

1 **New insights of fCO<sub>2</sub> variability in the tropical eastern Pacific Ocean using**  
2 **SMOS SSS**

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50 **Abstract**

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Complex oceanic circulation and air-sea interaction make the eastern tropical Pacific Ocean (ETPO) a highly variable source of CO<sub>2</sub> to the atmosphere. Although the scientific community have amassed 70,000 surface fugacities of carbon dioxide (fCO<sub>2</sub>) datapoints within the ETPO region over the past 25 years, the spatial and temporal resolution of this dataset is insufficient to fully quantify the seasonal to inter-annual variability of the region, a region where fCO<sub>2</sub> has been observed to fluctuate by >300 μatm.

Upwelling and rainfall events dominate the surface physical and chemical characteristics of the ETPO, with both yielding unique signatures in sea surface temperature and salinity. Thus, we explore the potential of using a statistical description of fCO<sub>2</sub> within sea-surface salinity-temperature space. These SSS/SST relationships are based on in-situ SOCAT data collected within the ETPO. This statistical description is then applied to high resolution (0.25°) SMOS sea surface salinity and OSTIA sea surface temperature in order to compute regional fCO<sub>2</sub>. As a result, we are able to resolve fCO<sub>2</sub> at sufficiently high resolution to elucidate the influence various physical processes have on the fCO<sub>2</sub> of the surface ETPO.

Normalised (to 2014) oceanic fCO<sub>2</sub> between July 2010 and June 2014 within the entire ETPO was 39 (+/-10.7) μatm supersaturated with respect to 2014 atmospheric partial pressures, and featured a CO<sub>2</sub> outgassing of 1.51 (+/- 0.41) mmol m<sup>-2</sup> d<sup>-1</sup>. Values of fCO<sub>2</sub> within the ETPO were found to be broadly split between southeast and a northwest regions. The north west, central and offshore regions were supersaturated, with wintertime wind jet driven upwelling found to be the first order control on fCO<sub>2</sub> values. This contrasts with the southeastern/ Gulf of Panama region, where heavy rainfall combined with rapid stratification of the upper water-column act to dilute dissolved inorganic carbon, and yield fCO<sub>2</sub> values undersaturated with respect to atmospheric fugacities of CO<sub>2</sub>.

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## 88 **1. Introduction**

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90 Perturbations to the global carbon cycle caused by anthropogenically driven increases in  
91 atmospheric partial pressures of CO<sub>2</sub> (pCO<sub>2</sub>) produce an acute requirement to understand inter-  
92 reservoir carbon fluxes (Le Quéré et al 2014). However, assessing these fluxes, especially the flux  
93 between the atmosphere and surface ocean is challenging, as pCO<sub>2</sub> within the oceanic reservoir  
94 varies considerably both spatially and temporally. Fortunately, considerable effort has been made in  
95 recording oceanic pCO<sub>2</sub> (or fugacity- fCO<sub>2</sub>) over the past 50 years, with roughly 10 million  
96 individual measurements of global surface ocean pCO<sub>2</sub> taken, processed, flagged and assembled  
97 into two large datasets: the surface ocean CO<sub>2</sub> atlas (SOCAT) and Lamont-Doherty Earth  
98 Observatory (LDEO) carbon dioxide database (Bakker et al., 2014; Takahashi et al., 2014). Both  
99 databases make heavy use of 'vessel of opportunity' derived pCO<sub>2</sub> data, resulting in a  
100 heterogeneous dataset with the majority of measurements collected within the tight confines of  
101 commercial shipping lanes.

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103 The first stage in efforts to estimate large scale air sea fluxes requires the extrapolation of these  
104 discrete surface pCO<sub>2</sub> observations over large areas of surface ocean. Completing a basic  
105 extrapolation of this vessel of opportunity based data results in a spatially-patchy pCO<sub>2</sub> field, as  
106 seen in the one degree gridded product available from SOCAT (Bakker et al., 2014). Therefore, in  
107 order to achieve improved spatial coverage, a frequently used solution is to fit data-driven  
108 diagnostic models (e.g. Feely et al. 2006; Park et al. 2010; Rödenbeck et al., 2013). These models  
109 use observed correlations between physical properties and the pCO<sub>2</sub> observed under these  
110 conditions. In addition, statistical criteria based on the surface ocean observations (for example,  
111 satellite imagery) and/ or neural networks have been used to identify biogeochemical provinces in  
112 order to improve the accuracy of the extrapolated field (Boutin et al. 1999; Cosca et al. 2003;  
113 Rangama et al., 2005; Landschützer et al., 2014). Although the quality of these extrapolation  
114 methods have been refined over the past few years, in part due to the increasing number of in situ  
115 measurements, the interannual variability of the global air-sea CO<sub>2</sub> flux obtained using different  
116 data-driven methods still substantially differ, and further work is required to unify our estimates and  
117 improve understanding of this air-sea flux.

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119 The recent (2010-present) availability of sea surface salinity from the Soil Moisture Ocean

120 Salinity (SMOS) mission has provided a new tool in statistically modelled pCO<sub>2</sub> studies, as now  
121 the relationship between surface temperature, salinity, density and pCO<sub>2</sub> can be utilised in high  
122 resolution statistical descriptions of pCO<sub>2</sub> within TS space. It is this relationship that is explored  
123 within this paper in order to quantify inter and intra annual variability within the oceanographically  
124 complex eastern tropical Pacific Ocean (ETPO) region between 4 °N and 18 °N, and east of 95 °W  
125 (Fig. 1).

126 The ETPO region is influenced by northern and southern hemisphere trade winds, the  
127 doldrums, strong seasonal wind jets, heavy rainfall, strong solar heating and the El Niño- Southern  
128 Oscillation (ENSO), (Kessler, 2006). Within the ETPO the Intertropical Tropical Convergence Zone  
129 (ITCZ) is neither zonally oriented, nor spatially fixed over the course of a year, thus wind fields are  
130 highly variable in both strength and direction (Kessler, 2006). Further wind variability is  
131 introduced by low altitude jets blowing through three low elevation gaps in the Central American  
132 Cordillera, with these jets observed predominantly between November and February (Kessler 2002,  
133 Fig. 1-schematically represented by orange arrows). These strong jet winds are generated from the  
134 pressure gradient force resulting from high-pressure synoptic midlatitude weather systems transiting  
135 North America towards the low pressure Equatorial Pacific during the winter months (Chelton et al.,  
136 2001). Jet wind velocities of up to 20-30 ms<sup>-1</sup> have been observed within the three gulfs, extending  
137 from the shoreline to at least 500 km into the ETPO (Chelton et al., 2001). The alignment of these  
138 jets is mainly meridional in the Gulf of Panama and Tehuantepec, and more zonal in the Gulf of  
139 Papagayo (Fig. 1). These jet winds result in strong wintertime upwelling in each of the three basins,  
140 with the northerly jet winds in the Gulfs of Tehuantepec and Panama promoting Ekman upwelling  
141 (Kessler, 2006). At 9 °N, 90 °W, the quasi permanent anticyclonic Costa Rica thermocline dome is  
142 energised by the westerly jet winds in the Gulf of Papagayo, again, resulting in upwelling (Fig 1.  
143 Kessler, 2002). However, the pressure disequilibrium that spawns these jet winds builds and  
144 subsides quickly (on a sub-weekly timescale), resulting in a highly fluctuating jet wind field, and  
145 nonlinearities in the oceanic (and hence upwelling) response to these episodic events (Kessler,  
146 2006). Recent studies have shown that satellite SSS are very well suited to capture the variability of  
147 these high SSS events (Grotsky et al., 2014; Reul et al., 2013).

148 During summertime, these jet winds are rare, and the northward deflection of the ITCZ over  
149 the ETPO result in very high levels of precipitation, particularly within the Gulf of Panama (Fig. 1;  
150 Alory et al., 2012). The effect of this rainfall is a strong freshening and stratification of the surface,  
151 especially in the Gulf of Panama region. For this reason, the Gulf of Panama is often referred to as  
152 the 'Pacific freshpool', with salinities of <30 frequently observed (Alory et al., 2012).

## 153 2. Data and Methods

### 154 2.1 Observations of in-situ fCO<sub>2</sub> , DIC and atmospheric CO<sub>2</sub>

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156 We use SOCAT data as the basis of a statistical description of pCO<sub>2</sub> within the ETPO. All  
157 CO<sub>2</sub> data within the SOCAT database are recorded as fugacity of CO<sub>2</sub> (fCO<sub>2</sub>); a measure of pCO<sub>2</sub>  
158 that is corrected for the non-ideal-gas behaviour carbon dioxide displays (Bakker et al., 2014). In  
159 this study we will use fugacity of CO<sub>2</sub>, owing to the improved accuracy of this measure over pCO<sub>2</sub>.  
160 The logistical importance of the region has proven beneficial, producing a large database of fCO<sub>2</sub>/  
161 SSS/SST observations, with ~ 70,000 surface carbon, temperature and salinity data-points collected  
162 between 1991 and 2011 (within the region depicted in Fig. 1). We use data (converted from pCO<sub>2</sub> to  
163 fCO<sub>2</sub>) collected between 1991 and 2013 within the LDEO v2013 database as a semi-independent  
164 dataset to test our SOCAT based statistical description of fCO<sub>2</sub> within TS space (Takahashi et al.  
165 2014). The LDEO database extends two years longer than SOCAT, making it a useful evaluation  
166 product.

