



# 1 Land-atmosphere interactions in the tropics

- 2
- 3 Pierre Gentine
- 4 Department of Earth and Environmental Engineering,
- 5 Earth Institute
- 6 Columbia University, New York, USA
- 7
- 8 Adam Massmann
- 9 Department of Earth and Environmental Engineering,
- 10 Earth Institute,
- 11 Columbia University, New York, USA
- 12
- 13
- 14 Benjamin R. Lintner
- 15 Department of Environmental Sciences
- 16 Rutgers, The State University of New Jersey
- 17 New Brunswick, NJ, USA
- 18
- 19 Sayed Hamed Alemohammad
- 20 Department of Earth and Environmental Engineering,
- 21 Earth Institute,
- 22 Columbia University, New York, USA

23

- 24 Rong Fu
- 25 Atmospheric and Ocean Sciences Department
- 26 University of California, Los Angeles
- 27





- 28 Julia K. Green
- 29 Department of Earth and Environmental Engineering,
- 30 Earth Institute,
- 31 Columbia University, New York, USA
- 32
- 33 Daniel Kennedy
- 34 Department of Earth and Environmental Engineering,
- 35 Earth Institute,
- 36 Columbia University, New York, USA
- 37
- 38 Jordi Vilà-Guerau de Arellano
- 39 Meteorology and Air Quality Group
- 40 Wageningen University, Wageningen, the Netherlands
- 41
- 42
- 43
- 44
- 45
- 46
- 47
- 48 Corresponding author address: Pierre Gentine, Department of Earth and Environmental
- 49 Engineering, Columbia University, NY 10027, USA.
- 50 E-mail: pg23288@columbia.edu
- 51 Phone number: +1-212-854-7287
- 52
- 53
- 54





#### 55 ABSTRACT

56

57 The continental tropics play a leading role in the terrestrial water and carbon cycles. Land-58 atmosphere interactions are integral in the regulation of surface energy, water and carbon fluxes 59 across multiple spatial and temporal scales over tropical continents. We review here some of the 60 important characteristics of tropical continental climates and how land-atmosphere interactions 61 regulate them. Along with a wide range of climates, the tropics manifest a diverse array of land-62 atmosphere interactions. Broadly speaking, in tropical rainforests, light and energy are typically more limiting than precipitation and water supply for photosynthesis and evapotranspiration; 63 whereas in savanna and semi-arid regions water is the critical regulator of surface fluxes and 64 65 land-atmosphere interactions. We discuss the impact of the land surface, how it affects shallow 66 clouds and how these clouds can feedback to the surface by modulating surface radiation. Some 67 results from recent research suggest that shallow clouds may be especially critical to land-68 atmosphere interactions as these regulate the energy budget and moisture transport to the lower 69 troposphere, which in turn affects deep convection. On the other hand, the impact of land surface 70 conditions on deep convection appear to occur over larger, non-local, scales and might be 71 critically affected by transitional regions between the climatologically dry and wet tropics.

#### 72 1 Introduction

73 Tropical ecosystems play a substantial role in regulating the global carbon and hydrologic 74 cycles. Tropical rainforests are one of the main terrestrial carbon sinks [Nakicenovic, 75 2000] but their projected response to a warming climate remains unclear because of 76 uncertainties associated with the representation of abiotic and biotic processes in models 77 as well as confounding factors such as deforestation and changes in land use and land 78 cover [Wang et al., 2009; Davidson et al., 2012; Fu et al., 2013; Saatchi et al., 2013; 79 Hilker et al., 2014; Boisier et al., 2015; Doughty et al., 2015; Gatti et al., 2015; Knox et 80 al., 2015; Saleska et al., 2016]. The ecosystems of tropical monsoonal and seasonal wet-81 dry climates are also important contributors to the global carbon cycle, especially with 82 respect to the interannual variability of the tropical terrestrial carbon sink [Poulter et al., 83 2014; Jung et al., 2017]. Some regions of the tropics have been further identified as hotspots of land-atmosphere 84

Some regions of the tropics have been further identified as hotspots of land-atmosphere interactions, modifying the regional climate [*Green et al.*, 2017] either locally, i.e. at horizontal scales on the order of a few boundary layer heights, regionally, at scales up to





