Additional objectives of Deep Argo may include dissolved oxygen measurements, and testing of ocean mixing and optical scattering sensors. The development of an emerging ocean bathymetry dataset using Deep Argo measurements is also described.

KEYWORDS

deep ocean, ocean observation, ocean heat content (OHC), sea level (SL), ocean deoxygenation, bathymetry accuracy, ocean mixing, sediment transport

1 Introduction

Earthi Energy Imbalance (EEI), the rate of net solar energy absorbed by the Earth, shows an increased trend in the past two decades at least partly as a result of increased atmospheric greenhouse gases (GHG; Loeb et al., 2021). Unless GHG emissions are sharply reduced, global warming is likely to exceed 1.5-21/2 above pre-industrial levels. The 1.51/2 threshold was recognized in the Paris Agreement as synonymous with extreme weather events, warmer ocean waters, and sea level rise, threatening human safety and ecosystem health (Allen et al., 2019). Due to its high heat capacity, the ocean delays the atmospheric warming, absorbing about 90% of the EEI (von Schuckmann et al., 2023). Together, the expansion of seawater from increasing temperature, import of continental water from Greenland and Antarctic ice sheet mass loss, and the melt of mountain glaciers, have resulted in a global mean sea level (GMSL) rise of 15-25 cm over 1901-2018 (1.28 to 2.17 mm per year; Fox-Kemper et al., 2021). The rate of rise has accelerated in the past two decades (Nerem et al., 2018) and reached 3.58 1025 mm per year over 2005-2015 (Llovel et al., 2019). Model projections indicate that this acceleration is likely to continue through the century (Fox-Kemper et al., 2021), increasing the extent and severity of ooding and coastal erosion (Martyr-Koller et al., 2021). Current estimates of global ocean heat content (OHC) inferred from sea-going platforms and sea level change assessed from satellite altimetry, lack the desired spatial resolution and precision to generate accurate projections of EEI response to climate policies (Meyssignac et al., 2019; Palmer, 2017; WCRP Global Sea Level Budget Group, 2018).

While the ocean sampling of the upper 2000 m has dramatically increased due to the revolutionary contributions of autonomous platforms, the status of deep-sea observing has remained fragmentary (Meyssignac et al., 2019; von Schuckmann et al., 2023) and opportunistic. Although the deep ocean below 2000 m represents half of the total ocean volume, only 10% of historical temperature and salinity pro les extend below 2000 m (Garcia et al., 2019). Because deep-ocean measurements are unevenly distributed in space and time, regional-to-global and seasonal-to-interannual variations are poorly resolved, and spurious patterns of ocean temperature and salinity uctuations can occur (Johnson and Lyman, 2020). Global decadal variations in deep-ocean volume due to temperature (thermosteric) changes contribute 10% of the GMSL budget (Chang et al., 2019) but the impact is signi cantly

higher (~30%) in the Southern Ocean where abyssal warming is strongest (Purkey and Johnson, 2010). The GMSL budget closes between 2005 and 2015 when using an approximate estimate of decadal variation of the ocean volume due to density (steric) changes (Llovel et al., 2019; Chen et al., 2022), but appears no longer closed between 2016 and 2019 (Chen et al., 2020). While part of the non-closure of the GMSL budget after 2015 can be explained by salty drift of conductivity-temperature-depth (CTD) sensors on some Argo oats, potential drift of satellite altimeters, and noise contamination of Gravity Recovery and Climate Experiment (GRACE/GRACE-FO) accelerometer measurements, the sparse sampling of the deep ocean, especially in the Southern Hemisphere, is likely another contributor (Barnoud et al., 2021). Even during 2005-2015 when the sea level budget closes at the global scale, a mismatch between sea level from altimetry and the sum of GRACE mass addition, Argo upper-ocean steric sea level, and shipboard deep-ocean steric sea level is observed at local to subbasin scale (Royston et al., 2020).

General circulation models are essential tools to monitor and forecast the in uence of ocean-atmosphere coupling on weather and climate (Stammer et al., 2019; Lellouche et al., 2021), but model drift in the deep ocean limits their ability to predict ocean heat and freshwater content (Palmer et al., 2011). Correction techniques developed to force model elds towards speci c deep-ocean conditions have shown discrepancies in simulated ocean temperature and salinity compared with observations (Uotila et al., 2019). The addition of global and frequent deep-ocean observations improves the representation of deep-ocean processes and can prevent unrealistic drifts (Palmer, 2017). Furthermore, deep-ocean observations decrease uncertainties in water mass properties (Gasparin et al., 2019; Gasparin et al., 2020), reduce bias in upper-ocean decadal predictability (e.g., Sevellec and Fedorov, 2013; Levin et al., 2019), and improve the consistency of assimilated in situ and satellite observations (Abdalla et al., 2021).

Partly owing to the decrease in oxygen solubility with ocean warming, dissolved oxygen (DO) concentrations have decreased (Schmidtko et al., 2017) and oceanic regions with low-oxygen concentration, termed quasi-permanent oxygen minimum zones (OMZs), have grown over the last ve decades. The spread of these inhospitable areas decreases biological diversity, reduces physiological adaptation, and depletes sh stocks (Garcon et al., 2019). Earth model simulations predict an expansion of OMZs (Busecke et al., 2022) and intensi cation of oxygen decline from the

surface to abyssal depths in the next decade (Kwiatkowski et al., 2020). Due to the scarcity of deep-ocean oxygen observations, the response of low-oxygen habitat size and distribution to physical and biogeochemical processes remains poorly understood (Levin, 2018; Oschlies et al., 2018), and the ability of marine biogeochemical models to simulate the response and feedback of marine ecosystems to climate change is limited (Seferian et al., 2020).

Deep Argo is an ongoing expansion of the Argo Program using autonomous robotic oats with the ability to cycle between the surface and the sea oor every 10 days, and transmit data via satellites (Roemmich et al., 2019a). Deep Argo is the deep component of the OneArgo design, endorsed by the U.N. Ocean Decade (Owens et al., 2022). At present, there are ~200 Deep Argo oats sampling some of the deepest regions of the ocean (Figure 1). The Argo community recommends increasing the number of active deep oats to form a 5162512 lobal array of 1200 Deep Argo the seasonally ice-free global ocean deeper than 2000 m. The capacity of the fully-implemented Deep Argo array to reduce errors in decadal trends of deep OHC and deep-ocean thermal expansion (Johnson et al., 2015; Meyssignac et al., 2019) is recognized by the EEI community (von Schuckmann et al., 2020; von Schuckmann et al., 2023) and the World Climate Research Program (WCRP) global sea level budget group (WCRP Global Sea Level Budget Group, 2018; Cazenave et al., 2019). Other bene ts of Deep Argo are to establish relationships between uctuations of the deep meridional overturning circulation and changes in ocean temperature and salinity and their representations in ocean reanalyzes and forecasts (Gasparin et al., 2020). The value of adding sensors on Deep Argo oats to assess full-depth ocean oxygen content, mixing, and sediment transport, and the capacity of Deep Argo oats to improve global ocean bathymetry are under study. Deep Argo oats are expected to play a key role in the Deep Ocean Observing Strategy (DOOS) initiative, a growing effort by the scienti c community to improve the deep-ocean observing system (Levin et al., 2019).

2 Requirements and technology status

2.1 Float and sensor requirements

In accordance with Argole practices, Deep Argo data are shared publicly in near real-time, usually in less than 24 hours. A further quality-controlled version is made available within 6 months on the two Argo Global Data Assembly Centers (GDAC) in netCDF les, as well as inserted onto the Global Telecommunications System (GTS) in the Binary Universal Form for the Representation (BUFR) format. For consistency with the Core Argo mission, Deep Argo oats operate on a nominal 10-day cycle between the surface and the sea oor (4000-m or 6000-m depth, depending on the oat model). For Deep Argo oats pro ling on descent and ascent, measurements on ascent should be collected at least in the upper 1000 m to provide near real-time data to ocean forecasting and seasonal prediction communities.

