



1 **Contrasting terrestrial carbon cycle responses to the two strongest El** 2 **Niño events: 1997-98 and 2015-16 El Niños**

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14

15 **Abstract**

16 The large interannual CO₂ variability is dominated by the response of terrestrial
17 biosphere to El Niño-Southern Oscillation (ENSO). However, behaviors of terrestrial
18 ecosystems differ in patterns and biological processes in different El Niño events. Here
19 we conduct a comprehensive comparison of the two strongest El Niño events in history,
20 namely, the recent 2015-16 event, and the earlier 1997-98 event in the context of multi-
21 event ‘composite’ El Niño. We analyze Mauna Loa CO₂ concentration, surface carbon
22 fluxes from three atmospheric inversions, and a mechanistic carbon cycle model
23 VEGAS. We find large differences in the carbon cycle responses, even though the two
24 El Niño events are of similar magnitude.

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26 We find that the land-atmosphere carbon flux (F_{TA}) anomaly in 1997-98 El Niño was



27 1.95 Pg C yr⁻¹ globally, but two times smaller during 2015-16 El Niño at 0.79 Pg C yr⁻
 28 ¹. We also find that F_{TA} had no obvious lagged response in 2015-16 El Niño, in contrast
 29 to that in 1997-98 El Niño. Separating the global flux by major geographical regions,
 30 during 1997-98, the fluxes in the tropics and extratropical northern hemisphere were
 31 1.98 and -0.04 Pg C yr⁻¹, respectively. During 2015-16, these were 1.07 and -0.4 Pg
 32 C yr⁻¹. Analysis of the mechanism shows that in the tropics, the widespread drier and
 33 warmer conditions caused the decrease in gross primary productivity (GPP, -1.11 Pg
 34 C yr⁻¹) and increase in terrestrial ecosystem respiration (TER, 0.49 Pg C yr⁻¹) in
 35 1997-98 El Niño. During 2015-16, in contrast, anomalously wet conditions occurred in
 36 Sahel and East Africa that caused increase in GPP, compensating its decrease over other
 37 tropical regions. As a result, the total 2015-16 tropical GPP and TER anomalies were
 38 0.29 and 1.23 Pg C yr⁻¹. GPP dominance during 1997-98 and TER dominance during
 39 2015-16 accounted for the phase difference in their F_{TA} . In extratropical northern
 40 hemisphere, we find that temperature was warmer both in 1997-98 and 2015-16 El
 41 Niños over North America, contributing to enhancements in GPP and TER. However,
 42 temperature over Eurasia was warmer in 2015-16 El Niño, opposing to the cooler in
 43 1997-98 and composite El Niño events. This warmer condition enhanced GPP and TER
 44 over the Eurasia in 2015-16 El Niño, compared to their suppressions in 1997-98 El
 45 Niño. The total extratropical northern hemisphere GPP and TER anomalies were 0.86
 46 and 0.74 Pg C yr⁻¹ in 1997-98 El Niño and 1.8 and 1.47 Pg C yr⁻¹ in 2015-16 El Niño.
 47 Additionally, we find that wildfires played less important roles in 2015-16 El Niño than
 48 in 1997-98 El Niño.

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52 **1 Introduction**

53 The atmospheric CO₂ growth rate has a significant interannual variability, greatly
54 influenced by the El Niño-Southern Oscillation (ENSO) (Bacastow, 1976; Keeling et
55 al., 1995). This interannual variability primarily stems from terrestrial ecosystems
56 (Bousquet et al., 2000; Zeng et al., 2005). Further, there is a general consensus that the
57 tropical terrestrial ecosystems account for the terrestrial carbon variability (Cox et al.,
58 2013; Peylin et al., 2013; Wang et al., 2016; Wang et al., 2013; Zeng et al., 2005). They
59 tend to anomalously release C flux during El Niño episodes, and uptake during La Nina
60 episodes (Wang et al., 2016; Zeng et al., 2005). Recently, Ahlstrom et al. (2015) further
61 suggested that ecosystems over the semi-arid regions dominated the terrestrial carbon
62 interannual variability with its 39% contribution.

63 The terrestrial dominance primarily results from the drive-response mechanisms in
64 climate variability (especially in temperature and precipitation) caused by ENSO and
65 plants/soil physiology (Jung et al., 2017; Tian et al., 1998; Wang et al., 2016; Zeng et
66 al., 2005). Directly, land-atmosphere C flux (F_{TA}) is mainly attributable to the
67 imbalance between the gross primary productivity (GPP) and terrestrial ecosystem
68 respiration (TER), according to $F_{TA} = TER - GPP + C_{fire}$ (C_{fire} is generally much
69 smaller than GPP or TER). So variations in each of them or both result in the variations
70 in F_{TA} .

71 Based on a dynamical global vegetation model (DGVM), Zeng et al. (2005) pointed
72 out that net primary productivity (NPP) contributed to almost three fourth of the tropical
73 F_{TA} interannual variability. Later, multi-model simulations involved in TRENDY
74 project consistently suggested NPP or GPP dominated the terrestrial carbon variability
75 (Ahlstrom et al., 2015; Piao et al., 2013; Wang et al., 2016).

76 These biological process analyses inferred that precipitation variation was the dominant



77 climate factor in controlling F_{TA} interannual variability (Ahlstrom et al., 2015; Qian et
78 al., 2008; Tian et al., 1998; Wang et al., 2016; Zeng et al., 2005). Quantitatively, Qian
79 et al. (2008) illustrated the contributions of the tropical precipitation and temperature
80 were 56 and 44% respectively, based on the model sensitivity experiments. Eddy
81 covariance network observations suggested that interannual C flux variability over the
82 tropical and temperate regions was controlled by precipitation, while boreal ecosystem
83 C fluxes were more subject to temperature and radiation (Jung et al., 2011). At the same
84 time, there was a significant positive correlation between atmospheric CO_2 growth rate
85 and mean tropical land temperature (Anderegg et al., 2015; Cox et al., 2013; Wang et
86 al., 2013; Wang et al., 2014). Sensitivity analysis indicated an about 3.5 Pg C yr^{-1}
87 anomaly in CO_2 growth rate with a 1°C increase in tropical land temperature, whereas
88 only a weaker interannual coupling existed between CO_2 growth rate and tropical land
89 precipitation (Wang et al., 2013). Therefore, these studies (Anderegg et al., 2015; Cox
90 et al., 2013; Wang et al., 2013; Wang et al., 2014) suggested the temperature dominance
91 in F_{TA} or CO_2 growth rate interannual variations, considering this strong emergent
92 linear relationship. Recently, in order to reconcile these contradictory reports, Jung et
93 al. (2017) illustrated that temporal and spatial compensatory effects in water availability
94 linked yearly global F_{TA} variability to temperature.

