



# 1 Climatic Controls on Metabolic Constraints in the Ocean

2

3 Precious Mongwe<sup>1</sup>, Matthew Long<sup>2</sup>, Takamitsu Ito<sup>3</sup>, Curtis Deutsch<sup>4</sup>, and Yeray

4 Santana-Falcón<sup>5</sup>

5 <sup>1</sup>Southern Ocean Carbon Climate Observatory (SOCCO), CSIR, Cape Town, South Africa

6 <sup>2</sup>Oceanography Section, Climate and Global Dynamics Laboratory, National Center for Atmospheric Research,  
7 Boulder, CO, United States of America

8 <sup>3</sup>School of Earth and Atmospheric Sciences, Georgia Institute of Technology, Atlanta, Georgia United States of  
9 America

10 <sup>4</sup>Department of Geosciences, Princeton University, Princeton, NJ, United States of America

11 <sup>5</sup>CNRM, Université de Toulouse, Météo-France, CNRS, Toulouse, 31057, France

12 **Corresponding Author:** Precious Mongwe (pmongwe@csir.co.za)

## 13 Abstract

14 Observations and models indicate that climate warming is associated with the loss of dissolved  
15 oxygen from the ocean. Dissolved oxygen is a fundamental requirement for heterotrophic marine  
16 organisms (except marine mammals) and, since the basal metabolism of ectotherms increases  
17 with temperature, warming increases organisms' oxygen demand. Therefore, warming and  
18 deoxygenation pose a compound threat to marine ecosystems. In this study, we leverage an  
19 ecophysiological framework and compilation of empirical trait data quantifying the temperature  
20 sensitivity and oxygen requirements of metabolic rates for a range of marine species  
21 ("ecotypes"). Using the Community Earth System Model Large Ensemble, we investigate how  
22 natural climate variability and anthropogenic forcing impact the ability of marine environments  
23 to support aerobic metabolisms on interannual to multi-decadal timescales. Warming and  
24 deoxygenation projected over the next several decades will yield a reduction in the volume of  
25 viable ocean habitat. We find that fluctuations in temperature and oxygen associated with natural  
26 variability are distinct from those associated with anthropogenic forcing in the upper ocean.  
27 Further, the joint temperature-oxygen anthropogenic signals emerges sooner than independently  
28 from natural variability. Our results demonstrate that anthropogenic perturbations underway in  
29 the ocean will strongly exceed those associated with the natural system; in many regions,  
30 organisms will be pushed closer to or beyond their physiological limits, leaving the ecosystem  
31 more vulnerable to extreme temperature-oxygen events.



## 32 1. Introduction

33 Dissolved oxygen ( $O_2$ ) is a fundamental metabolic requirement for heterotrophic marine  
34 organisms, excluding marine mammals (Portner, 2002; Keeling et al., 2010; Tiano et al., 2014).  
35  $O_2$  is declining due to warming, a tendency long predicted by models (Keeling et al., 2010; Long  
36 et al., 2016; Oschlies et al., 2018) and recently found evident at the global scale in compilations  
37 of in situ observations (Schmidtko et al., 2017; Ito et al., 2017). Deoxygenation is driven by the  
38 direct effect of reduced oxygen solubility with warming compounded by buoyancy-induced  
39 stratification in the upper ocean, which weakens the ventilation-mediated supply of fresh oxygen  
40 to the ocean interior. While the full ecological impacts of ocean deoxygenation remain uncertain,  
41 it is clear that the physiological impacts of oxygen loss on marine organisms can be considered  
42 explicitly in the context of warming: basal metabolic rates for ectothermic organisms depend on  
43 ambient temperature and increase with warming (Gillooly et al., 2001); thus, higher temperatures  
44 impose additional demand for oxygen to sustain aerobic respiration (Deutsch et al., 2015).  
45 Consequently, as the ocean warms, even present-day oxygen distributions may be insufficient to  
46 meet the oxygen demands of organisms living near key physiological thresholds (Deutsch et al.,  
47 2022).

48  
49 While model projections clearly demonstrate that warming and deoxygenation are consequences  
50 of human-driven climate change, it is important to recognize that natural climate variability also  
51 produces important fluctuations in these quantities. Indeed, evidence suggests that natural  
52 variability contributes to hypoxic events, such as those observed in the California Current, where  
53 fish and benthic-organism mortality has been associated with low- $O_2$  waters impinging on the  
54 continental shelf (Pozo Buil and Di Lorenzo, 2017; Howard et al., 2020). A clear understanding  
55 of how natural climate variability drives fluctuations in metabolic state and the associated  
56 implications for organisms is a critical context in which to view long-term climate warming.  
57 Given that the natural system is highly dynamic, climate change signals are often masked by  
58 decadal-scale variability (Ito and Deutsch, 2010). While numerous authors have considered  
59 detection and attribution of climate change for physical and biogeochemical variables (Rodgers  
60 et al., 2015; Long et al., 2016; Schlunegger et al., 2019), comparatively little attention has been  
61 devoted to explicitly characterizing the relative influence of natural and anthropogenic drivers of  
62 changes in the ocean's capacity to support aerobic life. In this study, we approach this challenge



63 by leveraging the concept of the Metabolic Index ( $\Phi$ ) introduced by Deutsch et al. (2015).  $\Phi$  is  
64 based on the notion that aerobic organisms can persist only where the ambient oxygen partial  
65 pressure ( $pO_2$ ) is sufficient to meet the requirements of sustaining respiration.  $\Phi$  incorporates an  
66 explicit representation of the dependence of metabolic oxygen demand on temperature, thus  
67 providing a framework to consider how joint oxygen and temperature variability constrain viable  
68 habitat in the ocean.

69

70 Many ocean organisms may already be under threat from deoxygenation (Hoegh-Guldberg and  
71 Bruno, 2010; Breitburg et al., 2018); however, ongoing climate-driven loss of oxygen raises  
72 important questions about the future of marine ecosystems: How will anthropogenic changes in  
73 dissolved oxygen and temperature impact the capacity of ocean habitats to support aerobic  
74 metabolism? What is the spatial and temporal distribution of changes in the ocean's metabolic  
75 state associated with climate variability? At what point can anthropogenic change in the ocean's  
76 metabolic state be distinguished from natural variability? This study addresses these questions  
77 using a combination of metabolic theory, a dataset set quantifying key physiological parameters  
78 for a collection of marine species adapted to specific environments ("ecotypes"), and the oxygen  
79 and temperature distributions simulated in the Community Earth System Model, version 1 Large  
80 Ensemble (CESM1-LE), which includes 34 members simulating ocean biogeochemistry under  
81 climate variability and change from 1920–2100 forced using historical data and the  
82 Representative Concentration Pathway Scenario 8.5 (RCP85) (Kay et al., 2015; Long et al.,  
83 2016).

84

85 This paper is organized as follows. Section 2 presents a brief overview of the relevant metabolic  
86 theory, the associated empirical datasets, and describes our approach to analysis. In Section 3 we  
87 present results quantifying the joint temperature-oxygen variability simulated in the CESM1-LE,  
88 evaluating the spatiotemporal structure of variability in marine ecotype habitat, including long-  
89 term trends based on the RCP8.5 scenario and time of emergence (ToE). The main outcomes of  
90 the results are synthesized in Section 4 and summarized in Section 5.

91

92 **2. Datasets and methods**



## 93 2.1 Metabolic index

94 Empirical studies measuring thermal tolerance and oxygen requirements in the laboratory on an  
95 array of marine organisms have enabled an assessment of lethal thresholds (Vaquer-Sunyer and  
96 Duarte, 2008; Rosewarne et al., 2016). These data coupled with recent advances in a theoretical  
97 framework enable both explanatory and predictive power in the context of a dynamic  
98 environment (Deutsch et al., 2015; Penn et al., 2018; Howard et al., 2020). The fundamental  
99 insights here are that basal metabolic rates for ectothermic marine organisms depend on ambient  
100 temperature and generally increase with warming (Gillooly et al., 2001). Increasing basal  
101 metabolic rates impose additional demand for oxygen. Organisms use oxygen dissolved in  
102 seawater and acquisition tends to be limited by diffusive processes; thus, oxygen supply is  
103 related to the ambient  $pO_2$ . The ratio of oxygen supply to temperature-dependent demand  
104 provides a critical indicator of the capacity for an organism to meet its metabolic requirements.  
105 Deutsch et al. (2015) formalized these concepts into a quantity termed the “Metabolic Index  
106 ( $\Phi$ )”, which is defined as the ratio of oxygen supply to an organism’s resting metabolic demand.  
107 Oxygen supply is parameterized according to a biomass-dependent scaling of  $pO_2$ , capturing  
108 variation in the efficiency with which organisms acquire and utilize  $O_2$ . This can be expressed as  
109  $S = \alpha_S B^\sigma pO_2$ , where  $\alpha_S$  is a mass-normalized coefficient expressing the rate of gas transfer  
110 between an organism and its environment and  $B^\delta$  is the scaling of supply with biomass,  $B$  (Piiper  
111 et al., 1971). Resting metabolic demand can be expressed using the Arrhenius equation as

$$112 \quad D = \alpha_D B^\delta \exp\left\{\frac{-E_d}{k_B} \left[\frac{1}{T} - \frac{1}{T_{ref}}\right]\right\},$$

