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Rapid rise in the global ocean carbon sink determined from atmospheric oxygen observations

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1 **Abstract:**

2 The rate of increase of atmospheric carbon dioxide (CO₂) in recent decades is only about
3 half that expected if all anthropogenic emissions had stayed airborne. The “missing” CO₂ is
4 absorbed by the oceans and land vegetation¹. Of these two sinks for CO₂, the ocean sink is
5 considered to be the better constrained²⁻⁴. Nevertheless, estimates of ocean uptake, using
6 data products that integrate surface ocean observations⁵⁻⁹, suggest a sink that is larger,
7 and increasing more rapidly since about 2000, than do global ocean models, so that they
8 now disagree substantially¹. Here we examine the recent history of the CO₂ sinks using
9 atmospheric observations of CO₂ and oxygen/nitrogen ratios from globally distributed
10 measurement locations¹⁰⁻¹². These provide a robust and independent separation of the
11 ocean and land contributions. This method indicates that the ocean sink increased rapidly
12 through the years 2000-2015, to values consistent with, though generally larger than, the
13 surface-ocean based data products, but inconsistent with models. Since 2015, our results
14 suggest this increase has stalled, but it remains 50% higher than most models predict.

Main

15 The rise in atmospheric carbon dioxide (CO₂) concentrations due to anthropogenic sources is
16 considerably slowed by net uptake by the oceans and land vegetation, which together
17 remove about half the amount of CO₂ emitted^{1,13}. The total of these sinks for CO₂ is well
18 constrained, since it is the difference between the CO₂ emitted by fossil fuel combustion and
19 the amount accumulating in the atmosphere, both of which are known with reasonable
20 accuracy^{1,14,15}. However, how much of the removed CO₂ goes into the land and how much to
21 the ocean is less certain. In particular the ocean sink, which is the less variable term and is
22 traditionally assumed to be better determined, is currently under scrutiny^{9,16-18}. The Global
23 Carbon Project (GCP) which publishes an annual accounting of the carbon budget, uses
24 ensembles of both ocean biogeochemical models (OBGMs) and data products based on
25 surface ocean observations to calculate this term¹. These two methods show steady
26 divergence over the last two decades, with the observation-based estimates increasing
27 faster than the models. The average of the data products for the CO₂ flux into the ocean now
28 exceeds the model average by about 0.8 petagrams of carbon per year (PgC yr⁻¹), which is
29 around one third of the total ocean sink¹. Both of these methods however have their
30 limitations: the models have relatively coarse spatial resolution and rely on simplified physics
31 and biology¹⁶, whereas the data products are based on observations that are sparse and
32 unevenly distributed in space and time, requiring extensive interpolation using machine
33 learning or regression techniques⁵. For these reasons, an independent estimate of the ocean
34 and land sinks is desirable.

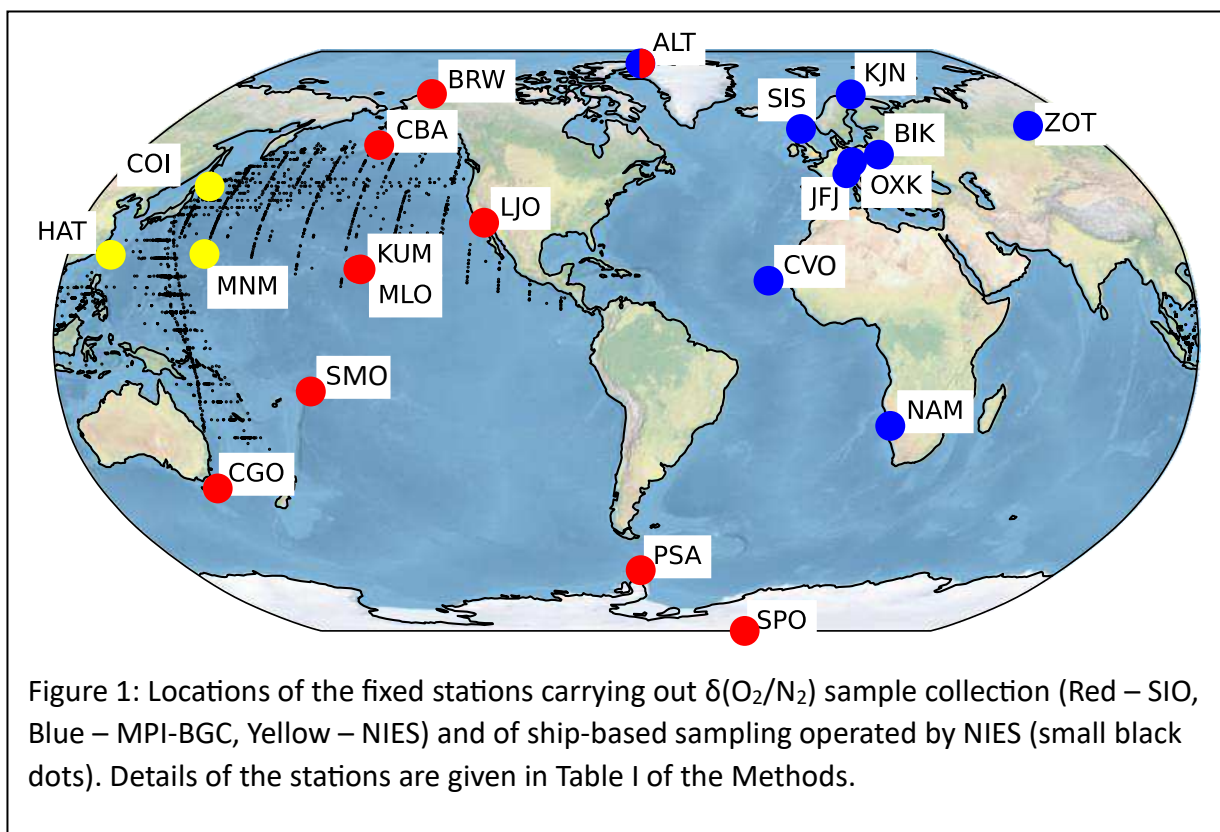
35 High precision measurements of atmospheric O₂/N₂ ratios provide such an independent
36 estimate of net land and ocean carbon sinks¹¹. The method takes advantage of the fact that
37 the net uptake of CO₂ by land vegetation is due to photosynthesis and releases oxygen,
38 whereas uptake of CO₂ by the oceans does not have a similarly anticorrelated O₂ flux. A
39 complication is that since the method was first proposed, it has become clear that the
40 oceans are losing dissolved oxygen due to ocean warming, resulting in a net O₂ source to the
41 atmosphere. The technique has been refined to account for this source of O₂¹⁹, and to use
42 the tracer atmospheric potential oxygen (APO)²⁰, a combination of O₂/N₂ and CO₂
43 measurements defined so that it is invariant to photosynthesis and therefore less subject to

44 local interference sources. We use this method here, applied to the much larger number of
45 atmospheric time series now available.

