# Can oxygen utilization rate be used to track the long-term changes of aerobic respiration in the mesopelagic Atlantic Ocean?

Haichao Guo<sup>1</sup>, Iris Kriest<sup>2</sup>, Andreas Oschlies<sup>3</sup>, and Wolfgang Koeve<sup>3</sup>

<sup>1</sup>GEOMAR Helmholtz Centre for Ocean Research Kiel <sup>2</sup>Helmholtz-Zentrum für Ozeanforschung Kiel (GEOMAR) <sup>3</sup>Helmholtz-Zentrum für Ozeanforschung Kiel, GEOMAR

June 20, 2023

## Can oxygen utilization rate be used to track the long-term changes of aerobic respiration in the mesopelagic Atlantic Ocean?

Haichao Guo<sup>1</sup>, Iris Kriest<sup>1</sup>, Andreas Oschlies<sup>1</sup>, and Wolfgang Koeve<sup>1</sup>

 $^1\mathrm{GEOMAR}$ Helmholtz Centre for Ocean Research Kiel, 24105, Kiel, Germany

## Key Points:

1

2

3

4

5

6

7	• 011	r model study confirms earlier findings that oxygen utilization rate (OUR) un-
-	dor	in model study communication $(R_{})$ in moscial again a second
8	der	estimates true respiration ( <i>n</i> <sub>true</sub> ) in mesoperagic ocean
9	• Des	pite OUR underestimate $R_{\rm true}$ , OUR can adequately estimate long-term changes
10	in I	$R_{\rm true}$ in the mesopelagic North Atlantic subtropical gyre
11	• OU	R cannot adequately estimate climate-driven changes in $R_{\rm true}$ in the mesopelagic
12	tro	pical South Atlantic where different water masses mix.

Corresponding author: Haichao Guo, hguo@geomar.de

#### 13 Abstract

Quantifying changes in oceanic aerobic respiration is essential for understanding 14 marine deoxygenation. Here we use an Earth system model to investigate if and to what 15 extent oxygen utilization rate (OUR) can be used to track the temporal change of true 16 respiration  $(R_{\rm true})$ .  $R_{\rm true}$  results from the degradation of particulate and dissolved or-17 ganic matter in the model ocean, acting as ground truth to evaluate the accuracy of OUR. 18 Results show that in thermocline and intermediate waters of the North Atlantic Sub-19 tropical Gyre (200m-1000m), vertically integrated OUR and  $R_{\rm true}$  both decrease by 0.2 20  $molO_2/m^2/yr$  from 1850 to 2100 under global warming. However, in the mesopelagic Trop-21 ical South Atlantic, integrated OUR increases by  $0.2 \text{ molO}_2/\text{m}^2/\text{yr}$ , while the  $R_{\text{true}}$  in-22 tegral decreases by 0.3 molO<sub>2</sub>/m<sup>2</sup>/yr. A possible reason for the diverging OUR and  $R_{\rm true}$ 23 is ocean mixing, which affects water mass composition and maps remote respiration changes 24 to the study region. 25

## <sup>26</sup> Plain Language Summary

The ocean is losing oxygen due to an imbalance in oxygen supply and aerobic res-27 piration. Therefore, monitoring temporal changes in the aerobic respiration rate contributes 28 to understanding marine deoxygenation. Based on simulations of an Earth system model, 29 we investigate an indirect diagnostic measure of the respiration rate (oxygen utilization 30 rate, OUR), calculated as the slope of the least square regression of the apparent oxy-31 gen utilization (AOU, saturated oxygen concentration minus local oxygen concentration) 32 and seawater age that can be computed from transient abiotic tracers. As the reference 33 to OUR, true respiration  $(R_{\rm true})$  is the oxygen consumption rate resulting from the degra-34 dation of organic matter in the model ocean. Results show that in the North Atlantic 35 Subtropical Gyre intermediate water (200m-1000m), both vertically integrated OUR and 36  $R_{\rm true}$  decrease by 0.2 molO<sub>2</sub>/m<sup>2</sup>/yr from 1850 to 2100. However, in the Tropical South 37 Atlantic intermediate water, the OUR integral increases by  $0.2 \text{ molO}_2/\text{m}^2/\text{yr}$  and the 38  $R_{\rm true}$  integral decreases by 0.3 molO<sub>2</sub>/m<sup>2</sup>/yr. We hypothesize that changes in ocean mix-39 ing over time, which can affect water mass composition and map remote respiration changes 40 to the study region, explain the discrepancy of OUR and  $R_{\text{true}}$  tendencies. 41

## 42 1 Introduction

Observations suggest that the dissolved oxygen concentration in the ocean has been substantially declining over the past 50 yrs (Schmidtko et al., 2017; Ito et al., 2017). This deoxygenation process has far-reaching impacts on global ocean ecosystems and biogeochemical processes (Stramma et al., 2008; Breitburg et al., 2018; Pezner et al., 2023). Better understanding ocean deoxygenation requires quantitatively assessing the temporal changes in the oxygen supply and sink processes, i.e., (i) oxygen solubility at the surface, (ii) ventilation, and (iii) aerobic respiration (Oschlies et al., 2018; Robinson, 2019).