87 a few hundreds of kilometers, or at large scales, over several of thousands of kilometers, 88 through coupling between the surface and the overlying atmosphere [Lintner and Neelin, 89 2009]. While tropical land-atmosphere interactions are often examined through the lens of coupling between land surface states (e.g., soil moisture) and rainfall, other aspects of 90 91 the coupling are also important. For example, even under nonprecipitating conditions, 92 surface radiation, temperature and vapor pressure deficit (VPD) may be altered [Lawton 93 et al., 2001; Pielke et al., 2016; Green et al., 2017] through coupling with clouds, 94 aerosols and shallow (non-precipitating) convection [Avissar and Nobre, 2002; Medvigy 95 et al., 2011; Seneviratne, 2013; Cook et al., 2014; Guillod et al., 2015; Krakauer et al., 2016; Martin et al., 2016; Green et al., 2017; Khanna et al., 2017; Martin et al., 2017; 96 97 Thiery et al., 2017; Vogel et al., 2017]. It is clear that the tropical energy, water, and carbon cycles cannot be understood in isolation; rather, interactions among these cycles 98 99 are critical, especially in determining whether the terrestrial tropics will act as a future 100 carbon sink or source [Zhang et al., 2015][Swann et al., 2015]. 101 The two-way interactions that occur between the land surface and overlying atmosphere 102 represent one of the more uncertain aspects of the terrestrial climate system, particularly

in the tropics [*Betts and Silva Dias*, 2010]. While the land surface is widely recognized as integral to the occurrence of important tropical climate phenomena such as monsoons [*Zeng and Neelin*, 1999; *Zeng et al.*, 1999], isolating and quantifying its precise role remains elusive. Indeed, such efforts have frequently been hampered by the paucity of observational data, not to mention the complex and multiple pathways through which land-atmosphere interactions can take place.

109 Several field campaigns have been conducted in the tropics with the purpose of advancing knowledge of land-atmosphere interactions. One of the first campaigns was 110 the Large-Scale Biosphere-Atmosphere Experiment in Amazonia (LBA) [Avissar et al., 111 2002; Keller et al., 2004], which aimed at refining our understanding of climatological, 112 113 ecological, biogeochemical and hydrological processes of the Amazon and their linkages, 114 in addition to the anthropogenic impacts (e.g., land-use land cover changes and 115 deforestation, in particular) on these. Among many other topics, LBA generated fundamental insights on the structure of the tropical atmosphere, processes generating 116 117 precipitation, and the seasonal variability of rainforest surface turbulent fluxes [Avissar 118 and Nobre, 2002; Betts et al., 2002; Laurent et al., 2002; Machado and Laurent, 2002;





119 Acevedo et al., 2004; Khairoutdinov and Randall, 2006; Fitzjarrald, 2007; Juárez et al., 120 2007; Restrepo-Coupe et al., 2013]. Much of the initial LBA research attempted to isolate 121 the effect of deforestation on precipitation, both in a local context as well as remotely via teleconnections [Avissar et al., 2002]. Much of this research pointed to deforestation 122 123 decreasing precipitation, albeit with uncertain magnitude. Even now, two decades after 124 the inception of LBA, the relationship between tropical deforestation and precipitation 125 remains uncertain, despite progress with respect to key processes such as the forest's role 126 in accessing deep water in the dry season, and cloud-cover's role in modulating energy 127 availability for photosynthesis [Betts and Dias, 2010]. Another noteworthy field campaign, the African Monsoon Multidisciplinary Analysis 128 129 (AMMA) campaign, focused on the West African monsoon system, especially the Sahel transition zone [Redelsperger et al., 2006; Boone et al., 2009b]. AMMA built upon 130 131 previous field work in the region [e.g. HAPEX-Sahel, Gourtorbe et al. 1993], and substantially advanced understanding of mesoscale convective systems and their 132 133 initiation the role of surface professes, and the vegetation water stress response in semi-134 arid regions [Lebel et al., 2009; Taylor et al., 2009; Boone et al., 2009a; Lohou et al., 135 2010; Couvreux et al., 2011a; 2011b]. More recently, the 2014-2015 Green Ocean Amazon (GO-Amazon) campaign [Martin et al., 2016] sought to quantify the impact of 136 137 atmospheric composition and aerosols under clean and polluted conditions on cloud formation and radiation over the basin, as well as on shallow to deep convection 138 139 development [Anber et al., 2015a; Tang et al., 2016; Giangrande et al., 2017]. 140 The remainder of this review article is organized as follows. We first review the typical 141 definitions of the tropics and of land-atmosphere interactions in section 2. In section 3 we 142 discuss the seasonality and characteristics of the climate of the tropics. The different types of feedbacks from local to non-local (i.e. remote influences) are then highlighted in 143 144 section 4, and we close in arguing that shallow cloud feedback and its impact on radiation 145 has received too little attention compared to precipitation feedback, in rainforests 146 especially.





