

1 A sensitivity analysis of the impact of rain on
2 regional and global sea-air fluxes of CO₂

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11

12 **Abstract**

13 The global oceans are considered a major sink of atmospheric carbon dioxide
14 (CO₂). Rain is known to alter the physical and chemical conditions at the sea
15 surface, and thus influence the transfer of CO₂ between the ocean and
16 atmosphere. It can influence gas exchange through enhanced gas transfer
17 velocity, the direct export of carbon from the atmosphere to the ocean, by
18 altering the sea skin temperature, and through surface layer dilution. However,
19 to date, very few studies quantifying these effects on global net sea-air fluxes
20 exist. Here, we include terms for the enhanced gas transfer velocity and the
21 direct export of carbon in calculations of the global net sea-air fluxes, using a 7-
22 year time series of monthly global climate quality satellite remote sensing
23 observations, model and in-situ data. The use of a non-linear relationship
24 between the effects of rain and wind significantly reduces the estimated impact
25 of rain-induced surface turbulence on the rate of sea-air gas transfer, when
26 compared to a linear relationship. Nevertheless, globally, the rain enhanced gas
27 transfer and rain induced direct export increase the estimated annual oceanic
28 integrated net sink of CO₂ by up to 6 %. Regionally, the variations can be larger,
29 with rain increasing the estimated annual net sink in the Pacific Ocean by up to
30 15% and altering monthly net flux by $> \pm 50\%$. Based on these analyses, the
31 impacts of rain should be included in the uncertainty analysis of studies that
32 estimate net sea-air fluxes of CO₂ as the rain can have a considerable impact,
33 dependent upon the region and timescale.

34 **1.0 Introduction**

35 The sea-air exchange of the greenhouse gas carbon dioxide (CO₂) is a critical part
36 of the climate system and a major factor in the biogeochemical development of
37 the oceans. It is widely accepted that more accurate and higher resolution
38 calculations of these gas exchanges (fluxes) are required if we are to fully
39 understand and predict our future climate. Such knowledge is also required for
40 understanding and monitoring chemical water quality (e.g. in relation to ocean
41 acidification).

42 The impact of raindrops falling on the sea can influence the rate of gas transfer
43 between the ocean and the atmosphere by increasing surface turbulence. The
44 addition of rainwater will change the temperature, salinity and carbonate
45 composition of surface waters, affecting the solubility and partial pressure of CO₂
46 (pCO_2) in the surface layer. Rain will also directly transfer dissolved CO₂ to the
47 ocean (termed wet deposition).

48 Early work by Ho *et al.* (1) highlighted how rain can significantly enhance gas
49 transfer and provided the first parameterisation of rain-driven gas transfer
50 velocity for freshwater environments. This work was extended towards
51 determining the physical mechanisms underlying this enhancement, showing
52 that the impact of rain in freshwater systems caused surface bubbles and waves
53 and that the enhancement of gas transfer was mainly due to an increase in
54 surface turbulence (2). It was subsequently shown that the rain-driven gas
55 transfer velocity was similar for freshwater and saltwater systems, although
56 differences in vertical mixing in the saltwater system due to stratification meant
57 that gas flux in the seawater system was lower (3). More recently, Zappa *et al.* (4)
58 showed that rain-induced turbulence was the main reason for rain enhanced gas
59 transfer in saltwater systems. They also showed that the gas transfer velocity
60 scaled with the turbulent dissipation rate.

61 Rain and wind effects were initially understood to combine linearly to influence
62 gas transfer velocity, k (3). However, Harrison *et al.* (5) showed that whilst rain
63 can contribute significantly to the total sea-air gas flux at low wind speeds, at
64 higher wind speeds the effects become negligible and a new non-linear
65 parameterisation for the gas transfer in field conditions was presented (5).

66 The changes in temperature, salinity and carbonate composition of surface
67 waters caused by the introduction of rainwater will alter the solubility and
68 partial pressure of CO₂ (pCO_2) in the surface layer. Data sets gathered from in-
69 situ or satellite remote sensing are selected or adjusted to represent the surface
70 water conditions that dominate sea-air gas flux (6). On the spatial and temporal
71 scales resolved by these data, stratification and the influence of rain on surface
72 waters will be represented and thus accounted for in exchange calculations.

73 However, rain events will occur at spatial and temporal scales that are not
74 resolved by the data used for these global studies, such as short, intense rain
75 showers. As such, the temporary dilution of surface water during rain events,
76 that has the potential to affect the sea-air gas flux, will not be resolved in this
77 analysis.

78 Dilution effects have received less research attention to date when compared
79 with the enhancement of gas transfer due to rain, although Turk *et al.* (7) provide
80 initial experimental evidence that dilution affects regional sea-air CO₂ flux.

81 Salinity gradients in the top few meters of the ocean surface due to freshwater
82 input from rain have been studied for their influence on remotely sensed salinity
83 measurements (8), and can be related to rain rate (9). Santos-Garcia *et al.* (10)
84 present a physics-based model that draws on very high resolution modeled
85 precipitation estimates (NOAA CMORPH) and surface wind data in order to
86 predict surface stratification due to rain. However, the spatial and temporal
87 resolution required to resolve individual rain events is not currently compatible
88 with the global data sets used in the calculations presented here.

89 Previous studies of the type presented here include Komori *et al.* (11), who
90 accounted for enhanced transfer velocity using results from their laboratory
91 tests and direct wet-deposition. They estimated that the global effect of rainfall
92 on net sea-air fluxes for the year 2001 was to increase the sink of atmospheric
93 CO₂ by <5%. Following this, Turk *et al.* (7) incorporated laboratory-derived
94 parameterisations of wet deposition, rain-enhanced k and surface pCO_2 dilution,
95 into flux estimates for a single location in the Western Equatorial Pacific. When
96 extrapolated across the region, these point values indicated an increased uptake
97 of CO₂, with the net flux in the Western Equatorial Pacific Ocean changing from a
98 source to a sink. The findings highlight the significant role that rain can play,
99 particularly in regions characterised by low winds and high precipitation and
100 support the need for the global and regional impact of rain to be considered in
101 gas flux studies.

