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Key Points:

- The South Subtropical Convergence (SSTC) at 40°S in the Atlantic Ocean shows an intensified biological carbon pump, amidst uncertainties
- Discrepancies in trends between oxygen utilization and nitrate concentrations provide two distinct estimates of the biological carbon pump
- Local and far-field processes affect remineralization at 40°S, but their relative contributions cannot yet be precisely quantified

Supporting Information:

Supporting Information may be found in the online version of this article.

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The Changing Biological Carbon Pump of the South Atlantic Ocean



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Abstract Global marine anthropogenic CO₂ inventories have traditionally emphasized the North Atlantic's role in the carbon cycle, while Southern hemisphere processes are less understood. The South Subtropical Convergence (SSTC) in the South Atlantic, a juncture of distinct nutrient-rich waters, offers a valuable study area for discerning the potential impacts of climate change on the ocean's biological carbon pump (C_{soft}). Using discrete observations from GLODAPv2.2022 and BGC-Argo at 40°S in the Atlantic Ocean from 1972 to 2023, an increase in dissolved inorganic carbon (DIC) of $+1.44 \pm 0.11 \mu\text{mol kg}^{-1} \text{ yr}^{-1}$ in surface waters was observed. While anthropogenic CO₂ played a role, variations in the contribution of C_{soft} were observed. Discrepancies emerged in assessing C_{soft} based on the tracers employed: when using AOU, C_{soft(AOU)} recorded an increase of $+0.20 \pm 0.03 \mu\text{mol kg}^{-1} \text{ yr}^{-1}$, while using nitrate as the reference, C_{soft(NO3)} displayed an increase of $+0.85 \pm 0.07 \mu\text{mol kg}^{-1} \text{ yr}^{-1}$. Key processes such as water mass composition shifts, changes in oxygenation, remineralization in the Southern Ocean, and the challenges they pose in accurately representing the evolving C_{soft} are discussed. These findings highlight that while global studies primarily attribute DIC increase to anthropogenic CO₂, observations at 40°S reveal an intensified biological carbon pump, showing that regional DIC changes are more complex than previously thought and emphasizing the need for better parameterizations to compute the BCP in the marine carbon budget.

Plain Language Summary The South Atlantic Ocean at 40°S has experienced changes in dissolved inorganic carbon (DIC) over the years, affecting its carbon composition. This study running from 1972 to 2023 showcased an increase in the DIC down to 2,000 m deep. While anthropogenic CO₂ has traditionally been seen as the major contributor, the biological carbon pump's activity, influenced by various ocean processes, emerged as a significant driver. Specifically, factors such as photosynthesis, organic matter remineralization, sea ice movements, and freshwater influx from melting ice play pivotal roles in dictating oxygen and nitrate levels, both crucial components to predict the biological carbon pump's contribution to DIC. Our findings emphasize that this enhanced biological pump might contribute as much as human-made CO₂ to DIC in certain ocean regions. If unchecked, these changes could recalibrate ocean carbon budgets and predictions with potential shifts in water mass compositions, demanding more vigilant future monitoring.

1. Introduction

The ocean has been acting as a sink for anthropogenic CO₂, absorbing an estimated 24% of anthropogenic CO₂ since the beginning of the industrial era, thus significantly mitigating climate change (Friedlingstein et al., 2023; Gruber et al., 2023). Part of this anthropogenic carbon remains in the form of dissolved inorganic carbon (DIC) and is transported into the ocean interior (Davila et al., 2022; Gruber et al., 2019; Khatiwala et al., 2009; Sabine & Tanhua, 2010; Sarmiento et al., 1992), and part of it is incorporated by marine organisms into organic matter or calcium carbonate (Heinze et al., 1991; Sarmiento et al., 1998; Volk & Hoffert, 1985).

The sequence of processes that store atmospheric carbon as biogenic matter at the surface ocean and sequester a small fraction of it in deep-sea sediments, where it can be stored permanently, is named the “biological pump” (Riebesell et al., 2009). While recent studies have focused on quantifying anthropogenic CO₂ uptake by the ocean at the air-sea interface, or its fate into the ocean interior as DIC, little is known about its effect on the removal by the biological carbon pump (BCP) over recent years, despite its importance for understanding future ocean carbon cycling. Biological processes in the upper ocean annually convert approximately 50–60 gigatons of dissolved inorganic carbon into organic matter (De La Rocha & Passow, 2014). Out of this amount, around 10% is

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transported out of the surface ocean in the form of organic carbon (De La Rocha & Passow, 2014). Through this natural and anthropogenic carbon export, the BCP lowers atmospheric CO₂ levels by ~200 ppm relative to a world without it (Parekh et al., 2006; R. G. Williams & Follows, 2011). It is thus essential that we understand the drivers and variability of the BCP and its vulnerability to current anthropogenic changes to predict future climate. However, the complexity of ecosystem functioning and composition makes estimates of both present-day and future organic carbon exports poorly constrained in models and observations (Henson et al., 2012, 2022; Laufkötter et al., 2016; Marsay et al., 2015).

The oceanic distribution of DIC is mainly controlled by three pumps: the biological processes of photosynthesis and remineralization (i.e., the soft-tissue pump, C_{soft}), the formation and dissolution of calcium carbonate (i.e., the carbonate pump, C_{carb}) and the uptake of natural and anthropogenic CO₂ (i.e., the solubility pump, C_{anth}; Gruber et al., 1996; Volk & Hoffert, 1985). While the solubility pump accounts for 30%–40% of ocean carbon export flux to the deep ocean (Gruber et al., 2002; Toggweiler et al., 2003), the remaining portion is transported from the sunlit surface to the deep ocean as photosynthetically fixed organic carbon by the soft tissue pump (Riebesell et al., 2009; Schlunegger et al., 2019). To quantify the rate of soft tissue and carbonate pump changes, one could monitor particulate organic and inorganic carbon sinking fluxes through time, but sediment trap data are sparse and associated with large uncertainties (Buesseler et al., 2007). Monitoring changes in DIC is also complicated because C_{soft}, C_{carb}, and C_{anth} cannot be readily distinguished from the DIC analyses themselves. Instead, one can unravel the different components of the carbon pump and changes therein by combining multiple proxies, such as the apparent oxygen utilization (AOU; the difference between the saturation oxygen concentration, [O_{2,sat}] and the observed oxygen concentration, [O₂]), or release rates of organic-matter degradation byproducts, that is, NO₃[−], PO₄^{3−} or alkalinity. For the carbonate pump, calcification and dissolution can be tracked in seawater by measuring dissolved calcium or alkalinity changes (C.-T. A. Chen, 1978; Feely et al., 2004). Carbon sequestration by the soft tissue pump can be estimated from nutrient and oxygen observations, taking advantage of the constant stoichiometric ratios of carbon, nutrients, and oxygen in marine organic matter in the open ocean, also known as the Redfield Ratio (Ito & Follows, 2005; Redfield, 1958). Using the latter, these variables are converted back to carbon units in the natural components of DIC (C_{carb}, and C_{soft}) as well as anthropogenic CO₂ (C_{anth}), allowing for the disentanglement of changes in each DIC component.

Although global evaluations of the oceanic CO₂ inventory have often highlighted the North Atlantic Ocean as the primary region of interest for anthropogenic CO₂ subduction, with a storage estimated at 23% of global ocean anthropogenic CO₂ (Khatiwala et al., 2013; Sabine et al., 2004), the importance of southern hemisphere processes, such as the formation of Antarctic Intermediate Water (AAIW), has recently gained attention (Groeskamp et al., 2016; Landschützer et al., 2015). The South Subtropical Convergence (SSTC), an intersection point of low-macronutrient subtropical gyre waters and high-macronutrient Antarctic Circumpolar Current waters, is a significant biogeochemical feature of the South Atlantic Ocean (Browning et al., 2014). This convergence zone results in strong downwelling and pronounced surface gradients in salinity and temperature, leading to distinct water column stratification with shallow mixing in a thermally homogeneous surface layer, increasing light availability and productivity (Browning et al., 2014). Such stratification can impact the marine carbon cycle by potentially inhibiting the transport of heat, oxygen, and carbon dioxide deeper into the ocean (Li et al., 2020).

