1 External forcing explains recent decadal variability of the ocean carbon sink

- 2
- 3 Galen A. McKinley¹, Amanda R. Fay¹, Yassir A. Eddebbar², Lucas Gloege¹,
- 4 Nicole S. Lovenduski³
- 5
- 6 Affiliations: ¹Lamont Doherty Earth Observatory / Columbia University, Palisades NY. ²Scripps
- 7 Institution of Oceanography, La Jolla, CA.³ University of Colorado Boulder, Boulder, CO.
- 8 *Corresponding author: Email: mckinley@ldeo.columbia.edu
- 9
- 10
- 11 Key points:
- The reduced ocean carbon sink in the decade of the 1990s can be explained by the slowed growth rate of atmospheric CO₂.
- The global sea surface temperature response to Mt Pinatubo in 1991 explains the intra decadal pattern of the ocean carbon sink in the 1990s.
- There will be an immediate reduction in ocean carbon uptake as atmospheric pCO₂
 responds to cuts in anthropogenic emissions.

The ocean has absorbed the equivalent of 39% of industrial-age fossil carbon emissions, 18 19 significantly modulating the growth rate of atmospheric CO₂ and its associated impacts on climate. Despite the importance of the ocean carbon sink to climate, our 20 21 understanding of the causes of its interannual-to-decadal variability remains limited. This hinders our ability to attribute its past behavior and project its future. A key period 22 of interest is the 1990s, when the ocean carbon sink did not grow as expected. Previous 23 explanations of this behavior have focused on variability internal to the ocean or 24 associated with coupled atmosphere/ocean modes. Here, we use an idealized upper ocean 25 box model to illustrate that two external forcings are sufficient to explain the pattern and 26 27 magnitude of sink variability since the mid-1980s. First, the global-scale reduction in the 28 decadal-average ocean carbon sink in the 1990s is attributable to the slowed growth rate 29 of atmospheric pCO₂. The acceleration of atmospheric pCO₂ growth after 2001 drove recovery of the sink. Second, the global sea surface temperature response to the 1991 30 31 eruption of Mt Pinatubo explains the timing of the global sink within the 1990s. These results are consistent with previous experiments using ocean hindcast models with and 32 33 without forcing from variable atmospheric pCO_2 and climate variability. The fact that variability in the growth rate of atmospheric pCO₂ directly imprints on the ocean sink 34 implies that there will be an immediate reduction in ocean carbon uptake as atmospheric 35 pCO₂ responds to cuts in anthropogenic emissions. 36

37

38 Plain Language Summary

39 Humans have added 440Pg of fossil fuel carbon to the atmosphere since 1750, driving up the atmospheric CO₂ concentration. But not all of this carbon remains in the atmosphere. The 40 ocean has absorbed 39%, substantially mitigating anthropogenic climate change. Though this 41 "ocean carbon sink" is a critical climate process, our understanding of its mechanisms remains 42 limited. Of great interest is the unexplained slow-down of the ocean carbon sink in the 1990s 43 44 and a subsequent recovery. In this work, we use a simple globally-averaged model to show that two processes external to the ocean are sufficient to explain the slowing of the ocean 45 carbon sink in the 1990s. First, a reduced rate of accumulation of carbon in the atmosphere 46 after 1989 reduced the atmosphere-ocean gradient that drives the ocean sink. Second, the 47 48 eruption of Mt Pinatubo led to changes in ocean temperature that modified the timing of the 49 sink from 1991 to 2001. We illustrate that the most important control on the decade-averaged magnitude of the ocean sink is variability in the growth rate of atmospheric CO₂. This implies 50 that as future fossil fuel emission cuts drive reduced growth of atmospheric CO₂, the ocean 51 52 sink will immediately slow down.

- 53
- 54
- 55

56

57 **1. Introduction**

58 The ocean has absorbed the equivalent of 39% of fossil carbon emissions since 1750, significantly 59 modulating the growth of atmospheric CO₂ and the associated climate change (Friedlingstein et 60 al. 2019; Le Quéré et al., 2018a,b; McKinley et al., 2017; Ciais et al., 2013). This sink is expected to grow and substantially mitigate atmospheric carbon accumulation for the next several centuries 61 (Randerson et al., 2015). Yet, we lack a detailed understanding of spatial and temporal variability 62 in the sink and its underlying mechanisms. Incomplete understanding of ocean flux variability 63 64 contributes to significant uncertainty in the global anthropogenic carbon budget for recent decades; this uncertainty is equivalent to $\sim 10\%$ of the atmospheric pCO₂ growth rate (Friedlingstein et al. 65 66 2019; Le Quéré et al., 2018a,b). This imbalance, in turn, limits the scientific community's ability to inform international efforts to curb fossil fuel emissions (Peters et al., 2017). 67

68 Recent studies have concluded that ocean hindcast models, which have long been used to assess the ocean sink, may significantly underestimate its variability (Gruber et al., 2019a; Le Quéré et 69 70 al., 2018b). These conclusions are based on comparisons to new observation-based gap-filled 71 products that suggest substantially larger interannual to decadal variability (Landschützer et al., 72 2015, 2016; Rödenbeck et al., 2013, 2015). It is possible, however, that these gap-filled products do not accurately represent the surface ocean carbon cycle given that their raw input data cover 73 74 only 1.5% of the global ocean in the last 3 decades, and only 3.5% in the most recent years (Bakker 75 et al., 2016). Variability may be amplified by the significant extrapolation that occurs when global 76 full-coverage maps are produced from these very sparse data (Fay et al., 2018; Rödenbeck et al., 77 2015).

78 The ocean carbon sink of the 1990s is of particular interest. During this time, the growth of the 79 ocean sink stalled from its expected growth (DeVries et al., 2019; Fay and McKinley, 2013; 80 Landschützer et al., 2015; Le Quéré et al., 2007; Lovenduski et al., 2008). Using a data-constrained ocean circulation model, it has been suggested that this slow-down was caused by excess 81 82 outgassing of natural carbon due to an anomalously strong overturning of the ocean's upper 1000m (DeVries et al., 2017). Changing patterns of wind-driven circulation in the Southern Ocean have 83 84 also been proposed (Gruber et al., 2019a; Landschützer et al., 2015). Consensus has yet to be 85 achieved on the mechanisms driving changes in the 1990s sink. It is critical that we accurately 86 quantify and diagnose this variability so that we can better project the future ocean carbon sink and, thus, the degree to which the ocean carbon sink will continue to mitigate global climate 87 88 change.

