



# From small scale variability to mesoscale stability in surface

#### ocean pH: implications for air-sea CO2 equilibration 2

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- 10 Correspondence to: Louise Delaigue (louise.delaigue@nioz.nl)
- 11 Abstract. One important aspect of understanding ocean acidification is the nature and drivers of pH variability in
- 12 surface waters on smaller spatial (i.e., areas up to 100 km<sup>2</sup>) and temporal (i.e., days) scales. However, there has been
- 13 a lack of high-quality pH data at sufficiently high resolution. Here, we describe a simple optical system for
- 14 continuous high-resolution surface seawater pH measurements. The system includes a PyroScience pH optode
- 15 placed in a flow-through cell directly connected to the underway supply of a ship through which near-surface
- 16 seawater is constantly pumped. Seawater pH is measured at a rate of 2 to 4 measurements min-1 and is cross-
- 17 calibrated using discrete carbonate system observations (total alkalinity, dissolved inorganic carbon and nutrients).
- 18 This setup was used during two research cruises in different oceanographic conditions: the North Atlantic Ocean
- 19 (December-January 2020) and the South Pacific Ocean (February-April 2022). Our findings reveal fine-scale
- 20 fluctuations in surface seawater pH across the North Atlantic and South Pacific Oceans. While temperature is a
- 21 significant abiotic factor driving these variations, it does not account for all observed changes. Instead, our results
- 22 23 highlight the interplay between temperature, biological activity, and water masses on pH. Notably, the variability
- differed between the two regions, suggesting differences in the dominant factors influencing pH. In the South
- 24 25 26 Pacific, biological processes appeared to be mostly responsible for pH variability, while in the North Atlantic,
- additional abiotic and biotic factors complicated the correlation between expected and observed pH changes.
- Although surface seawater pH exhibited fine-scale variations, it remained relatively stable over a 24-hour cycle due
- $\overline{27}$ to reequilibration with atmospheric CO2. Thus, for the regions and time periods studied, ocean basin-scale analyses
- based on discrete samples collected during traditional research cruise transects would still capture the necessary
- 28 29 variability for global CO<sub>2</sub> cycle assessments and their implications.

#### 30 1 Introduction

- 31 Ocean chemistry is changing due to the uptake of anthropogenic CO<sub>2</sub> from the atmosphere
- 32 (DeVries, 2022). The uptake of atmospheric CO<sub>2</sub> by the ocean's surface causes a rise in the
- 33 ocean's hydrogen ion concentration, a process commonly referred to as ocean acidification. As a
- 34 result, the hydrogen ion concentration of surface ocean seawater has increased by 30-40%, which
- 35 corresponds to a pH decrease of approximately 0.1 since around 1850 (Gattuso et al., 2015; Jiang
- 36 et al., 2019; Orr et al., 2005). These changes have already significantly impacted marine
- 37 organisms, especially marine calcifiers (Doney et al., 2020; Gattuso et al., 2015; Osborne et al.,
- 38 2020), and pH is projected to decline by ~0.3 by 2100 (Figuerola et al., 2021).
- 40 To understand these impacts and the underlying mechanisms more precisely, high-resolution 41 studies have significantly advanced our understanding of the upper ocean's carbon cycle.
- 42 However, gaps remain, particularly at fine spatio-temporal scales (e.g., variability over hours and
- 43 a few kilometers). At these scales, changes in pH over short time periods can be an important
- 44 control on the ocean's buffering capacity and response to CO2 uptake, highlighting the need for





further detailed observations (Cornwall et al., 2013; Egilsdottir et al., 2013; James et al., 2020; Qu et al., 2017; Wei et al., 2022).

Advancements in the last decade and a half have enhanced the capacity for accurate and precise in-situ pH measurements. For example, Martz et al. (2010) developed an autonomous sensor tailored for continuous deployment in marine environments that allows recording of high-resolution pH fluctuations. Autonomous surface vehicles in coastal upwelling systems have also been used to capture intricate partial pressure of  $CO_2$  ( $pCO_2$ ) and pH dynamics, even in complex environments where rapid biogeochemical changes occur due to natural phenomena like upwelling (Chavez et al., 2018; Cryer et al., 2020; Possenti et al., 2021). Staudinger et al. (2018) also developed an optode system capable of simultaneously measuring oxygen, carbon dioxide, and pH in seawater. This system was designed for extended deployment (i.e. days) in marine environments, enabling continuous monitoring (with measurement intervals between 1 second and 1 hour) of these parameters. Additionally, Sutton et al. (2019) detailed the implementation of autonomous seawater  $pCO_2$  and pH time series from 40 surface buoys, broadening the scope of observations at fixed time series sites. Staudinger et al. (2019) introduced fast and stable optical pH sensor materials specifically for oceanographic applications, enhancing the ability to measure pH under various environmental conditions.

These technological advancements have facilitated significant scientific progress. Field measurements conducted using submersible spectrophotometric sensors have revealed fine-scale variations in pH in coastal waters and shed light on localized acidification processes (Cornwall et al., 2013). The implementation of autonomous seawater  $pCO_2$  and pH time series as described by Sutton et al. (2019) has enhanced our ability to characterize sub-seasonal variability in the ocean. These efforts represent important progress that can be built upon to further understand fine-scale ocean dynamics.

Fine spatio-temporal scale variability in surface ocean pH is hard to capture because it is driven by a complex interplay of processes, including physical mixing, biological activity (i.e. photosynthesis and respiration), thermal variability, and air-sea CO<sub>2</sub> fluxes (Faassen et al., 2023; Hofmann et al., 2011; Price et al., 2012). Physical mixing moderates surface oceanic pH by redistributing dissolved CO<sub>2</sub>, nutrients, and heat throughout the water column. Mixing also mitigates extreme pH fluctuations by diluting surface concentrations of CO<sub>2</sub> during periods of high biological activity or temperature-induced CO<sub>2</sub> release (Egea et al., 2018; Li et al., 2019). Photosynthetic activity can decrease CO<sub>2</sub>, leading to an increase in pH during daylight hours, while respiration dominates at night, releasing CO<sub>2</sub> and lowering pH (Fujii et al., 2021; Jokiel et al., 2014). Warmer waters decrease CO<sub>2</sub> solubility and increase pH, while cooler waters increase solubility, promoting CO<sub>2</sub> uptake and decreasing pH, although the timescale of these processes differs, with some changes occurring instantaneously and others after equilibration (Zeebe & Wolf-Gladrow, 2001). Instantaneous changes are driven by physical and chemical reactions, while equilibration processes involve longer-term adjustments such as air-sea gas exchange and the mixing of surface waters with deeper layers (Emerson & Hedges, 2008). Increased atmospheric CO<sub>2</sub> boosts oceanic CO<sub>2</sub> uptake, lowering pH, while reduced atmospheric CO<sub>2</sub> would decrease oceanic CO<sub>2</sub>, raising pH (Caldeira & Wickett, 2005; Orr et al., 2005). Although each of these processes has its distinct impact on pH, their combined effects regulate the ocean's carbon cycle and its interaction with the atmosphere.





Recent studies on air-sea CO2 equilibration timescales have highlighted significant regional variations, particularly between the North Atlantic and South Pacific Oceans (Jones et al., 2014). In the North Atlantic, deeper mixed layers and unstable oceanic conditions lead to longer equilibration timescales compared to temperate or tropical regions, ranging from 3 to 6 months for the average North Atlantic latitude, and up to 18 months for regions above 55 °N (Jones et al., 2014). Cooler temperatures at higher latitudes increase CO<sub>2</sub> solubility, resulting in higher dissolved inorganic carbon (DIC), and upwelling brings DIC and total alkalinity (TA)-rich deep waters to the surface (Wu et al., 2019). These factors further increase the amount of CO<sub>2</sub> that needs to equilibrate with the atmosphere, further prolonging equilibration times. The South 

Pacific, with its shallower mixed layers and higher average temperatures, facilitates shorter equilibration times and enhances CO<sub>2</sub> uptake rates (i.e. 3 to 4 months; Jones et al., 2014). Wu et al. (2019) also showed that high biological productivity in these areas significantly impacts DIC,

potentially reducing surface DIC more quickly.