102 In synopsis, previous work studying the impact of rain on sea-air fluxes of CO₂
103 has focussed on laboratory studies (1-3, 5, 12), localised field studies, including

104 the Biosphere 2 model ocean (1, 4, 12), and the use of one-dimensional
105 numerical-models (3, 4, 7). Regional and global estimates of integrated net sea-
106 air fluxes have largely ignored the impact that rain can have, and most global
107 studies do not account for rain within their uncertainty analyses. The exception
108 to this is the work of Komori *et al.* (11) who applied laboratory-derived
109 parameterisations to study global sea-air fluxes for a single year (2001) using
110 model, climatology and Global Precipitation Climatology Project (GPCP) data.

111 The FluxEngine software tool offers an efficient mechanism to exploit up to 20
112 years of Earth observation (EO) and blended EO and in situ data, in order to
113 calculate global and regional estimations of sea-air CO₂ flux (6). Global gas
114 exchange requires a large and complex set of calculations. Variations or errors in
115 these calculations can hinder intercomparison between studies and are difficult
116 to identify without interrogating actual calculations procedures. The FluxEngine
117 has been created to provide a consistent set of calculations that reduce the
118 repeated effort required for studies in this field. It has been extensively verified
119 against known datasets to provide a common baseline for the international
120 community, such that its use minimises errors and helps maintain consistent
121 analysis between studies. The software tool and associated publications are open
122 access and can be accessed through the project website ([www.oceanflux-](http://www.oceanflux-ghg.org)
123 [ghg.org](http://www.oceanflux-ghg.org)). The source code is also open source. It is continually updated to keep
124 up with advances in the field and can be downloaded here,
125 github.com/oceanflux-ghg/FluxEngine.

126 The work in this paper uses FluxEngine to build upon and extend the work of
127 Komori *et al.* (11) by applying recent advances, parameterisations and tools in
128 order to characterise the potential global and regional impacts that rain can have
129 on the different components of the sea-air flux calculation. The components
130 considered are rain-induced gas transfer velocity and the direct wet deposition
131 of CO₂ by raindrops landing on the ocean surface. Results are presented as
132 monthly and annual net fluxes for global and regional seas, providing an inter-
133 annual and seasonal assessment of the net impact of rain on global flux of CO₂.
134 These estimates are driven by two different CO₂ climatologies, that presented by

135 Takahashi *et al.* (13), and that provided by SOCAT (14). These climatologies are
136 referenced to single years, 2000 and 2010 respectively. As such, inter-annual
137 variability is estimated solely through changes in sea surface temperature (SST),
138 wind and rainfall, and does not reflect changes in $p\text{CO}_{2w}$. Inter-annual results are
139 analysed in terms of the sensitivity of the global and regional estimates to rain,
140 identifying regions where rain can have a significant impact on sea-air CO_2 gas
141 exchange, whilst acknowledging the unknown effect of changes in $p\text{CO}_{2w}$. The
142 final part of the paper includes a discussion of the impact of rain-driven dilution
143 of the surface layer, including an initial analysis of the impact of rain-driven
144 variations in the sea skin temperature (SST_{skin}).

145 **2.0 Methods**

146 The global impact of rain on sea-air CO_2 fluxes is studied using monthly, multi-
147 year data. The following sections describe the datasets used as well as the
148 methods for calculating the monthly sea-air CO_2 fluxes, the rain-driven gas
149 transfer velocity and the wet deposition of CO_2 . Calculations were undertaken
150 using the FluxEngine open source processing toolbox (6). This toolbox allows
151 users to easily parameterize and generate global and regional sea-air CO_2 flux
152 estimates. For this study the toolbox was extended to allow rain induced transfer
153 and wet deposition to be included in the air-sea gas flux parameterisation. Here
154 we study the four major ocean basins, Atlantic, Indian, Pacific and Southern.
155 Detailed definitions of these regions, verification of the system and the range of
156 configurations available are presented in Shutler *et al.* (6).

157 **2.1 Datasets**

158 To characterise the sea surface, we first used satellite EO data from the European
159 Space Agency (ESA) Sea Surface Temperature Climate Change Initiative data
160 (version 1.1.1) for SST_{skin} , (K) (15) and ESA GlobWave for wind speed at 10 m,
161 U_{10} (m s^{-1}) (16). Both of these datasets are calibrated, bias corrected, well-
162 characterised with known uncertainties and designed for use in climate studies.
163 For ice cover we use satellite based, Special Sensor Microwave Imager (SSM/I)
164 global percentage ice cover data (17, 18). These datasets have been re-gridded