While the contribution of (i) can straightforwardly be estimated from observations and was found to explain about 15% of the observed decline in oceanic oxygen inventory from 1960 to 2010 (Schmidtko et al., 2017), contributions from (ii) and (iii) are more difficult to quantify. Here we address (iii), i.e. changes in aerobic respiration.

Direct measurements of respiration rates over the recent decades are still sparse because of methodological limitations (Del Giorgio & Williams, 2005; Robinson, 2019). For example, rates obtained via enzymatic ETS (electron transport system) represent potential rates and are thus only a proxy for microbial activity, which has to be converted to actual respiration rates. This conversion is not always straightforward as it depends on community composition of organisms (Filella et al., 2018), which might change quickly in space and time. Likewise, rates obtained through oxygen uptake in sea water incubations integrate over the entire auto- and heterotrophic community, and depend on the availability of organic substrate. They may thus also represent just a snapshot of biogeochemical processes at a certain location. On the other hand, respiration estimates calculated from the decay of particle flux with depth, integrate over larger space and longer time scales, but are affected by mixing and advection that may laterally decouple the signals recorded at different depths (Waniek et al., 2000). In this case, it can be difficult to disentangle the effects of the hydrodynamic components from the biological ones.

Here, we focus on the classic Oxygen Utilization Rate (OUR) method as an indi-68 rect measure of marine respiration. OUR is defined as the slope of the least square re-69 gression of apparent oxygen utilization (AOU, the difference between saturated oxygen 70 concentration,  $[O_2^{\text{sat}}]$ , and actual oxygen concentration,  $[O_2^{\text{obs}}]$ ) and seawater age (t) on 71 potential density surfaces (equation 1; Jenkins, 1987). Seawater age is defined as the time 72 elapsed since the water had been last in contact with the atmosphere. Notably, AOU might 73 overestimate the true oxygen utilization (TOU) due to an incomplete equilibration of 74 the sea surface oxygen (Ito et al., 2004; Duteil et al., 2013). Still, instead of using AOU 75 directly as proxy for respiration, OUR is determined by the regression between AOU and 76 the seawater age. The OUR derived from AOU would not differ much from that derived 77 from TOU if water parcels share the same biases from surface disequilibration and thus 78 do not affect isopycnal oxygen utilization gradients (Sonnerup et al., 2019). 79

In the real ocean, AOU is a readily available property of seawater, while the age 80 of a water mass is typically derived from transient abiotic tracers like sulphur hexaflu-81 oride (SF6), chlorofluorocarbons (CFCs, e.g. CFC-11, CFC-12), or radioactive elements 82 (e.g.  ${}^{3}H$ ,  ${}^{39}Ar$ ,  ${}^{14}C$ ) (Fine, 2011; Fine et al., 2017; Stöven et al., 2015). These age trac-83 ers have been intensively measured globally over the last three decades (Fine et al., 2017), 84 which allows, theoretically, the widespread use of OUR and reconstruction of respira-85 tion over decades in parts of the ocean. Note that current approaches to estimate sea-86 water age (e.g., tracer age, transient time distribution TTD) still include some uncer-87 tainties and biases, like the assumption of perfect saturation of transient tracers at the 88 time of water mass formation (Stöven et al., 2015). 89

$$OUR = \frac{\partial AOU}{\partial t} = \frac{\partial ([O_2^{\text{sat}}] - [O_2^{\text{obs}}])}{\partial t}$$
(1)

Despite several field studies (e.g., Jenkins, 1987; Brea et al., 2004; Sonnerup et al., 91 2013, 2015, 2019; Álvarez-Salgado et al., 2014) and model studies (e.g., Koeve & Kähler, 92 2016) focusing on the comparison between OUR and other independent respiration es-93 timations at specific time points, to our knowledge, no study has attempted to evalu-94 ate the potential ability of OUR on tracking temporal changes in aerobic respiration un-95 der a changing climate. One reason might be the above-mentioned difficulties in obtain-96 ing direct observations of in-situ respiration rates in the real ocean as OUR reference. 97 Here, we employ a high-resolution Earth system model (for which we have perfect spa-98 tial and temporal coverage of "sampling", and knowledge of the true respiration rate) 99 to address this issue. In particular, we examine the relationship and its temporal vari-100 ations between OUR (calculated from simulated AOU and ideal age) and true respira-101 tion from 1850 to 2100 for a global warming scenario in two selected study areas in the 102 Atlantic Ocean. 103

#### 104 2 Method

90

#### **2.1 Model description**

The model used in this study is the Flexible Ocean and Climate Infrastructure (FOCI) Earth system model (Matthes et al., 2020) coupled to ocean biogeochemistry as detailed by Chien et al. (2022). It includes an atmosphere, a land biosphere, an ocean circula-

