text{mol kg}^{-1}$, varying from 294 to $315 \mu\text{mol kg}^{-1}$ (Fig. A1). The drift correction reduced the variability of $c_G(\text{O}_2)$ in the selected potential density range from a standard deviation of $7.3 \mu\text{mol kg}^{-1}$ to a standard deviation of $2.4 \mu\text{mol kg}^{-1}$ (Fig. 9a).

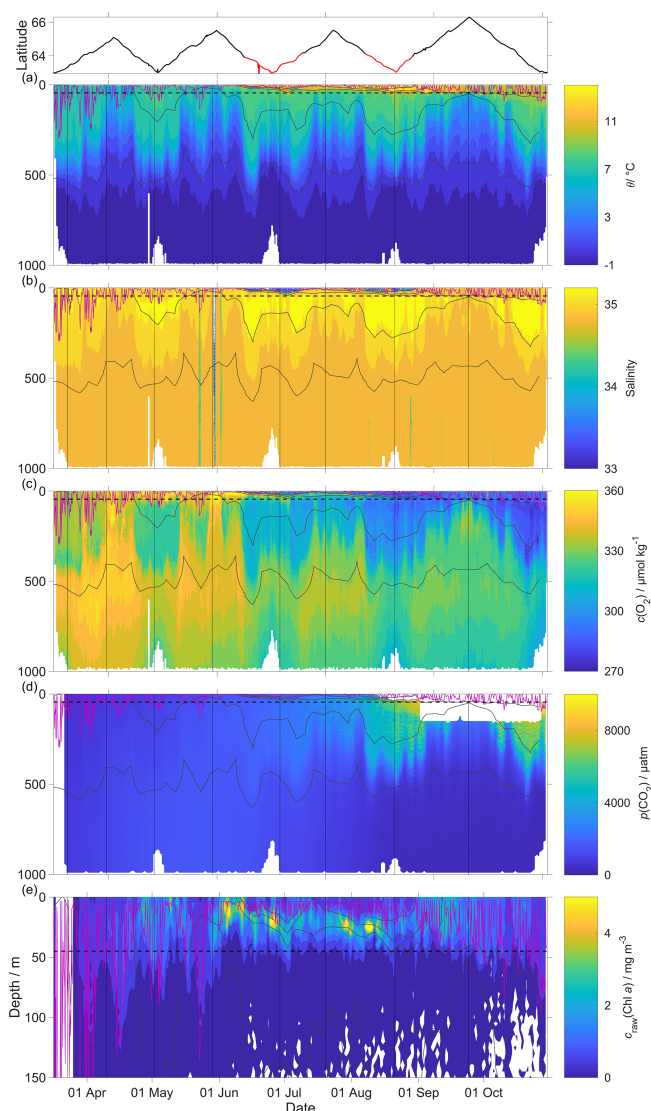
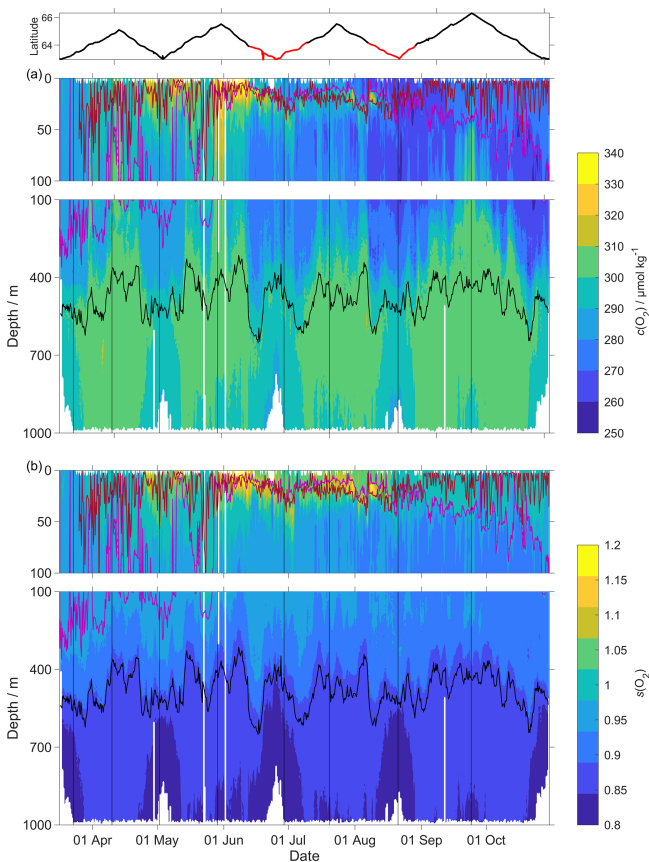


Figure 8. Raw glider data for all 703 dives with latitude of the glider trajectory at the top (black: NwAC; red: NCC, separated by a S of 35). (a) Temperature θ , (b) salinity S , (c) oxygen concentration $c_G(\text{O}_2)$, (d) uncorrected CO₂ optode output $p_u(\text{CO}_2)$ and (e) chlorophyll a concentration $c_{\text{raw}}(\text{Chl } a)$. The white space means that the sensors did not measure any data. The pink line is z_{mix} calculated using a threshold criterion of $\Delta\theta = 0.5 \text{ }^\circ\text{C}$ to the median θ in the top 5 m (Obata et al., 1996; Monterey and Levitus, 1997; Foltz et al., 2003). The dotted black line designates z_{lim} , used as a depth limit to calculate N . Contoured black lines represent isopycnals.

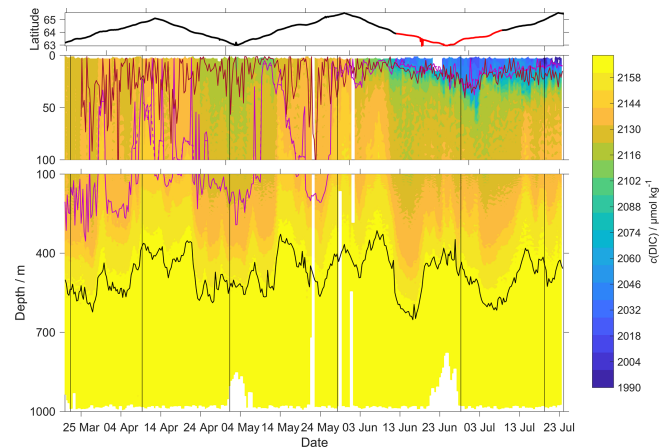
Following drift, lag and scale corrections, glider fugacity $f_G(\text{CO}_2)$ derived from Eq. (2) had a mean difference of $(2 \pm 22) \mu\text{atm}$ to the discrete samples ($n = 55$; not shown) and $c(\text{DIC})$ had a mean difference of $(3 \pm 11) \mu\text{mol kg}^{-1}$ (Fig. 10). $p(\text{CO}_2)$ and $f(\text{CO}_2)$ are almost identical, but $f(\text{CO}_2)$ takes into account the non-ideal nature of the gas phase. The optode was able to capture the temporal and spatial variability showing that NCC had a lower DIC concentra-

Table 2. Uncertainty associated with $N(\text{DIC})$ and $N(\text{O}_2)$ input variables calculated by a Monte Carlo approach.