Simultaneously, the Southern Ocean is known for its high macronutrient supply, primarily from the upwelling of nutrient-rich deep waters (Moore et al., 2001; Tagliabue et al., 2014). These waters provide nitrate that fuels primary production. As organic matter produced in the Southern Ocean sinks to the SSTC region, it undergoes remineralization, converting back into inorganic nutrients. Nitrate concentrations can thus indicate the extent of remineralization in an area (Moore et al., 2001; Tagliabue et al., 2014), with higher concentrations suggesting that more organic matter has been remineralized. This enhanced remineralization, fueled by the increased nitrate supply from the Southern Ocean waters, may then contribute to a stronger soft tissue pump and carbon export to deeper waters (Boyd & Trull, 2007; Sarmiento, 2006). However, it is important to note that the Southern Ocean is a High Nutrient, Low Chlorophyll (HNLC) area, meaning that despite the abundance of macronutrients, primary productivity is often limited by the availability of micronutrients such as iron (Boyd & Trull, 2007). Behrenfeld et al. (2006) have shown a decrease in net primary productivity (NPP) with global warming in major parts of the Southern Ocean. This indicates that primary productivity is not uniformly enhanced and is influenced by factors such as nutrient availability and climate change impacts.

Although not historically prominent, anticipated changes in the BCP are emerging due to specific climatic and oceanographic drivers. Behrenfeld et al. (2006) reported alterations in ocean productivity linked to climate change, with increased stratification in some regions affecting primary production. Polovina et al. (2008) identified the expansion of subtropical gyres and associated shifts in phytoplankton communities in the Pacific. Furthermore, Orr et al. (2005) demonstrated that escalating CO_2 concentrations lead to ocean acidification, challenging marine calcifying organisms crucial for carbon sequestration. Considering these documented changes in diverse oceanic regions and the critical nature of the SSTC around 40°S in the Atlantic Ocean, it is plausible that similar dynamics will impact the BCP in this region.

In this study, we aim to quantify changes in the biological carbon pump at 40°S in the Atlantic Ocean. We leverage data from the GEOTRACES 40°S cruise, the first zonal expedition at this latitude in the South Atlantic to measure DIC. We join up all the other GLODAPv2.2022 cruises going perpendicularly and recent autonomous marine observations from the BGC-Argo array into a single analysis. This time series of marine carbonate chemistry measurements is used to calculate the rates of change in key variables and used to disentangle changes in the natural components of DIC (soft tissue pump, C_{soft} and carbonate pump, C_{carb}) from anthropogenic CO_2 (C_{anth}). We examine the underlying reasons for the divergence in BCP estimations, probing into the shifts in water mass composition within formation areas and delving into the effects of diminished oxygenation and heightened remineralization in the Southern Ocean. BCP estimates are then aligned with global carbon budgets, revealing potential gaps in our understanding, particularly how regional changes in the biological carbon pump, driven by climate change, might feedback into these estimates.

2. Materials and Methods

2.1. Discrete Measurements From GLODAPv2.2022

Data from 11 cruises along latitude 40°S in the Atlantic Ocean were extracted from the Global Ocean Data Analysis Project version 2 (GLODAPv2.2022; Lauvset et al. (2022), Figure 1; see Table S1 in Supporting Information S1) and analyzed. From these, only the GEOTRACES 40°S cruise occupied the entire latitudinal transect (Expocode #740H20111224, 2011, RRS James Cook; cruise #4095 in GLODAPv2.2022), while the rest either covered it partially or crossed it latitudinally. Only stations with a bottom depth beyond 1,000 m were considered as to remove data falling on the continental shelf and/or slope. Rather than using original cruise data, the analysis benefited from the GLODAPv2.2022 applied adjustment of properties, making the analysis internally consistent. Hydrographic parameters included temperature (T), practical salinity (S_p), dissolved oxygen ($[\text{O}_2]$), dissolved inorganic carbon (DIC), total alkalinity (A_T) and nitrate (NO_3^-). Only data with a quality control deemed “Good” (Flag = 2) were used. This analysis considered the full depth of the water column, with an emphasis on the central and intermediate waters during the discussion as most change has been observed in these waters (Piñango et al., 2022).

2.2. Autonomous Data From BGC-Argo Floats

Float data were downloaded from the Argo Global Data Assembly Center (AOML) for an 8-year period (2015–2023). Only months matching GLODAPv2.2022 were kept (i.e., October, November, December, January, February, and March). Selected floats all fell within a defined geographical area matching GLODAPv2.2022 (38°S–42°S, 15°W–50°E, Figure 1). All selected float profiles included temperature (T), practical salinity (S_p), dissolved oxygen ($[\text{O}_2]$), pH on the total scale (pH_T) and nitrate (NO_3^-) (i.e., 9 floats, 73 profiles; see Table S2 in Supporting Information S1). All BGC-Argo profiles used here were downloaded as Delayed Mode files, which are designed for scientific exploitation and represent the highest quality of data to possibly extract climate-related trends (Bittig et al., 2019). Only adjusted data that were flagged as “Good” (QC = 1) were used in this study, except for T and S_p , for which estimated values (QC = 8) were used when adjusted data was missing. Each float profile was treated as the equivalent of a GLODAPv2.2022 cruise in the analysis.

2.3. Derived Variables

For all BGC-Argo profiles, A_T was calculated using Python SciPy nonlinear least-square fitting (v.1.9.3; Virtanen et al., 2020) with T, S_p and depth as input parameters. More information can be found in Supporting Information S1, along with a comparison to Lee et al. (2006) A_T estimates (Text S1 and Figures S1–S3 in Supporting Information S1).

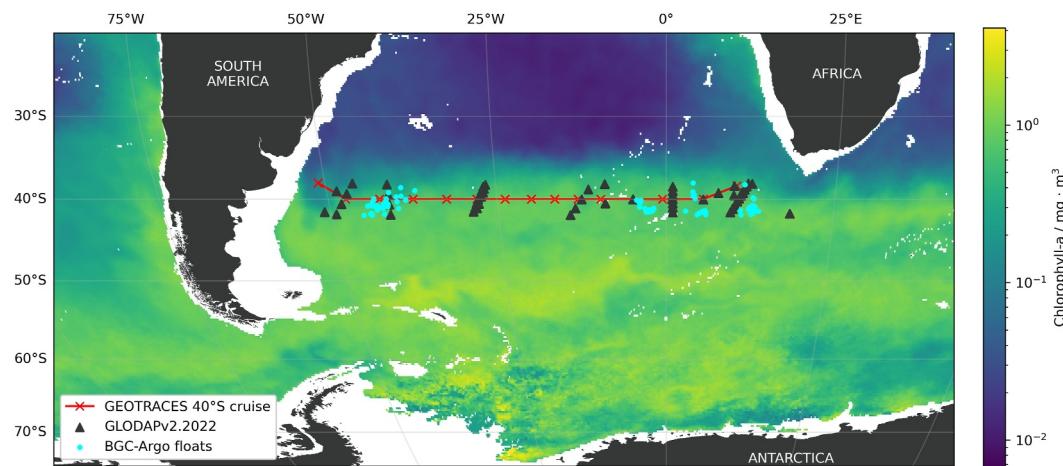


Figure 1. Map of the study region in the South Atlantic Ocean. Red lines with crosses show the trajectory of the longitudinal GEOTRACES 40°S cruise. Black triangles show the location of GLODAPv2.2022 cruises. Turquoise blue dots show the BGC-Argo float data. Color scale of the background map shows chlorophyll-a concentration (log-scale; Chl-a data downloaded from CMEMS following Sauzède et al., 2016).

Neutral density surfaces (γ^n) were calculated from T , S_p , pressure, latitude and longitude using the EOS-80 Legacy toolbox for MATLAB® (MathWorks®, USA). Absolute salinity (S_A) and potential temperature (θ) were calculated from T , S_p , and pressure using the Gibbs-SeaWater Oceanographic Toolbox for MATLAB® (MathWorks®, USA). Apparent oxygen utilization (AOU) was calculated from θ , S_A and $[O_2]$ using the combined fit coefficients from Garcia and Gordon (1992). These calculations were carried out for all data sets without the above variables either already measured or calculated. The Mixed Layer Depth (MLD) for each station during the GEOTRACES 40°S cruise was determined using a density criterion, with density calculated from T , S_p , and pressure data according to the Python implementation of the Gibbs SeaWater (GSW) Oceanographic Toolbox of TEOS-10 (v. 3.6.16; McDougall & Barker, 2011). The MLD was defined as the depth at which the water's density increases by 0.03 kg m^{-3} compared to the surface. Then, the mean MLD and mean neutral density across all stations for that cruise were calculated, resulting in an average MLD of 54m ($\gamma^n \approx 25.8$). We performed this calculation to distinguish the upper layer from the rest of the water column, which is essential for separating deeper water biogeochemical processes which fundamentally differ from those in the surface layer. Our results specifically focus on the waters below the MLD.