165 onto a $1^\circ \times 1^\circ$ grid where each grid value was the statistical mean of all
 166 contributing data (6). For surface salinity (S), we use the World Ocean Atlas
 167 salinity data provided within Takahashi *et al.* (13). For in-water pCO_2 (pCO_{2W}) we
 168 use two different data sets. Firstly, the climatological data from Takahashi *et al.*
 169 (13) with a reference year 2000 and an estimated global increase in pCO_{2W} of 1.5
 170 $\mu\text{atm yr}^{-1}$ (eq. 5). Moving further away from this reference year, the estimated
 171 temporal correction for the CO_2 climatology becomes less robust. Thus, the study
 172 using (13) was limited to the years 1999 – 2006, where the correction is most
 173 appropriate. The flux estimates are expected to be strongly dependent on the
 174 accuracy of the pCO_{2W} climatological data. In order to examine this sensitivity, an
 175 alternative climatological pCO_{2W} dataset was also used, which is derived from the
 176 SurfaceOcean CO_2 Atlas (SOCAT) (14). Notably here, the reference year is 2010.
 177 The timescales between these two pCO_{2W} data sets do not match, but they do
 178 overlap, allowing the impact of the choice of pCO_{2W} dataset to be determined.
 179 We calculate atmospheric pCO_2 (pCO_{2A}) using modelled air pressure (P) and
 180 climatological concentration of CO_2 in dry air ($X_{CO_{2A}}$) from the NCEP CFSR model
 181 (13). The pCO_{2W} , P and $X_{CO_{2A}}$ data were linearly interpolated to the same $1^\circ \times 1^\circ$
 182 grid as the other datasets. For rain rate we used the daily $1^\circ \times 1^\circ$ GPCP, version
 183 2.2 (19). There is still considerable debate about the absolute magnitudes of the
 184 global distribution of precipitation and its seasonal variation (20, 21), although
 185 the GPCP dataset is widely accepted as one of the most reliable.

186 **2.2 Sea-air CO_2 flux**

187 The sea-air flux of CO_2 (F , $\text{g m}^{-2} \text{s}^{-1}$), is calculated using the product of a gas
 188 transfer velocity, k (m s^{-1}), and the difference in CO_2 concentration (g m^{-3})
 189 between the base [CO_{2AQW}] and the top [CO_{2AQ0}] of a thin (~ 10 to $250 \mu\text{m}$)
 190 boundary layer at the sea surface:

$$191 \quad F = k \left([CO_{2AQW}] - [CO_{2AQ0}] \right) \quad (1)$$

192 The concentration of CO_2 in seawater is the product of its solubility, α (g m^{-3}
 193 μatm^{-1}), and its fugacity, fCO_2 (in μatm). Gas solubility is a function of salinity and

194 temperature and as such, it varies across the aqueous boundary layer. Equation
195 (1) then becomes:

$$196 \quad F = k (\alpha_w fCO_{2W} - \alpha_s fCO_{2A}) \quad (2)$$

197 where the subscripts denote values in water (*W*), at the sea-air interface (*S*) and
198 in air (*A*). For simplicity we can substitute partial pressure for fugacity because
199 their values differ by <0.5% over the temperature range considered (22).

200 Therefore we estimate the sea-air flux using:

$$201 \quad F = k (\alpha_w pCO_{2W} - \alpha_s pCO_{2A}) \quad (3)$$

202 Climatological estimates of pCO_{2W} (pCO_{2Wclim}) must be adjusted to the SST for the
203 period of study. Following previous studies (23-25), the pCO_{2W} values were
204 corrected to reflect SST using the relationship provided by Takahashi *et al.* (13):

$$205 \quad pCO_{2W} = pCO_{2Wclim} (\exp (0.0423(SST - T_{clim})) - 4.35 \times 10^{-5} [SST^2 - T_{clim}^2]) \quad (4)$$

206 where T_{clim} is the temperature from the Takahashi *et al.* (13) climatology in °C,
207 and SST is estimated as $SST_{skin} + 0.17$ and converted to °C (26).

208 pCO_{2A} (in μatm) was calculated by including a global average increase of 1.5
209 $\mu\text{atm yr}^{-1}$ using: (13)

$$210 \quad pCO_{2A} = X_{CO2A}(P - pH_2O) + 1.5(y - 2000) \quad (5)$$

211 where y is the year, P is the daily average air pressure (in μatm), X_{CO2A} is the
212 zonal mean molar fraction of CO_2 in the dry atmosphere (in parts per million)
213 and pH_2O is the saturation vapour pressure in μatm (27):

$$214 \quad pH_2O = 1013.25 \exp[24.45 - (67.45(100/SST_k)) - (4.85 \ln(SST_k/100)) - 0.00054S] \quad (6)$$

216 where salinity, S is on the Practical Salinity Scale and air temperature, and SST_k is
217 subskin sea surface temperature in Kelvin.

218 **2.3 Rain impacts**

219 The sea-air flux due solely to wet deposition is estimated using (11):

$$220 \quad F_{DIC} = -Rn \alpha pCO_{2A} \quad (7)$$

221 where Rn is the rain rate in mm h^{-1} and α is the solubility of CO_2 in fresh water,
 222 calculated for local air temperature, but with salinity set to 0, using the
 223 formulation in Wanninkhof (28).

224 Initial laboratory experiments derived a linear increase in the transfer velocity
 225 during rain events, dependent on Rn , (1)

$$226 \quad k_{total} = k_{wind} + k_{rain} \quad (8a)$$

227 where,

$$228 \quad k_{rain} = (0.929 + 0.679 Rn - 0.0015 Rn^2) \quad (8b)$$

229 Recent work has shown how the rain influences the gas transfer velocity in a
 230 nonlinear fashion (5). Therefore, the total gas transfer velocity (k_{total}) due to
 231 wind and rain is defined as:

$$232 \quad k_{total} = k_{wind} + [1 - \exp(-a\beta)] k_{rain} \quad (9)$$

233 where $a = 0.3677$ and $\beta = KEF_r / KEF_w$, where KEF_r is the kinetic energy flux due
 234 to rain, and KEF_w is that imparted to the water by surface winds. Harrison *et al.*
 235 (5) assume a Laws-Parsons raindrop-size distribution to derive a simplified
 236 relationship, $KEF_r = 0.0112Rn$ and define $KEF_w = \rho_a u^{*3}$, where ρ_a is the density of
 237 air (in kg m^{-3}) defined as $\rho_a = P / (R SST_k)$, where P , is the air pressure (in Pa) and
 238 R is the specific gas constant for dry air (in $\text{J kg}^{-1} \text{K}^{-1}$). The friction velocity u^* (in
 239 ms^{-1}) is given by $u^{*2} = C_D U_{10}^2$, where C_D is the drag coefficient as defined by
 240 Yelland and Taylor (29).