tion, a sea-ice, and an ocean biogeochemistry component. The oceanic components ap-109 ply the ORCA05 grid, corresponding to a tripolar grid with  $0.5^{\circ} \times 0.5^{\circ}$  nominal hor-110 izontal resolution and 46 vertical levels with thicknesses varying from 6 m at the surface 111 to 250 m in the deep ocean. Tracer diffusion is aligned along isopycnals, with a diffu-112 sion coefficient of  $600 \text{ m}^2 \text{s}^{-1}$ . The biogeochemical component of FOCI, MOPS (Model 113 of Oceanic Pelagic Stoichiometry) includes nine compartments, of which five are calcu-114 lated in phosphorus units, namely phytoplankton, zooplankton, particulate detritus (DET), 115 dissolved organic matter (DOM), and phosphate. The abiotic tracers include oxygen, 116 nitrate, dissolved inorganic carbon, and alkalinity (Kriest & Oschlies, 2015; Chien et al., 117 2022). The effects of iron limitation on marine primary productivity are not explicitly 118 resolved in MOPS. With some modifications due to a slow-down of respiration in low-119 oxygen environments (see Kriest & Oschlies, 2015; Chien et al., 2022), the flux profile 120 of particulate organic matter in MOPS follows a "Martin Curve" (Martin et al., 1987), 121 where the exponent is derived from a constant decay rate  $(0.05 \text{ d}^{-1})$  and linearly increas-122 ing sinking speed (w = 0.0354 z m/d). 123

We here describe details on how FOCI simulates ideal age and true respiration rate. 124 The ideal age tracer works like a "clock," which increases one day per day since the wa-125 ter parcel has left the surface. The "clock" is set to zero when the water reaches the sur-126 face of the ocean (Thiele & Sarmiento, 1990). In FOCI, the ideal age is set to zero in 127 the upper 10 m. The remineralization rate in MOPS is temperature-independent and 128 depends only on substrate availability and oxygen concentration. Oxygen concentration 129 constrains remineralization rate only in the oxygen deficit zone, which is not the case for 130 the two sections analyzed here. Hence, we do not need to account for the oxygen depen-131 dence of remineralization in this paper, but refer readers to Chien et al. (2022, Appendix 132 A1) for details. The true respiration rate,  $R_{\rm true}$ , is the oxygen consumption rate for aer-133 obic remineralization of DET and DOM in each grid box, as described in Equation 2: 134

$$R_{\text{true}} = (\lambda'_{\text{DET}} \cdot DET + \lambda'_{\text{DOM}} \cdot DOM) \cdot R_{\text{O2:P}}$$
(2)

where  $\lambda'_{\text{DET}}$  (0.05d<sup>-1</sup>) and  $\lambda'_{\text{DOM}}$  (0.17yr<sup>-1</sup>) are the temperature-independent decay rates of DET and DOM, respectively.  $R_{\text{O2:P}} = 165.08044$  denotes the calibrated stoichiometric oxygen demand of aerobic remineralization (Chien et al., 2022).  $R_{\text{true}}$  is computed at every model time step and acts as the ground truth of its proxy, OUR, in the model ocean.

135

The experimental set-up is detailed by Chien et al. (2022), and we show the schematic 141 figure (Figure S1) and some technical details in the supplementary material Text S1. In 142 brief, the model was integrated for 750 years under a pre-industrial partial pressure of 143  $CO_2$  as a total spin-up. Branching off from this spinup state, transient climate-change 144 and pre-industrial control (esm-piControl) simulations were carried out for 250 years, 145 respectively. The transient simulations include 165 years (1850 to 2014) historical sim-146 ulation and 85 years (2015 to 2099) projection under the Shared Socioeconomic Path-147 ways 585 (SSP-585) scenario (esm-ssp-585; Eyring et al., 2016). The esm-piControl sim-148 ulation shares the same time period as transient simulations but without anthropogenic 149 effects (zero-emission of  $CO_2$ ). Concerning the model's projection sensitivity to initial 150 conditions and spin-up, three transient simulations and esm-piControl simulations were 151 employed. These simulations branch off from the 730th, 740th, and 750th year of the spin-152 up simulation, respectively (for brevity simply referred to as ensemble members 1, 2, 3). 153 Here, we only present details of ensemble member 1, but provide the statistical analy-154 sis of simulated OUR and true respiration for all ensemble members in Table 1 to sup-155 port the robustness of the finding. 156

#### 157 2.2 Model analysis

We confine our analysis to the mesopelagic zone, i.e. the depth range 200m to 1000m. 158 We consider OUR estimates unreliable in the upper 200m because of possible seasonal 159 effects such as subsurface warming by absorption of solar radiation (e.g., Dietze & Os-160 chlies, 2005). In addition, AOU is also substantially modified by photosynthesis in the 161 photic zone. In the deep ocean below 1000m, respiration proceeds at a very low rate (Del Gior-162 gio & Duarte, 2002; Williams, 1981). Although 76% of total ocean volume is below 1000m, 163 the respiration in the ocean interior only contributes 3% of globally integrated oceanic 164 165 respiration in the FOCI model.

We select two main study sections, referred to as the North Atlantic Subtropical 166 Gyre (NASG section, 60°W-30°W, 20°N-25°N) and the Tropical South Atlantic (TSA 167 section, 35°W-5°W, 15°S-20°S), respectively (Figure 1). Both sections approximately 168 fit the criteria for selecting sections used by Jenkins (1987): (a) sections should follow 169 the advective flow direction (Figure S2) and (b) should be approximately perpendicu-170 lar to isolines of ideal age and AOU. These conditions are considered necessary criteria 171 to minimize the effects of mixing of different source water types, which could otherwise 172 misguide attempts to estimate local respiration rates from OUR. Also, the NASG region 173 was selected since it is one of the regions projected to suffer the most significant reduc-174 tion of net primary production in climate projections, though with large uncertainties 175 (Kwiatkowski et al., 2020; Tagliabue et al., 2021). This might impact local respiration 176 rates and oxygen concentrations. Finally, the NASG and TSA sections experience dif-177 ferent circulation patterns and water masses compositions (Poole & Tomczak, 1999), which 178 might lead to a different behaviour of OUR under a changing climate. 179