Here, we investigate how surface seawater pH varies on fine spatio-temporal scales (i.e areas up to 100 km², and days) across different ocean basins (i.e., North Atlantic and South Pacific Oceans) and identify abiotic and biotic factors driving these variations. We use direct, high-frequency measurements of surface seawater pH and estimate TA to resolve the rest of the carbonate system. These data are used to analyse how key biogeochemical processes, such as temperature, hydrodynamic mixing and biological activity, influence fine-scale spatio-temporal variability in pH.

# 2 Materials and Procedures

# **2.1 Study** areas

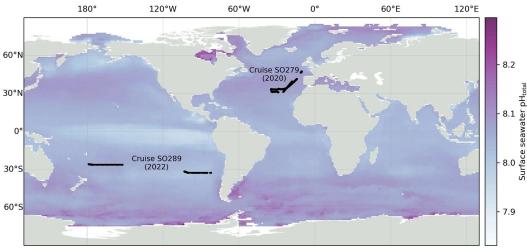


Figure 1. Locations of pH measurements during two oceanographic cruises used in this study: SO279 in the North Atlantic (December 2020) and SO289 in the South Pacific (February-April 2022). Surface seawater pH<sub>total</sub> for December 2022 from the OceanSODA product is shown in the background (Gregor & Gruber, 2020).

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- 121 Two datasets from separate oceanic regions were used: research expeditions SO279 in the North
- Atlantic Ocean and SO289 in the South Pacific Ocean, both on the German R/V Sonne (Fig. 1). 122

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- 124 Expedition SO279 in December 2020, conducted in the Azores Region of the North Atlantic
- 125 Ocean (Fig. 1), was part of the NAPTRAM research program investigating the transport
- 126 pathways of plastic and microplastic debris (Beck et al., 2021). The data collection included
- 127 discrete samples from the CTD rosette (n=77; Delaigue et al., 2021a), with measurements of
- 128 DIC, TA and nutrients (silicate, phosphate ammonium, nitrite, and nitrate + nitrite); discrete
- 129 samples from the underway water system (UWS; n=51; Delaigue et al., 2021b) also measuring
- 130 these parameters; and a high-resolution UWS timeseries of ocean surface pH (over 43,000
- 131 datapoints; Delaigue et al., 2021c).

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- 133 The South Pacific GEOTRACES Cruise SO289, from Valparaiso (Chile) to Noumea (New
- 134 Caledonia) under the GEOTRACES GP21 initiative, was conducted between 18 February and 08
- 135 April, 2022 (Fig. 1; Achterberg et al., 2022). The data collection also included discrete carbonate
- chemistry and nutrient samples from the CTD rosette (n=395; Delaigue, Ourradi, Ossebar, et al., 136
- 137 2023), discrete samples from the UWS (n=32; Delaigue, Ourradi, et al., 2023a) and another high-
- 138 resolution UWS timeseries of ocean surface pH from the optode system (over 78,000 datapoints;
- 139 Delaigue, Ourradi, et al., 2023b).

#### 140 2.2 Integrated Shipboard Optode System for Continuous pH Measurements

- 141 We used a pH optode (PHROBSC-PK8T, for pH range 7.0 – 9.0 on the total scale; PyroScience
- 142 GmbH), made of a robust cap adapter fiber with a stainless-steel tip (length=10cm,
- 143 diameter=4mm) and disposable plastic screw cap with an integrated pH sensor. The optode was
- 144 connected to a meter combined with a pressure-stable optical connector (optical feed-through;
- 145 OEM module Pico-pH-SUB; PyroScience GmbH; Fig. 2). Briefly, the optical pH sensor is
- 146 constructed using the PyroScience REDFLASH technology, which uses a pH neutral reference
- 147 indicator and a pH responsive luminescent dye. These elements are activated using a specifically
- 148 tuned orange-red light with a wavelength ranging between 610-630 nm, which triggers a bright
- 149 luminescence emission in the near-infrared (NIR) band, spanning from 760-790 nm. At elevated
- 150 pH, the fluorescence from the pH marker is diminished, leaving only the NIR emission of the
- 151 reference indicator noticeable. As the acidity increases, the pH marker is protonated, which
- 152 results in a heightened NIR luminescence that is detected along with the emissions of both
- 153 indicators. The measurement approach uses red excitation light modulated in a sinusoidal
- 154 manner, leading to a similar modulation of the NIR emission, albeit with a phase discrepancy.
- 155 This phase variation is registered by the PyroScience OEM module and subsequently converted into a total pH measurement. 156

- 158 Automatic temperature compensation of the optical pH sensor was achieved using a flexible
- 159 Teflon-coated temperature probe (Pt100 Temperature Probe, PyroScience GmbH; Fig. 2)
- soldered onto the OEM module. The optode was placed in a closed flow-through cell directly 160
- 161 connected to the underway supply of the ship through which seawater was pumped at a constant
- 162 rate (6 L/min for SO279 and 9 L/min for SO289) and stirred using a magnetic stirrer (Fig. 2).
- 163 The entire setup was kept inside a closed box to isolate the optical instrument from any other
- 164 light source (Fig. 2). All seawater first went through a thermo-salinograph close to the water



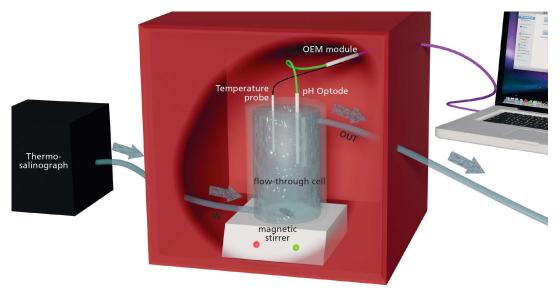


source, which also measured salinity, temperature and chlorophyll-a fluorescence (chl-a), before going through the pH setup (Fig. 2).

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Figure 2. Schematic representation of the optical continuous pH measurement system. Arrows indicate the direction of flow-through tubing. The system consists of the following components: a thermo-salinograph; a flow-through cell; a magnetic stirrer; a fiber-based pH optode; a flexible Teflon-coated temperature probe; a fiber-optic meter OEM module Pico-pH-SUB and a portable computer. All elements inside the red box are in the dark to avoid any light disturbance.

### 2.3 Initial calibration and underway measurements

Direct measurements of surface water (~3 m depth) pH were carried out at a frequency of 2 measurements min<sup>-1</sup> for the North Atlantic cruise and 4 measurements min<sup>-1</sup> for the South Pacific cruise.

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A one-point calibration of the temperature probe was performed against a thermometer inside a water bath (Lauda Ecoline RE106). A two-point calibration of the pH sensor was performed using PyroScience pH buffer capsules (pH 2 or pH 4 for the acidic calibration point, pH 10 or pH 11 for the basic calibration points). Buffers were prepared by dissolving each capsule's powder into 100 mL MilliQ water. A pH offset adjustment was made using certified reference material (CRM, batches #189, #195 and #198; provided by Andrew Dickson, Scripps Institution of Oceanography).

# 2.4 Discrete sampling and analysis for other CO<sub>2</sub> parameters

The underway seawater system was sub-sampled from the cell every 12 hours via silicone tubing for TA and DIC following an internationally established protocol (Dickson et al., 2007). TA was sampled in Azlon<sup>TM</sup> HDPE wide neck round 150 mL bottles filled to the neck and poisoned with 50 μL saturated HgCl<sub>2</sub>. DIC was sampled into Labco Exetainer® 12 ml borosilicate vials and





poisoned with 15  $\mu$ L saturated HgCl<sub>2</sub>. Samples were stored at 4 °C whenever possible and kept in the dark until analysis.