2.4. Marine Carbonate System Parameters

Depending on available variables, remaining marine carbonate parameters were calculated from any pair of DIC, A_T , pH_T using PyCO2SYS v1.8.1 (Humphreys et al., 2022), and with the carbonic acid dissociation constants from Sulpis et al. (2020), bisulfate dissociation constant from Dickson (1990), the total boron:chlorinity from Uppström (1974) and the hydrogen fluoride dissociation constant from Dickson and Riley (1979). When available for GLODAPv2.2022, phosphate and silicate were included in the carbonate system calculations but their inclusion did not affect the results significantly as their concentrations are relatively low in the open ocean (mean difference of -0.007 for pH and 1.51% for pCO_2). Calculated parameters include pH on the total scale (pH_T) and the seawater partial pressure of CO_2 (pCO_2) for GLODAPv2.2022 data, and DIC and pCO_2 for BGC-Argo data.

2.5. Interpolations

For each year, all variable data were clustered and interpolated vertically to γ^n levels using Piecewise Cubic Hermite Interpolating Polynomial (PCHIP) fits to observations (Figure 2; Fritsch & Carlson, 1980; Humphreys et al., 2016). Briefly, the PCHIP method uses monotonic cubic splines to find the value of new points. This allows for comparability between all cruises across the transect as averaging on isopycnals mimics real oceanic mixing processes occurring primarily along isopycnals, or, more precisely, along neutral density surfaces. Data were not extrapolated beyond the measured observational γ^n levels.

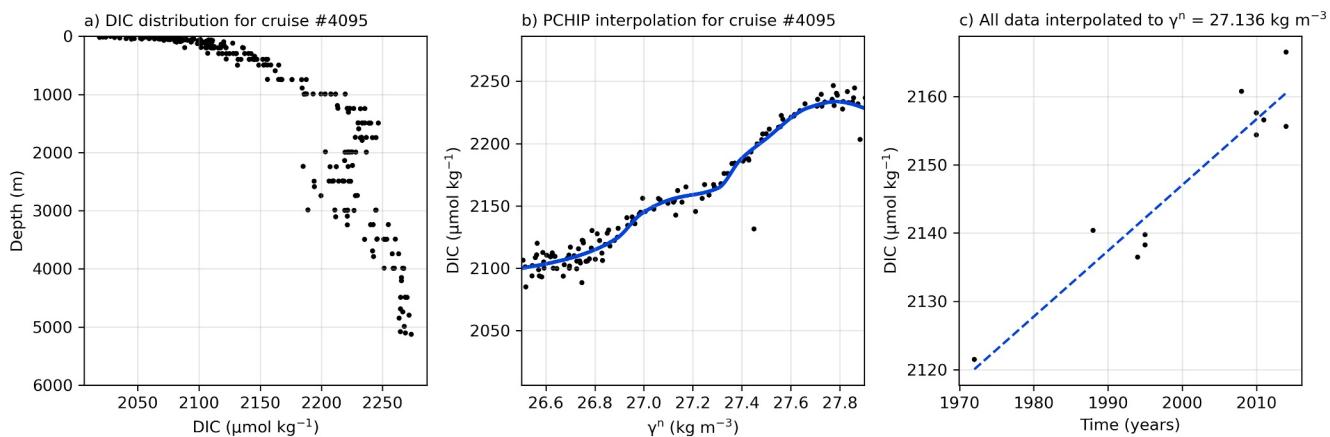


Figure 2. Analysis routine for each cruise and float profile, here showing for GEOTRACES 40°S longitudinal cruise: (a) the vertical distribution of DIC; (b) the PCHIP interpolation for DIC along neutral density levels; and (c) the linear regression at a given neutral density level.

2.6. Rates of Change

For each variable, ordinary least square regressions were used to determine their rate of change at each γ^n level (Figure 2). All rates of change are reported as $\frac{dX}{dt} \pm U$, where X is the variable and U is the associated uncertainty. Regressions were achieved using the Python library SciPy v.1.9.3 (Virtanen et al., 2020). All rates of change were calculated with the same time range for all variables, that is, from 1972 to 2023, except for silicate and phosphate for which only GLODAPv2.2022 data are available up to 2014.

2.7. CO₂ Components

Three pumps contribute to the distribution of DIC: photosynthesis and remineralization (C_{soft}), the formation and dissolution of calcium carbonate (C_{carb}), and uptake of natural and anthropogenic CO₂ (C_{anth} ; Gruber et al., 1996; Volk & Hoffert, 1985). Thus, DIC changes can be described as

$$\Delta T_{\text{DIC}} = \Delta C_{\text{soft}} + \Delta C_{\text{carb}} + \Delta C_{\text{anth}}. \quad (1)$$

Biological activity converts dissolved inorganic nutrients to particulate organic matter, which is transported down the water column through gravitational settling and active transport by marine organisms. During their settling in deeper waters, carbon and nutrients are returned to their dissolved inorganic forms through remineralization while taking up O₂ and thus increasing AOU (Redfield, 1963). This process drives the soft tissue pump, which is defined as

$$\Delta C_{\text{soft(AOU)}} = -R_{C/O_2} \cdot \Delta \text{AOU}, \quad (2a)$$

$$\Delta C_{\text{soft(NO}_3^-)} = R_{C/NO_3^-} \cdot \Delta \text{NO}_3^-, \quad (2b)$$

where $R_{C/X}$ is the increase in C as a fraction of X variable consumption during remineralization. Changes in nitrate (ΔNO_3^-) which, like changes in AOU (ΔAOU), reflect biological processes as well as physical mixing in the ocean. Specifically, ΔNO_3^- indicates the net effect of nitrate consumption during phytoplankton growth and release during remineralization. Thus, Equations 2a and 2b assume that changes occur under conditions influenced by both preformed properties—those characteristics established at the surface before subduction—and isopycnal mixing, which refers to mixing along surfaces of equal density.

In this study, we used the ratio from Anderson and Sarmiento (1994) as follows:

$$\begin{aligned} \text{P: N: C: } &-O_2 \\ &1: 16: 117: 170, \end{aligned} \quad (3)$$

Thus, R_{C/O_2} and R_{C/NO_3^-} can be assumed to be a constant value of -0.688 ± 0.092 and 7.31 ± 0.092 respectively (Anderson & Sarmiento, 1994).

The formation and dissolution of $CaCO_3$ makes up the carbonate pump, where an increase in C is coupled with a two-fold increase in A_T (Wolf-Gladrow et al., 2007):

$$\Delta C_{\text{carb}} = 0.5 \cdot (\Delta A_T - R_{N/O_2} \cdot \Delta \text{AOU}), \quad (4)$$

where R_{N/O_2} is a ratio of -0.0941 ± 0.0081 (Anderson & Sarmiento, 1994).

The remaining term, namely the solubility pump, relies on the uptake of anthropogenic CO_2 (ΔC_{anth}) and the CO_2 air-sea disequilibrium (ΔC_{diseq}) at the time the water lost contact with the atmosphere (Gruber et al., 1996). Assuming $\Delta C_{\text{diseq}} = 0$, that is, no significant long-term trend in air-sea CO_2 disequilibrium, the accumulated anthropogenic ΔC_{anth} is defined as

$$\Delta C_{\text{anth}} = \Delta C_{\text{sol}} - \Delta C_{\text{diseq}} \approx \Delta C_{\text{sol}}, \quad (5)$$

where ΔC_{sol} represents the change in carbon content due to solubility processes, reflecting the amount of anthropogenic CO_2 absorbed by the ocean through physical and chemical processes. In this context, considering $\Delta C_{\text{diseq}} = 0$ is justified based on the assumption that the long-term trends in air-sea CO_2 disequilibrium do not significantly affect the solubility pump on the timescales we are examining. While short-term and regional variations in disequilibrium are acknowledged, the overall impact on long-term global carbon uptake is considered minimal for this analysis (Jones et al., 2014; Nowicki et al., 2024).

2.8. Determining the Transition Depth for C_{soft} Significance

To determine the neutral density level at which the difference between $C_{\text{soft}(NO_3)}$ and $C_{\text{soft}(AOU)}$ stops being statistically significant, we conducted an analysis incorporating the individual errors of each estimate. For each neutral density level, the estimates of $C_{\text{soft}(NO_3)}$ and $C_{\text{soft}(AOU)}$ were adjusted by their respective errors to generate additional sample values. Specifically, for each estimate, three values were considered: the estimate itself, the estimate plus its error, and the estimate minus its error. These samples were then used in a paired t-test to compare the distributions of $C_{\text{soft}(NO_3)}$ and $C_{\text{soft}(AOU)}$ at each neutral density level.

The t-test was applied iteratively from the deepest depths toward shallower depths, testing for significant differences at each neutral density level using a significance level of 0.05. The analysis identified the neutral density level where the p-value first exceeded this threshold, indicating that the differences between $C_{\text{soft}(NO_3)}$ and $C_{\text{soft}(AOU)}$ were no longer statistically significant.

2.9. Natural CO_2 Inventory Change

The study area considered here is bound by latitudes -40.5 to -39.5° and longitude -50 to 15° for a total surface area of $616,300 \text{ km}^2$.