Along the chosen sections, we compare the diagnosed OUR and the true respira-180 tion, integrated vertically over the mesopelagic zone rather than on individual density 181 surfaces. We notice a sharp change of density surface depth in the transient simulations, 182 e.g., in the NASG section, the depth of potential density surface 26.5  $kg/m^3$  increases 183 from  $257.1 \pm 51.4$  to  $420.7 \pm 4.3$  m from 1850 to 2099. Such changes of isopycnal mean 184 depth can, in itself, induce significant changes in the true respiration on the respective 185 density surfaces (Figure S3). Vertical displacement of isopycnals may mislead the anal-186 ysis of how the biology-induced respiration evolves under changing climate. Hence we 187 decide to explore temporal trends of vertically integrated  $R_{\rm true}$  and OUR in this study. 188

We derive the vertical integrals of  $R_{\text{true}}$  and OUR from 200m to 1000m as follows. 189 Firstly, we calculate the potential density from potential temperature, salinity, and sur-190 face reference pressure (0 decibar) using the 1980 UNESCO International Equation of 191 State (Millero & Poisson, 1981), and remap data from z-coordinates to sigma0-coordinates. 192 The OUR is calculated for every 0.1  $kg/m^3$  density surface interval from 24.1 to 28.0 kg/m<sup>3</sup> 193 by using the linear least square regression of AOU versus ideal age. We also calculate 194 the area-weighted mean true respiration and mean depth for each density surface. Af-195 terwards, we remap OUR and mean  $R_{\rm true}$  onto z-coordinates using the area-weighted 196 depth of the corresponding density surfaces. Finally, the vertical integral is obtained as 197 the sum of the density layer thickness times the variable (OUR or mean  $R_{\rm true}$ ) at each 198 grid point; see details in the supplementary material Text S2. The transformation of co-199 ordinates forth and back does not induce biases on either the vertical distribution of true 200 respiration rate or the trend of integrated  $R_{\rm true}$  (Figure S4). 201

## <sup>202</sup> 3 Results and discussion

Depth (200m to 1000m) integrated OUR underestimates integrated  $R_{true}$  along the NASG section by around 1.9 fold (Figure 2a), and most of the underestimation occurs in the upper ocean (Figure 2b). This underestimate may be caused by the spatial heterogeneity of respiration on isopycnals, as found for idealized isopycnals with prescribed



Figure 1. Distributions of ideal age (shading), AOU (red lines), and outcrop locations (lightblue lines) in the North and the South Atlantic Ocean on three isopycnal surfaces. We use the 2090-2099 mean in the esm-piControl simulation and exclude the waters above the respective hemispheric winter mixing depth. Winter surface density outcrops are calculated from March (northern hemisphere) and September (southern hemisphere) mean temperature and salinity. Blue lines represent chosen study sections.

patterns of respiration (Koeve & Kähler, 2016). High respiration occurs near the out-207 crops (shallower ocean) due to the high availability of substrates and low sinking speed 208 of detritus. But, mixing with nearby surface waters (with high oxygen concentration) 209 does not allow the imprint of respiration, AOU, to be well preserved. On the other hand, 210 respiration far from outcrops (deeper ocean) can be better preserved in terms of AOU 211 because of smaller mixing losses to surface waters. However, respiration in deeper wa-212 ters is often much lower because much organic matter has already been consumed in the 213 water column above, and because the sinking speed of particulate organic matter increases 214 with depth (Berelson, 2001). In contrast to respiration, the water aging rate is the same 215 everywhere. Therefore, more of the idealized age tracer is preserved on isopycnals com-216 pared to AOU, and consequently the OUR diagnosed from the ratio of AOU gradient 217 to ideal age gradient is smaller than  $R_{\rm true}$ . The underestimate of OUR compared to the 218  $R_{\rm true}$ , and the vertical distribution of their difference, has also been found in the other 219 two ensemble members with similar magnitude (Table 1, Figure S5ab, S6ab). 220

At the 95% confidence level, the esm-piControl experiment shows an OUR trend of  $-0.241 \pm 0.239 \text{ mmol/m}^2/\text{yr}^2$ , while the trend of true respiration is  $+0.282\pm0.191 \text{ mmol/m}^2/\text{yr}^2$ along the NASG section (Figure 2c). These small drifts in the true respiration and OUR in the esm-piControl simulations might result from a too short spin-up or from internal variability in the Earth system model (Matthes et al., 2020). Over the entire period from 1850 to 2100, the Pearson correlation coefficient between the diagnosed OUR and the true respiration in esm-piControl is 0.48.