95 Apart from these long-term time series studies on the interannual F_{TA} or CO_2 growth
96 rate variability, we should keep in mind that the response of terrestrial carbon cycle to
97 every El Niño/La Niña event has its unique behaviors such as in the strength, spatial
98 pattern, biological process, and so on (Schwalm, 2011). For example, the wildfires
99 played an important role in F_{TA} anomalies during 1997-98 El Niño (van der Werf et al.,
100 2004). Recently, one of the three strongest El Niño events in recorded history occurred
101 in 2015-16 years (<https://www.esrl.noaa.gov/psd/enso/current.html>). Given the



disturbance of the El Chichón eruption in 1982-83 El Niño episode, we here attempt to comprehensively compare the responses of terrestrial ecosystems to the two strongest El Niños in 1997-98 and 2015-16 years in the context of multi-event ‘composite’ El Niño, based on DGVM VEGAS in its Near-Real Time framework, inversion datasets (CAMS, MACC, and CarbonTracker) and so on. Our purpose is to clarify the distinctions in responses of biological processes in these two extreme events. This paper is organized as follows: Section 2 describes the mechanistic carbon cycle model used, its drivers, and reference datasets. Section 3 presents the results about the total terrestrial C flux anomalies and spatial patterns along with their mechanisms. Finally, discussions and concluding remarks are illustrated in Sect. 4.

112

113 **2 Model and datasets**

114 **2.1 Mechanistic carbon cycle model and its drivers**

In this study, we used the state-of-the-art VEGAS version 2.4 in its Near-Real Time framework to investigate the responses of terrestrial ecosystems to El Niño events. VEGAS has been widely used to study the terrestrial carbon cycle on its seasonal cycle, interannual variability, and long-term trend (Zeng et al., 2005; Zeng et al., 2004; Zeng et al., 2014). And it extensively participated in the international carbon modelling project, such as the Coupled Climate-Carbon Cycle Model Intercomparison Project (C⁴MIP) (Friedlingstein et al., 2006), TRENDY project (Sitch et al., 2015) and Multi-scale Synthesis and Terrestrial Model Intercomparison Project (MsTMIP) (Huntzinger et al., 2013). The detailed descriptions on its model structure, biological processes, and so on can be referred to the appendix in Zeng et al. (2005). We ran VEGAS on the 0.5°×0.5° horizontal resolution from 1901 till the end of year 2016, and focus on the period from 1980 to 2016.



127 The climate fields used to force VEGAS are as follows: (1) Precipitation datasets are
128 generated by combining the Climatic Research Unit (CRU) Time-series (TS) Version
129 3.22 (University of East Anglia Climatic Research Unit et al., 2014) , NOAA's
130 PRECipitation REConstruction over Land (PREC/L) (Chen et al., 2002), and NOAA
131 NCEP climate anomaly monitoring system-outgoing longwave radiation precipitation
132 index (CAMS-OPI) (Janowiak and Xie, 1999). (2) Temperature is adopted from the
133 CRU TS3.22 before the year 2013, and generated by combining CRU 1981-2010
134 climatology and the Goddard Institute for Space Studies (GISS) Surface Temperature
135 Analysis (GISTEMP) (Hansen et al., 2010) after 2013. (3) Downward shortwave
136 radiation is retrieved from the driver datasets in MsTMIP (Wei et al., 2014) before the
137 year 2010, and repeated the value of the year 2010 after it. Additionally, the gridded
138 cropland and pasture land use datasets are integrated from the History Database of the
139 Global Environment (HYDE) (Klein Goldewijk et al., 2011) with an linear
140 extrapolation in 2016.

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142 **2.2 Reference datasets**

143 We here take a series of reference datasets as a comparison with the simulation of
144 VEGAS. The atmospheric CO₂ concentrations are from the monthly in-situ CO₂
145 datasets at Mauna Loa Observatory, Hawaii (Keeling et al., 1976). The Niño 3.4
146 (120°W–170°W, 5°S–5°N) sea surface temperature anomalies (SSTA) are adopted from
147 the NOAA Extended Reconstructed Sea Surface Temperature (ERSST), version 4
148 (Huang et al., 2015), with a 3-month running average. We take Copernicus Atmosphere
149 Monitoring Service (CAMS, 1980–2015), Monitoring atmospheric composition &
150 climate (MACC, 1980–2014) inversion results (Chevallier, 2013), and
151 CarbonTracker2016 (200001–201512) with the CarbonTracker Near-Real Time results



152 in 2016 (Peters et al., 2007) to compare with VEGAS. Fire emissions come from the
153 Global Fire Emissions Database, Version 4 (GFEDv4) from 1997 through 2014
154 (Randerson et al., 2015). Owing to the high correlation between solar-induced
155 chlorophyll fluorescence (SIF) and terrestrial GPP (Guanter et al., 2014), we take the
156 monthly satellite SIF from the GOME2_F version 26 from 2007 till 2016 (Joiner et al.,
157 2012). Another, we adopt the Enhanced Vegetation Index (EVI) from MODIS
158 MOD13C2 (Didan, 2015) to compare with the simulated leaf area index (LAI)
159 anomalies.

160 In order to get the anomalies during the El Niño events, we first remove the long-term
161 climatology in each dataset for getting rid of seasonal cycle signals, and then detrend
162 them based on the linear regression, because the trend is not caused by the interannual
163 variability.

164

165 **3 Results**

166 **3.1 Total terrestrial C flux anomalies**

167 Three strongest El Niño events (1982-83, 1997-98, and 2015-16) occurred from 1980
168 to 2016 with their maximum SST anomalies above 2.0 (Fig. 1a). El Niño event tends
169 to make the atmospheric CO₂ growth rate anomalously increase (Fig. 1b), so there are
170 two significant anomalously increased CO₂ growth rate corresponding to 1997-98 and
171 2015-16 El Niño events. Though the maximum increase in 2015-16 is a little smaller
172 than that in 1997-98. Owing to the diffuse light disturbance (Mercado et al., 2009) of
173 the eruption of Mount. El Chichón during 1982-83 El Niño event on the canonical
174 coupling between CO₂ growth rate anomalies and El Niño events, we mainly focus on
175 1997-98 and 2015-16 El Niño events in this study. The interannual variability of
176 atmospheric CO₂ growth rate principally originates from the terrestrial ecosystems (Fig.



177 1c). The correlation coefficient between CO₂ growth rate anomalies and global F_{TA}
178 simulated by VEGAS is 0.64 ($p < 0.05$). In order to evaluate the performance of
179 VEGAS simulation on the interannual time scale, we at the same time present CAMS,
180 MACC and CarbonTracker inversion results. We find that CAMS and MACC
181 inversions are nearly the same, both having the correlation coefficient about 0.60 ($p <$
182 0.05) with VEGAS. From 2000 through 2016, CarbonTracker is highly correlated with
183 VEGAS ($r = 0.71$, $p < 0.05$). These high correlation coefficients between VEGAS and
184 reference datasets underscore that VEGAS can well capture the terrestrial carbon cycle
185 interannual variability.