Changes in $C_{\text{soft}(AOU)}$ and $C_{\text{soft}(NO_3)}$ were multiplied by a density factor of 1028 to convert units from $\mu\text{mol kg}^{-1} \text{ yr}^{-1}$ to $\mu\text{mol m}^3 \text{ yr}^{-1}$. The resulting values were integrated along the water column by sum ($C_{\text{soft}(AOU)\text{-int}}$ and $C_{\text{soft}(NO_3)\text{-int}}$). Assuming lateral homogeneity, $C_{\text{soft}(AOU)\text{-int}}$ and $C_{\text{soft}(NO_3)\text{-int}}$ were further converted to $\text{GtC m}^2 \text{ yr}^{-1}$ and integrated over the surface of the study area.

2.10. Uncertainty Propagation

For GLODAPv2.2022, uncertainties were assigned based on Table 3 of Lauvset et al. (2022) for all data dated from after 1994 (i.e., first use of CRMs). For data prior to 1994, an uncertainty of $\pm 17.2 \mu\text{mol kg}^{-1}$ was assigned (Dickson, 1992). Most BGC-Argo floats included the error for each adjusted variable. Some uncertainties were missing for T and S_p , for which values were assigned based on N. L. Williams et al. (2017) and Mignot et al. (2019), except for the calculated A_T for which the fit RMSE was used ($5.5 \mu\text{mol kg}^{-1}$, see Text S1 in Supporting Information S1). Uncertainties for variables calculated using PyCO2SYS v1.8.1 were propagated using the independent uncertainty argument (Humphreys et al., 2022).

In the process of quantifying analysis uncertainty, a multifaceted approach was employed to ensure comprehensive error propagation. To assess the uncertainty in the rates of change, a differential analysis was undertaken, calculating the derivatives for all variables under consideration (i.e., forward-finite differences). Each variable was incrementally altered by adding or subtracting a value derived from its assigned uncertainty. This alteration was specific to the variable's nature, and the impact of these perturbations on the overall outcome was studied through a repeated linear regression on the data set, now modified to reflect these adjustments. The differences between the coefficients of the original and modified data sets provided insights into the sensitivity of each variable. After normalization, these differences gave the true derivatives. By juxtaposing these derivatives with known measurement uncertainties, a comprehensive error term for each variable's rate of change was determined, offering a robust assessment of propagated uncertainties. This step provided a foundational understanding of how minor perturbations in each variable could influence the outcome.

Subsequently, to further refine the understanding of the variability and potential uncertainties inherent in the data set, a bootstrapping technique was applied. Essentially, the data for specific cruises was selectively omitted in a series of simulations, creating a variety of modified data sets. This non-parametric statistical method facilitated the estimation of the distribution of sample statistics by resampling with replacement, thereby offering insights into the potential variability of the results.

Finally, for each measured variable, an internal bias was determined for each cruise or float profile using the standard deviation of each linear regression from the rates of change estimation. All uncertainties were then synthesized to produce a combined error estimate. This layered approach ensured a robust assessment of analysis uncertainty, capturing both the immediate sensitivities of the variables and the broader variability in the data set.

3. Results

Multi-decadal rates of change for all available and derived carbonate parameters were calculated using GLODAPv2.2022 and BGC-Argo data along latitude 40°S for the period 1972–2023. Changes in individual variables are reported in Figure 3, while the calculated change in each component of DIC is shown in Figure 4. For all variables, little to no change was observed deeper than ~2,000 m in the water column; thus, only data down to this depth are presented here.

On average, surface waters became warmer and fresher, which was reflected in A_T with a slight negative change. Slight decreases of $-0.04 \pm 0.0006^\circ\text{C yr}^{-1}$ in θ and $-0.01 \pm 0.0002 \text{ yr}^{-1}$ in S_p were observed in the top ~300 m ($\gamma^n < 27.0 \text{ kg m}^{-3}$), with virtually no change deeper in the water column (Figures 3a and 3b). This was closely mirrored by A_T , which showed a maximum decrease of $-0.49 \pm 0.11 \mu\text{mol kg}^{-1} \text{ yr}^{-1}$ within the top ~200 m ($\gamma^n < 26.8 \text{ kg m}^{-3}$), while showing no change deeper in the water column (Figure 3c).

Surface waters also showed signs of deoxygenation and increased AOU (Figure 3h), while the nitrate pool increased (Figure 3i). $[\text{O}_2]$ decreased down to ~900 m ($\gamma^n \approx 27.4 \text{ kg m}^{-3}$) with a minimum $\Delta[\text{O}_2]$ of $-0.56 \pm 0.05 \mu\text{mol kg}^{-1} \text{ yr}^{-1}$ (Figure 3g). Deeper than 150 m, the AOU increased by a maximum of $+0.54 \pm 0.02 \mu\text{mol} \cdot \text{kg}^{-1} \cdot \text{yr}^{-1}$, from the surface down to ~2,000 m ($\gamma^n \approx 27.9 \text{ kg m}^{-3}$; Figures 3h and 3i). Both AOU and $[\text{O}_2]$ converged toward no change at around ~300 m ($\gamma^n \approx 27.0 \text{ kg m}^{-3}$) before continuing their respective decrease/increase. An important increase in NO_3^- was observed with a peak of $+0.14 \pm 0.01 \mu\text{mol} \cdot \text{kg}^{-1} \cdot \text{yr}^{-1}$ for the first ~150 m ($\gamma^n < 26.7 \text{ kg m}^{-3}$), while slowly converging toward no change by ~1,000 m ($\gamma^n \approx 27.5 \text{ kg m}^{-3}$; Figure 3i).

For the analysis of the N/P ratio and $[\text{Si}]$, the data is limited to measurements up to 2014. This limitation arises because the BGC-Argo floats are not equipped to measure phosphate and silicate, thus only GLODAPv2.2022 data was used. Nonetheless, within this data set, there was no significant change observed in the N/P ratio (Figure 3l). Additionally, $[\text{Si}]$ displayed an increase of $+0.15 \pm 0.08 \mu\text{mol} \cdot \text{kg}^{-1} \cdot \text{yr}^{-1}$ near $\gamma^n = 27.4$ with no substantial alterations observed in the remaining portions of the water column (Figure 3k).

Across the study period, seawater at 40°S became more acidic due to a change in DIC partially caused by biological processes, calcium carbonate formation and CO_2 uptake. A small decrease of $-0.24 \pm 0.06 \mu\text{mol kg}^{-1} \text{ yr}^{-1}$ in the carbonate pump, C_{carb} , was observed within the first ~200 m ($\gamma^n < 26.8 \text{ kg m}^{-3}$) of the water column (Figure 4), correlating closely with the change in A_T (Figure 3c). DIC showed an increase above ~2,000 m ($\gamma^n \approx 27.9 \text{ kg m}^{-3}$; Figures 3d and 4) with a maximum $+1.44 \pm 0.11 \mu\text{mol kg}^{-1} \text{ yr}^{-1}$ in the surface waters ($\gamma^n \approx 26.7 \text{ kg m}^{-3}$; Figures 3d and 4). It is interesting to note that the change in DIC was higher down to 2,000 m