89 **2.** Methods

In this study, we compare the ocean carbon sink since 1980 as estimated from six ocean hindcast
 models, four observationally-based products, and a theoretical upper ocean box model. We
 supplement our analysis with results from nine ocean hindcast models run in constant climate
 model (DeVries et al., 2019). Supporting information provides additional methodological details.

94 **2.1 Models and Products**

95 Six (6) ocean hindcast models for 1980-2017 are the primary basis for this analysis (Table S1).

96 Ocean hindcast models are gridded, three dimensional representations of the evolution of the ocean

- 97 state for recent decades. These models have been forced at the surface by winds, heat and buoyancy
- 98 fluxes from reanalyses of the past atmospheric state. The models we analyze are those used in the 99 recent versions of the Clobal Carbon Budget (La Quéré et al. 2018a h)
- 99 recent versions of the Global Carbon Budget (Le Quéré et al., 2018a,b).

100 Four (4) observationally-based products are also utilized, chosen because they all cover 1985-2016

101 (Table S2). While models have full coverage in space and time, in situ observations only cover a102 small fraction of the global surface ocean. These observation-based products utilize gap-filling

103 techniques to estimate values in all periods and areas not directly observed. Interpolation methods

104 fill spatial and temporal gaps by assuming statistical relationships with neighboring or similar areas

- 105 with available observations.
- 106 Ensemble means of the 4 observationally-based products and the 6 hindcast models are calculated.

107 **2.2 Flux analysis**

Model fluxes span years 1980-2017, while observationally-based products span the 1985-2016 period (Table S1, S2). An area-weighted annual mean timeseries is calculated for each model and product for regions of interest: i) global, ii) east equatorial Pacific biome (Fay & McKinley, 2014), and iii) global, excluding east equatorial Pacific biome. We note that individual models and products utilize different methods for the flux calculation, including wind speed products and parameterizations. The standard approach of our field is to use the mean of these estimates as the current best-estimate of the air-sea flux (DeVries et al., 2019; Le Quéré et al., 2018a,b).

2.3 pCO₂ analysis

For pCO₂, hindcast models span 1980-2017, while observationally-based products cover 1985-2016. Detrended pCO₂^{ocean} is obtained by removing the atmospheric trend of 1.70 μ atm/yr (the observed average annual atmospheric pCO₂ change for 1980 to 2017) from each time series. Δ pCO₂ is calculated at annual timescales by surface ocean pCO₂ - atmospheric pCO₂ [Δ pCO₂ = pCO₂^{ocean} - pCO₂^{atmosphere}].

121 **2.4** Accounting for open ocean areas without observationally-based estimates

122 There are large differences in the spatial extent of the observationally-based products compared to 123 the hindcast models. This difference, when left unaccounted for, results in significant discrepancies 124 between models and products for both global mean flux and global mean pCO₂. We correct for 125 spatial coverage differences by calculating the mean flux or mean pCO₂ in the area of each model 126 where there is no coverage for a specific data product. By calculating this offset from 6 models for each individual data-product's spatial coverage, we generate a product-specific offset for the 127 128 annual time series that corrects for the difference in spatial coverage. As expected, observationally-129 based products with more complete spatial coverage have smaller offsets (Table S2). The only exception to this data-product correction process is the JENA product (Rödenbeck et al., 2013) 130 131 because it is produced with full spatial coverage and is primarily used in this analysis on its coarser native grid. 132

133 134

2.5 Upper Ocean Box Model

The box model (Fig S1) solves for the time change of Dissolved Inorganic Carbon (DIC) in singleupper ocean box (Equation 1).

137 138

$$\frac{dDIC}{dt} = \frac{v}{v} \left(DIC_{deep} - DIC \right) - \frac{k_w s_o \rho}{dz} \left(pCO_2^{ocean} - pCO_2^{atmosphere} \right)$$
(1)

139 The first term on the right of Equation 1 is the overturning circulation (v) acting on the surface to 140 depth gradient of DIC. V is the volume of the global ocean box, $V = A^*dz$. Our value for the global 141 area removes ice-covered regions. The second term on the right of Equation 1 is the air-sea exchange of CO₂. The rate of flux is set by a piston velocity (k_w) , solubility (S_0) and density (ρ) 142 143 over the depth of the box (dz = 200m), multiplied by the ocean to atmosphere difference of pCO₂ (Wanninkhof, 2014). So and pCO2^{ocean} are calculated using full carbon chemistry (Dickson and 144 Millero, 1987; Mehrbach et al., 1973) given inputs of temperature, salinity, alkalinity and DIC. 145 146 Parameters choices are globally representative values outside the tropics (Table S3). Consistent with current understanding of the drivers of ocean uptake of anthropogenic carbon (Gruber et al., 147 148 2019b), the biological pump is assumed constant over time. This leads to our abiotic formulation. 149 Thus, we remove from the DIC_{deep} concentration the amount of carbon that would be vertically 150 supplied, and then removed biologically. We take a global mean DIC_{deep} concentration of 2320 151 mmol/m³ and a biological pump component of this of 265 mmol/m³ (Sarmiento and Gruber 2006, 152 Table 8.2, Figure 8.4.2, organic + carbonate) leading to $DIC_{deep} = 2055 \text{ mmol/m}^3$.

153

NOAA ESRL surface marine boundary layer annual mean observed xCO2^{atmosphere} is used to force 154 the model. This is the same $xCO_2^{atmosphere}$ data used to force the ocean hindcast models and 155 observationally-based products, and in conversion to pCO2^{atmosphere}, the water vapor correction is 156 applied. pCO₂^{atmosphere} is interpolated linearly to monthly resolution, using the annual mean value 157 158 at the mid-point of each year. Temperature is held at a constant global surface ocean value, except if the impact of volcanoes is included (Fig. S2). This estimate of the forced sea surface temperature 159 160 (SST) response to the El Chichon and Mt Pinatubo volcanic eruptions is based on the Community Earth System Model Large Ensemble (CESM LENS) (Eddebbar et al., 2019). The global-mean 161 expression of this forced temperature anomaly extends to several hundred meters depth in CESM 162 163 LENS. The box model is time stepped with monthly resolution for 1959-2018. In all cases, the box model is spun up from 1959-1979 using observed pCO2^{atmosphere} and the values presented in Table 164 165 S3.

166

167 The mean uptake flux in the box model is most sensitive to the depth of the box and the global 168 overturning rate (Fig S3, Supporting Information). We use dz = 200m and set other parameters to 169 result in a mean flux and ΔpCO_2 consistent with the ocean models and observationally-based 170 products.