## 147 2 Definitions of land-atmosphere interactions in the tropics

148 2.1 What (where) are the Tropics?

149 There exist multiple definitions of the Tropics. On the one hand, the Tropics can be 150 defined spatially as the area between the Tropics of Cancer and Capricorn, located at 151  $\sim 23\frac{1}{2}^{\circ}$  N and  $\sim 23\frac{1}{2}^{\circ}$  S, respectively. On the other hand, it is sometimes useful to define the Tropics in terms of underlying climate or physical characteristics. One such 152 153 physically-motivated definition of the Tropics is the region over which mean top-of-the-154 atmosphere solar incoming radiation exceeds outgoing radiation (reflected shortwave and 155 outgoing longwave), which occurs equatorward of  $\sim 35^{\circ}$ . Another definition is the region 156 near the equator where the Coriolis effect is small and planetary scale equatorial wave dynamics are dominant, which strongly affects the dynamics, as we elaborate below 157 [Sobel et al., 2001; Sobel and Bretherton, 2003; Lintner and Chiang, 2005; Raymond and 158 159 Zeng, 2005].

160 Over land, the Tropics are often defined biogeographically, as in the traditional Köppen 161 climate classification scheme [*Köppen*, 1884]: tropical regions are divided into three 162 main groups—tropical rainforest, tropical monsoon, and tropical wet and dry (or 163 savanna)—all of which are characterized by annual mean temperatures exceeding 18°C 164 but which differ in terms of precipitation amount and seasonality.

The latitudes between the Tropics of Cancer and Capricorn encompass some regions of large-scale subsidence and limited rainfall, including drylands and deserts, which we largely neglect here, even though land-atmosphere coupling processes within these regions is clearly of interest. Thus, throughout this manuscript, we define the Tropics as the latitudinal band between -15° S and 15° N, as it captures most of the wet regions of the Tropics while excluding many of the more arid regions at higher latitudes.

171 2.2 How to define land-atmosphere interactions?

172 There are typically two main definitions of land-atmosphere interactions:

173 2.2.1 Surface turbulent fluxes

While many potential definitions of land-atmosphere interactions exist, we propose a definition of land-atmosphere interactions as the study of turbulent fluxes and associated momentum, energy, water and trace gases exchanges between the biosphere and the





177 atmosphere [Goulden et al., 2004; Fisher et al., 2009; Restrepo-Coupe et al., 2013]. 178 Surface turbulent flux measurements in the tropics are usually obtained from eddy-179 covariance methods, typically above the canopy [Baldocchi et al., 2001]. Observing 180 turbulent fluxes is challenging in tropical environments for many reasons including 181 logistics, maintenance and the harsh environment such as intense rainfall, high wind, and 182 relative humidity, which impacts the sensors [Campos et al., 2009; Da Rocha et al., 183 2009; Restrepo-Coupe et al., 2013; Zahn et al., 2016; Chor et al., 2017; Gerken et al., 184 2017]. In light of these challenges, it is perhaps not surprising that even the best estimates 185 of surface turbulent fluxes manifest large uncertainties [Mueller et al., 2011]. Apart from site level measurements, remote sensing observations can provide 186 information about surface turbulent fluxes and other relevant quantities over tropical land 187 regions. There is considerable uncertainty in upscaling point observations to larger areas. 188 189 Remote sensing observations are useful to generalize and compare fluxes across the tropics even if they are not as direct as point observations, which are limited to  $\sim 10$  local 190 191 stations across the wet tropics. We emphasize that there are considerable uncertainties in 192 remote sensing and reanalysis estimates of rainfall [Washington et al., 2013; Levy et al., 193 2017], radiation [Jimenez et al., 2011], and surface turbulent fluxes [Alemohammad et al.,

194 2016].