Variable	Error	Reference/method
$c(\text{DIC})$	$11 \mu\text{mol kg}^{-1}$	Standard deviation of the differences to discrete water samples
S	0.01	Standard deviation of glider salinities for $\sigma_0 > 1028 \text{ kg m}^{-3}$ and latitude $> 64^\circ \text{ N}$
θ	0.3°C	Standard deviation of glider temperature for $\sigma_0 > 1028 \text{ kg m}^{-3}$ and latitude $> 64^\circ \text{ N}$
$C_{\text{atm}}(\text{CO}_2)$	1.5 mmol m^{-3}	Standard deviation of $C_{\text{atm}}(\text{CO}_2)$
$C(\text{CO}_2)$	0.8 mmol m^{-3}	Standard deviation of the differences to discrete water samples
$k_w(\text{CO}_2), k_w(\text{O}_2)$	20 %	Wanninkhof (2014)
z_{mix}	9 m	Standard deviation for z_{mix} based on thresholds $\Delta T = 0.1^\circ \text{C}$ (Sprintall and Roemmich, 1999), 0.2°C (Thompson, 1976) and 0.8°C (Kara et al., 2000)
$c(\text{O}_2)$	$2.4 \mu\text{mol kg}^{-1}$	Standard deviation of glider oxygen concentrations for $\sigma_0 > 1028 \text{ kg m}^{-3}$ and latitude $> 64^\circ \text{ N}$

**Figure 9.** (a) $c(\text{O}_2)$; (b) $s(\text{O}_2) = c(\text{O}_2)/c_{\text{sat}}(\text{O}_2)$ with z_{DCM} (red line), z_{mix} (pink line) five-point median z_{mix} (dotted pink line). The black line indicates $\sigma_0 = 1028 \text{ kg m}^{-3}$. The top panel indicates glider latitude (black: NwAC; red: NCC).

tion than NwAC. Restricting the $f(\text{CO}_2)$ comparison to the discrete samples in the top 10 m gave a mean difference of $(19 \pm 31) \mu\text{atm}$ ($n = 6$). We also compared glider $f_G(\text{CO}_2)$ with the Surface Ocean CO₂ Atlas (SOCAT) $f(\text{CO}_2)$ (Bakker et al., 2016) data in the region during the deployment (Fig. 11). During the whole deployment, there was general agreement between $f_G(\text{CO}_2)$ and $f_{\text{SOCAT}}(\text{CO}_2)$. $f_G(\text{CO}_2)$

**Figure 10.** $c(\text{DIC})$ contour plot with z_{DCM} (red line), z_{mix} (pink line) five-point median z_{mix} (dotted pink line). The black line indicates $\sigma_0 = 1028 \text{ kg m}^{-3}$. The top panel indicates glider latitude (black: NwAC; red: NCC).

varied between 204 and $391 \mu\text{atm}$, while $f_{\text{SOCAT}}(\text{CO}_2)$ varied between 202 and $428 \mu\text{atm}$ (Fig. 11).

Our results are in agreement with Jeansson et al. (2011), who found the surface NCC was the region with the lowest $c(\text{DIC})$ values ($2083 \mu\text{mol kg}^{-1}$) in the Norwegian Sea. This was confirmed during our deployment because $c(\text{DIC})$ was $(2081 \pm 39) \mu\text{mol kg}^{-1}$ in the NCC region and $(2146 \pm 27) \mu\text{mol kg}^{-1}$ in the NwAC region (Fig. 10) and $c(\text{O}_2)$ was $> 300 \mu\text{mol kg}^{-1}$ in the NwAC and $< 280 \mu\text{mol kg}^{-1}$ in the NCC.

3.2 Air–sea exchange

The surface water was supersaturated with oxygen all summer (Fig. 12). From May, this supersaturation drove a continuous O₂ flux from the sea to the atmosphere. However, the flux varied throughout the deployment, having a median of $25 \text{ mmol m}^{-2} \text{ d}^{-1}$ (5th percentile: $-31 \text{ mmol m}^{-2} \text{ d}^{-1}$; 95th percentile: $88 \text{ mmol m}^{-2} \text{ d}^{-1}$). Prior to the spring period of increased Chl *a* inventory, the supersaturation var-

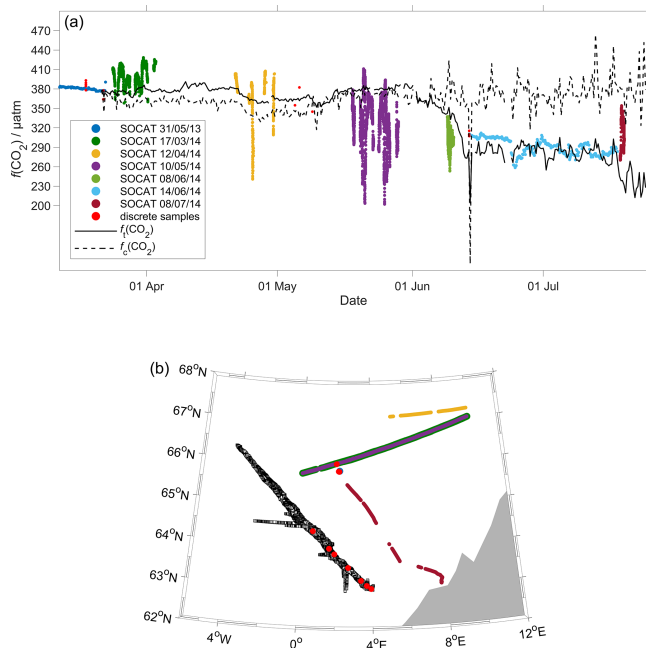


Figure 11. Comparison between surface $f(\text{CO}_2)$ from 2014 SOCAT and CO_2 optode on the glider. **(a)** The black lines are the median glider $f(\text{CO}_2)$ in the top 10 m, with $f_c(\text{CO}_2)$ (dotted line) corresponding to regression 1 (Fig. 6a) and $f_t(\text{CO}_2)$ (continuous line) to regression 2 (Fig. 6a). Discrete samples collected during the deployment are shown as red dots, with the other coloured dots representing cruises in the SOCAT database (Bakker et al., 2016). **(b)** Glider and SOCAT data positions (same colours as in panel a).

