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Unexpected decline of the ocean carbon sink under record-high sea surface temperatures in 2023

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17 Abstract

In 2023, sea-surface temperatures (SST) reached record highs. Based on historical responses, this SST
 anomaly would suggest an increased oceanic CO₂ uptake (-0.11±0.03 PgC yr⁻¹). In contrast, our

20 observation-based estimates reveal that the global non-polar ocean absorbed about 10% less carbon than

21 expected (+0.16 \pm 0.28 PgC yr⁻¹). This weakening occurred despite reduced CO₂ outgassing in the tropics

22 associated with El Niño. Hence, the decline in CO_2 uptake in 2023 materialised entirely in the 23 extratropics, driven primarily by elevated SSTs in the Northern Hemisphere. Two ocean

24 biogeochemical models demonstrate that in the subtropical North Atlantic the thermally-induced

reduction in CO₂ uptake was strongly mitigated by the depletion of dissolved inorganic carbon in the

surface mixed layer. Such negative feedbacks cause an overall muted response of the ocean carbon sink

- 27 to the record high SSTs, but this resilience may not persist under long-term warming or more severe SST
- 28 extremes.

Main 29

In 2023, sea surface temperatures (SST) across most of the ocean reached record highs^{1–3}. Global 30

31 warming is a main driver of this anomalous state of the surface ocean, but even when accounting for the 32 linear trend in SST over the past 34 years, the annual mean anomaly of +0.21±0.02 °C was the largest

33

ever observed between 50°S and 65°N (Fig. 1a). Apart from global warming, a strong El Niño was an

- 34 important contributor to this unprecedented SST anomaly¹. The spatial pattern of the SST anomalies 35 represented in many parts the typical response to this phenomenon (Fig. 1b), but unusually high
- 36 temperatures in the North Atlantic made 2023 distinct from previous years⁴.
- 37 Here, we aim to establish how these extreme SSTs impacted the ocean's uptake of CO_2 from the
- 38 atmosphere. This understanding is critical given that the ocean currently removes about a quarter of the annual anthropogenic CO₂ emissions from the atmosphere^{5,6}, but inevitable further global warming⁷ and 39
- the increasing occurrence of anomalously high SSTs⁸⁻¹⁰ might affect the functioning of this sink. 40

It is well established that warming reduces the solubility of CO₂ in seawater, favouring an increased 41

outgassing of CO₂ to the atmosphere¹¹. Under isochemical conditions, that is when the dissolved 42

43 inorganic carbon (DIC) concentration and alkalinity (TA) remain constant, each 1°C rise in temperature 44 increases the fugacity of CO_2 (fCO₂) by ~4%¹². Thus, in the absence of any compensating mechanism,

the 2023 SST anomaly of +0.2°C would have raised fCO₂ by 4 µatm, largely eliminating the mean sea-45

air fCO₂ gradient (Δ fCO₂) over the non-polar global ocean¹³ and ceasing the uptake of CO₂ from the 46

47 atmosphere.

48 However, non-thermal processes — such as changes in ocean circulation, mixing, and biogeochemical

- 49 processes — can compensate for these thermal effects by modifying DIC and TA concentrations¹⁴.
- 50 Thermal and non-thermal drivers are often in a delicate balance, which is well documented for the seasonal cycle of surface ocean $fCO_2^{12,15,16}$. In some cases, non-thermal processes even overcompensate 51
- 52 the direct temperature effect. This was observed during previous El Niño years, when the oceanic uptake
- 53 of CO_2 became unusually strong (Fig. 1a) despite anomalously high global SSTs. This strengthening of
- 54 the ocean carbon sink during El Niño resulted from the reduced outgassing of CO₂ in the eastern
- 55 equatorial Pacific by roughly 0.1 - 0.2 PgC yr⁻¹, due to reduced upwelling of cold and CO₂-rich waters¹⁷
- . In contrast, the seasonal cycle of fCO₂ in the subtropics is thermally controlled, so that exceptionally 56
- warm SSTs are associated with an anomalous outgassing of CO₂¹⁸. Hence, the overall response of the 57
- 58 ocean carbon sink to unusual warming depends sensitively on the regional distribution of the SST
- 59 anomalies and the outcome of the "tug of war" between the thermal and non-thermal drivers of the
- 60 surface ocean carbon cycle.

To quantify the impact of 2023's record high SSTs on the oceanic CO₂ uptake, we employ four 61

- observation-based fCO₂ products^{19–22}. The fCO₂ products are machine-learning-based statistical models 62
- 63 trained on in-situ fCO₂ observations from version 2023 of the Surface Ocean CO₂ Atlas (SOCAT) that
- 64 provides observations through December 2022²³. We mapped surface ocean fCO₂ through the end of
- 2023, using observed predictor fields for this year and assuming that the diagnosed relationships 65
- between fCO_2 and the predictors remain robust. This assumption was validated through prediction skill 66 assessments based on truncated training data and comparisons to limited fCO₂ observations from 2023
- 67 68 (see SI)^{23–26}. From the mapped fCO₂ fields, we computed the sea-to-air CO₂ flux (FCO₂) as the product of
- 69 Δ fCO₂, the wind-dependent gas transfer velocity (k_w), and the solubility of CO₂ in seawater (K₀) (see
- 70 Methods). Additionally, we employed hindcast simulations from two global ocean biogeochemical
- models (GOBM)^{27,28}, physically forced with reanalysis data²⁹, to explore how physical and 71
- 72 biogeochemical anomalies in the ocean interior shape the surface anomalies.
- We focus our study on low- and mid-latitudes between ~50°S and ~65°N, referred to as the global non-73
- 74 polar ocean. The Arctic and parts of the Southern Ocean are excluded due to remaining challenges in
- 75 mapping interannual and seasonal flux anomalies in these regions, as highlighted by both previous
- 76 studies^{30,31} and our own analysis (Extended Data Figs. 3,4).