Required accuracies for Deep Argo salinity, temperature, and pressure measurements are to approach 2002 PSS-78, 2001 2 and 2 and 2 abar respectively (Roemmich et al., 2019a), in line with shipboard measurements from the Global Ocean Ship-Based Hydrographic Investigations Program (GOSHIP; Katsumata et al., 2022). It is recommended that technology development continues to reach targeted values. With salinity accuracy target ve times more stringent than Core (0-2000 m) Argo (Wong et al., 2020), Deep Argo can ef ciently supplement the high-quality shipboard reference database used for Core Argo quality control and has therefore the capacity to improve the post-calibration of salinity measurements. The accuracy, precision, and vertical resolution of

FIGURE 1

Locations of the 197 Deep Argo oats active in May 2023. This gure shows locations of 4000 m capable Deep Arvor and Deep NINJA, and 6000 m capable Deep SOLO and Deep APEX oats, color coded by national program. The background colors indicate ocean bottom depth: <2000 m (white), 2000-3000 m (light gray), 3000-4000 m (light blue), 4000-5000 m (blue), and >5000 m (dark gray). Data courtesy of OceanOPS.

the Deep Argo CTD sensor in uence the temperature-salinity relation, and estimates of steric height and vertical strati cation as outlined below.

Steric height is vital for sea level budget analyses (Cazenave et al., 2019) as well as geostrophic circulation studies (Zilberman et al., 2020). Errors the size of Deep Argo target accuracies (12002 PSS-78, 12001) and 121 and 121 applied to full-depth (0 12000 m) pro les at 2012 in the Brazil Basin result in sea level errors of about 120, 120, 120, and 120 cm, as well as depth-integrated transport errors (assuming zero velocity at the bottom) of 121, 120, and 125 10 cm s⁻¹ for salinity, temperature, and pressure, respectively. With these targeted values, salinity accuracy has the largest impact on sea level, geostrophic velocity, and transport error budgets.

Mean temperature-salinity relations are used to diagnose deep water-mass distributions (Johnson, 2008) and ocean circulation (Zilberman et al., 2020), and their variations are important in monitoring changes in deep and bottom water properties and formation rates (Purkey et al., 2019). Again, salinity accuracy is likely the limiting factor, but decadal changes in the temperature-salinity relationships of Antarctic Bottom Water (e.g. van Wijk and Rintoul., 2014) and the constituents of North Atlantic Deep Water (Yashayaev, 2007) near their sources are several times Deep Argo target accuracies, and therefore easily quanti able.

Strati cation variations on small vertical scales are needed to estimate mixing related to internal waves and tides (Whalen et al., 2012), while vertical strati cation on longer length scales is appropriate to study features such as variability in the pycnocline between deep and bottom waters, which may change ocean mixing and diffusion rates (Zhang et al., 2021). For compatibility with Core Argo sampling recommendation in the upper 2000 m, Deep Argo oat longevity, and ocean mixing parameterizations, the recommended vertical resolution of Deep Argo pro les is 2-dbar bin-averaging in the upper 2000 m, and may increase to 10-dbar, or even 25-dbar, vertical resolution below 2000 m.

2.2 Readiness of oat and sensor technology

Deep Argo oat technology has been extensively tested in regional pilot arrays (Figure 1). Several oat designs have proven their ability to collect deep-ocean pro les to 4000 m (e.g., Deep NINJA and Deep Arvor) and 6000 m (e.g., Deep SOLO and Deep APEX) (Kobayashi et al., 2013; Petzrick et al., 2014; Le Reste et al., 2016; Roemmich et al., 2019b) and new oat designs are under development to further increase the diversity of the Deep Argo eet.

Deep Argo CTD sensor speci cations of temperature, salinity, and pressure accuracies approach the targeted values described in section 2.1, and speci cations of stability are compatible with the Deep Argo targets. Both extended-depth SBE41-CP and SBE61 models from SeaBird Scienti c (SBS) are operational. A pilot Deep Argo CTD instrument from RBR is under testing. The ability of the SBE61 to reach the targeted temperature accuracy of 2001 has been demonstrated from comparisons of shipboard rosettemounted SBE-61 CTDs with shipboard (SBE-911) CTD observations (Roemmich et al., 2019a). Work has been achieved

to improve our understanding of the impact of temperature and pressure variations on conductivity and pressure measurements (Kobayashi et al., 2021). A compressibility correction of the conductivity cell used on the extended-depth SBE41-CP and SBE61 has been successfully implemented, reducing salinity bias observed in the eld from comparisons of Deep Argo-mounted extended-depth SBE41-CP and SBE-61 with shipboard reference pro les to \$2002 PSS-78 (Foppert et al., 2021; Walicka et al., 2022), consistent with the Deep Argo target. Once the pressure dependent salinity bias is adjusted, a time-varying correction of salinity can be applied as long as salinity drift is linear in time (Owens and Wong, 2009; Cabanes et al., 2016). A description of quality control procedures and best practices for analysis of Deep Argo temperature, salinity and pressure is provided in the report by Walicka et al. (2022). New pressure sensors with accuracy and stability speci cations exceeding the Deep Argo target are undergoing testing in the eld. Although reaching the ambitious accuracy targets described in section 2.1 across all Deep Argo CTD sensors is a challenging task, ongoing work is showing great potential for success. To that end, it is essential that research institutions continue independent assessment of sensor accuracy and stability, the sensor sability to maintain targeted accuracy over the oat lifetime, using Deep Argo oat comparisons with highquality shipboard reference data, and maintain collaborative work with the industry to optimize calibration and data quality control procedures. Maintaining a diversity of distinct Deep Argo oats and sensors from several manufacturers is encouraged to avoid supply chain issues, and to effectively identify, understand, and resolve potential oat and sensor failure.

3 Value of Deep Argo data

The amount of deep-ocean pro les collected from Deep Argo oats in the South Australian, Australian Antarctic, Southwest Paci c, and Brazil basins in less than 7 years approaches that of deep-ocean data accumulated from ships over the past 70 years (Figure 2). This new dataset has enabled breakthroughs in our understanding of OHC, ocean circulation, and sea level change.

For instance, strong decadal-scale warming rates of the cold and dense Antarctic Bottom Water (AABW) that spreads over much of the abyssal ocean (depth > 4000 m) have been quanti ed combining Deep Argo data with historical data in the Brazil Basin (Johnson et al., 2020), and Argentine Basin (Johnson, 2022) with trend estimates that are up-to-date and several times more certain than those from repeat hydrography alone (Figures 3A, B). Deep Argo data alone are measuring shorter time-scale trends, such as accelerated warming in the bottom waters of Southwest Paci c Basin since 2014 (Johnson et al., 2019), with increasing certainty as the record length increases (Figure 3C). Other scienti c advances using Deep Argo data with other data sources include a warming-to-cooling reversal of the Lower North Atlantic Deep Water observed in the subpolar North Atlantic in the mid-2010s (Desbruy) et al., 2022; Figure 3D).

Small clusters of Deep Argo oats have demonstrated their ability to detect the reversal of a long-term freshening trend in the

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FIGURE 2

The sampling density per 5 matitude by 5 monitude of temperature and salinity pro les deeper than 2000 m. (A) Deep Argo pro les in regional pilot arrays deployed between 2014 and 2023. (B) Historical non-Argo pro les from the World Ocean Database (WOD) 2018 collected between 1950 and 2023 (Garcia et al., 2019).

Australian Antarctic Basin, at higher temporal and spatial resolution than historical shipboard CTD occupations (Thomas et al., 2020; Foppert et al., 2021). By providing simultaneous temperature and salinity pro les over broad swaths of ocean basins, Deep Argo has proven effective at mapping large-scale circulation pathways of dense water masses formed near Antarctica (Thomas et al., 2020; Foppert et al., 2021), in the subpolar North Atlantic Ocean (Racape et al., 2019; Petit et al., 2022), and in the Southwest Paci c Ocean (Zilberman et al., 2020; Figures 4A, B) with unprecedented detail. Seasonal uctuations in the recirculation of the Deep Western Boundary Current of the Southwest Paci c Basin, and associated forcing mechanisms, were detected for the rst time (Zilberman et al., 2020; Figures 4C), as was a slowdown of ow of AABW into the Argentine Basin (Johnson, 2022). In the North Atlantic Ocean, Deep Argo pro les have been added to the In Situ Analysis System (ISAS, Gaillard et al., 2016; Kolodziejczyk et al., 2021) to provide monthly gridded elds of full-depth temperature and salinity, and to improve estimates of subannual-to-annual full-depth regional steric sea level change. Deep Argo data have even been used to describe regional patterns of deep internal wave activity (Johnson, 2022).