171 **3.** Results/Discussion

Global air-sea CO_2 flux variability estimated by the ensemble means of ocean hindcast models and of observationally-based products (Gruber et al., 2019a; Landschützer et al., 2016) are highly correlated (r = 0.95) (Fig 1a, Table S4). The decadal variability of the air-sea CO_2 flux is not driven by a single region, but instead is largely globally coherent (DeVries et al., 2019; Le Quéré et al., 2019b) This single CO_2 flux is primerile determined by the difference of the surface energy of

the atmosphere pCO₂, $\Delta pCO_2 = pCO_2^{ocean} - pCO_2^{atmosphere}$ (Fay & McKinley, 2013; Landschützer 177 178 et al., 2015; Lovenduski et al., 2007; McKinley et al., 2017) with a more negative ΔpCO₂ driving 179 a greater ocean uptake. From 1991 to 1993, ΔpCO_2 experiences a negative or neutral anomaly on the global average and at most latitudes over the 91% of ocean that is south of 45°N (Fig 2, S3). 180 181 From 1993-1995 and again for 1999-2001, ΔpCO_2 anomalies were positive at most latitudes. In 1997-1998, the El Niño event drove negative ΔpCO_2 anomalies in the tropics (Fig 2), but the El 182 Niño-Southern Oscillation (ENSO) cycle does not dominate the global mean decadal variability 183 (Fig 2, Fig S5). Following 2001, ΔpCO₂ anomalies become much more negative at all latitudes 184 185 outside the tropics and in the global average (Fig 2).

From 1980 to 2017, pCO₂^{atmosphere} increased from 330 µatm to 394 µatm, and pCO₂^{ocean} follows the 186 187 atmosphere on the long-term (Fig 3a). Detrending reveals the detailed features of these timeseries (Fig 3b). Growth of pCO₂^{atmosphere} slowed significantly with respect to the long-term trend starting 188 in the late 1980s (Fig 3b, Sarmiento 1993). This change was due in part to a pause of growth in 189 190 fossil fuel emissions from 1989 to 1994 when fossil fuel emissions were approximately constant 191 at 6.1 PgC/yr (Friedlingstein et al. 2019; Sarmiento et al., 2010). Increased uptake of carbon by the terrestrial biosphere also contributed significantly to this slow down (Brovkin et al., 2010; 192 Sarmiento et al., 2010; Angert et al., 2004). Though some studies have attributed the increased 193 194 land carbon sink to temperature and radiation changes caused by Mt. Pinatubo's 1991 eruption, 195 there is not a consensus with respect to the mechanisms on land (Frölicher et al., 2011; Brovkin et 196 al., 2010; Sarmiento et al., 2010; Angert et al., 2004; Lucht et al., 2002).

197 Deviations in the evolution of pCO₂^{ocean} from the evolution of pCO₂^{atmosphere} drive Δ pCO₂ changes 198 in the observationally-based products and hindcast models (Fig 3c). Since global mean ΔpCO_2 is only ~5 µatm (Fig 3c), anomalies of a few µatm in either pCO₂^{atmosphere} or pCO₂^{ocean} can 199 significantly impact the air-sea CO₂ flux. In 1992, growth of pCO₂^{ocean} abruptly slowed in both the 200 201 hindcast models and the observationally-based products (Fig 3b) and ΔpCO_2 becomes more negative (Fig 3c). From 1992 through 2001, pCO2^{ocean} increases more rapidly than pCO2^{atmosphere}, 202 driving a positive change in ΔpCO_2 over this period (Fig 3c). For 2002-2011 and beyond, pCO_2^{ocean} 203 grows more slowly than the strongly accelerating $pCO_2^{atmosphere}$, leading to increasingly negative 204 205 ΔpCO₂ over 2002-2011 (Fig 3c).

Given that global mean changes in the ocean sink (Fig 1) are found to occur at most latitudes (Fig 2), we hypothesize an important role for external forcing. To explore these drivers, we apply the upper ocean box model (Fig S1). The key processes captured by this model are (1) ventilation to the surface of deep waters with low anthropogenic carbon content, and (2) air-sea gas exchange. First, we ask: Are the changes in the growth rate of pCO₂^{atmosphere} (Fig 3a,b) sufficient to substantially modify the global ocean carbon sink?

To test this, the box model is forced only with changes in the observed $pCO_2^{atmosphere}$. When the growth rate of $pCO_2^{atmosphere}$ slows in the late 1980s, growth of pCO_2^{ocean} gradually slows in

213 growth rate of $pCO_2^{atmosphere}$ slows in the late 1980s, growth of pCO_2^{ocean} gradually slows in 214 response (Fig 3b, red dashed line). pCO_2^{ocean} achieves a few years after the $pCO_2^{atmosphere}$ minimum

in 1994, consistent with the long equilibration timescale for carbon due to carbonate chemistry

215 In 1794, consistent with the long equilibration timescale for earbon due to carbonate enclinisity 216 (Fig S6). Considering the temporal evolution of the ΔpCO_2 (Fig 3c) and flux over 1988-1994, the

rapid slowdown of pCO₂^{atmosphere} growth would have caused an outgassing pulse centered on 1993

- 218 (Fig 1b). Beyond 1994, $pCO_2^{atmosphere}$ growth returned to approximately its long-term growth rate 219 and then grew more rapidly after 2001 (Fig 3b). Since pCO_2^{ocean} lags behind $pCO_2^{atmosphere}$, ΔpCO_2 220 grows more negative as the atmosphere accelerates (Fig 3b,c). For the box model forced only with 221 $pCO_2^{atmosphere}$, this increasingly negative ΔpCO_2 drove a steady increase in ocean uptake from 1994 222 onward (Fig 1b).
- This same response of the global mean air-sea CO₂ flux to pCO₂^{atmosphere} forcing occurs in two 223 types of three-dimensional ocean models with constant circulation. Both an ensemble of ocean 224 hindcast models with constant climate forcing (DeVries et al., 2019) and the data-constrained 225 226 constant circulation Ocean Circulation Inverse Model (OCIM) (DeVries, 2014) have the same flux 227 response as in the box model forced only with variable pCO₂^{atmosphere} (Fig 1b). Correlations across these models are almost perfect (r=0.97-0.98, dashed lines in Fig 1b) and remain very high even 228 229 when independently detrended (r>0.91, Table S4). This correspondence emphasizes the critical role of variability in the growth of pCO₂^{atmosphere} to variability in the ocean sink. It also serves as 230 231 evidence that the box model is realistically estimating the timing and magnitude of this response.
- **232** Though it is the slowed growth of $pCO_2^{atmosphere}$ in the early 1990s that causes the mean 1990s sink

to be only 0.1 PgC/yr larger than the sink of the 1980s (Table S5), the pattern of the sink change

in the 1990s is clearly inconsistent with ocean hindcast models with variable climate (compare

bold green to red dash in Fig 1b), and thus is also inconsistent with the observationally-based

236 products (Fig 1a). An additional mechanism is required.