77 **Results**



78 Impact of record-high SSTs on the oceanic CO₂ uptake



88 The four observation-based fCO₂ products infer for 2023 an anomalous weakening of the ocean carbon 89 sink by +0.16±0.28 PgC yr⁻¹ integrated over our global non-polar analysis region (Fig. 1bd, Table 1). 90 Note that we report sea-to-air CO_2 fluxes, that is, an ocean CO_2 uptake is negative, and an outgassing is 91 positive. The 2023 flux anomaly corresponds to a roughly 10% reduction in CO₂ uptake. We computed 92 this anomaly relative to a baseline that accounts for the linear trend from 1990 through 2022 (Fig. 1b, 93 Methods), representing the increase in the uptake of CO_2 due to rising atmospheric CO_2 ⁶. The decline of 94 the ocean carbon uptake in 2023 is not unprecedented over the past 34 years, but it is unusual to occur in 95 a year with record high SST (Fig. 2a). Based on the (largely El Niño-driven) relationship between 96 annual mean anomalies in SST and ocean CO₂ uptake over the global non-polar ocean (Fig. 2a), one 97 would have expected for 2023 an anomalously strong uptake of -0.12 ± 0.03 PgC yr⁻¹ (Fig. 1b,2a). Thus, 98 something must have been different in 2023 compared to previous exceptionally warm years.



99 Fig. 2: Relationship between annual mean SST and sea-to-air CO₂ flux anomalies from 1990 to 2023 relative to the 100 respective long-term trends. Anomalies are shown for (a) the global, non-polar ocean, (b) the North Atlantic 101 subpolar seasonally stratified biome (NA-SPSS) (c) the North Atlantic subtropical permanently stratified biome 102 (NA-STPS), and (d) the Pacific Equatorial Upwelling East biome (PEQU-E). Symbols and error bars represent the 103 mean and standard deviation across the ensemble of four observation-based fCO₂ products, with the El Niño years 104 2023, 2015 and 1997 highlighted in warm colours. The grey lines and ribbons indicate linear regressions and 68% 105 confidence intervals across all annual mean anomalies from 1990 to 2022. A biome map and correlation plots for 106 other biomes are provided in the Extended Data Figs. 2 and 3, respectively.

- 107 The exceptional reduction of the ocean CO₂ uptake in 2023 was not caused by the eastern equatorial
- 108 Pacific (PEQU-E). This biome actually experienced a reduced outgassing of CO₂ by -0.06±0.02 PgC yr
- 109 ⁻¹, i.e., an anomalous uptake. This CO₂ flux anomaly is very consistent with the expectation based on the
- 110 SST anomaly in 2023 (Fig. 2d), and matches the response to previous El Niño events^{17,32,33}. Slightly
- reinforced by reduced CO₂ outgassing from the tropical Atlantic and Indian Ocean (Fig. 1d), the flux
- anomalies integrated over the global tropics amounted to -0.07±0.05 PgC yr⁻¹ (Table 1). This reduced
- 113 outgassing from the tropics alone matches the expected uptake anomaly for the global ocean (Fig. 1a).
- Hence, the decline of the CO_2 uptake must have occurred entirely in the extratropical ocean. Here, widespread positive SST anomalies, especially in the northern hemisphere (Fig. 1c), triggered an anomalous outgassing of CO_2 (Table 1).
- Among the extratropical regions, the North Atlantic experienced the highest, most persistent and extensive SST anomalies in 2023. In particular the subtropical permanently stratified biome (NA-STPS) faced an unprecedented annual mean SST anomaly of $+0.48\pm0.05$ °C, which caused a substantial decline of the CO₂ uptake by $+0.04\pm0.01$ Pg C yr⁻¹ (Table 1). This response fits the historic relationship between the CO₂ flux and SST anomalies very well (Fig. 2c), although the SST anomaly exceeds the previous record over the past 34 years by more than 50%. The two seasonally stratified biomes in the subtropical (NA-STSS) and subpolar (NA-SPSS) North Atlantic experienced similar CO₂ flux density
- anomalies as the NA-STPS, despite weaker annual mean SST anomalies (Table 1). However, the
- integrated CO₂ flux anomalies in the two seasonally stratified biomes are substantially lower than in

126 STPS, due to their smaller surface area. In total, the North Atlantic contributed a flux anomaly of $+0.07\pm0.01$ PgC yr⁻¹ to the weakening of the ocean carbon sink in 2023.

128 In 2023, the CO₂ uptake weakened also in all biomes of the North Pacific (Table 1). The strongest 129 anomalous outgassing occurred in the subtropical permanently stratified (NP-STPS) biome ($+0.05\pm0.06$ 130 PgC yr⁻¹). The two smaller biomes (NP-STSS and NP-SPSS) contributed less to the weakening of the 131 global ocean carbon sink, although the anomalies of their flux densities were about twice as strong as in 132 the STPS biome (Fig. 1d, Table 1). Together, the North Pacific contributed +0.11±0.12 Pg C yr⁻¹ to the 133 total flux anomaly for the year 2023.

134 The subtropical permanently stratified biomes of the southern hemisphere reveal a more mixed picture

135 with regard to the 2023 anomalies. Even though all three basins showed a positive SST anomaly, the CO

136 ₂ fluxes responded differently. While the STPS biomes in the Indian and Pacific also showed anomalous

137 outgassing in 2023, the Atlantic STPS actually experienced a minor anomalous uptake. Considering

138 also the STSS biome of the Southern Ocean, the extratropics of the southern hemisphere contributed in

total $+0.06\pm0.14$ Pg C yr⁻¹ to the global flux anomaly.

140 In the analysis above, we formally quantified the uncertainty of our estimates as the standard deviation 141 across four observation-based fCO₂ products. For most parts of the tropics and the North Atlantic, we 142 have high confidence in our estimates given that the fCO₂ products agree in sign and magnitude (Table 143 1, Extended Data Fig. 4). Due to the larger spread in the North Pacific and southern hemisphere biomes, 144 we assign only medium confidence to our estimates in these regions. In particular, the fCO₂-Residual 145 product suggests a substantially stronger weakening of the CO₂ sink in the North Pacific (Extended Data 146 Figs. 1,4), reflecting the general tendency of this product to generate a strong interannual variability 147 (Extended Data Fig. 1). The spread in the integrated flux anomalies from the North Pacific and southern 148 hemisphere propagates to our estimates for the non-polar global ocean (Table 1, Extended Data Fig. 4), which range from a minor strengthening of the sink by -0.05 Pg yr⁻¹ (CMEMS) to a strong weakening by 149 150 almost +0.5 Pg yr⁻¹ (fCO₂-Residual). As the fCO₂ products agree (mostly) in sign but not in magnitude, 151 we assign a medium confidence to our flux anomaly estimates for the non-polar global ocean. However, 152 when compared against the expectation of an anomalous strengthening of the sink (-0.12±0.03 PgC yr⁻¹ 153) instead of the zero flux anomaly at the baseline (Fig. 1b), the consistency of our estimates is higher. 154 Hence, we have high confidence in our conclusion that the ocean carbon sink in 2023 was indeed weaker

than one would expect based on the SST anomaly.