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All TA and DIC analysis was carried out at the Royal Netherlands Institute for Sea Research, Texel (NIOZ). The analysis was calibrated using certified reference material (CRM, batches #189, #195 and #198; provided by Andrew Dickson, Scripps Institution of Oceanography). TA was determined using a Versatile INstrument for the Determination of Total inorganic carbon and titration Alkalinity (VINDTA 3C #017 and #014, Marianda, Germany). The instrument performed an open-cell, potentiometric titration of a seawater subsample with 0.1 M hydrochloric acid (HCl). Results were then recalculated using a modified least-squares fitting as implemented by Calkulate v3.1.0 (Humphreys and Matthews, 2024). DIC concentrations were determined using either the VINDTA system (SO279 samples and part of SO289 samples) or the QuAAtro Gas Segmented Continuous Flow Analyzer (CFA, SEAL Analytical; SO289 samples). Briefly, the VINDTA measures DIC by acidifying a seawater sample, which releases CO<sub>2</sub> that is then quantified through a coulometric titration cell. Similarly, the QuAAtro CFA uses acidification to liberate CO<sub>2</sub>, which then discolours a slightly alkaline phenolphthalein pink coloured solution which is measured spectrophotometrically at 520 nm (Stoll et al., 2001).

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For cruise SO279, nutrient samples were gathered using 60 mL syringes made of high-density polyethylene, which were connected to a three-way valve by tubing, drawing directly from the CTD-rosette bottles to avoid air exposure. Immediately upon collection, the samples were taken to the laboratory for processing, where they were filtered through a dual-layer filter with pore sizes of 0.8 and 0.2 μm. All samples were stored at -20°C in a freezer except Si, which were stored at 4°C in a cold room until analysis back at NIOZ. Nutrients were analysed using a QuAAtro Continuous Flow Analyser. Measurements were made simultaneously on four channels together Si, PO<sub>4</sub>, NH<sub>4</sub>, NO<sub>3</sub>, and NO<sub>2</sub>. All measurements were calibrated with standards diluted in low nutrient seawater (LNSW) in the salinity range of the stations (approx. 34 - 37) to ensure that analysis remained within the same ionic strength. Prior to analysis, all samples were brought to laboratory temperature in about one to two hours. To avoid gas exchange and evaporation during the runs with NH<sub>4</sub> analysis, all vials including the calibration standards were covered with Parafilm under tension before being placed into the auto-sampler, so that the sharpened sample needle easily penetrated through the film leaving only a small hole. Silicate samples were measured separately on a TRAACS Gas Segmented Continuous Flow Analyser (manufactured by Bran+Lubbe, now SEAL Analytical) following Strickland and Parsons (1972). A sampler rate of 60 samples per hour was also used for all analyses. Calibration standards were diluted from stock solutions of the different nutrients in 0.2 µm filtered LNSW diluted with de-ionised water to obtain approximately the same salinity as the samples and were freshly prepared every day. This diluted LNSW was also used as the baseline water for the analysis and in between the samples. Each run of the system had a correlation coefficient of at least 0.9999 for 10 calibration points. The samples were measured from the lowest to the highest concentration, i.e., from surface to deep waters in order to reduce carry-over effects. Concentrations were recorded in µM at an average container temperature of 23.0 °C and later converted to \(\mu\mod \)/kg by dividing the recorded concentration by the sample density, calculated following Millero and Poisson (1981).

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For cruise SO289, nutrient analysis was carried out on seawater from every Niskin bottle triggered at various depths during each cast. The seawater was transferred into 15 mL

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238 polypropylene vials. Each container and its cap were rinsed three times with seawater before 239 filling. If immediate analysis was not possible, samples were stored in a fridge at 4 °C in the 240 dark. Macro nutrients were analyzed onboard using a segmented flow injection analysis with a 241 Seal Analytical QUAATRO39 auto-analyzer that includes an XY2-autosampler. For nanomolar 242 nutrient analysis, a modified setup with 1000 mm flow cells was employed. The setup was designed to analyze four channels: total oxidized nitrogen (TON), silicate, nitrite, and phosphate, 243 244 using methods outlined in QuAAtro Applications: Method Nos. Q-068-05 Rev. 11, Q-066-05 245 Rev. 5, Q-070-05 Rev. 6, and Q-064-05 Rev. 8, respectively. To ensure analytical consistency 246 and validate the data, each run was checked against Certified Reference Material for Nutrients in Seawater (RMNS). Nutrient analyses were further validated using KANSO CRM, with specific 247 248 lot numbers for macromolar and nanomolar nutrient concentrations.



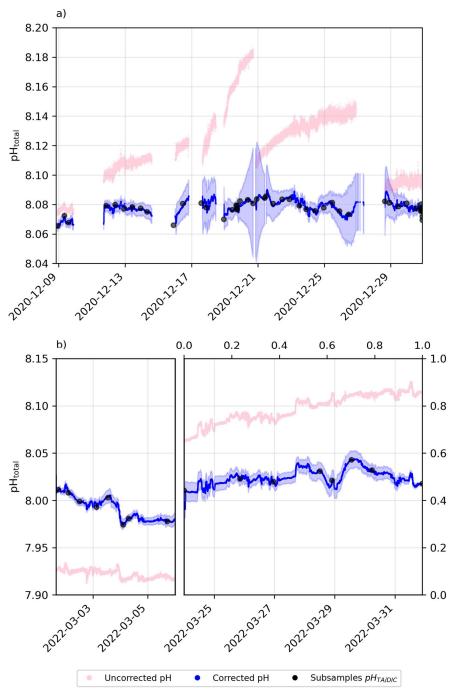


Figure 3. Underway pH drift correction using  $pH_{total(TA/DIC)}$  subsamples for a) cruise SO279 and b) cruise SO289. Raw pH measurements (pink); corrected pH values and uncertainty from bootstrapping (blue); subsample  $pH_{total(TA/DIC)}$  depicted as black circles.





#### 2.5 Post-cruise correction

- To ensure the reliability of the dataset, an initial screening was conducted to identify and flag
- 255 unreliable continuous pH data points, primarily attributable to optode stabilization issues.
- 256 Specifically, data points exhibiting unreasonable drift patterns, inconsistent with expected optode
- and surface ocean pH behavior, were flagged as unreliable.

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- 259 pH<sub>TA, DIC</sub> values were calculated from TA and DIC UWS discrete samples and were then aligned
- with the continuous pH data to calculate the offset between the  $pH_{TA, DIC}$  subsamples and the
- 261 continuous pH data (Fig. 3). A Piecewise Cubic Hermite Interpolating Polynomial (PCHIP;
- 262 (Fritsch & Carlson, 1980) was fitted to the offset as a way to continuously correct pH variations
- across the entire pH timeseries (Fig. 3).

# 2.6 Estimation of other biogeochemical parameters

- 265 TA was estimated for the continuous pH dataset using the empirical equations presented by Lee
- et al. (2006). For the North Atlantic, the corresponding equation was used (see Fig. S1 in
- 267 supplementary information):

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$$TA_{NA} = 2305 + 53.97(SSS - 35) + 2.74(SSS - 35)^{2} - 1.16(SST - 20) + 0.040(SST - 20)^{2}$$
(1)

- 269 In contrast, the (Sub)tropics equation (SBT, Eq. 2) was applied for the South Pacific region, as
- 270 the cruise mostly followed the 32.5°S longitude and this equation was determined to offer the
- best fit to the local temperature (SST) and salinity (SSS) equation (see Fig. S1 in supplementary
- 272 information):

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$$TA_{SBT} = 2305 + 58.66(SSS - 35) + 2.32(SSS - 35)^{2} - 1.41(SST - 20) + 0.040(SST - 20)^{2}$$
(2)

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- TA estimates were used together with pH to solve the rest of the marine carbonate system (i.e.
- 276 DIC and fCO<sub>2</sub>) using PyCO2SYS (Version 1.8.2; (Humphreys et al., 2022), with the carbonic
- acid dissociation constants of Sulpis et al. (2020), the bisulfate dissociation constant of Dickson
- 278 (1990), the total boron to chlorinity ratio of Uppström (1974), and the hydrogen fluoride
- dissociation constant of Dickson and Riley (1979).