167 In addition to surface fCO<sub>2</sub>, we use dissolved inorganic carbon (DIC) concentrations within  
168 the ETPO measured during World Ocean Circulation Experiment cruise P19 in 1993 (Fig. 1, WOCE  
169 data available from <http://cchdo.ucsd.edu/>), and weekly dry-air CO<sub>2</sub> mole fractions (xCO<sub>2</sub>)  
170 measured at Fanning Island (3.5° N, 159° W, as part of the Scripps CO<sub>2</sub> program,  
171 <http://scrippsco2.ucsd.edu/>). We derive fCO<sub>2</sub> in the atmosphere according to:

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$$173 \quad f\text{CO}_2 \text{ air} = x\text{CO}_2 (p_{\text{atm}} - p_{\text{H}_2\text{O}}) \quad (1)$$

174

175 Where  $P_{\text{atm}}$  is the atmospheric pressure taken from the ERA interim product  
176 ([www.apps.ecmwf.int/datasets/](http://www.apps.ecmwf.int/datasets/)) and  $p_{\text{H}_2\text{O}}$ , the saturated water pressure (Weiss, 1974).

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### 178 2.2. Calculating air sea fluxes of CO<sub>2</sub>

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180 The air-sea flux of CO<sub>2</sub> ( $\text{mmol m}^{-2} \text{d}^{-1}$ ) is derived using the difference between surface  
181 ocean and atmospheric fCO<sub>2</sub> values, the solubility of CO<sub>2</sub> in standard seawater ( $\alpha \text{CO}_2$  calculated  
182 using the values given by Weiss 1974) and the gas transfer velocity  $k\text{CO}_2$  (cm/h):

$$183 \quad F = k\text{CO}_2 \alpha\text{CO}_2 (f\text{CO}_2 \text{ sea} - f\text{CO}_2 \text{ air}) \quad (2)$$

184 k is calculated using the 10 metre wind speed based parameterisation described by Sweeney et al.  
185 2007, where:

$$186 \quad k=0.27U^2(660/Sc)^{0.5} \quad (3)$$

187 where Sc is the Schmidt number for CO<sub>2</sub> (Wanninkhof, 1992). We use the 0.25°, daily resolution  
188 10 m wind speed product from the Advanced Scatterometer (ASCAT,  
189 [www.knmi.nl/scatterometer/](http://www.knmi.nl/scatterometer/)), and air fCO<sub>2</sub> is calculated assuming water vapour saturation at the  
190 boundary layer, then using the description in Eq. 1.

191 We note that when calculating air-sea gas exchange, it is not practical to derive a unique k  
192 parameter for all oceanic conditions and locations of interest, therefore, through necessity, k is  
193 calculated using the non-specific parameterisation of Sweeney et al. 2007. Influences on k, such as  
194 rainfall, have been observed to enhance gas transfer velocities in laboratory experiments (Liss and  
195 Johnson, 2014). However, applying the results from these laboratory experiments into a modified  
196 gas transfer velocity within the ETPO region is challenging, and requires further field experiments  
197 to validate. It is for this reason, and the role rainfall has in reducing DIC, (thus bringing ΔfCO<sub>2</sub>  
198 closer to atmospheric equilibrium and limiting the influence on air sea fluxes brought about by an  
199 enhanced gas transfer velocity), we do not attempt to correct for the influence rainfall has on air-  
200 sea exchanges.

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### 202 **2.3 Satellite observations of SST and SSS**

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204 Sea surface temperatures have been continuously measured from space via satellites since the  
205 AVHRR mission in 1981, and sea surface salinity has been measured since the launch of the  
206 European Space Agency's Soil Moisture and Ocean Salinity (SMOS) satellite mission in November  
207 2009 (Kerr et al. 2010). In this paper, we use satellite T and S data to help in the interpretation of  
208 spatial variability of fCO<sub>2</sub> from the ETPO.

209 SMOS SSS maps at 0.25° resolution (running average over 100x100km<sup>2</sup>) produced by the  
210 LOCEAN SMOS group (combining SMOS ascending and descending passes) were used within  
211 this paper (Boutin et al. 2013, data available at [www.cats.ifremer.fr/Products/Available-products-fromCEC\\_OS/Locean-v2013](http://www.cats.ifremer.fr/Products/Available-products-fromCEC_OS/Locean-v2013)). SMOS SSS data are not used prior to June 2010 due to variations in  
212 sensor configuration tested during the in-orbit sensor commissioning phase (Corbella et al., 2011).  
213 Analysing the accuracy of the SMOS product, using in-situ point observations of SSS can be  
214 completed using both SOCAT SSS data (where the salinity is derived from ship mounted  
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216 thermosalinograph-TSG), and Argo profiles made within the region (S Fig. 1). Binning all Argo  
217 and TSG near surface salinity data (that overlaps the SMOS observational period) into the same  
218 monthly, 0.25 degree structure as the SMOS monthly 0.25 ° product enables direct comparison. It is  
219 found that the RMSE between SMOS and these binned in-situ data is 0.24 psu, with a minimal  
220 offset (of 0.04 psu, S Fig. 1). This variability is a function of both the intrinsic variability of SSS  
221 within the EPTO (for example, caused by localised rainfall, riverine outflow, upwelling events that  
222 may occur within each SMOS observational pixel), and a measurement error made in the SMOS  
223 data. Estimates of this intrinsic variability within the tropics is 0.1 psu, but in certain regions (such  
224 as the western warmpool and ETPO) may feature variabilities of 0.4 psu or higher (Delcroix et al.  
225 2005). If the average intrinsic variability for the Tropical Pacific (of 0.1 psu) is used, the  
226 measurement noise within the SMOS product is 0.14 psu. This measurement noise is included in  
227 our fCO<sub>2</sub> error estimates (as detailed below). We suggest that this is a satisfactory noise to signal  
228 ratio, given that the region features SSS variability of upto 8 psu (Fig. 2).

229 SSS was combined with sea surface temperatures from the operational sea surface  
230 temperature and sea ice analysis (OSTIA) system; an optimal interpolation of multiple microwave  
231 and infrared satellite-based data sources (Donlon et al., 2012). This SST daily product features a  
232 native resolution of 0.05°, which was subsequently re-sampled over the same spatial and temporal  
233 grid as the SMOS data. As auxiliary datasets for near surface T and S, data from the Argo float  
234 array were used.

#### 235 **2.4 The basis of using a SST/SSS statistical fCO<sub>2</sub> model in the ETPO**

236

237 Using an fCO<sub>2</sub> model based on surface T and S properties requires the carbon properties of  
238 water masses to be quantified. We define three surface water-masses within the study region, based  
239 on the near-surface (0-100 m depth) T and S characteristics of the ETPO recorded by ~6000 Argo  
240 profiles between 17 °N and 4 °N, and east of 95 °W (Fig. 2). These water-masses are also  
241 independently featured within studies by Kessler (2002, 2006):

242 1)An ETPO surface water, with temperatures >27°C and salinities >33.5 (Fig 2. ‘B’).

243 2)Deep water that predominantly exists below the thermocline (defined by Kessler as the 20  
244 °C isotherm) and is only expressed at the surface of the ETPO during periods of upwelling  
245 (Fig 2. ‘C’).

246 3)Rain influenced water, with salinities <33.5 (Fig 2. ‘A’).

247 Here, the water-mass at 'C' (referred to as 'deep water' herein) is predominantly found at  
248 depths greater than 80 m, whilst the near surface water is typically ETPO surface water and warmer  
249 than 27.5 °C ('A' and 'B'). Occasionally, the deep water-mass can be observed at shallower depths,  
250 advected upwards during times of strong upwelling (Fig. 2). Rainfall influenced waters display a  
251 wide range of salinities (from 29 to 34.5) within the upper 20 m (as observed in the salinity  
252 variation between 'A' and 'B'). At low salinities, densities as low as 1018 kg m<sup>-3</sup> are observed; an  
253 eight kg m<sup>-3</sup> difference between this surface water and deep water (featuring densities of 1026 kg  
254 m<sup>3</sup>). This large density gradient is an important feature of the ETPO, as periods when stratification  
255 and high thermoclines strengths are prevalent, the upwelling of deep water is inhibited (Fiedler and  
256 Talley, 2006).

257 To identify the carbon properties of these water-masses, DIC concentrations between the  
258 surface and 100 m depth within the ETPO measured during WOCE transect P19 are shown as red  
259 circles and associated concentrations plotted within TS space (Fig. 2). Deep water 'C' was  
260 observed to have DIC concentrations > 2200 µmol kg<sup>-1</sup>, contrasting values between 'A' and 'B' of  
261 <1950 µmol kg<sup>-1</sup>. End-member mixing occurs at intermediate temperatures between the two  
262 watermasses 'B' and 'C', exhibiting in the gradient observed along the 34.2- 34.75 isohalines in  
263 figure 2. Assuming that upwelling results in the expression of deep water at the surface, upwelling  
264 will act to increase the surface inventory of DIC, leading to increased fCO<sub>2</sub>.

265 Rainfall results in surface layer dilution, thus reducing DIC between 'A' and 'B' (Turk et al.,  
266 2010). When the fCO<sub>2</sub> of both endmembers in this system (rainwater and surface waters) are at  
267 equilibrium with the atmosphere, this dilution results in a lowering of fCO<sub>2</sub> within the sea surface,  
268 and ingassing of CO<sub>2</sub>. Turk et al's (2010) study in the western Pacific warm pool have indicated  
269 that decreases of 30- 40 µatm in surface fCO<sub>2</sub> values can result from rainfall alone, with these  
270 effects strongest under highly stratified conditions. This suggests that rainfall could be an important  
271 influence on ETPO fCO<sub>2</sub>, particularly within the rainfall dominated Panama Basin region. Finally,  
272 although biological processes influence DIC (and hence fCO<sub>2</sub>), observations of DIC within this  
273 region suggest that physical processes (such as upwelling or rainfall) are the first order control on  
274 DIC. For this reason, coupled to the lack of net community production data within the region, we  
275 have elected to concentrate our efforts on quantifying these physical processes.