The true respiration along the NASG section shows a long-term drift-corrected de-228 creasing trend along with global warming in the transient simulations (Figure 2d). Here 229 we use the transient simulation minus esm-piControl to remove the trend in the latter 230 and isolate the climate change signal. The true respiration decreases with the rate of 0.872 231  $\pm 0.268 \text{ mmol/m}^2/\text{yr}^2$  from 1850 to 2100, indicating a total decline of up to 10.6 % of 232 local mesopelagic respiration. Overall, mesopelagic respiration is reduced by about 0.2233 mmol/m<sup>2</sup>/yr in the NASG section by the end of this century. Average trend and total 234 decline are the same in the other two ensemble members at the 95% confidence level (Ta-235 ble 1). A decline in respiration can be related to enhanced stratification (Figure S9a) and 236 reduced nutrient supply, as has been suggested as an explanation for reduced net pri-237 mary production in the tropical and temperate regions in earlier model simulations (Bopp 238 et al., 2013; Kwiatkowski et al., 2020). 239

For the high emission climate change scenario simulated here, the drift-corrected 240 vertical OUR integral is suitable to track the long-term trend of the vertical  $R_{\rm true}$  in-241 tegral along the NASG section (Figure 2d). The OUR integral trend is -0.802  $\pm$  0.333 242 mmol/m<sup>2</sup>/yr<sup>2</sup>, which overlaps statistically with the  $R_{\rm true}$  changes of -0.872  $\pm$  0.268 mmol/m<sup>2</sup>/yr<sup>2</sup>. 243 Besides, the Pearson correlation coefficient between the diagnosed OUR integral and  $R_{\rm true}$ 244 integral is 0.54 for the transient minus esm-piControl. In the other two ensemble mem-245 bers, the trends of OUR and  $R_{\rm true}$  integral versus time also overlap at the 95% confi-246 dence level (Table 1). 247

In our second study region, the TSA section, the vertically integrated OUR amounts 248 to 83.3% of the  $R_{\rm true}$  integral in the esm-piControl simulation (Figure 3a), and the un-249 derestimation of true respiration by OUR occurs between 200m and 300m (Figure 3b). 250 A similar explanation as put forward for the NASG section above can be proposed here. 251 Below around 320m, however, OUR overestimates true respiration and even increases 252 with depth between 300m to 400m (Figure 3b). Both vertical integrals of OUR and  $R_{\rm true}$ 253 do not show a significant temporal trend in the esm-piControl simulation of the ensem-254 255 ble member 1 (Figure 3c), but the Pearson correlation coefficient between them is only 0.193. When calculating the transient tendencies corrected by the esm-piControl, ver-256 tically integrated OUR and  $R_{\text{true}}$  show significantly diverging trends (Figure 3d). The 257  $R_{\rm true}$  integral decreases with a rate of  $1.311\pm0.197 \text{ mmol/m}^2/\text{yr}^2$ , i.e., between 1850 and 258 2100, mesopelagic respiration decreases by about  $0.3 \text{ mmol/m}^2/\text{yr}$ . In contrast, the OUR 259



Figure 2. Comparison between vertically integrated OUR and true respiration (200 - 1000m) along the NASG section (see blue line in Figure 1a) in simulation ensemble member 1. Panel (a) shows absolute vertically integrated OUR (blue) and true respiration (red) from the transient simulation (thick) and the esm-piControl simulation (thin). Panel (b) is the vertical distribution of OUR (blue) and true respiration (red) in the last year of the esm-piControl simulation. The shading indicates one standard deviation, while OUR standard deviation is too small to be visible. All  $r^2$  values of the linear regressions of AOU against ideal age on density surfaces are above 0.92. Panel (c) shows the integrated OUR (thick blue line) and true respiration (thick red line) anomalies relative to the respective time-averaged values of the esm-piControl simulation. The thin solid lines are the linear least square regression lines of integrated OUR anomalies (blue) and integrated OUR (blue) and true respiration (red), and their linear least square regression lines of integrated OUR anomalies (red) over time. (d) shows the transient minus esm-piControl integrated OUR (blue) and true respiration (red), and their linear least square regression lines over time.



Figure 3. As Figure 2, but for the TSA section (see blue line in Figure 1d). In panel (b), all  $r^2$  values of the linear regressions of AOU against ideal age on density surfaces are above 0.94. The standard deviation of true respiration rates and OUR is too small to be visible.

integral shows a significant increasing trend of  $0.905 \pm 0.257 \text{ mmol/m}^2/\text{yr}^2$ . In other words, OUR integral suggests that the mesopelagic respiration in the TSA section will increase by 0.2 mmol/m<sup>2</sup>/yr by 2100. The other two ensemble members show similar features (Table 1, Figure S7, S8). To sum up the above results, we propose that along the TSA section, the trend of the OUR integral is not primarily determined by changes in local respiration.

What are potential reasons for the diverging trends in the vertical integrals of OUR 266 and local true respiration  $(R_{\rm true})$  along the TSA section? First, local  $R_{\rm true}$  decreases with 267 time, likely because of the enhanced stratification (Figure S9a) as seen also at the NASG 268 section and in other model results (Bopp et al., 2013; Kwiatkowski et al., 2020). The in-269 creasing OUR may indicate either an increasing AOU gradient along the section, decreas-270 ing age gradient, or a combination of both (see Equation 1). Here we propose two pos-271 sible reasons for the increasing OUR in the TSA section: (i) Ocean circulation changes 272 in a warming ocean might change water mass composition in the study region, that is 273 water masses with different biogeochemical histories are recombined over time in differ-274 ent ways, potentially causing a change in tracer distribution, e.g. a steepening of the AOU 275 versus age regression line. (ii) In regions where source water masses observed in the study 276 region form, ocean biogeochemistry also responds to climate change. For example, in the 277 region of formation of Subantarctic Mode Waters (SAMW), we found a strong increase 278

**Table 1.** Ratio of vertically (200m to 1000m) integrated true respiration ( $R_{true}$ ) and OUR, trend of  $R_{true}$  and OUR over time and its 95% confidence interval, and Pearson correlation coefficient between integrated  $R_{true}$  and OUR (with p-value below 0.05). "n.s." represents the non-significant outcomes (with p-value above 0.05).