# 280 2.7 Projected pH variability

- 281 The derived parameters pH<sub>temp</sub>, pH<sub>sal</sub>, and pH<sub>temp,sal</sub> were calculated while holding TA and DIC
- constant at the average values of each diurnal cycle. These calculations allowed for the
- 283 exploration of changes in pH under varying environmental conditions by solely altering either
- temperature, salinity, or both simultaneously. pH<sub>TA, fCO2</sub> was also computed while holding TA
- and fCO<sub>2</sub> constant at their daily mean values but with varying temperature and salinity to model
- the pH distribution under fully equilibrated conditions. All calculations were done using the
- same options in PyCO2SYS described in Sect. 2.6.



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# 2.8 Identification of full diurnal cycles

- To account for the influence of geographic location on temporal measurements, timestamps from Coordinated Universal Time (UTC) were converted to Local Solar Time (LST). This conversion was necessary to align time-sensitive data with the true solar position at each measurement
- location, thereby facilitating more accurate comparisons of environmental data across different geographic regions and improving the analysis of diurnal processes.

The conversion process involved calculating the mean longitudinal position for each date within the dataset. Subsequently, a time offset was determined based on the average longitude, assuming a standard rate of Earth's rotation. This offset was then applied to the original UTC timestamps, resulting in a modified dataset with timestamps adjusted to reflect LST:

 $DateTime_{LST} = DateTime_{UTC} + \left(\frac{Longitude_{mean}}{15}\right).$  (3)

- 300 Following the conversion to LST, the dataset was further processed to isolate complete diurnal
- 301 cycles to ensure that only data representing full 24-hour cycles were included. The North
- 302 Atlantic dataset included 7 complete diurnal cycles, while the South Pacific dataset included 11
- 303 complete diurnal cycles.

### 304 **2.9** Uncertainty propagation

# 2.9.1 Uncertainty in pH measurements and post-correction

- The final uncertainty in pH measurements was estimated by combining uncertainties from two main sources: (1) the uncertainty in the TA and DIC measurements used to calculate pH<sub>(TA/DIC)</sub> and (2) the correction of the UWS pH measurements using pH<sub>TA, DIC.</sub>
- First, the precision in TA and DIC were determined based on the RMSE from repeated
   measurements of a known standard water sample in the laboratory (NIOZ; 0.92 umol/kg and
   μmol/kg, respectively). Then a Monte Carlo simulation was applied to the calculated pH<sub>TA</sub>,
   plc to obtain a pH<sub>RMSE</sub> for each subsample pH<sub>TA, DIC</sub>.
- Next, the pH measurements obtained from the optode were corrected using pH<sub>(TA, DIC)</sub> from the discrete measurements of TA and DIC. The uncertainty in the pH correction was assessed using a bootstrapping approach (n=1000 iterations), where a fraction (50%) of the discrete samples was randomly omitted in each iteration and the selected fraction of the discrete samples varied within its own pH<sub>RMSE</sub>.
- 320
- The variation in each subsample's pH<sub>TA, DIC</sub> captured the likely variability in TA and DIC measurements, while omitting different subsets of data allowed for the estimation of how
- sensitive the pH correction is to which set of subsamples are used for calibration.

## 324 **2.9.2** Uncertainty in pH diurnal patterns

- To ensure the observed patterns in pH over the full 24-hour cycles were not artifacts of sampling
- bias or other anomalies, a Monte Carlo simulation (n = 1000 iterations) was employed on each
- diurnal cycle's hourly data analysis. This simulation randomly selected 50% of the data points for





- 328 each hour (i.e. 50% of 120 hourly measurements for the North Atlantic dataset and 50% of 240
- 329 hourly measurements for the South Pacific dataset), repeatedly calculating the mean pH to assess
- 330 the consistency and robustness of the hourly trends. This component of the uncertainty proved
- insignificant (i.e., errors bars were smaller than the symbols in Fig. 4 and 5).

# 2.10 CO<sub>2</sub> air-sea flux dynamics

- 333 The difference in pCO<sub>2</sub> between the air and ocean surface dictates the direction of gas exchange
- and determines whether the ocean acts as a CO<sub>2</sub> source or sink to the atmosphere. This CO<sub>2</sub>
- exchange between the air and sea, referred to as CO<sub>2</sub> flux, can be computed based on the
- 336 relationship:

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$$F = k_{\rm w} \times K_0 \times (p{\rm CO}_{2\rm sw} - p{\rm CO}_{2\rm sir}) \tag{4}$$

- where F represents the flux of  $CO_2$  across the air-sea interface,  $k_w$  is the gas transfer velocity,
- $K_0$  is the solubility constant and  $pCO_{2_{SW}}$  is the partial pressure of  $CO_2$  in sea water, representing
- 341 the concentration of dissolved  $CO_2$  that is in equilibrium with the atmosphere and  $pCO_{2air}$  is the
- partial pressure of CO<sub>2</sub> in the atmosphere above the ocean surface. For the fluxes, a positive
- value shows the ocean acts as a source (i.e., releasing CO<sub>2</sub> to the atmosphere), while a negative
- value shows it acts as a sink (i.e., absorbing CO<sub>2</sub> from the atmosphere). The parameterisation
- from Ho et al. (2006) was used to determine the gas transfer velocity. All fluxes were computed
- using the pySeaFlux package (v2.2.2; (Fay et al., 2021).
- 347
- Flux calculations were performed for each complete diurnal cycle, followed by the computation
- of the mean flux for each cycle. Additionally, the mean flux for the diurnal cycle was also
- 350 calculated from the daily mean inputs (wind speed, temperature, salinity, and  $pCO_2$ ), and
- 351 specifically computed for the hours 12 AM and 12 PM (LST) to examine temporal variations
- within each cycle.





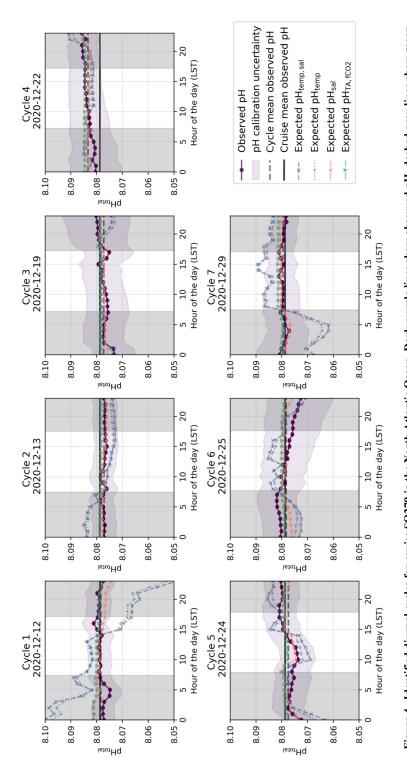
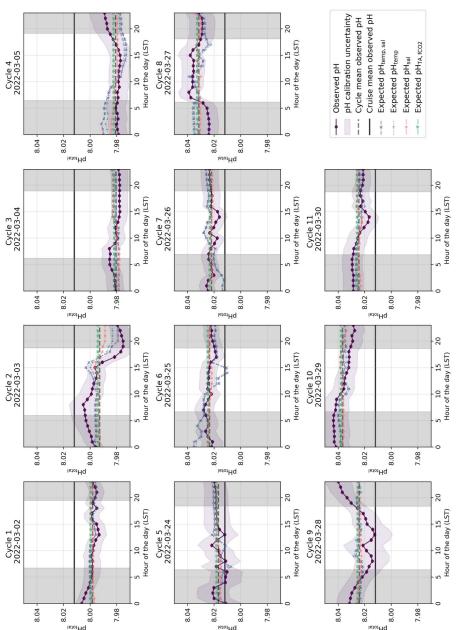