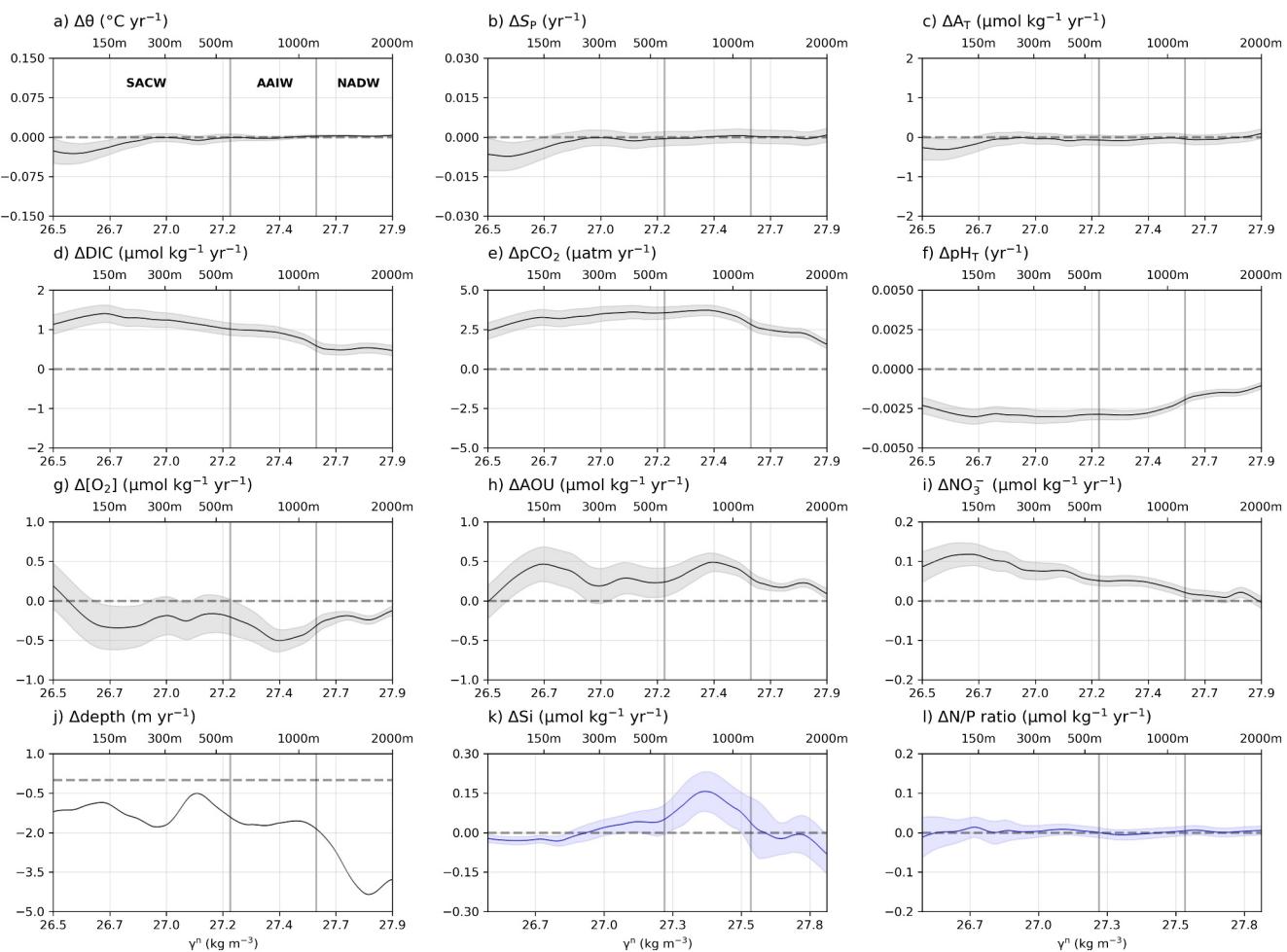


Figure 3. Change in biogeochemical parameters across 40°S in the Atlantic Ocean for all available data. Approximate depths are shown at the top of each plot along water masses as part of subplot a. Note that blue panels k and l only include data up to 2014 as BGC-Argo floats are not equipped for phosphate and silicate measurements. MLD is off the x-axis (54m, $\gamma^n \approx 25.8$).

when using all available data, most likely indicative of a deeper penetration depth within the last decade, and noticeable thanks to the high-resolution of BGC-Argo data. Seawater $p\text{CO}_2$ showed a corresponding average increase of $+3.83 \pm 0.15 \text{ }\mu\text{atm yr}^{-1}$, before slowly converging back to no change deeper than 2000 m ($\gamma^n \approx 27.9 \text{ kg m}^{-3}$; Figure 3e). These changes were closely followed by an average decrease of $-0.002 \pm 0.0001 \text{ yr}^{-1}$ in pH_T and a maximum decrease of $-0.003 \pm 0.0001 \text{ yr}^{-1}$ in the shallow subsurface ($\sim 150 \text{ m}$, $\gamma^n < 26.7 \text{ kg m}^{-3}$; Figure 3f), where the pH maximum is often located (Arroyo et al., 2022). This change in pH_T is consistent with the most recent globally averaged rate of surface ocean pH change, $-0.0016 \pm 0.0006 \text{ yr}^{-1}$ (Garcia-Soto et al., 2021).

Not all DIC changes were attributed to C_{anth} , as there also appeared to be a significant increase in the soft tissue pump (Figure 4). If calculated using AOU, $\text{C}_{\text{soft(AOU)}}$ increased by $+0.20 \pm 0.03 \text{ }\mu\text{mol} \cdot \text{kg}^{-1} \cdot \text{yr}^{-1}$ down to $\sim 2,000 \text{ m}$ ($\gamma^n \approx 27.9 \text{ kg m}^{-3}$; Figure 4), while being close to no change near $\sim 300 \text{ m}$ ($\gamma^n \approx 27.0 \text{ kg m}^{-3}$; Figure 4). The corresponding $\text{C}_{\text{anth(AOU)}}$ thus represented most of the DIC change, with an increase of $+1.48 \pm 0.13 \text{ }\mu\text{mol} \cdot \text{kg}^{-1} \cdot \text{yr}^{-1}$ close to the surface ($\gamma^n < 26.7 \text{ kg m}^{-3}$; Figure 4). However, if calculated using NO_3^- , $\text{C}_{\text{soft(NO}_3^-)}$ showed an increase ~ 4 times greater than $\text{C}_{\text{soft(AOU)}}$, with $+0.85 \pm 0.07 \text{ }\mu\text{mol kg}^{-1} \text{ yr}^{-1}$ in the surface waters ($\gamma^n < 26.8 \text{ kg m}^{-3}$; Figure 4), and $\text{C}_{\text{anth(NO}_3^-)}$ then contributing half of $\text{C}_{\text{anth(AOU)}}$, with an average increase of $+0.67 \pm 0.14 \text{ }\mu\text{mol kg}^{-1} \text{ yr}^{-1}$ down to $1,000 \text{ m}$ ($\gamma^n \approx 27.5 \text{ kg m}^{-3}$; Figure 4). The *t*-test revealed that the difference between $\text{C}_{\text{soft(NO}_3^-)}$ and $\text{C}_{\text{soft(AOU)}}$ stopped being statistically significant below 500 m ($\gamma^n = 27.15$).

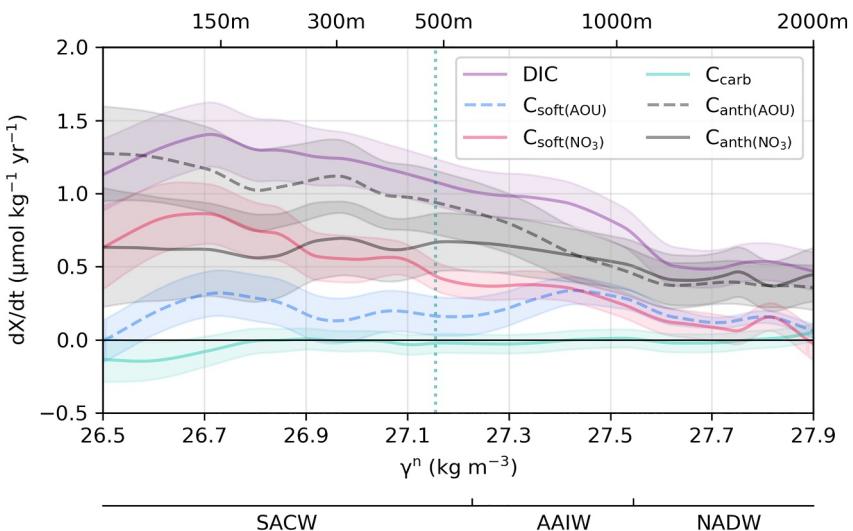


Figure 4. Rates of change in DIC (purple) and its components, C_{carb} (turquoise), $C_{\text{soft}(\text{NO}_3)}$ estimated from NO_3^- (red), $C_{\text{soft}(\text{AOU})}$ estimated from AOU (dashed blue), with corresponding $C_{\text{anth}(\text{NO}_3)}$ in solid black and $C_{\text{anth}(\text{AOU})}$ in dashed black. Results are for the compilation of all available data. Approximate depths are shown at the top and water masses at the bottom of the axes. MLD is off the x-axis (54 m, $\gamma^n \approx 25.8$). The vertical line (dotted green) indicates the neutral density level at which the difference between $C_{\text{soft}(\text{NO}_3)}$ and $C_{\text{soft}(\text{AOU})}$ stops being statistically significant (~ 500 m, $\gamma^n = 27.15$). Separated contributions from both C_{soft} estimates can also be found in Figure S4 in the Supporting Information S1.

Within the study area, the depth-integrated increase in (a) the soft tissue pump was approximately $+0.004 \pm 0.002 \text{ GtC yr}^{-1}$ for $C_{\text{soft}(\text{AOU})}$ and $+0.006 \pm 0.002 \text{ GtC yr}^{-1}$ for $C_{\text{soft}(\text{NO}_3)}$, (b) the anthropogenic component was approximately $+0.008 \pm 0.003 \text{ GtC yr}^{-1}$ for $C_{\text{anth}(\text{AOU})}$ and $+0.007 \pm 0.003 \text{ GtC yr}^{-1}$ for $C_{\text{anth}(\text{NO}_3)}$ and (c) the carbonate pump C_{carb} was approximately insignificant ($< 2 \times 10^{10} \text{ GtC yr}^{-1} \pm 0.001 \text{ GtC yr}^{-1}$).