Region	simulations	ensemble members	$R_{\rm true}/{\rm OUR}$	trend of $R_{\rm true}^a$ (mmol/m <sup>2</sup> /yr <sup>2</sup> )	trend of $OUR^a$ (mmol/m <sup>2</sup> /yr <sup>2</sup> )	correlation coefficient
NASG		1	1.9	$0.282{\pm}0.191$	$-0.241 \pm 0.239$	0.48
	esm-piControl	2	1.9	n.s.	$-0.288 \pm 0.246$	0.51
		3	1.9	$0.433 {\pm} 0.205$	n.s.	0.45
	transient minus	1	1.9	$-0.872 \pm 0.268$	$-0.802 \pm 0.333$	0.54
	esm-piControl	2	1.9	$-0.833 \pm 0.296$	$-0.844 \pm 0.357$	0.54
		3	1.9	$-0.981 {\pm} 0.272$	$-0.754 {\pm} 0.362$	0.55
TSA		1	1.2	n.s.	n.s.	0.19
	esm-piControl	2	1.2	$0.206{\pm}0.137$	n.s.	0.13
		3	1.2	$0.191{\pm}0.150$	$-0.316 {\pm} 0.188$	n.s.
	transient minus	1	1.0	$-1.311 \pm 0.197$	$0.905 {\pm} 0.257$	-0.19
	esm-piControl	2	1.0	$-1.783 \pm 0.200$	$0.520{\pm}0.268$	n.s.
		3	1.0	$-1.323 \pm 0.182$	$0.826{\pm}0.257$	-0.17

<sup>a</sup> For trends: positive number means increase and negative number means decrease.

in true respiration (Figure S9b). The associated AOU can then propagate with the SAMW into the study region and impact the regression relationship between AOU and age. All the above processes may be summed as an 'apparent' OUR contribution induced by the mixing-driven intrusion of AOU and age signals from outside of the study section. For clarification, we suggest to rewrite the OUR equation as follows:

$$OUR = R_{\rm true} + OUR_{\rm mixing} \tag{3}$$

in which,  $R_{\text{true}}$  is local OUR induced by the degradation of DOM and DET along the section, and  $OUR_{\text{mixing}}$  is 'apparent' OUR induced by the mixing of waters from different origins, yielding imprints of ocean biogeochemistry and water age from outside the study section (Sonnerup et al., 1999). We hypothesize that depending on the region under consideration,  $R_{\text{true}}$  might be either the main (e.g., in the NASG section) or a minor driver (e.g., in the TSA section) of OUR change.

To reliably infer temporal changes in  $R_{\rm true}$  from OUR under global warming, it is 291 required to remove the mixing-induced contributions to AOU and age. This has been 292 addressed in several studies assuming steady-state conditions, e.g. by applying the Op-293 timal Multi-Parameter (OMP) analysis (e.g., Karstensen & Tomczak, 1998; Brea et al., 294 2004; Alvarez-Salgado et al., 2014). However, this method requires temporally constant 295 properties (e.g., potential temperature, salinity) of water masses in their formation re-296 gions (Álvarez et al., 2014), which may limit its straightforward application in a chang-297 ing ocean. 298

#### <sup>299</sup> 4 Conclusions

284

Our study confirms the potential ability of OUR to track the trends of true respiration during the time period between 1850 and 2100, and may hence contribute to our understanding of drivers of ocean deoxygenation in parts of the ocean. However, there is also a risk that temporal trends diagnosed from OUR can be an unreliable indicator of trends in true respiration, and may even be of opposite sign to trends of true respiration in other parts of the ocean (as in the example of the tropical South Atlantic). As one potential reason of diverging trends between local respiration and OUR, we propose that climate-driven changes in ocean mixing, in water mass composition and in biogeochemical properties of individual water masses can map onto the local OUR. A quantification of climate-driven changes in aerobic respiration from OUR requires separating mixing-induced changes in AOU and age, including temporal changes of water masses composition and properties. This is challenging, as it requires careful consideration of climate-driven changes in (remote) areas of water mass formation, whenever mixing of multiple water masses is involved.

## <sup>314</sup> 5 Data availability statement

The model code is provided by Chien et al. (2022) at https://doi.org/10.5281/zenodo.6772175. The model outputs used in this paper, together with the scripts for data processing are available at: https://hdl.handle.net/20.500.12085/e7d53204-df0c-4973-8628-63dad7dd140.