Figure 4. Identified diurnal cycles for cruise SO279 in the North Atlantic Ocean. Dark purple lines show observed pH, dashed grey lines show mean observed pH over the full diurnal cycle, and black lines show overall mean pH for all diurnal cycles combined for that cruise. The remaining shows lines), varying temperature alone (dashed blue lines), varying salinity alone (dashed orange lines) and constant TA and fCO2 (dashed green lines). expected pH using constant TA and DIC (averaged for each respective diurnal cycle) and varying temperature and salinity (dashed light purple Grey areas are night hours.







cycle) and varying temperature and salinity (dashed light purple lines), varying temperature alone (dashed blue lines), varying salinity alone (dashed orange lines) and constant TA and fCO2 (dashed green lines). Grey areas are night hours. combined for that cruise. The remaining shows expected pH using constant TA and DIC (averaged for each respective diurnal Figure 5. Identified diurnal cycles for cruise SO289 in the South Pacific Ocean. Dark purple lines show observed pH, dashed grey lines show mean observed pH over the full diurnal cycle, and black lines show overall mean pH for all diurnal cycles



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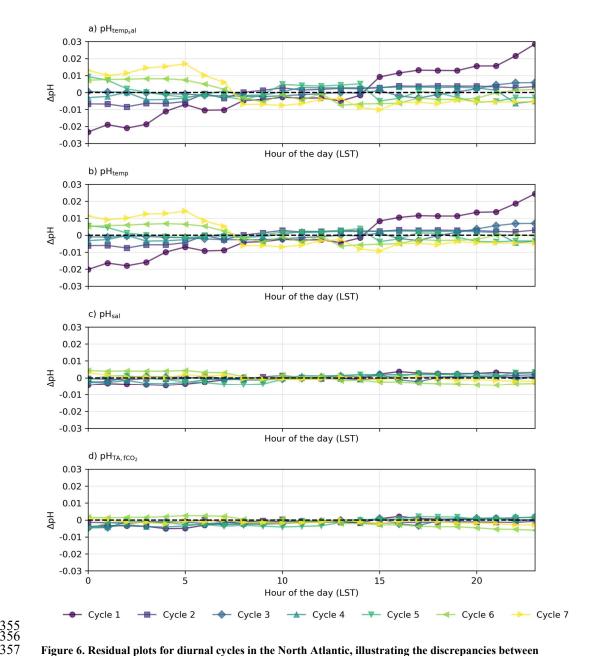


Figure 6. Residual plots for diurnal cycles in the North Atlantic, illustrating the discrepancies between observed pH measured by the optode (pH $_{obs}$ ) and a) pH calculated from constant TA and DIC with varying temperature and salinity (pH $_{temp,sal}$ ), b) varying temperature only (pH $_{temp}$ ), c) varying salinity only (pH $_{sal}$ ) and d) constant TA and fCO $_2$  with varying temperature and salinity . Horizontal dashed lines at y=0 indicate no deviation between observed and calculated pH.





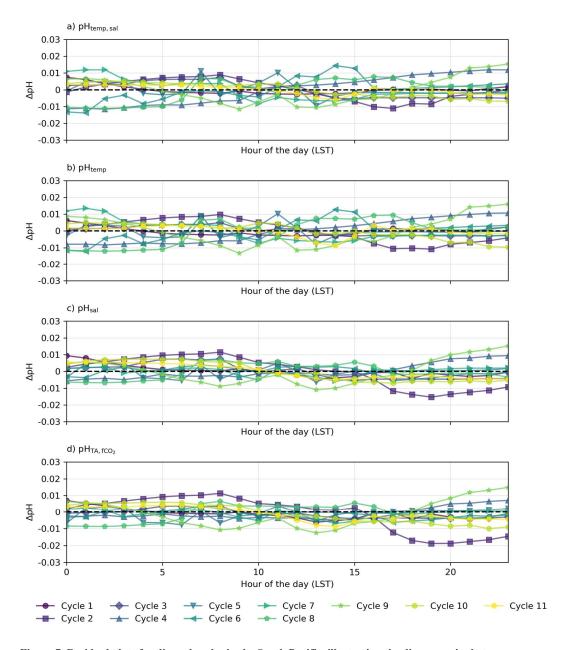


Figure 7. Residual plots for diurnal cycles in the South Pacific, illustrating the discrepancies between observed pH measured by the optode ( $pH_{obs}$ ) and a) pH calculated from constant TA and DIC with varying temperature and salinity ( $pH_{temp,sal}$ ), b) varying temperature only ( $pH_{temp}$ ), c) varying salinity only ( $pH_{sal}$ ) and d) constant TA and  $fCO_2$  with varying temperature and salinity. Horizontal dashed lines at y=0 indicate no deviation between observed and calculated pH.





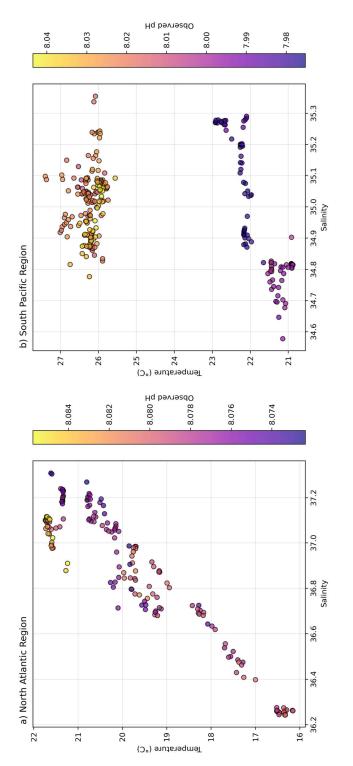
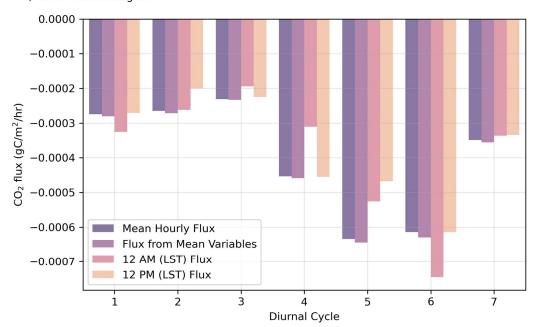


Figure 8. T/S diagram with observed pH for a) the North Atlantic region and b) the South Pacific region.









# b) South Pacific Region

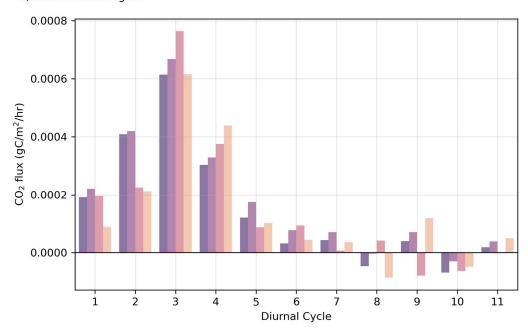


Figure 9. Sensitivity analysis of mean  $CO_2$  flux compared to flux calculated from mean inputs, fluxes for 12 AM (LST) and fluxes for 12 PM (LST). Top panel a) shows cruise SO279 in the North Atlantic while bottom panel b) shows cruise SO289 in the South Pacific.