186 There are altogether 10 El Niño events from 1980 through 2016, each with different
187 duration and strength (Table 1). According to El Niño definition, we can find that these
188 10 El Niño events can be categorized into 2 weak (with a 0.5 to 0.9 SSTA), 3 moderate
189 (1.0 to 1.4), 2 strong (1.5 to 1.9), and 3 very strong (≥ 2.0) events. In 1997-98 El Niño,
190 the positive SSTA lasted from April 1997 to June 1998, while positive SSTA happened
191 in winter 2014, and extended to June 2016 in 2015-16 El Niño (Fig. 2a). However,
192 every El Niño event always peaks in winter (November or December) (Fig. 2a).

193 Considering this phase-lock phenomenon in El Niño events, we make a composite
194 analysis (getting rid of 1982-83 and 1991-92 because of the diffuse radiation
195 disturbances) as the background responses of terrestrial carbon cycle to El Niño events.

196 We can easily find that evolutions of F_{TA} anomalies in VEGAS, mean of CAMS and
197 MACC, and CarbonTracker in composite, 1997-98, and 2015-16 El Niño events are
198 closely consistent with Mauna Loa CGR anomalies (Fig. 2b, c, and d). Peaks of F_{TA}
199 and Mauna Loa CGR anomalies in 1997-98 and 2015-16 El Niños are much stronger
200 than those in composite analysis. Importantly, there were significant terrestrial lagged
201 responses in composite and 1997-98 El Niño events, with the peak of F_{TA} anomaly in



202 March to April in El Niño decaying year (Fig. 2b and c), consistent with previous
 203 studies (Qian et al., 2008; Wang et al., 2016). But this lagged terrestrial response
 204 disappeared in Mauna Loa CGR, VEGAS and CarbonTracker in 2015-16 El Niño (Fig.
 205 2d). Further, in June 2016, the F_{TA} anomaly of VEGAS and CarbonTracker significantly
 206 dropped (sign changed), but Mauna Loa CGR dropped a little (no sign changed) (Fig.
 207 2d). Similar phenomenon also occurred earlier from April to July 2015. In addition, we
 208 can know that the anomalous C release caused by El Niño lasts about from July in the
 209 El Niño developing year to October in the El Niño decaying year (Fig. 2b, c, and d).
 210 For simplicity, we calculate the total anomalies in next context during this period for
 211 all El Niño events, taking the terrestrial lagged responses into account (Wang et al.,
 212 2016).
 213 According to major geographical regions, we separate global F_{TA} anomaly into
 214 extratropical northern hemisphere (23°N – 90°N), tropical regions (23°S – 23°N), and
 215 extratropical southern hemisphere (60°S – 23°S). Because F_{TA} anomaly over the
 216 extratropical southern hemisphere is generally smaller, we mainly present the
 217 evolutions of F_{TA} over the extratropical northern hemisphere and tropical regions in Fig.
 218 3. Comparing the global and tropical F_{TA} anomalies, we find that F_{TA} anomalies in
 219 tropical regions dominate the global F_{TA} in these events (Fig. 3b, d and f), in accord
 220 with previous conclusions (Peylin et al., 2013; Zeng et al., 2005). Additionally, F_{TA}
 221 anomalies over the extratropical northern hemisphere are nearly neutral in VEGAS
 222 during composite and 1997-98 El Niño events (Fig. 3a and c). But we find that there
 223 were obvious anomalous uptakes from April to September in 2016 simulated by
 224 VEGAS (Fig. 3e), compensating the release over the tropics (Fig. 3f). These anomalous
 225 uptakes made the global negative F_{TA} anomalies from May to September in 2016 (Fig.
 226 2d). Similar anomalous uptake happened over the extratropical northern hemisphere



earlier from April to July 2015. These anomalous uptakes in VEGAS are to some extent consistent with results in CarbonTracker, and well account for the global F_{TA} drops mentioned above in these periods. Comparing the behaviors between Mauna Loa CGR and F_{TA} anomalies, we can now clearly find that Mauna Loa CGR, coming from the tropical observatory, does not reflect the signals over the extratropical northern hemisphere in time (Fig. 2d and Fig. 3e).

Because F_{TA} mainly stems from the difference between TER and GPP, we present TER and GPP anomalies in Fig. 4 in order to well explain the F_{TA} anomalies. We find that anomalous negative GPP dominated the F_{TA} anomaly in tropics during composite and 1997-98 El Niño episodes with the significant lagged responses (peak at about May in El Niño decaying year) (Fig. 4b and d). Besides, obvious positive TER anomalies occurred from October 1997 to April 1998 (Fig. 4d), contributing to tropical C release in this period (Fig. 3d). In contrast, we find that anomalous positive TER dominated the F_{TA} anomaly in tropics during 2015-16 El Niño episode without obvious lags (Fig. 4f), accounting for the disappearance of terrestrial F_{TA} lagged response (Fig. 2d). In the extratropical northern hemisphere, increased GPP and TER from April to October in composite and 1998 were nearly identical (Fig. 4a and c), making neutral F_{TA} anomalies (Fig. 3a and c). But increased GPP was stronger than increased TER from April to July 2015 and from April to September 2016 (Fig. 4e), resulting in the anomalous uptake in F_{TA} (Fig. 2d and Fig. 3e).

Quantitatively, we calculated the total C flux anomalies from July in El Niño developing year till October in El Niño decaying year. The composite global F_{TA} anomaly during El Niño events in VEGAS is about $0.71 \text{ Pg C yr}^{-1}$, dominated by tropical ecosystems with $0.74 \text{ Pg C yr}^{-1}$ (Table 2). These anomalies are comparable to the mean of CAMS and MACC inversion results with 0.92 ± 0.01 globally and



252 $0.66 \pm 0.03 \text{ Pg C yr}^{-1}$ in tropics. In these two extreme cases, a very strong anomalous C
 253 release occurred in 1997-98 El Niño episode with a value of $1.93 \text{ Pg C yr}^{-1}$, close to
 254 $2.57 \text{ Pg C yr}^{-1}$ in CAMS and MACC inversions, while only $0.79 \text{ Pg C yr}^{-1}$ was released
 255 in 2015-16 El Niño episode, comparable to $0.82 \text{ Pg C yr}^{-1}$ in CarbonTracker. But F_{TA}
 256 anomalies in tropical regions dominated the global F_{TA} anomalies in both cases with
 257 respective values of 1.98 and $1.07 \text{ Pg C yr}^{-1}$ in VEGAS. Moreover, anomalous C uptake
 258 simulated by VEGAS over the extratropical northern hemisphere cancelled 0.40 Pg C
 259 yr^{-1} anomalous release in tropics in 2015-16 El Niño, while it was neutral (-0.04 Pg C
 260 yr^{-1}) in 1997-98 El Niño. And the F_{TA} anomaly was relatively smaller in the
 261 extratropical southern hemisphere.