46 **Calculation of atmospheric potential oxygen from time series observations.**

47 We use flask sample observations from analyses made at three laboratories: Scripps
48 Institution of Oceanography (SIO) in the USA¹⁰, the National institute for Environmental
49 Studies (NIES) in Japan²¹, and the Max Planck Institute for Biogeochemistry (MPI-BGC) in
50 Germany²². These laboratories report atmospheric CO₂ mole fractions and O₂/N₂ ratios from
51 21 fixed stations, as well as an extensive network of ship observations in the Pacific operated
52 by NIES (Figure 1). Further details of the data sources and the stations are in Table 1 in the
53 Methods section.

54 Oxygen measurements are reported as $\delta(\text{O}_2/\text{N}_2)$ ratios, in units of per meg¹¹. Since dry air



55 consists of about 21% oxygen, a change of one micromole of oxygen per mole of air is
56 equivalent to about $(0.21)^{-1}$ or 4.8 per meg. The different laboratories use distinct calibration
57 scales, and we have therefore adjusted all O₂ data to the SIO2017 O₂ scale (see Methods).
58 With these adjustments, we calculated a δAPO value for each observation of XCO₂, the CO₂
59 mole fraction, and $\delta(\text{O}_2/\text{N}_2)$ according to the equation of Manning and Keeling¹⁹:

60
$$\delta\text{APO}=\delta(\text{O}_2/\text{N}_2) + \frac{\alpha_B}{X\text{O}_2} (X\text{CO}_2 - 350) \quad (\text{i})$$

61 Here, α_B is the average molar ratio of O_2 released to CO_2 absorbed during land
62 photosynthesis, assumed to be the same as the ratio of CO_2 released to oxygen absorbed in
63 terrestrial respiration, this parameter is taken to be 1.1 +/- 0.05, and $X\text{O}_2 = 0.209436$, the
64 approximate molar ratio of O_2 in dry air, inherent to the SIO2017 calibration scale. 350 is an
65 arbitrary value of the SIO scale.

66 The δAPO and CO_2 flask data records for individual stations, and the ship data after grouping
67 into ten-degree latitude bands, were curve-fitted as described in the Methods section to
68 obtain representations of the seasonal cycles and the de-seasonalized rate of change at each
69 site. The procedure included a low-pass filter with a gaussian kernel having a 40-day half-
70 width, short enough to retain the inter-annual variations in the rate of change.

71 **Global rate of decline of atmospheric potential oxygen**

72 δAPO is defined to be invariant to the land sink. It is however declining in the atmosphere
73 due to fossil fuel combustion and the removal of CO_2 by the ocean without a corresponding
74 release of oxygen. We therefore seek to define a rate of change of δAPO appropriate to the
75 atmosphere as a whole.

76 Figure 2 shows rates of change of δAPO calculated over five-year periods from 1990 to 2020
77 from each of the stations and latitudinal ship averages. For each station the rates of change
78 were constructed by first calculating annual averages by integrating the smoothed curve fits
79 from the start to the end of each year, then fitting a continuous, piece-wise linear function
80 with hinge points at 1995.0, 2000.0, ... 2015.0 to those annual averages (see Methods).

81 Values are only shown for stations where an annual average could be calculated for every
82 year in a given 5-year interval. Rates of decline increase with time, and are somewhat
83 location dependent, but for periods after 2000, the agreement between the stations is quite
84 good. The standard deviation of the stations for each 5 year period after 2000 ranges from
85 0.45-0.60 per meg per year, which is only about 5% of the rate of decline.

86 To construct a trend for the atmosphere as a whole, we created a representation for δ APO in
 87 the marine boundary layer as a function of latitude and time. For this we used only the
 88 stations near the ocean and near sea level (see Table I in Methods), except for the South
 89 Pole. The smoothed curves of the individual stations and latitudinal ship averages were
 90 sampled at regular intervals of 0.1 years and 4th-order polynomial functions of the sine of
 91 latitude fitted to these samples at each time slice. The result is a “flying carpet” plot as a
 92 function of latitude and time (Figure 3). This 3-D surface was then integrated with latitude
 93 over the area of the Earth’s surface, to give a representative time series for global mean
 94 surface δ APO.