#### 3 Results and Discussion

376 We first analyse the effects of temperature and salinity on pH across different diurnal cycles and 377 regions (Sect. 3.1). We examine the pH expected from temperature and salinity variations alone 378 (pH<sub>temp,sal</sub>), disregarding changes in TA or DIC. If pH<sub>obs</sub> aligns with pH<sub>temp,sal</sub>, it suggests that 379 recent temperature changes, such as day-night cycles, primarily control pH. In this context, 380 "recent" is relative to the air-sea CO<sub>2</sub> equilibration timescale, i.e., temperature change that 381 happened recently enough that its effect on pH has not been modified by subsequent gas 382 exchange. Next, we assess the influence of water mass variations on pH by considering 383 pH<sub>TA,fCO2</sub>, which accounts for constant fCO<sub>2</sub> instead of DIC alongside constant TA and varying 384 temperature and salinity. Alignment of pHobs with pHTA,fCO2 indicates that slower or long-ago 385 processes control pH. For example, an observed change in temperature may be due to spatial 386 variability, with the ship passing through different water masses that have had different 387 temperatures for long enough to reequilibrate with atmospheric CO<sub>2</sub>. We then explore the role of 388 biological activity and its interaction with abiotic factors by looking at the discrepancies between 389 pH<sub>obs</sub> and both pH<sub>temp,sal</sub> and pH<sub>TA,fCO2</sub> (Sect. 3.2). Finally, we distinguish between temporal and 390 spatial variability in our measurements (Sect. 3.3), considering implications for air-sea CO<sub>2</sub> 391 equilibration timescales at fine spatio-temporal scales (Sect. 3.4).

# 3.1 Influence of temperature and salinity

The diurnal cycles of pH observed in the North Atlantic and South Pacific Oceans are significantly influenced by temperature. In most cycles, observed pH stays within  $\pm 0.01$  of pH<sub>temp,sal</sub> thus indicating that temperature and salinity together drive pH changes (Fig. 6 and 7). However, salinity alone does not appear to strongly influence pH in any observed cycle. For both ocean basins, pH<sub>sal</sub> typically remains close to the mean pH of each cycle, rather than impacting the observed pH significantly. This consistency suggests that daily salinity variations do not exert a primary influence on the observed pH (Fig. 4 and 5; Fig. 6c; Fig. 7c). Thus, the rest of this section focuses on the effect of temperature on the observed pH.

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On short timescales (minutes and faster), temperature influences pH by affecting the dissociation of carbonic acid (H<sub>2</sub>CO<sub>3</sub>) into bicarbonate (HCO<sub>3</sub><sup>-</sup>) and carbonate (CO<sub>3</sub><sup>2</sup>-) ions. As temperature changes, the dissociation constants of carbonic acid change, which in turn alters the concentrations of these ions and the overall pH of the solution. This dependency shifts the equilibrium of these species with changing temperature and affects pH over short timescales (i.e., seconds to minutes; (Egleston et al., 2010; Stumm & Morgan, 2012). However, the overall response can be lessened by the stabilizing effect of alkalinity, which keeps the pH steady despite temperature changes (Frankignoulle, 1994; Middelburg et al., 2020). The alkalinity of seawater allows the carbonate system and other weak acid-base equilibria to maintain pH by absorbing or releasing protons, which reduces the impact of temperature changes on pH (Dickson & Millero, 1987; Egleston et al., 2010). Typically, the most pronounced temperature fluctuations—highest temperatures during daylight and lowest at night—occur in surface waters. This is due to surface waters' lower thermal inertia and direct exposure to the atmosphere. In contrast, deeper waters have a smaller thermal response because they are insulated from direct atmospheric influence and have a higher heat capacity, which buffers them against rapid temperature changes.





In the North Atlantic, the observed variability in pH generally matches the calculated pH<sub>temp</sub> and pH<sub>temp,sal</sub>, with most cycles' residuals within ±0.01 (Fig. 4 and Fig. 6a and b, Cycles 2, 3, 4, 5 and 6). However, some cycles match better than others. Cycle 3 demonstrates a consistent alignment between observed pH and expected pH<sub>temp,sal</sub> throughout the day, with minimal oscillations around the mean (mean residuals for pH<sub>temp,sal</sub> of less than -0.001; Fig. 4; Fig. 6a and b). A similar pattern is observed in Cycle 4, where pH variations are minimal (residuals range from -0.007 to 0.003 for pH<sub>temp,sal</sub>, with a mean of less than -0.001; Fig. 4). In contrast, in Cycle 5, pH<sub>obs</sub> and pH<sub>temp,sal</sub> show close agreement throughout the day with residuals close to zero, indicating a stable match despite slight deviations from that day's cycle mean. Cycle 6 also displays more pronounced daily fluctuations in pH<sub>temp</sub> and pH<sub>temp,sal</sub> with residuals from -0.008 to  $\pm 0.008$  (Fig. 6) but that still closely follow pH<sub>obs</sub> (within the  $\pm 0.01$  residuals; Fig. 4). In the South Pacific, similar patterns are evident in Cycles 1, 3, 4, and 6, with consistent and minimal fluctuations around the cycle mean (Fig. 5). Cycle 3 in particular shows most stability, with all expected pH values remaining within a narrow range of the observed pH (i.e. mean residuals for pH<sub>temp,sal</sub> of less than 0.001; Fig. 5; Fig. 7a and b). Similarly, in Cycles 4, 10 and 11, pH<sub>obs</sub> closely match pH<sub>temp,sal</sub> across the day, and exhibit minimal residuals (< 0.01; Fig. 5 and 7a and b).

However, notable discrepancies occur in pH observations versus calculated pH $_{temp}$  and pH $_{temp,sal}$  in the North Atlantic. Specifically, in Cycle 1, while pH $_{obs}$  remains relatively stable, there are significant daily fluctuations in pH $_{temp}$  and pH $_{temp,sal}$ , with residuals ranging from -0.023 to 0.029 and a mean of less than -0.001 (Fig. 4; Fig. 6a, b, and c). Cycles 5 and 7 also show discrepancies: although pH $_{obs}$  generally matches the cruise mean across all cycles, pH $_{temp,sal}$  exhibits greater deviations, indicating that some other process has eliminated the temperature signal in the observed pH (Fig. 4; Fig. 6a and b). In the South Pacific, similar patterns emerge. Cycle 5 shows significant hourly fluctuations in pH $_{obs}$ , which closely track the deviations in pH $_{temp}$  and pH $_{temp,sal}$ , markedly diverging from the more stable pH $_{sal}$  (Fig. 5; Fig. 7a and b). Additionally, in Cycle 2, the initial excess and subsequent shortfall in pH $_{temp}$  and pH $_{temp,sal}$ , with residuals from -0.011 to 0.008, suggest a comparable moderated effect by solar heating (Fig. 4). This pattern is

consistent with observations in the North Atlantic. This overriding influence, likely due to solar

heating, impacts the diurnal pH fluctuation patterns, particularly in cycles where pH<sub>temp</sub> and

450 pH<sub>temp,sal</sub> start higher and decrease throughout the day.

While changes in pH and fCO<sub>2</sub> due to temperature change occur virtually instantaneously (i.e., seconds to minutes), CO<sub>2</sub> equilibration between the atmosphere and the ocean is considerably slower (i.e., weeks to months; Jones et al. (2014). This means that the timescale of temperature change significantly influences how a water masses' pH adjusts to temperature changes (Takahashi et al., 2009). For instance, when a water mass cools during the night, it quickly alters pH and causes CO<sub>2</sub> undersaturation due to increased solubility, so observed pH would agree with pH<sub>temp</sub> (Feely et al., 2008). However, as the ship travels through different water masses that vary in temperature, but are each in equilibrium with the atmosphere, observed pH may not align with pH<sub>temp</sub> or pH<sub>temp,sal</sub> calculated based on temperature shifts alone. This raises an issue regarding the timescales of temperature change in relation to chemical equilibration, as it is uncertain whether these changes provide the system enough time for fCO<sub>2</sub> (and DIC) to shift and maintain equilibrium with the atmosphere, or if they happen more rapidly, such that disequilibrium remains.