The fully implemented global Deep Argo 1200- oat array will diminish the spread in plausible regional-to-basin-scale OHC trends of the deep ocean due to sparse historical sampling, while lling temporal gaps at subseasonal to interannual time scales not often sampled by repeat hydrographic sections (Desbruytés et al., 2016). Deep Argo thermosteric and halosteric sea level sampling will decrease the observed regional-to-basin-scale mismatch between sea level from altimetry and the sum of GRACE mass addition, Argo upper-ocean steric sea level, and shipboard deep-ocean steric sea level (Royston et al., 2020). Deep Argo has the capacity to reduce errors in global decadal trends of deep OHC from the 2006 W m -2 (Meyssignac et al., 2019) and deep-ocean thermal expansion from 23 to 23 mm decade -1 (Johnson et al., 2015; Desbruytés et al., 2016). As the array expands, Deep Argo will become increasingly useful to improve knowledge of the



C D

Mean deep-ocean temperature trends using Deep Argo. Temperature trends are shown in (m@yr -1) vs. pressure (lines) with 5%95% con dence limits (shading). (A) Temperature trends in the Argentine Basin computed by comparing all pairs of Deep Argo temperature pro les collected from January 2021 to April 2023 with nearby historical pro les collected from 1972 to 1998 (blue, updated following Johnson, 2022) and computed from repeat hydrography sections carried out in the Argentine Basin between the early 1990% and the mid-2010% (orange, from Desbruy) et al., 2016). (B) Temperature trends in the Brazil Basin computed by comparing all pairs of Deep Argo pro les reported from May 2019 to April 2023 within nearby WOCE pro les from 1989-1995 (blue, updated following Johnson et al., 2021) and computed from repeat hydrography sections carried out in the Brazil Basin between the early 1990% and the mid-2010% (orange, from Desbruy) et al., 2016). (C) Temperature trends in the Southwest Paci c Basin evaluated from Deep Argo data collected from June 2014 to April 2023 (updated from Johnson et al., 2019). (D) Temperature trends in the Irminger Sea evaluated from Deep Argo during 2016 % 2021 (from Desbruy) et al., 2022). This last trend is density-compensated and only the layer > 36.94 m⁻³ occupied by Lower North Atlantic Deep Water is considered (note the different x-axis scale).

regional-to-basin-scale structure and subseasonal-to-interannual uctuations of the large-scale deep-ocean circulation, and advance understanding of the drivers for these changes. In particular, Deep Argo will provide a new means to validate numerical model projections of the impact of meltwater release from Greenland and Antarctica on the deep-ocean overturning circulation (Swingedouw et al., 2022; Li et al., 2023; Zhou et al., 2023).

4 Global implementation

To be sustainable, it is essential that the global implementation of the Deep Argo array stems from achievable long-term resources of international Argo partners and production capability of the oat and sensor manufacturers. Since each Deep Argo oat replaces a Core Argo oat in the OneArgo design (Roemmich et al., 2019a; Owens et al., 2022), a minimum lifetime of at least 4 years, approaching the averaged 5.5-year Core Argo oat longevity, is desired across all Deep Argo oat models (Zilberman et al., 2019). Assuming an averaged Deep Argo oat longevity of 4 years, implementing a 1200- oat global array size would require 300

oat deployments per year among all international partners. Some active Deep Argo oat models have already achieved a longevity exceeding 5 years. Engineering advances will continue to increase the averaged Deep Argo oat lifetime in order to reduce the size of annual oat deployments needed to re-seed the array. The fully implemented Deep Argo array will require additional support from international Deep Argo partners as the averaged capital purchase of Deep Argo oats across all active models is two to three times that of Core Argo. The data quality control and deployment costs for Deep and Core Argo oats are similar.

Additional recommendations should be considered in order to optimize deep-ocean sampling cost and ef ciency. Prior to reaching the 1200- oat design target, Deep Argo oat deployments should prioritize deep (> 2000 m) regions of the subpolar North Atlantic, Southern Ocean, and tropical oceans, and abyssal (> 4000 m) plains of the ocean interior, where highest deep-ocean temperature uctuations have been observed. Deep Argo oats deployed in these regions may park at depths greater than 1000 m to minimize oat dispersion, hence optimizing their value. For instance, programming parking depth to coincide with reduced background ow such as within 500-m above the sea oor in abyssal plains, and

A B

С

FIGURE 4
Spatial structure and seasonal variability of the deep-ocean circulation in the Southwest Pacic Cocean using Deep Argo. (A) Positions of Deep Argo pro les collected from January 2016 to December 2019 are indicated in squared red symbols. The shaded gray color indicates bottom depth shallower than 4800 m depth, and the white background color shows regions deeper than 4800 m. The Samoan Islands (SI), Samoan Basin (SB), Manihiki Plateau (MP), Penrhyn Basin (PB), Louisville Ridge (LR, thin black line), Tonga Kermadec Ridge (TKR, thick black line), Hikurangi Plateau (HP), Chatham Rise (CR, dashed black line), and New Zealand are indicated (from Zilberman et al., 2020). (B) Absolute transport function integrated between 1800 and 4100 dbar for 2016 2019. The white background color indicates regions shallower than 4800 m (from Zilberman et al., 2020). (C) Seasonal variability of dynamic height referenced to 4800 dbar between 0 and 4800 dbar between 46 and 11 2 are 2016 2019 (from Zilberman et al., 2020).

above deep boundary currents near topography, has proven an effective way to limit horizontal displacement during parking (Zilberman et al., 2020). Floats may also be programmed to park at the core depth of local deep water masses in order to assess regional circulation (Racape et al., 2019). Once a global array is achieved, it is recommended that all Deep Argo oats should park at 1000 m in order to facilitate homogeneous oat sampling (Roemmich et al., 2019a) and increase the velocity database from Argo oat trajectories to reduce sampling error (Zilberman et al., 2023).

5 Ongoing evolution of the Deep Argo array

Sensors to measure ocean properties additional to temperature, salinity, and pressure are under testing to assess their performance in the eld and increase value of the Deep Argo data set. Any additional sensors should be compatible with the Deep Argo oat longevity target and should not impede the achievement of a global Deep Argo array. The opportunistic measurement of bathymetry from Deep Argo oats, which does not require supplementary sensor measurements, is under evaluation.

5.1 Dissolved oxygen

Full-depth Deep Argo DO data can help constrain oceanic sources and sinks of oxygen (Riser and Johnson, 2008) and provide better assessments and predictions of global ocean deoxygenation (Schmidtko et al., 2017). Combined with Deep Argo temperature and salinity, DO measurements can improve our understanding of water-mass formation (e.g., Johnson, 2008), ocean ventilation (Piron et al., 2016), ocean circulation and mixing (Racape et al., 2019) and the impact of these processes on oxygen change in the deep ocean (Coppola et al., 2017; Ulses et al., 2021). Using temperature, salinity, and DO in transfer functions allows estimates of concentrations of nutrients, carbonate system parameters (Sauzille et al., 2017), and ultimately anthropogenic carbon storage and export. About 30% of active Deep Argo oats are equipped with DO sensors. The best accuracy reached in the eld by Deep Argo DO sensors is 1-2 it folkg ⁻¹ and requires a careful correction of the data to account for sensor drift (Bittig et al., 2018; Bittig et al., 2019), time response (Gordon et al., 2020) and pressure dependent response of the sensor (Bittig et al., 2018; Racape et al., 2019). The current accuracy of Deep Argo DO sensors is suitable for all the above-mentioned applications, but needs to be increased to 0.5 it to resolve deoxygenation trends (Gregoire et al, 2021).