241 The wind speed parameterised gas transfer velocity, k_{wind} , was estimated
 242 following the method in (13) such that, $k_{wind} = 0.26(U_{10})^2 (Sc/660)^{-1/2}$, where Sc
 243 represents the Schmidt number of the gas in question. k_{wind} was used to calculate
 244 a reference flux, F_{ref} , in which no contribution from rain was included:

$$245 \quad F_{ref} = k_{wind} (\alpha_W pCO_{2W} - \alpha_S pCO_{2A}) \quad (10)$$

246 The $[1 - \exp(-a\beta)] k_{rain}$ term in equation (9) represents the enhancement of the
 247 gas transfer velocity due to rain rate through a non-linear relationship with wind
 248 speed. The combined wind and rain sea-air CO_2 flux is then given by:

249 $F_{k-rain} = k_{total} \alpha_W (pCO_{2W} - pCO_{2A})$ (11)

250 The total sea-air flux, F_T , that includes the rain impacts described above is then
251 the sum of the gas transfer and the wet deposition components:

252 $F_T = F_{k-rain} + F_{DIC}$ (12)

253 The contributions from rain effects were then calculated as the difference
254 between the rain affected flux values, (F_T , F_{k-rain} and F_{DIC}) and F_{ref} . In this study, F
255 values represent the sea-air CO₂ flux. Positive values represent an outgassing of
256 CO₂ from the ocean to the atmosphere, whilst negative values represent a
257 transfer (sink) of CO₂ from the atmosphere to the oceans.

258 **2.4 Integrated net sea-air fluxes**

259 Integrated fluxes over a given region are calculated from the monthly mean flux
260 at each pixel, adjusted for ice and the pixel's total area, which is calculated
261 assuming the Earth to be an oblate spheroid. Missing data values are accounted
262 for using a regional average and added to the integrated net flux from valid data
263 values, to give an estimate of the total regional integrated net flux (6). Global
264 values are estimated by treating the entire globe as a single region. We refer the
265 reader to the (6) for a detailed description of the integrated net flux tool which is
266 part of the FluxEngine.

267 **2.5 Uncertainties**

268 An ensemble approach was adopted to assess the uncertainties in F_T . Random
269 errors were used to perturb input data for multiple runs, according to known
270 variability in the input data sets. Uncertainties in the rain data set are provided
271 through the GPCP (19) as a variance for each datum, σ_i^2 , which includes both
272 algorithm and random sampling errors that can vary in time and space. Bias is
273 considered to be zero (19). Using the values presented in Land *et al.* (25), the
274 variabilities of U_{10} , SST and pCO_{2W} were estimated as published global standard
275 deviations that do not vary in time or space.

276 Following the method used by Land *et al.* (25), a random noise signal was
277 generated for each parameter and used to perturb the input data. For rain rates,

278 noise was added by using a value drawn at random from a normal distribution
279 with mean X_i equal to the original value and standard deviation σ_i equal to the
280 uncertainty value provided, $N(X_i, \sigma_i)$. Resulting rain rates less than zero were set
281 to 0. For U_{10} , SST and pCO_{2w} , noise was added by using a value drawn at random
282 from a log-normal distribution, with a (natural) log mean equal to the log of the
283 original data point and the published log standard deviation, $\exp[N(\ln X_i, \sigma)]$.
284 This process was repeated with 10 different perturbations for the year 2000,
285 producing 10 separate sets of monthly and annual results. The uncertainty
286 estimates provided here are the standard deviation of results across these 10
287 runs. The cumulative effect of the uncertainties on each parameter was used as
288 an indication of total uncertainty in the resulting flux. This assumes that all of the
289 errors are uncorrelated. In reality, there will be some inter-dependence between
290 the input parameters, which will affect the stated errors in CO₂ flux.

291 **3.0 Results**

292 The following sections present the results for the global oceans and the
293 individual oceanic basins.

294 **3.1 Annual integrated net sea-air CO₂ fluxes**

295 For the CO₂ climatology reference year (2000), the estimated annual global sea-
296 air CO₂ flux without any rain impacts is -1.4 Pg C yr⁻¹. The FluxEngine has been
297 validated with previous outputs in this research field (6) and these annual net
298 integrated values are consistent with the original publication of Takahashi *et al.*
299 (13). When the CO₂ climatology was used to study subsequent years (eq. 5),
300 applying SST, wind and other data from each year, the annual values are
301 consistently negative and between -1 Pg C yr⁻¹ and -1.6 Pg C yr⁻¹, i.e. the global
302 ocean is a net sink of CO₂ (Figures. 1, 2 & Table 1). Global estimates are
303 significantly lower during the years 1999 and 2000, meaning the net sink of CO₂
304 is at its greatest. Notably, these years correspond to a strong La Niña event.

305 Figure 1. Mean monthly CO₂ flux between January 1999 and December 2005 for
306 a reference dataset (no rain components)

307 Figure 2. Annual (right axis, solid lines) and monthly (left axis, dashed lines)
 308 global net sea-air CO₂ flux, without the effects of rain, F_{ref} , and with the effects of
 309 rain, $F_T = F_{DIC} + F_{k-rain}$.

310 The change to global net sea-air flux due to direct wet deposition of CO₂, F_{DIC} ,
 311 varies from -60 to -64 Tg C yr⁻¹ (Figures. 3,4 & Table 1). The effect of rain
 312 enhancing gas transfer velocity, for a non-linear model (eq. 9), varies from 3 to 6
 313 Tg C yr⁻¹ (Figures. 3,5 & Table 1). Assuming a linear sum of these components
 314 gives an effect on annual global net sea-air CO₂ flux of -56 to -58 Tg C yr⁻¹
 315 (Figures. 3,6 & Table 1). When compared to the estimated annual net integrated
 316 sea-air CO₂ flux without any rain impacts, this equates to an increase in the
 317 global oceanic CO₂ sink of 3.5 to 6%.