113 where  $\alpha_D$  is a species-specific basal metabolic rate,  $B^\delta$  is the scaling of this rate with biomass,  $E_d$   
114 (eV) is the temperature dependence of oxygen supply,  $T$  is temperature,  $T_{ref}$  is the reference  
115 temperature (15°C), and  $k_B$  is the Boltzmann constant (Gillooly et al., 2001). Gas transfer is  
116 kinematically slow at low temperatures, and hence organism viability can be limited by the  
117 energy to acquire oxygen at low temperatures, thus  $E_o$  varies with temperature. Here we account  
118 for this by adding the temperature dependence ( $dE_o/dT$ ) to  $E_o$  in equations above ( $E_o + \frac{dE_o}{dT}(T -$   
119  $T_{ref})$ ), using the mean value of  $dE_o/dT = 0.022$  eV consistent with Deutsch et al. (2020). The  
120 Metabolic Index can thus be written as the ratio of  $S/D$ :

$$121 \quad \Phi = A_o pO_2 \exp\left\{\frac{-E_o}{k_B} \left[\frac{1}{T} - \frac{1}{T_{ref}}\right]\right\}, \quad (1)$$



122

123 where  $A_o = \alpha_s/\alpha_D$  ( $l/atm$ ) is the hypoxic tolerance,  $E_o = E_d - E_s$  ( $E_s$  is the temperature  
124 dependence of oxygen supply) (Deutsch et al., 2015; Penn et al., 2018). The exponent,  $\varepsilon = \sigma -$   
125  $\delta$ , is the allometric scaling of the supply to demand ratio with biomass, is typically near zero.  
126 Therefore, in the analysis that follows, we presume unit biomass and thus neglect potential  
127 impacts of variations in biomass.

128

129 If  $\Phi$  falls below a critical threshold value of 1, conditions are physiologically unsustainable: an  
130 organism cannot meet its basic resting metabolic oxygen requirements. Conversely, values of  $\Phi$   
131 above 1 enable organismal metabolic rates to increase by a factor of  $\Phi$  above resting levels,  
132 permitting critical activities such as feeding, defence, growth, and reproduction. Thus, for a  
133 given environment and species,  $\Phi$  provides an estimate of the ratio of maximum sustainable  
134 metabolic rate to the minimum rate necessary for basal metabolism. Deutsch et al. (2015)  
135 inferred the ratio of active to resting energetic demand by examining the biogeographic  
136 distribution of several species, finding that range boundaries coincide with values of  $\Phi = 1.5-7$ .  
137 This threshold, termed critical rate ( $\Phi_{crit}$ ), represents the minimum metabolic index required for  
138 an organism to sustain an active metabolic state, which is a more meaningful ecological  
139 threshold than requirements for resting metabolism. Therefore, in this study, we define a quantity  
140  $\Phi'$ , which is derived by dividing  $\Phi$  by  $\Phi_{crit}$ ; equivalently, this yields an adjusted definition of the  
141 hypoxic tolerance trait,  $A_c = A_o / \Phi_{crit}$ , where  $A_c$  is termed the “ecological hypoxia tolerance”  
142 consistent with Howard et al. (2020). Where  $\Phi' > 1$  (i.e.,  $\Phi > \Phi_{crit}$ ) an organism can sustain an  
143 active metabolic rate; where  $\Phi' < 1$  (i.e.,  $\Phi < \Phi_{crit}$ ),  $O_2$  is insufficient and an active metabolic  
144 state is not viable. Henceforth, our analysis uses  $\Phi'$  to characterize ecotypes viability.

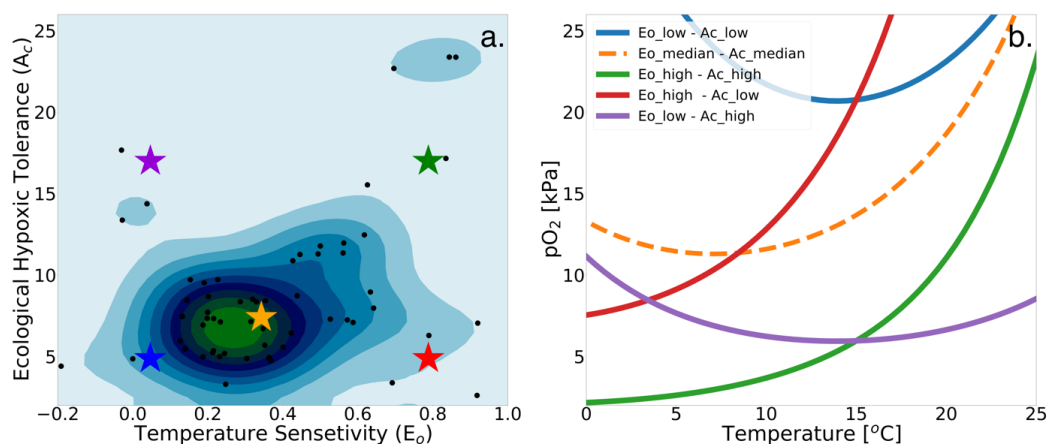
145

## 146 **2.2 Physiological dataset**

147 We make use of a dataset describing physiological parameters for a collection of 61 marine  
148 ecotypes spanning a range of ecological hypoxic tolerances ( $A_c$ ) and temperature sensitivities  
149 ( $E_o$ ) (Penn et al., 2018; Deutsch et al., 2020, Figure 1a). We illustrate how the physiological traits  
150  $E_o$  and  $A_c$  constrain habitat viability in the context of distributions of  $pO_2$  and temperature in the  
151 marine environment in Figure 1b, which shows the minimum  $pO_2$  (i.e.,  $pO_2$  at  $\Phi_{crit}$ ) required to  
152 sustain an active metabolic state as a function of temperature for five combinations of  $E_o$  and  $A_c$ .



153 The five combinations are derived from sampling the probability distributions of  $E_o$  and  $A_c$   
154 (Figure 1a) at the 10<sup>th</sup>, 50<sup>th</sup>, and 90<sup>th</sup> percentile values (illustrated by colored stars in Figure 1a  
155 and corresponding curves in Figure 1b). We assume that the trait distributions are independent,  
156 which is a reasonably modest simplification;  $E_o$  is represented by a normal distribution and  $A_c$  by  
157 a lognormal distribution function (Figure S1). The  $pO_2$  at  $\Phi_{crit}$  curves shown in Figure 1b  
158 delineate regions of  $pO_2$ -temperature space that are habitable (above the curve) and  
159 uninhabitable (below the curve). The reversing curvature of  $pO_2$  at  $\Phi_{crit}$  in Figure 1b at low  
160 temperature captures the decrease of the organism's oxygen acquisition efficiency in cooler  
161 conditions yielding cold intolerance.

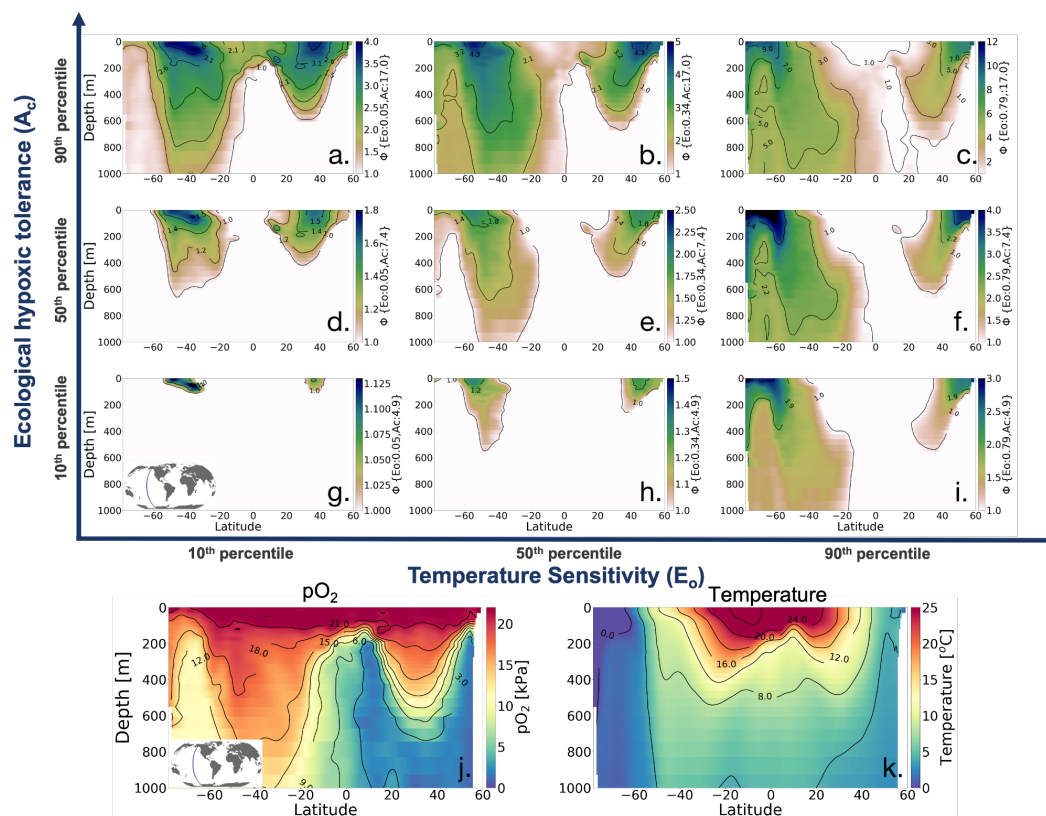


162  
163 **Figure 1.** Physiological traits determining hypoxic tolerance. (a) Scatter plot of 61 marine ecotypes for which  
164 empirically derived estimates of activation energy ( $E_o$ ) and the ecological hypoxic tolerance ( $A_c$ ) have been  
165 determined (Penn et al., 2018). The color shows the density of occurrence for the 61 marine ecotypes in the  $A_c - E_o$   
166 trait space. (b) The minimum  $pO_2$  required to sustain an active metabolic state (i.e.,  $pO_2$  at  $\Phi_{crit}$ , Deutsch et al., 2020)  
167 for five combinations of  $A_c$  and  $E_o$  corresponding to the stars in panel “a”; these are combinations of the 10th, 50th,  
168 90th percentile values for each parameter.