156 **Table 1**: Annual mean CO₂ fluxes (FCO₂) and sea surface temperatures (SST) in 2023 together with the respective 157 anomalies relative to a linear regression baseline. SST estimates are regional averages (°C), whereas FCO₂ 158 estimates are given both as integrals (PgC yr⁻¹) and regional averages (mol m⁻² yr⁻¹). All estimates represent the 159 mean \pm standard deviation over four observation-based fCO₂ products. Biomes abbreviations stand for: Subpolar 160 seasonally stratified (SPSS), subtropical seasonally stratified (STSS), subtropical permanently stratified (STPS), 161 the eastern and western equatorial Pacific (PEQU-E and PEQU-W), the tropical Atlantic and Indian Ocean 162 (AEQU and Equ. Ind.), South Atlantic (SA), South Pacific (SP), and Southern Ocean (SO), according to the biome 163 mach shown in the Extended Data Fig. 2. The anomaly estimates from this table are also presented in Extended 164 Data Fig. 4.

		SST (°C)		FCO₂ (mol m ⁻² yr ⁻¹)		FCO₂ (Pg yr ⁻¹)	
Region	Biome	Abs.	Anom.	Abs.	Anom.	Abs.	Anom.
North Atl.	SPSS	+8.73±0.08	+0.13±0.05	-2.61±0.17	+0.18±0.08	-0.29±0.02	+0.02±0.01
	STSS	+19.02±0.03	+0.25±0.03	-2.19±0.25	+0.12±0.1	-0.16±0.02	+0.01±0.01
	STPS	+26.02±0.05	+0.48±0.05	-0.22±0.04	+0.14±0.03	-0.06±0.01	+0.04±0.01
	Total	+20.63±0.02	+0.36±0.02	-1.12±0.08	+0.15±0.01	-0.51±0.03	+0.07±0.01
North Pac.	SPSS	+8.91±0.1	+0.37±0.03	-1.09±0.28	+0.2±0.3	-0.18±0.05	+0.03±0.05
	STSS	+19.47±0.05	+0.43±0.05	-2.2±0.19	+0.22±0.11	-0.21±0.02	+0.02±0.01
	STPS	+26±0.03	0±0.03	-0.35±0.15	+0.1±0.13	-0.18±0.08	+0.05±0.06
	Total	+21.5±0.01	+0.13±0.01	-0.74±0.17	+0.14±0.16	-0.58±0.14	+0.11±0.12
NH extratropics		+21.18±0.01	+0.21±0.01	-0.88±0.12	+0.14±0.1	-1.09±0.16	+0.17±0.12
Tropics	PEQU-E	+27.45±0.08	+1.21±0.08	+1.67±0.16	-0.31±0.13	+0.3±0.03	-0.06±0.02
	PEQU-W	+29.57±0.07	+0.04±0.03	+0.23±0.06	+0.05±0.03	+0.04±0.01	+0.01±0.01
	AEQU	+27.57±0.03	+0.23±0.08	+0.19±0.18	-0.11±0.13	+0.02±0.02	-0.01±0.01
	Equ. Ind.	+28.32±0.03	+0.01±0.06	+0.31±0.1	-0.02±0.08	+0.1±0.04	-0.01±0.03
	Total	+28.28±0.05	+0.33±0.05	+0.59±0.06	-0.09±0.06	+0.46±0.05	-0.07±0.05
ics	SA-STPS	+22.94±0.03	+0.11±0.04	-0.21±0.05	-0.06±0.06	-0.05±0.01	-0.01±0.01
SH extratrop	SP-STPS	+22.29±0.02	+0.11±0.03	-0.29±0.19	+0.06±0.14	-0.19±0.13	+0.04±0.09
	South. Ind.	+22.35±0.03	+0.11±0.09	-1.15±0.05	+0.13±0.11	-0.24±0.01	+0.03±0.02
	SO-STSS	+13.2±0.08	+0.28±0.07	-2.24±0.27	+0.02±0.14	-0.79±0.1	+0.01±0.05
	Total	+20.21±0.02	+0.15±0.04	-0.87±0.15	+0.04±0.1	-1.26±0.23	+0.06±0.14
Global non-polar		+22.36±0.02	+0.21±0.02	-0.55±0.11	+0.05±0.08	-1.89±0.38	+0.16±0.28
Polar	SO-SPSS	+3.88±0.11	+0.13±0.04	-0.6±0.59	-0.17±0.38	-0.22±0.22	-0.06±0.14
	SO-ICE	-0.95±0.07	+0.09±0.04	-0.57±0.29	-0.05±0.2	-0.13±0.05	-0.01±0.04
	Arctic	-0.14±0.41	-0.07±0.09	-1.5±0.58	+0.29±0.12	-0.14±0.02	+0.03±0.02
Global		+18.74±0.01	+0.19±0.01	-0.58±0.15	+0.03±0.1	-2.38±0.65	+0.12±0.43

165 Driver attribution and seasonality of CO₂ flux anomalies

166 The annual mean CO_2 flux anomaly over the non-polar global ocean in 2023 was primarily (>95%) 167 driven by anomalies in ΔfCO_2 (Fig. 3a, S5). The ΔfCO_2 anomaly triggered positive CO₂ flux anomalies 168 (i.e. a weaker sink) in all regions that experienced a positive SST anomaly, except in the eastern equatorial Pacific (Figs. 1a,3a). Anomalies in wind speed further modulated the annual mean CO₂ flux at 169 170 regional scales (Fig. 3b), for example the central equatorial Pacific or the North Pacific off Japan, but had a small net impact on globally integrated fluxes (Extended Data Fig. 5). The modulation of ΔfCO_2 -171 172 driven flux anomalies through wind anomalies (i.e. the cross term $\Delta fCO_2 \times k_w K_0$) is negligible on 173 regional scales, as well as for the globally integrated flux anomalies (Fig. 3c, Extended Data Fig. 5), and 174 will not be discussed further.