## 318 Acknowledgments

We acknowledge discussions with colleagues from the Biogeochemical Modelling research 319 unit at GEOMAR, and Respiration in the Mesopelagic Ocean (ReMO) SCOR working 320 321 group. We would like to thank useful comments from two anymous reviewers, which greatly improved the manuscript. The authors acknowledge funding from the European Union's 322 Horizon 2020 research and innovation program under grant agreement no. 820989 (project 323 COMFORT, Our common future ocean in the Earth system-quantifying coupled cycles 324 of carbon, oxygen and nutrients for determining and achieving safe operating spaces with 325 respect to tipping points). The authors would like to thank Chia-Te Chien for provid-326 ing FOCI model data. We wish to acknowledge use of the Ferret program of NOAA's 327 Pacific Marine Environmental Laboratory for analysis and graphics featured in this pa-328 per. This is a contribution to Subtopic 6.3. 'The future biological carbon pump' of the 329 Helmholtz Research Program 'Changing Earth – Sustaining our Future'. 330

#### 331 References

- Álvarez, M., Brea, S., Mercier, H., & Álvarez-Salgado, X. A. (2014).Mineraliza-332 tion of biogenic materials in the water masses of the south atlantic ocean. i: 333 Assessment and results of an optimum multiparameter analysis. Progress in 334 Oceanography, 123, 1–23. 335 Alvarez-Salgado, X. A., Alvarez, M., Brea, S., Mémery, L., & Messias, M. (2014).336 Mineralization of biogenic materials in the water masses of the south atlantic 337 338 ocean. ii: Stoichiometric ratios and mineralization rates. Progress in Oceanography, 123, 24-37. 339 Berelson, W. M. (2001).Particle settling rates increase with depth in the ocean. 340 Deep Sea Research Part II: Topical Studies in Oceanography, 49(1-3), 237-341 251.342 Bopp, L., Resplandy, L., Orr, J. C., Doney, S. C., Dunne, J. P., Gehlen, M., ... 343 others (2013).Multiple stressors of ocean ecosystems in the 21st century: 344 projections with cmip5 models. Biogeosciences, 10(10), 6225-6245. 345 Brea, S., Alvarez-Salgado, X. A., Alvarez, M., Pérez, F. F., Mémery, L., Mercier, 346 H., & Messias, M.-J. (2004). Nutrient mineralization rates and ratios in the 347 eastern south atlantic. Journal of Geophysical Research: Oceans, 109(C5). 348 Breitburg, D., Levin, L. A., Oschlies, A., Grégoire, M., Chavez, F. P., Conley, D. J., 349 ... others (2018).Declining oxygen in the global ocean and coastal waters. 350 Science, 359(6371), eaam7240. 351 Chien, C.-T., Durgadoo, J. V., Ehlert, D., Frenger, I., Keller, D. P., Koeve, W., ... 352 others (2022). Foci-mops v1-integration of marine biogeochemistry within the 353 flexible ocean and climate infrastructure version 1 (foci 1) earth system model. 354 Geoscientific Model Development, 15(15), 5987–6024. 355 Del Giorgio, P., & Duarte, C. M. (2002). Respiration in the open ocean. Nature. 356 420(6914), 379-384.357 Del Giorgio, P., & Williams, P. (2005). Respiration in aquatic ecosystems. OUP Ox-358 ford. 359 Dietze, H., & Oschlies, A. (2005). Modeling abiotic production of apparent oxygen 360 utilisation in the oligotrophic subtropical north atlantic. Ocean Dynamics, 55, 361 28 - 33.362 Duteil, O., Koeve, W., Oschlies, A., Bianchi, D., Galbraith, E., Kriest, I., & Matear, 363 **R**.. (2013).A novel estimate of ocean oxygen utilisation points to a reduced 364 rate of respiration in the ocean interior. Biogeosciences, 10(11), 7723-7738. 365 Evring, V., Bony, S., Meehl, G. A., Senior, C. A., Stevens, B., Stouffer, R. J., & 366 Taylor, K. E. (2016). Overview of the coupled model intercomparison project 367 phase 6 (cmip6) experimental design and organization. Geoscientific Model 368 Development, 9(5), 1937-1958.369
- Filella, A., Baños, I., Montero, M. F., Hernández-Hernández, N., Rodríguez-Santos,
  A., Ludwig, A., ... Arístegui, J. (2018). Plankton community respiration and ets activity under variable co2 and nutrient fertilization during a mesocosm study in the subtropical north atlantic. *Frontiers in Marine Science*, 5, 310.
- Fine, R. A. (2011). Observations of cfcs and sf6 as ocean tracers. Annual review of marine science, 3, 173–195.
- Fine, R. A., Peacock, S., Maltrud, M. E., & Bryan, F. O. (2017). A new look at
   ocean ventilation time scales and their uncertainties. Journal of Geophysical
   *Research: Oceans*, 122(5), 3771–3798.
- Ito, T., Follows, M., & Boyle, E. (2004). Is aou a good measure of respiration in the oceans? *Geophysical research letters*, 31(17).
- Ito, T., Minobe, S., Long, M. C., & Deutsch, C. (2017). Upper ocean o2 trends:
   1958–2015. *Geophysical Research Letters*, 44 (9), 4214–4223.
- Jenkins, W. J. (1987). 3 h and 3 he in the beta triangle: Observations of gyre ventilation and oxygen utilization rates. *Journal of Physical Oceanography*, 17(6), 763–783.