- 237 Volcanic eruptions of El Chichon in 1982 and Mt. Pinatubo in 1991 injected large quantities of 238 sulfate aerosols into the stratosphere and dramatically altered global air and sea surface 239 temperatures (Church et al., 2005). The forced response to these eruptions was a substantial 240 oceanic uptake of carbon and oxygen for the following 2-3 years (Eddebbar et al., 2019). Modern 241 earth system models indicate that a significant negative anomaly in global sea surface temperatures 242 (SST) was driven by the eruptions (Eddebbar et al., 2019). For the diagnostic box model, we apply 243 the same magnitude of forced global SST cooling estimated by these models, 0.1°C in 1982 and 244 0.2°C in 1991 (Fig S2) to evaluate the impact on ocean carbon sink variability.
- 245 With this volcano-forced SST variability applied to the box model, strong coolings in 1982 and in 1991 drive a rapid drop of pCO₂^{ocean} (Fig 3b, solid red) and a strong uptake anomaly (Fig 1, red 246 247 bold). The reduced flux that would have occurred in the early 1990s due to the slowing of pCO₂^{atmosphere} (Fig 1b, red dash) was overwhelmed by the rapid cooling due to Mt. Pinatubo and 248 thus, a strong uptake pulse occurred (Fig 1b, red bold). The re-warming and excess DIC in the 249 surface ocean that follow Mt. Pinatubo elevates pCO₂^{ocean} relative to pCO₂^{atmosphere} through 2001, 250 251 leading to ΔpCO_2 becoming less negative over this period (Fig 3c). Thus, the sink stagnates from 252 the early to late 1990s (Fig 1). The net effect of both forcings is that the reduced sink of the early 1990s caused by the slowed pCO₂^{atmosphere} growth rate is shifted to the late 1990s by the rapid 253 cooling and slow re-warming caused by Mt. Pinatubo. In summary, the climate variability 254 255 mechanism that led to a neutral 1990s intra-decadal trend of the ocean carbon sink (DeVries et al. 256 2019) contains a major contribution from the ocean's response and recovery from Pinatubo 257 cooling, i.e. a response to external forcing from volcanos.

CO₂ fluxes from the diagnostic box model forced with both observed pCO₂^{atmosphere} and the 258 259 volcanoes' impacts on sea surface temperature are highly correlated to the ensemble mean of the 260 observationally-based products for their overlapping periods (r = 0.89, 1985-2016) and hindcast 261 models (r=0.92, 1980-2017) (Fig 1a, Table S4). The simplicity of these global mechanisms and 262 the strong correspondence of the resulting decadal variability to the products and the models supports the conclusion that global air-sea CO₂ flux variability since 1980 has been significantly 263 driven by external forcing from (1) the changing $pCO_2^{atmosphere}$ growth rate and (2) in the 1990s, 264 265 the surface ocean temperature effects of Mt. Pinatubo (Fig 4).

266 Since ocean carbon uptake is enhanced with the eruption of large volcanos, the effect on pCO₂^{atmosphere} would ideally be modeled interactively. Unfortunately, land carbon sink 267 uncertainties preclude this. The sea surface temperature effects of Pinatubo led to an increased 268 269 ocean sink of approximately 0.5 PgC/yr (Fig.1b), but estimates of the land sink anomaly at this time are much larger and uncertain, ranging 1-2 PgC/yr (Sarmiento et al. 2010; Angert et al. 2004; 270 271 Sarmiento 1993). In fact, the post-Pinatubo period is one of maximum uncertainty in the post-1960 Global Carbon Budget (Friedlingstein et al. 2019, Peters et al. 2017). Soon after the eruption of 272 Mt. Pinatubo, Sarmiento (1993) noted the coincident slowdown in growth of pCO₂^{atmosphere} and 273 274 reported that ¹³C records at that time suggested a terrestrial driver, while O₂/N₂ records suggested an oceanic driver. A modern reconsideration of ¹³C and O₂/N₂ records may lead to better 275 276 understanding of this partitioning.

277 This analysis illustrates that externally forced variability played an important role in recent decadal 278 variability of the ocean carbon sink. However, the total climate variability in any variable is the sum of forced variability caused by drivers external to the system, and internal variability due to 279 280 system dynamics (Deser et al., 2012). We have recently reviewed the many previous studies on mechanisms of ocean carbon sink variability (McKinley et al., 2017). Processes discussed have 281 included the variable upper ocean circulation, wind and circulation patterns in the Southern Ocean, 282 283 and modes of coupled atmosphere/ocean variability in both hemispheres (DeVries et al., 2017, 2019; Gruber et al., 2019a; Landschützer et al., 2015, 2019; Lovenduski et al., 2007). These 284 analyses have focused on variability internal to the ocean or associated with coupled 285 286 atmosphere/ocean modes (McKinley et al., 2017), but have not been able to comprehensively explain the global-scale decadal variability. Here, we illustrate that the observed changes can, to 287 288 first order, be attributed to two forcings external to the ocean.

289 Previous studies have also typically focused on a single model or a single observationally-based product. However, for the best estimate of the real ocean's flux variability, it is common practice 290 291 to use the ensemble mean of ocean models and/or of observationally-based products 292 (Friedlingstein et al. 2019; DeVries et al., 2019; Le Quéré et al., 2018a,b), which is our approach. 293 Only the internal variability that is represented in most ensemble members will be preserved in the 294 ensemble average. Averaging damps internal variability of the individual members and thus amplifies the common forced component (McKinley et al., 2016, 2017; Deser et al., 2012). Our 295 results illustrate that the current best estimate of the real ocean's flux variability, based on this 296 297 ensemble average, can be explained largely with forced mechanisms. However, because we do not 298 have enough information to determine which member of the ensemble best approximates the 299 ocean's true internal variability, this current best estimate potentially underestimates the full 300 impact of internal variability in the carbon sink of the real ocean.