In the North Atlantic, significant mixing due to ocean currents, eddies, and upwelling introduces substantial variability in temperature, salinity, and pH across different water masses (Fig. 6a and b). This variability is further influenced by the Gulf Stream and North Atlantic Drift, which create complex structures of distinct water masses (Liu & Tanhua, 2021). These dynamic conditions adjust the rate of CO<sub>2</sub> absorption or release, impacting the timescale over which ocean-atmosphere equilibrium is achieved. But in this ocean basin, pH<sub>TA,fCO2</sub> (i.e. proxy for gasexchange equilibrium) follows pH<sub>obs</sub> very closely, indicating that temperature changes have been sufficient for the system to reach re-equilibration. This is true both for individual cycles' diurnal distributions (Fig. 4), and when comparing the cycles' residuals together, which are virtually zero (Fig. 6d), even during temperature gradient crossings (Fig. S4).

In contrast, the South Pacific displays more predictable water mass dynamics primarily driven by physical processes like mixing and advection, which introduce water masses with consistent temperatures, salinities, and pH, resulting in a more homogeneous water column. This stability is influenced by large-scale gyres and trade winds, contributing to the uniformity of surface waters (Vallis, 2017). This is reflected by the consistent pH<sub>obs</sub>. Residuals are further from zero although most cycles fall within  $\pm 0.01$ , indicating a lesser degree of re-equilibration (Fig. 7).

Thus, this implies that, in both ocean basins, gas exchange plays a significant role in stabilizing pH for all cycles by re-equilibrating with atmospheric  $CO_2$  despite temperature fluctuations – although it likely does more so in the North Atlantic than in the South Pacific. To first order, this also suggests that the temperature change happened long enough ago for re-equilibration. However, the rates of gas exchange in the North Atlantic and South Pacific are from 3 to 6 months and 3 to 4 months, respectively (Jones et al., 2014). Thus, the consistency between  $pH_{TA,fCO_2}$  and  $pH_{obs}$  likely represents the crossings of water masses that are in equilibrium with the atmosphere and that equilibrated a long time ago (i.e., weeks to months) with different TA, DIC, and temperature baselines. Thus, the observed temperature changes as the ship traverses different water masses are likely attributable to spatial variations rather than an immediate response to recent temperature shifts. However, some pH cycles observed do not fully conform to these patterns, suggesting that other factors could be influencing the pH beyond the effects of temperature and gas exchange alone.

Ultimately, although we observe temperature-induced variations throughout each diurnal cycle, when comparing all cycles'  $pH_{obs}$ ,  $pH_{temp}$  and  $pH_{temp,sal}$  within both ocean basins, residuals are well within  $\pm 0.02$  (Fig. 6a and b and Fig. 7a and b). Thus, despite observed 24-hour fluctuations, pH remains relatively stable with respect to temperature variations across both ocean basins. This stability implies that even though there are short-term fluctuations in pH due to temperature changes, overall the pH stays relatively consistent, and thus a stable ocean-atmosphere  $CO_2$  equilibrium.

### 3.2 Influence of biological activity

While temperature and salinity effects on pH have been addressed in Sect. 3.1, deviations from expected patterns may be shaped by biological processes. The process of photosynthesis during daylight consumes CO<sub>2</sub>, leading to a rise in pH, whereas respiration and decomposition at night, which release CO<sub>2</sub>, lower pH (Falkowski, 1994; Falkowski & Raven, 2013). The balance https://doi.org/10.5194/egusphere-2024-2853 Preprint. Discussion started: 30 October 2024 © Author(s) 2024. CC BY 4.0 License.





between photosynthesis and respiration hence affects pH, and should result in a diurnal pH cycle.

These biological processes can also influence the rate at which CO<sub>2</sub> equilibrates between the

ocean and atmosphere. For instance, intense photosynthetic activity during daylight hours can

rapidly deplete seawater  $pCO_2$  in surface waters, potentially accelerating seawater  $CO_2$  uptake

from the atmosphere. Conversely, nighttime respiration can increase  $pCO_2$ , slowing the

outgassing process and thus extending the equilibration timescale. However, given an

equilibration timescale of months, these day and night changes likely get averaged out over

longer periods, resulting in an overall steady-state CO<sub>2</sub> flux and pH when observed over longer

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These biological processes are particularly evident in some cycles from the North Atlantic where the expected pH<sub>temp,sal</sub> changes due to temperature and salinity do not align with the observed data, but do follow the distribution expected from biological activity (Fig. 4). This is especially evident for Cycle 1, which shows an increase in pH during daylight hours and a decrease in pH during night hours, with a peak in pH in late afternoon and the lowest pH occurring in the early morning before sunrise (Fig. 4). Other cycles likely reflect the combined impact of biological processes and temperature effects, as the observed pH does not fully align with the distribution expected from any of temperature, salinity or biological activity (Fig. 4 and 6). This is also evident in the South Pacific Ocean, where some cycles show significant variability within pH<sub>obs</sub> that are not mirrored by the expected pH<sub>temp,sal</sub> changes (Fig. 5, Cycles 2, 8). For example, Cycle 2 exhibits significant deviations in pH with higher observed values in the morning and a drop in the afternoon, indicating biological influence (Fig. 5).

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The interplay between photosynthesis during the day and respiration at night, with possible contributions of other biological processes, may be behind the more pronounced peaks and troughs of Cycles 2 and 9 (Fig. 6; Duarte and Agusti (1998). Nitrogen fixation consumes hydrogen ions, increasing pH, whereas denitrification releases hydrogen ions, thereby decreasing pH (Richardson, 2000; Zehr & Kudela, 2011). However, without nutrient and oxygen data, we cannot directly assess whether these processes also impacted the observed pH.

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Although biological activity likely accounts for some variability, not all cycles exhibit the same residual, presumably biotic pattern of variation (Fig. 4 and 5). This variability further complicates assessments of air-sea CO<sub>2</sub> equilibration, as localized biological conditions may transiently alter CO<sub>2</sub> dynamics, influencing the timescale for reaching equilibrium. This is more obvious in the South Pacific, where some cycles display more pronounced night-time stability (Fig. 5, Cycles 3, 6, 7 and 11), while others have noticeable day-time fluctuations that could align with photosynthetic activity, which typically increase during daylight hours (Fig. 5, Cycle 8; (Duarte & Agusti, 1998; Raven & Johnston, 1991). Cycle 8 in the South Pacific shows a peak in observed pH around midday, likely due to increased photosynthesis, followed by a decrease in the evening (Fig. 5). Other cycles do not show this pattern as clearly (i.e., they even show a decrease of pH during the day), suggesting that, for some cycles, respiration may dominate over photosynthesis also during daytime (Fig. 5). For example, Cycle 9 displays significant variations in pH throughout the day, with notable decreases during daylight hours (Fig. 5). Additionally, as some cycles appear to conform to temperature-based pH expectations, the only minor deviations observed suggest biological activity to be minor or represent a balanced biological system where photosynthesis and respiration are in near-equilibrium (Fig. 5, Cycles 3 and 6).

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Despite the strong impact of both abiotic and biotic factors on pH, some cycles exhibit fine-scale trends that cannot be solely attributed to temperature fluctuations or biological activity. The fine-scale trends observed, especially in Cycles 1 and 7 (Fig. 4), exceed what can be attributed to temperature-induced changes alone and cannot be explained by biological activity, given the atypically high carbon fixation rates required to explain the pH offsets (Basu & Mackey, 2018; Wang et al., 2023). Indeed, the required biological carbon fixation would need to exceed 687 mg C m³ day to explain the offset (Text S7). Typical rates in open ocean waters are much lower, generally ranging from 50 to 150 mg C m³ day⁻¹ in nutrient-poor regions and can reach up to 1000 C m³ day⁻¹ in upwelling zones during phytoplankton blooms, which is not the case here (Basu & Mackey, 2018; Wang et al., 2023).