## 276 **2.5 Processing SOCAT data**

277

278 Surface fCO<sub>2</sub> responds to changing levels of atmospheric fCO<sub>2</sub>, which is increasing at  
279 approximately 20 µatm per decade, (Takahashi et al. 2009, Fig. 3). In order to calculate



280 contemporary air sea fluxes, the multi-year SOCAT data must first be temporally corrected for the  
281 inter-annual trend of increasing atmospheric CO<sub>2</sub>. The mean annual increase in fCO<sub>2</sub> within the  
282 central equatorial Pacific (5 °N and 5 °S) between 1979 and 1990 (excluding El Niño events) was  
283 estimated at a rate of 1.1 µatm +/- 0.3 per year (Takahashi et al., 2009). A second estimate, using  
284 data collected between 1990 and 2003 (within the region 10 °N to 5 °S) calculated an annual  
285 increase of 2.0 µatm +/- 0.2 (Takahashi et al. 2009). Correcting 1991 data to contemporary July  
286 2014 values would equate to a not-inconsiderable difference depending on which of these two  
287 estimates was chosen. Therefore, in order to choose the optimum correction for the smaller  
288 subregion of the ETPO, all available SOCAT fCO<sub>2</sub> data 1991-2011 were binned by year the data  
289 was collected (Fig. 3). Fitting a linear regression to the average bin value of this data, resulted in an  
290 average annual increase of 1.95 µatm/ yr (+/- 0.38 µatm). This linear regression was found to fall  
291 within the 5<sup>th</sup> and 95<sup>th</sup> percentile values for all bins, with variability in mean and median values  
292 caused by heterogeneous sampling of the region and inter-annual variability (Fig. 3, percentiles  
293 plotted in blue). Our calculated rate of increase is both within the range of the annual atmospheric  
294 fCO<sub>2</sub> increase measured at Fanning Island (1.8 µatm/ yr +/- 0.1 µatm) suggesting that fCO<sub>2</sub> within  
295 the ETPO tracks the rate of increasing atmospheric fCO<sub>2</sub> (Fig. 3). Therefore, we apply a  
296 1.95 µatm per year correction to the SOCAT data in order to normalise the fCO<sub>2</sub> data to 1<sup>st</sup> July  
297 2014 values, with all results herein also normalised to this date.

298 Although the spatial distribution of fCO<sub>2</sub> SOCAT observations within the ETPO are irregular,  
299 the variability can be constrained once the TS properties of the surface ocean are accounted for (as-  
300 per DIC, Fig. 2, Fig. 4). Data from the upper 100 m of Argo profiles completed in each of the three  
301 gulfs (Panama, Papagayo and Tehuantepec, as defined in Fig. 1) show strong TS similarities  
302 between each of the three gulf regions; low salinity (<34) waters are observed in each gulf, with a  
303 thermocline separating surface-water and deep water at 80-100 m depth. The T-S properties of  
304 these Argo profiles are replicated at the surface, within SOCAT SST and SSS data (Fig. 4).  
305 However, the relative numbers of observations that fall into each water-mass definition is different.  
306 The warm ETPO surface water is dominant in all three regions for the majority of samples, but  
307 cooler (<20 °C) deep water is occasionally expressed at the surface, with this signal strongest in the  
308 Gulf of Tehuantepec, followed by the Gulf of Papagayo, and seldom observed in the Gulf of  
309 Panama. This contrasts observations of low salinity water, which are very frequent within the Gulf  
310 of Panama, but rare in the other two gulfs (Fig. 4). Although both SOCAT and Argo data suffer  
311 from sparse sampling, we suggest that both datasets indicate that the relative dominance of each  
312 water-mass within each of the three gulfs is different. There are very few SOCAT observations

313 within the Offshore (OS) region (Fig. 1), thus we do not show a TS diagram from this region,  
314 however Argo profiles from this region are included in Fig. 2, highlighting that the OS also shares  
315 the same watermasses as the three gulf regions.

316 The distribution of DIC concentrations within TS space is mirrored by  $f\text{CO}_2$  values; deep  
317 water features higher  $f\text{CO}_2$  values than ETPO surface water, and low salinity water feature the  
318 lowest  $f\text{CO}_2$  values. The intermediate values of  $f\text{CO}_2$  observed between each of these watermasses  
319 suggest end-member mixing of  $f\text{CO}_2$  between waters-masses. Finally, there are no watermasses that  
320 share TS characteristics, but feature very different DIC concentrations or  $f\text{CO}_2$  values, meaning that  
321 DIC/  $f\text{CO}_2$  values are unique within TS space in this region (Fig. 2 and 4). Therefore, as both DIC  
322 and  $f\text{CO}_2$  values behave pseudo-conservatively within TS space, it is possible to construct a  
323 statistical description of  $f\text{CO}_2$  using solely SSS and SST.

## 324 **2.6 Fitting a statistical description- the look-up-table**

325

326 The similarities of T,S and  $f\text{CO}_2$  properties of the water-masses observed both by Argo and within  
327 the SOCAT database across all three gulfs enable data from the entire ETPO to be considered as a  
328 single system (Fig. 5A). As the variance between individual  $f\text{CO}_2$  observations occupying the  
329 same location within T S space is low (Fig 5B) we use a look-up-table (LUT) to describe  $f\text{CO}_2$  as  
330 a function of T and S. This LUT technique, although not previously used to estimate oceanic  $f\text{CO}_2$   
331 has proven useful for estimating net primary productivity using satellite observations (Zhao et al.  
332 2006). The LUT in this study uses a mesh of equal sized bins within T-S space. Observations of  
333  $f\text{CO}_2$  within these discrete TS bins are collected, and an average  $f\text{CO}_2$  within TS space calculated.  
334 As each bin is fully independent of neighbouring bins, non-linear and/or skewed  $f\text{CO}_2$  distributions  
335 within TS space can be accounted for, and hence an improved synthetic  $f\text{CO}_2$  product, with a lower  
336 root-mean-squared-error (RMSE) attained compared to using an alternative linear statistical  
337 description, such as a basic linear fit.

338 By combining all SOCAT data from the ETPO onto a single T-S diagram, a look-up-table  
339 (LUT) can be produced in order to describe  $f\text{CO}_2$  as a function of T and S (Fig. 5). The LUT was  
340 constructed by completing a linear interpolation of binned 0.1 (salinity) x 0.1 °C SOCAT data from  
341 the entire ETPO region. The number of observations per bin is depicted in the right-hand panel.  
342 Here, the highest number of observations fall in a narrow region of TS space, between salinities of  
343 32-34.5 and 25-30 °C (Fig. 5). We tested the quality of fit of the SOCAT based LUT, using LDEO  
344 v2013 data. Here  $f\text{CO}_2$  values were computed using the LUT using T and S from the LDEO v2013  
345 database, with this computed  $f\text{CO}_2$  data compared to the measured (annual increase corrected)

346 fCO<sub>2</sub> values. The LUT could be applied to 96.2 % of all LDEO v2013 T-S measurements made  
347 between 1990 and 2013, with the remaining TS measurements falling outside the TS boundaries of  
348 the LUT.

349 The average root-mean squared error (RMSE) of the LUT was 16.8  $\mu\text{atm}$  between the LUT  
350 computed and measured fCO<sub>2</sub>. This RMSE was asymmetrically distributed; highest at temperatures  
351 of 22-26 °C/ salinities of 34.4, and lowest at warmer temperatures/ lower salinities and also at  
352 colder temperatures (Fig. 5). This suggests that fCO<sub>2</sub> variability is low in aged surface waters, and  
353 high in recently upwelled and warming water. As there are a number of measurements within the  
354 region of higher error (Fig. 5), we hypothesize that this larger uncertainty is not due to lack of  
355 observations, but rather to mixing / heating processes that allow water with slightly different fCO<sub>2</sub>  
356 values to occupy the same TS space. For example, the fCO<sub>2</sub> value of water at salinities of 34.5 and  
357 25 °C could either result from the stoichiometric mixing of two waters at salinities of 34.5, and  
358 temperatures of 18 and 32 °C, or solely from warmed 34.5 /18 °C water. Upwelling, solar radiation,  
359 horizontal advection, biological productivity and diapycnal mixing processes all influence this  
360 system, and need to be accounted for in order to quantify completely the formation and fCO<sub>2</sub>  
361 observed within this water, but are outside the scope of this study using the data available in this  
362 region. However, this LUT technique works efficiently in determining the first-order variability of  
363 the system within the ETPO. In order to calculate an uncertainty on this LUT derived synthetic  
364 fCO<sub>2</sub> data, we complete mean squared error calculation, based on the variance of fCO<sub>2</sub> data within  
365 each cell of the LUT, and the uncertainty on SMOS SSS (as described above.) We assume that the  
366 SST and SSS measurements within SOCAT, and the OSTIA SST product are accurate. This error is  
367 propagated through the air-sea flux estimates in order to calculate an error on the air sea flux values  
368 (as reported in table 1).