301 What is the range of magnitude of internal variability that may be occurring in addition to the 302 forced variability that we identify? Individual observationally-based products have a range of detrended flux variability from 0.14-0.30 PgC/yr, while the ensemble mean variability is 0.19 303 304 PgC/yr (1 σ for 1985-2016). For the hindcast models, the range is 0.10-0.20 PgC/yr and the 305 ensemble mean variability is 0.11 PgC/yr for the same years. Our estimate of the amplitude of externally-forced variability from the box model is 0.14 PgC/yr (Fig 1). By this measure, externally 306 307 forced variability as estimated by the box model is approximately equal to the total amplitude 308 common to the hindcast models, but is only about 70% of the variability common to the products. 309 On top of this, the individual models and products suggest a wide range of additional internal variability. In future studies, separation of the forced component of ocean carbon sink variability 310 driven by changing pCO₂^{atmosphere} and volcanos from the total variability in individual models and 311 products should help to clarify the patterns, magnitudes, and physical and biogeochemical 312 mechanisms of internal variability in the real ocean. For diagnostic (Friedlingstein et al. 2019; Le 313 Quéré et al., 2018a,b) and predictive purposes (Randerson et al., 2015) it is critical to also 314 315 determine which model and observationally-based estimates best represent both the internal and 316 forced variability of the real ocean.

317 Though our box model is sufficient to represent the global-mean behavior of the externally-forced 318 ocean carbon sink in recent decades, other mechanisms may increase in importance in the future. 319 As climate changes have increased impact on ocean physics and biogeochemistry, feedbacks on 320 the carbon sink of increasing magnitude can be expected. Future reduction in the overturning 321 circulation, or increased re-emergence of waters already carrying a high anthropogenic carbon load 322 would reduce the sink. A weaker biological pump would also damp net ocean carbon uptake 323 (Kwon et al. 2009). The reduced buffer capacity of the surface ocean should grow in importance 324 over time, particularly under high emission scenarios (Fassbender et al. 2017). As mitigation of CO₂ emissions occurs, the growth rate of pCO₂^{atmosphere} will slow. With this reduced external 325 326 forcing, the imprint of internal variability on the sink should become more evident. Improved 327 understanding of both internal and external mechanisms is essential to continue to accurately 328 diagnose the evolving ocean carbon sink, and to improve model-based predictions.

329 **4.** Conclusions

330 We have shown that externally forced variability is sufficient to explain a significant portion of current model and observationally-based best-estimates of the recent decadal variability of the 331 global ocean carbon sink (Fig 1a). The reduced ocean carbon sink in the decade of the 1990s was 332 driven by a slowed growth rate of pCO₂^{atmosphere}. The intra-decadal timing of the slowed growth 333 rate in the 1990s was due to the surface ocean temperature response to the Mt. Pinatubo eruption 334 335 in 1991. Volcano-driven cooling first led to an anomalously large sink, and then as surface ocean temperatures recovered, pCO_2^{ocean} was elevated causing the sink to slow. In the box model, only 336 this SST response is needed to replicate the behavior of the observationally-based products and 337 338 the ocean hindcast models (Fig 1), but it would be of great value to perform a deeper analysis of the upper ocean response to Mt. Pinatubo with future studies. From 2001 on, the recovery of the 339 340 global ocean carbon sink is attributable to the enhanced growth rate of $pCO_2^{atmosphere}$ (Fig 4).

341 Implications for the future ocean carbon sink are several. First, we note the relative importance of 342 external vs internal drivers of ocean sink change can be expected to change, and thus both must be

understood. Regarding external forcing, future large volcanic eruptions cannot be predicted, and 343 it is difficult to predict the detailed future of pCO₂^{atmosphere}. Thus, these are now identified as 344 additional sources of uncertainty in decadal predictions and long-term projections (Lovenduski et 345 346 al., 2019; McKinley et al., 2017). The timescales on which additional human interventions in the 347 climate system, such as solar radiation management or nuclear conflict, would mimic these 348 externally forced changes and modify the ocean carbon sink should be considered (Lovenduski et al. 2020; Lauvset et al., 2017). Finally, since the changing growth rate of pCO₂^{atmosphere} is the 349 primary driver of recent variability in the ocean carbon sink, the ocean sink should be expected to 350 slow as reductions in the pCO₂^{atmosphere} growth rate occur in response to climate change mitigation 351 352 efforts (Peters et al., 2017). It is important that this critical feedback on the atmospheric CO_2 353 content be accurately estimated and accounted for in policy making.

354 Acknowledgments: Funding from many countries and agencies has supported the collection of surface ocean pCO₂ data, for development of ocean hindcast models and observationally-based 355 356 products, and for international coordination. Ocean hindcast models with real climate are available 357 from https://www.globalcarbonproject.org/carbonbudget/18/data.htm; ocean hindcast models with constant climate are available from DeVries et al. (2019); and observationally-based products 358 359 available from https://www.nodc.noaa.gov/ocads/oceans/ are 360 SPCO2 1982 2015 ETH SOM FFN.html (SOM-FFN), http://www.bgc-jena.mpg.de/ 361 CarboScope/ https://doi.org/10.6084/m9.figshare.7894976.v1 (JENA), (CSIR), and 362 http://dods.lsce.ipsl.fr/invsat/CMEMS/ (LSCE). The code for the upper ocean box model is 363 available (https://doi.org/10.6084/m9.figshare.11983947.v1). G.A.M., A.R.F and L.G. were 364 supported by NASA NNX17AK19G and by Columbia University. G.A.M., A.R.F. and N.S.L. 365 were supported by NSF OCE-1948664 and OCE-1558225. N.S.L. was also supported by NSF 366 OCE-1752724, NSF PLR-13009540, and the Open Philanthropy Project. This work would not be 367 possible without the efforts of many scientists who have collected surface ocean pCO₂ data and 368 contributed it to the SOCAT database, and to the developers of the observationally-based products 369 based on these data. We thank also the scientists who have contributed their ocean hindcast model 370 results to the Global Carbon Project. Leadership from the International Ocean Carbon 371 Coordinating Project (IOCCP) and the Global Carbon Project (GCP) has been essential to the 372 success of these efforts. 373

374 **References**

- 375
- Angert, A., Biraud, S., Bonfils, C., Buermann, W., & Fung, I. (2004). CO2 seasonality indicates
 origins of post-Pinatubo sink. *Geophysical Research Letters*, *31*(11).
- 378
- Bakker, D. C. E., Pfeil, B., Landa, C. S., Metzl, N., O'Brien, K. M., Olsen, A., et al. (2016). A
 multi-decade record of high-quality fCO₂ data in version 3 of the Surface Ocean CO₂
 Atlas (SOCAT), Earth Syst. Sci. Data, 8, 383–413.
- 382
- Brovkin, V., Loren, S., Jungclaus, J., Raddatz, T., Timmreck, C., Reick, C., et al. (2010).
 Sensitivity of a coupled climate-carbon cycle model to large volcanic eruptions during the last millennium. *Tellus B: Chemical and Physical Meteorology*, 62(5), 674-681.