Work is underway to bring DO accuracy to this target and to determine through observing system simulation experiments (OSSEs) an optimal deployment strategy, and the required amount of Deep Argo oats equipped with DO sensors. Deep Argo will supplement DO data from repeat shipboard hydrography (Testor et al., 2018), and in return, shipboard observations will help validate and calibrate Deep Argo pro les (Racape et al., 2019). It is expected that Deep Argo oats with DO sensors could play a key role in the expansion of the Global Ocean Oxygen Database and Atlas (Gregoire et al., 2021).

5.2 Bathymetry

Ocean bathymetry has a fundamental impact on the pathway and strength of ocean currents (Wilson et al., 2022), and amplitude of ocean mixing (Mashayek et al., 2017), and can in uence heat and freshwater exchange between the ocean and atmosphere (de Boer et al., 2022). Variations in the shape of the sea oor can dictate the magnitude and frequency of earthquakes (Passarelli et al., 2022) and the generation and propagation of tsunamis (Salaree and Okal, 2020). Less than 24% of the sea oor topography from the General Bathymetric Chart of the Ocean (GEBCO) dataset is constrained by shipboard sounding measurements or other direct measurements, and is limited mostly to Exclusive Economic Zones (Tozer et al., 2019; W1/2 et al., 2019; GEBCO Bathymetric Compilation Group, 2022). Deep Argo oats are not equipped with an echosounder, but bottom depth can be estimated when near-constant pressure values are recorded over a set period of time when the oat contacts the sea oor. The main challenge in estimating Argo-based bathymetry is the horizontal displacement that occurs between the grounding position at depth and the position obtained at the surface, once a connection between the oat and satellite positioning system is established (Ollitrault and Rannou, 2013). Estimated horizontal errors using the Bluelink ReANalysis 2020 (BRAN2020; Chamberlain et al., 2021) and GLobal Ocean ReanalYsiS (GLORYS12; Lellouche et al., 2021) ocean reanalyzes are 1-2 km in widespread regions of the deep-ocean interior, 2-8 km in equatorial regions, and reach 8-18 km at ocean western boundaries and near topography where ocean currents are strongest. Preliminary comparisons between historical Argo and GEBCO bathymetry sourced from echo sounding measurements show encouraging agreement (van Wijk et al., 2022). Based on the averaged Deep Argo oat grounding rate of 60% over the past 4 years, a 1200- oat array could contribute over 25,000 additional deep-ocean bathymetry measurements every year to global bathymetry datasets such as the Nippon Foundation ¿GEBCO Seabed 2030 Project (Mayer et al., 2018).

5.3 Ocean mixing

Ocean mixing controls the vertical transport of water masses within the meridional overturning circulation and therefore strongly affects the oceanic uptake of heat, oxygen, and anthropogenic carbon (Cimoli et al., 2023). Upcoming work includes testing the ability of pilot temperature and velocity shear microstructure sensors mounted on Deep Argo oats to improve deep-ocean mixing estimates. At present, deep ocean mixing is measured from either extremely sparse focused

process studies using tracer, moorings, and microstructure measurements (Waterhouse et al., 2014; Ferron et al., 2017; MacKinnon et al., 2017) or from ne-scale parameterizations using Argo and ship-based CTD pro les (Whalen et al., 2015; Ferron et al., 2016). However, these parametrizations rely on assumptions that may not be valid for all ocean dynamics (Vladoiu et al., 2021) leading to systematic biases in some areas of the ocean (Lele et al., 2021), particularly close to sea oor topography. Modern turbulent packages offer low-power, inexpensive options (e.g., Le Boyer, 2021) that have already successfully been integrated onto upper 2000 m ocean observing platforms (e.g., Nagai et al., 2015). The slow pro ling speeds of Deep Argo oats and limited turbulence noise induced by the oat-sensor apparatus, coupled with new technology and vetted onboard processing (e.g., Hughes et al., 2023) have reduced the amount of transmitted data required to compute mixing estimates reliably.

5.4 Optical scattering

Deep scattering measurements are needed to improve our understanding of the fate of sinking particles produced at the surface and their role in burying organic carbon in sediments, to quantify sediment resuspension and transport near the ocean bottom, and to measure transport and redistribution of trace metals (Lam and Bishop, 2008; Estapa et al., 2015). Another application of deep-ocean scattering data is to assess the environmental impacts of deep-sea mining by establishing a baseline for the concentration of particles present below 2000 m and inferring variations from this baseline due to mining operations. Emerging plans involve testing the performance of pilot optical scattering sensors (Gardner et al., 2018) on Deep Argo oats to measure the concentration of particles present in deep-sea water. Deeprated optical scattering sensors are commercially available and have been eld tested to 6000 m (e.g., Gardner et al., 2018; Ray et al., 2020), but have not yet been implemented on Deep Argo oats. The high sensitivity of optical scattering to minerals and large particles (Briggs et al., 2020) implies that no increase in sensor sensitivity nor accuracy is needed to quantitatively resolve the deep-ocean signal. Sensor calibration and data quality control will be studied to address and correct the effect of pressure cycling on the accuracy and stability of optical scattering measurements (Hu et al., 2019).

6 Conclusion

Deep Argo oats are autonomous pro lers with limited carbon footprint and environmental impact, and are rapidly improving deep-ocean sampling coverage. A strategy to implement the global 1200 Deep Argo oat array has been carefully designed to resolve regional-to-global variations of temperature and salinity over the full-ocean depth at subseasonal-to-decadal time scale. Ongoing challenges are to increase the diversity of Deep Argo oats and CTD sensor models, extend the averaged Deep Argo oat longevity to > 4 years, improve sensor performance and re ne real-time and delayed-mode Deep Argo data quality control procedures to achieve targeted values of temperature, salinity, and pressure accuracies across all CTD sensors. Concurrent observations from Core and Deep Argo, repeat hydrography, and

satellite networks will enable identi cation and correction of bias and drift in sea level anomalies measured from satellite altimeters, ocean mass addition from space gravimetry, and salinity measurements from Argo CTD sensors. The deployment of the global Deep Argo array, combined with Core Argo, repeat hydrography and satellite altimetry and gravimetry missions, has the capacity to improve assessments and projections of the global planetary heat, freshwater, and sea level budgets, and strength of the overturning circulation. While Deep Argoite priority is the full implementation of the backbone 1200- oat array extending Argo temperature and salinity pro ling to the abyssal ocean, emerging projects are under evaluation to address ocean deoxygenation and carbon sequestration, ocean mixing, sediment transport, and improve bathymetry estimates.

Author contributions

NZ: Conceptualization, Investigation, Writing 1/2 original draft, Writing Vzreview and editing. VT: Investigation, Writing Vzoriginal draft, Writing V2review and editing. BK: Writing V2review and editing. MA: Writing V2review and editing. XA: Writing V2review and editing. KB: Writing Vzreview and editing. NB: Writing Vzreview and editing. ZC: Writing Vareview and editing. CC: Writing Vareview and editing. LC: Writing V2review and editing. GDV3: Writing V2 review and editing. DD: Writing Vzreview and editing. DF: Writing Writing √2review and editing. FG: Writing √2review and editing. BH: Writing Vareview and editing. SH: Writing Vareview and editing. GJ: Writing Vareview and editing. TK: Writing Vareview and editing. AL: Writing 1/2 review and editing. WL: Writing 1/2 review and editing. PO: Writing V2review and editing. SP: Writing V2review and editing. ER: Writing Vareview and editing. DR: Writing Vareview and editing. MS: Writing Vzreview and editing. PS: Writing Vzreview and editing. KW: Writing V2review and editing. LW: Writing V2 review and editing. EW: Writing V2review and editing.

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Con ict of interest

The authors declare that the research was conducted in the absence of any commercial or nancial relationships that could be construed as a potential con ict of interest.