318 Figure 3 The monthly mean global CO₂ flux attributed to the enhancement of
 319 transfer velocity (both non-linear, F_{rain-k} and non-linear, $F_{rain-k (linear)}$) and Direct
 320 deposition, F_{DIC} , TgC month⁻¹

321 Figure 4. The mean effect of wet deposition on monthly CO₂ flux between January
 322 1999 and December 2005, $F_{DIC} - F_{ref}$

323 Figure 5 The effect of rain on monthly CO₂ flux between January 1999 and
 324 December 2005, given a non-linear model of transfer velocity (eq. 9), $F_{k-rain} - F_{ref}$.

325 Figure 6. The combined effect of wet deposition and non-linear gas transfer
 326 velocity on CO₂ flux between Jan 1999 and Dec 2005, $(F_{DIC} + F_{k-rain}) - F_{ref}$.

327

Table 1. Annual global integrated net flux, F_T (Tg C yr⁻¹) with and without rain (left hand columns) and the impact of each rain component on F_T (Tg C yr⁻¹) (right hand columns), where $F_T = F_{DIC} + F_{k-rain}$, (non-linear).

	Net CO2 Flux, F , Tg C yr ⁻¹		Effect on CO2 flux, ΔF , Tg C yr ⁻¹			
	Reference, F_{ref}	F_T	F_T	F_{DIC}	F_{k-rain} , linear	F_{k-rain} , non-linear
1999	-1584.24	-1640.88	-56.64	-59.70	-27.76	3.06
2000	-1427.66	-1484.11	-56.45	-60.47	-18.29	4.03
2001	-1094.42	-1150.03	-55.61	-60.08	-2.45	4.47
2002	-1097.07	-1153.14	-56.07	-62.85	2.56	6.78
2003	-1057.83	-1114.33	-56.50	-62.40	4.27	5.90
2004	-997.91	-1053.65	-55.74	-62.19	3.96	6.45
2005	-1014.53	-1073.01	-58.48	-63.59	2.67	5.11
2006	-1122.76	-1180.59	-57.83	-63.67	2.13	5.84

328

329 Comparison of flux estimations between those made using a linear relationship
 330 between wind and rain (eq. 8) and those made with a non-linear relationship

331 (eq. 9) showed notably different results. The non-linear parameterisation
 332 decreased the oceanic CO₂ sink, whilst the linear parameterisation increased the
 333 oceanic CO₂ sink (Figures, 3 & 7). The linear parameterisation also exhibited
 334 seasonal variations with magnitude up to 10 times greater than the non-linear
 335 parameterisation (Figure 3 & Table 2). A similar effect was observed in average
 336 global flux, where the linear parameterisation showed significantly higher
 337 geographic variability, ranging from an average of -0.4 to 0.4 Tg C month⁻¹,
 338 compared to -0.02 to 0.02 Tg C month⁻¹ for the non-linear parameterisation.
 339 Following recommendations in Harrison *et al.* (5), the non-linear
 340 parameterisation was adopted and the remaining results in this paper that
 341 include k_{rain} or F_{k-rain} refer to the non-linear parameterisation (eq. 9).

342 Figure 7. The mean effect of rain on monthly CO₂ flux between January 1999 and
 343 December 2005, given a linear model (Ho 2004), $F_{k-rain(linear)} - F_{ref}$. Note different
 344 scale compared to figures 4,5 and 6.

Table 2. Monthly global integrated net flux, F_T (Tg C yr⁻¹) with and without rain (left hand columns) and the impact of each rain component on F_T (Tg C yr⁻¹) (right hand columns), where $F_T = F_{DIC} + F_{k-rain}$ (non-linear).

Month	Net CO2 Flux, F_T Tg C mnth ⁻¹		Effect on CO2 flux, ΔF , Tg C mnth ⁻¹			
	Reference, F_{ref}	F_T	F_T	F_{DIC}	$F_{k-rain, linear}$	$F_{k-rain, non-linear}$
Jan	-139.08	-143.92	-4.84	-5.33	-1.31	0.49
Feb	-114.47	-118.83	-4.37	-4.76	-0.14	0.40
Mar	-119.72	-124.62	-4.90	-5.49	0.00	0.59
Apr	-112.40	-117.11	-4.71	-5.21	-0.58	0.50
May	-117.18	-122.16	-4.97	-5.33	-2.09	0.35
Jun	-87.63	-92.30	-4.67	-4.99	-1.64	0.32
Jul	-53.22	-57.80	-4.58	-5.02	1.22	0.44
Aug	-27.04	-31.72	-4.68	-5.11	2.29	0.43
Sep	-21.46	-25.93	-4.47	-4.96	2.33	0.49
Oct	-76.81	-81.46	-4.66	-5.13	0.44	0.47
Nov	-135.23	-140.01	-4.78	-5.11	-2.19	0.33
Dec	-166.98	-171.88	-4.90	-5.30	-2.43	0.39

345

346 3.2 Spatial Variability

347 In general, F_{DIC} dominates the combined effect of rain on CO₂ sea-air flux and the
 348 global distribution follows that of the precipitation estimates (Figure 4).

349 However, the strongest reductions in sea-air CO₂ flux were in higher latitudes,
 350 where k_{rain} and wet deposition combined and a cumulative reduction in sea-air
 351 flux was observed. Reductions in sea-air flux were also observed in tropical
 352 regions with high rainfall, which represent an increase in the estimated oceanic
 353 sink of CO₂. In tropical areas with lower rainfall, an increase in net sea-air
 354 transfer was observed, decreasing the estimated oceanic sink of CO₂ (Figure 6).