ied between 0 to $10 \mu\text{mol kg}^{-1}$. $\Phi(\text{O}_2)$ had a median of $-1.4 \text{ mmol m}^{-2} \text{ d}^{-1}$ (5th percentile: $-49 \text{ mmol m}^{-2} \text{ d}^{-1}$; 95th percentile: $23 \text{ mmol m}^{-2} \text{ d}^{-1}$). Then, during the spring period of increased Chl *a* inventory, the surface concentration increased by over $35 \mu\text{mol kg}^{-1}$, causing a peak in $\Phi(\text{O}_2)$ of $140 \text{ mmol m}^{-2} \text{ d}^{-1}$. A second period of increased Chl *a* inventory was encountered in June and had a larger $\Phi(\text{O}_2)$ up to $118 \text{ mmol m}^{-2} \text{ d}^{-1}$, driven by supersaturation of $68 \mu\text{mol kg}^{-1}$. The fluxes were smaller than during spring and were associated with an increase of $c_{\text{raw}}(\text{Chl } a)$ from 2.5 mg m^{-3} to the summer maximum of 4.0 mg m^{-3} . However, prior to the increased spring Chl *a* inventory, $\Phi(\text{O}_2)$ showed a few days of influx into seawater caused by a decrease of θ from 7.6 to 5.9°C that increased $C_{\text{sat}}(\text{O}_2)$. The influx at the beginning of the deployment is partly due to the $\Delta_{\text{bub}}(\text{O}_2)$ correction that resulted in $[1 + \Delta_{\text{bub}}(\text{O}_2)]c_{\text{sat}}(\text{O}_2) > c(\text{O}_2)$ for $U > 10 \text{ m s}^{-1}$. In August, the surface supersaturation decreased to $2.3 \mu\text{mol kg}^{-1}$ and $\Phi(\text{O}_2)$ decreased to a monthly minimum of $-7.6 \text{ mmol m}^{-2} \text{ d}^{-1}$. In the second half of September, the surface water became undersaturated by $-2.6 \mu\text{mol kg}^{-1}$, causing O_2 uptake with a median flux of $-13 \text{ mmol m}^{-2} \text{ d}^{-1}$ (5th percentile: $-39 \text{ mmol m}^{-2} \text{ d}^{-1}$; 95th percentile: $10 \text{ mmol m}^{-2} \text{ d}^{-1}$).

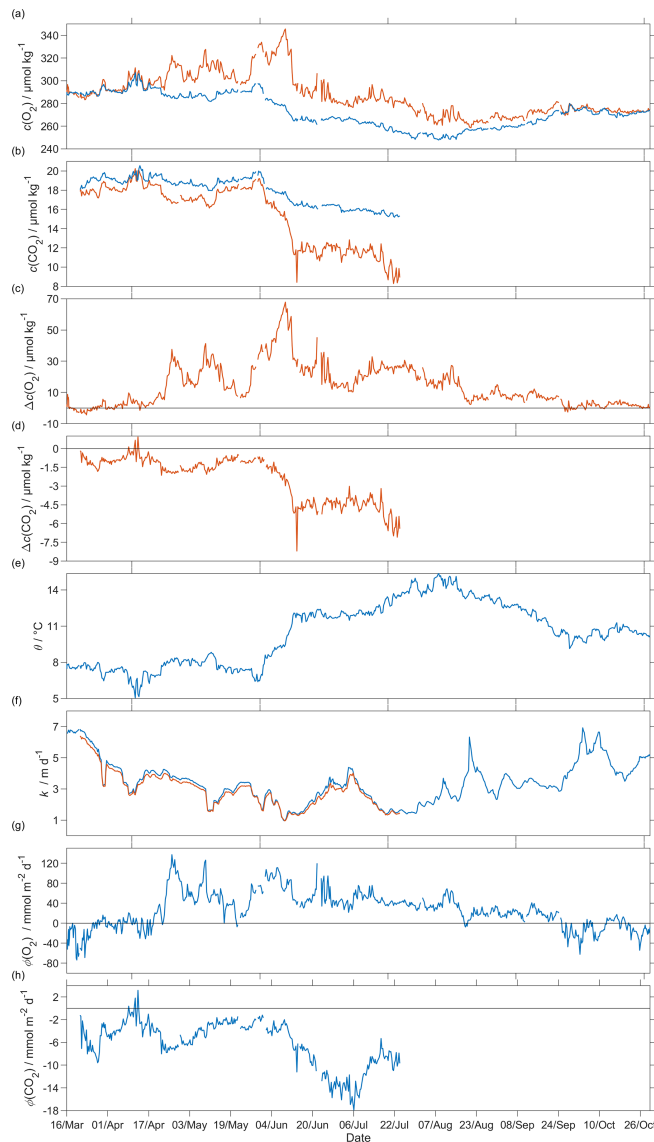


Figure 12. Air–sea flux of O_2 and CO_2 during spring and summer for CO_2 and during spring, summer and autumn for O_2 , **(a)** $c_{\text{sat}}(\text{O}_2)$ in blue and $c(\text{O}_2)$ in red, **(b)** $c_{\text{sat}}(\text{CO}_2)$ in blue and $c(\text{CO}_2)$ in red, **(c)** $\Delta c(\text{O}_2) = c(\text{O}_2) - c_{\text{sat}}(\text{O}_2)$, **(d)** $\Delta c(\text{CO}_2) = c(\text{CO}_2) - c_{\text{sat}}(\text{CO}_2)$, **(e)** sea surface temperature, **(f)** $k_w(\text{O}_2)$ in blue and $k_w(\text{CO}_2)$ in red normalised back to 50 d (Reuer et al., 2007), **(g)** oxygen air–sea flux $\Phi(\text{O}_2)$ and **(h)** CO_2 air–sea flux $\Phi(\text{CO}_2)$. The flux from sea to air is positive, while that from air to sea is negative.