262 In biological processes, GPP ($-1.11 \text{ Pg C yr}^{-1}$) and TER ($0.49 \text{ Pg C yr}^{-1}$) in tropics
 263 together drove the anomalous F_{TA} in 1997-98, while TER ($1.23 \text{ Pg C yr}^{-1}$) partly
 264 cancelled by GPP ($0.29 \text{ Pg C yr}^{-1}$) drove the anomalous F_{TA} in 2015-16 (Table 2). These
 265 data confirmed that GPP played the more important role in 1997-98, while TER
 266 dominance occurred in 2015-16 El Niño episode. In the extratropical northern
 267 hemisphere, GPP and TER cancelled each other. They had respective 0.20 and 0.12 Pg
 268 C yr^{-1} in composite analysis and 0.86 and $0.74 \text{ Pg C yr}^{-1}$ in 1997-98 El Niño, making
 269 the nearly neutral F_{TA} anomaly there. But GPP ($1.80 \text{ Pg C yr}^{-1}$) was stronger than TER
 270 ($1.47 \text{ Pg C yr}^{-1}$) in 2015-16 El Niño, causing the significant C uptake. Additionally, F_{TA}
 271 anomaly caused by wildfires also played an important role in 1997-98 El Niño episode
 272 with globally $0.46 \text{ Pg C yr}^{-1}$ in VEGAS, consistent with GFED fire data product (0.82
 273 Pg C yr^{-1}). The effect of wildfires on F_{TA} anomaly in 1997-98 El Niño episode has been
 274 suggested by van der Werf et al. (2004). But it was close to zero ($0.08 \text{ Pg C yr}^{-1}$) in
 275 2015-16 El Niño episode.

276



277 **3.2 Spatial features and its mechanisms**

278 Regional responses of terrestrial ecosystems to El Niño events are inhomogeneous,
 279 principally according to the anomalies in climate variability. In composite El Niño
 280 analysis (Fig. 5a), land consistently releases C flux in tropics, while it anomalously
 281 uptakes C flux over North America as well as the central and eastern Europe. These
 282 regional responses are generally consistent with the CAMS and MACC inversion
 283 results (Fig. 5d).

284 In 1997-98 El Niño episode, tropical responses were analogous to composite results
 285 except the stronger releases. North America and central and eastern China had stronger
 286 C uptake, whereas Europe and Russia had stronger C release (Fig. 5b). However, in
 287 2015-16 El Niño episode, anomalous C uptake happened over the Sahel and east Africa,
 288 compensating the C release over the other tropical regions (Fig. 5c). It made the total
 289 F_{TA} anomaly in tropics in 2015-16 smaller than that in 1997-98 (Fig. 3d and f, and Table
 290 2). North America had anomalous C uptake, similar to that in composite and 1997-98
 291 El Niño, while central and eastern Russia also had anomalous C uptake in 2015-16 El
 292 Niño (Fig. 5c), opposing to C release in composite and 1997-98 El Niño. This opposing
 293 behavior of boreal forests over the central and eastern Russia clearly contributed to the
 294 total uptake over the extratropical northern hemisphere (Table 2). Moreover, we can
 295 clearly find that these regional responses in 2015-16 El Niño episode are significantly
 296 consistent with the CarbonTracker result (Fig. 5f).

297 In order to better make the explanations on these regional C flux anomalies, we present
 298 the main climate variabilities of soil wetness (mainly caused precipitation) and air
 299 temperature, as well as the biological processes of GPP and TER in Fig. 6. In the
 300 composite analyses, the soil wetness is generally reduced in tropics (Fig. 6a), making
 301 the widespread decrease in GPP (Fig. 6b), verified by model sensitivity experiments



(Qian et al., 2008). At the same time, air temperature is anomalously warmer, contributing to the enhancement in TER. But the drier conditions in the semi-arid regions such as Sahel, South Africa, and Australia, restrict the enhancement in TER induced by warmer temperature (Fig. 6d). Higher air temperature over the North America largely enhances GPP and TER, while cooler conditions over the Eurasia will reduce them (Fig. 6b–d). Wetter conditions over part of North America and Eurasia also to some extent benefit GPP and TER (Fig. 6a).

Comparing the composite results (Fig. 6a–d) and 1997-98 El Niño episode (Fig. 6e–h), we can easily find that the regional patterns are almost identical except the difference in magnitude. In contrast, there are some differences in 2015-16 El Niño episode. Over the Sahel and East Africa, the soil wetness increased induced by more precipitation (Fig. 6i), dynamically making the air temperature cooler (Fig. 6k). This wetter condition largely benefit GPP (Fig. 6j), compensating the decreased GPP over the other tropical regions. It caused in total the increased GPP in tropics, opposing to composite and 1997-98 El Niño episode (Table 2). More soil moisture also contributed to increase in TER over the Sahel (Fig. 6l), contrary to that in 1997-98 El Niño episode (Fig. 6h). This spatial compensation in GPP together with the widespread increased TER well accounted for the TER dominance in tropics during 2015-16 El Niño episode. Besides, increased GPP resulted in the anomalous C uptake here (Fig. 5c) which partly compensated the anomalous C release over the other tropical regions. It in some degree made the tropical smaller F_{TA} in 2015-16 El Niño episode than that in 1997-98 El Niño episode. Another obvious difference happened over the Eurasia with almost opposite signals in 1997-98 and 2015-16 El Niño episodes. Air temperature during 2015-16 El Niño episode over the Eurasia was anomalously higher, opposing to the cooler during composite and 1997-98 El Niño (Fig. 6c, g, and k). This warmer condition enhanced



327 GPP and TER (Fig. 6j and l), contrary to their suppressions in composite and 1997-98
 328 El Niño (Fig. 6b, d, f, and h). This phenomenon explained stronger GPP and TER
 329 anomalies and anomalous C uptake over the whole extratropical northern hemisphere
 330 (Table 2).

331 Recently, more attentions have been paid on SIF as an effective indicator for GPP
 332 (Guanter et al., 2014). Therefore, we here try to make a comparison between simulated
 333 GPP and SIF variabilities on the interannual timescale. Though there are noisy signals
 334 in SIF, we can find that SIF was anomalously positive over the USA, part of Europe,
 335 and East Africa, and negative over the Amazon and South Asia during the 2015-16 El
 336 Niño episode, corresponding to the increased and decreased GPP respectively (Fig. 7a
 337 and c). The correspondences over the other regions were not significant. In addition,
 338 MODIS EVI anomalously increased over the America, Southern South America, part
 339 of Europe, Sahel, and East Africa, but decreases over the Amazon, Northern Canada,
 340 central Africa, South Asia, and Northern Australia (Fig. 7d). These EVI anomalies were
 341 well corresponding to simulated LAI anomalies (Fig. 7b). These good correspondences
 342 between simulated GPP (LAI) and SIF (EVI) give us more confidence in VEGAS
 343 simulations.