355 Regional Analysis

356 Table 3 provides the estimated sea-air CO₂ flux for the four regions, representing
 357 the main oceanic basins. The effect of rain on gas transfer alters the annual
 358 regional oceanic basin net integrated sea-air CO₂ flux by -0.03 to 4.5 Tg C yr⁻¹ and
 359 is primarily positive, decreasing the oceanic sink. Wet deposition alters the
 360 annual regional oceanic basin net integrated sea-air CO₂ transfer by -2 to -32 Tg
 361 C yr⁻¹, increasing the oceanic sink of CO₂.

Table 3. Annual integrated net flux with rain components, F_T (Tg C yr⁻¹) from 1999 – 2006, for each of the ocean basins, and the impact of each rain component on F_T (Tg C yr⁻¹), where $F_T = F_{DIC} + F_{k-rain}$ and all-rain = $F_T - F_{ref}$.

Year	Atlantic				Indian				Pacific				Southern			
	F_T	All-rain	F_{DIC}	F_{k-rain} non-linear	F_T	All-rain	F_{DIC}	F_{k-rain} non-linear	F_T	All-rain	F_{DIC}	F_{k-rain} non-linear	F_T	All-rain	F_{DIC}	F_{k-rain} non-linear
99	-493	-13.2	-14.0	0.8	-355	-11.4	-12.3	0.9	-686	-28.3	-29.6	1.3	-68.1	-2.4	-2.4	-0.02
00	-504	-13.1	-13.7	0.6	-334	-11.6	-12.6	1.0	-517	-28.1	-30.4	2.3	-91.8	-2.3	-2.3	-0.03
01	-473	-12.5	-13.1	0.6	-270	-10.7	-11.6	0.9	-326	-27.3	-30.0	2.7	-16.9	-2.0	-2.0	-0.01
02	-564	-13.7	-14.2	0.6	-290	-11.4	-13.0	1.6	-198	-27.2	-31.7	4.5	-69.9	-2.3	-2.3	-0.02
03	-469	-13.5	-14.3	0.8	-322	-11.8	-13.3	1.5	-281	-27.7	-31.2	3.5	-15.9	-2.2	-2.2	-0.01
04	-447	-13.4	-14.0	0.6	-362	-11.2	-12.8	1.6	-182	-28.0	-31.6	3.6	-36.9	-1.7	-2.2	0.01
05	-385	-13.9	-14.7	0.8	-430	-12.1	-13.4	1.3	-182	-28.3	-31.3	2.9	-47.8	-2.6	-2.5	-0.01
06	-377	-13.7	-14.5	0.8	-446	-11.8	-13.2	1.4	-296	-28.1	-31.6	3.5	-33.4	-2.9	-2.9	-0.02

362 Regionally, during 1999 and 2000, the Pacific Ocean shows the most negative
 363 sea-air flux values, $F_T = -686$ and -517 Tg C yr⁻¹, respectively (Table 3). However,
 364 subsequent years show this reducing by approximately 60% to a 2001-2006
 365 mean of $F_T = -244$ Tg C yr⁻¹, meaning that between 2002 and 2006, less CO₂ is
 366 absorbed by the Pacific Ocean than both the Atlantic and Indian Oceans. The
 367 substantial differences between global flux during 1999-2000 and subsequent

368 years (Figure 2 & Table 1), particularly evident in results from the Pacific (Table
369 3), agree with previous results (30). The values for $p\text{CO}_2$ have been fixed by the
370 climatology (eq. 5) and variations in overall flux estimations can be attributed to
371 changes in wind speed and water temperature during these years. The observed
372 differences in 1999-2000 are likely to be related to the strong La Niña event
373 during this time.

374 All regions consistently show an overall reduction in annual CO_2 flux due to rain
375 effects, increasing the oceanic CO_2 sink. The Pacific reduction is the strongest,
376 varying between 5% and 15% of the total estimated flux from this region. The
377 change in estimated CO_2 flux in the Atlantic is approximately half the magnitude
378 of that in the Pacific, with that in the Indian Ocean slightly less again. As such,
379 rain effects comprise between 2.4 and 4% of annual net flux in the Atlantic and
380 Indian ocean basins, increasing the oceanic CO_2 sink. The Southern Ocean
381 exhibits the smallest net change to CO_2 flux due to rain effects. However, due to
382 the low net total CO_2 flux in this region, the predicted changes due to rain
383 represent between 3% and 14% of total flux, again increasing the oceanic sink of
384 CO_2 .

385 **3.3 Temporal variability**

386 The monthly estimated global net CO_2 flux shows a strong, consistent seasonal
387 cycle (Figure 2). The influence of rain is again dominated by wet deposition, F_{DIC} .
388 The global net influence of F_{DIC} varies between -5 and -5.5 Tg C month⁻¹ (Table 2),
389 representing an increase in the oceanic CO_2 sink each month. During September,
390 this represents 20% of the global net flux, F_{ref} , reducing to 3% for December. A
391 seasonal pattern can be observed in F_{DIC} , although this is not consistent for all
392 years.

393 The influence of the non-linear k_{rain} term is between 0.3 and 0.5 Tg C month⁻¹ and
394 increases net flux, which represents a decrease in the oceanic CO_2 sink (Figures
395 3,5 & Table 1). This represents between 0.2% of total flux during December and
396 2% during September.

397 **3.4 Errors and uncertainty**

398 Random noise was used to perturb the input data for the year 2000, as described
399 in section 2.5. Due to the perturbation of input signals with random noise, annual
400 global CO₂ flux values varied with a standard deviation of 0.7 Tg C month⁻¹, taken
401 across all 10 ensemble runs. This represents 0.7% of the estimated net
402 integrated global flux values, which is of the same order as the 0.5 % random
403 error reported in CO₂ sea-air fluxes in the Arctic seas (25).