The CO_2 flux from March to July was always from the air to the sea (Fig. 12), with a median of $-5.2 \text{ mmol m}^{-2} \text{ d}^{-1}$ (5th percentile: $-14 \text{ mmol m}^{-2} \text{ d}^{-1}$; 95th percentile: $-1.5 \text{ mmol m}^{-2} \text{ d}^{-1}$). An opposite flux direction is expected for $\Phi(\text{O}_2)$ and $\Phi(\text{CO}_2)$ during the productive season when net community production is the main driver of concentration changes. After the summer period of increased Chl *a* inventory, the flux had a median

of $-11 \text{ mmol m}^{-2} \text{ d}^{-1}$ (5th percentile: $-16 \text{ mmol m}^{-2} \text{ d}^{-1}$; 95th percentile: $-6.8 \text{ mmol m}^{-2} \text{ d}^{-1}$), in agreement with previous studies that classified the Norwegian Sea as a CO₂ sink (Takahashi et al., 2002; Skjelvan et al., 2005). $\Phi(\text{CO}_2)$ for the discrete samples from 18 March to 14 June ($n = 13$) varied from 0.1 to $-13 \text{ mmol m}^{-2} \text{ d}^{-1}$.

3.3 $N(\text{O}_2)$

To capture the entire euphotic zone, we calculated $N(\text{O}_2)$ and $N(\text{DIC})$ using an integration depth of $z_{\text{lim}} = 45 \text{ m}$ because the mean deep chlorophyll maximum (DCM) depth was $z_{\text{DCM}} = (20 \pm 18 \text{ m})$ (Fig. 9). For comparison, the mixed layer depth was deeper, varied more strongly and had a mean value of $z_{\text{mix}} = (68 \pm 78 \text{ m})$, using a threshold criterion of $\Delta\theta = 0.5^\circ \text{C}$ to the median θ value in the top 5 m of the glider profile (Obata et al., 1996; Monterey and Levitus, 1997; Foltz et al., 2003).

The two N values were calculated as the difference in inventory changes between two transects when the glider moved in the same direction.

During the deployment, we sampled two periods of increased Chl *a* inventory, the first one in May and a second one in June. The chlorophyll *a* inventory ($I_{\text{raw},z_{\text{lim}}}(\text{Chl } a)$) was calculated integrating $c_{\text{raw}}(\text{Chl } a)$ to z_{lim} . To remove outliers, we used a five-point moving mean of $I_{\text{raw},z_{\text{lim}}}(\text{Chl } a)$.

The $N(\text{O}_2)$ changes were dominated by $\Phi(\text{O}_2)$, which had an absolute median of $34 \text{ mmol m}^{-2} \text{ d}^{-1}$ (5th percentile: $4.3 \text{ mmol m}^{-2} \text{ d}^{-1}$; 95th percentile: $86 \text{ mmol m}^{-2} \text{ d}^{-1}$), followed by $I(\text{O}_2)$, which had a median of $15 \text{ mmol m}^{-2} \text{ d}^{-1}$ (5th percentile: $2.3 \text{ mmol m}^{-2} \text{ d}^{-1}$; 95th percentile: $29 \text{ mmol m}^{-2} \text{ d}^{-1}$), $F_V(\text{O}_2)$, which had an absolute median of $0.3 \text{ mmol m}^{-2} \text{ d}^{-1}$ (5th percentile: $0 \text{ mmol m}^{-2} \text{ d}^{-1}$; 95th percentile: $1.0 \text{ mmol m}^{-2} \text{ d}^{-1}$) and $E(\text{O}_2)$, which had a median of $0 \text{ mmol m}^{-2} \text{ d}^{-1}$ (5th percentile: $-1.2 \text{ mmol m}^{-2} \text{ d}^{-1}$; 95th percentile: $0 \text{ mmol m}^{-2} \text{ d}^{-1}$).

At the beginning of May, $I_{\text{raw},z_{\text{lim}}}(\text{Chl } a)$ increased to 97 mg m^{-2} and $N(\text{O}_2) = (95 \pm 16) \text{ mmol m}^{-2} \text{ d}^{-1}$. After this period, $I_{\text{raw},z_{\text{lim}}}(\text{Chl } a)$ decreased to 49 mg m^{-2} and $N(\text{O}_2) = (-4.6 \pm 1.6) \text{ mmol m}^{-2} \text{ d}^{-1}$. During the summer $I_{\text{raw},z_{\text{lim}}}(\text{Chl } a)$ increased to 110 mg m^{-2} , which caused a sharp increase of $N(\text{O}_2)$ to $(126 \pm 25) \text{ mmol m}^{-2} \text{ d}^{-1}$. $I_{\text{raw},z_{\text{lim}}}(\text{Chl } a)$ remained higher than 50 mg m^{-2} until the end of June when $N(\text{O}_2)$ was $(31 \pm 9) \text{ mmol m}^{-2} \text{ d}^{-1}$. The passage of the glider from NwAC to NCC was accompanied by a drop of surface $c(\text{O}_2)$ from 330 to $280 \mu\text{mol kg}^{-1}$ (Fig. 9) that resulted in lower $\Phi(\text{O}_2)$ and $N(\text{O}_2)$ values (Fig. 13). At the same time, $I_{\text{raw},z_{\text{lim}}}(\text{Chl } a)$ decreased to 35 mg m^{-2} showing that the decrease of $N(\text{O}_2)$ depended on the passage to NCC and a decrease of biological production. After the beginning of August, $I_{\text{raw},z_{\text{lim}}}(\text{Chl } a)$ decreased to 49 mg m^{-2} and $N(\text{O}_2)$ turned negative with a minimum of $(-23 \pm 25) \text{ mmol m}^{-2} \text{ d}^{-1}$. In October, during the last glider transect, $I_{\text{raw},z_{\text{lim}}}(\text{Chl } a)$ continued de-