344 At last, wildfires as important disturbances for F_{TA} always release C flux. Though F_{TA}
 345 anomalies caused by wildfires are generally smaller than GPP or TER anomalies, they
 346 played an important role in 1997-98 El Niño episode (Globally $0.46 \text{ Pg C yr}^{-1}$ in
 347 VEGAS and $0.82 \text{ Pg C yr}^{-1}$ in GFED) (Table 2), consistent with the previous study (van
 348 der Werf et al., 2004). Here we show the F_{TA} anomalies caused by wildfires in Fig. 8.
 349 The correlation coefficients between simulated global F_{TA} anomalies caused by
 350 wildfires and GFED fire data product are 0.40 (unsmoothed) and 0.61 (smoothed) (Fig.
 351 8a), confirming that VEGAS has certain capability in simulating this disturbance. In



1997-98 El Niño episode, satellite-based GFED data showed that F_{TA} anomalies caused by wildfires mainly happened over the tropical regions, such as Amazon, Central Africa, South Asia, and Indonesia (Fig. 8d). VEGAS also simulated the positive F_{TA} over these tropical regions (Fig. 8b). The total tropical F_{TA} anomalies caused by fires were $0.39 \text{ Pg C yr}^{-1}$ in VEGAS and $0.72 \text{ Pg C yr}^{-1}$ in GFED (Table 2). In 2015-16 El Niño episode, wildfires also resulted in positive F_{TA} anomalies over Amazon, South Asia, and Indonesia, but their magnitudes were smaller than those in 1997-98 El Niño episode, because it was much drier in 1997-98 El Niño episode than in 2015-16 El Niño episode (Fig. 6e and i). In addition, the wetter conditions over the East Africa in 2015-16 El Niño episode depressed the occurrences of wildfires with the negative F_{TA} anomalies (Fig. 8c). The tropical F_{TA} anomaly in total was $0.13 \text{ Pg C yr}^{-1}$ in VEGAS (Table 2). Therefore, we can find that wildfires played less important roles in 2015-16 than in 1997-98 El Niño episode. F_{TA} anomalies caused by wildfires over the extratropics are much weaker than those over the tropics, and their correspondences between VEGAS and GFED are poorer (Table 2 and Fig. 8b and d).

367

368 **4 Conclusions and Discussions**

Climate anomalies in magnitudes and patterns caused by different El Niño events are inconsistent, so responses of terrestrial ecosystems remain uncertain to different El Niño events (Schwalm, 2011). In this study, we comprehensively compare the impacts of the two strongest El Niño events in history, namely, the recent 2015-16, and earlier 1997-98 events in the context of multi-event ‘composite’ El Niño on the terrestrial carbon cycle, relying on VEGAS in its Near-Real Time framework, inversion datasets and so on. Main conclusions are drawn as follows:

(1) Simulations indicate that F_{TA} anomaly in 2015-16 El Niño episode was globally



377 0.79 Pg C yr⁻¹, nearly two times smaller than that in 1997-98 El Niño (1.95 Pg C
 378 yr⁻¹), confirmed by inversion results. We also find that F_{TA} had no obvious lagged
 379 response in 2015-16 El Niño, in contrast to that in 1997-98 El Niño. Separating the
 380 global flux, we find that fluxes in the tropics and extratropical northern hemisphere
 381 were 1.07 and -0.4 Pg C yr⁻¹ during 2015-16 El Niño episode respectively, while
 382 these were 1.98 and -0.04 Pg C yr⁻¹ during 1997-98 event. Tropical F_{TA} anomalies
 383 dominated global F_{TA} anomalies in both extreme El Niño events.

384 (2) Mechanistic analysis indicates that anomalously wetter conditions happened over
 385 the Sahel and East Africa during 2015-16 El Niño episode, resulting in the increase
 386 of GPP, which compensated the reduction of GPP over the other tropical regions. It
 387 caused in total the increased GPP in tropics (0.29 Pg C yr⁻¹), opposing to composite
 388 analysis (-0.80 Pg C yr⁻¹) and 1997-98 El Niño (-1.11 Pg C yr⁻¹). Spatial
 389 compensation in GPP and widespread increased TER (1.23 Pg C yr⁻¹) well
 390 explained the TER dominance in 2015-16 El Niño episode, opposing to GPP
 391 dominance in 1997-98 event. Different biological dominance accounted for the
 392 phase difference in F_{TA} responses in 1997-98 and 2015-16 El Niños.

393 (3) Higher air temperature over the North America largely enhanced GPP and TER
 394 both in 1997-98 and 2015-16 El Niño episodes. However, air temperature during
 395 2015-16 El Niño episode over the Eurasia was anomalously higher, opposing the
 396 cooler in 1997-98 El Niño episode. This warmer condition benefited GPP and TER,
 397 well accounting for stronger GPP (1.80 Pg C yr⁻¹) and TER (1.47 Pg C yr⁻¹)
 398 anomalies and anomalous C uptake (-0.40 Pg C yr⁻¹) over the extratropical
 399 northern hemisphere during 2015-16 El Niño.

400 (4) Wildfires, frequently happening in tropics, played an important role in F_{TA}
 401 anomalies during 1997-98 El Niño episode, confirmed by VEGAS simulation and



satellite-based GFED fire product. But VEGAS simulation indicates that the tropical F_{TA} caused by wildfires during 2015-16 El Niño episode was relatively smaller than that during 1997-98 El Niño episode. This result was mainly because the tropical weather was much drier in 1997-98 El Niño than that in 2015-16 El Niño.

Data Availability

In this study, all the datasets can be freely accessed. Mauna Loa monthly CO_2 records are available at <https://www.esrl.noaa.gov/gmd/ccgg/trends/data.html>. ERSST4 Niño3.4 index can be accessed from <http://www.cpc.ncep.noaa.gov/data/indices/ersst4.nino.mth.81-10.ascii>. CAMS and MACC inversions are available at <http://apps.ecmwf.int/datasets/>. CarbonTracker datasets can be found at <https://www.esrl.noaa.gov/gmd/ccgg/carbontracker/>. GFEDv4 global fire emissions are downloaded at https://daac.ornl.gov/cgi-bin/dsviewer.pl?ds_id=1293. Satellite SIF datasets are retrieved from http://avdc.gsfc.nasa.gov/pub/data/satellite/MetOp/GOME_F/MetOp-A/level3/. MODIS enhanced vegetation index (EVI) datasets are downloaded from https://lpdaac.usgs.gov/dataset_discovery/modis/modis_products_table/mod13c2_v006.

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Center for SIF datasets, and Land Processes Distributed Active Archive Center for
MODIS EVI datasets.