4. Discussion

Although the biological pump at 40°S has clearly strengthened, the magnitude of the increase remains uncertain, depending on whether NO_3^- or AOU is used to estimate the soft tissue pump component. Below, we discuss this uncertainty and the implications of our findings, first looking at the physical factors and subsequently the influence of biology. Attributing the observed BCP changes to different driving factors is beyond the scope of this study, although it deserves further analyses. Here, we discuss how different physical and biogeochemical processes may qualitatively explain our results, as well as the observed discrepancies between the use of NO_3^- and AOU as a proxies for C_{soft} estimates.

4.1. Influence of Physical Factors on BCP Dynamics

4.1.1. Melting Ice

Variability in oxygen content at 40°S is largely influenced by sea ice dynamics affecting gas exchange, temperature and salinity (Hofmann et al., 2011). The stratification induced by meltwater creates a barrier that inhibits deep water mixing. As a result, vertical water exchange between the surface and deeper ocean layers is slowed, retaining carbon and nutrients in the upper water column for longer, marking a transition from a system that exports carbon and nutrients to one that retains them (Gjelstrup et al., 2022; Priest et al., 2023; von Appen et al., 2021). However, the melting of sea ice cools seawater, enhancing the solubility of CO_2 and O_2 without affecting NO_3^- (Loose et al., 2009). Subsequent air-sea gas exchange will thereby increase DIC and $[\text{O}_2]$, but it will not alter AOU because $[\text{O}_2]$ changes to match its new saturation level ($[\text{O}_2]_{\text{sat}}$). Enhanced gas exchange due to stronger surface winds in the South Atlantic, which is a net sink for CO_2 , would also increase surface ocean $p\text{CO}_2$ and DIC (Wanninkhof, 2014). This increase in DIC occurs because cooler seawater from melting sea ice can absorb more CO_2 from the atmosphere, assuming that atmospheric CO_2 levels are higher than those in the surface ocean. The impact varies with the supersaturation or undersaturation of surface waters with respect to atmospheric CO_2 and O_2 (Council, 2010; Figuerola et al., 2021). In scenarios of reduced gas exchange due to diminished sea

ice in the Southern Ocean, there should be a decrease in DIC and AOU, as the melting sea ice enhances the ocean's capacity to absorb atmospheric CO₂ and O₂, which would lead the concentrations of these dissolved gases to increase, thus changing C_{soft(AOU)} but not NO₃⁻ and C_{soft(NO₃)}. Increased biological productivity, driven by micronutrient influx, releases O₂ into the ocean, potentially leading to local oversaturation; however, this biological O₂ production is often offset by enhanced remineralization at depth, where O₂ is consumed, thereby maintaining overall similar AOU levels.

Melting land ice also enhances ocean stratification and introduces micronutrients such as iron, fueling biological productivity in near-surface waters (Lannuzel et al., 2010, 2016; Morley et al., 2020). Sediment-derived iron, accounting for 54 ± 15% of total iron flux, is carried from continental shelves via benthic diffusion and sediment resuspension (De Jong et al., 2012; Tian et al., 2023). This transport may influence nutrient dynamics and productivity as far as 3,500 km from the Antarctic Peninsula (De Jong et al., 2012). An influx of micronutrients leads to a higher uptake of DIC and also nitrate NO₃⁻, resulting in a decrease in C_{soft(NO₃)}. These changes, predominantly affecting near-surface photosynthesis, align with the Redfield Ratio, which predicts a decrease in NO₃⁻ proportional to the decrease in DIC. However, this would impact AOU and hence also C_{soft(AOU)} less because O₂ more quickly re-equilibrates with the atmosphere than DIC (Emerson & Hedges, 2008). Enhanced remineralization at greater depths due to the increased productivity would change DIC, AOU, and NO₃⁻ in line with the Redfield ratio (Henley et al., 2020). As the composition of the water we observe at 40°S is the result of transport of this water from south to north, it also reflects the initial preformed value from the enhanced productivity near the Antarctic as well as processes happening at 40°S and during transport (Morley et al., 2020). Although NO₃⁻ is generally consumed by biological uptake, localized upwelling or mixing processes can occasionally lead to higher concentrations of NO₃⁻ in surface waters, reflecting regional variations in nutrient dynamics.

A component of the observed increase in DIC at 40°S could hence reflect both the melting of sea and land ice, which in turn impact the BCP. But unlike for NO₃⁻, the AOU change from increased productivity should quickly be masked by air-sea gas exchange, making it impossible to quantify this process using AOU alone. Therefore, the observed changes in AOU at 40°S are not due to melting processes. While changes in NO₃⁻ could result from these changing endmember conditions, they are influenced by multiple factors such as nutrient dynamics, biological productivity, and vertical water mixing, making NO₃⁻ an unreliable sole indicator. Consequently, melting ice resulting from global change influences both estimates, although C_{soft(AOU)} may provide a more reliable interpretation. Future research should focus on developing a variable to quantify ice melt in this context, which would enhance our understanding of these processes.

4.1.2. Interior Ocean Mixing

Recent studies have highlighted changes in ocean circulation patterns, which most likely are also reflected at 40°S. Decrease in the ages of Subantarctic Mode Water (SAMW) and Circumpolar Deep Water (CDW) have been observed across the South Atlantic Ocean since the 1990s, suggesting enhanced ventilation, a phenomenon partly attributed to shifts in westerly winds (Fine et al., 2017; Tanhua et al., 2017; Waugh et al., 2013). This trend in SAMW indicates potentially increased isopycnal mixing, which elevates surface DIC due to the upward transport of DIC-rich deep waters, which may drive part of the observed increase in DIC at 40°S (Figure 3d). Additionally, Wei et al. (2022), focusing on a transect from the Rio Grande Rise to the Mid-Atlantic Ridge, demonstrated increases in diapycnal diffusivities, again indicating intensified mixing, thus implying enhanced vertical circulation of nutrient-rich deep waters. This observation is consistent with the observed rise in NO₃⁻ and C_{soft(NO₃)} at 40°S (Figures 3i and 4). Both the isopycnal and diapycnal mixing introduce oxygen-poor deep water to the surface, increasing SACW AOU, especially in regions with high remineralization (Figure 3h; Fine et al., 2017; Tanhua et al., 2017; Waugh et al., 2013). However, AOU does not increase significantly in the top layers of the profile (Figure 3g), which would be expected with the upwelling of deep oxygen-poor waters. This suggests that the apparent increase in remineralization within the BCP as inferred from DIC, AOU, and NO₃⁻ measurements is primarily due to the enhanced advection of older deep waters to the surface. However, increased winds also drive enhanced upwelling, which similarly elevates DIC, AOU, and NO₃⁻ in the SACW (Section 4.1.3). Therefore, the observed discrepancy between C_{soft(NO₃)} and C_{soft(AOU)} at 40°S cannot be solely attributed to wind-driven upwelling but is also influenced by other processes such as enhanced vertical mixing and advection of older waters.

Piñango et al. (2022) also reported increased organic matter remineralization in the AAIW, likely indicating deoxygenation. This is due to an enhanced flux of organic matter, which increases microbial oxygen consumption, and enhanced ventilation. While the latter introduces oxygen, it also brings in more anthropogenic carbon, exacerbating the oxygen demand for remineralization. The influx of anthropogenic carbon boosts the biological pump's activity, leading to greater production and sinking of organic matter. As this organic matter decomposes, microbial respiration intensifies, consuming more oxygen and further depleting oxygen levels in intermediate waters, thus amplifying the overall deoxygenation effect. This is especially true along the AAIW from 50°S to 30°S, where increased AOU was observed at a mean rate of AOU change of $0.23 \pm 0.68 \mu\text{mol kg}^{-1} \text{yr}^{-1}$ south of 30°S, which concurs with our results (Figure 3; Piñango et al., 2022). This is consistent with the deoxygenation trend reported by Santos et al. (2016) for the AAIW in the South Atlantic subtropical gyre from 1960 to 2015, which was $-0.18 \pm 0.04 \mu\text{mol kg}^{-1} \text{yr}^{-1}$. It also aligns with the AOU rate observed by Fontela et al. (2021) in the Argentine Basin, which was $0.38 \pm 0.13 \mu\text{mol kg}^{-1} \text{yr}^{-1}$. Additionally, these results agree with the AAIW deoxygenation documented by Schmidtko et al. (2017) over the last two decades.

A component of the observed changes in DIC, AOU, NO_3^- and therefore the BCP at 40°S is thus mechanistically linked to enhanced mixing through an increased influence of deeper waters on overlying water masses.