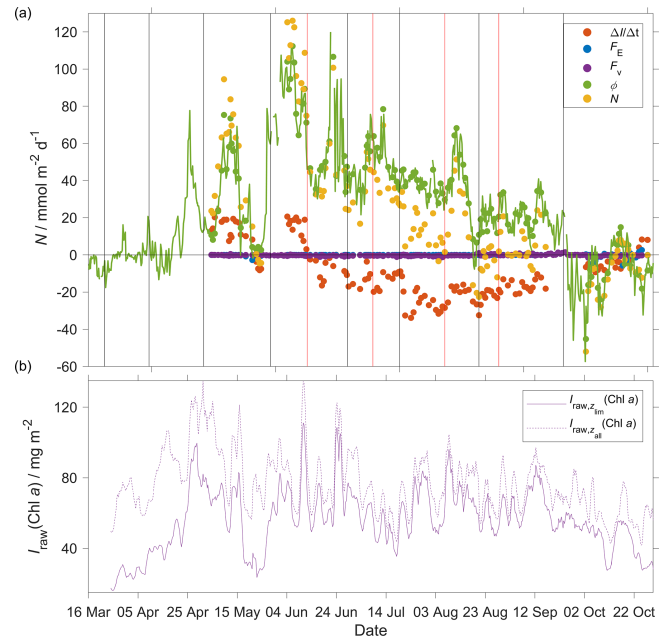


Figure 13. (a) Components of the $N(\text{O}_2)$ calculation: $\Delta I(\text{O}_2)/\Delta t$ (red), $E(\text{O}_2)$ (blue), $F_V(\text{O}_2)$ (violet), $\Phi(\text{O}_2)$ (green) with $k_w(\text{O}_2)$ weighted over 50 d, $N(\text{O}_2)$ (yellow). (b) Chl *a* inventory in the top 45 m, $I_{\text{raw},z_{\text{lim}}}(\text{Chl } a)$ (violet). Chl *a* inventory for the whole water column, $I_{\text{raw},z_{\text{all}}}(\text{Chl } a)$ (dotted violet line). The vertical black lines represent each glider transect. Between the two vertical red lines, the glider was in the NCC region.

creasing to 27 mg m^{-2} leading to the minimum $N(\text{O}_2)$ of $(-52 \pm 11) \text{ mmol m}^{-2} \text{ d}^{-1}$.

Integrating $N(\text{O}_2)$ from March to October gives a flux of $4.9 \text{ mol m}^{-2} \text{ a}^{-1}$ (Table 3; discussed in Sect. 4.2).

3.4 $N(\text{DIC})$

In the case of $N(\text{DIC})$, the main drivers were the inventory changes with an absolute median of $29 \text{ mmol m}^{-2} \text{ d}^{-1}$ (5th percentile: $1.3 \text{ mmol m}^{-2} \text{ d}^{-1}$; 95th percentile: $57 \text{ mmol m}^{-2} \text{ d}^{-1}$), followed by $\Phi(\text{CO}_2)$, which had an absolute median of $7.0 \text{ mmol m}^{-2} \text{ d}^{-1}$ (5th percentile: $0.8 \text{ mmol m}^{-2} \text{ d}^{-1}$; 95th percentile: $15 \text{ mmol m}^{-2} \text{ d}^{-1}$), $F_V(\text{DIC})$, which had an absolute median of $0.2 \text{ mmol m}^{-2} \text{ d}^{-1}$ (5th percentile: $0 \text{ mmol m}^{-2} \text{ d}^{-1}$; 95th percentile: $1.3 \text{ mmol m}^{-2} \text{ d}^{-1}$) and $E(\text{DIC})$, which had a median of $0 \text{ mmol m}^{-2} \text{ d}^{-1}$ (5th percentile: $0 \text{ mmol m}^{-2} \text{ d}^{-1}$; 95th percentile: $3.4 \text{ mmol m}^{-2} \text{ d}^{-1}$). During the period of increased Chl *a* inventory $N(\text{DIC})$ was $(21 \pm 4.5) \text{ mmol m}^{-2} \text{ d}^{-1}$. Later, $I_{\text{raw},z_{\text{lim}}}(\text{Chl } a)$ decreased to 30 mg m^{-2} driving $N(\text{DIC})$ to negative values with a minimum of $(-2.7 \pm 5.0) \text{ mmol m}^{-2} \text{ d}^{-1}$. In the next transect, the glider measured the maximum $I_{\text{raw},z_{\text{lim}}}(\text{Chl } a)$ of 111 mg m^{-2} that increased $N(\text{DIC})$ to $(85 \pm 4.5) \text{ mmol m}^{-2} \text{ d}^{-1}$. This maximum was reached during a transect when the glider moved in NCC, which had a

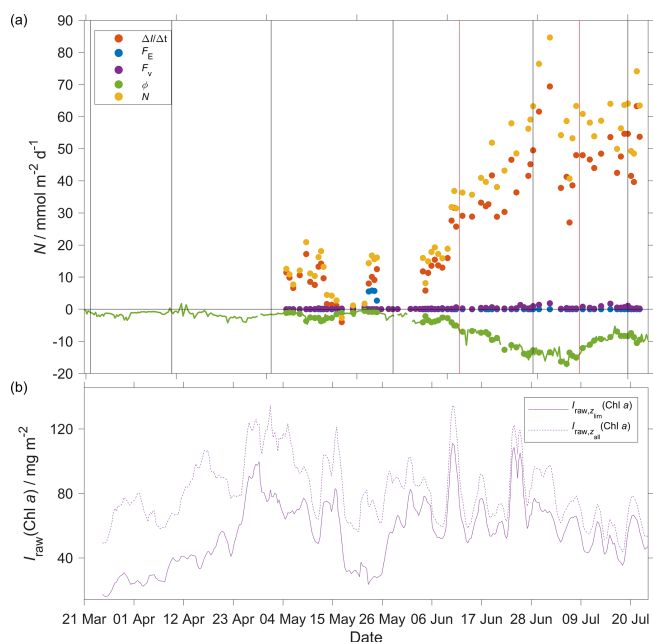


Figure 14. (a) Components of the $N(\text{DIC})$ calculation: $\Delta I(\text{DIC})/\Delta t$ (red), $E(\text{DIC})$ (blue), $F_V(\text{CO}_2)$ (violet), $\Phi(\text{CO}_2)$ (green) with $k_w(\text{CO}_2)$ weighted over 50 d, $N(\text{DIC})$ (yellow). (b) Chl *a* inventory in the top 45 m, $I_{\text{raw},z_{\text{lim}}}(\text{Chl } a)$ (violet). Chl *a* inventory for the whole water column, $I_{\text{raw},z_{\text{all}}}(\text{Chl } a)$ (dotted violet line). The vertical black lines represent each glider transect. Between the two vertical red lines, the glider was in the NCC region.

$c(\text{DIC})$ of $2080 \mu\text{mol kg}^{-1}$ at the surface compared with the $2150 \mu\text{mol kg}^{-1}$ in NwAC and drove a continuous positive $N(\text{DIC})$, which had a minimum of $(36 \pm 7.4) \text{mmol m}^{-2} \text{d}^{-1}$ (Fig. 14).

Integrating $N(\text{DIC})$ from March to July gives a flux of $3.3 \text{mol m}^{-2} \text{a}^{-1}$ (Table 3; discussed in Sect. 4.2).

4 Discussion

4.1 Sensor performance

This study presents data from the first glider deployment with a CO₂ optode. The initial uncalibrated $p_u(\text{CO}_2)$ measured by the CO₂ optode had a median of $604 \mu\text{atm}$ (5th percentile: $566 \mu\text{atm}$; 95th percentile: $768 \mu\text{atm}$), whereas the $p(\text{CO}_2)$ of discrete samples varied from 302 to $421 \mu\text{atm}$.