387 388	Buitenhuis, E. T., Rivkin, R. B., Sailley, S., & Le Quéré, C. (2010). Biogeochemical fluxes through microzooplankton. <i>Global biogeochemical cycles</i> , 24(4).
389 390 391 392 393	Canadell, J. G., Ciais, P., Gurney, K., Le Quéré, C., Piao, S., Raupach, M. R., & Sabine, C. L. (2011). An international effort to quantify regional carbon fluxes. <i>Eos, Transactions</i> <i>American Geophysical Union</i> , 92(10), 81-82.
393 394 395	Church, J. A., White, N. J., & Arblaster, J. M. (2005). Significant decadal-scale impact of volcanic eruptions on sea level and ocean heat content. <i>Nature</i> , <i>438</i> (7064), 74.
397 398 399 400	Ciais, P., Sabine, C., Bala, G., Bopp, L., Brovkin, V., Canadell, J. et al. (2013). The physical science basis. Contribution of working group I to the fifth assessment report of the intergovernmental panel on climate change. <i>IPCC Climate Change</i> , 465-570.
401 402 403	Denvil-Sommer, A., Gehlen, M., Vrac, M., & Mejia, C. (2019). LSCE-FFNN-v1: a two-step neural network model for the reconstruction of surface ocean pCO 2 over the global ocean. <i>Geoscientific Model Development</i> , 12(5), 2091-2105.
404 405 406 407	Deser, C., Phillips, A., Bourdette, V., & Teng, H. (2012). Uncertainty in climate change projections: the role of internal variability. <i>Climate dynamics</i> , <i>38</i> (3-4), 527-546.
408 409 410	DeVries, T. (2014). The oceanic anthropogenic CO2 sink: Storage, air-sea fluxes, and transports over the industrial era. <i>Global Biogeochemical Cycles</i> , <i>28</i> (7), 631-647.
411 412 413	DeVries, T., Holzer, M., & Primeau, F. (2017). Recent increase in oceanic carbon uptake driven by weaker upper-ocean overturning. <i>Nature</i> , <i>542</i> (7640), 215.
413 414 415 416	DeVries, T., Le Quéré, C., Andrews, O., Berthet, S., Hauck, J., Ilyina, T., et al. (2019). Decadal trends in the ocean carbon sink. <i>Proceedings of the National Academy of Sciences</i> , <i>116</i> (24), 11646-11651.
417 418 419 420	Dickson, A. G., & Millero, F. J. (1987). A comparison of the equilibrium constants for the dissociation of carbonic acid in seawater media. <i>Deep Sea Research Part A.</i> <i>Oceanographic Research Papers</i> , 34(10), 1733-1743.
421 422	Dickson, A. G., Sabine, C. L., and Christian, J. R. (Eds.): Guide to Best Practices for Ocean CO2 Measurements, PICES Special Publication, IOCCP Report No. 8, 2007. 8825
423 424 425	Dlugokencky, E.J., Thoning, K.W., Lang, P.M., Tans, P.P. (2017) NOAA Greenhouse Gas Reference from Atmospheric Carbon Dioxide Dry Air Mole Fractions from the NOAA ESRL Carbon Cycle Cooperative Global Air Sampling Network.
427 428 429 430	 Doney, S. C., Lima, I., Feely, R. A., Glover, D. M., Lindsay, K., Mahowald, N., et al. (2009). Mechanisms governing interannual variability in upper-ocean inorganic carbon system and air–sea CO2 fluxes: Physical climate and atmospheric dust. <i>Deep Sea Research Part</i> <i>II: Topical Studies in Oceanography</i>, 56(8-10), 640-655.

431	
432	Eddebbar, Y. A., Rodgers, K. B., Long, M. C., Subramanian, A. C., Xie, S. P., & Keeling, R. F.
433	(2019). El Niño-Like Physical and Biogeochemical Ocean Response to Tropical
434	Eruptions. <i>Journal of Climate</i> , <i>32</i> (9), 2627-2649.
435	
436	Fassbender, A. J., Sabine, C. L., & Palevsky, H. I. (2017). Nonuniform ocean acidification and
437	attenuation of the ocean carbon sink. Geophysical Research Letters, 44(16), 8404–8413.
438	
439	Fay, A. R., & McKinley, G. A. (2013). Global trends in surface ocean pCO2 from in situ
440	data. <i>Global Biogeochemical Cycles</i> , 27(2), 541-557.
441	
442	Fay, A. R., & McKinley, G. A. (2014). Global open-ocean biomes: mean and temporal
443	variability. <i>Earth System Science Data</i> , 6(2), 273-284.
444	
445	Fay, A. R., Lovenduski, N. S., McKinley, G. A., Munro, D. R., Sweeney, C., Gray, A. R., et al.
446	(2018). Utilizing the Drake Passage Time-series to understand variability and change in
447	subpolar Southern Ocean pCO2. <i>Biogeosciences</i> , 15, 3841-3855.
448	
449	Friedlingstein, P., Jones, M.W., O'Sullivan, M., Andrew, R.M, Hauck, J. et al. (2019), Global
450	Carbon Budget 2019, Earth Syst. Sci. Data, 11(4), 1783–1838.
451	
452	Frolicher, I., Joos, F., & Raible, C. (2011). Sensitivity of atmospheric CO2 and climate to
453	explosive volcanic eruptions. <i>Biogeosciences</i> , 8(8), 2317-2339.
454	Concern L. Labelat A.D. Kale C. Mantaine DMC (2010) A communities concerned of the
455	Gregor, L., Lebenot, A.D., Kok, S., Monteiro, P.M.S. (2019) A comparative assessment of the
450	uncertainties of global surface ocean CO_2 estimates using a machine learning ensemble
457	(CSIR-MIL6 version 2019a) – have we fit the wall? Geosci. Model Dev. Discuss
458	Cryber N. Clean M. Mikeloff Eleteber S. E. Danay, S. C. Dytkiewicz, S. Fellows, M. L. et
459	ol (2000) Occorris courses sinks and transport of atmospheric CO2. Clobal
400	<i>Biogeochemical Cycles</i> , 23(1)
401	Diogeochemical Cycles, 25(1).
402	Gruber N. Landschützer P. & Lovenduski N. S. (2019a). The variable Southern Ocean carbon
464	sink Annual review of marine science 11 159-186
465	sink. Annual review of marine science, 11, 159-160.
466	Gruber N. Clement D. Carter B. R. Feely, R. A. Van Heuven, S. Honnema, M. et al.
467	(2019b) The oceanic sink for anthronogenic CO2 from 1994 to
468	2007 Science 363(6432) 1193-1199
469	2007. Science, 505 (0152), 1155 1155.
470	Hauck I Lenton A Langlais C & Matear R (2018) The fate of carbon and nutrients
471	exported out of the Southern Ocean <i>Global Biogeochemical Cycles</i> , 32(10), 1556-1573
472	
473	Jacobson, A. R., Mikaloff Fletcher, S. E., Gruber, N., Sarmiento, J. L. and Gloor M. (2007). A
474	joint atmosphere-ocean inversion for surface fluxes of carbon dioxide: 1. Methods and
475	global-scale fluxes, Global Biogeochem, Cycles, 21, GB1019.
476	doi:10.1029/2005GB002556