195 While direct, satellite-based retrievals of turbulent fluxes of carbon (i.e. gross primary 196 production (GPP))) and water would be most suitable for the study of tropical land-197 atmosphere interactions, such retrievals are beyond current remote sensing capability. 198 However, some recent work demonstrates that existing satellite observations may still be 199 leveraged to study surface turbulent fluxes in the tropics. Alemohammad et al. [2016] 200 applied a machine learning algorithm based on remotely-sensed Solar-Induced 201 Fluorescence (SIF), called WECANN (Water Energy and Carbon Artificial Neural 202 Network) to derive surface turbulent fluxes. WECANN reproduces the seasonality in the 203 wet tropics and exhibits plausible interannual. In contrast to the normalized difference 204 vegetation index (NDVI) or many other vegetation indices which are indirect byproducts 205 of photosynthesis, SIF (at the leaf scale) is directly related to the ecosystem-scale 206 photosynthesis rate, providing important information on the impact of stressors on 207 photosynthesis and is available from existing remote sensing platforms [Frankenberg et 208 al., 2011; Joiner et al., 2011; Frankenberg et al., 2012; Joiner et al., 2013; Frankenberg





209 et al., 2014; Guanter et al., 2014; Lee et al., 2015; Duveiller and Cescatti, 2016; Liu et al., 2017; Thum et al., 2017; Alexander et al., n.d.]. SIF is thus an important indicator of 210 211 the rates of photosynthesis and transpiration through stomatal (small pores at the leaf surface) opening [Alemohammad et al., 2017]. Indeed, during photosynthesis plants take 212 213 up CO<sub>2</sub> from the atmosphere while releasing water to the atmosphere through stomata. 214 WECANN performs well compared to eddy-covariance observations and has less 215 uncertainty compared to many other retrievals (see [Alemohammad et al., 2017]). We 216 note that recent developments in observations of SIF seem to indicate that the major 217 fraction of the SIF signal might be related to chlorophyll photosynthetically active radiation and that changes in SIF yield (equivalent to light use efficiency) may account 218 219 for only a small fraction of the observed SIF signal [Du et al., 2017]. This is still an open topic to better understand what is actually observed by SIF remote sensing. 220

221 2.2.2 Weather and climate feedback

A second definition of land-atmosphere interactions relates to the feedback between surface processes (radiation, surface turbulent fluxes) and the overlying atmosphere, which may occur across multiple temporal and spatial scales. Throughout this manuscript, we highlight contribution of three types of feedbacks:

- 1) feedbacks between the surface and low-level clouds, including surface fog and shallowconvection;
- 228 2) feedbacks between the surface and deep convection, i.e. deep raining clouds extending
- above the freezing level;
- 230 3) feedbacks between the surface and large-scale circulation.

231 The distinction between shallow and deep convection remains elusive, as these have been

- 232 regarded as both fundamentally distinct or as a continuum, in both observations and
- 233 model convection parameterizations [Khairoutdinov and Randall, 2006; Bretherton and
- 234 Park, 2009; Park and Bretherton, 2009; Rio et al., 2009; Wu et al., 2009; Del Genio and
- 235 Wu, 2010; Hohenegger and Bretherton, 2011; Böing et al., 2012; D'Andrea et al., 2014;
- 236 Rochetin et al., 2014b]. We will loosely refer to shallow convection as convection
- confined below the freezing level (typically less than 3km deep) and comprising non-
- precipitating clouds with motions of small scale (typically less than a km in thehorizontal).