We applied corrections for drift (using deep-water samples as a reference point), sensor lag and calibrated the CO₂ optode against co-located discrete samples throughout the water column.

Atamanchuk et al. (2014) reported that the sensor was affected by a lag that varied from 45 to 264 s depending on temperature. These values were determined in an actively stirred beaker. However, in this study, the sensor was mounted on

a glider and was not actively pumped, which increased the response time to 23 min (25th quartile: 18 min; 75th quartile: 30 min). Also, the optode was affected by a continuous drift from 637 to $5500 \mu\text{atm}$ that was larger than the drift found by Atamanchuk et al. (2015a) which increased by $75 \mu\text{atm}$ after 7 months.

In this study, the drift- and lag-corrected sensor output showed a better correlation with the CO₂ concentration $c(\text{CO}_2)$ than with $p(\text{CO}_2)$. The latter two quantities are related to each other by the solubility that varies with θ and S (Weiss, 1974) (Eq. 2). The better correlation with $c(\text{CO}_2)$ was probably related to an inadequate temperature parameterisation of the sensor calibration function. Including both temperature and temperature squared in the calibration gave a better fit with $c(\text{CO}_2)$ than with $p(\text{CO}_2)$ but overall still a lower calibration residual for the former. The sensor output depends on the changes in pH that are directly related to the changes of $c(\text{CO}_2)$ in the membrane and – indirectly – $p(\text{CO}_2)$, via Henry's law. The calibration is supposed to correct for the temperature dependence of the sensor output (Atamanchuk et al., 2014). So the fact that the sensor output correlated better with $c(\text{CO}_2)$ than $p(\text{CO}_2)$ is perhaps due to a fortuitous cancellation of an inadequate temperature parameterisation and Henry's law relationship between $c(\text{CO}_2)$ and $p(\text{CO}_2)$.

The calibrated optode output captured the $c(\text{DIC})$ changes in space and time with a standard deviation of $11 \mu\text{mol kg}^{-1}$ compared with the discrete samples. $c(\text{DIC})$ decreased from 2130 to $2000 \mu\text{mol kg}^{-1}$ and increased with depth to $2170 \mu\text{mol kg}^{-1}$. This shows the potential of the sensor for future studies that aim to analyse the carbon cycle using a high-resolution dataset.

The optode-derived CO₂ fugacity $f_G(\text{CO}_2)$ had a mean bias of $(1.8 \pm 22) \mu\text{atm}$ compared with the discrete samples. These values are comparable with a previous study when the CO₂ optode was tested for 65 d on a wave-powered profiling cRAWLER (PRAWLER) from 3 to 80 m (Chu et al., 2020), which had an uncertainty between 35 and $72 \mu\text{atm}$. The PRAWLER optode was affected by a continuous drift of $5.5 \mu\text{atm d}^{-1}$ corrected using a regional empirical algorithm that uses $c(\text{O}_2)$, θ , S and σ_θ to estimate A_T and $c(\text{DIC})$.

4.2 Norwegian Sea net community production

Increases in $N(\text{O}_2)$ and $N(\text{DIC})$ were associated with increases in depth-integrated $c_{\text{raw}}(\text{Chl } a)$, designated as periods of increased Chl *a* inventory $I_{\text{raw}}(\text{Chl } a)$, at the beginning of May and in June. During May, $I_{\text{raw}}(\text{Chl } a)$ reached 135mg m^{-2} . In June, $I_{\text{raw}}(\text{Chl } a)$ reached again 135mg m^{-2} . Between these two periods, $N(\text{DIC})$ briefly turned negative, indicating remineralisation of the high Chl *a* inventory material during this period. The period of increased Chl *a* inventory coincided with a surface temperature increase from 7 to 11°C and shoaling of the mixed layer from 200 to 20m . $c(\text{O}_2)$ reached a summer maximum of $340 \mu\text{mol kg}^{-1}$ and

Table 3. Net community production (N) estimates in the Norwegian Sea (with integration depth z_{lim}). Falck and Anderson (2005) used year-round data from 1960 to 2000 between 62 and 70° N and from 1991 to 1994 at OWSM. Skjelvan et al. (2001) used year-round data from 1957 to 1970 and from 1991 to 1998 between 67.5° N, 9° E, and 71.5° N, 1° E, and along 74.5° N from 7 to 15° E. Kivimäe (2007) used year-round data from 1955 to 2005, and Falck and Gade (1999) used year-round data from 1955 to 1988 in all of the Norwegian Sea. While the previous studies report annual N estimates, the present study derives $N(\text{O}_2)$ between March and October and $N(\text{DIC})$ between March and July.

Study	$N(\text{DIC})/$ $\text{mol m}^{-2} \text{a}^{-1}$	$N(\text{O}_2)/$ $\text{mol m}^{-2} \text{a}^{-1}$	$N(\text{O}_2)/$ $N(\text{DIC})$	$z_{\text{lim}}/$ m	Variables used derive N
Falck and Anderson (2005, annual)	3.4	–		100	$c(\text{NO}_3^-)$, $c(\text{PO}_4^{3-})$, $c(\text{DIC})$
Skjelvan et al. (2001, annual)	–	2.6		300	$c(\text{O}_2)$, $c(\text{PO}_4^{3-})$
Kivimäe (2007, annual)	–	11 (4.7 to 18.3)		z_{mix} until 100 m	$c(\text{O}_2)$
Falck and Gade (1999, annual)	–	3.9		30	$c(\text{O}_2)$
This study, March to July	3.1	4.1	1.3	30	$c(\text{O}_2)$, $c(\text{DIC})$
This study, March to July	3.3	4.2	1.3	45	$c(\text{O}_2)$, $c(\text{DIC})$
This study, March to July	3.3	3.7	1.1	100	$c(\text{O}_2)$, $c(\text{DIC})$
This study, March to October		5.0			
This study, March to October		4.9			
This study, March to October		3.6			

$c(\text{DIC})$ decreased to a summer minimum of 1990 $\mu\text{mol kg}^{-1}$. In both cases, the main components of the N changes were the inventory and air–sea flux, while the smallest driver was the entrainment. Also, the glider sampled two different water masses characterised by different $c(\text{DIC})$ and $c(\text{O}_2)$. This might be the cause of the smaller values $N(\text{O}_2)$ and higher values $N(\text{DIC})$ in June and July in NCC compared to NwAC (Figs. 13 and 14). Another explanation might be a consumption of O₂ due to remineralisation and a delay in the response of the $c(\text{DIC})$ that was lowered during the two blooms. A fully functional CO₂ optode in the second part of the deployment would have helped to uncover the cause of the higher $N(\text{DIC})$ than of $N(\text{O}_2)$.