169  
170 To illustrate how the trait combinations of  $E_o$  and  $A_c$  exert control on the geographic distribution  
171 of organisms in the marine environment (Deutsch et al., 2020), we use observations of  $pO_2$  and T  
172 along a zonal transect of the Pacific Ocean and plot  $\Phi'$  for nine combinations of  $E_o$  and  $A_c$   
173 percentile values (Figure 2). The colorbar in Figures 2a-i show the metabolic index for an active  
174 state ( $\Phi'$ ); regions with values above one are habitable (color), while regions with values below



175 one are uninhabitable (white) on the basis of metabolic constraints (other ecological  
176 considerations are not considered). The subplots in the upper portion of the figure are arranged  
177 according to the same trait axes shown in Figure 1a;  $E_o$  increases horizontally from left to right  
178 and  $A_c$  increases from the bottom to the top. For the trait combination in the bottom left (low  $E_o$ ,  
179 low  $A_c$ ; Figure 2g), metabolism is relatively insensitive to temperature, and tolerance for low  
180  $pO_2$  is poor. Thus, ecotypes with low  $E_o$  and low  $A_c$  are restricted to high latitude surface waters,  
181 where temperatures are cool, and  $pO_2$  is abundant (Figure 2g). As  $E_o$  increases from left to right,  
182 metabolic rates become more sensitive to temperature. Then, habitat is gained at depth, where  
183 temperatures are cooler and higher temperature sensitivity confers an advantage (Figure 2g–i).  
184 From the bottom to the top, the increase in tolerance of low  $pO_2$  conditions increases habitability  
185 in regions of low  $pO_2$ , enabling organisms to expand beyond high-latitude surface waters (Figure  
186 2g-a). The biogeographic range for organisms with high  $A_c$  is modulated by  $E_o$ ; as temperature  
187 sensitivity increases, ecotype viability at high latitudes is increased, but tropical surface waters  
188 become less viable (Figure 2 a-c). Henceforth, our analysis will utilize the metabolic index of the  
189 median ecotype ( $E_o = 0.34$ ,  $A_c = 7.4$ ; Figure 2e) for illustrative purposes; i.e., all metabolic index  
190 figures refer to this median ecotype unless otherwise stated.



191

192 **Figure 2.** Annual mean metabolic index ( $\Phi$ ) for nine combinations of the ecological traits  $E_o$  (metabolic  
 193 temperature sensitivity) and  $A_c$  (ecological hypoxic tolerance) along a transect in the Pacific Ocean based on a  
 194 climatology from the World Ocean Atlas dataset (Garcia et al., 2014). The percentile values of each trait are: 10<sup>th</sup> ( $E_o$   
 195 = 0.04,  $A_c$  = 4.8), 50<sup>th</sup> ( $E_o$  = 0.34,  $A_c$  = 7.4), and 90<sup>th</sup> ( $E_o$  = 0.79,  $A_c$  = 17.0). The lower panels show  $pO_2$  and  
 196 temperature from the WOA dataset. Note that the colorbar range differs by panel and values where  $\Phi' < 1$  are  
 197 omitted, thus the color shows only areas where an active metabolic state can be sustained.

198

### 199 2.3 Earth system model simulations

200 This study is based on the CESM1-LE, described in detail by Kay et al. (2015). The CESM1-LE  
 201 included 34 ensemble members integrated from 1920–2100 under historical and RCP8.5 forcing.  
 202 The ensemble was generated by adding round-off level ( $10^{-14}$  K) perturbations to the air  
 203 temperature field at initialization in 1920; this small difference yields rapidly diverging model  
 204 solutions due to the chaotic dynamics intrinsic to the climate system, thus developing ensemble  
 205 spread representative of internal variability (Kay et al., 2015). Briefly, the CESM1-LE uses the



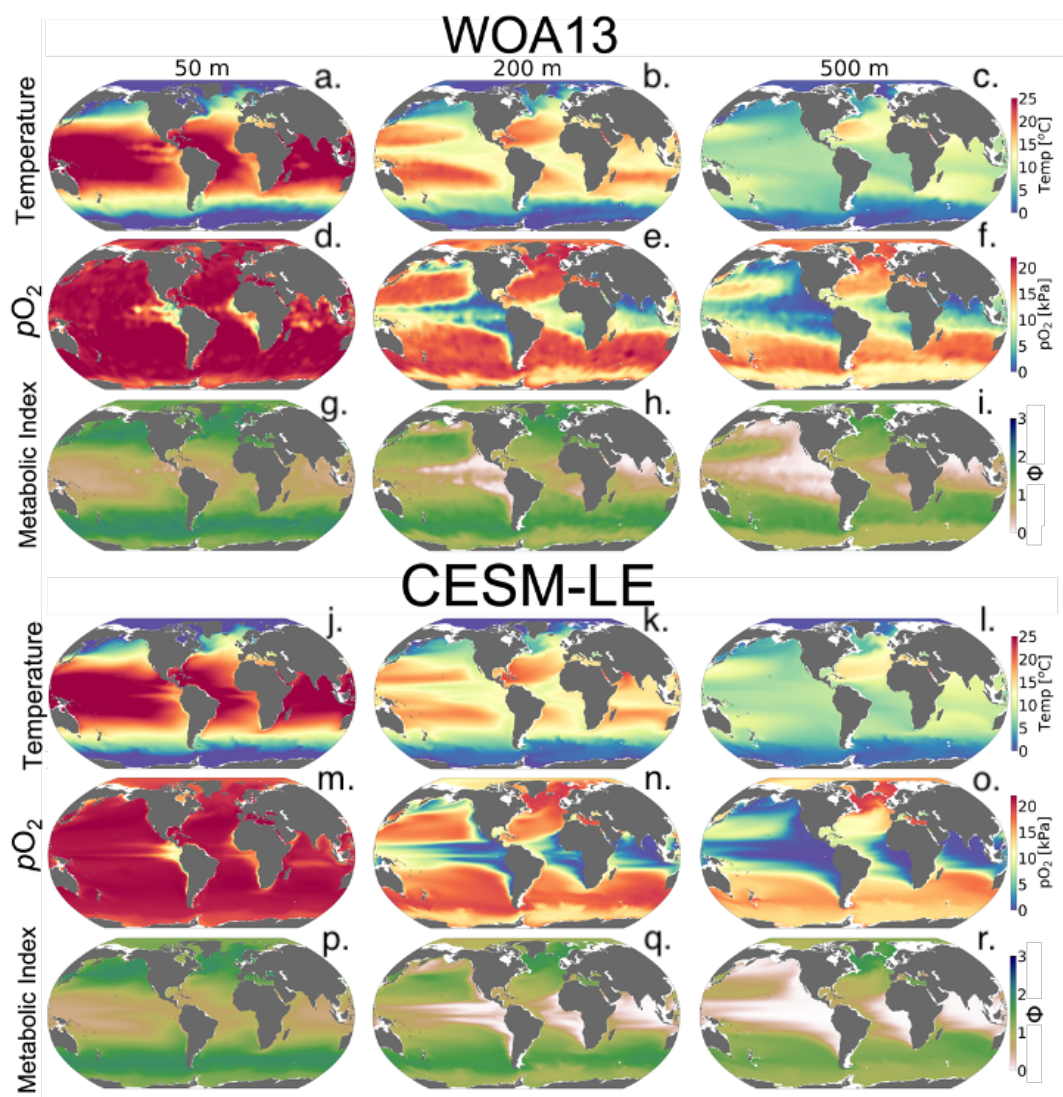


206 Community Earth System Model, version 1 (Hurrell et al., 2013), with a horizontal resolution of  
207 nominally  $1^\circ$  in all components. The ocean component is Parallel Ocean Program version 2,  
208 (Smith et al., 2010) with sea ice simulated by the Los Alamos Sea Ice Model version 4 (Hunke  
209 and Lipscomb, 2010). Ocean biogeochemistry was represented by the Biogeochemical Elemental  
210 Cycling (BEC) model (Moore et al., 2013; Lindsay et al., 2014).

211

212 Our analysis focuses on three depths: 50 m representing near-surface dynamics, the epipelagic  
213 zone at 200 m, and the mesopelagic zone at 500 m.  $pO_2$  was calculated using the Garcia and  
214 Gordon. (1992) solubility formulation. For convenience, we use the period 1920–1965 to define  
215 a minimally-perturbed natural state, as this period is prior to the development of substantial  
216 anthropogenic trends in ocean oxygen and temperature (Long et al., 2016). We also examine  
217 distributions over the last three decades of the 21st century (2070–2099) to evaluate the projected  
218 climate-change signal under RCP8.5. We use the mean across all 34 ensemble members to  
219 quantify the deterministic, “forced” response of the climate system to anthropogenic influence  
220 (Deser et al., 2012). The ensemble spread is thus indicative of the amplitude of variations  
221 attributable to natural variability.