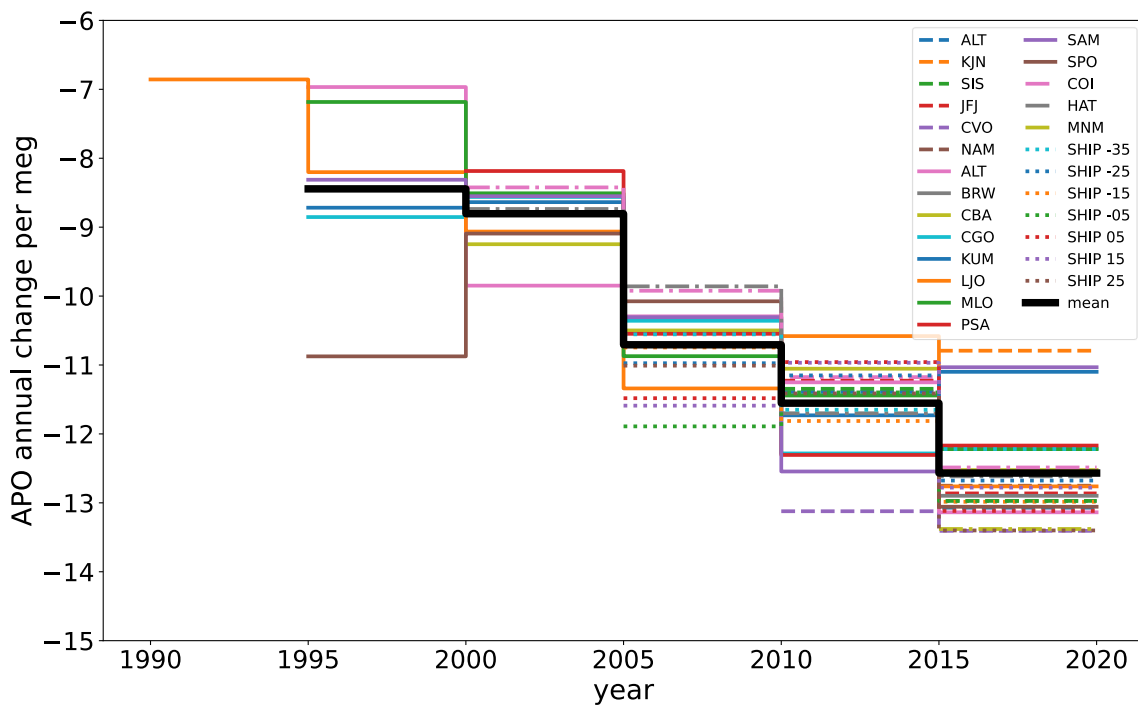


Figure 2: Rates of change of δ APO for individual stations and latitudinally averaged ship data, averaged over five-year intervals 1990-1995, 1995-2000, etc up to 2015-2020. Individual stations are distinguished by different colours and identified by their three-letter codes. SIO stations are solid lines, MPI-BGC stations are dashed lines and NIES stations are dot-dash lines. The latitudinally binned NIES ship data are dotted lines, identified by the central latitude between 25°N and 35°S. The black line is an unweighted mean of all stations and ship data in each 5-year period.

95

96

97 **Ocean oxygen outgassing**

98 An estimate of this flux is needed in order that the atmospheric observations can be used to
99 quantify the ocean and land CO₂ sinks. Deoxygenation of the ocean is driven by several
100 mechanisms related to climate warming and global change²³. The solubility of O₂ in water
101 decreases with increasing temperature, which will drive a direct loss of O₂ from surface
102 waters to the atmosphere due to ocean warming. In addition, the ventilation of deeper
103 water is expected to reduce as the oceans become more vertically stratified, resulting in less
104 uptake of oxygen in temperate and cold regions. It has also been shown that some plankton
105 communities are stimulated to fix more carbon as CO₂ partial pressure rises²⁴. These
106 processes lead to reduced uptake or greater release of oxygen in surface waters, with
107 corresponding depletion at depth. Over recent decades, the uptake and release of oxygen by

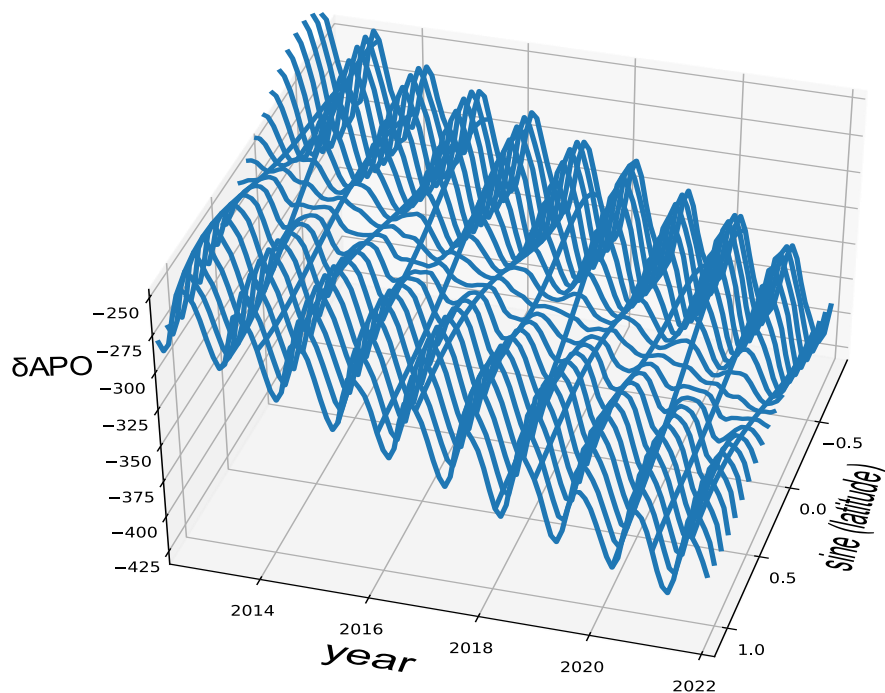


Figure 3: Wireframe representation of δAPO (in per meg) in the marine boundary layer as a function of time and latitude. The northern and southern hemispheres are dominated by seasonal cycles with maxima in the summer, which inter-leave in the equatorial regions. Only the last 12 years are shown, to better show the structure.

108 the ocean are expected therefore to be significantly out of balance with a net source to the
109 atmosphere.

110 Keeling and Garcia²⁵ used the relationship between ocean heat uptake and degassing of
111 oxygen as a proxy for O₂ outgassing. Ito et al²⁶ confirmed that the rate of loss of O₂ in the top
112 1000 m of the ocean is closely tied to ocean heat uptake, deriving a linear relationship of 8.2
113 +/- 0.7 nmol O₂ per joule of heat gained. The surprisingly tight correlation suggests that this
114 O₂ loss is closely related to interaction with the atmosphere, so it is likely that this oxygen is
115 escaping at the surface into the atmosphere. We used their result to quantify the O₂ flux out
116 of the ocean from heat uptake data. Annual estimates of heat uptake averaged over 5 yearly
117 periods presented in Figure 2 were calculated from the Levitus World Ocean Atlas (WOA)
118 estimates²⁷: <https://www.ncei.noaa.gov/data/oceans/>.

119 The rate of loss of upper ocean O₂ calculated by this method for the years 1990 to 2020 is 65
120 Tmol yr⁻¹, with year-to-year standard deviation of 33 Tmol yr⁻¹. For comparison, the decline
121 of global ocean O₂ from 2004-2022 found in the recently published GOBAI data product for
122 marine oxygen concentrations is 45±5 Tmol yr⁻¹ over the upper 1000m, or 84±2 Tmol yr⁻¹
123 over the upper 2000m²⁸ (calculated from Table B5 of that publication). These are estimates
124 of O₂ deoxygenation in the ocean interior, which is not necessarily equivalent to outgassing
125 to the atmosphere, however the global O₂ flux out of the ocean cannot at present be reliably
126 calculated using surface observations alone. An estimate is available from ECCO-Darwin, a
127 state estimation of ocean physics and biogeochemistry which assimilates both physics and
128 biogeochemical data, which gives a value of 62 Tmol yr⁻¹ for the period 1994-2017²⁸.