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References

Abdalla, S., Kolahchi, A. A., Ablain, M., Adusumilli, S., Bhowmick, S. A., Alou-Font, E., et al. (2021). Altimetry for the future: Building on 25 years of progress. Adv. Space Res. 68 (2), 319§363. doi: 10.1016/j.asr.2021.01.022

Allen, M., Antwi-Agyei, P., Aragon-Durand, F., Babiker, M., Bertoldi, P., Bind, M., et al. (2019). Technical Summary: Global warming of 1.5 & An IPCC Special Report on the impacts of global warming of 1.5 & above pre-industrial levels and related global greenhouse gas emission pathways, in the context of strengthening the global response

to the threat of climate change, sustainable development, and efforts to eradicate poverty. Intergovernmental Panel Climate Change.

Barnoud, A., Pfeffer, J., Guerou, A., Frery, M.-L., Simeon, M., Cazenave, A., et al. (2021). Contributions of altimetry and Argo to non-closure of the global mean sea level budget since 2016. Geophysical Res. Lett. 48, e2021GL092824. doi: 10.1029/2021GL092824

- Bittig, H. C., Ktzinger, A., Neill, C., van Ooijen, E., Plant, J. N., Hahn, J., et al. (2018). Oxygen optode sensors: principle, characterization, calibration, and application in the ocean. Front. Mar. Sci. 4. doi: 10.3389/fmars.2017.00429
- Bittig, H., Maurer, T., Plant, J., Schmechtig, C., Wong, A., Claustre, H., et al. (2019). A BGC-argo guide: planning, deployment, data handling and usage. Front. Mar. Sci. 6. doi: 10.3389/fmars.2019.00502
- Briggs, N., Dallilomo, G., and Claustre, H. (2020). Major role of particle fragmentation in regulating biological sequestration of CO_2 by the oceans. Science 367 (6479), 7911/793. doi: 10.1126/science.aay1790
- Busecke, J. J. M., Resplandy, L., Ditkovsky, S. J., and John, J. G. (2022). Diverging fates of the Paci c Ocean oxygen minimum zone and its core in a warming world. AGU Adv. 3, e2021AV000470.
- Cabanes, C., Thierry, V., and Lagadec, C. (2016). Improvement of bias detection in Argo oat conductivity sensors and its application in the North Atlantic. Deep-Sea Res. Part 1: Oceanographic Res. Papers 114, 128₺236. doi: 10.1016/j.dsr.2016.05.007
- Cazenave, A., Hamlington, B., Horwath, M., Barletta, V. R., Benveniste, J., Chambers, D., et al. (2019). Observational requirements for long-term monitoring of the global mean sea level and its components over the altimetry era. Front. Mar. Sci. 6 (582). doi: 10.3389/fmars.2019.00582
- Chamberlain, M. A., Oke, P. R., Fiedler, R. A. S., Beggs, H. M., Brassington, G. B., and Divakaran, P. (2021). Next generation of Bluelink ocean reanalysis with multiscale data assimilation: BRAN2020. Earth Syst. Sci. Data Discuss 13, 56631/5688. doi: 10.5194/essd-2021-194
- Chang, L., Tang, H., Wang, Q., and Sun, W. (2019). Global thermosteric sea level change contributed by the deep ocean below 2000 m estimated by Argo and CTD data. Earth Planetary Sci. Letters. 524, 115727. doi: 10.1016/j.epsl.2019.115727
- Chen, J., Cazenave, A., Dahle, C., Llovel, W., Panet, I., Pfeffer, J., et al. (2022). Applications and challenges of GRACE and GRACE follow-on satellite gravimetry. Surveys Geophysics 43, 305½345. doi: 10.1007/s10712-021-09685-x
- Chen, J., Tapley, B., Wilson, C., Cazenave, A., Seo, K.-W., and Kim, J.-S. (2020). Global ocean mass change from GRACE and GRACE Follow-On and altimeter and Argo measurements. Geophys. Res. Lett. 47, e2020GL090656. doi: 10.1029/2020GL090656
- Cimoli, L., Mashayek, A., Johnson, H. L., Marshall, D. P., Naveira Garabato, A. C., Whalen, C. B., et al. (2023). Signi cance of diapycnal mixing within the Atlantic meridional overturning circulation. AGU Adv. 4, e2022AV000800. doi: 10.1029/2022AV000800
- Coppola, L., Prieur, L., Taupier-Letage, I., Estournel, C., Testor, P., Lefevre, D., et al. (2017). Observation of oxygen ventilation into deep waters through targeted deployment of multiple $Argo-O_2$ oats in the north-western Mediterranean Sea in 2013. J. Geophys. Res.: Oceans 122 (8), 6325&6341. doi: 10.1002/2016jc012594
- de Boer, A. M., Hutchinson, D. K., Roquet, F., Sime, L. C., Burls, N. J., and Heuze, C. (2022). The impact of southern ocean topographic barriers on the ocean circulation and the overlying atmosphere. J. Climate 35, $5805\sqrt[6]{5821}$. doi: 10.1175/JCLI-D-21-0896.1
- Desbruytés, D. G., Bravo, E. P., Thierry, V., Mercier, H., Lherminier, P., Cabanes, C., et al. (2022). Warming-to-cooling reversal of over ow-derived water masses in the Irminger Sea during 2002/2021. Geophysical Res. Lett. 49, e2022GL098057. doi: 10.1029/2022GL098057
- Desbruyiés, D. G., Purkey, S. G., McDonagh, E. L., and Johnson, G. C. (2016). and King, B. A. Deep and abyssal ocean warming from 35 years of repeat hydrography. Geophys. Res. Lett. 43 (10), ,356½310. doi: 10.1002/2016GL070413
- Estapa, M. L., Breier, J. A., and German, C. R. (2015). Particle dynamics in the rising plume at Piccard Hydrothermal Field, Mid-Cayman Rise. Geochem. Geophys. Geosyst. 16, 2762 2774. doi: 10.1002/2015GC005831
- Ferron, B., Bouruet Aubertot, P., Cuypers, Y., Schroeder, K., and Borghini, M. (2017). How important are diapycnal mixing and geothermal heating for the deep circulation of the Western Mediterranean? Geophys. Res. Lett. 44, 7845\[2012]78154. doi: 10.1002/2017GL074169
- Ferron, B., Kokoszka, F., Mercier, H., and Lherminier, P. (2016). Variability of the diapycnal mixing along the A25 Greenland-Portugal transect repeated from 2002 to 2012. J. Phys. Oceanography 46, 1989∮2003. doi: 10.1175/JPO-D-15-0186.1
- Foppert, A., Rintoul, S. R., Purkey, S. G., Zilberman, N., Kobayashi, T., Sallada, B., et al. (2021). Deep Argo reveals bottom water properties and pathways in the Australian-Antarctic Basin. J. Geophys. Res.: Oceans 126, 11/218. doi: 10.1029/2021JC017935
- Fox-Kemper, B., Hewitt, H. T., Xiao, C., Adgeirsdo ttir, G., Drijfhout, S. S., Edwards, T. L., et al. (2021). Leryosphere and sea level change Lan Climate change 2021: The Physical Science Basis. Contribution of Working Group I to the Sixth Assessment Report of the Intergovernmental Panel on Climate Change. Eds. V. Masson-Delmotte, P. Zhan, A. Pirani, S. L. Connors, C. Pean, S. Berger, N. Caud, Y. Chen, L. Goldfarb, M. I. Gomis, M. Huang, K. Leitzell, E. Lonnoy, J. B. R. Matthews, T. K. Maycock, T. Water eld, O. Yelekci, R. Yu and B. Zhou (Cambridge, United Kingdom and New York, NY, USA: Cambridge University Press), 12111/1362. doi: 10.1017/9781009157896.011
- Gaillard, F., Reynaud, T., Thierry, V., Kolodziejczyk, N., and von Schuckmann, K. (2016). In-situ based reanalysis of the global ocean temperature and salinity with ISAS: variability of the heat content and steric height. J. Clim. 29 (4), 1305½323. doi: 10.1175/JCLI-D-15-0028.1
- Garcia, H. E., Boyer, T. P., Baranova, O. K., Locarnini, R. A., Mishonov, A. V., Grodsky, A., et al. (2019). World ocean atlas 2018: product documentation. A. Mishonov Tech. Editor.
- Garcon, V., Karstensen, J., Palacz, A., Telszewski, M., Aparco Lara, T., Breitburg, D., et al. (2019). Multidisciplinary observing in the world ocean⊯oxygen minimum zone