### 369 **3 Results**

#### 370 **3.1 Bi-monthly variability along ship tracks**

371 Bimonthly fCO<sub>2</sub> variability can be seen in the (1991-2011 annual fCO<sub>2</sub> increase corrected)  
372 SOCAT observations plotted in figure 6. Here, the top six panels display SOCAT observations of  
373 fCO<sub>2</sub>, and the lower six panels display fCO<sub>2</sub> calculated from SOCAT SST and SSS observations  
374 using the LUT. As the majority of SOCAT data is collected using ships of opportunity, resolution  
375 within shipping lanes in the ETPO is excellent; (for example multiple observations are made in  
376 regions south and west of the Panama Canal, at 9.1 °N, 79.7 °W ), but very sparse outside of these  
377 shipping lanes. However, acknowledging these issues in data resolution, patterns in fCO<sub>2</sub> can still

378 be detected. The lowest  $f\text{CO}_2$  is typically observed within the Gulf of Panama and close to the  
379 coast. Higher  $f\text{CO}_2$  is observed in the Gulfs of Tehuantepec and Papagayo. Inter seasonal variability  
380 is also observed, with the highest  $f\text{CO}_2$  ( $>440 \mu\text{atm}$ ) occurring between November to February, with  
381 lower  $f\text{CO}_2$  values ( $<400 \mu\text{atm}$ ) occurring across the entire region during the summer months (May-  
382 August). The data also show a very high degree of variability at small spatial and temporal scales,  
383 for example during November- December within the Gulf of Papagayo. It is this variability that  
384 needs to be resolved using the LUT, when coupled to satellite SST and SSS observations at native  
385 resolution. Thus, it is important to confirm that the LUT is able to recreate this variability in  $f\text{CO}_2$   
386 using the initial (SOCAT) T and S conditions. Here, we find that the LUT performs well, with most  
387 of the measured  $f\text{CO}_2$  variability also observed within the  $f\text{CO}_2$  calculated using the LUT (Fig. 6).

### 388 **3.2 $f\text{CO}_2$ and fluxes by region- the influence of wind and upwelling**

389 Using the LUT, SMOS and OSTIA data, the small scale variability and features within the  
390 ETPO can be resolved more thoroughly than possible through SOCAT data alone. Bimonthly  
391 averaged (between July 2010- June 2014) SSS, SST,  $f\text{CO}_2$ ,  $\text{CO}_2$  flux and windspeed are plotted in  
392 figure 7. Jet wind velocities over the Gulf of Tehuantepec are at their peak between October to  
393 February, thus optimising Ekman upwelling. During these months the SSS of the Gulf increases,  
394 whilst the SST decreases (due to the influences of upwelling). Increased  $f\text{CO}_2$  is observed across  
395 most of the Gulf of Tehuantepec, with peak outgassing occurring as a narrow band, centred  
396 underneath the axis of the jet wind. We suggest that this is due to the complimentary nature of ocean  
397 physics in this region- high windspeeds promote Ekman upwelling of high DIC deep water, whilst  
398 increasing the  $k$  component within the air-sea flux parameterisation, thereby maximising  $\text{CO}_2$   
399 outgassing.

400

## 401 **4 Discussion**

402 The Gulf of Papagayo shares many similarities with the Tehuantepec Gulf, however, jet winds  
403 are more zonally aligned (reducing Ekman pumping strengths, (Alexander et al., 2012)), and occur  
404 later into winter compared to Tehuantepec; being strongest between November-February. This  
405 results in a lag between peak wintertime  $f\text{CO}_2$  values seen within the two gulfs (Fig. 7 and 8). The  
406 Gulf of Papagayo also features elevated  $f\text{CO}_2$  values during the summer months which are not  
407 seen in either the Gulf of Tehuantepec or Panama. This is due to westerly winds maintaining the  
408 vorticity of the Costa Rican dome structure, enabling the continued upwelling of deep water at the  
409 core of this dome throughout the summer season (Grotsky et al. 2014, Kessler 2006, Fig. 7).

410 The seasonally averaged wind velocities in the Gulf of Panama superficially resemble the  
411 wind patterns in the two northern gulfs- jet winds are observed between January- February (Fig. 7).  
412 However, excluding a small area of ocean directly underneath the wind jet axis during winter, the  
413 region remains a small net sink of carbon throughout the year (Fig. 8). We suggest that this contrast  
414 between the Panama gulf and the rest of the ETPO is due to the high rainfall within this region,  
415 resulting in the dilution of DIC as described above.

416

#### 417 **4.1 fCO<sub>2</sub> and the influence of rainfall**

418 A significant proportion of atmospheric water exported from above the Atlantic basin into the  
419 Pacific basin is precipitated into the ETPO. This rainfall is intensified within the ETPO during the  
420 summertime, due to a northwards shift of the ITCZ towards the Panama coast (Xie et al., 2005). As  
421 a result, the Gulf of Panama receives net precipitation of 180-220 cm per year, with peak rainfall of  
422 20 mm per day during July- August (Alory et al., 2012). This large freshwater flux, coupled with  
423 light southerly winds between March and June, and the southerly (thus downwelling promoting),  
424 winds between July and December result in the semi-permanent stratification of the water-column  
425 (Fig. 7, Alory et al., 2012). Qualitatively, this stratification can be observed in the SOCAT and Argo  
426 data in Fig. 4 by the scarcity of deep water observations in the Gulf of Panama compared to the  
427 Gulfs of Tehuantepec and Papagayo. Additionally, the thermocline is observed at deeper depths in  
428 the Argo profiles taken in the Gulf of Panama, compared to either the Gulf of Tehuantepec or  
429 Papagayo (Fig. 4).

430 The influence of stratification/ rainfall on fCO<sub>2</sub> values are seen in figure 7. Here, the lowest  
431 salinity and fCO<sub>2</sub> are observed during the summer months (during peak rainfall season, Alory et al.  
432 2012). This low fCO<sub>2</sub>, stratified system persists until January, when intensification of the south  
433 equatorial current results in the export of the low fCO<sub>2</sub>/fresh surface layer towards the south west  
434 (as seen by the elongation of the freshpool during January- March SSS). This export of water,  
435 coupled with the dry-season within the Panama Gulf appears to weaken the stratification, with the  
436 result that sporadic jet winds enable Ekman upwelling, thus increasing SSS (Alory et al., 2012, Fig.  
437 7) and average fCO<sub>2</sub> values between January-April (Fig. 8).

#### 438 **4.2 Air-sea fluxes from July 2010- June 2014**

439 The ETPO between 2010- June 2014 had an annual average  $\Delta$  fCO<sub>2</sub> 39.4  $\mu$ atm (+/- 10.7)  
440 and a outgassing CO<sub>2</sub> flux of 1.51 mmol m<sup>-2</sup> d<sup>-1</sup> (+/-0.41 , Fig. 8/ Table 1). The ETPO, (the

441 boundaries of which defined as per Fig. 1), has an area of  $1.9 \times 10^{12} \text{m}^2$ , therefore, net outgassing  
442 from the ETPO equates to  $10 \times 10^{12} \text{g}$  (0.01 Pg) of carbon per year.

443 To examine both monthly and inter-annual variability in air-sea fluxes, the ETPO is split into  
444 subregions; the Gulfs of Tehuantepec, Papagayo, Panama and the Offshore region (Fig. 1, 8 and  
445 Table 1). Here, although variable, the consistent seasonal cycle between summer and winter  
446 observed in the bimonthly data in figure 7 is present for each individual year. Peak outgassing  
447 within the Gulfs of Tehuantepec, Papagayo and the South Equatorial Current occurs during winter  
448 (November -February). This contrasts the Gulf of Panama, which is a net sink of atmospheric  $\text{CO}_2$   
449 except during occasional upwelling events during January -February. However, ingassing within of  
450 the Gulf of Panama is insufficient to offset outgassing from the rest of the ETPO and the region  
451 remains a net source of  $\text{CO}_2$  year round.

452 In addition to seasonal variability, interannual variability is observed within the ETPO. Across  
453 the entire basin, the ETPO during the years 2010-2012 were on average cooler and saltier than  
454 subsequent years (especially 2012-13). This resulted in higher annually averaged outgassing and  
455  $\text{fCO}_2$  values (Fig. 8, Table 1). Concurrently, during 2010-2012, the Gulf of Panama featured a low  
456 salinity minima (of 28), and a small freshpool footprint (the boundary of which is defined by the  
457 33 isohaline (Alory et al., 2012)). The extreme low salinity observed within the eastern region of  
458 Gulf of Panama during 2010 to 2012 resulted in lower  $\text{fCO}_2$  and hence stronger ingassing than  
459 2012-2014 values, highlighting the effect of rainwater dilution of DIC. However, this localised  
460 ingassing was insufficient to offset the higher outgassing observed in the other regions of the ETPO  
461 (Fig. 8, Table 1). Also noteworthy is that average wind velocities were similar during 2010 and  
462 2012 compared to subsequent years, so the increased outgassing observed during these years  
463 cannot be attributed to increased wind-mediated Ekman upwelling of deep water alone.

464 We note that 2010 to the end of 2011 featured La Niña conditions, and suggest that this could  
465 be causal to the differences in  $\text{CO}_2$  fluxes, higher salinities and higher  $\text{fCO}_2$  observed during this  
466 time period. A study using NCEP (National Centers for Environmental Prediction) reanalysis data  
467 within this region observed cooler waters and more frequent upwelling of sub-thermocline deep  
468 water during La Niña events. A shallower thermocline depth exists during La Niña events, which  
469 results from the uplift of the water-column by intensified coastal Kelvin waves. Mechanistically,  
470 this shallow thermocline reduces the strength of upwelling required for the expression of deep  
471 water at the surface, thus decreasing average SSTs observed during La Niña periods (Alexander et  
472 al., 2012).

473 Furthermore, Alexander et al's (2012) study also suggests that although jet winds are the first  
474 order control on SST and thermocline depth during winter; it is changes in the thermocline depth

475 (rather than changes in jet winds) that result in observed El Niño Southern Oscillation (ENSO)  
476 variability in SST within the ETPO. Variability in thermocline depth, and by association,  
477 variability in the ease by which deep water can be advected towards the surface suggest that  
478 (ENSO) could drive variability within the surface  $f\text{CO}_2$  observations. However, with only four  
479 years of data, and no El Niño phase for intercomparison we are unable to draw definitive  
480 conclusions.