222



223  
224 **Figure 3.** Mean-state comparison with observations. The climatological mean of (top rows) temperature (°C),  
225 (middle rows)  $pO_2$  (kPa), and the (bottom rows) metabolic index for active metabolism ( $\Phi$ ) for the median ecotype  
226 ( $E_o = 0.34$ ,  $A_c = 7.4$ ); three depths are shown (left) 50 m, (center) 200 m, and (right) 500. Top panels show the  
227 WOA13 dataset and the bottom panels show CESM1-LE.

228

229 We compared the CESM1-LE (1920 - 1965) with the World Ocean Atlas, version 2013  
230 (WOA2013) dataset (Garcia et al., 2014), an observationally-based, gridded climatology (Figure  
231 3a-i). CESM1-LE generally provides a reasonable representation of  $pO_2$  and temperature



232 distributions at the selected depths (Figure 3); however, there are important biases to  
233 acknowledge in the context of interpreting the results. Temperature magnitudes are generally  
234 well simulated in the CESM1-LE, showing a root mean square error (RMSE)  $< 1.3$  °C, and  
235 pattern correlation coefficient (PCC)  $> 0.98$  in all three selected depths (50 m, 200 m, and 500)  
236 (Table 1). Temperature magnitudes are slightly underestimated at 50 m and 200 m (mean bias of  
237  $< 0.3$ °C), and overestimated by  $0.41$  °C at 500 m. Note that since our comparison uses CESM1-  
238 LE data from 1920-1965, some discrepancy in temperature might be expected from the signal of  
239 climate warming present in the WOA observations.  $pO_2$  is also reasonably well captured by the  
240 CESM1-LE (PCC  $< 0.95$ ), but magnitudes are slightly underestimated at depth, showing a mean  
241 bias of  $-1.63$  kPa and  $-2.1$  kPa at 200 m and 500 m with respect to WOA13 (Table 1). Regions of  
242 low  $pO_2$  waters are too extensive in CESM1-LE (Figure 3n-o) and there is a slight degradation of  
243 skill with depth for  $pO_2$  fields (Table 1). The underestimation of  $pO_2$  leads to a slight  
244 underestimation of  $\Phi'$  with respect to WOA13 (Figure 3 p-r); however,  $\Phi'$  computed from the  
245 model fields demonstrates that the dominant spatial patterns are well captured by the CESM1-LE  
246 despite magnitudes that are slightly too low (i.e., Figure 1, c, l). These differences ultimately  
247 matter most near the hypoxic zones and at the boundaries of habitable zones like the OMZs.  
248  
249



250 **Table 1.** Summary statistics for the comparison of CESM1-LE with the World Ocean Atlas dataset (Garcia et al.,  
251 2014). The columns include the mean bias, pattern correlation coefficient (PCC), and root mean square error  
252 (RMSE) at 50 m, 200 m, and 500 m.

	<b>Mean bias</b>	<b>R</b>	<b>RMSE</b>
	<b>Temperature [°C]</b>		
<b>50 m</b>	-0.17	0.99	1.22
<b>200 m</b>	-0.25	0.99	1.22
<b>500 m</b>	0.10	0.98	0.63
	<b>pO<sub>2</sub> [kPa]</b>		
<b>50 m</b>	0.05	0.99	1.91
<b>200 m</b>	-1.17	0.96	5.96
<b>500 m</b>	-1.46	0.95	6.28
	<b>Metabolic index</b>		
<b>50 m</b>	0.01	0.99	0.02
<b>200 m</b>	-0.09	0.97	0.05
<b>500 m</b>	-0.15	0.96	0.08

253

254



255        **2. Results**

256

257                **3.1 Joint temperature- $pO_2$  natural variability and forced trends**

258

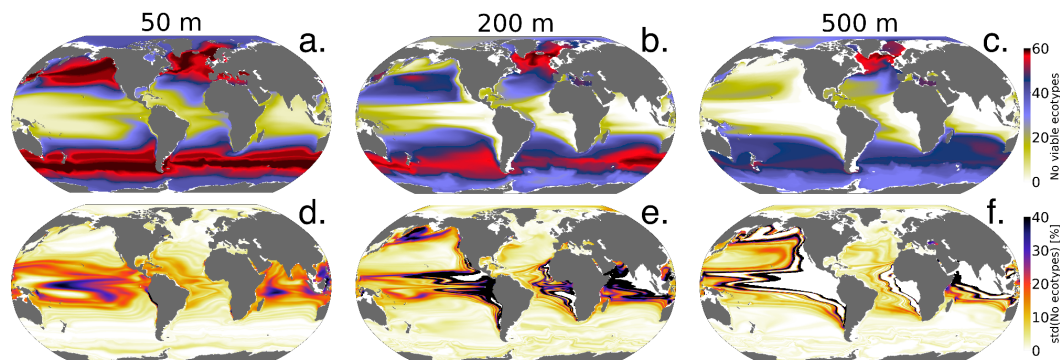
259    The spatial distribution of the number of viable ecotypes is shown in Figure 4 for the  
260    “unperturbed” climate (1920-1965). Our intention here is not to quantify the actual  
261    biogeographic range of organisms in the environment, but rather to illustrate the ocean’s ability  
262    to support respiration by marine ectotherms given the metabolic capacities afforded within the  
263    trait space of extant organisms. High latitude environments do not impose strong aerobic  
264    constraints (cold intolerance notwithstanding), thus over much of the Southern Ocean, North  
265    Atlantic, and Arctic Ocean almost all 61 ecotypes can sustain respiration. The tropical oceans  
266    impose the strongest aerobic constraints, restricting the viability of ecotypes that do not have  
267    high-hypoxia tolerance ( $A_o$ ). For example, less than 25 ecotypes are viable over much of the  
268    tropical surface ocean (Figure 4a); low concentrations of oxygen at depth impose even stronger  
269    constraints, and no ecotypes are viable in the core of OMZs (Figure 4b, c). The spatial patterns of  
270    the number of viable ecotypes is tightly controlled by temperature at the surface, since  $pO_2$  is  
271    mostly near saturated levels; at depth, however,  $pO_2$  is the dominant driver of geographic  
272    patterns in ecotype viability (Figures 2-4). Temperature generally decreases with depth, reducing  
273    the metabolic oxygen demand. However, since  $pO_2$  also decreases with depth and displays  
274    greater lateral heterogeneity,  $pO_2$  emerges as the dominant constraint of spatial structure in  
275    ecotype viability at depth.

276

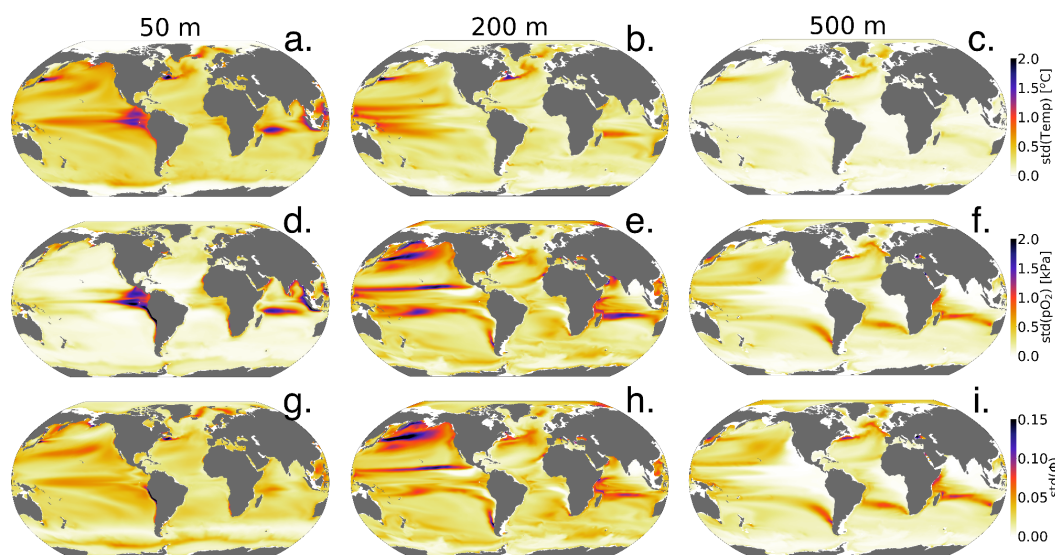
277    The standard deviation of annual anomalies using all CESM1-LE ensemble members provides  
278    insight into the amplitude of natural variability (Figure. 5). Temperature and  $pO_2$  show similar  
279    patterns of natural variability in the upper ocean, both showing particularly large variance in the  
280    western tropical Pacific and Indian Ocean (Figure 5 a, d). Spatial variation in the magnitude of  
281    temperature variability generally decreases with depth, but  $pO_2$  displays even relatively larger  
282    variability at depth with respect to the surface in some regions (Figure 5 a–f). The joint  $pO_2$ -  
283    temperature variability manifests in variations of  $\Phi'$  (Figure 5g-i). Natural variability in  $\Phi'$   
284    computed for the median ecotype shows spatial patterns similar to temperature in the upper-