175 To quantify the causes for the 2023 anomalies in ΔfCO_2 , we further decomposed the main contributor,

176 which is the oceanic fCO₂ anomaly, into a thermal and a non-thermal component (see Methods). The

177 regions with the strongest SST anomalies reveal by definition the largest thermal component of the fCO

anomaly, i.e., the eastern equatorial Pacific, subpolar North Pacific, and subtropical North Atlantic
 (Fig. 3d). The same regions experienced also the strongest non-thermal component of the fCO₂ anomaly

180 (Fig. 3e). Hence, the resulting total fCO₂ anomaly remained comparably small at regional scales (Fig.

181 3f), and were slightly positive (+0.05±0.1 µatm) when averaged over the global non-polar ocean

182 (Extended Data Fig. 6).



Fig. 3: Attribution of 2023 annual mean flux anomalies and fCO₂ decomposition. (a-c) Attribution of flux anomalies to their primary drivers, that is, (a) the CO₂ fugacity gradient between ocean and atmosphere (Δ fCO₂), (b) the product of the gas transfer velocity and the solubility of CO₂ (k_wK₀), which is primarily controlled by wind speed, as well as (c) the cross product of both drivers (Δ fCO₂ × k_wK₀). (d-f) Decomposition of (f) the total fCO₂ anomalies into the (d) thermal and (e) non-thermal components. Stippling indicates regions where the ensemble standard deviation is higher than the absolute ensemble mean.

- 189 To dive further into the attribution of the 2023 CO₂ flux anomalies, we contrast their seasonal evolution
- 190 and drivers for three biomes. For this purpose, we corroborate the observation-based surface anomalies
- 191 with vertically resolved simulations from two GOBMs. Our GOBMs simulate a similar seasonal
- evolution of the SST and CO₂ flux anomalies in 2023 compared to the observation-based estimates (Fig.
- **193 4**, Extended Data Fig. 7) and are therefore considered reliable tools to interpret the underlying physical
- 194 and biogeochemical processes.
- 195 The seasonal evolution of SST anomalies in the eastern equatorial Pacific (PEQU-E) resembles that of
- 196 previous El Niño events. A positive SST anomaly of around 2°C established gradually over the first half
- 197 of 2023 and remained rather stable thereafter (Fig. 4a). This SST evolution was mirrored in a reduction
- 198 of the outgassing of CO₂ (i.e., a growing CO₂ sink) by up to 0.5 mol m⁻² yr⁻¹. The reduced outgassing was
- 199 primarily a consequence of negative ΔfCO_2 anomalies, but reinforced by anomalously low wind speeds.

200 The negative ΔfCO_2 anomalies in PEQU-E resulted from reduced upwelling of remineralized DIC. To 201 quantify this well-known non-thermal driver, we subtract the salinity-normalised alkalinity (sTA) 202 anomaly from the anomaly in salinity-normalised dissolved inorganic carbon (sDIC), and convert the 203 anomalous decline of sDIC-sTA (~- 20 µmol kg⁻¹; Fig. 4b) into an equivalent fCO₂ reduction of roughly 204 40 µatm (see Methods). This DIC-driven fCO₂ reduction is already muted by the reduced outgassing of 205 CO₂ that occurred over the course of 2023, but it still remains substantially stronger than the 206 temperature-driven increase of fCO_2 by ~30 µatm at the end of the year. Hence, the non-thermal 207 component clearly won the "tug of war" in PEQU-E.

208 In the subtropical permanently stratified biome of the North Atlantic (NA-STPS), the monthly SST 209 anomalies peaked in summer at around +1°C and remained well above the range of past anomalies for 210 the remainder of 2023. In contrast to PEQU-E, the resulting CO₂ flux anomaly in NA-STPS was 211 positive, that is, the CO_2 uptake in winter became weaker and the outgassing in summer became 212 stronger. This anomalous outgassing was primarily driven by thermally-driven ΔfCO_2 anomalies (Fig. 213 4a, Extended Data Figs. 5,6) and slightly enhanced by weak winds that further reduced the CO_2 uptake 214 from January to April (Extended Data Fig. 7). The mixed layer depth simulated by the GOBMs was 215 anomalously shallow in 2023, suggesting increased stratification and reduced mixing of remineralized 216 DIC into the surface layer (Extended Data Fig. 8). However, the simulated sDIC-sTA anomalies of -3 217 µmol kg⁻¹ in NA-STPS were much weaker than in PEQU-E. This sDIC-sTA anomaly was established 218 primarily in summer, when the surface CO₂ flux anomalies were also strongest (Fig. 4a). Integrated over 219 the summertime mixed layer depth of ~50m, the sDIC-sTA inventory decreased by roughly 0.15 mol m 220 $^{-2}$ over the course of 2023 (Fig. 4b), which is almost identical to the cumulative CO₂ flux anomaly. This 221 suggests that the surface CO_2 flux anomaly was the primary driver of the inventory anomaly in the 222 surface mixed layer. Hence, the reduced mixing of DIC into the surface layer — due the increased 223 stratification — was either negligibly small or nearly balanced by reduced primary production of 224 organic matter (Extended Data Fig. 8). Overall, the thermal driver won the "tug of war" and determined 225 the CO₂ flux anomalies during the onset of the SST anomaly in NA-STPS.