404 **3.5 Comparison with SOCAT Climatology**

405 Estimations of F_T using both rain effects, F_{DIC} and k_{rain} , were repeated with the
406 SOCAT pCO_{2w} climatological data (14) replacing that of (13). The SOCAT
407 reference year is 2010 and the trend in equation 5 was applied moving back in
408 time from 2010. As such, the two climatologies represent global pCO₂ values for
409 different years, adjusted for changes due to increased levels of atmospheric CO₂,
410 but inter-annual variability is not explicitly resolved. Furthermore, pCO_2 values
411 from Takahashi *et al.* (13) have been smoothed to best represent idealised non
412 El-Niño conditions, whilst the SOCAT derived data set does not include such
413 adjustments (14). Estimates were made between 2004 and 2006 both with rain
414 effects, k_{rain} and F_{DIC} and without. During this period, results using SOCAT give an
415 average CO₂ flux of -1600 Tg C yr⁻¹ compared to -1120 Tg C yr⁻¹ using (13).
416 Importantly for this study, the estimated effect of rain on annual net integrated
417 CO₂ transfer was in general agreement, with an average difference in global CO₂
418 transfer of 57 Tg C yr⁻¹ using (13) and 42 Tg C yr⁻¹ using SOCAT. The effect of rain
419 was to increase the oceanic sink in both cases.

420 **4.0 Discussion**

421 In this work, the choice of a linear or non-linear parameterisation of the relative
422 importance of wind and rain on gas transfer velocity is shown to have significant
423 impact on the estimation of CO₂ flux (Figure 3). Using a non-linear term (5), both
424 temporal and spatial variability are diminished and the average net CO₂ transfer
425 is decreased. Thus, relative to the previous linear parameterisation, importance

426 of rain for gas transfer at a global level is diminished, meaning that F_{DIC} is the
427 more important process for the impact of rain on CO₂ flux between the ocean and
428 air.

429 This research provides a comprehensive global study into the effect of rain.
430 However, the practicalities of capture, processing and storage of global data sets
431 mean that it is often necessary to compromise on spatial and/or temporal
432 resolution. In the case of the rain data from GPCP, the global data set is available
433 as monthly averages in mm day⁻¹, averaged spatially over 1° x 1°. At these scales,
434 it is not possible to resolve intense episodic or extreme events. This raises three
435 areas for consideration. Firstly, the transfer velocity during a single day of heavy
436 rain within a month will not be equivalent to that calculated using equation 8b,
437 based on a monthly average rain rate. Secondly, as discussed above, the lack of
438 knowledge of actual rain rates during these episodes will prevent the direct
439 estimation of the extent of temporary surface dilution and its impact on gas
440 exchange. Finally, correlation between wind and rain within the month will also
441 affect the gas exchange and again, cannot be predicted.

442 Taking these three areas in turn, the first is surface dilution. Rain falling onto the
443 ocean will influence the chemical properties of surface waters. As such, it could
444 decrease the pCO_{2W} and directly affect CO₂ exchange. Observational studies from
445 Biosphere ocean experiments provide evidence for the formation of freshwater
446 layers (3, 4). There is also in-situ evidence from the Pacific region, with direct
447 measurements of decreased salinity at the surface during and after rain events
448 (31). These sources identify a peak in the freshwater layer after approximately 1
449 hour of persistent rain and highlight changes in surface stratification up to two
450 days after the rain event.

451 Experimental data to estimate the effect of surface dilution on CO₂ exchange exist
452 and Turk *et al.* (7) consider dilution for a point in the Western Equatorial Pacific.
453 The same temporary changes to surface water composition have been seen to
454 affect remotely-sensed salinity measurements (8) and methods have been
455 proposed to relate these to rain rate (9), or use physical modeling to predict
456 their existence, in order to better understand variability in remote sensing data.

457 In theory, such methods could be applied to predict the impact of freshwater
458 layers on gas exchange. However, the spatial and temporal scales of the
459 estimates made here are limited by the global data sets and are not sufficient to
460 resolve individual events.

461 In addition to chemical dilution, rain falling on the sea surface could affect *SST*.
462 Gas solubility is a function of salinity and temperature and changing *SST* will
463 affect the CO₂ balance across the surface, altering exchange between air and
464 atmosphere through equations 10 and 11. The high temperature dependency of
465 pCO_{2w} suggests that this could be an important process to consider (32, 33).
466 Gosnell *et al.* (34) used a modeling study to investigate the relative temperature
467 of the rain to the sea surface through estimations of the changing temperature of
468 raindrops. In their experiment, a maximum 0.2K difference occurs at maximum
469 rain rates (100 mm hr⁻¹) from maximum height (5000 m). As an initial
470 investigation, temperature differences were applied to the *SST* input data for the
471 reference year 2000. A constant bias of 0.2 K was subtracted from surface *SST* for
472 calculations where the rain rate exceeds 1 mm hr⁻¹. The observed differences
473 were negligible, resulting in the flux being altered (reduced or increased) by up
474 to 0.02% of monthly regional integrated net CO₂ flux. These results imply that
475 the rain-induced temperature differences have a negligible effect on the air-sea
476 gas fluxes and significantly less than the total uncertainties (0.7%) calculated in
477 section 3.3.