285 surface ocean (50 m), but is more similar to  $pO_2$  at depth. Thus, variations in  $\Phi'$  tend to be  
 286 temperature-dominated near the surface, but are more strongly controlled by  $pO_2$  variability at  
 287 depth.  $\Phi'$  also shows the most extensive natural variability at 200 m consistent with the  
 288 variability of  $pO_2$ . The number of viable species shows more dramatic fluctuations than  
 289 variations in the median ecotype  $\Phi'$ ; variations in the number of viable ecotypes exceed 30% on  
 290 annual timescales in the tropical upper ocean and near OMZ boundaries in the water column  
 291 (Figure 4 c–d). This reflects the fact that interannual variability can preclude habitability for  
 292 some regions of the  $A_c$ - $E_o$  trait space, but these variations do not necessarily impact viability for  
 293 the median ecotype (Figure 1). In the tropical surface ocean, high temperatures ( $>25^\circ\text{C}$ ), and  
 294 saturated surface ( $pO_2 > 20$  kPa) require high hypoxia tolerance ( $A_c$ ), but permit a range of  
 295  $E_o$  values (Figure 1b, 2a-b). Ecotypes with larger temperature sensitivity (high  $E_o$ ) are  
 296 particularly responsive to variations in temperature.  
 297



298  
 299 **Figure 4.** Metabolic constraints on trait-space viability. Top row: the number of ecotypes from the physiological  
 300 trait database that are viable (total = 61) in the CESM1-LE over the period 1920–1965. Bottom row: the standard  
 301 deviation (expressed as a percent of the mean) in the number of viable ecotypes, reflecting fluctuations driven by  
 302 natural variability.  
 303



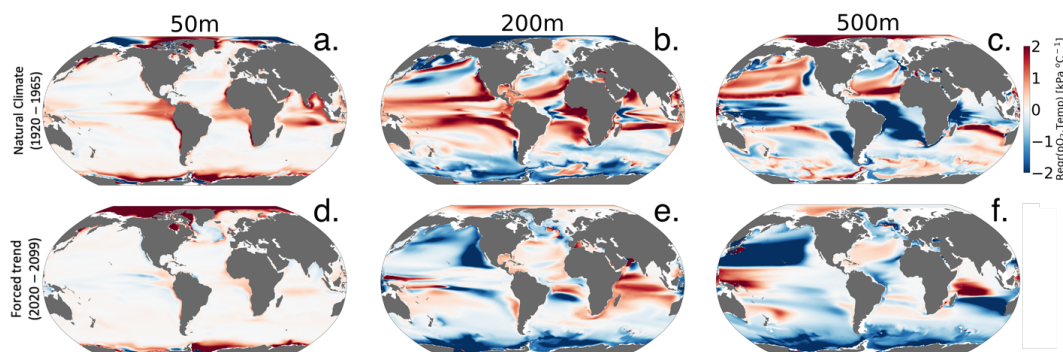
304  
305 **Figure 5.** The amplitude of natural variability in the ocean’s metabolic state. The panels show the standard deviation  
306 of annual-mean anomalies of all ensemble members over the period 1920–1965 for (top row) temperature (°C),  
307 (middle row)  $pO_2$  (kPa), and (bottom row) the metabolic index (unitless) of the median ecotype ( $E_o = 0.34$ ,  $A_c = 7.4$ ).  
308  
309 CESM1-LE simulates nearly homogeneous warming between 1920–1965 and 2070–2099 in the  
310 surface ocean (50 m) under RCP8.5, with an exception of the so-called North Atlantic warming  
311 hole (Figure 6a). Both modelling and observational studies have linked the North Atlantic  
312 warming hole to the slowing of the Atlantic overturning circulation with climate change (Keil et  
313 al., 2020). The magnitude of ocean warming generally diminishes with depth except in the North  
314 Atlantic, where, despite reductions, the overturning circulation effectively propagates  
315 anthropogenic heat anomalies into the ocean interior.  $pO_2$  shows heterogeneous changes between  
316 1920–1965 and 2070–2099 (Figure 6 d-f). In the upper ocean,  $pO_2$  changes are generally small (<  
317 1 kPa) because the near-surface is kept close to saturation via photosynthetic oxygen production  
318 and air-sea equilibration. At depth, however,  $pO_2$  shows long-term changes linked to  
319 accumulated effects of respiration and changes in circulation (Ito et al., 2017). At 200 m for  
320 example, the Pacific Ocean displays a basin-wide mean reduction in  $pO_2$  of 2 kPa (~30%), while  
321 the Atlantic and Indian basins gain about >2 kPa (~ 10 - 35%) by the end of the century. The  
322 largest long-term  $pO_2$  loss (>3 kPa) occurs in the North Pacific while the largest  $pO_2$  gain (~2







347 associated with adiabatic vertical displacement of isopycnals, or “heave”, which has the effect of  
 348 translating background gradients in properties vertically in the water column. Upward movement  
 349 of a deep isopycnal surface would yield a negative temperature anomaly and a negative  $pO_2$   
 350 anomaly (positive correlation), as the deeper, colder waters have greater oxygen utilization  
 351 signatures associated with longer ventilation age. Negative correlations between  $pO_2$  and  
 352 temperature could manifest from ventilation processes, where enhanced subduction of surface  
 353 water yields anomalously cold water masses that are enriched in oxygen. The sign of these  
 354 epipelagic  $pO_2$ -T correlations shows some similarity to those associated with the externally  
 355 forced climate (Figure 6e), but the latter is characterized by a greater prevalence of  
 356 anticorrelation, most notably in the North Pacific ocean. At 500 m depth, the relationship  
 357 between temperature and  $pO_2$  in the natural climate is almost a mirror image of the epipelagic  
 358 (Figure 7c); the tropics generally display negative correlations, while polar regions show positive  
 359 correlations (Figure 7 e). The  $pO_2$ -T relationship in the forced trend at 500 m is dominated by  
 360 broad regions of deeply negative correlations, with the most pronounced effect again in the  
 361 North Pacific. The negative relationship is consistent with a ventilation signal, as buoyancy-  
 362 induced stratification from warming curtails the introduction of new oxygen into the ocean  
 363 interior. The predominantly negative  $pO_2$ -T relationship associated with the forced trend is  
 364 indicative of the compounding effects of climate change on metabolic state, increasing metabolic  
 365 demand while simultaneously reducing oxygen supply.



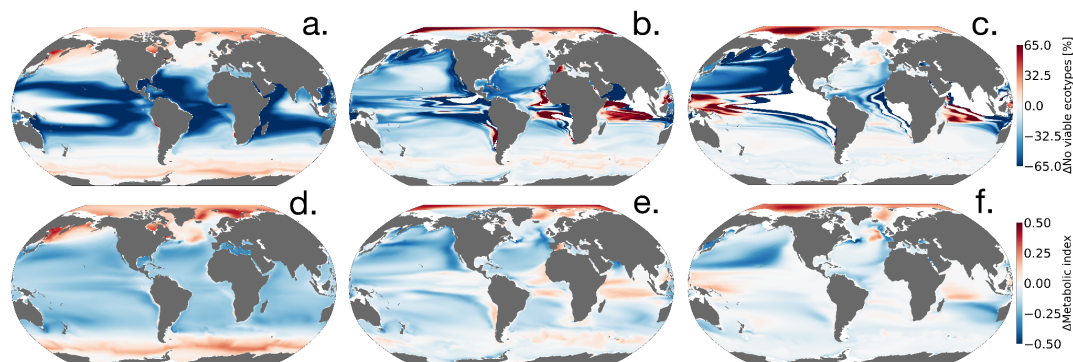
366  
 367 **Figure 7.** Regression of annual means  $pO_2$  versus temperature ( $\text{kPa } ^\circ\text{C}^{-1}$ ) for (top row) interannual variability and  
 368 (bottom row) the forced trend (difference between 2020–2099 and 1920–1965). The columns show the regressions  
 369 computed at different depths, 50 m, 200 m, and 500 m, respectively.  
 370



### 371 3.2 Long-term habitat changes

372

373 Figure 8 shows the climate-driven changes in  $\Phi'$  for the median ecotype, as well as the impacts  
374 of climate change on the number of viable ecotypes. Notably, while  $pO_2$  in the near-surface  
375 ocean is relatively insensitive to climate change (Figure 6d), there are reductions in  $\Phi'$  in the  
376 tropics (Figure 9d), owing to the direct impacts of warming. These changes are associated with  
377 deep reductions in the number of viable ecotypes in the tropics (Figure 8a). There are modest  
378 increases in  $\Phi'$  and ecotype viability at high-latitudes; metabolic state in these regions is affected  
379 by cold intolerance, thus warming broadens the viable region of trait space. Additionally,  
380 reductions in sea ice cause an increase in  $pO_2$ , as gas exchange becomes more effective at  
381 restoring equilibrium oxygen concentrations. The number of viable ecotypes shows more intense  
382 patterns than those in the median ecotype  $\Phi'$  in the upper ocean (Figure 8). This is partly because  
383 ecotypes predicted to lose viability in the tropical regions ( $\sim 50\%$ ) are at the extremes of the  $A_c$ -  
384  $E_o$  distribution (Figure 1) and not captured by the median ecotype  $\Phi'$ . Nevertheless, outside the  
385 tropical regions, the median ecotype gives a good indication of the anthropogenic impact to  
386 marine ectotherms. The projected habitat loss in the epipelagic-pelagic North Pacific ( $> 50\%$ )  
387 and habitat gain in the epipelagic-pelagic Southern Indian Ocean ( $\sim 40\%$ ) and pelagic western  
388 tropical regions ( $\sim 40\%$ ) are consistent with a decrease in the median ecotype  $\Phi'$ . Note that the  
389 most pronounced effects on habitat are associated with regions where climate change drives a  
390 strongly negative  $pO_2$ -temperature relationship (Figure 7).