226 In the subpolar seasonally stratified biome of the North Atlantic (NA-SPSS), the monthly peak SST 227 anomalies were about as high as in NA-STPS, but more confined to the summer months and less 228 exceptional when compared to the variability in previous decades (Fig. 4a). Likewise, the CO_2 flux 229 anomalies were less exceptional. In contrast to NA-STPS, wind anomalies played a major role in 230 regulating the 2023 CO₂ fluxes in NA-SPSS (Extended Data Figs. 5,7). Strong winds until May 231 favoured the natural CO₂ sink, whereas weak winds throughout the rest of 2023 reduced the natural sink 232 and thereby reinforced the anomalous outgassing triggered by positive ΔfCO_2 anomalies. The 233 interpretation of the GOBM simulations in the NA-SPSS biome is rather complex. The simulated 234 summertime temperature increase was confined to a very shallow surface layer (10-20m) and triggered 235 a higher ΔfCO_2 anomaly than in the fCO₂ products, due to a weaker non-thermal fCO₂ anomaly in the 236 GOBMs (Extended Data Fig. 7). However, GOBMs are known to underestimate the non-thermal fCO₂ 237 component in the seasonal cycle¹⁶ in the NA-SPSS biome, which also hampers the interpretation of 238 biogeochemical drivers of the 2023 anomalies.



239 Fig. 4: Seasonal anomaly evolution for three key biomes. (a,b) Monthly mean surface anomalies of (a) the CO₂ 240 flux density, with triangles indicating the direction of the prevailing absolute CO₂ flux into (downwards) or out of 241 (upwards) the ocean, and (b) SST. Red and grey lines represent the fCO_2 product ensemble mean, whereas the 242 orange line represents the GOBM ensemble mean for 2023. (c,d) Vertical mean monthly anomalies of (c) salinity-243 normalised DIC (sDIC) anomalies, adjusted for the anomaly in salinity-normalised TA (sTA) to capture the sDIC 244 anomaly that is directly linked to fCO2 anomalies, and (d) temperature. The sDIC anomalies are presented as 245 changes since January 2023, to remove legacy anomalies from 2022 and achieve direct comparability to the 246 instantaneous surface flux anomalies. Black lines indicate the mean mixed layer depth (MLD). All vertical 247 anomalies represent the mean of the two GOBMs.

248 **Discussion**

In this study, we demonstrated that widespread and record-high SST led to a reduction in the ocean's carbon sink in 2023. The anticipated anomalous CO₂ uptake in the tropics was outweighed by an

increase in CO₂ outgassing from the non-polar extratropics (Fig. 5b). Among these regions, the
 anomalous CO₂ outgassing in the subtropical North Atlantic, driven by unusually strong and persistent
 SST anomalies (Fig. 5a), stood out as a strong and robust feature.

254 Interestingly, the depletion of DIC in NA-STPS by the end of 2023, reflected by the sDIC-sTA anomaly 255 of -3 μ mol kg⁻¹ (Fig. 4b), had an effect on surface ocean fCO₂ (~6 μ atm) that was similar in magnitude 256 but opposite in direction compared to that of the SST anomaly (+0.5°C). This suggests that the DIC-257 depletion fully compensated the thermal driver by the end of the year. The resulting decline of the 258 anomalous outgassing towards winter is indeed evident in our GOBM simulations, but less so in the 259 observation-based fCO₂ products (Fig. 4a), suggesting that the DIC-depletion feedback may not be well 260 captured by the underlying statistical models. If the DIC-depletion remained stable during the first half 261 of 2024, the anomalous outgassing in NA-STPS may have ceased, despite the continued elevated SSTs. 262 However, mixing during wintertime 2023/24 might have replenished the surface DIC pool. Hence, we 263 expect a neutral to to reduced sink strength in NA-STPS in the first half of 2024.

264 In contrast to the near-balance of the thermal and non-thermal drivers in NA-STPS, the non-thermal 265 DIC anomaly remained dominant in PEQU-E. By the end of 2023, the integrated anomaly of sDIC-sTA 266 over the top 100 m had reached -2 mol m^{-2} , which far exceeded the CO₂ flux anomaly of about -0.1 mol 267 m⁻² that occurred cumulatively over the course of the year (Fig. 4). Hence, the remaining negative DIC 268 anomaly could potentially drive further reduced outgassing, provided that these water masses stay in 269 contact with the atmosphere. However, as SST anomalies in PEQU-E decreased during the first half of 270 2024, CO_2 outgassing in this biome most likely returned to normal levels. Hence, we expect a rather 271 neutral sink strength in PEQU-E in the first half of 2023. If indeed the reduced outgassing in the tropics 272 fades out, while the extratropics remain anomalously warm, the global ocean carbon sink might have 273 continued to be weak in a weak state in early 2024. Whether these projected CO₂ flux anomalies indeed 274 materialise in 2024 remains to be confirmed through fCO₂ observations.

275 The limitation of thermally-induced CO_2 outgassing by DIC-depletion – that we observed for the 276 warming event in 2023 – is also crucial when considering the ocean carbon sink's long-term response to 277 global warming. Between 2000 and 2019, warming weakened the oceanic uptake of CO₂³⁴, but the 278 impact was much weaker than one would expect from the decreased solubility alone (i.e. the thermal 279 driver). This is because the temperature-induced outgassing of CO_2 and the reduced upwelling of DIC 280 (both non-thermal drivers) caused a negative feedback that compensated the initial perturbation. While 281 these negative feedbacks have now been demonstrated for historic warming SST extremes and historic 282 trends in the ocean carbon sink, it remains unclear if these stabilising mechanisms of the ocean carbon 283 sink remain effective under extreme SST events in the future, which are expected to become more frequent, intense, and longer lasting⁸, or under progressing global warming. Deviations from the 284 285 negative feedbacks observed in the past could for example occur if longer lasting SST extremes cause 286 stronger limitations of the CO₂ outgassing due to DIC-depletion, or if the efficiency of the biological 287 carbon pump becomes more strongly affected. To keep track of potential surprises in the ocean carbon 288 sink, the continued, revived, and extended observation of the ocean with high quality fCO_2 289 measurements remains indispensable, and it needs to be accompanied with an improved understanding 290 of the fCO₂ mapping skill at seasonal to interannual time scales and across ocean biomes 30,31 .



Fig. 5: (a) Bivariate map of 2023 annual mean anomalies in the CO₂ flux (FCO₂) and corresponding SST

- anomalies. (b) Annual mean fCO_2 anomalies (grey) and their decomposition into the thermal (red) and non-thermal
- (blue) component, for the eastern equatorial Pacific (PEQU-E) and subtropical permanently stratified biome of the
- 294 North Atlantic (NA-STPS). All estimates are based on the ensemble mean of four fCO₂ products.