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577 **Tables and Figures:**

578 **Table 1** Lists of El Niño events from 1980 till 2016.

| No. | El Niño Events | Duration (mo) | Maximum Nino3.4 Index (°C) |
|-----|-----------------|---------------|----------------------------|
| 1 | Apr1982–Jun1983 | 15 | 2.1 |
| 2 | Sep1986–Feb1988 | 18 | 1.6 |
| 3 | Jun1991–Jul1992 | 14 | 1.6 |
| 4 | Oct1994–Mar1995 | 6 | 1.0 |
| 5 | May1997–May1998 | 13 | 2.3 |
| 6 | Jun2002–Feb2003 | 9 | 1.2 |
| 7 | Jul2004–Apr2005 | 10 | 0.7 |
| 8 | Sep2006–Jan2007 | 5 | 0.9 |
| 9 | Jul2009–Apr2010 | 10 | 1.3 |
| 10 | Nov2014–May2016 | 19 | 2.3 |

579

580 **Table 2** Carbon flux anomalies during El Niño events, calculated as the mean from July

581 in the El Niño developing year to October in El Niño decaying year. Flux units are in

582 Pg C yr⁻¹.

| Zones | El Niños | Inversions | | VEGAS Model | | | | GFED |
|----------|------------------------|---|------------------------------------|-----------------|-------|-------|-------------------|-------------------|
| | | F _{TA} (CAMS+MACC) ^a | F _{TA} (CarbonTracker) | F _{TA} | GPP | TER | C _{fire} | C _{fire} |
| | composite ^b | 0.92±0.01 | – | 0.71 | –0.76 | –0.20 | 0.15 | – |
| Global | 1997-98 | 2.57±0.04 | – | 1.93 | –0.11 | 1.36 | 0.46 | 0.82 |
| | 2015-16 | – | 0.82 | 0.79 | 1.79 | 2.50 | 0.08 | – |
| | composite | 0.20±0.02 | – | –0.09 | 0.20 | 0.12 | –0.01 | – |
| NH | 1997-98 | 0.40±0.07 | – | –0.04 | 0.86 | 0.74 | 0.07 | 0.11 |
| | 2015-16 | – | 0.18 | –0.40 | 1.80 | 1.47 | –0.06 | – |
| Tropical | composite | 0.66±0.03 | – | 0.74 | –0.80 | –0.22 | 0.16 | – |



| | | | | | | | | |
|----|-----------|-----------------|------|-------|-------|-------|------|-------|
| | 1997-98 | 2.12 ± 0.14 | – | 1.98 | –1.11 | 0.49 | 0.39 | 0.72 |
| | 2015-16 | – | 0.53 | 1.07 | 0.29 | 1.23 | 0.13 | – |
| | composite | 0.07 ± 0.01 | – | 0.06 | –0.16 | –0.10 | 0.00 | – |
| SH | 1997-98 | 0.05 ± 0.02 | – | –0.01 | 0.13 | 0.13 | 0.00 | –0.01 |
| | 2015-16 | – | 0.11 | 0.12 | –0.31 | –0.19 | 0.00 | – |

^aIt represents the mean value of CAMS and MACC inversion results with the uncertainty of their standard deviation.

^bComposite analyses exclude 1982-83, 1991-92, and 2015-16 El Niños, because the former two cases are disturbed by eruptions of El Chichón and Pinatubo, and the latter is not covered by inversion datasets.

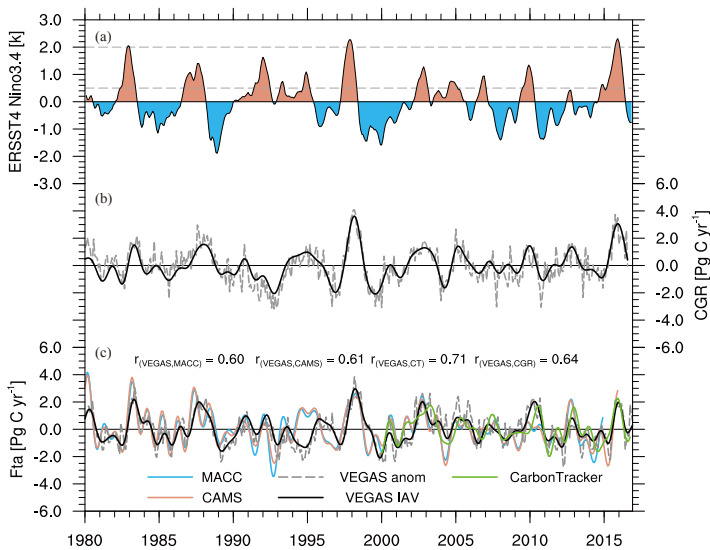
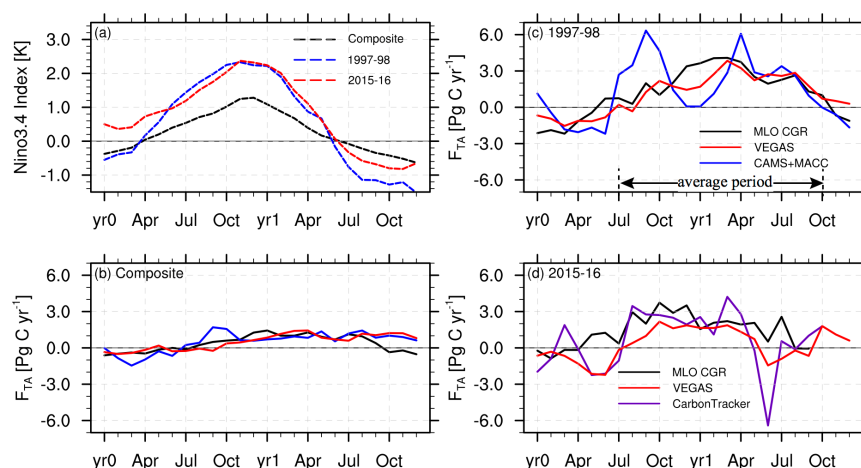


Figure 1. Interannual variabilities (IAV) in sea surface temperature anomaly (SSTA) and carbon cycle. (a) ERSST4 Niño3.4 Index (units: K). It is the 3-month running averaged SST anomaly for the Niño 3.4 region (5°N–5°S, 120°–170°W). (b) IAV in MLO CO₂ growth rate (CGR, units: Pg C yr⁻¹). CGR is calculated as the difference