129 To reflect the large uncertainty on the ocean outgassing term we assigned a 1-σ value of 25
130 Tmol yr⁻¹ to this flux, implying that with ~95 % confidence (i.e. 2-σ) the flux lies between 15
131 and 115 Tmol yr⁻¹. This is the largest uncertainty in the calculation of the ocean sink (Table S3
132 in the supplementary information gives contributions of all the uncertainties considered).
133 Uncertainty on the land sink is larger than that of the ocean sink because it is more sensitive
134 to errors in the total amount of fossil fuel CO₂ emitted.

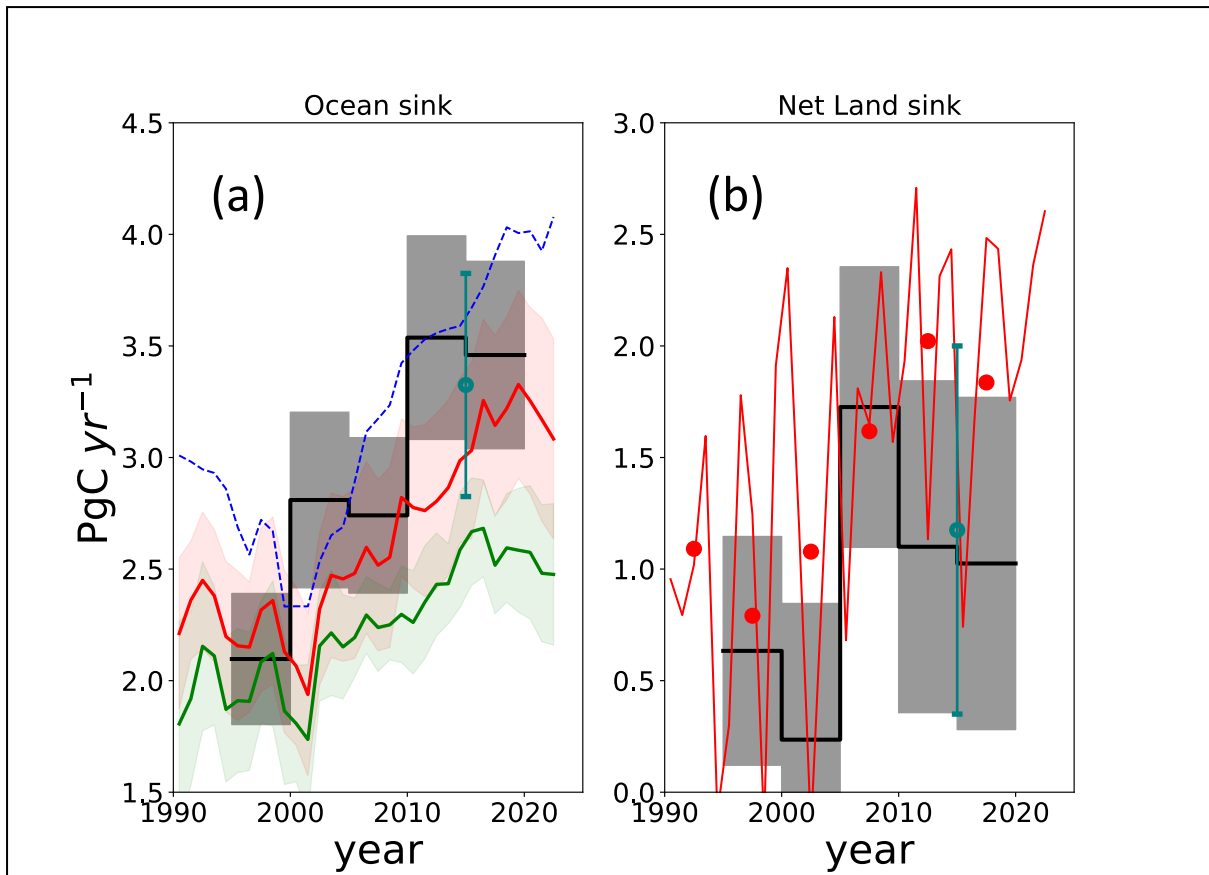


Figure 4: (a) Ocean and (b) net land carbon sinks (black lines) calculated from δ APO and CO_2 trends in the atmosphere averaged over 5-year periods, with $1-\sigma$ uncertainties (grey bands). The data sources, and equations from Manning and Keeling, are given in the Methods. Coloured lines: sink estimates from the Global Carbon Project budgets¹: For the ocean sink, green is the mean of GCP ocean models, red is the mean of surface ocean CO_2 data products, with colour shading showing the standard deviations of the individual submissions. The blue dashed line is a data product (not included with the others) using surface ocean data corrected for skin temperature variations in the very near surface⁹. For the land sink the GCP estimate (red lines) and its 5-year average (red dots) are shown. The teal-coloured data point on each plot is the estimate from the 2023 global carbon budget for the 2010-2020 decade¹, using the O_2/N_2 method based on three stations only.

135

136 **CO₂ sink estimates**

137 Figures 4a and b show our best estimates for the ocean and net land sinks for atmospheric
 138 CO_2 , respectively, calculated using the marine boundary layer δ APO product shown in Figure
 139 3, integrated annually and over the Earth's surface. The sinks have been calculated as means
 140 for each five-year period between 1995 and 2020: numerical values are given in the
 141 Supplementary information. The grey shading shows $1-\sigma$ uncertainties calculated from 100
 142 Monte-Carlo runs using randomly chosen parameters from gaussian uncertainty

143 distributions (see Methods). Since any definition for the global atmospheric average
144 concentration derived from a set of individual stations is open to question, we also repeated
145 the calculations with the change in δAPO taken as the mean of the stations shown in Figure
146 2, and the standard deviation of the stations as the uncertainty (see Figure S1 in the
147 supplementary Information). This deliberately simplistic approach makes no attempt to
148 weight the stations to represent regions of the atmosphere. However, the results are similar
149 to those of Figure 4, and within $1\text{-}\sigma$ bounds. This suggests that the method used to combine
150 the station data to produce a global estimate of APO is not crucial to the results.