295 Methods

296 Data base

297 This study relies on four fCO₂ products and two global ocean biogeochemical models, for which 298 technical details are provided in the Extended Data Tables 1 and 2, respectively. These data sources 299 constitute a subset of those used in the Global Carbon Budget⁵ and the second iteration of the REgional Carbon Cycle Assessment and Processes project (RECCAP2)^{35,36}. The observation-based sea surface 300 301 temperature fields used as predictor variables in the fCO₂ products are also used for our analysis of SST 302 trends and anomalies. The GOBMs simulations used here are equivalent to those considered as 303 "simulation A" in RECCAP2, that is, they are forced with (i) reanalysis data to represent the observed 304 climate variability over the hindcast period and (ii) historic atmospheric CO₂ observations to represent 305 anthropogenic emissions.

306 Anomaly determination against moving baseline

307 All anomalies determined in this study are expressed relative to a moving baseline, in order to remove

the long-term trends that are driven by the growth in atmospheric CO_2 or global warming. The baseline

estimate for any particular year (mostly 2023) was determined from a linear regression model that was

310 fitted to the historic observations from 1990 through 2022, and then predicted for the year of interest.
311 The underlying data are either annual or monthly mean values. The data for 2023 were excluded from

312 The underlying data are entire annual of monthly mean values. The data for 2023 were excluded nom 312 the regression to achieve a baseline estimate that is unbiased from the actual anomaly in 2023. Linear

313 regression models were applied to determine the baseline for any variable of interest, except for the

- atmospheric and surface ocean fCO_2 , for which a quadratic fit was used to better approximate the actual
- 315 growth rate. Finally, anomalies are calculated by subtracting the predicted baseline value from the
- 316 observed or estimated value.

317 Expected FCO₂ anomaly in 2023

318 To determine the expected FCO_2 anomaly in 2023 for the global non-polar ocean, we fitted linear

regression models of the integrated annual mean FCO₂ anomaly as a function of the annual mean SST anomaly to the hindcast estimates from 1990 through 2022 of our four fCO₂ products. The intercepts (in

321 PgC yr⁻¹) and the slopes (in PgC yr⁻¹ $^{\circ}$ C⁻¹) of these four regression models are: -1.3×10⁻¹⁴ and -0.61

322 (CMEMS), 1.9×10^{-15} and -0.74 (fCO₂-Residual), -4.6×10^{-15} and -0.39 (OceanSODAv2), and 1.8×10^{-16}

323 and -0.43 (SOM-FFN).

324 Based on these regression models, the expected FCO₂ anomaly in 2023 was calculated for each fCO₂

325 product from the SST anomaly in 2023. The 2023 SST anomalies (in °C) and the derived expected FCO

326 ₂ anomalies (in PgC yr⁻¹) are: 0.19 and -0.12 (CMEMS), 0.2 and -0.15 (fCO₂-Residual), 0.22 and -0.09

- 327 (OceanSODAv2), and 0.23 and -0.1 (SOM-FFN). The mean and standard deviation of this expected
- 328 FCO₂ anomaly is -0.11 ± 0.03 PgC yr⁻¹ (Fig. 1b).

329 **Computation and attribution of flux anomalies**

330 The CO_2 flux (FCO₂) across the air-sea interface is determined as the product of the fugacity difference 331 between ocean and atmosphere (ΔfCO_2), the gas transfer velocity (k_w), and the solubility of CO_2 in

332 seawater (K_0) and is scaled with the fractional ice coverage (f_{ice}) according to:

333
$$FCO_2 = \Delta fCO_2 \times (k_w K_0) \times (1-f_{ice})$$

- **334** The product k_wK₀ is largely temperature independent, because the temperature dependence in k_w and K
- $_{0}$ largely cancel out. Hence, $k_w K_0$ depends primarily on the prevailing wind speed. Applying a first order
- **336** Taylor decomposition, one can attribute flux anomalies to their drivers as:

(1)

337 $'FCO_2 = '\Delta fCO_2 \times (k_w K_0)_{\text{baseline}} + \Delta fCO_{2,\text{baseline}} \times (k_w K_0) + '\Delta fCO_2 \times '(k_w K_0)$ (2)

338 , where prime (') symbols and the index "baseline" denote anomalies and detrended baseline estimates,

respectively. The modulation by the fractional ice coverage is neglected in the attribution, as our study

- focuses on the ice-free ocean. We compute the flux anomaly contributions initially on the original grid
- of our estimates (1 monthly, $1^{\circ}x1^{\circ}$) and then average the components in space and time (e.g. to compute
- biome annual means).

343 Thermal and non-thermal decomposition of fCO₂ anomalies

To assess the mechanistic drivers causing the 2023 anomalies in ΔfCO_2 , we decomposed the main contributor to this anomaly, which is the surface ocean fCO_2 anomaly, into a thermal and a non-thermal component based on the sea surface temperature anomalies. We perform this decomposition initially on the original grid of our estimates (1 monthly, 1°x1°) and then average the components in space and time (e.g. to compute biome annual means).

Specifically, we determine in a first step the thermally-driven fCO₂ anomaly ('fCO_{2,thermal}) according to equation (3):

351
$$'fCO_{2,thermal} = fCO_{2,baseline} \times exp(\gamma_T \times 'SST) - fCO_{2,baseline}$$
 (3)

352 , where fCO_{2,baseline} is the monthly baseline value of fCO₂, γ_T is the temperature sensitivity of fCO₂ 353 (0.0423 K⁻¹)¹², and 'SST is the monthly anomaly in sea surface temperature. Note that the fCO_{2,baseline} 354 inherits a seasonal cycle and is expressed in absolute values that are similar to the observed fCO_2 values. 355 In contrast, 'SST represents only the deviation of the observed SST from the expected baseline value, 356 i.e. it is a numerically small value of positive or negative sign and does not follow a classical seasonal 357 cycle. As a consequence, the variable 'fCO_{2,thermal} computed according to equation (3) is also a 358 numerically small value of positive or negative sign and does not follow a typical seasonal pattern. In this regard, our thermal anomaly component 'fCO_{2,thermal} differs from the widely used thermal component 359 360 of fCO₂ that is defined as fCO_{2,thermal} = fCO_{2,mean} × $exp(y_T \times (SST_{obs} - SST_{mean}))$. In this classical decomposition of absolute fCO₂ values (instead of anomalies), SST – SST_{mean} and hence also fCO_{2,thermal} 361 362 follow a classical seasonal cycle, and the value of fCO_{2,thermal} has the same order of magnitude as fCO₂ 363 itself. $fCO_{2,thermal}$ can be considered as the seasonal cycle of fCO_2 driven solely by the seasonal cycle in 364 SST.