391

392 **Figure 8.** Net change in the number of habitable ecotypes in percentage (top row). Net metabolic index change  
393 (2070 - 2099 vs. 1920 - 1965) for the median ecotype [ $E_o = 0.34$ ,  $A_c = 7.4$ ] (bottom row). At 50m (first column),  
394 200m (second column) and 500m (third column).



395

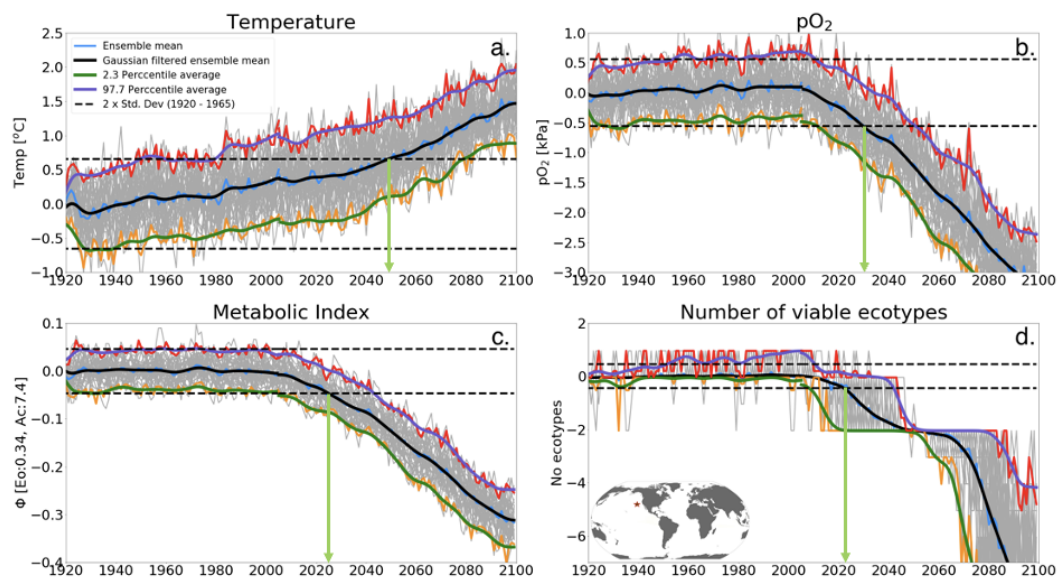
### 396 **3.3 Time of Emergence**

397

398 In this section, we examine the “time of emergence” (ToE, Hawkins and Sutton, 2012), the point  
399 when forced changes in  $pO_2$ , temperature and  $\Phi'$  can be distinguished from the background  
400 natural variability. We define ToE as the time when the magnitude of change in the ensemble  
401 mean of a particular variable exceeds two standard deviations of the natural climate (1920 -  
402 1965). This is illustrated in Figure 9 for a single grid point in the North Pacific at 200 m. At this  
403 location, the forced trend in temperature shows a monotonic increase, while  $pO_2$  shows a  
404 monotonic decrease; as a result,  $\Phi'$  for the median ecotype and the number of viable ecotypes  
405 decrease over time. The anti-correlation between  $pO_2$  and temperature exacerbates trends in  $\Phi'$ ,  
406 and hence the forced trend of the median ecotype  $\Phi'$  emerges from natural noise earlier than  
407 either  $pO_2$  or temperature do alone (Figure 10a-c). Note that although the ToE of ecotype  
408 viability change is directly derived from changes in  $\Phi'$ , it is binary counted; changes in ecotype  
409 viability are counted in whole numbers and this creates a step-function temporal-spatial variation  
410 (Figure 9d). Consequently, this step-function-like feature of ecotype viability creates  
411 discontinuities even in spatial patterns of ToE (Figure 10 j-l) as also shown in the natural  
412 variance in Figure 4 d-f.

413

414



415

416

**Figure 9.** Time of emergence (ToE) of the climate forcing signal for (a) temperature, (b) pO<sub>2</sub> (c) the metabolic index of the median ecotype [ $E_o = 0.34, A_c = 7.4$ ], and (d) the number of viable ecotypes for a single model grid in

417

the North Pacific at 200 m. ToE (green arrows) is defined as the time when the forced trend signal (ensemble

418

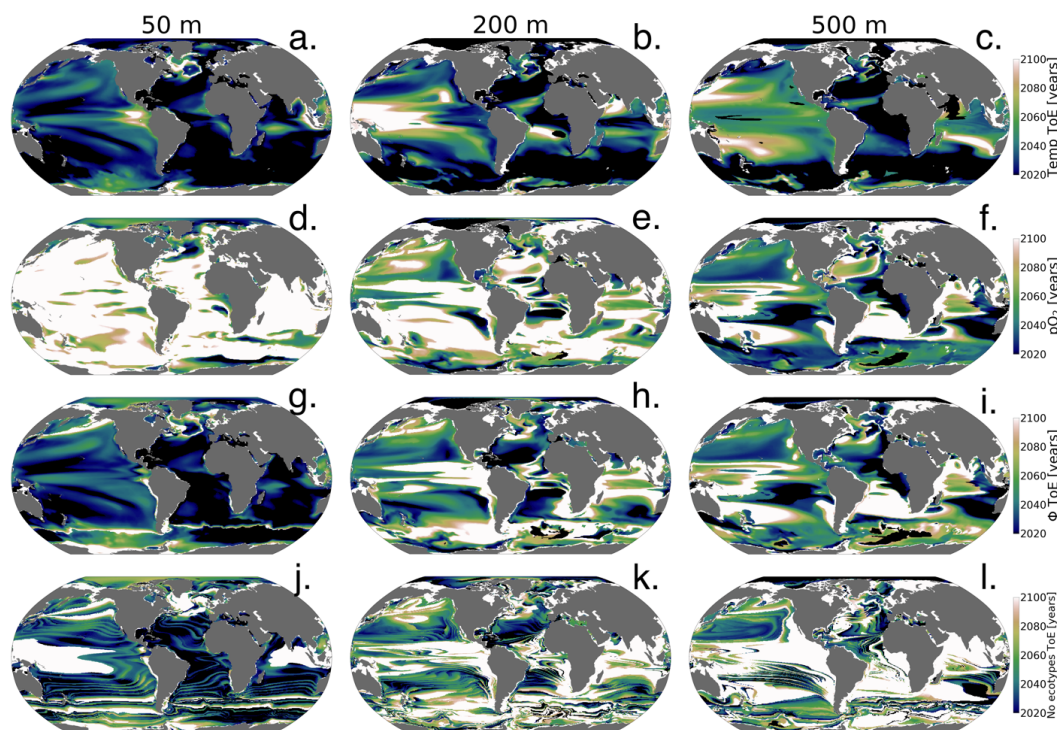
member time series) is above two standard deviations (black dotted line) of all ensemble members for the period

419

1920 - 1965.

420

421



422  
423 **Figure 10.** Time of emergence (ToE) of the climate forcing signal for temperature,  $pO_2$ ,  $\phi$ , and the number of  
424 viable ecotypes. ToE is defined as the time when the forced trend signal (ensemble member time series) is above  
425 two standard deviations of all ensemble members for the period 1920 - 1965.

426  
427 The ToE of  $pO_2$  and temperature are inverted with depth; temperature emerges earliest in the  
428 upper ocean while  $pO_2$  emerges earlier at depth and later or shows no emergence in the upper  
429 ocean (Figure 10 a-f). This feature is consistent with larger upper ocean temperatures long-term  
430 changes and greater  $pO_2$  changes at depth. Near-surface ocean temperature has mostly already  
431 emerged by 2020 and is predicted to have almost completely emerged by the late 2060s under  
432 RCP85 (Figure 10 a-c). The early emergence of temperature from natural noise also persists for  
433 regions of relatively low natural variance at depth, e.g., the Southern Ocean and Atlantic Basin  
434 Gyres. Regions of the largest natural variability (see Figure 5) like the subtropical-subpolar  
435 Pacific however do not emerge until close to the end of the century. For  $pO_2$ , anthropogenic  
436 changes in the upper ocean generally do not emerge from natural noise before the end of the  
437 century except for the Arctic Ocean and Eastern Antarctic. In the Arctic Ocean and Eastern



438 Antarctic  $pO_2$  gain is related to sea-melt emergence by the mid-2050s (Figure 10a). The median  
439 ecotype  $\Phi'$  ToE shows spatial patterns that are coherent with temperature ToE in the upper ocean  
440 with exception of polar regions. In contrast, they are consistent with  $pO_2$  ToE patterns at depth;  
441 this is consistent with net long-term  $\Phi'$  changes in Figure 9d. The emergence of the  
442 anthropogenic signal in ecotype viability closely resembles the median ecotype  $\Phi'$  spatial  
443 patterns but showing non-harmonious spatial patterns due to the step-function-like counting  
444 feature of viability changes. It shows that the predicted  $\sim 50\%$  ecotype viability loss in the  
445 tropics (Figure 6a) may already be distinguishable from natural variability by the mid-2030s. In  
446 the North Pacific, the predicted  $> 50\%$  ecotype viability loss in the epipelagic-pelagic regions is  
447 predicted to start emerging in the 2040s at 500 m and 2080s at 200 m (Figure 10 k-l).