151 Also shown in Figure 4 are sink estimates from the GCP¹. For the ocean sink we show values
152 from the average of ocean models, the average of surface ocean data products, and a data
153 product (not included in that average) that accounts for near-surface skin effect temperature
154 variations, which significantly increases the flux. The GCP net land sink estimates are highly
155 variable from year to year, (red line) so for better comparison we also show 5-year averages
156 (red dots). The most recent global carbon budget includes an estimate for the two sinks by
157 the O_2/N_2 method for the decade of 2010-20, also shown in the figure, which however uses
158 only three stations as distinct from the 25 records used here; nevertheless, the estimate for
159 this decade is similar to ours.

160 The O_2/N_2 method suggests a rapidly increasing ocean sink for the period from 1995 to
161 2015. It suggests a value consistent with the upper end of the data products, indeed, it most
162 closely matches the data product that takes into account skin temperature effects and is
163 substantially larger than the others. By contrast, it does not match the ocean model
164 estimates well. However, the O_2/N_2 method suggests the rapid increase did not continue
165 after 2015, unlike the data products (which continue to increase throughout the decade of
166 the 2010s) and more in accord with the trend of model estimates for that period.

167 **Conclusions**

168 The good number of stations with O_2/N_2 observations now available give a coherent view of
169 the rate of change of oxygen in the atmosphere, which translates comparatively directly into
170 estimates of the ocean sink for anthropogenic CO_2 . Uncertainties are on the order of 0.45
171 PgC yr^{-1} ($1\text{-}\sigma$) when integrated over five year periods, with errors in estimating the
172 outgassing of oxygen from the ocean making the largest contribution. The values obtained

173 for the ocean sink are broadly consistent with observational approaches used by the Global
174 Carbon Project and by their models before 2000 (which is not surprising, as models were
175 originally adjusted³ to be consistent with estimates of the ocean sink²⁹ compatible with
176 results from the O₂/N₂ method at that time¹⁹. However, our results indicate a more rapid
177 increase of the ocean sink than do models through the period up to 2015, with the ocean
178 sink in recent years substantially greater than models suggest. Data products based on the
179 gas exchange equation and surface ocean CO₂ observations are more consistent with the
180 O₂/N₂ results up to 2015, with the O₂/N₂ method supporting values towards the higher end
181 of the values given by those estimates.

182 The period of 2000-2015 includes the densest coverage of surface ocean carbon
183 observations³⁰, during which we might expect calculations of the CO₂ ocean sink based on
184 observations (as in the data products shown in Figure 4a) to be at their most reliable - before
185 that time coverage is quite limited. This is also the period when the rate of increase of the
186 ocean sink in the data products agrees best with the O₂/N₂ method. Since 2015, the
187 coverage of these primary data has declined substantially. The pCO₂ data products are
188 reliant on the use of machine learning techniques to interpolate the carbon observations to
189 assist in filling in the always-sparse primary carbon data. Poor data coverage over extended
190 periods is however not readily visible in the final data products, and there is a concern that
191 they may tend to extrapolate past well-observed changes into the less-well-observed near-
192 past and present. This might help to explain at least some of the recent discrepancy in
193 trends diagnosed from $\delta(\text{O}_2/\text{N}_2)$ and surface pCO₂ observations. To help resolve the ongoing
194 uncertainties about the fate of anthropogenic CO₂ these long-term observation
195 programmes, both in the atmosphere and oceans, need to be supported and continued.

196 **Methods**

197 **Sources of O₂/N₂ and CO₂ flask data**

198 The earliest observations were begun by SIO in 1989, and through most of the 1990s all the
199 available O₂/N₂ observations we use are from their network only. The number of
200 observations increased greatly by 2000 with the commencement of the NIES program in the
201 Pacific, while in the 2000s they were joined by observations in the Atlantic sector made by
202 MPI-BGC.

203 Bi-monthly averaged data for $\delta(\text{O}_2/\text{N}_2)$ and XCO₂ from flasks, from the network operated by
204 Scripps Institution^{10,19} were downloaded from the UCSD site in May 2023. The archive 2022-
205 12-02 was used: <https://library.ucsd.edu/dc/object/bb56403139>.

206 Flask data for $\delta(\text{O}_2/\text{N}_2)$ and XCO₂ from the MPI-BGC at Jena, Germany²² were obtained from
207 the archive maintained by MPI-BGC: <https://doi.org/10.17617/3.ZQX1LI>. We used the values
208 calculated after correcting O₂/N₂ ratios for micro-leaks during flask storage, using the
209 observed Ar/N₂ ratios²².

210 Flask data for $\delta(\text{O}_2/\text{N}_2)$ and XCO₂ for the stations and ship-based sampling network operated
211 by the National Institute for Environmental Studies^{12,21} was downloaded from the Global
212 Environmental Database, <https://db.cger.nies.go.jp/ged/en>, in September 2023.

213 δAPO was calculated for each record using equation (i) in the text.

214 The $\delta(\text{O}_2/\text{N}_2)$ and XCO₂ records, their locations and date of the start of the record are shown
215 in Table I. The records all continue until January 2020 or beyond.

216 **Table 1: Stations having records of $\delta(\text{O}_2/\text{N}_2)$ and XCO₂ in flask samples of air, and their use**
217 **in our calculations.**

Institute	Station	ID	Lat(N)	Lon(E)	Record starts (decimal year)	Used for MBL product (fig3)?	Used for fig 2?
MPI-BGC	Alert	ALT	82.45	-62.51	2005.3	Yes	Yes
MPI-BGC	Kjølnes	KJN	70.85	29.22	2014.8	No	Yes
MPI-BGC	Zotto	ZOT	60.8	89.35	2006.8	No	No
MPI-BGC	Shetland	SIS	59.85	-1.28	2005.3	Yes	Yes
MPI-BGC	Białystok	BIK	52.23	23.03	2005.9	No	No