Based on the anomaly components 'fCO_{2,thermal} and 'fCO₂, we determine the non-thermally driven fCO₂ a
 nomaly ('fCO_{2,non-thermal}) as:

367
$$'fCO_{2,non-thermal} = 'fCO_2 - 'fCO_{2,thermal}$$

(4)

368 While our definition of 'fCO_{2,non-thermal} according to equation (4) resembles the definition of the fCO₂ 369 residual in previous works^{19,37}, it differs in that it does not inherit a classical seasonal cycle. Likewise, 370 our anomaly component 'fCO_{2,non-thermal} differs from the widely used^{15,38} non-thermal component of fCO₂ 371 that is defined as fCO_{2,non-thermal} = fCO_{2,obs} × exp(γ_T × (SST_{mean} – SST_{obs})) and describes the fCO₂ 372 seasonality that would occur if SST remained at the annual mean, but all other processes followed their 373 natural seasonal cycle.

374 **Conversion from DIC to fCO**₂ anomalies

- 375 To convert DIC anomalies into fCO₂ anomalies, it is important to consider alkalinity (TA) anomalies
- that occur simultaneously, because the fraction of the DIC anomaly that is caused by the TA anomaly
- has no effect on fCO₂. The conversion can formally be derived by considering the sensitivity of fCO_2 to
- **378** changes in either DIC or $TA^{14,39}$:

(7)

(9)

- $379 \quad \gamma_{\text{DIC}} = (\Delta f CO_{2,\text{DIC}} / f CO_2) / (\Delta \text{DIC} / \text{DIC})$ (5)
- 380 and
- $381 \quad \gamma_{TA} = (\Delta fCO_{2,TA} / fCO_2) / (\Delta TA / TA)$ (6)
- 382 Given that the total change in fCO₂ is the sum of the change driven by TA and DIC:
- $\Delta fCO_2 = \Delta fCO_{2,DIC} + \Delta fCO_{2,TA}$
- and assuming that to first order $\gamma_{DIC} \approx -\gamma_{TA}$ and DIC \approx TA, one can insert the two sensitivities (5 and 6) in equation (7) and derive the expression:

$$\Delta fCO_2 = \gamma_{DIC} (fCO_2 / DIC) (\Delta DIC - \Delta TA)$$
(8)

387 One can interpret the term $\Delta DIC - \Delta TA$ as the effective change in DIC that is not compensated for by a 388 change in TA. Intuitively, a positive DIC anomaly that is not fully balanced by a TA anomaly would 389 lead to a positive fCO₂ anomaly.

- Interestingly, one can further introduce the approximation that the carbonate ion concentration $[CO_3^{2-}]$ 391 \approx TA - DIC. Hence, the changes in fCO₂ can also be expressed as:
- **392** $\Delta fCO_2 = -\gamma_{DIC} (fCO_2 / DIC) \Delta [CO_3^{2-}]$
- 393 This equation can be interpreted such that a positive anomaly in carbonate ion concentration is
- 394 equivalent to a negative anomaly in fCO₂. Hence, our Fig. 4c could be redrawn with an inverted colour
- 395 scale and showing $[CO_3^{2-}]$ instead of DIC -TA.
- 396 In this study, we used the salinity-normalised anomalies in DIC and TA for ΔDIC and ΔTA, 397 respectively. We approximated the absolute values of DIC, fCO_2 , and γ_{DIC} as 2000 µmol kg⁻¹, 420 µatm, 398 and 10. It should be noted that this conversion only serves as a first-order approximation of the
- **399** relationship between fCO₂ and DIC anomalies on biome scales.

400 Data availability

- 401 The surface ocean fCO₂ observations that were used for the training and validation of the fCO₂ products
 402 are publicly available under:
- 403 SOCATv2023: <u>https://doi.org/10.25921/r7xa-bt92</u>
- 404 SOCATv2024: https://doi.org/10.25921/9wpn-th28
- For two stations, 2023 observations were sourced from SOCATv2024 and analysed in detail. The directaccess to these data is possible through the following DOIs:
- 407 TAO170W: 10.3334/cdiac/otg.tsm_tao170w_0n
- 408 BATS: 10.25921/r7vk-e838
- 409 The fCO₂ products and GOBM simulations analysed in this study will be made available through
- 410 Zenode and receive a doi upon acceptance of this article for publication.

411 Code availability

- 412 All code required to perform the analysis and prepare figures presented in this study is available under:
- 413 <u>https://github.com/jens-daniel-mueller/heatwave_co2_flux_2023</u>
- 414 Upon acceptance of this article for publication, the latest version of the code will be made available415 through Zenodo and receive a doi.

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516





518 Extended Data Fig. 1: Same as Fig. 1, but showing the fCO₂ products and GOBMs individually.



- 519 Extended Data Fig. 2: Maps of (a) ocean biomes and (b) fCO₂ observations available from SOCATv2024 for the
- 520 historical period and 2023. The abbreviations in (a) stand for the subpolar seasonally stratified (SPSS), subtropical
- seasonally stratified (STSS), subtropical permanently stratified (STPS) biomes. Note that the equatorial Pacific
 biome is split into an eastern and western part at the black line. These biomes are used to regionally integrate or
- 523 average spatially resolved estimates. The labels in panel (b) show the three stations for which 2023 observations
- are compared to the fCO₂ product estimates (see SI): 1 = Bermuda Atlantic Time Series; 2 = Equatorial Pacific; 3 =
- 525 VOS Line (Gran Canaria Barcelona).