- Gardner, W. D., Richardson, M. J., Mishonov, A. V., and Biscaye, P. E. (2018). Global comparison of benthic nepheloid layers based on 52 years of nephelometer and transmissometer measurements. Prog. Oceanogr 168, 1001/11. doi: 10.1016/j.pocean.2018.09.008
- Gasparin, F., Guinehut, S., Mao, C., Mirouze, I., Remy, E., King, R. R., et al. (2019). Requirements for an integrated in situ atlantic ocean observing system from coordinated observing system simulation experiments. Front. Mar. Sci. 6. doi: 10.3389/fmars.2019.00083
- Gasparin, F., Hamon, M., Remy, E., and Le Traon, P. (2020). How deep argo will improve the deep ocean in an ocean reanalysis. J. $Clim~33~(1), 77\slashed{k}94$. doi: 10.1175/JCLI-D-19-0208.1
- GEBCO Bathymetric Compilation Group (2022). The GEBCO_2022 Grid a continuous terrain model of the global oceans and land. NERC EDS Br. Oceanographic Data Centre NOC. doi: 10.5285/e0f0bb80-ab44-2739-e053-6c86abc0289c
- Gordon, C., Fennel, K., Richards, C., Shay, L. K., and Brewster, J. K. (2020). Can ocean community production and respiration be determined by measuring high-frequency oxygen pro les from autonomous oats? Biogeosciences 17, 4119∤⁄4134. doi: 10.5194/bg-17-4119-2020
- Gregoire, M., Garcon, V., Garcia, H., Breitburg, D., Isensee, K., Oschlies, A., et al. (2021). A global ocean oxygen database and atlas for assessing and predicting deoxygenation and ocean health in the open and coastal ocean. Front. Mar. Sci. 8. doi: 10.3389/fmars.2021.724913
- Hu, L., Zhang, X., and Perry, M. J. (2019). Light scattering by pure seawater: Effect of pressure. Deep Sea Res. Part I: Oceanographic Res. Papers 146, 103j/409. doi: 10.1016/j.dsr.2019.03.009
- Hughes, K., Moum, J. N., and Rudnick, D. (2023). A turbulence data reduction scheme for autonomous and expendable pro ling oats. Ocean Sci. 19, 193½207. doi: 10.5194/os-19-193-2023
- Johnson, G. C. (2008). Quantifying antarctic bottom water and north atlantic deep water volumes. J. Geophys. Res. Oceans 113 (C5), 13. doi: C0502710.1029/2007jc004477
- Johnson, G. C. (2022). Antarctic bottom water warming and circulation slowdown in the argentine basin from analyses of deep argo and historical shipboard temperature data. Geophys. Res. Lett. 49, e2022GL100526. doi: 10.1029/2022GL100526
- Johnson, G. C., and Lyman, J. (2020). Warming trends increasingly dominate global ocean. Nat. Clim. Change 10, 757 $\c/2$ 761. doi: 10.1038/s41558-020-0822-0
- Johnson, G. C., Cadot, C., Lyman, J., McTaggart, K. E., and Steffen, E. L. (2020). Antarctic bottom water warming in the Brazil basin: 1990s through 2020, from WOCE to deep argo. Geophys. Res. Lett. doi: 10.1029/2020GL089191
- Johnson, G. C., Lyman, J. M., and Purkey, S. G. (2015). Informing deep argo array design using argo and full-depth hydrographic section data. J. Atmos. Oceanic Technol. 32, 2187 ₹2198. doi: 10.1175/JTECH-D-15-0139.1
- Johnson, G. C., Purkey, S. G., Zilberman, N. V., and Roemmich, D. (2019). Deep argo quanti es bottom water warming rates in the southwest paci c basin. Geophys. Res. Lett. 46, 2662∤2669. doi: 10.1029/2018GL081685
- Katsumata, K., Purkey, S. G., Cowley, R., Sloyan, B. M., Diggs, S. C., Moore, T. S., et al. (2022). GO-SHIP Easy Ocean: Gridded ship-based hydrographic section of temperature, salinity, and dissolved oxygen. Sci. Data 9, 103. doi: 10.1038/s41597-022-01212-w
- Kobayashi, T., Sato, K., and King, B. A. (2021). Observed features of salinity bias with negative pressure dependency for measurements by SBE 41CP and SBE 61 CTD sensors on deep pro ling oats. Progr. Oceanogr. 198, 102686. doi: 10.1016/j.pocean.2021.102686
- Kobayashi, T., Watanabe, K., and Tachikawa, M. (2013). Deep NINJA collects pro les down to 4,000 meters. Sea Technology. 54, 41½44.
- Kolodziejczyk, N., Prigent-Mazella, A., and Gaillard, F. (2021). SEANOE. doi: 10.17882/52367
- Kwiatkowski, L., Torres, O., Bopp, L., Aumont, O., Chamberlain, M., Christian, J., et al. (2020). Twenty- rst century ocean warming, acidi cation, deoxygenation, and upper ocean nutrient decline from CMIP6 model projections. Biogeosciences 17 (13), 3439\?470. doi: 10.5194/bg-17-3439-2020
- Lam, P. J., and Bishop, J. K. B. (2008). The continental margin is a key source of iron to the HNLCNorth Paci c Ocean. Geophys. Res. Lett. 35. doi: 10.1029/2008GL033294
- Le Boyer, A. (2021). Modular, exible, low-cost microstructure measurements: the Epsilometer. J. Atmos. Oceanic Technol. 38 (3), 657/668. doi: 10.1175/JTECH-D-20-0116.1
- Lele, R., Purkey, S. G., Nash, J. D., MacKinnon, J. A., Thurnherr, A. M., Whalen, C. B., et al. (2021). Abyssal heat budget in the southwest paci c basin. J. Phys. Oceanogr. 51 (11), 3317∦3333. doi: 10.1175/JPO-D-21-0045.1
- Lellouche, J.-M., Greiner, E., Bourdalle-Badie, R., Garric, G., Melet, A., Drevillon, M., et al. (2021). The copernicus global 1/12∰0ceanic and sea ice GLORYS12 reanalysis. Front. Earth Sci. 9. doi: 10.3389/feart.2021.698876
- Le Reste, S., Dutreuil, V., Andre, X., Thierry, V., Renaut, C., Le Traon, P.-Y., et al. (2016). Deep-Arvor: A new pro ling oat to extend the Argo observations down to 4000-m depth. J. Atmos. Oceanic Technol. 33, 1039∮4055. doi: 10.1175/JTECH-D-15-0214.1
- Levin, L. A. (2018). Manifestation, drivers, and emergence of open ocean deoxygenation, annu. Rev. Mar. Sci. 10 (1), 2291/260. doi: 10.1146/annurev-marine-121916-063359
- Levin, L. A., Bett, B. J., Gates, A. R., Heimbach, P., Howe, B. M., Janssen, F., et al. (2019). Global observing needs in the deep ocean. Front. Mar. Sci. 6. doi: 10.3389/fmars.2019.00241
- Li, Q., England, M. H., Hogg, A. M., Rintoul, S. R., and Morrison, A. K. (2023). Abyssal ocean overturning slowdown and warming driven by Antarctic meltwater. Nature 615 (7954), 841½847. doi: 10.1038/s41586-023-05762-w