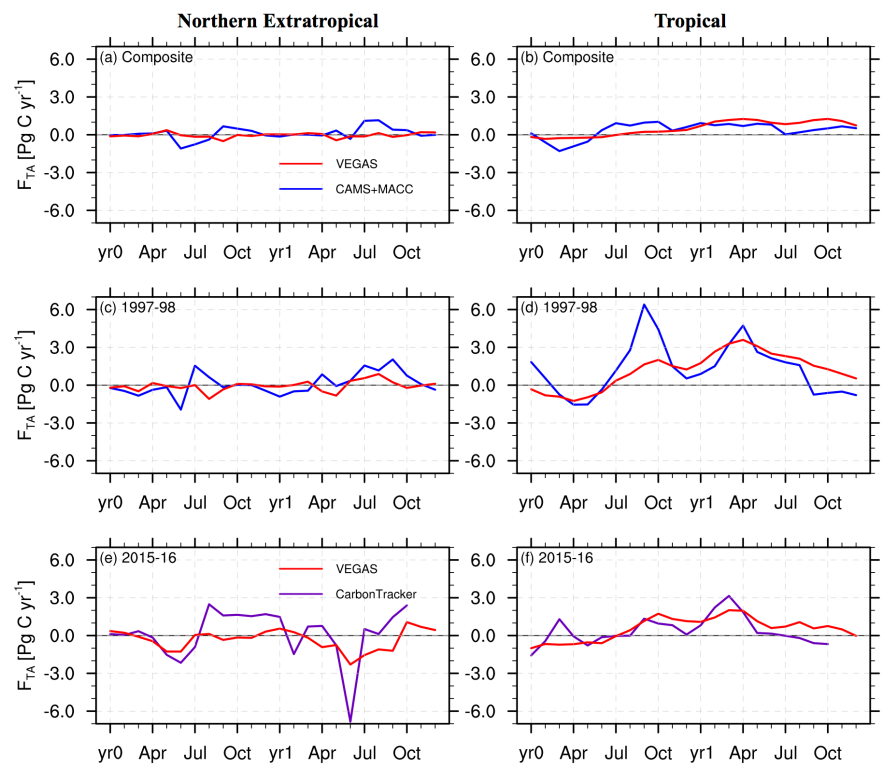
594 between the monthly mean in the adjacent years. The dashed line is the detrended
 595 anomaly and the solid line is smoothed by the butterworth filtering. (c) IAV in land-
 596 atmosphere carbon fluxes (F_{TA} , units: Pg C yr^{-1}). Blue and orange solid lines are the
 597 smoothed results of MACC and CAMS inversions. Gray dashed line is the detrended
 598 anomaly and black one is the smoothed result in VEGAS model simulation. The green
 599 solid line is the smoothed CarbonTracker result.
 600



601
 602 **Figure 2.** Evolutions of global F_{TA} along with the development of El Niño. (a) shows
 603 the SSTA in composite (in black), 1997-98 (in blue), and 2015-16 (in red) El Niño
 604 events. (b) illustrate the F_{TA} anomalies in El Niño composite analysis. The black solid
 605 line denotes Mauna Loa CGR; red and blue lines show the VEGAS and mean of CAMS
 606 and MACC inversions, respectively. (c) shows the F_{TA} anomalies during 1997-98 El
 607 Niño events. And the arrows demonstrate the time periods during which we calculate
 608 the C flux anomalies in table and below space figures. (d) demonstrate the F_{TA}
 609 anomalies during 2015-16 El Niño events. And the purple line denotes result of
 610 CarbonTracker2016 and CarbonTracker Near Real-Time datasets.



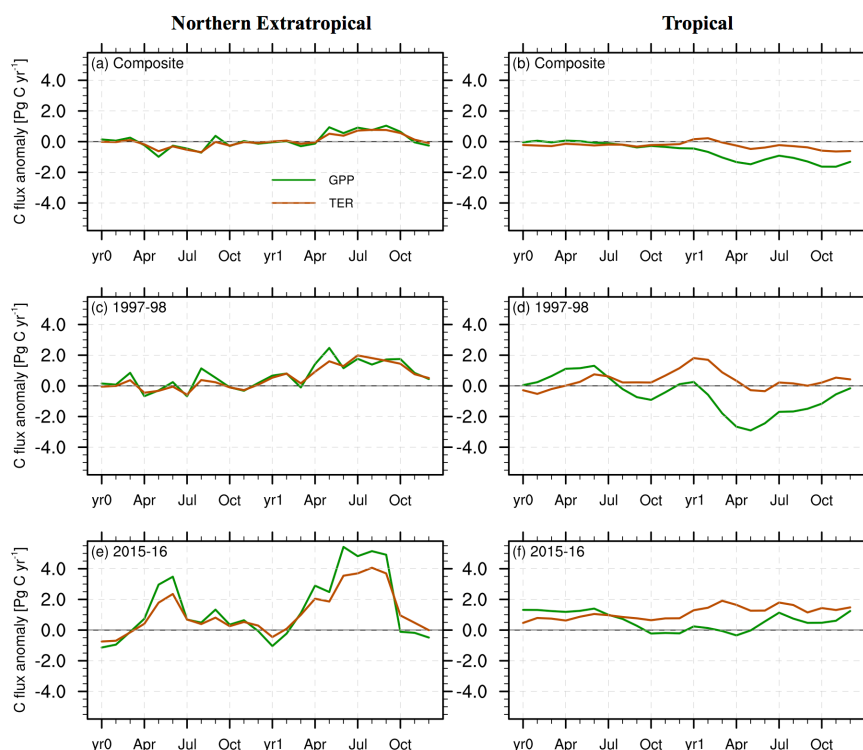
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613 **Figure 3.** Evolutions of F_{TA} over the extratropical northern hemisphere and tropical
614 regions along with the development of El Niño. (a–b) show the composite results with
615 VEGAS simulation (red solid line) and mean of CAMS and MACC inversions (blue
616 solid line). (c–d) show the F_{TA} anomalies in 1997-98 El Niño. (e–f) demonstrate the
617 FTA anomalies in 2015-16 El Niño with VEGAS (red solid line) and CarbonTracker
618 (purple solid line).

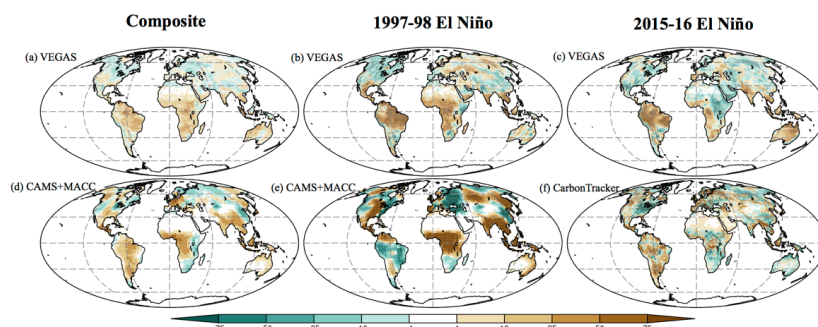
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621 **Figure 4.** Evolutions of gross primary productivity (GPP, green lines) and terrestrial
 622 ecosystem respiration (TER, brown lines) over the extratropical northern hemisphere
 623 and tropical regions along with the development of El Niño. (a–b) show the El Niño
 624 composite results. (c–d) show the results in 1997-98 El Niño event. And (e–f)
 625 demonstrate the results in 2015-16 El Niño event.