### 481 **4.3 Previous work**

482 The results in figures 7 and 8 represent the first attempt to quantify  $f\text{CO}_2$  within the ETPO region at  
483 high resolution from observations. Previous work encompassing the ETPO include three basin wide  
484 or global  $f\text{CO}_2$ /  $p\text{CO}_2$  studies, by Takahashi et al., (2009), Ischii et al., (2014) and Landschützer et  
485 al., (2014). These studies, as discussed in the introduction, are based on the extrapolation of  $p\text{CO}_2$   
486 directly, or the extrapolation of  $p\text{CO}_2$  using a neural network technique, and feature spatial  
487 resolutions of  $4^\circ \times 5^\circ$ ,  $4^\circ \times 5^\circ$  and  $1^\circ \times 1^\circ$  respectively. Of these three studies, only Takahashi et al.  
488 (2009) features a dataset of calculated  $\Delta p\text{CO}_2$  that includes the ETPO region, thus allowing direct  
489 comparison with our calculations. Collocating the  $\Delta p\text{CO}_2$  results from Takahashi et al. (2009)  
490 within the ETPO regions defined in Fig. 1, the Gulfs of Tehuantepec, Papagayo, Panama and the  
491 Offshore region featured values of +28.9, +54.1, +17.1 and +19.5  $\mu\text{atm}$  respectively (Table. 1). This  
492 suggests that the Takahashi et al. (2009) study differentiates the low  $\Delta p\text{CO}_2$  Panama Gulf region  
493 from the higher values observed in the Gulfs of Tehuantepec and Papagayo. The resolution used in  
494 Takahashi's study is too coarse to identify any mesoscale features (such as upwelling), or  
495 interannual variability within the region. However, we find average  $\Delta p\text{CO}_2$  from Takahashi et al.  
496 (2009) of 29.1  $\mu\text{atm}$  was within the error limits of this study (Table 1).

497 Applying the same windspeed product, and gas transfer parametrisation used in this study to the  
498 ETPO  $\Delta p\text{CO}_2$  values from Takahashi et al. (2009), we find that all regions are net outgassing, with  
499 the Gulf of Papagayo dominant within the ETPO (table 1). The ingassing observed within the Gulf  
500 of Panama for our study is not replicated in Takahashi et al. 2009. The improved resolutions  
501 featured in Landschützer et al., (2014) resulted in some mesoscale features being observed, such as  
502 the increased  $p\text{CO}_2$  values within the Gulf of Papagayo. However, this work was not able to  
503 identify the low  $p\text{CO}_2$  conditions within the Gulf of Panama, most likely due to the same  
504 biogeochemical province description being applied to the entire ETPO. In addition to these studies,  
505 our work distinguishes the importance of jet winds in increasing  $p\text{CO}_2$ , and quantifies the strong  
506 inter annual variability within the region.

## 507 **5. Conclusions**

508 Estimating surface  $f\text{CO}_2$  and air sea fluxes of  $\text{CO}_2$  within the global oceans has advanced  
509 considerably over the past decade, assisted by the assembly of large standardised atlases of surface  
510 observations (such as the SOCAT database). However, although these databases boast  
511 measurements in the millions, a challenge remains in gauging seasonal or sub-mesoscale  $f\text{CO}_2$   
512 variability in the oceans. Sampling through the use of commercial volunteering observation vessels  
513 may introduce bias, with most data existing within the narrow confines of the major global  
514 shipping lanes or for only a few months of the year. However, the quantification of  $f\text{CO}_2$  within TS  
515 space, used in conjunction with observations of surface SST and SSS (made possible through the  
516 recent availability of SSS from satellite) has proved highly useful in improving our understanding  
517 of  $f\text{CO}_2$  variability at much improved spatial and temporal resolutions. We have demonstrated a  
518 technique using SOCAT data to identify the  $f\text{CO}_2$  signatures of water-masses within the ETPO  
519 region, namely, the high  $f\text{CO}_2$  deep water, the near equilibrium ETPO surface water, and the  
520 undersaturated rainfall diluted surface waters. From this, we used a LUT technique, in order to  
521 produce a description of the  $f\text{CO}_2$  content of an ETPO surface water using satellite SST and SSS.  
522 The highest outgassing and surface  $f\text{CO}_2$  were observed during the winter period  
523 (November -March), in the Gulfs of Tehuantepec, Papagayo and in the south equatorial current.  
524 The first order control on these upwelling events and hence  $f\text{CO}_2$  in the ETPO are strong wind jets  
525 blowing through low altitude gaps in the Central American cordillera. The Gulf of Panama  
526 remained net undersaturated on average, due to dilution effects of heavy rainfall and the  
527 stratification of the water column. Although wind jets were observed in the Gulf of Panama, the  
528 exceptionally low density of the water within this region appears to limit upwelling, and any  
529 upwelling that occurs is directly underneath the wind jet axis. Inter-annual variability was observed  
530 within the region, with the location of the western extent of the freshpool moving westwards  
531 considerably between 2010 and 2014. Previous work within this region suggest that changes in  
532 thermocline depth related to ENSO are likely to influence  $f\text{CO}_2$  within this region. The region is a  
533 net contributor to atmospheric  $\text{CO}_2$ , with average sea to air fluxes (over the four years of  
534 observations) of  $1.5 \text{ mmol m}^{-2} \text{ d}^{-1}$ , with all regions of the ETPO outgassing year-round, except  
535 the rainfall diluted Gulf of Panama/ Freshpool region.

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542

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547

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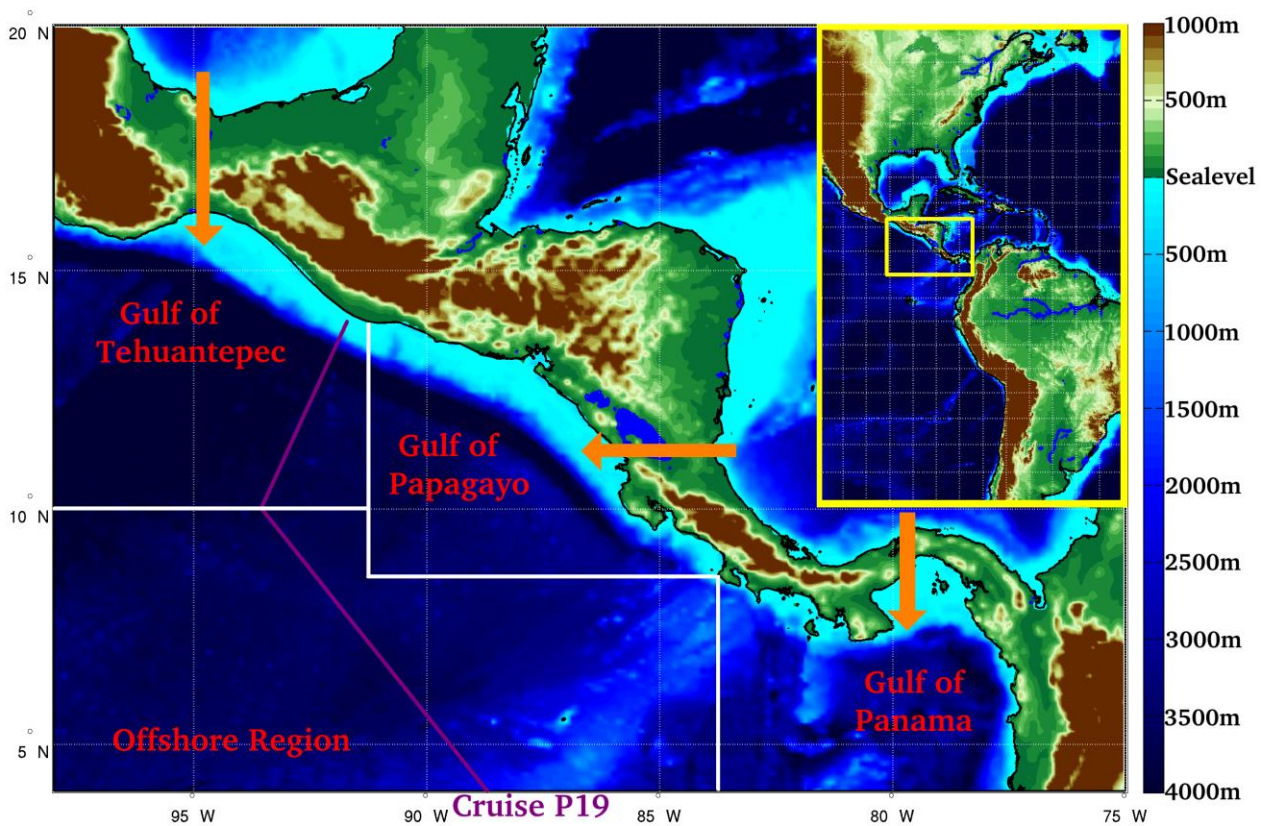
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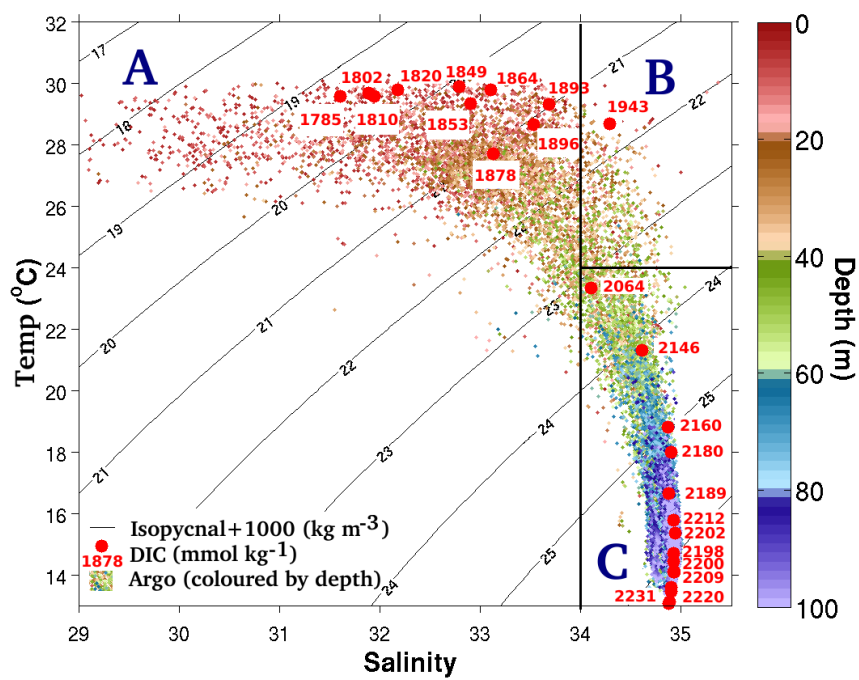
639 **Figure**