- Llovel, W., Purkey, S., Meyssignac, B., Blazquez, A., Kolodziejczyk, N., and Bamber, J. (2019). Global ocean freshening, ocean mass increase and global mean sea level rise over 2005 № 2015. Sci. Rep. 9, 17717. doi: 10.1038/s41598-019-54239-2
- Loeb, N. G., Johnson, G. C., Thorsen, T. J., Lyman, J. M., Rose, F. G., and Kato, S. (2021). Satellite and ocean data reveal marked increase in earthischeating rate. Geophys. Res. Lett. 48e2021GL093047. doi: 10.1029/2021GL093047
- MacKinnon, J. A., Zhao, Z., Whalen, C. B., Waterhouse, A. F., Trossman, D. S., Sun, O. M., et al. (2017). Climate process team on internal wave-driven ocean mixing. Bull. Am. Meteorol. Soc. 98, 2429∤2454. doi: 10.1175/BAMS-D-16-0030.1
- Martyr-Koller, R., Thomas, A., Schleussner, C. F., Nauels, A., and Lissner, T. (2021). Loss and damage implications of sea-level rise on Small Island Developing States. Curr. Opin. Environ. Sustain 50, 245∦259. doi: 10.1016/j.cosust.2021.05.001
- Mashayek, A., Ferrari, R., Merri eld, S., Ledwell, J. R., St Laurent, L., Naveira Garabato, A., et al. (2017). Topographic enhancement of vertical turbulent mixing in the Southern Ocean. Nat. Commun. 8, 14197. doi: 10.1038/ncomms14197
- Mayer, L. A., Jakobsson, M., Allen, G., Dorschel, B., Falconer, R., Ferrini, V., et al. (2018). The nippon foundation ½2GEBCO seabed 2030 project: the quest to see the world ≤ oceans completely mapped by 2030. Geosciences 8, 63. doi: 10.3390/geosciences8020063
- Meyssignac, B., Boyer, T., Zhao, Z., Hakuba, M. Z., Landerer, F. W., Stammer, D., et al. (2019). Measuring global ocean heat content to estimate the Earth energy imbalance. Front. Mar. Sci. 6, 432. doi: 10.1002/2015JC011288
- Nagai, T., Inoue, R., Tandon, A., and Yamazaki, H. (2015). Evidence of enhanced double-diffusive convection below the main stream of the Kuroshio Extension. J. Geophys. Res. Oceans 120, 84021/8421. doi: 10.1002/2015JC011288
- Nerem, R. S., Beckley, B. D., Fasullo, J. T., Hamlington, B. D., Masters, D., and Mitchum, G. T. (2018). Climate-change variven accelerated sea-level rise detected in the altimeter era. Proc. Natl. Acad. Sci. 115 (9), 2022 22205. doi: 10.1073/pnas.1717312115
- Ollitrault, M., and Rannou, J.-P. (2013). ANDRO: an argo-based deep displacement dataset. J. Atmos. Ocean. Technol. 30 (4), 759\frac{1}{2}788. doi: 10.1175/JTECH-D-12-00073.1
- Oschlies, A., Brandt, P., and Stramma, L. (2018). and Schmidtko, S. Drivers and mechanisms of ocean deoxygenation. Nat. Geosci. 11 (7), 4671/473. doi: 10.1038/s41561-018-0152-2
- Owens, W. B., and Wong, A. P. S. (2009). An improved calibration method for the drift of the conductivity sensor on autonomous CTD pro ling oats by -S climatology. Deep-Sea Res. Part I 56, 450½457. doi: 10.1016/j.dsr.2008.09.008
- Owens, W. B., Zilberman, N., Johnson, K. S., Claustre, H., Scanderbeg, M., Wijffels, S., et al. (2022). OneArgo: A new paradigm for observing the global ocean. Mar. Technol. Soc J. 56 (3), 84½90. doi: 10.4031/MTSJ.56.3.8
- Palmer, M. D. (2017). Reconciling estimates of ocean heating and earthi≰radiation budget. Curr. Clim. Change Rep. 3, 78₺86. doi: 10.1007/s40641-016-0053-7
- Palmer, M. D., McNeall, D. J., and Dunstone, N. J. (2011). Importance of the deep ocean for estimating decadal changes in Earth Aradiation balance. Geophys. Res. Lett. 38, L13707. doi: 10.1029/2011GL047835
- Passarelli, L., Cesca, S., Nooshiri, N., and Jonsson, S. (2022). Earthquake ngerprint of an incipient subduction of a bathymetric high. Geophys. Res. Lett. 49, e2022GL100326. doi: 10.1029/2022GL100326
- Petit, T., Thierry, V., and Mercier, H. (2022). Deep through- ow in the bight fracture zone. Ocean Sci. 18, 1055½4071. doi: 10.5194/os-18-1055-2022
- Petzrick, E., Truman, J., and Fargher, H. (2014). Pro ling from 6000 meter with the APEX-Deep oat. 2013 OCEANS San Diego, CA, USA, (2013) pp. 1₺3. doi: 10.23919/OCEANS.2013.6741074
- Piron, A., Thierry, V., Mercier, H., and Caniaux, G. (2016). Observation of basin-scale deep convection in the Irminger Sea with Argo oats in the 2011-2012 winter. Deep-Sea Res. Part | 109, 76∦90. doi: 10.1016/j.dsr.2015.12.012
- Purkey, S. G., and Johnson, G. C. (2010). Warming of Global Abyssal and Deep Southern Ocean waters between the 1990s and 2000s: Contributions to global heat and sea level rise budgets. J. Clim 23 (23), 6336⊮6351. doi: 10.1175/2010JCLI3682.1
- Purkey, S. G., Johnson, G. C., Talley, L. D., Sloyan, B. M., Wijffels, S. E., Smethie, W., et al. (2019). Unabated bottom water warming and freshening in the South Paci c Ocean. J. Geophys. Res. Oceans 124 (3), 1778∤4794. doi: 10.1029/2018jc014775
- Racape, V., Thierry, V., Mercier, H., and Cabanes, C. (2019). ISOW spreading and mixing as revealed by Deep-Argo oats launched in the Charlie-Gibbs Fracture Zone. J. Geophys. Res.: Oceans 124, 6787≱6808. doi: 10.1029/2019JC015040
- Ray, D., KameshRaju, K. A., Srinivas Rao, A., et al. (2020). Elevated turbidity and dissolved manganese in deep water column near 10½% & Central Indian Ridge: studies on hydrothermal activities. Geo-Mar. Lett. 40, 619½628. doi: 10.1007/s00367-020-00657-5
- Riser, S., and Johnson, K. (2008). Net production of oxygen in the subtropical ocean. Nature 451, 323 $\mbox{$\not 25}$ 225. doi: 10.1038/nature06441
- Roemmich, D., Alford, M. H., Claustre, H., Johnson, K., King, B., Moum, J., et al. (2019a). On the future of argo: A global, full-depth, multi-disciplinary array. Front. Mar. Sci. 6. doi: 10.3389/fmars.2019.00439
- Roemmich, D., Sherman, J. T., Davis, R. E., Grindley, K., McClune, M., Parker, C. J., et al. (2019b). Deep SOLO: A full-depth pro-ling oat for the Argo Program. J. Atmos. Oceanic Technol. 36 (10), 1967 1981. doi: 10.1175/jtech-d-19-0066.1