386	Karstensen, J., & Tomczak, M. (1998). Age determination of mixed water masses
387	using cfc and oxygen data. Journal of Geophysical Research: Oceans, 103(C9),
388	18599 - 18609.
389	Koeve, W., & Kähler, P. (2016). Oxygen utilization rate (our) underestimates ocean
390	respiration: A model study. Global Biogeochemical Cycles, 30(8), 1166–1182.
391	Kriest, I., & Oschlies, A. (2015). Mops-1.0: towards a model for the regulation of
392	the global oceanic nitrogen budget by marine biogeochemical processes. Geo-
393	scientific Model Development, 8(9), 2929–2957.
394	Kwiatkowski, L., Torres, O., Bopp, L., Aumont, O., Chamberlain, M., Christian,
395	J. R., others (2020). Twenty-first century ocean warming, acidification,
396	deoxygenation, and upper-ocean nutrient and primary production decline from
397	cmip6 model projections. <i>Biogeosciences</i> , 17(13), 3439–3470.
398	Martin, J. H., Knauer, G. A., Karl, D. M., & Broenkow, W. W. (1987). Vertex: car-
399	bon cycling in the northeast pacific. Deep Sea Research Part A. Oceanographic
400	Research Papers, 34(2), 267–285.
401	Matthes, K., Biastoch, A., Wahl, S., Harlaß, J., Martin, T., Brücher, T., others
402	(2020). The flexible ocean and climate infrastructure version 1 (foci1): mean
403	state and variability. Geoscientific Model Development, 13(6), 2533–2568.
404	Millero, F. J., & Poisson, A. (1981). International one-atmosphere equation of state
405	of seawater. Deep Sea Research Part A. Oceanographic Research Papers, 28(6),
406	625-629.
407	Oschlies, A., Brandt, P., Stramma, L., & Schmidtko, S. (2018). Drivers and mecha-
408	nisms of ocean deoxygenation. Nature Geoscience, 11(7), 467–473.
409	Pezner, A. K., Courtney, T. A., Barkley, H. C., Chou, WC., Chu, HC., Clements,
410	S. M., others (2023). Increasing hypoxia on global coral reefs under ocean
411	warming. Nature Climate Change, 1–7.
412	Poole, R., & Tomczak, M. (1999). Optimum multiparameter analysis of the wa-
413	ter mass structure in the atlantic ocean thermocline. Deep Sea Research Part
414	I: Oceanographic Research Papers, 46(11), 1895–1921.
415	Robinson, C. (2019). Microbial respiration, the engine of ocean deoxygenation. Fron-
416	tiers in Marine Science, 533.
417	Schmidtko, S., Stramma, L., & Visbeck, M. (2017). Decline in global oceanic oxygen
418	content during the past five decades. Nature, 542(7641), 335–339.
419	Sonnerup, R. E., Chang, B. X., Warner, M. J., & Mordy, C. W. (2019). Timescales
420	of ventilation and consumption of oxygen and fixed nitrogen in the eastern
421	tropical south pacific oxygen deficient zone from transient tracers. Deep Sea
422	Research Part I: Oceanographic Research Papers, 151, 103080.
423	Sonnerup, R. E., Mecking, S., & Bullister, J. L. (2013). Transit time distributions
424	and oxygen utilization rates in the northeast pacific ocean from chlorofluoro-
425	carbons and sulfur hexafluoride. Deep Sea Research Part I: Oceanographic
426	Research Papers, 72, 61–71.
427	Sonnerup, R. E., Mecking, S., Bullister, J. L., & Warner, M. J. (2015). Transit time
428	distributions and oxygen utilization rates from chlorofluorocarbons and sulfur
429	hexafluoride in the s outheast p acific o cean. Journal of Geophysical Research:
430	Oceans, 120(5), 3761-3776.
431	Sonnerup, R. E., Quay, P. D., & Bullister, J. L. (1999). Thermocline ventilation and
432	oxygen utilization rates in the subtropical north pacific based on cfc distribu-
433	tions during woce. Deep Sea Research Part 1: Oceanographic Research Papers,
434	4b(5), 777-805.
435	Stoven, T., Tanhua, T., Hoppema, M., & Bullister, J. (2015). Perspectives of tran-
436	signt tracer applications and limiting cases. Ocean Science, 11(5), 699–718.
437	Stramma, L., Johnson, G. C., Sprintall, J., & Mohrholz, V. (2008). Expanding
438	oxygen-minimum zones in the tropical oceans. <i>science</i> , $320(5876)$ , $655-658$ .
439	Tagliabue, A., Kwiatkowski, L., Bopp, L., Butenschön, M., Cheung, W., Lengaigne,
440	M., & Vialard, J. (2021). Persistent uncertainties in ocean net primary produc-

tion climate change projections at regional scales raise challenges for assessing impacts on ecosystem services. *Frontiers in Climate*, 149.

- Thiele, G., & Sarmiento, J. (1990). Tracer dating and ocean ventilation. Journal of
   *Geophysical Research: Oceans*, 95(C6), 9377–9391.
- Waniek, J., Koeve, W., & Prien, R. D. (2000). Trajectories of sinking particles and
   the catchment areas above sediment traps in the northeast atlantic. Journal of
   Marine Research, 58(6), 983-1006.
- Williams, P. (1981). Microbial contribution to overall marine plankton metabolism direct measurements of respiration. Oceanologica Acta, 4(3), 359–364.