4.1.3. Wind-Driven Circulation

Increased wind-driven circulation has been invoked to explain changes in marine biogeochemical cycles, including effects on DIC and nutrients (England et al., 2014; Keppler & Landschützer, 2019; Liu et al., 2023).

The rate of gas exchange between the ocean and atmosphere is controlled by a set of physical processes that scale with wind speed (Wanninkhof, 2014). Because the South Atlantic Ocean is a net sink for CO_2 (i.e., seawater $p\text{CO}_2$ is lower than atmospheric $p\text{CO}_2$), enhanced gas exchange due to stronger surface winds would increase surface ocean $p\text{CO}_2$ and DIC. The same process decreases AOU, which is positive in surface waters here (Figure 3h), due to the absorption of atmospheric oxygen. In contrast, dissolved nutrients such as NO_3^- (and therefore $\text{C}_{\text{soft}(\text{NO}_3)}$), which are primarily controlled by biological uptake and nutrient cycling, are not directly affected by intensified gas exchange. A component of the observed changes in $\text{C}_{\text{soft}(\text{NO}_3)}$ and $\text{C}_{\text{soft}(\text{AOU})}$ could hence reflect the impact of increased wind-driven circulation. However, the complexity of these interactions makes it impossible to quantify this process using C_{soft} alone. This highlights the need to consider wind-driven effects in conjunction with other drivers, such as enhanced vertical mixing and advection of older waters (Section 4.1.2).

However, as discussed in Section 4.1.2, increased winds drive enhanced upwelling due to intensified wind shear over the surface ocean. Deep waters are enriched in DIC, AOU, and NO_3^- due in a large part to the remineralization of organic matter, so enhanced upwelling elevates all these properties in the SACW roughly in line with the Redfield ratio (Figures 3d, 3h, and 3i; England et al., 2014; Liu et al., 2023). Accordingly, changes in wind-driven circulation would increase both $\text{C}_{\text{soft}(\text{NO}_3)}$ and $\text{C}_{\text{soft}(\text{AOU})}$ in SACW, but likely in agreement with the Redfield ratio and true change in C_{soft} . Therefore, this aspect alone cannot cause the observed discrepancy between $\text{C}_{\text{soft}(\text{NO}_3)}$ and $\text{C}_{\text{soft}(\text{AOU})}$ at 40°S and the influence of interior ocean mixing must hence also be considered (Section 4.1.2).

Additionally, increased mixing due to stronger winds leads to a more uniform distribution of DIC in the upper part of the water column as the surface mixed layer thickens (England et al., 2014), as seen at 40°S (Figure 3j). This mixing increases the AOU near the surface and lowers it at depth. Mixing also results in a more even distribution of nutrients like NO_3^- throughout the water column, redistributing these vertically rather than straightforwardly increasing or decreasing the soft tissue pump. This highlights that the observed changes in DIC, AOU, and NO_3^- are not solely driven by biological processes but are also influenced by physical mixing. It is hence essential to understand the complex interplay of factors affecting these variables. However, unlike NO_3^- , the changes in AOU and DIC are influenced by both physical mixing and air-sea gas exchange, making it impossible to quantify the exact contribution of each process to the observed changes.

4.2. Biological Effects on Quantifying the BCP

Previous sections have highlighted the significant influence of physical processes on the observed changes in DIC, AOU, and NO_3^- at 40°S. Melting ice, both sea and land, affects gas exchange and nutrient dynamics,

leading to changes in DIC and $[O_2]$ without altering NO_3^- directly. Enhanced interior ocean mixing, driven by shifts in ocean circulation, has been shown to elevate surface DIC and NO_3^- due to upward transport of nutrient-rich deep waters. Wind-driven circulation further contributes by increasing gas exchange rates and driving upwelling, which enriches surface waters with DIC, AOU, and NO_3^- .

In addition to physical processes' influence, the increase in NO_3^- at 40°S, especially in the first 500 m (Figure 3i), could reflect increased remineralization, possibly caused by enhanced organic matter export at 40°S (Boyd & Trull, 2007; Sarmiento, 2006). This should be accompanied by an $[O_2]$ decline in the same waters, increasing AOU, which in the surface mixed layer could then be attenuated by oxygen exchange with the atmosphere. Consequently, $C_{soft(AOU)}$ in the near surface would underestimate the actual change in C_{soft} , while in deeper waters beneath the mixed layer (without the associated air-sea gas exchange causing a bias in $C_{soft(AOU)}$), $C_{soft(AOU)}$ would be consistent with $C_{soft(NO_3)}$ (Figure 4). However, $C_{soft(AOU)}$ and $C_{soft(NO_3)}$ differ significantly from each other down to around 500 m (or 27.15 kg m^{-3}), which is deeper than the mixed layer here (54 m, $\gamma^n \approx 25.8$), so this is not a complete explanation for the discrepancy.

In shallower waters, other factors might explain the observed discrepancy between $\Delta C_{soft(AOU)}$ and $\Delta C_{soft(NO_3)}$, such as changes in the stoichiometry of organic matter. Variations from standard plankton biomass elemental ratios (i.e., the Redfield ratio; Redfield, 1958) have been observed spatially and temporally (Inomura et al., 2022; Martiny et al., 2013; Tanioka & Matsumoto, 2020), which may contribute to the difference between $C_{soft(AOU)}$ and $C_{soft(NO_3)}$ observed here (Figure 4; Anderson & Sarmiento, 1994). South of 40°S, the carbon to nitrogen (C:N) ratio is closely aligned with the Redfield ratio (Johnson et al., 2022). Conversely, north of 40°S, dissolved organic matter (DOM) has higher C:N ratios, with 10.4 ± 4.1 at 30°S, 14.0 ± 4.8 at 35°S, and 9.7 ± 1.7 at 40°S, all surpassing the Redfield ratio of 6.6 (Johnson et al., 2022). The elevated C:N observed at 40°S could then also result from a changing balance between water masses from the north and the south. Organic matter with higher C:N requires more oxygen for its remineralization, thereby accelerating the consumption of dissolved oxygen (Matsumoto et al., 2020; Szewczyk et al., 2023). Climate change is expected to cause significant stoichiometric shifts in plankton biomass, with warmer temperatures and rising CO_2 levels promoting higher C:P and N:P ratios (Ayo et al., 2017; DeVries, 2018; DeVries et al., 2017; Toseland et al., 2013; van de Waal et al., 2010; Yvon-Durocher et al., 2017) which may result in increased oxygen consumption during remineralization and thus higher AOU (DeVries, 2018). Additionally, variability in the rates of change across the water column could result from changing remineralization depth due to temperature-dependent remineralization (Boscolo-Galazzo et al., 2018, 2021), shifts in particle injection pumps (Boyd et al., 2019), or changes in DOC remineralization with circulation or degradation dynamics (Hansell, 2013).

The enhanced remineralization and possible changes in C:N observed at 40°S may be due to increased diatom populations further south (Arrigo et al., 1999, 2015; Soppa et al., 2016). Diatoms are reliant on silicate and significant nitrate consumers. Also, enhanced inputs of iron are known to stimulate diatom productivity (Section 4.1.1). Diatoms secrete transparent exopolymeric particles (TEP), which act like glue to hold aggregates together, leading to faster sinking of marine particles and a more efficient biological pump (J. Chen & Thornton, 2015; Toullec & Moriceau, 2018). TEP production also increases the C:N ratio (Kim et al., 2021; Passow, 2002). Boosted silicate availability, most likely due to enhanced upwelling and glacial runoff (Henley et al., 2020), may accordingly reduce NO_3^- levels at the sea surface and increase NO_3^- and AOU in the AAIW due to enhanced sinking and remineralization of algal diatom biomass. If driven by diatoms, we would expect the increased NO_3^- and AOU in deeper waters to be accompanied by an increase in [Si] in AAIW (Cael et al., 2021), which is what we observed at 40°S (Figure 3k). Either through remote or via locally enhanced productivity and increased abundance of diatoms, caused by higher iron and [Si], could in part be responsible for the changes in the BCP that we observe.

4.3. Consequences of Changes in the Biological Carbon Pump

4.3.1. Implications for Marine CO_2 Sink

Previous studies have shown that the South Atlantic Ocean is a sink of atmospheric CO_2 , with an average net air-to-sea CO_2 flux of 0.3 Pg C yr^{-1} (Takahashi et al., 1997, 2002). Recent work has highlighted strong seasonality in the South Atlantic CO_2 flux, acting as a strong sink during the spring when most primary production takes place, and shifting toward a source during autumn (Lencina-Avila et al., 2016; Padin et al., 2010). Monitoring whether the South Atlantic acts as a source or sink of CO_2 is vital for understanding its role in the global carbon cycle as it

directly influences the atmospheric CO₂ concentration and therefore the Earth's climate (Lencina-Avila et al., 2016; Padin et al., 2010). As these variations are connected to the biological productivity of the South Atlantic Ocean, it is especially important to understand the biological pump's contribution to changes in DIC, which has a direct impact on whether the ocean basin acts as a source or sink of CO₂.