640

641 Figure 1.

642 The topography of the Eastern Tropical Pacific Ocean and Isthmus of Panama, plotted using  
643 GEBCO bathymetry. The three gulfs with the region, Panama (84° W to coastline of panama),  
644 Papagayo (84 to 91 °W, north of 8 °N) and Tehuantepec (west of 91 °W, north of 10 °N), and the  
645 Offshore region are marked in white. The transect of WOCE cruise P19 is indicated in purple. The  
646 path of jet winds are marked by orange arrows.



647

648 Figure 2.

649 The total library of Argo profiles collected within the ETPO in the upper 100 metres. Numbered red  
 650 dots indicate the total dissolved carbon concentration at specific TS values as measured during  
 651 cruise WOCE cruise p19.

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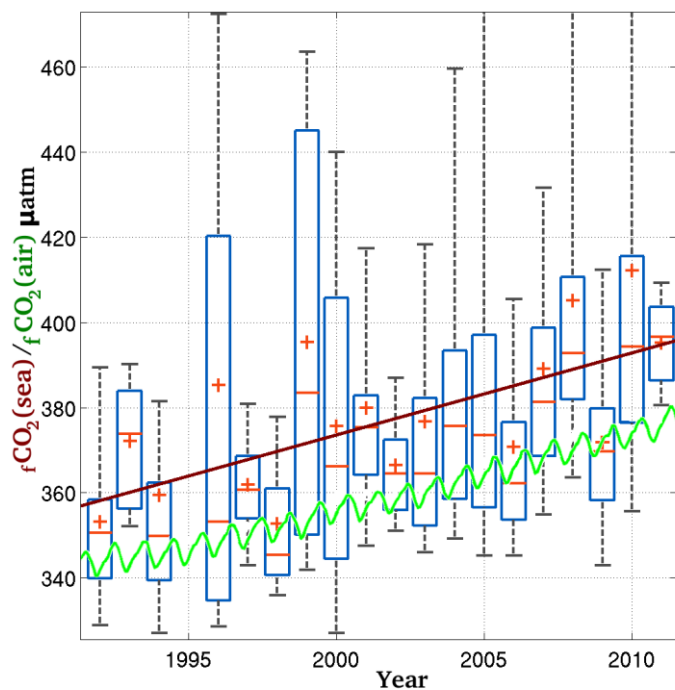
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665 Figure 3.

666 Yearly binned pCO<sub>2</sub> measurements from the SOCAT database. The interpolated average rate of  
 667 ETPO fCO<sub>2</sub> increase is shown as a red line, with the atmospheric fCO<sub>2</sub> data shown as a green line.  
 668 The blue boxes represent the 5<sup>th</sup> and 95<sup>th</sup> percentile pCO<sub>2</sub>, with the small red lines indicating yearly  
 669 averages, and the red crosses yearly median.

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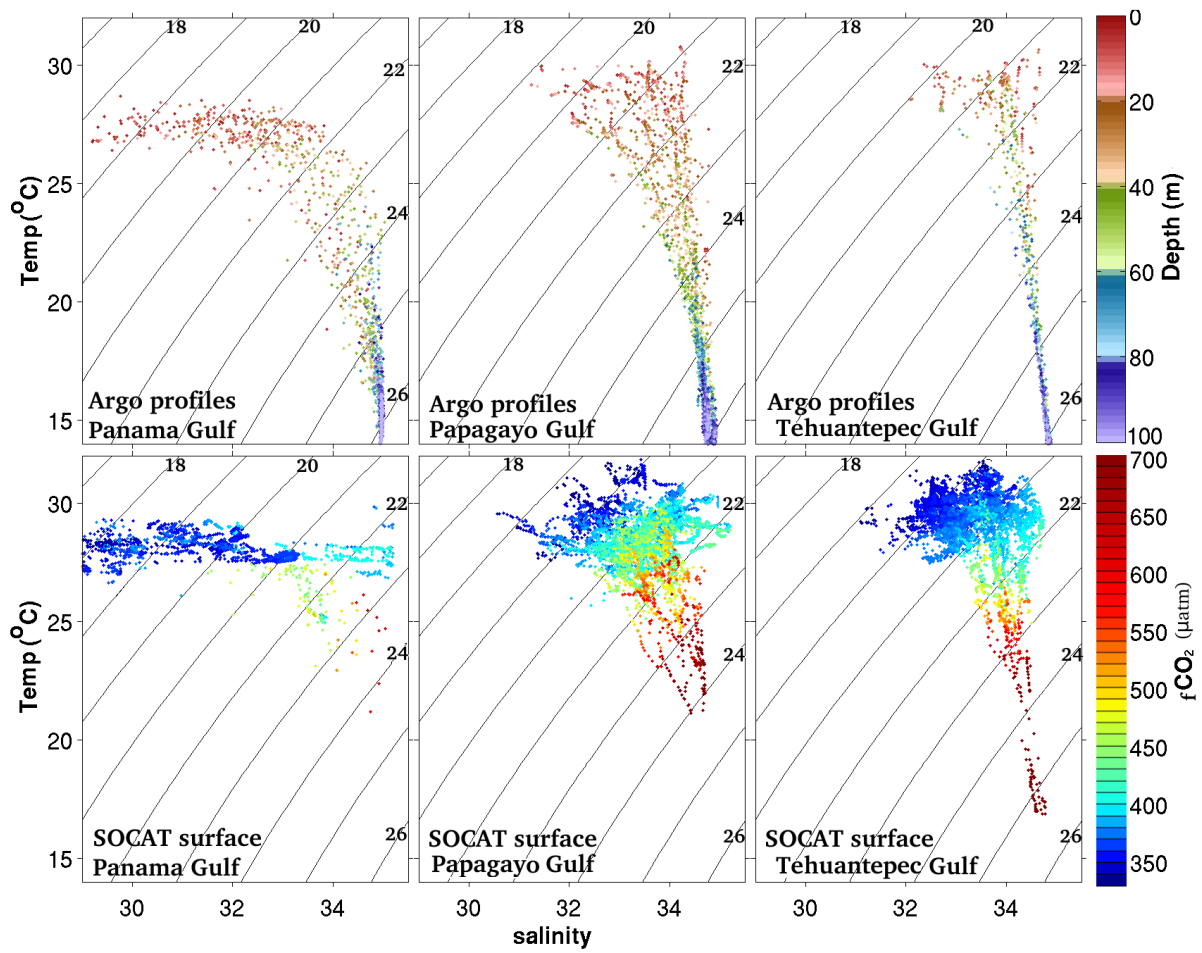
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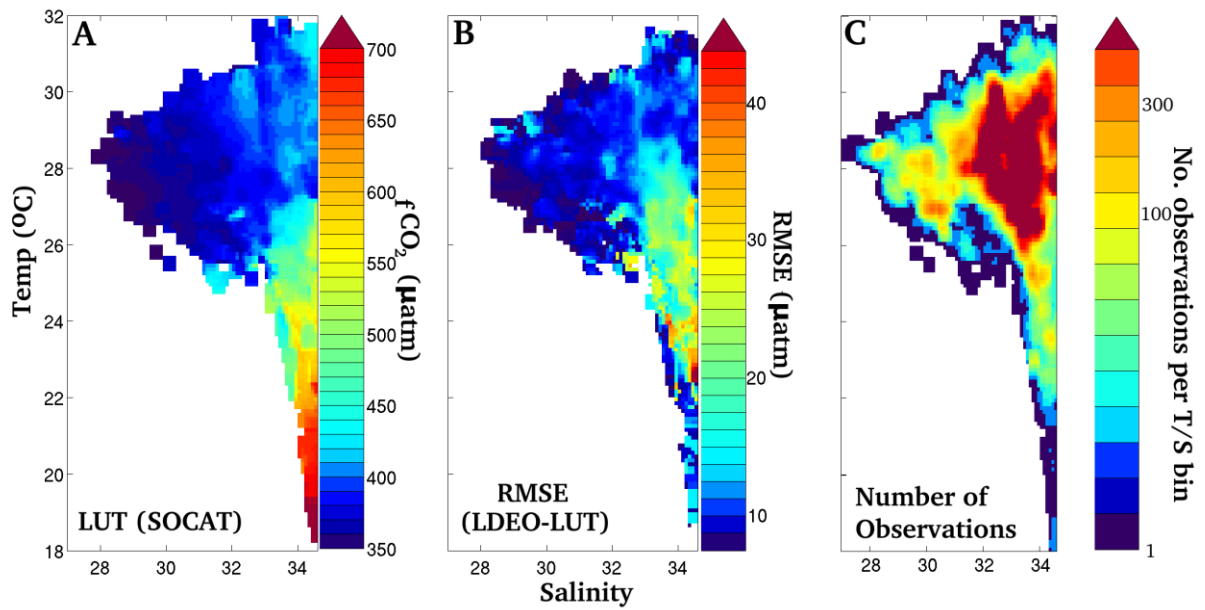
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681 Figure 4.