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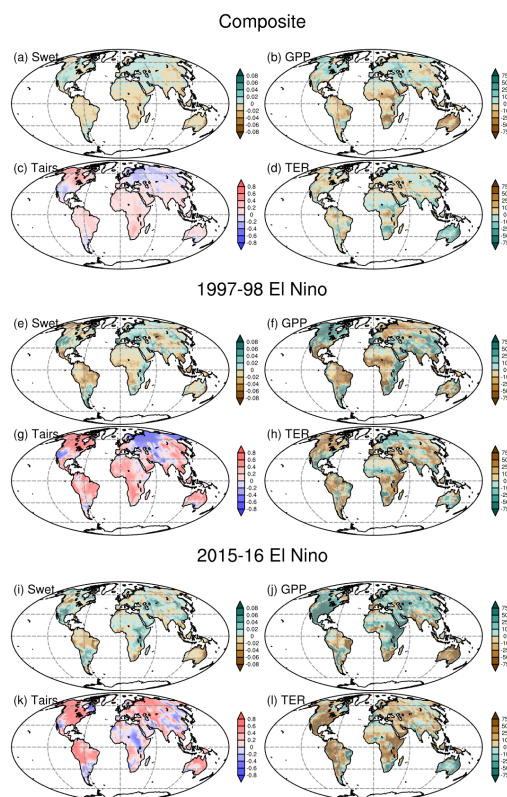


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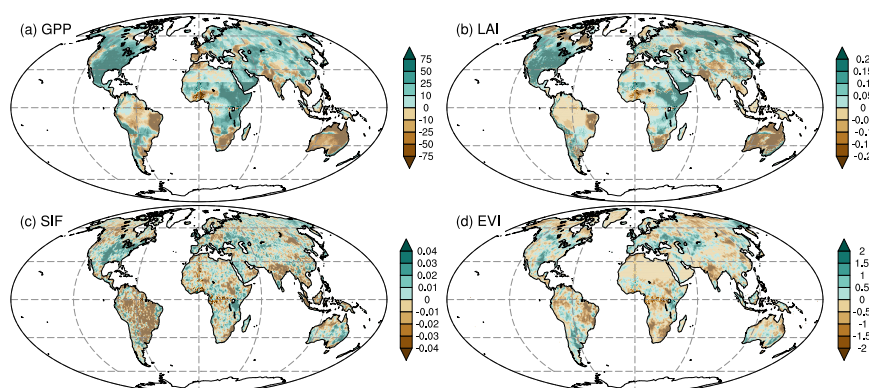
628 **Figure 5.** Spatial F_{TA} anomalies calculated from July in El Niño developing year to
 629 October in El Niño decaying year (units: Pg C yr^{-1}). (a), (b), and (c) show the results of
 630 composite, 1997-98, and 2015-16 El Niño events simulated by VEGAS, respectively.
 631 (d) and (e) represent the averaged results of CAMS and MACC in composite and 1997-
 632 98 El Niños. (f) shows 2015-16 El Niño F_{TA} anomaly in CarbonTracker.

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 636 **Figure 6.** Anomalies in soil wetness, air temperature (units: K), gross primary
 637 productivity (GPP, $\text{g C m}^{-2} \text{ yr}^{-1}$), and terrestrial ecosystem respiration (TER, g C m^{-2}
 638 yr^{-1}) from July in El Niño developing year to October in El Niño decaying year in
 639 composite, 1997-98, and 2015-16 El Niño episodes, respectively. (a–d) represent the
 640 results in composite analyses. (e–h) represent the anomalies during 1997-98 El Niño
 641 episode. And (i–l) show the anomalies during 2015-16 El Niño episode.
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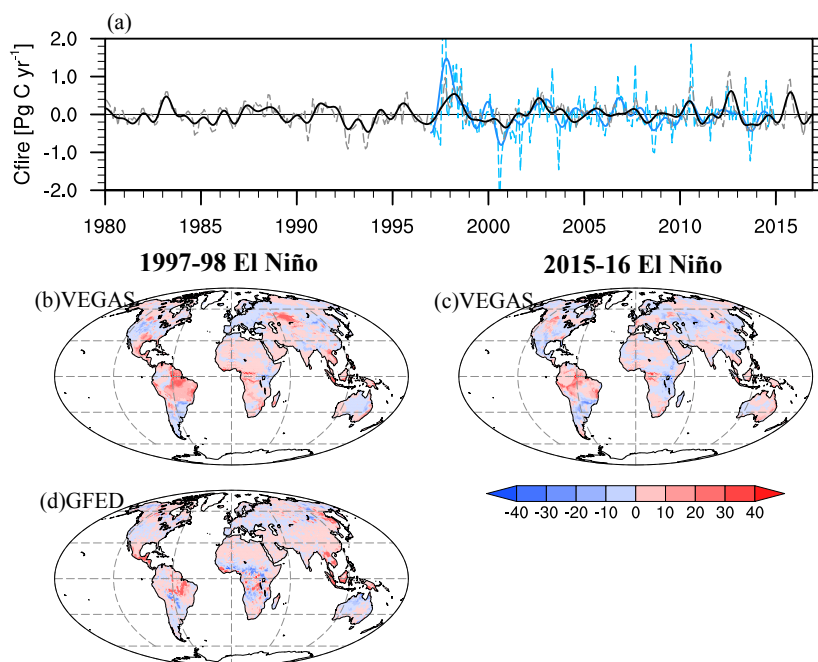


643
 644 **Figure 7.** Spatial anomalies in (a) simulated GPP by VEGAS (units: $\text{g C m}^{-2} \text{ yr}^{-1}$), (b)
 645 simulated leaf area index (LAI, units: $\text{m}^2 \text{ m}^{-2}$), (c) solar-induced chlorophyll
 646 fluorescence (SIF, units: $\text{mW m}^{-2} \text{ nm}^{-1} \text{ sr}^{-1}$), and (d) MODIS enhanced vegetation index
 647 (EVI, $\times 10^{-2}$) from July 2015 to October 2016.

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 652 **Figure 8.** F_{TA} anomalies induced by wildfires. (a) global total anomalies (Pg C yr^{-1}).
 653 The dashed gray and solid black lines represent the detrended and smoothed by
 654 butterworth filtering anomalies simulated by VEGAS. The dashed and solid blue lines
 655 represent the GFED results. (b) spatial F_{TA} anomaly ($\text{g C m}^{-2} \text{ yr}^{-1}$) in 1997-98 El Niño
 656 episode in VEGAS. (c) spatial F_{TA} anomaly in 2015-16 El Niño episode in VEGAS. (d)
 657 GFED anomaly in 1997-98 El Niño episode.
 658
 659