The observed increase in surface ocean *p*CO₂ at 40°S is consistent with the growth in atmospheric *p*CO₂ (Figure 3e). If the changes in the BCP were to lead seawater *p*CO₂ to rise faster or slower than atmospheric *p*CO₂, this would cause a decrease or increase respectively of the ocean CO₂ sink (DeVries et al., 2017). However, the enhanced BCP that we find suggests that while biological processes are contributing to increased carbon sequestration (Section 4.2), physical processes such as increased mixing may counteract this effect by bringing CO₂-rich waters to the surface (Section 4.1), thereby maintaining the observed consistency with atmospheric *p*CO₂ growth.

4.3.2. Carbon Export and BCP Role in Climate Responses

Our findings at 40°S indicate that the biological pump is experiencing shifts due to a mix of changes in water mass formation regions and biological factors across the Atlantic Ocean. Model results suggest that the anticipated increase in iron supply and improved light availability for phytoplankton—owing to enhanced near-surface stratification and prolonged ice-free periods—will likely amplify primary production. This could, in turn, boost carbon export around the Antarctic region (Henley et al., 2020). These observations at 40°S may be an early sign of the impact of global change in the Southern Ocean, assuming that our estimates have accounted for the effects of advection and transport on the BCP.

Building on these observations, the depth-integrated increase in organic carbon export flux was between 0.004 ± 0.002 (C_{soft(AOU)}) and 0.006 ± 0.002 GtC yr⁻¹ (C_{soft(NO₃)}) for the study area, while the change in the carbonate pump was negligible ($<2 \times 1010$ GtC yr⁻¹ ± 0.001 GtC yr⁻¹). For comparison, a recent estimate of global export production accounting for both POC and DOC was 8.37 ± 1.57 GtC yr⁻¹ (Sulpis et al., 2023), which amounts to 0.017 ± 0.003 GtC yr⁻¹ when scaled down to the study area. This suggests that the increase in the BCP that we observed could represent an increase in the amount of carbon remineralized by 23%–35% each year. However, if the study region is a biological “hotspot” with high baseline productivity and remineralization, the global average would be underestimated for this region, so the 23%–35% increase is an upper bound. The anthropogenic component remained the highest, with $+0.008 \pm 0.003$ GtC yr⁻¹ for C_{anth(AOU)} and $+0.007 \pm 0.003$ GtC yr⁻¹ for C_{anth(NO₃)} which, compared to the global average of 0.005 ± 0.001 for 2022 (2.8 ± 0.4 GtC yr⁻¹; Friedlingstein et al., 2023), implies that the study area also has a higher anthropogenic CO₂ uptake relative to the global average.

It is crucial to note that these estimates assume that changes in export production are fully reflected in changes in remineralization rates at these depths. However, the full variance of these processes throughout the entire water column might not be captured, and changes in remineralization and circulation can also influence C_{soft} independently of direct changes in export production. Thus, while our results show an increase in local export production, it is important to differentiate between export production, carbon remineralization, and sequestration, as these differences in C_{soft} may not correspond directly to changes in export production over these timescales. Additionally, these observations should not be interpreted as being indicative of a consistent global trend, as highlighted by Henson et al. (2022), who found no consistent global increase in export production. While we cannot precisely quantify the influence of each driver on C_{soft} due to the complexity of these interactions, both C_{soft(AOU)} and C_{soft(NO₃)} estimates consistently indicate an increase, suggesting an overall intensification of the BCP.

Regardless of its size as a fraction of the baseline remineralization rate, the magnitude of the DIC increase associated with the BCP at 40°S is of the same order of magnitude as the anthropogenic increase in DIC (Figures 3d and 4). Thus, while changes in the biological carbon pump are often considered less significant than anthropogenic CO₂ uptake, they should still be considered in global carbon budgets.

4.3.3. Anthropogenic Influence and CO₂ Dynamics

Since the early 1960s, the primary driver behind the long-term trend in the ocean carbon sink has been the rising uptake of anthropogenic CO₂ (Gruber et al., 2023). From 2004 to 2019, the global oceanic DIC pool increased at

an average rate of $3.2 \pm 0.7 \text{ Pg C yr}^{-1}$, with no statistically detectable difference between the total DIC change and C_{anth} accumulation between 2004 and 2020 (Keppler et al., 2023). This implies no global net change in C_{soft} but does not rule out a spatial redistribution driven by various factors including ocean warming, alterations in marine biology, and other physical changes within the oceans, as discussed above for 40°S. Thus, our study may be an example of this redistribution effect—as also witnessed in the northeast Atlantic Ocean (Humphreys et al., 2016). The role of the biological carbon pump relative to anthropogenic CO_2 uptake in changing the marine DIC pool may be more important than previously thought.

However, these findings must be interpreted with caution due to the seasonal biases inherent to the data collection and the variable nature of nutrient concentrations, which challenge the reliability of estimated long-term change estimates. Observations, primarily during the austral summer, may not fully capture the year-round dynamics of the BCP, nutrient fluxes, or the changing properties of water masses, potentially skewing our understanding of the BCP's role in DIC variations and affecting the capacity for C_{anth} uptake by the ocean. Furthermore, Behrenfeld et al. (2006) highlighted the significant impact of climate-driven variability, such as the El Niño-Southern Oscillation, on surface phytoplankton abundance. Thus, this underscores the importance of considering both seasonal and interannual variability.

Finally, although current observations reveal negligible net changes in global C_{soft} , it is essential to acknowledge that model projections indicate potential divergences in these trends on centennial timescales, largely due to the long residence times of C_{soft} (Cao & Zhang, 2017; Siegel et al., 2023). Additionally, the spatial redistribution of C_{soft} may modify nutrient dynamics and alter residence times, which, over extended periods, are likely to result in significant alterations in both C_{soft} and export production (Henson et al., 2022; Wilson et al., 2022).

5. Conclusions

At 40°S in the Atlantic Ocean, from 1972 to 2023, DIC increased, down to approximately 2,000 m ($\gamma^n \approx 27.9 \text{ kg m}^{-3}$), with a near-surface maximum rate of $1.44 \pm 0.11 \mu\text{mol kg}^{-1} \text{ yr}^{-1}$. Although at least half of this change can be attributed to anthropogenic CO_2 accumulation, the intensification of the BCP in this region is evident, with contributions ranging from $0.20 \pm 0.03 \mu\text{mol kg}^{-1} \text{ yr}^{-1}$ to $0.85 \pm 0.07 \mu\text{mol kg}^{-1} \text{ yr}^{-1}$ (using $C_{\text{soft(AOU)}}$ and $C_{\text{soft(NO}_3)}$, respectively). While we cannot definitively select one soft tissue pump estimation over another due to inherent uncertainties in the influence of different processes, the concurrent increase in both $C_{\text{soft(AOU)}}$ and $C_{\text{soft(NO}_3)}$ estimates, despite their discrepancies, consistently points toward an intensification of the BCP at this location.

This study at 40°S in the Southern Ocean has also provided valuable insights into the complex interplay of physical, biological, and anthropogenic factors influencing the dynamics of the BCP. Our investigation suggests that no single factor alone accounts for the observed changes in the efficiency of the BCP; rather, it is the cumulative effect of changes in sea ice dynamics, ocean stratification, wind patterns, and biological activity, including photosynthesis and organic matter degradation, which themselves are impacted by ongoing global changes that are not in equilibrium. The precise quantification of each driver's influence and interplay should be the focus of future research.

These findings also highlight the implications of these changes for CO_2 sink behavior and carbon export. The observed increase in $p\text{CO}_2$ at 40°S and the depth-integrated increase in organic carbon export flux imply shifts in the region's role in the global carbon cycle. These shifts are not only vital for understanding the evolution of atmospheric CO_2 concentration but also for potential future climate change mitigation efforts.

Finally, the study addresses the anthropogenic influence on CO_2 dynamics. The increasing uptake of anthropogenic CO_2 and its interplay with the biological carbon pump indicate that the role of the BCP in DIC changes might be more intricate and significant than previously assumed. This highlights the need for comprehensive consideration of the biological carbon pump in global carbon budgets and climate models.

Conflict of Interest

The authors declare no conflicts of interest relevant to this study.

Data Availability Statement

All data and code used in this analysis are available as a GitHub repository (Delaigue, 2024). PyCO2SYS v1.8.1 (Humphreys et al., 2022) was used to solve for the carbonate system (Humphreys et al., 2024). The Gibbs-SeaWater (GSW) Oceanographic Toolbox was used to calculate neutral density (McDougall & Barker, 2011). Figures were made with Python version 3.9 (van Rossum & Drake Jr, 2009).

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