682 TS data from the three main gulfs within the ETPO both from Argo floats (coloured by depth-top  
 683 three plots) and from surface SOCAT T and S data, (coloured by fCO<sub>2</sub> -bottom three plots).



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686 Figure 5.

687 A, the Look Up Table derived from the position of ETPO SOCAT  $f\text{CO}_2$  measurements within T S  
 688 space.

689 B the root mean squared error of the LUT  $f\text{CO}_2$  observations, showing the variance between  $f\text{CO}_2$   
 690 observed within the same TS space in the LUT.

691 C the number of LDEO measurements per  $0.1^\circ \times 0.1^\circ$  salinity/ temperature bins that went into  
 692 generating the LUT

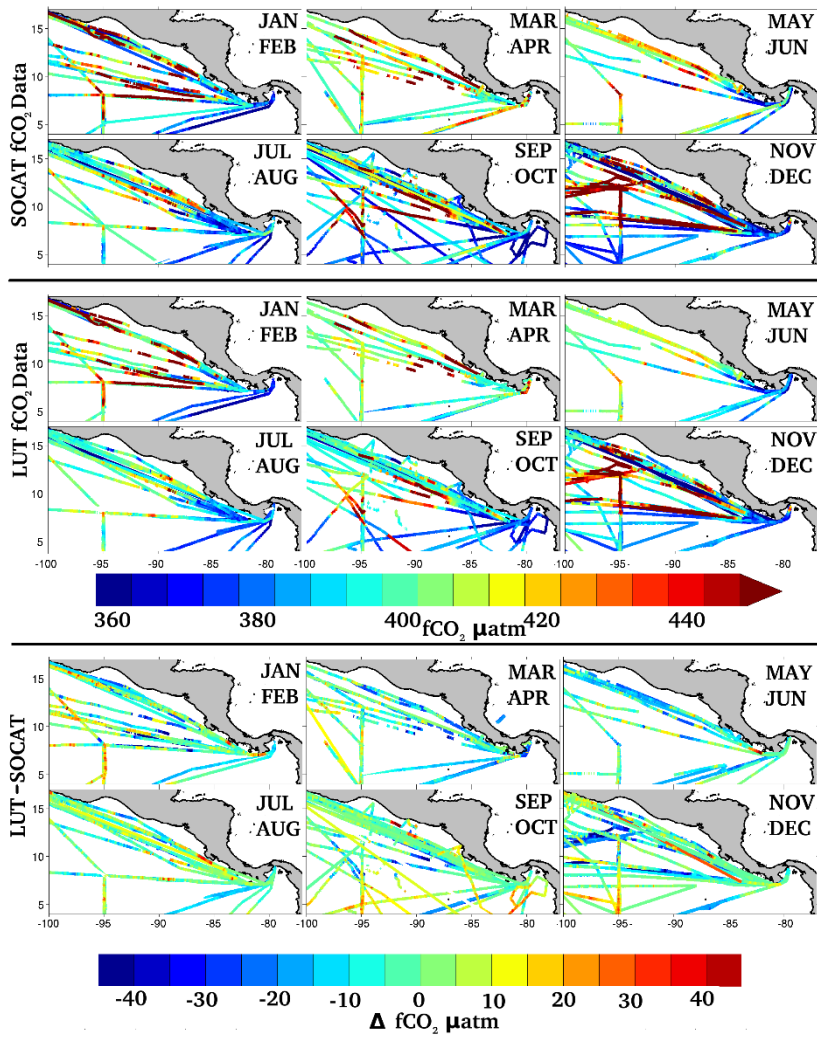
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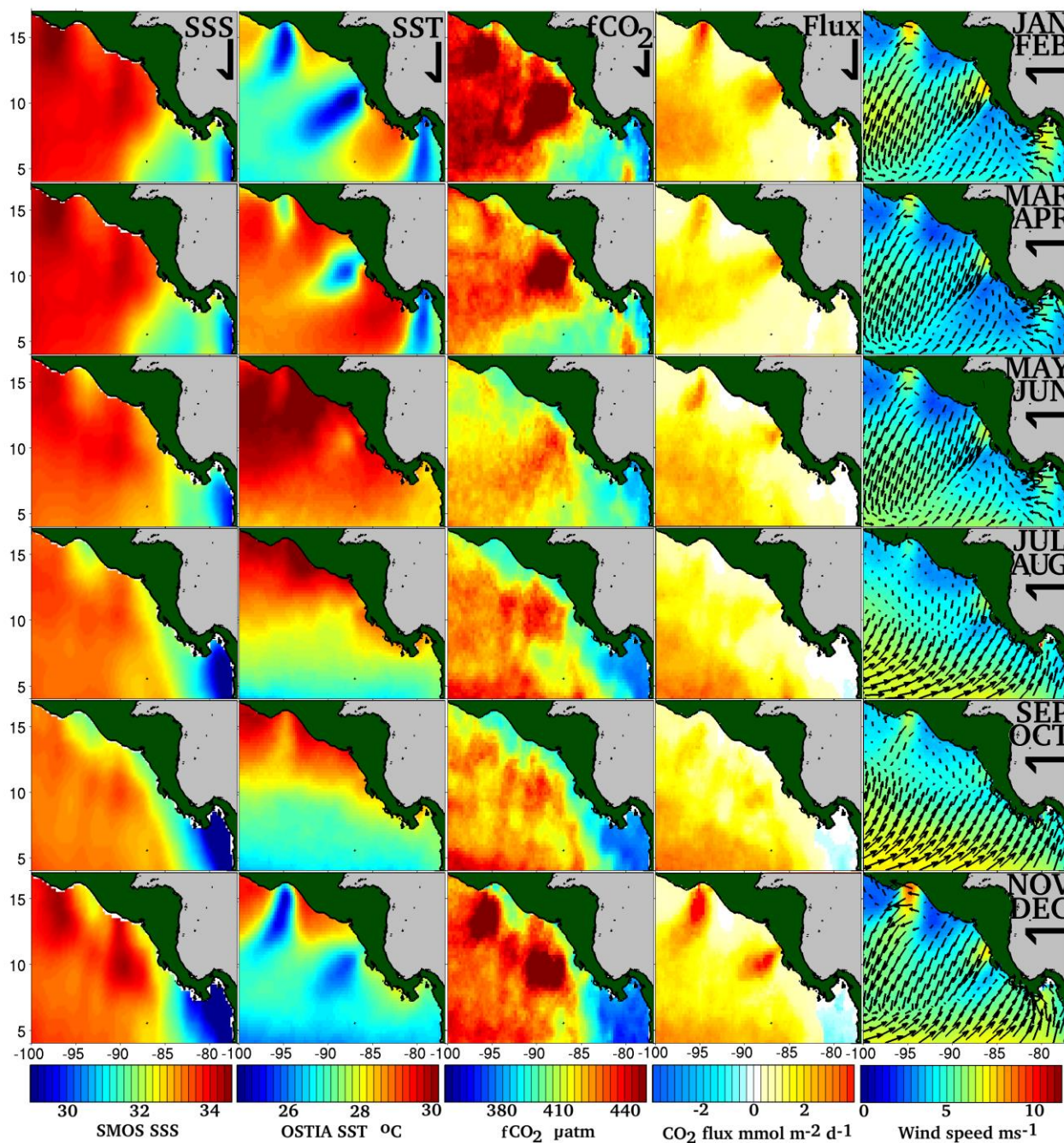
698 Figure 6.

699 SOCAT fCO<sub>2</sub> bimonthly (Jan+Feb, Mar+Apr, May+Jun, Jul+Aug, Sep+Oct, Nov+Dec)

700 observations within the ETPO corrected for the annual CO<sub>2</sub> increases, LUT derived bimonthly fCO<sub>2</sub>