MPI-BGC	Ochsenkopf	OXK	50.03	11.81	2007.1	No	No
MPI-BGC	Jungfrauoch	JFJ	46.55	7.98	2008.0	No	Yes
MPI-BGC	Cape Verde	CVO	16.85	-24.69	2007.3	Yes	Yes
MPI-BGC	Namibia	NAM	-23.56	15.05	2013.6	Yes	Yes
Scripps	Alert	ALT	82.45	-62.51	1989.9	Yes	Yes
Scripps	Barrow	BRW	71	-156	2010.9	Yes	Yes
Scripps	Cold Bay	CBA	55	-163	1995.6	Yes	Yes
Scripps	Cape Kumukahi	KUM	20	-155	1993.5	Yes	Yes
Scripps	La Jolla	LJO	33	-117	1989.4	Yes	Yes
Scripps	Mauna Loa	MLO	20	-156	1991.0	No	Yes
Scripps	American Samoa	SMO	-14	-170	1993.5	Yes	Yes
Scripps	Cape Grim	CGO	-41	150	1991.0	Yes	Yes
Scripps	Palmer Station	PSA	-65	-64	1996.7	Yes	Yes
Scripps	South Pole	SPO	-90	0	1991.9	Yes	Yes
NIES	Cape Ochiishi	COI	43.2	145.5	1998.9	Yes	Yes
NIES	Hateruma Island	HAT	24.1	123.8	1997.6	Yes	Yes
NIES	Minamitorishima	MNM	24.3	154.0	2011.8	Yes	Yes
NIES	Ship data		-35 – 45	130-240	2001.8	Yes	Yes

218

219 **Adjustments to the O₂/N₂ scales**

220 NIES data and MPI-BGC data were adjusted to bring them onto the SIO 2017 scale, as
221 follows:

222 **NIES O₂/N₂ data:** A comprehensive intercomparison study by Aoki et al³¹ of O₂/N₂
223 calibrations determined linear relationships between SIO and NIES scales. We used the data
224 they provide (their table 4) to adjust the NIES data also to the SIO scale. In addition, we
225 applied a drift in the offset of 0.28 per meg per year for the period since March 2010, which
226 Rödenbeck et al³² report is obtained by a linear fit to differences between regular
227 simultaneous samples collected at La Jolla, San Diego, since that time.

228 **MPI-BGC O₂/N₂ data:** We used the O₂/N₂ data after “Argon correction” as reported by
229 Heimann et al²². MPI-BGC data are calibrated to a scale referenced to the version of the SIO
230 scale current in 2007. Figure S2 in the supplementary information shows a comparison of
231 the flask records from Alert, during the period since 2005 when both SIO and MPI-BGC
232 report flask observations there. There is good agreement but a systematic offset, which from
233 2005 to early 2015 and after early 2020 averages about 6.9 per meg, but in the intervening
234 time is higher, about 13.4 per meg. Rödenbeck et al¹ provide a discussion of these offsets
235 and conclude that a micro-leak in the measurement system caused the enhanced offset

236 between 2015 and 2020. To align the MPI-BGC data with the current SIO scale, we applied
237 the larger offset between March 2015 and May 2020, and the smaller offset to the
238 remainder of the data.

239 **Curve fitting the δAPO time series of individual stations**

240 Curve fitting to the δAPO time series was performed on the fixed stations and also on the
241 ship-based observations after grouping into ten-degree latitude bands. Using least squares
242 regression, a curve with three harmonics to represent the seasonal cycle and a cubic
243 polynomial to represent the annual trend was fitted to the δAPO_{data} records as follows:

$$244 \quad \delta APO_{model} = a_0 + \sum_{i=1}^3 a_i t^i + \sum_{f=1}^3 (b_f \sin(2\pi f t) + c_f \cos(2\pi f t)) \quad (ii)$$

245 Where a_0, a_i, b_f and c_f are constants found by the fitting procedure and t is time in years.
246 The residuals of the data from the model ($\delta APO_{data} - \delta APO_{model}$) were input to a low pass
247 filter with a Gaussian kernel and a 40-day half-width at half height. Finally, this residual curve
248 was added to the δAPO_{model} to produce a smoothed record that captures the lower
249 frequency seasonal and annual variability of the data. For gaps in the time series longer than
250 200 days, a gap was left in the smoothed record, except in the case of the LJO record which
251 has a substantial gap in 1990. This was filled by linear interpolation, since no other station
252 was operating at that time. Figures S4-S29 show the 19 station time series and seven
253 latitude-averaged groups of ship observations with the smoothed curve fits.

254 **Other data sources**

255 The Excel spread sheet linked to the Global Carbon Budget 2023¹ was the source of data on
256 the annual release of CO₂ from coal, oil, natural gas and cement-production since 1990, and
257 the burden of CO₂ in the atmosphere. We used values for the O₂/CO₂ stoichiometry of these
258 processes as quoted by Keeling and Manning¹⁰.

259 **Calculation of sinks**

260 We used the equations for the ocean sink (O) and the net land sink (L) given by Manning and
261 Keeling¹⁹, with the addition of a small correction term due to oxygen sources from industrial
262 processes as recently compiled by Battle et al³³. The equations are:

$$263 \quad O = [(-\Delta(\delta APO) \times M_{atm} \times X_{O_2} \times m_c) + \alpha_B \times F - FOD + (Z_{eff} + I) \times m_c] \times 1/\alpha_B$$

264 And

$$265 \quad L = F - O - \Delta XCO_2 \times M_{atm} \times m_c$$

266 Where

$$267 \quad Z_{eff} = Z - \Delta N_2 \times \frac{X_{N_2}}{X_{O_2}}$$

$$268 \quad Z = H \times \frac{d[O_2]}{dT}$$

$$269 \quad \Delta N_2 = H \times \frac{dS_{N_2}}{dT} \times \frac{1}{c_p}$$

270 Definitions are:

271 F = fossil fuel source in gC yr⁻¹.

272 FOD = oxygen demand of the fossil fuel source in moles per year multiplied by m_c (see
273 below: also, effectively therefore in gC yr⁻¹).

274 $\Delta(\delta APO)$ is the change in δAPO in mole fraction per year.

275 ΔXCO_2 is the change in atmospheric CO₂ in mole fraction per year.

276 M_{atm} is the number of moles in the atmosphere.

277 m_c is the molecular weight of carbon (equals 12).