478 Global rain rate retrievable through the GPCP is the monthly average for a 1° x 1°
479 grid square. Here, this has been used in equations 8 & 9 to calculate the gas
480 transfer velocity, k as a combination of wind, k_{wind} and rain, k_{rain} , assuming
481 constant and consistent rain rate throughout the area and throughout the month.
482 In reality, the rain will fall at varying rates during a month and within a grid
483 square, which will cause variability in k_{rain} , as well as the ratio of kinetic energy
484 flux between wind and rain, β , which governs the contribution of k_{rain} . Heavy rain
485 for two days in a month and no other rain, will not affect k_{total} by the same
486 magnitude as the same rain spread over the month. However, Rn used in these
487 studies will be the same and the temporal (or spatial) variability cannot be

488 accounted for. In order to examine how this will affect the overall outcome, the
489 $[1 - \exp(-a\beta)] k_{rain}$ term in equation 8 was calculated for an example average
490 monthly rain rate of 1mm/hr spread over a varying number of days in a month.
491 At low wind speeds ($u < 10 \text{ m s}^{-1}$), spreading the rain over the month (as is
492 assumed with a monthly mean) gives a higher estimate for k_{rain} , and
493 subsequently, k_{total} , than shorter duration heavier rain. However, at higher wind-
494 speeds ($u > 10 \text{ m s}^{-1}$), the β ratio means that low rain rates are estimated to have
495 little effect. As such, the shorter duration, heavier rain produces a higher
496 estimate for the influence of rain on k_{total} . This means that, the methodology may
497 be under-estimating the effect of rain in higher latitudes and overestimating in
498 lower latitudes. Nevertheless, F_{k-rain} has an impact on the overall flux that is a
499 factor of 10 smaller than that of direct deposition, F_{DIC} , limiting the overall
500 impact of variability on global results. When examining regionally, the effect will
501 become more important and in the future, more detailed data for the pattern of
502 rainfall would be beneficial, particularly to small regional studies in areas where
503 F_{k-rain} is relatively important, such as those with high rainfall and low wind
504 speeds..

505 Within a grid-square and during a month, there will also be variability in the
506 wind strength. It is a combination of wind and rain rate will govern the extent
507 and duration of surface dilution, as well as the effect of temporal and spatial
508 variability. Typically, in mid-latitudes, rain events are associated with developing
509 low pressure systems and so correspond with stronger winds than average for a
510 region. In tropical latitudes, precipitation occurs both in storms and in large
511 convective systems with relatively gentle low-level convergence (Plate 4 of
512 Quartly *et al.* (35)). More recently, Quartly *et al.* (36) confirmed the
513 predominance of rain at low wave heights for a number of regions in the Atlantic,
514 with, for some seasons, rainfall in mid-latitudes being roughly five times as likely
515 at low sea state than at high. Further work should look towards measuring the
516 instantaneous relationship between wind and rain. There are a number of
517 different remote-sensing technologies that can make estimates of the rain rate at
518 the Earth's surface. Dual-frequency altimeters can provide simultaneous
519 estimates of wind speed, wave height and rain rate. These could support studies

520 of the correlation of these conditions (35), or even direct measurement of the β
521 ratio

522 It must also be noted that here we have assumed that the GPCP precipitation
523 data characterises only rainfall, whereas precipitation also includes sleet, ice and
524 snow. However, we have no information on how much snow and sleet falls
525 globally each year so the impact of this assumption is unknown.

526 **5.0 Conclusions**

527 This paper has presented analysis of the impact of rain on global and regional
528 sea-air CO₂ fluxes and the oceanic net sink of CO₂. The work has exploited the
529 open source FluxEngine software, cloud computing, advanced methods for
530 estimating rain induced sea-air gas fluxes and an extensive dataset of climate
531 quality satellite Earth observation, in situ, model and re-analysis data.

532 The results demonstrate a non-negligible effect of rain when estimating global
533 and regional integrated net sea-air CO₂ fluxes. Differences of approximately 6%
534 in annual global CO₂ flux have been estimated, which means that rain serves to
535 increase the oceanic CO₂ sink.

536 Implementing the non-linear relationship between rain and wind, as
537 recommended by Harrison *et al.* (5), over the linear relationship originally
538 proposed by Ho *et al.* (12), significantly reduces the spatial and temporal
539 variability with which rain enhancement of gas transfer rate affects CO₂ flux. This
540 serves to diminish the importance of this rain induced gas transfer in the effect
541 of rain on integrated net sea-air CO₂ fluxes

542 Globally, the observed changes are dominated by the influence of wet deposition,
543 F_{DIC} . The influence of rain varies regionally and is greatest in the Pacific Ocean
544 where it represents up to 15% of the annual regional net flux, and up to 50% of
545 monthly net flux. It is also important in the Southern Ocean, due to the low
546 overall CO₂ sink estimate, where it represents 13% of annual net flux. Regional
547 fluxes are more variable, with up to 16% modulation of the annual integrated net

548 CO₂ flux due to rain, which can be responsible for turning the region from a net
549 source to a net sink.

550 Therefore we conclude that the impacts of rain should be included in the
551 uncertainty analysis of studies that estimate integrated net sea-air fluxes of CO₂.
552 However, for regional or short-term studies, results suggest that rain can have a
553 considerable impact on the fluxes, dependent upon the region and timescale and
554 may need to be considered directly in sea-air CO₂ flux estimates.

555 Three key limitations of current global datasets for deriving more accurate
556 measures of the effect of rain on gas transfer have been highlighted. Further
557 work to exploit con-incident wind and rain data sets and associated
558 development of a generalised parameterisation relating wind and rain rate to the
559 concentration balance of trace gases across the interface offers significant
560 potential in this area.

561 **Acknowledgements**

562 The authors would like to thank C. J. Zappa and W. R. McGillis for their
563 constructive discussions about this work. This work was funded by the
564 European Space Agency (ESA) Support to Science Element (STSE) through the
565 OceanFlux Greenhouse Gases project (contract 4000104762/11/I-AM) and the
566 OceanFlux Greenhouse Gases Evolution project (contract 4000112091/14/I-LG),
567 as well as through the NERC RAGNARoCC project, (grant ref. NE/K002473/1)

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