Table 3 shows estimates of net community production (N) in the Norwegian Sea. All other studies used ships to gather observations. The estimated N in the four other studies varied from 2.6 to 11.1 $\text{mol m}^{-2} \text{a}^{-1}$ for $N(\text{O}_2)$ and was 3.4 for $N(\text{DIC})$. In our glider study, we obtained between March and July a $N(\text{DIC})$ of 3.3 $\text{mol m}^{-2} \text{a}^{-1}$ and a $N(\text{O}_2)$ of 4.2 $\text{mol m}^{-2} \text{a}^{-1}$, in agreement with these studies. The ratio of $N(\text{O}_2)$ and $N(\text{DIC})$ for an integration depth of 45 m gave a photosynthetic quotient (PQ) of 1.3, in agreement with the Redfield ratio of 1.45 ± 0.15 (Redfield, 1963; Anderson, 1995; Anderson and Sarmiento, 1994; Laws, 1991). The $N(\text{O}_2)$ estimate is influenced primarily by the air–sea exchange flux $\Phi(\text{O}_2)$ (median: 34 $\text{mmol m}^{-2} \text{d}^{-1}$), followed by the inventory change (15 $\text{mmol m}^{-2} \text{d}^{-1}$). In contrast, $N(\text{DIC})$ is dominated by the inventory change (–29 $\text{mmol m}^{-2} \text{d}^{-1}$), followed by $\Phi(\text{CO}_2)$ (–7.0 $\text{mmol m}^{-2} \text{d}^{-1}$). This reflects the slower gas-exchange time constant of CO₂ compared with O₂, due to DIC buffering. To compare our results with previous studies, we also used $z_{\text{lim}} = 30$ m (Falck and Gade, 1999) and 100 m (Falck and Anderson, 2005; Kivimäe, 2007). The cal-

culated $N(\text{DIC}; 30 \text{ m})$ was 3.1 $\text{mol m}^{-2} \text{a}^{-1}$, $N(\text{DIC}; 100 \text{ m})$ was 3.4 $\text{mol m}^{-2} \text{a}^{-1}$, $N(\text{O}_2; 30 \text{ m})$ was 4.1 $\text{mol m}^{-2} \text{a}^{-1}$ and $N(\text{O}_2; 100 \text{ m})$ was 3.7 $\text{mol m}^{-2} \text{a}^{-1}$. The $N(\text{DIC}; 100 \text{ m})$ value is in agreement with the value of 3.4 $\text{mol m}^{-2} \text{a}^{-1}$ given by Falck and Anderson (2005). However, the latter estimate was for the entire year, whereas our estimate only covers the months from March to July. $N(\text{O}_2)$ was similar for $z_{\text{lim}} = 30$ and 45 m but lower for $z_{\text{lim}} = 100$ m because of O₂ consumption during organic matter remineralisation below the euphotic zone. The PQ value at 30 was 1.3 and at 100 m decreased to 1.1. Extending $N(\text{O}_2)$ to October increased $N(\text{O}_2; 30 \text{ m})$ and $N(\text{O}_2; 45 \text{ m})$ to 5.0 and 4.9 $\text{mol m}^{-2} \text{a}^{-1}$, respectively. Instead, $N(\text{O}_2; 100 \text{ m})$ decreased to 3.6 $\text{mol m}^{-2} \text{a}^{-1}$, confirming the consumption of O₂ below the euphotic zone. The calculated $N(\text{O}_2)$ until October was in agreement with the previous studies that varied between 2.6 and 11 $\text{mol m}^{-2} \text{a}^{-1}$.

Some of the previous $N(\text{DIC})$ estimates derived $c(\text{DIC})$ from other variables such as $c(\text{O}_2)$, $c(\text{PO}_4^{3-})$, $c(\text{NO}_3^-)$, assuming Redfield ratios P : N : C : O₂ 1 : 16 : 106 : –138 (Redfield, 1963). During photosynthesis $c(\text{PO}_4^{3-})$ and $c(\text{NO}_3^-)$ are taken up by phytoplankton to form organic matter and are released again after remineralisation of the organic matter giving an indication of NCP changes. Our $N(\text{DIC})$ estimate was 3.3 $\text{mol m}^{-2} \text{a}^{-1}$ and is similar to 3.4 $\text{mol m}^{-2} \text{a}^{-1}$ estimated by Falck and Anderson (2005) who used $c(\text{DIC})$ samples directly. The carbon/nutrient ratios vary between water masses and during photosynthesis (Thomas et al., 1999; Copin-Montégut, 2000; Osterroht and Thomas, 2000; Körtzinger et al., 2001).

The difference of the annual $N(\text{O}_2)$ and $N(\text{DIC})$ with the previous studies can also be caused by the yearly variability of N in the Norwegian Sea. In fact, Kivimäe (2007) saw an annual variability of $N(\text{O}_2)$ from 1955 to 2005 between 4.7

and $18.3 \text{ mol m}^{-2} \text{ a}^{-1}$. In order to understand what is causing these interannual changes, it is important capture inventory and air–sea changes. Also, this study showed that the Norwegian Sea spring, summer and autumn N is strongly affected by the time of sampling. For that reason, N estimated from low-resolution datasets makes the result strongly dependent on the time of sampling. To quantify this interannual variability in N , more high-resolution studies are needed.

5 Conclusions

To the best of our knowledge, this study represents the first glider deployment of a CO₂ optode. The CO₂ optode together with a O₂ optode shows the potential of using these sensors on autonomous observing platforms like Seagliders to quantify the interactions between biogeochemical processes and the marine carbonate system at high spatiotemporal resolution. The deployment helped to uncover NCP and air–sea flux variability over a period of 8 months.

Despite all the problems (drift, lag and poor calibration), the CO₂ optode data could be used to quantify dissolved inorganic carbon concentration variations. The temporal sampling resolution was 106 s in the top 100 m (increasing to 381 s from 500 to 1000 m). This could be improved to less than 10 s, but this would reduce the length of the deployment due to the limited glider battery capacity. With better calibration and stability improvements, the CO₂ optode could be routinely used to measure the carbonate system on gliders, floats and surface vehicles. Glider deployments up to 8 months are possible thanks to the sensor's low power consumption of 8 mW at 5 s sampling intervals and 7 mW at 60 s sampling intervals (Atamanchuk et al., 2014). Combined with other novel sensors that measure another DIC-related quantity such as A_T or $c(\text{DIC})$, CO₂ optodes on gliders could help provide estimates of NCP, air–sea flux, respiration and remineralisation and aragonite saturation.