477 478 Kalnay, E., Kanamitsu, M., Kistler, R., Collins, W., Deaven, D., Gandin, L., et al. (1996). The 479 NCEP/NCAR 40-year reanalysis project. Bulletin of the American meteorological 480 Society, 77(3), 437-472. 481 482 Kwon, E. Y., F. Primeau, and J. L. Sarmiento (2009), The impact of remineralization depth on 483 the air-sea carbon balance, Nature Geoscience, 2(9), 630-635, doi:10.1038/NGEO612. 484 Landschützer, P., Gruber, N., Haumann, F. A., Rödenbeck, C., Bakker, D. C., Van Heuven, S., et 485 486 al. (2015). The reinvigoration of the Southern Ocean carbon sink. Science, 349(6253), 487 1221-1224. 488 489 Landschützer, P., Gruber, N., & Bakker, D. C. (2016). Decadal variations and trends of the 490 global ocean carbon sink. Global Biogeochemical Cycles, 30(10), 1396-1417. 491 492 Landschützer, P., Gruber, N., Bakker, D. C. E., & Landschützer, P. (2017). An updated 493 Observation-Based Global Monthly Gridded Sea Surface pCO2 and Air-sea CO2 Flux 494 Product from 1982 Through 2015 and Its Monthly Climatology. NCEI 495 Accession, 160558. 496 497 Landschützer, P., Ilvina, T., & Lovenduski, N. S. (2019). Detecting Regional Modes of 498 Variability in Observation-Based Surface Ocean p CO2. Geophysical Research 499 Letters, 46(5), 2670-2679. 500 501 Lauvset, S. K., Tjiputra, J., & Muri, H. (2017). Climate engineering and the ocean: effects on 502 biogeochemistry and primary production. *Biogeosciences*, 14(24), 5675-5691. 503 504 Le Quéré, C., Rödenbeck, C., Buitenhuis, E. T., Conway, T. J., Langenfelds, R., Gomez, A., et 505 al. (2007). Saturation of the Southern Ocean CO₂ sink due to recent climate 506 change. Science, 316(5832), 1735-1738. 507 508 Le Quéré, C., Andrew, R. M., Friedlingstein, P., Sitch, S., Pongratz, J., Manning, A. C., et al. 509 (2018). Global carbon budget 2017. Earth Syst. Sci. Data, 10(1), 405-448. 510 511 Le Quéré, C., Andrew, R. M., Friedlingstein, P., Sitch, S., Hauck, J., Pongratz, J., et al. (2018). 512 Global carbon budget 2018. Earth System Science Data (Online), 10(4). 513 Lovenduski, N. S., Gruber, N., Doney, S. C., & Lima, I. D. (2007). Enhanced CO2 outgassing in 514 515 the Southern Ocean from a positive phase of the Southern Annular Mode. Global 516 *Biogeochemical Cycles*, 21(2). 517 518 Lovenduski, N. S., Gruber, N., & Doney, S. C. (2008). Toward a mechanistic understanding of 519 the decadal trends in the Southern Ocean carbon sink. Global Biogeochemical 520 *Cycles*, 22(3). 521

522 523 524	Lovenduski, N. S., Yeager, S. G., Lindsay, K., & Long, M. C. (2019). Predicting near-term variability in ocean carbon uptake. <i>Earth System Dynamics</i> , <i>10</i> (1), 45-57.
525 526 527 528	Lovenduski, N. S., Harrison, C. S., Olivarez, H., Bardeen, C. G., Toon, O. B., Coupe, J., Robock, A., Rohr, T., and Stevenson S. (2020), The Potential Impact of Nuclear Conflict on Ocean Acidification, <i>Geophys Res Lett</i> , 47(3), 1535, doi:10.1007/s10584-012-0475-8.
529 530 531 532	Lucht, W., Prentice, I. C., Myneni, R. B., Sitch, S., Friedlingstein, P., Cramer, W et al. (2002). Climatic control of the high-latitude vegetation greening trend and Pinatubo effect. <i>Science</i> , <i>296</i> (5573), 1687-1689.
533 534 535 536	McKinley, G. A., Pilcher, D. J., Fay, A. R., Lindsay, K., Long, M. C. and Lovenduski, N. S. (2016), Timescales for detection of trends in the ocean carbon sink, <i>Nature</i> , <i>530</i> (7591), 469–472.
537 538 539 540	McKinley, G. A., Fay, A. R., Lovenduski, N. S., & Pilcher, D. J. (2017). Natural variability and anthropogenic trends in the ocean carbon sink. <i>Annual review of marine science</i> , <i>9</i> , 125-150.
541 542 543 544	Mehrbach, C., Culberson, C. H., Hawley, J. E., & Pytkowicx, R. M. (1973). Measurement of the apparent dissociation constants of carbonic acid in seawater at atmospheric pressure 1. <i>Limnology and Oceanography</i> , <i>18</i> (6), 897-907.
545 546 547 548	Naegler, T., P. Ciais, K. Rodgers, and I. Levin (2006), Excess radiocarbon constraints on air-sea gas exchange and the uptake of CO2 by the oceans, <i>Geophys Res Lett</i> , <i>33</i> (11), L11802, doi:10.1029/2005GL025408.
549 550 551	Paulsen, H., Ilyina, T., Six, K. D., & Stemmler, I. (2017). Incorporating a prognostic representation of marine nitrogen fixers into the global ocean biogeochemical model HAMOCC. <i>Journal of Advances in Modeling Earth Systems</i> , 9(1), 438-464.
553 554 555 555	Peters, G. P., Andrew, R. M., Canadell, J. G., Fuss, S., Jackson, R. B., Korsbakken, J. I., et al. (2017). Key indicators to track current progress and future ambition of the Paris Agreement. <i>Nature Climate Change</i> , 7(2), 118.
557 558 559 560	Randerson, J. T., Lindsay, K., Munoz, E., Fu, W., Moore, J. K., Hoffman, F. M., et al. (2015). Multicentury changes in ocean and land contributions to the climate-carbon feedback. <i>Global Biogeochemical Cycles</i> , 29(6), 744-759.
561 562 563	Reynolds, R. W., Rayner, N. A., Smith, T. M., Stokes, D. C., & Wang, W. (2002). An improved in situ and satellite SST analysis for climate. <i>Journal of climate</i> , <i>15</i> (13), 1609-1625.
564 565 566 567	Rödenbeck, C., Keeling, R. F., Bakker, D. C., Metzl, N., Olsen, A., Sabine, C., & Heimann, M. (2013). Global surface-ocean p (CO2) and sea-air CO2 flux variability from an observation-driven ocean mixed-layer scheme.