Extended Data Fig. 3: Same as Fig. 2, but showing additional estimates for various biomes and regions asprovided also in Table 1.

21



Extended Data Fig. 4: Anomaly estimates from Table 1 displayed as bars (fCO₂ product ensemble mean) with
 uncertainty ranges (fCO₂ product ensemble standard deviation). In addition, estimates for the individual fCO₂
 products and GOBMs are displayed as symbols.



Extended Data Fig. 5: Same estimates as shown in Fig. 3a-c, but averaging the attribution of the 2023 annual mean flux anomalies over the ocean biomes shown in Extended Data Fig. 2. Flux anomalies are attributed to their primary drivers, that is, the CO_2 fugacity gradient between ocean and atmosphere (ΔfCO_2), the product of the gas transfer velocity and the solubility of CO_2 ($k_w K_0$), which is primarily controlled by wind speed, as well as the cross

535 product of both drivers ($\Delta fCO_2 \times k_w K_0$).



536 Extended Data Fig. 6: Same estimates as shown in Fig. 3d-f, but averaging the 2023 fCO₂ anomaly
537 decomposition over the ocean biomes shown in Extended Data Fig. 2. Total fCO₂ anomalies (grey) are
538 decomposed into the thermal (red) and non-thermal (blue) components.



Extended Data Fig. 7: Same as Fig. 3a, but showing the four fCO₂ products and two GOBMs individually. In
addition to the three biomes in Fig. 3, estimates for the non-polar global ocean are displayed. Furthermore, the
decomposition of the fCO₂ anomalies into their thermal and non-thermal components is shown.



542 Extended Data Fig. 8: Annual mean surface anomaly maps based on the mean of two GOBM simulations.

543 544 545 **Extended Data Table 1:** Details of fCO_2 products. *All fCO_2 products and underlying predictor fields were regridded to a common spatial resolution of $1^{\circ}x1^{\circ}$ and monthly temporal resolution prior to the analysis presented

in this study.

Product name	CMEMS	fCO2-Residual	SOM-FFN	OceanSODAv2
Reference	Chau et al. (2022)	Bennington et al. (2022)	Landschützer et al. (2016)	Gregor et al. (2024)
Interpolation method	An ensemble of 100 feed- forward neural network models	EXtreme Gradient Boosting (XGB) algorithm applied to fCO_2 -Residual (= pCO_2 - pCO_2 -T). For extension beyond training, $xCO_{2,atm}$ trend is added to reconstructed fCO_2 field.	Clustering-regression approach with self organising map (SOM) and feed-forward neural network (FFN)	Ensemble of clustering- regression approaches
Native resolution*	0.25°x0.25°; monthly	1°x1°; monthly	1°x1°; monthly	0.25°x0.25°; 8-daily
Sea surface fCO ₂ training data	SOCATv2023; monthly, 1°x1°	SOCATv2023; monthly, 1°x1°	SOCATv2023; monthly, 1°x1°	SOCATv2023; monthly, 1°x1°
SST predictor	CMEMS	OISST	HadiSST	ESA
SSS predictor	CMEMS	EN4	EN4	SODA/ESA
SSH predictor	CMEMS	-	-	Ssalto/Duacs
Chl-a predictor	CMEMS	Globcolor merged CHL	Globcolor merged CHL	ESA
MLD predictor	ECCO2	deBoyer climatology	Merged MIMOC and deBoyer climatologies	SODA
K _w formulation	Wanninkhof (2014); a adjusted for global mean piston velocity = 16.5 cm/h; scaled by 1-f _{ice}	Wanninkhof (1992) with Schmidt number from Wanninkhof (2014); scaled to bomb 14C (Fay et al., 2021); scaled by 1-f _{ice}	Wanninkhof (1992); a adjusted for global mean piston velocity = 16cm/h; scaled by 1-f _{ice}	Wanninkhof (1992); scaled by 1-f _{ice}
Wind-speed	ERA5	CCMP2, ERA5, JRA55	ERA5	average of ERA5, JRA55, and NCEP1
Atmospheric CO ₂	xCO ₂ from CAMS inversion; sea level pressure from ERA5; water vapor correction from Weiss (1980)	xCO ₂ from NOAA MBL; sea level pressure from ERA5; water vapor correction from Dickson (2007)	xCO ₂ from NOAA MBL; sea level pressure from NCEP; water vapor correction from Dickson (2007)	xCO ₂ from NOAA MBL; sea level pressure from ERA5; water vapor correction from Dickson (2007)
Sea-ice fraction (f _{ice})	OSTIA/CMEMS product	NOAA OISST	Rayner et al. (2003)	OSTIA

Model name **CESM-ETHZ** FESOM-REcoM Reference Doney et al. (2009); Gürses et al. (2023) Lindsay et al. (2014); Yang and Gruber (2016) Vertical layers 60 46 Native horizontal lon: 1.125° 1°; refined in equatorial, coastal, and polar regions lat: 0.27° - 0.53° resolution POP2 model using Levitus data and a state of rest, DIC and TA (GLODAPv2 preindustrial); N, Si, and O2 **Initial conditions** carbonate chemistry from GLODAPv2 pre-industrial (WOA13); Fe (PISCES, corrected with observations), temperature and salinity from PHC3.0 climatology Spin up procedure 180 years with CORE forcing and 1850 pCO₂, switch to Spun up from 1611-1957 with JRA55-do v1.3, repeating JRA forcing for 14 years; the year 1961. 3x cycling through JRA with historical forcing 180 years (CORE) + 14 years (JRA) Length of spin up 1850-1957 JRA-55 v1.3 JRA55-do v1.3 **Physical forcing** K_w parameters a = 0.31 (Wanninkhof 1992), 1-fice a = 0.251 (Wanninkhof 2014), 1-fice Atmospheric CO₂ monthly, global mean monthly, global mean xCO₂ (Friedlingstein et al. 2022) xCO₂ (Friedlingstein et al. 2022)

Extended Data Table 2: Details of GOBM simulations. fice is the fractional ice coverage.

Supplementary Files

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• MuellerOceanCarbonSink2023SI.pdf