701 calculated using SOCAT SST and SSS observations and LUT-SOCAT observations

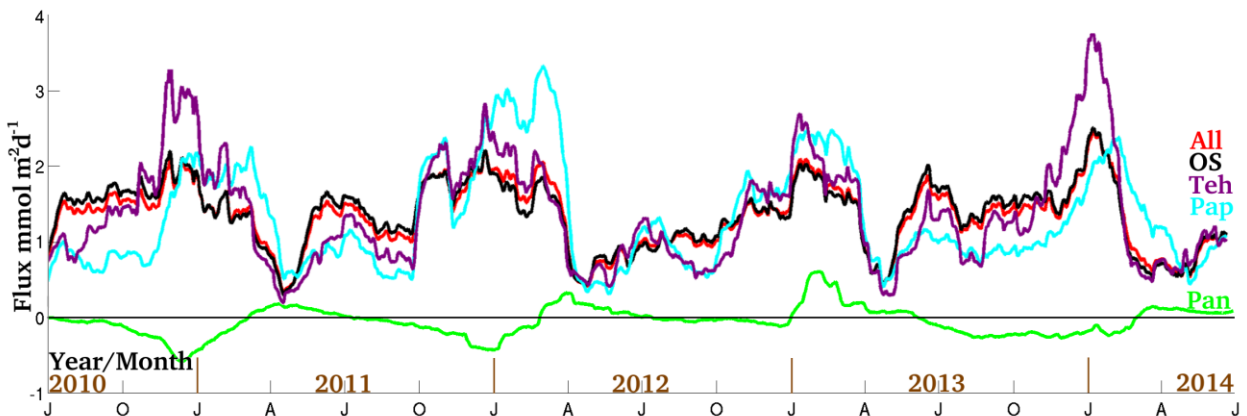
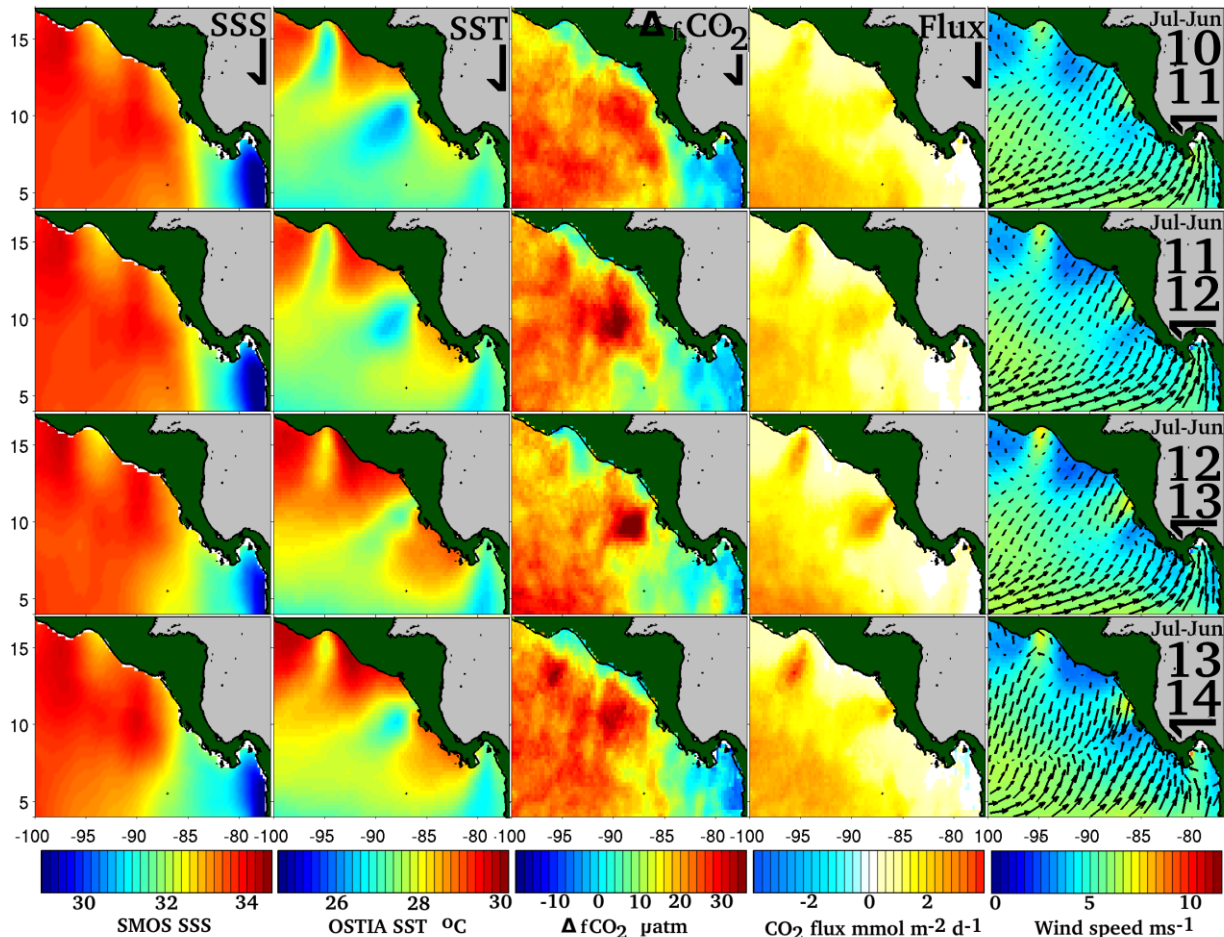




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703 Figure 7.

704 July 2010- June 2014 average SSS, SST, LUT derived  $f\text{CO}_2$ , air sea fluxes and wind vectors for  
 705 the ETPO, split bimonthly (Jan+Feb, Mar+Apr, May+Jun, Jul+Aug, Sep+Oct, Nov+Dec).



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707 Figure 8.

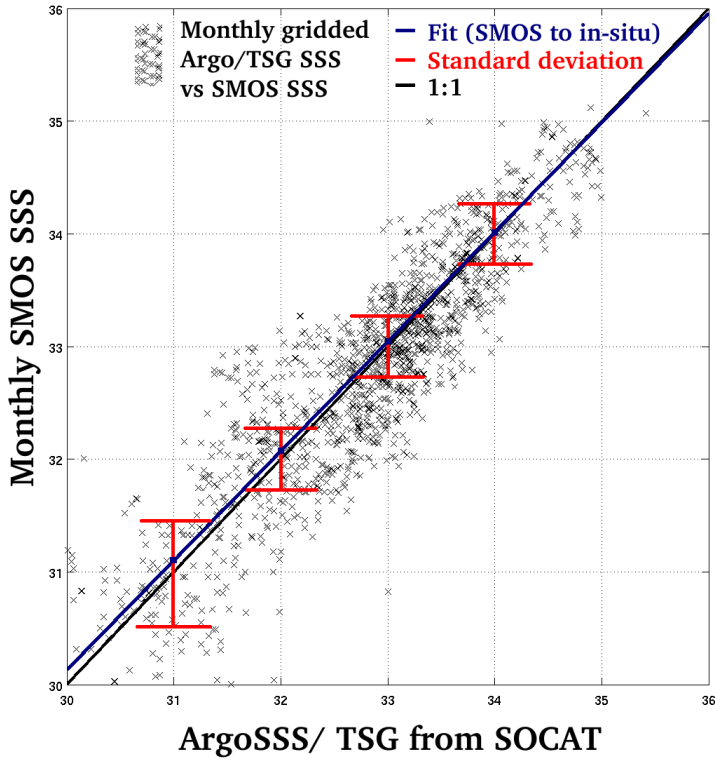
708 Upper: Yearly average SSS, SST LUT derived  $\Delta f\text{CO}_2$ , air sea fluxes and wind vectors for the ETPO  
 709 for July to June 2010+2011, 2011+2012, 2012+2013 and 2013+2014.

710 Lower: The continuous LUT derived  $f\text{CO}_2$  fluxes from the entire ETPO (red line), the Gulfs of  
 711 Tehuantepec (purple), Papagayo (blue), Panama (green) and the South Equatorial Current (black).

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Supplementary Figure 1.

Monthly 0.25 degree SMOS SSS matchups with near surface Argo and SOCAT TSG data (that have been binned into the same monthly/ 0.25 degree grid as the SMOS SSS data). The red error bars indicate the standard deviation, the blue line is a linear regression between the two SSS datasets.

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744 Table 1.

745 Annual averaged values for each region, as reported in figure 8 and average values from Takahashi  
746 et al. 2009.

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Date/ Location	$\Delta fCO_2$ ( $\mu atm$ )	CO2 Flux (mmol m <sup>-2</sup> d <sup>-1</sup> )	CO2 Flux error (+/- mmol m <sup>-2</sup> d <sup>-1</sup> )	Wind (ms <sup>-2</sup> )	Salinity	Temp (°C)
2010-11 ETPO	40.4	1.55	0.39	4.81	33.3	27.9
2010-11 OS	41.1	1.62	0.40	5.18	33.2	27.7
2010-11 Panama	-4.7	-0.19	-0.03	4.57	28.3	27.4
2010-11 Papagayo	34.8	0.95	0.20	4.17	33.7	28.2
2010-11 Tehuantepec	40	1.55	0.33	3.72	33.6	29.0
2011-12 ETPO	42	1.56	0.43	4.59	33.3	27.9
2011-12 OS	43.2	1.60	0.48	4.97	33.1	27.7
2011-12 Panama	-4.1	-0.21	0.05	4.02	28.4	27.5
2011-12 Papagayo	34.1	1.09	0.34	4.01	33.7	27.9
2011-12 Tehuantepec	49.3	1.64	0.36	3.90	33.7	28.4
2012-13 ETPO	35.3	1.41	0.40	4.88	33.2	28.6
2012-13 OS	37.1	1.50	0.43	5.11	33.0	28.3
2012-13 Panama	-2.1	-0.11	0.08	3.92	28.7	27.3
2012-13 Papagayo	32.8	1.08	0.37	4.49	33.6	28.5
2012-13 Tehuantepec	32.5	1.52	0.43	4.08	33.6	29.3
2013-14 ETPO	39.9	1.53	0.41	4.67	33.3	28.0
2013-14 OS	39.3	1.63	0.45	5.09	33.0	27.8
2013-14 Panama	-2.8	-0.16	0.05	3.88	27.5	27.7
2013-14 Papagayo	37.9	1.15	0.38	4.15	33.7	28.0
2013-14 Tehuantepec	47.7	1.57	0.43	3.92	33.7	28.6
Average ETPO	39.4	1.51	0.41	4.7	33.3	28.10
Average OS	40.2	1.59	0.44	5.1	33.1	27.88
Average Panama	-3.4	-0.17	0.04	4.1	28.2	27.48
Average Papagayo	34.9	1.07	0.32	4.2	33.7	28.15
Average Tehuantepec	42.4	1.57	0.42	3.91	33.7	28.83
Takahashi 2009. ETPO	29.1	0.78	N/A	4.74	33.3	28.10
Takahashi 2009. OS	19.5	0.65	N/A	5.09	33.1	27.88
Takahashi 2009. Pan	17.1	0.54	N/A	4.10	28.2	27.48
Takahashi 2009. Pap	54.1	1.83	N/A	4.21	33.7	28.15
Takahashi 2009. Tec	28.9	0.71	N/A	3.91	33.7	28.83

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