240 An important point is that shallow convection is frequently generated by thermals rooted 241 in the boundary layer and is thus ultimately related to surface sensible (H) and latent heat 242 (LE) flux and their partitioning [Gentine et al., 2013a; 2013b; de Arellano et al., 2014]. 243 The impact of surface heat fluxes and their partitioning on shallow convection is 244 demonstrated in the Amazon in Figure 1. Shallow convection frequently occurs over the 245 vegetated surface away from the ocean; also, over cooler and more humid river basins, shallow clouds are virtually absent [Gentine et al., 2013a; Rieck et al., 2014; 2015]. In 246 247 addition, shallow convection is strongly influenced by the diurnal cycle of surface 248 radiation and surface turbulent heat fluxes [Gentine et al., 2013a; 2013b; de Arellano et 249 al., 2014].

On the other hand, we use the term deep convection in association with deep, 250 251 precipitating clouds. Deep convection may be triggered by boundary layer thermals 252 [D'Andrea et al., 2014; Guillod et al., 2014; Rochetin et al., 2014a; 2014b; Anber et al., 2015a] as well as other processes such as radiative destabilization [Anber et al., 2015b], 253 254 meso- and large-scale circulations [Werth and Avissar, 2002; Roy et al., 2003], cold pools 255 (cold density currents due to rain evaporation that cools the air within precipitating downdrafts) [Engerer et al., 2008; Del Genio and Wu, 2010; Böing et al., 2012; Feng et 256 257 al., 2015; Torri et al., 2015; Gentine et al., 2016; Heever, 2016; Drager and van den Heever, 2017] and wave activity [Kuang, 2008; 2010]. As such, deep convection may be 258 259 viewed as less dependent on the surface state compared to shallow convection.

Over the central Amazon a large fraction of wet season precipitation occurs during the 260 261 nighttime (Figure 2). Moreover, during the daytime in both the dry and the wet seasons, 262 the diurnal cycle reflects not only locally surface-triggered deep convection [Khairoutdinov and Randall, 2006; Ghate and Kollias, 2016] but also mesoscale 263 264 convective systems propagating on daily time scales throughout the Amazon basin 265 [Ghate and Kollias, 2016]. However, during the dry season, precipitation occurs more 266 frequently with the "popcorn type" deep convection that is more locally triggered and 267 thus directly related to the state of the land surface [Ghate and Kollias, 2016] (see an 268 example here https://youtu.be/c2-iguZziPU).

- 269 Current generation climate models struggle to represent both shallow and deep
- convection over continents [Guichard et al., 2004; Bechtold et al., 2013; Yin et al., 2013;
- 271 D'Andrea et al., 2014; Couvreux et al., 2015], and especially in the tropics, as they





272 exhibit substantial errors in the phasing and intensity of both the diurnal and seasonal cycles of convection [Bechtold et al., 2013], as well as biases in the climatological 273 274 distribution of rainfall over land. For example, over the Amazon, many climate models 275 underestimate surface precipitation, evapotranspiration, and specific humidity [Yin et al., 276 2013], with the dry bias in moisture extending upwards into the lower free troposphere {Lintner:2017gm}. Such biases are largely thought to reflect deficiencies or errors in how 277 278 convection is represented in models [Yano and Plant, 2012; Stevens and Bony, 2013; 279 Bechtold et al., 2014]. Indeed, in current generation climate models, cloud processes 280 occur at scales smaller than resolved grid-scale prognostic variables and therefore need to be parameterized, i.e. represented as a function of the resolved-scale variables. This is 281 282 important as it means that climate models do not explicitly represent the small-scale 283 convective physics of the climate system. We do note, however, that cloud resolving 284 models which include explicit convection at scales of ~1km alleviate many of the biases 285 observed in climate models, especially in terms of the diurnal cycle of convection or the 286 sign and magnitude of the feedbacks between deep convection and surface evaporative 287 fraction [Taylor et al., 2013; Anber et al., 2015a]. Nonetheless, due to convective wave coupling in the Tropics, a simple prescription of lateral boundary conditions in small-288 domain cloud-resolving model may be problematic, as the convective scales ultimately 289 290 interact and are coupled with the planetary scales. With a sufficiently large domain and 291 fine enough resolution, coupling between the convective scales and planetary scales may 292 be explicitly resolved, but simulations of this nature are likely too be computationally too 293 expensive for many applications. However, techniques exist to represent the effect of 294 large-scale dynamics on the convective scales, which, when combined with cloud 295 resolving simulations, yield powerful tools for understanding land-atmosphere interactions in the tropics, as we elaborate further below. 296