448

#### 449 **4. Discussion**

450

451 The human-induced rapid warming of the planet has been shown to drive ocean deoxygenation  
452 (Ito et al., 2017; Schmidtko et al., 2017; Long et al., 2016). Higher metabolic oxygen demand at  
453 higher temperatures (Gillooly et al., 2001; Deutsch et al., 2015, 2022) raises concerns about the  
454 ability of marine ectotherms to support aerobic respiration in the future. This study set out to  
455 characterize the anticipated climate change signal in the ocean's metabolic state in the context of  
456 natural variability using the metabolic theory as a basis to examine the capacity of the  
457 environment to support ectothermic marine heterotrophs.

458

459 The spatial variation in  $pO_2$  and temperature in the unperturbed natural climate state set  
460 biogeographic boundaries based on ectotherms' physiological performance. The resilience of  
461 these ectotherms' biogeographic structure to natural variability and long-term climate warming is  
462 perturbed by the joint  $pO_2$ -temperature changes, effectively measured by the metabolic index  
463 ( $\Phi'$ ). An increase in the capacity of the organisms to support aerobic respiration increases  $\Phi'$ ; for  
464 example by ocean cooling or increase in oxygen supply contrary, warming and decrease in  
465 oxygen supply decrease  $\Phi'$ . There are exceptions in extremely low-temperature environments,  
466 where aerobic respiration is also limited by kinematic gas transfer into the organism in addition  
467 to environmental oxygen supply. Relative changes in  $pO_2$  and temperature in the natural



468 variability and forced trend, therefore, regulate ectotherms' resilience to environmental changes.

469 Under the RCP85 climate scenario, the ocean generally warms homogeneously but concurrent

470  $pO_2$  changes are heterogeneous and vary with depth. Thus, the characteristics of these  $pO_2$ -

471 temperature forced trend changes determine when the climate change impact on marine

472 ectotherms can be distinguishable from natural variability.

473

474 In the surface ocean,  $pO_2$  is generally abundant and relatively uniform, and thus spatial

475 temperature variations have a dominant constraint on the spatial variations of organismic

476 metabolic state. The warmest parts of the surface ocean, the tropical oceans, can only support

477 about 10-20 (~ 30%) of the 61 ecotypes while cooler regions in extratropics have nearly 100%

478 viability. Moreover, since warming anomalies propagate from the surface, the surface tropical

479 oceans also show the largest natural variance in temperature and ecotype viability in the surface

480 ocean. This is because extremely warm temperatures in the surface tropics ( $>25^\circ\text{C}$ ) are mainly

481 suited for organisms with high-temperature sensitivity ( $E_o$ ), which are relatively fewer, and

482 mostly close to their physiological limits (Storch et al., 2014). Large natural variability in these

483 warmest parts of the tropical surface ocean precludes the forced trend signal from emerging from

484 the natural variability in the ecotype viability by end of the century although the ocean warms the

485 largest in the surface. Nevertheless, the large warming trends in the surface ocean generally

486 emerge relatively early (the 2020s) from natural variability in both temperature and ecotype

487 viability in most regions. Minimal changes in surface  $pO_2$  in the forced trend affirm that surface

488 ocean marine ectotherms are mainly perturbed by temperature in the context of anthropogenic

489 changes. In polar regions, warming has a counterintuitive effect on marine ectotherms with

490 respect to most parts of the surface ocean. There, warming helps organisms escape extreme cold

491 intolerances by enhancing membrane kinematic gas transfer which enhances  $\Phi'$  and thus ecotype

492 richness in the future.

493

494 In the epipelagic and pelagic regions (200 m and 500 m), the temperature forced trend and

495 natural variability are smaller compared to the surface ocean, while concurrent  $pO_2$  changes are

496 larger than the surface ocean. Thus,  $pO_2$  and temperature play a more intricate role in

497 perturbing marine ectotherm habitats in the context of anthropogenic warming with respect to

498 the surface ocean, where temperature plays a dominant role. At depth, contrasting the regression



499 between  $pO_2$  and temperature in the natural climate, and forced trends provides an instructive  
500 framework to analyzing ectotherms' long-term changes. Regions showing distinct correlations  
501 between temperature and  $pO_2$  in the forced trends relative to the natural variability show a  
502 weakening metabolic resilience; loss of habitat and emerging relatively early from natural  
503 variability. For example, in the pelagic - epipelagic North Pacific, temperature- $pO_2$  regressions  
504 switched from a positive correlation in the unperturbed climate to a strong negative correlation in  
505 the forced trend. Consequently, the pelagic-epipelagic North Pacific is projected to lose nearly  
506 half of the present climate ecotype viability by the end of the century. This loss of pelagic -  
507 epipelagic North Pacific habitat is projected to emerge earliest at 500 m (the 2030s) where  
508 anthropogenic  $pO_2$  losses are larger than at 200 m. On the other hand, in the Arctic Ocean and  
509 some parts of the Southern Ocean, concomitant  $pO_2$ -temperature correlations in the forced trends  
510 result in the preservation of the marine habitat and even slight enhancements.

511

## 512 **5. Conclusions**

513

514 The joint temperature-oxygen metabolic framework in this study provides additional insight into  
515 the impact of climate change on marine ecosystems in comparison to the independent oxygen or  
516 temperature analysis. We here showed that while warming is the leading order driving  
517 mechanism of climate change, the direct effect of warming on marine ecosystems is mostly in  
518 the upper ocean. Climate change-related oxygen loss is a major driver of marine ecosystem stress  
519 in addition to warming at depth. Incorporating organismal physiological sensitivity to oxygen-  
520 temperature changes in the metabolic framework provides insight into how climate impacts the  
521 biogeographic structure of marine habitat. We find that underway forced trends perturbations in  
522  $pO_2$  and temperature will strongly exceed those associated with the natural system in many parts  
523 of the upper ocean, mostly pushing organisms in these environments closer to or beyond their  
524 physiological limits. Climate warming is expected to drive significant marine habitat loss in the  
525 surface tropical oceans and epipelagic - pelagic North Pacific Basin, while gaining marginal  
526 habitat viability in the surface Arctic Ocean and some parts of the Ocean Southern.

527

## 528 **6. Competing interests**





529 The contact author has declared that none of the authors has any competing interests

530

## 531 **7. Acknowledgments**

532

533 PM, ML, CD and TI were funded by the National Science Foundation (NSF) grant agreement  
534 No. 1737158. PM and YSF were also funded by the European Union's Horizon 2020 research  
535 and innovation programme under grant agreement No. 820989 (COMFORT). ). We also would  
536 like to acknowledge the data access and computing support provided by the NCAR Cheyenne  
537 HPC.

## 538 **8. Author contribution**

539

540 PM and ML designed the study approach. PM developed the model code and analysis with  
541 feedback from ML, CD and TI. PM prepared the manuscript with contributions from all co-  
542 authors.

543

## 544 **9. Data access**

545

546 The CESM1 large ensemble data used in this study can be accessed in this location:

547 <https://www.cesm.ucar.edu/community-projects/lens/data-sets>

548

## 549 **10. References**

550

551 Breitburg, D., Levin, L. A., Oschlies, A., Grégoire, M., Chavez, F. P., Conley, D. J., Garçon, V.,  
552 Gilbert, D., Gutiérrez, D., Isensee, K., Jacinto, G. S., Limburg, K. E., Montes, I., Naqvi, S. W.  
553 A., Pitcher, G. C., Rabalais, N. N., Roman, M. R., Rose, K. A., Seibel, B. A., Telszewski, M.,  
554 Yasuhara, M., and Zhang, J.: Declining oxygen in the global ocean and coastal waters,  
555 <https://doi.org/10.1126/science.aam7240>, 5 January 2018.

556

557 Deser, C., Phillips, A., Bourdette, V., and Teng, H.: Uncertainty in climate change projections:  
558 The role of internal variability, *Clim Dyn*, 38, 527–546, [https://doi.org/10.1007/s00382-010-](https://doi.org/10.1007/s00382-010-0977-x)  
559 0977-x, 2012.