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Comparing the impact of biological activity based on the offsets between pH<sub>temp</sub> and pH<sub>obs</sub> with chlorophyll-a fluorescence data also shows no clear pattern (Fig. S2 and S3). This is expected as 569 570 fluorescence does not necessarily reflect instantaneous photosynthetic activity. Instead, 571 fluorescence primarily indicates the presence and abundance of phytoplankton. Therefore, while fluorescence can provide insights into the overall biomass of phytoplankton, it does not directly 572 573 correlate with photosynthesis and respiration. The absence of a clear correlation between 574 fluorescence and the daily pH cycle, with some cycles even showing a decrease in pH during 575 daytime, confirms that the influence of different water masses is important in shaping the local 576 high-resolution pH profiles.

# 3.3 Variability and stability in spatio-temporal pH patterns

578 The North Atlantic is characterized by complex interactions among water masses, leading to 579 higher turbulence and susceptibility to rapid and significant changes in water mass properties and 580 circulation patterns, influenced by both past and present climatic conditions (Lynch-Stieglitz, 581 2017; Gebbie, 2014). This complexity is partly due to the Atlantic Meridional Overturning 582 Circulation (AMOC), which is sensitive to various climate perturbations and has historically 583 undergone significant reorganizations, especially during climatic transitions such as the Last 584 Glacial Maximum (Curry and Oppo, 2005; Duplessy et al., 1988). As examined in Sect. 3.1, 585 considerable spatial variability is observed in the North Atlantic (Fig. 8 and Fig. S6), and this 586 heterogeneity introduces complexity in deciphering the relative contributions of abiotic and 587 biotic factors to pH fluctuations (Gruber & Sarmiento, 2002). Residual plots for the North 588 Atlantic suggest that the observed discrepancies between pHobs and both pHtemp and pHtemp, are 589 insignificant, indicating that variability in water mass characteristics—which include but are not 590 limited solely to temperature and salinity—do not dominate the observed pH variability (Fig. 6 591 and 8). Therefore, the hourly fluctuations in pHobs and the observed discrepancies between pHobs 592 and pH<sub>temp/sal</sub> across various cycles in this ocean basin may not be attributed to pronounced 593 spatial variability (Fig. 4). The T/S diagram also demonstrates this spatial variability, showing a 594 wide range of pH values correlated with differing salinity and temperature profiles across various 595 water masses (Fig. 8), reinforcing that spatial dynamics do not significantly influence pH 596 variation in this region. This is also supported by the relative stability observed in Cycle 4, where 597 the ship's consistent positioning (i.e. on station) highlights the role of temporal variability rather 598 than spatial variability (Fig. S4). Although water could still be moving spatially around the ship, 599 the variability could be due to some degree to different water masses rather than being limited to 600 true in-situ temporal variability.





In contrast, the South Pacific cruise predominantly exhibits a stable transit through more homogeneous surface waters as no SST gradient is observed, although it is not entirely free from disruptions (Fig. 8 and Fig. S5; Qu and Lindstrom (2004). This suggests a minor role of interactions between diverse water masses. Despite the hourly fluctuations observed, all diurnal cycles in the South Pacific tend to cluster closely around the mean pH for their respective 24-hour periods, reflecting substantial stability within each cycle but considerable variability across different days, particularly evident among consecutive cycles (Fig. 5). This underscores a clear temporal delineation of diurnal cycles influenced by DIC changes (Fig. 5).

# 3.4 Implications for air-sea CO<sub>2</sub> fluxes

Our findings indicate that while temperature and salinity predominantly govern diurnal pH fluctuations, additional variability arises from water mass dynamics and biological activity. Temperature rapidly affects pH, but the slower rate of CO<sub>2</sub> equilibration with the atmosphere moderates its impact on air-sea CO<sub>2</sub> fluxes. Although biological processes markedly influence daily pH cycles, they do not fully account for the observed variability, especially in cases where unusually high carbon fixation rates would be required to explain observed pH (Sect. 3.2).

 $\mathrm{CO}_2$  fluxes were calculated for each complete diurnal cycle (Fig. 9). The flux calculations were performed by computing the mean flux for each cycle from daily mean inputs (wind speed, temperature, salinity, and  $p\mathrm{CO}_2$ ) and specifically for the hours 12 AM and 12 PM (LST) to examine temporal variations within each cycle (Fig. 9). Despite different methods of calculation, the  $\mathrm{CO}_2$  fluxes remained relatively consistent, indicating that the variations in pH and thus  $\mathrm{CO}_2$  flux do not significantly affect the air-sea  $\mathrm{CO}_2$  exchange in either basin.

This consistent result shows that despite the processes influencing pH in these two ocean basins during the study period, the surface ocean's interaction with atmospheric CO<sub>2</sub> over longer timescales (eg. months) can minimize the effect on pH, leading to relatively low variability on short spatiotemporal scales (kilometers/days).

## 4. Conclusions

High-resolution pH data is crucial for understanding the subtle yet important variations in ocean surface water pH across different regions. Diurnal cycles and hourly fluctuations in pH, with varying magnitudes, can be observed in both the North Atlantic and South Pacific basins. Variability in pH is influenced not only by temperature but also by interactions between various water masses and biological activity. These factors together drive observed pH fluctuations, often diverging from predictions based solely on temperature. However, although the processes responsible for these pH variations are well-understood, high-resolution data highlight the challenge of identifying a dominant factor at the fine-scale due to their complex interplay. 

In both the North Atlantic and South Pacific, the close correlation between  $pH_{TA,fCO2}$  and observed pH across diurnal cycles suggests a significant role of gas exchange in stabilising pH despite temperature fluctuations. The temperature changes must therefore have happened more slowly than the air-sea  $CO_2$  equilibration timescales for these ocean basins. The observed, relatively constant pH is likely due to the ship passing through different water masses that had already equilibrated with the atmosphere.





| 645<br>646<br>647<br>648<br>649                             | Thus, despite short-term fluctuations in pH, there were no major changes in ocean-atmosphere CO <sub>2</sub> disequilibrium, indicating that, for the regions and time periods studied here, broader ocean-basin scale analyses based on lower resolution datasets would still be able to capture the necessary variability in surface ocean CO <sub>2</sub> chemistry for accurate global CO <sub>2</sub> cycle assessments.  |
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| 650<br>651<br>652   | Data availability. The hydrographic and biogeochemical data presented here, together with the processing code will be freely available online at <a href="https://github.com/louisedelaigue/NA-SP-HIGH-RES-PH">https://github.com/louisedelaigue/NA-SP-HIGH-RES-PH</a> by the time of publication.   |
| 653   | Supplement. The supplement related to this article is available online.  |
| 654<br>655<br>656<br>657                                    | <i>Author contributions</i> . LD and MPH conceptualized the project. CG, LD, MPH and YO curated the data. LD, MPH and GJR performed the investigation. LD conceptualized the methodology, used the necessary software, visualized the data and prepared the original draft of the paper. LD, GJR, MPH, YO and CG reviewed and edited the paper.  |
| 658<br>659  | Competing interests. The contact author has declared that neither they nor their co-authors have any competing interests.  |
| 660<br>661  | Disclaimer. Publisher's note: Copernicus Publications remains neutral with regard to jurisdictional claims in published maps and institutional affiliations.   |
| 662<br>663<br>664<br>665<br>666<br>667<br>668<br>669<br>670 | Acknowledgements. We are grateful to the officers and crew of the R/V Sonne and technical support from GEOMAR for their support and assistance, and the opportunity to join cruises SO279 and SO289. Special acknowledgment goes to Paul Battermann who greatly assisted in the sampling aboard the FS Sonne during SO289. We thank Karel Bakker and Sharyn Ossebaar for their help in the lab – this work would not be possible without them. LD also wishes to thank Sorbonne Université and the Institut de la mer de Villefranche (France), in particular the OMTAB team, for hosting her during the later stage of this research project, as well as Jean-Pierre Gattuso, Pierrick Lemasson and Elsa Simon for the many brainstorming sessions. Figure 2 schematic made by cartographic design - faculty of Geosciences - Utrecht University. |
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