- Royston, S., Vishwakarma, B. D., Westaway, R. M., Rougier, J., Sha, Z., and Bamber, J. L. (2020). Can we resolve the basin-scale sea level trend budget from GRACE ocean mass? J. Geophys. Res.: Oceans 125, e2019JC015535. doi: 10.1029/2019JC015535
- Salaree, A., and Okal, E. A. (2020). Effects of bathymetry complexity on tsunami propagation: a spherical harmonics approach. Geophys. J. Int. 223 (1), 632≵647. doi: 10.1093/gji/ggaa334
- Sauzie, R., Bittig, H. C., Claustre, H., Pasqueron de Fommervault, O., Gattuso, J.-P., Legendre, L., et al. (2017). Estimates of water-column nutrient concentrations and carbonate system parameters in the global ocean: A novel approach based on neural networks. Front. Mar. Sci. 4. doi: 10.3389/fmars.2017.00128
- Schmidtko, S., Stramma, L., and Visbeck, M. (2017). Decline in global oceanic oxygen content during the past ve decades. Nature 542, 335½339. doi: 10.1038/nature21399
- Seferian, R., Berthet, S., Yool, A., Palmieri, J., Bopp, L., Tagliabue, A., et al. (2020). Tracking improvement in simulated marine biogeochemistry between CMIP5 and CMIP6. Curr. Clim. Change Rep. 6, 95∤19. doi: 10.1007/s40641-020-00160-0
- Sevellec, F., and Fedorov, A. V. (2013). Model bias reduction and the limits of oceanic decadal predictability: Importance of the deep ocean. J. Climate 26, 3688∤3707. doi: 10.1175/JCLI-D-12-00199.1
- Stammer, D., Bracco, A., AchutaRao, K., Beal, L., Bindoff, N. L., Braconnot, P., et al. (2019). Ocean climate observing requirements in support of climate research and climate information. Front. Mar. Sci. 6. doi: 10.3389/fmars.2019.00444
- Swingedouw, D., Houssais, M.-N., Herbaut, C., Blaizot, A.-C., Devilliers, M., and Deshayes, J. (2022). AMOC recent and future trends: A crucial role for oceanic resolution and Greenland melting? Front. Clim. 4. doi: 10.3389/fclim.2022.838310
- Testor, P., Bosse, A., Houpert, L., Margirier, F., Mortier, L., Legoff, H., et al. (2018). Multiscale observations of deep convection in the northwestern mediterranean sea during winter 2012 ₹2013 using multiple platforms. J. Geophys. Res.: Oceans 123, 1745 ₹2 1776. doi: 10.1002/2016JC012671
- Thomas, G., Purkey, S. G., Roemmich, D., Foppert, A., and Rintoul, S. R. (2020). Spatial variability of antarctic bottom water in the Australian antarctic basin from 2018-2020 captured by deep argo. Geophys. Res. Lett. 47. doi: 10.1029/2020gl089467
- Tozer, B., Sandwell, D. T., Smith, W. H. F., Olson, C., Beale, J. R., and Wessel, P. (2019). Global bathymetry and topography at 15 arc sec: SRTM15+. Earth Space Sci. 6, 1847/1864. doi: 10.1029/2019EA000658
- Ulses, C., Estournel, C., Fourrier, M., Coppola, L., Kessouri, F., Leftze, D., et al. (2021). Oxygen budget of the north-western Mediterranean deep-convection region. Biogeosciences 18, 9371/960. doi: 10.5194/bg-18-937-2021
- Uotila, P., Goosse, H., Haines, K., Chevallier, M., Barthelemy, A., Bricaud, C., et al. (2019). An assessment of ten ocean reanalyses in the polar regions. Clim. Dyn. 52, 1613½4650. doi: 10.1007/s00382-018-4242-z
- van Wijk, E. M., Hally, B., Wallace, L. O., Zilberman, N., and Scanderbeg, M. (2022). Can Argo oats help improve bathymetry? Int. Hydrographic Rev. 28, 2261/230. doi: 10.58440/ihr-28-n08
- van Wijk, E. M., and Rintoul., S. R. (2014). Freshening drives contraction of Antarctic Bottom Water in the Australian Antarctic Basin. Geophys. Res. Lett. 41 (5), 1657∤4664. doi: 10.1002/2013gl058921
- Vladoiu, A., Bouruet-Aubertot, P., Cuypers, Y., Ferron, B., Schroeder, K., Borghini, M., et al. (2021). Contrasted mixing ef ciency in energetic versus quiescent regions: insights from microstructure measurements in the Western Mediterranean Sea. Prog. Oceanogr. 195, 102594. doi: 10.1016/j.pocean.2021.102594
- von Schuckmann, K., Cheng, L., Palmer, M. D., Hansen, J., Tassone, C., Aich, V., et al. (2020). Heat stored in the Earth system: where does the energy go? Earth Syst. Sci. Data 12, 2013½2041. doi: 10.5194/essd-12-2013-2020
- von Schuckmann, K., Min \cancel{k} 6, A., Gues, F., Cuesta-Valero, F. J., Kirchengast, G., Adusumilli, S., et al. (2023). Heat stored in the Earth system 1960 \cancel{k} 2020: Where does the energy go? Earth Syst. Sci. Data Discuss 15, 1675 \cancel{k} 4709. doi: 10.5194/essd-15-1675-2023
- Walicka, K., King, B., Thierry, V., Cabanes, C., Velez-Belch@P., Coatanoan, C., et al. (2022). D3.4: Report of the outcome of the comparative study for the deep Argo quality control processing. Zenodo. doi: 10.5281/zenodo.8366455
- Waterhouse, A. F., MacKinnon, J. A., Nash, J. D., Alford, M. H., Kunze, E., Simmons, H. L., et al. (2014). Global patterns of diapycnal mixing from measurements of the turbulent dissipation rate. J. Phys. Oceanogr 44 (7), 1854\(\frac{1}{2}\)4872. doi: 10.1175/JPO-D-13-0104.1
- WCRP Global Sea Level Budget Group (2018). Global sea-level budget 1993∦present. Earth Syst. Sci. Data 10, 1551∦1590. doi: 10.5194/essd-10-1551-2018
- Whalen, C. B., MacKinnon, J. A., and Talley, L. D. (2015). and Waterhouse, A. F. Estimating the mean diapycnal mixing using a nescale strain parameterization. J. Phys. Oceanogr. 45, 1174½1188. doi: 10.1175/JPO-D-14-0167.1
- Whalen, C. B., Talley, L. D., and MacKinnon, J. A. (2012). Spatial and temporal variability of global ocean mixing inferred from Argo pro les. Geophys. Res. Lett. 39. doi: 10.1029/2012gl053196
- Wilson, E. A., Thompson, A. F., Stewart, A. L., and Sun, S. (2022). Bathymetric control of subpolar gyres and the overturning circulation in the southern ocean. J. Phys. Oceanogr 52 (2), 205∤223. doi: 10.1175/JPO-D-21-0136.1
- W ½ , A.-C., Snaith, H., Amirebrahimi, S., Devey, C. W., Dorschel, B., Ferrini, V., et al. (2019). Sea oor mapping ½ the challenge of a truly global ocean bathymetry. Front. Mar. Sci. 6. doi: 10.3389/fmars.2019.00283

Wong, A. P. S., Wijffels, S. E., Riser, S. C., Pouliquen, S., Hosoda, S., Roemmich, D., et al. (2020). Argo data 1999-2019: two million temperature-salinity pro les and subsurface velocity observations from a global array of pro ling oats. Front. Mar. Sci. 7. doi: 10.3389/fmars.2020.00700

Yashayaev, I. (2007). Hydrographic changes in the labrador se-2005. Prog. Oceanogr 73 (3-4), $242 \not\ge 276$. doi: 10.1016/j.pocean.2007.04.015

Zhang, H. L. J., Whalen, C. B., Kumar, N., and Purkey, S. G. (2021). Decreased strati cation in the abyssal southwest paci c basin and implications for the energy budget. Geophys. Res. Lett. 48 (19). doi: 10.1029/2021gl094322

Zhou, S., Meijers, A. J. S., Meredith, M. P., Povl Abrahamsen, E., Holland, P. R., Silvano, A., et al. (2023). Slowdown of Antarctic Bottom Water export driven by

climatic wind and sea-ice changes. Nat. Clim. Chang. 13, $701\slash\hspace{-0.065cm}/709$. doi: 10.1038/s41558-023-01695-4

Zilberman, N., King, B., Purkey, S., Thierry, V., and Roemmich, D. (2019). Report on the 2nd Deep Argo Implementation Workshop (Hobart). Available at: https://archimer.ifremer.fr/doc/00507/61873/.

Zilberman, N. V., Roemmich, D. H., and Gilson., J. (2020). Deep-ocean circulation in the southwest paciac ocean interior: estimates of the mean ow and variability using deep argo data. Geophys. Res. Lett. 47, (13). doi: 10.1029/2020gl088342

Zilberman, N. V., Scanderbeg, M., Gray, A. R., and Oke, P. R. (2023). Scripps Argo trajectory-based velocity product: global estimates of absolute velocity derived from core, biogeochemical, and deep argo oat trajectories at parking depth. J. Atmos. Oceanic Technol. 40 (3), 361½374. doi: 10.1175/JTECH-D-22-0065.1