278 I is the number of moles of O₂ released by industry in a year³³.

279 Z is the moles of O₂ released from the ocean due to heating in a year. ΔN_2 is the moles of
280 nitrogen released by heating in a year. Z_{eff} is the molar correction to the calculated ocean
281 sink for these releases. Z is calculated as the product of H , the yearly uptake of heat by the
282 ocean, and $\frac{dO_2}{dT}$, the rate of change of oxygen concentration in seawater with heat input (see
283 discussion in text) while ΔN_2 is calculated as the product of $\frac{dS_{N_2}}{dT}$, the rate of change of N₂
284 solubility with temperature, and $1/c_p$ where c_p is the specific heat capacity of seawater. Table
285 S4 in the supplementary information gives five-year-average values for F , FOD , Z_{eff} and I .

286 Constants are given in the table below.

Constant	Value
M_{atm}	1.77×10^{20} mol
$\frac{dS_{N_2}}{dT}$	1.03×10^{-5} mol l ⁻¹ K ⁻¹ (at 288.15 K)
$\frac{d[O_2]}{dH}$	8.2×10^{-9} mol J ⁻¹
X_{O_2}	0.20943 6
X_{N_2}	0.78084
C_p	4108 Joules l ⁻¹ K ⁻¹

287

288 **Smoothed model fits to Individual station records**

289 The supplementary information file contains plots of δ APO for all the stations and
290 latitudinally binned ship-based observations used in this study.

291 **Sources of uncertainty**

292 The Monte Carlo results in figure 4 include five sources of uncertainty. We assumed 1-sigma
293 values as follows: (1) a 5% margin of error for the rate of fossil fuel emissions: (2)
294 uncertainties in the ratio of O₂: consumed to CO₂ released by fossil fuel combustion, of 0.03-
295 0.04 mol mol⁻¹ for the major fuel types¹⁰; (3) a 5% uncertainty for α_B , the molar ratio of O₂
296 released to CO₂ taken up in land photosynthesis; (4) the uncertainty in the rate of decline of
297 global δ APO, calculated from the fitting procedure when using the global surface product, or
298 the standard deviation of individual stations when using the mean of stations; (5) the
299 uncertainty in ocean outgassing of O₂ of 25 Tmol yr⁻¹ as discussed in the main text.

300 The individual contributions of each these sources of uncertainty to the calculation of sinks,
301 are given in Table S3 in the supplementary information. For the ocean CO₂ sink as shown in
302 Figure 4a, ocean outgassing is the largest term, and error in the O:C ratio for combustion of
303 coal, oil and gas fossil fuels is the next largest. The other terms have comparatively small
304 influences on the ocean sink, but the fossil fuel uncertainty has a much larger effect on the
305 calculation of the land sink. Overall the terms add in quadrature giving values of \sim 0.45 PgC
306 yr⁻¹ for the ocean sink and \sim 0.75 PgC yr⁻¹ for the land sink.

307 **Data and code availability**

308 The atmospheric O₂/N₂ and CO₂ flask data are available from the data archives of the three
309 primary data providers (SIO, NIES and MPI-BGC) at the URLs given above in the Methods
310 Section. Data from the Global Carbon Budget is available from spreadsheets linked to those
311 publications, at the ICOS Carbon Portal: [https://www.icos-cp.eu/science-and-impact/global-](https://www.icos-cp.eu/science-and-impact/global-carbon-budget/)
312 [carbon-budget/](https://www.icos-cp.eu/science-and-impact/global-carbon-budget/).

313 The python code used to process the data is available on request from the authors at U.
314 Exeter.

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325 **Author information**

326 Contributions: Design by EK, AJW and ACM. Writing by EK and AJW with contributions from
327 all authors. Data contributions from RFK, YT, MH, and SZ. Calculations and coding by AJW
328 and EK.

329

330 **References**

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- 332 1 Friedlingstein, P. *et al.* Global Carbon Budget 2023. *Earth Syst. Sci. Data* **15**, 5301-5369
333 (2023). <https://doi.org:10.5194/essd-15-5301-2023>
- 334 2 DeVries, T. *et al.* Magnitude, Trends, and Variability of the Global Ocean Carbon Sink From
335 1985 to 2018. *Global Biogeochemical Cycles* **37** (2023).
336 <https://doi.org:10.1029/2023GB007780>
- 337 3 Le Quéré, C. *et al.* Trends in the sources and sinks of carbon dioxide. *Nature Geoscience* **2**,
338 831-836 (2009). <https://doi.org:10.1038/ngeo689>
- 339 4 Wanninkhof, R. *et al.* Global ocean carbon uptake: Magnitude, variability and trends.
340 *Biogeosciences* **10**, 1983-2000 (2013). <https://doi.org:10.5194/bg-10-1983-2013>
- 341 5 Landschützer, P. *et al.* A neural network-based estimate of the seasonal to inter-annual
342 variability of the Atlantic Ocean carbon sink. *Biogeosciences* **10**, 7793-7815 (2013).
343 <https://doi.org:10.5194/bg-10-7793-2013>