568 569 570 571	 Rödenbeck, C., Bakker, D. C., Gruber, N., Iida, Y., Jacobson, A. R., Jones, S., et al. (2015). Data-based estimates of the ocean carbon sink variability–first results of the Surface Ocean pCO₂ Mapping intercomparison (SOCOM). <i>Biogeosciences</i>, <i>12</i>, 7251-7278.
572 572	Sarmiento, J. L. (1993), Carbon-Cycle - Atmospheric CO2 Stalled, Nature, 365(6448), 697–698.
573	Sarmiento I I & Gruber N (2006) Ocean biogeochemical dynamics Princeton University
575	Press
576	
577	Sarmiento, J. L., Gloor, M., Gruber, N., Beaulieu, C., Jacobson, A. R., Mikaloff Fletcher, S. E.,
578	et al. (2010). Trends and regional distributions of land and ocean carbon
579	sinks. <i>Biogeosciences</i> , 7(8), 2351-2367.
580	
581	Schwinger, J., Goris, N., Tjiputra, J. F., Kriest, I., Bentsen, M., Bethke, I., et al. (2016).
582	Evaluation of NorESM-OC (versions 1 and 1.2), the ocean carbon-cycle stand-alone
583	configuration of the Norwegian Earth System Model (NorESM1). Geoscientific Model
584	Development, 9, 2589-2622.
585	
586	Seferian, R., Delire, C., Decharme, B., Voldoire, A., y Melia, D. S., Chevallier, M., et al. (2016).
587	Development and evaluation of CNRM Earth system model-CNRM-ESMI.
200	Takahashi T. S. Sutharland P. Wanninkhof C. Swaanay, P. Eaaly, D. Chinman, P. Halas, G.
203	Friederich E Chavez and C Sabina (2000). Climatelogical mean and decadal change in
590	surface ocean pCO^2 and net sea-air CO2 flux over the global oceans. Deen Seq
592	Research Part II 56(8-10) 554–577
593	Research 1 an 11, 50(0 10), 554 577.
594	Wanninkhof, R. (2014) Relationship between wind speed and gas exchange over the ocean
595	revisited. Limnology and Oceanography: Methods, 12(6):351-362.
596	





599 Fig 1: Air-sea CO₂ flux of anthropogenic carbon from observationally-based products (blue), 600 hindcast models (green) and upper ocean diagnostic box model (red); negative flux into the ocean. (A) global (bold), with range of individual members (shading), (B) anomalies of air-sea 601 CO₂ flux for the hindcast models with constant climate and variable pCO₂^{atmosphere} (green 602 dashed), and variable climate and variable pCO2^{atmosphere} (solid green); box model with only 603 pCO2^{atmosphere} forcing (dashed red) and both pCO2^{atmosphere} and volcano-driven SST forcing (solid 604 red); and the constant circulation Ocean Circulation Inverse Model (dash black, DeVries et al. 605 2014) that imposes variable pCO₂^{atmosphere}. In B, dashed lines are correlated at 0.97-0.99 and the 606 607 solid lines 0.92 (Table S4). In A, the mean flux of the observationally-based products is 608 increased by 0.45 PgC/yr (Jacobson et al. 2007) to account for the outgassing of natural carbon 609 supplied by rivers to the ocean.



Fig 2: Latitudinal mean anomaly ΔpCO₂ (µatm) from the ensemble mean of the
 observationally-based products. Anomaly is calculated from the 1990-1999 mean. Annual
 ΔpCO₂ time series overlaid in black for global (solid) and global excluding east equatorial
 Pacific biome (dashed).



Fig 3: Trends of pCO₂^{atmosphere} (black) and pCO₂^{ocean} (colors) (A) with trend, (B) detrended 618 with the long-term pCO₂^{atmosphere} trend (1.70 μ atm/yr from 1980 to 2017), (C) Δ pCO₂ (= 619 $pCO_2^{ocean} - pCO_2^{atmosphere}$). Observationally-based products mean (blue), hindcast model mean 620 (green) and upper ocean diagnostic box model (red). The box model is forced with only 621 $pCO_2^{atmosphere}$ (dashed) and with both $pCO_2^{atmosphere}$ and volcano-associated SST change (solid). 622 623 Hindcast models without the water vapor correction applied to their atmospheric pCO₂ time 624 series are corrected to account for that difference. Fig S6 expands on these results by including 625 additional box model forcing scenarios.

Figure 4



626

- 627 Fig 4: Mechanisms of recent decadal variability of the ocean carbon sink. (A) The reduced
- 628 sink of the 1990s (black arrow) was due to a slowing of the $pCO_2^{atmosphere}$ growth rate, and the
- 629 rapid cooling and slower warming recovery in response to the eruption of Mt. Pinatubo (red
- 630 arrows). (**B**) In the 2000s and beyond, $pCO_2^{atmosphere}$ growth accelerates, leading to enhanced
- **631** ΔpCO_2 and sink growth.

632











Figure 4

pCO₂atm

Stronger Sink