560



- 561 Deutsch, C., Ferrel, A., Seibel, B., Pörtner, H. O., and Huey, R. B.: Climate change tightens a  
562 metabolic constraint on marine habitats, *Science* (1979), 348, 1132–1135,  
563 <https://doi.org/10.1126/science.aaa1605>, 2015.  
564
- 565 Deutsch, C., Penn, J. L., and Seibel, B.: Metabolic trait diversity shapes marine biogeography,  
566 *Nature*, 585, 557–562, <https://doi.org/10.1038/s41586-020-2721-y>, 2020.  
567
- 568 Deutsch, C., Penn, J. L., Verberk, W. C. E. P., Inomura, K., Endress, M.-G., and Payne, J. L.:  
569 Impact of warming on aquatic body sizes explained by metabolic scaling from microbes to  
570 macrofauna, <https://doi.org/10.1073/pnas>, 2022.  
571
- 572 Garcia, H. E. and Gordon, L. I.: Oxygen solubility in seawater: Better fitting equations,  
573 <https://doi.org/10.4319/lo.1992.37.6.1307>, 1992.  
574
- 575 Garcia, H. E. , Boyer, T. P. , Locarnini, R. A. , Antonov, J. I. , Mishonov, A. V. , Baranova, O.  
576 K. , Melissa, M. Z. , Reagan, J. R. , and Johnson, D. R. : WORLD OCEAN ATLAS 2013  
577 Volume 3: Dissolved Oxygen, Apparent Oxygen Utilization, and Oxygen Saturation, 75,  
578 <https://doi.org/10.7289/V5XG9P2W>, 2014.  
579
- 580 Gillooly, J., Brown, J., West, G., Savage, V., Charnov, E., Gillooly, J. F., Brown, J. H., West, G.  
581 B., Savage, V. M., and Charnov, E. L.: Effects of size and temperature on metabolic rate  
582 Recommended Citation, 2001.  
583
- 584 Hawkins, E. and Sutton, R.: Time of emergence of climate signals, *Geophys Res Lett*, 39,  
585 <https://doi.org/10.1029/2011GL050087>, 2012.  
586
- 587 Hoegh-Guldberg, O. and Bruno, J. F.: The Impact of Climate Change on the World’s Marine  
588 Ecosystems, *New Series*, 328, 1523–1528, <https://doi.org/10.1126/science.1185779>, 2010.  
589 Howard, E. M., Penn, J. L., Frenzel, H., Seibel, B. A., Bianchi, D., Renault, L., Kessouri, F.,  
590 Sutula, M. A., Mcwilliams, J. C., and Deutsch, C.: Climate-driven aerobic habitat loss in the  
591 California Current System, 2020.  
592
- 593 Hunke, E. C. and Lipscomb, W. H.: CICE: the Los Alamos Sea Ice Model Documentation and  
594 Software User’s Manual Version 4.1 LA-CC-06-012, 2010.  
595
- 596 Hurrell, J. W., Holland, M. M., Gent, P. R., Ghan, S., Kay, J. E., Kushner, P. J., Lamarque, J. F.,  
597 Large, W. G., Lawrence, D., Lindsay, K., Lipscomb, W. H., Long, M. C., Mahowald, N., Marsh,  
598 D. R., Neale, R. B., Rasch, P., Vavrus, S., Vertenstein, M., Bader, D., Collins, W. D., Hack, J. J.,  
599 Kiehl, J., and Marshall, S.: The community earth system model: A framework for collaborative



- 600 research, *Bull Am Meteorol Soc*, 94, 1339–1360, <https://doi.org/10.1175/BAMS-D-12-00121.1>,  
601 2013.  
602
- 603 Ito, T. and Deutsch, C.: A conceptual model for the temporal spectrum of oceanic oxygen  
604 variability, *Geophys Res Lett*, 37, <https://doi.org/10.1029/2009GL041595>, 2010.  
605
- 606 Ito, T., Minobe, S., Long, M. C., and Deutsch, C.: Upper ocean O<sub>2</sub> trends: 1958–2015, *Geophys*  
607 *Res Lett*, 44, 4214–4223, <https://doi.org/10.1002/2017GL073613>, 2017.  
608
- 609 Kay, J. E., Deser, C., Phillips, A., Mai, A., Hannay, C., Strand, G., Arblaster, J. M., Bates, S. C.,  
610 Danabasoglu, G., Edwards, J., Holland, M., Kushner, P., Lamarque, J. F., Lawrence, D.,  
611 Lindsay, K., Middleton, A., Munoz, E., Neale, R., Oleson, K., Polvani, L., and Vertenstein, M.:  
612 The community earth system model (CESM) large ensemble project : A community resource for  
613 studying climate change in the presence of internal climate variability, *Bull Am Meteorol Soc*,  
614 96, 1333–1349, <https://doi.org/10.1175/BAMS-D-13-00255.1>, 2015.  
615
- 616 Keeling, R. F., Körtzinger, A., and Gruber, N.: Ocean deoxygenation in a warming world, *Ann*  
617 *Rev Mar Sci*, 2, 199–229, <https://doi.org/10.1146/annurev.marine.010908.163855>, 2010.  
618 Keil, P., Mauritsen, T., Jungclaus, J., Hedemann, C., Olonscheck, D., and Ghosh, R.: Multiple  
619 drivers of the North Atlantic warming hole, *Nat Clim Chang*, 10, 667–671,  
620 <https://doi.org/10.1038/s41558-020-0819-8>, 2020.  
621
- 622 Lindsay, K., Bonan, G. B., Doney, S. C., Hoffman, F. M., Lawrence, D. M., Long, M. C.,  
623 Mahowald, N. M., Moore, J. K., Randerson, J. T., and Thornton, P. E.: Preindustrial-control and  
624 twentieth-century carbon cycle experiments with the Earth system model CESM1(BGC), *J Clim*,  
625 27, 8981–9005, <https://doi.org/10.1175/JCLI-D-12-00565.1>, 2014.  
626
- 627 Long, M. C., Deutsch, C., and Ito, T.: Finding forced trends in oceanic oxygen, *Global*  
628 *Biogeochem Cycles*, 30, 381–397, <https://doi.org/10.1002/2015GB005310>, 2016.  
629
- 630 Moore, J. K., Lindsay, K., Doney, S. C., Long, M. C., and Misumi, K.: Marine ecosystem  
631 dynamics and biogeochemical cycling in the community earth system model [CESM1(BGC)]:  
632 Comparison of the 1990s with the 2090s under the RCP4.5 and RCP8.5 scenarios, *J Clim*, 26,  
633 9291–9312, <https://doi.org/10.1175/JCLI-D-12-00566.1>, 2013.  
634
- 635 Oschlies, A., Brandt, P., Stramma, L., and Schmidtko, S.: Drivers and mechanisms of ocean  
636 deoxygenation, <https://doi.org/10.1038/s41561-018-0152-2>, 1 July 2018.  
637



- 638 Penn, J. L., Deutsch, C., Payne, J. L., and Sperling, E. A.: Temperature-dependent hypoxia  
639 explains biogeography and severity of end-Permian marine mass extinction, *Science* (1979), 362,  
640 <https://doi.org/10.1126/science.aat1327>, 2018.  
641
- 642 Piiper, J., Dejours', P., Haab, P., and Rahn, H.: CONCEPTS AND BASIC QUANTITIES IN  
643 GAS EXCHANGE PHYSIOLOGY, *Respiration Physiology*, 292–304 pp., 1971.
- 644 Portner, H. O.: Climate variations and the physiological basis of temperature dependent  
645 biogeography: systemic to molecular hierarchy of thermal tolerance in animals, *Comparative*  
646 *Biochemistry and Physiology Part A*, 739–761 pp., 2002.  
647
- 648 Pozo Buil, M. and Di Lorenzo, E.: Decadal dynamics and predictability of oxygen and  
649 subsurface tracers in the California Current System, *Geophys Res Lett*, 44, 4204–4213,  
650 <https://doi.org/10.1002/2017GL072931>, 2017.  
651
- 652 Rodgers, K. B., Lin, J., and Frölicher, T. L.: Emergence of multiple ocean ecosystem drivers in a  
653 large ensemble suite with an Earth system model, *Biogeosciences*, 12, 3301–3320,  
654 <https://doi.org/10.5194/bg-12-3301-2015>, 2015.  
655
- 656 Rosewarne, P. J., Wilson, J. M., and Svendsen, J. C.: Measuring maximum and standard  
657 metabolic rates using intermittent-flow respirometry: A student laboratory investigation of  
658 aerobic metabolic scope and environmental hypoxia in aquatic breathers, *J Fish Biol*, 88, 265–  
659 283, <https://doi.org/10.1111/jfb.12795>, 2016.  
660
- 661 Schlunegger, S., Rodgers, K. B., Sarmiento, J. L., Frölicher, T. L., Dunne, J. P., Ishii, M., and  
662 Slater, R.: Emergence of anthropogenic signals in the ocean carbon cycle, *Nat Clim Chang*, 9,  
663 719–725, <https://doi.org/10.1038/s41558-019-0553-2>, 2019.  
664
- 665 Schmidtko, S., Stramma, L., and Visbeck, M.: Decline in global oceanic oxygen content during  
666 the past five decades, *Nature*, 542, 335–339, <https://doi.org/10.1038/nature21399>, 2017.
- 667 Smith, R., Jones, P., Briegleb, B., Bryan, F., Danabasoglu, G., Dennis, J., Dukowicz, J., Eden,  
668 C., Fox-Kemper, B., Gent, P., Hecht, M., Jayne, S., Jochum, M., Large, W., Lindsay, K.,  
669 Maltrud, M., Norton, N., Peacock, S., Vertenstein, M., and Yeager, S.: The Parallel Ocean  
670 Program (POP) Reference Manual Ocean Component of the Community Climate System Model  
671 (CCSM) and Community Earth System Model (CESM) 1, 2010.  
672
- 673 Storch, D., Menzel, L., Frickenhaus, S., and Pörtner, H. O.: Climate sensitivity across marine  
674 domains of life: Limits to evolutionary adaptation shape species interactions, *Glob Chang Biol*,  
675 20, 3059–3067, <https://doi.org/10.1111/gcb.12645>, 2014.  
676



677 Tiano, L., Garcia-Robledo, E., Dalsgaard, T., Devol, A. H., Ward, B. B., Ulloa, O., Canfield, D.  
678 E., and Peter Revsbech, N.: Oxygen distribution and aerobic respiration in the north and south  
679 eastern tropical Pacific oxygen minimum zones, *Deep Sea Res 1 Oceanogr Res Pap*, 94, 173–  
680 183, <https://doi.org/10.1016/j.dsr.2014.10.001>, 2014.  
681  
682 Vaquer-Sunyer, R. and Duarte, C. M.: Thresholds of hypoxia for marine biodiversity, 2008.  
683  
684