344 6 Gloege, L., Yan, M., Zheng, T. & McKinley, G. A. Improved Quantification of Ocean Carbon
345 Uptake by Using Machine Learning to Merge Global Models and pCO₂ Data. *Journal of*
346 *Advances in Modeling Earth Systems* **14** (2022). <https://doi.org:10.1029/2021MS002620>
347 7 Gregor, L. & Gruber, N. OceanSODA-ETHZ: A global gridded data set of the surface ocean
348 carbonate system for seasonal to decadal studies of ocean acidification. *Earth System Science*
349 *Data* **13**, 777-808 (2021). <https://doi.org:10.5194/essd-13-777-2021>
350 8 Rödenbeck, C., Devries, T., Hauck, J., Le Quéré, C. & Keeling, R. F. Data-based estimates of
351 interannual sea-Air CO₂ flux variations 1957-2020 and their relation to environmental
352 drivers. *Biogeosciences* **19**, 2627-2652 (2022). <https://doi.org:10.5194/bg-19-2627-2022>
353 9 Watson, A. J. *et al.* Revised estimates of ocean-atmosphere CO₂ flux are consistent with
354 ocean carbon inventory. *Nature Communications* **11** (2020). [https://doi.org:10.1038/s41467-](https://doi.org:10.1038/s41467-020-18203-3)
355 [020-18203-3](https://doi.org:10.1038/s41467-020-18203-3)
356 10 Keeling, R. F. & Manning, A. C. in *Treatise on Geochemistry: Second Edition* Vol. 5 385-404
357 (2014).
358 11 Keeling, R. F. & Shertz, S. R. Seasonal and Interannual Variations in Atmospheric Oxygen and
359 Implications for the Global Carbon-Cycle. *Nature* **358**, 723-727 (1992).
360 12 Tohjima, Y., Mukai, H., MacHida, T., Hoshina, Y. & Nakaoka, S. I. Global carbon budgets
361 estimated from atmospheric O₂N₂ and CO₂ observations in the western Pacific region over a
362 15-year period. *Atmospheric Chemistry and Physics* **19**, 9269-9285 (2019).
363 <https://doi.org:10.5194/acp-19-9269-2019>
364 13 2021, I. Climate Change 2021: The Physical Science Basis. Contribution of Working Group I to
365 the Sixth Assessment Report of the Intergovernmental Panel on Climate Change 2391 (2021).
366 14 Andrew, R. M. A comparison of estimates of global carbon dioxide emissions from fossil
367 carbon sources. *Earth System Science Data* **12**, 1437-1465 (2020).
368 <https://doi.org:10.5194/essd-12-1437-2020>
369 15 Keeling, C. D., Bacastow, R. B. & Bainbridge, A. E. Atmospheric carbon dioxide variations at
370 Mauna Loa Observatory, Hawaii. *TELLUS* **28**, 538-551 (1976).
371 <https://doi.org:10.3402/tellusa.v28i6.11322>
372 16 Hauck, J. *et al.* Consistency and Challenges in the Ocean Carbon Sink Estimate for the Global
373 Carbon Budget. *Frontiers in Marine Science* **7** (2020).
374 <https://doi.org:10.3389/fmars.2020.571720>
375 17 Gray, A. R. *et al.* Autonomous Biogeochemical Floats Detect Significant Carbon Dioxide
376 Outgassing in the High-Latitude Southern Ocean. *Geophysical Research Letters* **45**, 9049-
377 9057 (2018). <https://doi.org:10.1029/2018GL078013>
378 18 Long, M. C. *et al.* Strong Southern Ocean carbon uptake evident in airborne observations.
379 *Science* **374**, 1275-1280 (2021). <https://doi.org:10.1126/science.abi4355>
380 19 Manning, A. C. & Keeling, R. F. Global oceanic and land biotic carbon sinks from the Scripps
381 atmospheric oxygen flask sampling network. *Tellus Series B-Chemical and Physical*
382 *Meteorology* **58**, 95-116 (2006).
383 20 Stephens, B. B. *et al.* Testing global ocean carbon cycle models using measurements of
384 atmospheric O₂ and CO₂ concentration. *Global Biogeochemical Cycles* **12**, 213-230 (1998).
385 <https://doi.org:10.1029/97GB03500>
386 21 Tohjima, Y. *et al.* Analysis of seasonality and annual mean distribution of atmospheric
387 potential oxygen (APO) in the Pacific region. *Global Biogeochemical Cycles* **26** (2012).
388 <https://doi.org:10.1029/2011GB004110>
389 22 Heimann, M. J., A., Brand, W., Lavric, J., Moossen, H., and Rothe, M., . The atmospheric flask
390 sampling program of MPI-BGC, Version 13, 2022. (Max Planck Insitute for Biogeochemistry,
391 Jena, Germany, 2022).
392 23 Oschlies, A., Brandt, P., Stramma, L. & Schmidtko, S. Drivers and mechanisms of ocean
393 deoxygenation. *Nature Geoscience* **11**, 467-473 (2018). [https://doi.org:10.1038/s41561-018-](https://doi.org:10.1038/s41561-018-0152-2)
394 [0152-2](https://doi.org:10.1038/s41561-018-0152-2)

395 24 Riebesell, U. *et al.* Enhanced biological carbon consumption in a high CO₂ ocean. *Nature* **450**,
396 545-548 (2007). <https://doi.org:10.1038/nature06267>

397 25 Keeling, R. F. & Garcia, H. E. The change in oceanic O₂ inventory associated with recent
398 global warming. *Proceedings of the National Academy of Sciences of the United States of*
399 *America* **99**, 7848-7853 (2002). <https://doi.org:10.1073/pnas.122154899>

400 26 Ito, T., Minobe, S., Long, M. C. & Deutsch, C. Upper ocean O₂ trends: 1958–2015.
401 *Geophysical Research Letters* **44**, 4214-4223 (2017). <https://doi.org:10.1002/2017GL073613>

402 27 Levitus, S. *et al.* World ocean heat content and thermosteric sea level change (0-2000m),
403 1955-2010. *Geophysical Research Letters* **39** (2012). <https://doi.org:10.1029/2012GL051106>

404 28 Carroll, D. *et al.* The ECCO-Darwin Data-Assimilative Global Ocean Biogeochemistry Model:
405 Estimates of Seasonal to Multidecadal Surface Ocean pCO₂ and Air-Sea CO₂ Flux. *Journal of*
406 *Advances in Modeling Earth Systems* **12** (2020). <https://doi.org:10.1029/2019MS001888>

407 29 Denman, K. *et al.* Chapter 7, Climate Change 2007, : The physical Science Basis. Contribution
408 of Working Group 1 to the fourth assessment report of the Intergovernmental Panell on
409 Climate Change. 499-587 (2007).

410 30 Bakker, D. C. E. *et al.* A multi-decade record of high-quality fCO₂ data in version 3 of the
411 Surface Ocean CO₂ Atlas (SOCAT). *Earth System Science Data* **8**, 383-413 (2016).
412 <https://doi.org:10.5194/essd-8-383-2016>

413 31 Aoki, N. *et al.* Intercomparison of O₂/N₂ratio scales among AIST, NIES, TU, and SIO based on
414 a round-robin exercise using gravimetric standard mixtures. *Atmospheric Measurement*
415 *Techniques* **14**, 6181-6193 (2021). <https://doi.org:10.5194/amt-14-6181-2021>

416 32 Rödenbeck, C. *et al.* The suitability of atmospheric oxygen measurements to constrain
417 western European fossil-fuel CO₂ emissions and their trends. *Atmospheric Chemistry and*
418 *Physics* **23**, 15767-15782 (2023). <https://doi.org:10.5194/acp-23-15767-2023>

419 33 Battle, M., Raynor, A., Kesler, S. & Keeling, R. The Impact of Industrial Activity on the Amount
420 of Atmospheric O₂. *Tellus B: Chemical and Physical Meteorology*, **75**, 65-75 (2023).
421 <https://doi.org:10.16993/tellusb.1875>

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