During our deployment, we calculated O₂ and DIC-based NCP over the spring and summer period. In the future, extended deployments could be used to estimate annual (full-year) NCP. To have an accurate estimate of annual NCP, at least one additional glider deployment is needed to have continuous coverage (Binetti et al., 2020). Similar deployments can be used in other areas of the globe to fill gaps in $N(\text{DIC})$ and $N(\text{O}_2)$. In particular, glider deployments have potential in undersampled areas of the globe such as the Southern Ocean and the Arctic. Also, they can be used in well-studied areas such as North Sea and Mediterranean Sea to reduce monitoring costs and compare NCP estimates with previous studies that used other sampling strategies.

Appendix A

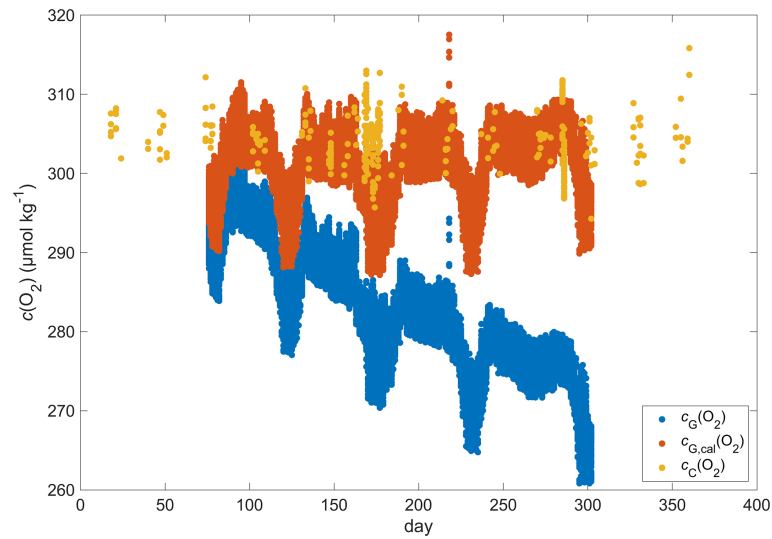


Figure A1. Discrete samples $c_C(\text{O}_2)$ (yellow), raw glider oxygen $c_G(\text{O}_2)$ (blue) and drift-corrected glider oxygen $c_{G,\text{cal}}(\text{O}_2)$ (red) for a potential density $> 1028 \text{ kg m}^{-3}$ at depths less than 1000 m.

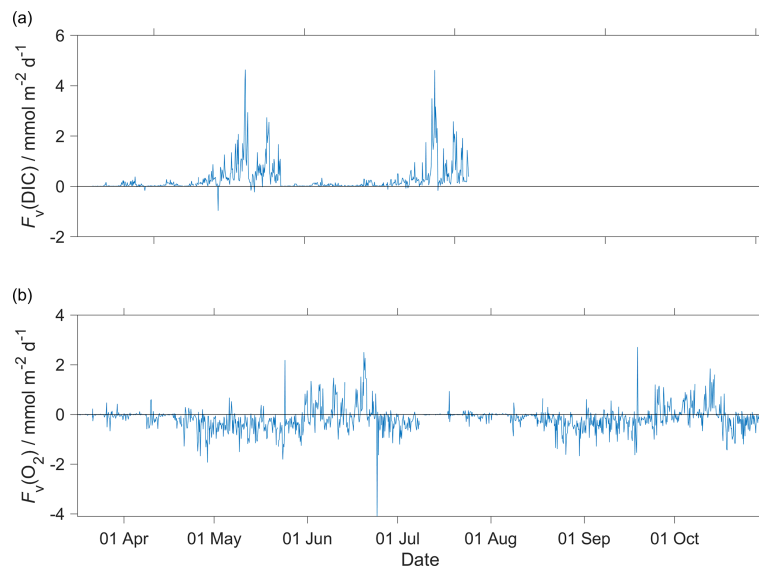


Figure A2. Diapycnal mixing (F_v) calculated for the glider descent and ascent for (a) $c(\text{DIC})$ and (b) O_2 at the mixed layer depth (z_{mix}) when deeper than 45 m (z_{lim}) and at z_{lim} when z_{mix} was shallower than 45 m. In the calculations, we used a vertical eddy diffusivity (K_z) of $10^{-5} \text{ m}^2 \text{ s}^{-1}$ (Naveira Garabato et al., 2004).

Appendix B: List of symbols

Symbols	Quantity (unit)
A_T	total alkalinity ($\mu\text{mol kg}^{-1}$)
b	backscatter signal (engineering units)
c	amount content ($\mu\text{mol kg}^{-1}$)
C	amount concentration (mmol m^{-3})
Chl a	chlorophyll a
DIC	dissolved inorganic carbon
E	entrainment flux ($\text{mmol m}^{-2} \text{d}^{-1}$)
F_v	diapycnal eddy diffusion flux ($\text{mmol m}^{-2} \text{d}^{-1}$)
$f(\text{CO}_2)$	fugacity of CO ₂ (μatm)
I	inventory (mmol m^{-2})
K_z	diapycnal eddy diffusivity ($\text{m}^2 \text{s}^{-1}$)
N	net community production ($\text{mmol m}^{-2} \text{d}^{-1}$)
$p(\text{CO}_2)$	partial pressure of CO ₂ (μatm)
S	practical salinity (‰)
t	time (s)
U	wind speed (m s^{-1})
x	dry mole fraction (mol mol^{-1})
z_{DCM}	depth of the deep chlorophyll maximum (m)
z_{lim}	integration depth (m)
z_{mix}	mixed layer depth (m)
Φ	air–sea flux ($\text{mmol m}^{-2} \text{d}^{-1}$)
φ	CO ₂ optode CalPhase (°)
σ_0	potential density (kg m^{-3})
θ	Celsius temperature (°C)
τ	response time (s)

Data availability. The glider data are available from the Norwegian Marine Data Centre (NMDC) at <https://doi.org/10.21335/NMDC-1654657723> (Possenti and Castano Primo, 2020).

Author contributions. The CO₂ sensor development was led at the University of Gothenburg by DA and AT. Sensor implementation, sampling and analysis for dissolved inorganic carbon content and total alkalinity were led by IS at NORCE. The data were analysed by LP, JK and MPH. LP wrote the first draft manuscript. All authors contributed to further editing and writing.

Competing interests. The authors declare that they have no conflict of interest.

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