### 297 3 Characteristics of the tropics

#### 298 3.1 Weak temperature gradient approximation – nonlocality

299 One key concept in tropical climate is the Weak Temperature Gradient (WTG) 300 approximation. In the tropical free troposphere, horizontal gradients of temperature (and 301 pressure) are small in part because of the relative weakness of the Coriolis parameter (as





302 on large-scales, geostrophic balance holds poleward of ~5 degrees). Homogenization 303 occurs over a spatial scale comparable to the Rossby radius of deformation, which is 304 inversely proportional to the Coriolis parameter. In midlatitudes, the Rossby radius is of order  $10^2$  km (similar to climate model resolution). In the tropics, the Rossby radius is 305 306 typically an order of magnitude larger. Consequently, localized convection, and the 307 diabatic heating associated with condensation and freezing of water, cannot be viewed in 308 isolation from the large-scale in the tropics: in other words, in the tropical free 309 troposphere, the temperature and pressure fields rapidly adjust to localized perturbations, 310 effectively spreading the effect of these perturbations. In addition, it is relatively straightforward to show that adiabatic cooling, associated with large-scale vertical ascent 311 in the presence of a vertical gradient of dry static energy  $h = c_p T + gz$ , effectively 312 balances the diabatic heating rate Q, which in rainy regions of the tropics is mostly 313 314 associated with convective processes. This further emphasizes the coupling between 315 diabatic heating and large-scale ascent. Since the introduction of WTG, related and 316 refined frameworks, such as weak pressure gradient [Romps, 2012a; 2012b] or damped 317 gravity waves [Wang et al., 2013], have been proposed. It should be emphasized that the WTG framework is only valid in the free troposphere, above the boundary layer, as it 318 319 relates to wave dynamics in a stratified atmosphere.

The WTG framework has been used in single-column model and cloud-resolving models of the tropics [*Sobel et al.*, 2007; *Daleu et al.*, 2012; 2014; *Sentić and Sessions*, 2017] to obtain boundary conditions consistent with convective activity in the domain, thus avoiding the issues of inconsistent boundary forcing alluded to in section 2.2.2. While the WTG framework has often been applied in an oceanic context, [*Anber et al.*, 2015a] have demonstrated its utility in studying the coupling between regional land surface processes and larger-scale circulation, as discussed in Section XX.

#### 327 3.2 Surface turbulent fluxes climatology and seasonality

328 Given that few flux towers are available across the tropics, we use WECANN 329 [*Alemohammad et al.*, 2017] to calculate surface flux climatologies across the continental 330 tropics. WECANN has been validated against available flux tower data and outperforms 331 other products in terms of reproducing both the seasonality and interannual variability 332 [*Alemohammad et al.*, 2017]. While remote sensing retrievals are not perfect and cannot 333 be considered the truth, they do provide spatially extensive data coverage, including





regions with sparse (or no) site-level measurements (e.g., Congo), which are hard to upscale to larger scale. In what follows, we evaluate climatologies of evapotranspiration (ET) and gross primary production (GPP) against precipitation (based on GPCP 1DD v1.2 [*Huffman et al.*, 2001]) and net radiation (based on CERES SYN [*Kato et al.*, 2013]) (Figure 4 to Figure 8).

339 We first focus on the main tropical rainforests and the northeastern savanna (or Cerrado) 340 region of Brazil (Figure 4). In the wetter part of the Amazon, net radiation, Rn, peaks in 341 the dry season (August to November) (Figure 4) when precipitation and cloud cover-342 especially shallow cloud cover, including fog-are reduced, [Anber et al., 2015a]. As a result of the reduced cloud cover, incident surface solar radiation increases, and both GPP 343 344 (Figure 6) and ET (Figure 7) increase in the dry season (Figure 4). As discussed further 345 in the next section, the forest in the climatologically wetter Amazon is primarily light 346 limited, while water stress there is moderate in the dry season. The seasonal cycle is more 347 pronounced for GPP than for ET (Figure 4), as canopy rain interception comprises a large 348 fraction of total ET in the wet season [Scott et al., 1997; Oleson et al., 2008; Miralles et 349 al., 2010; Sutanto et al., 2012; van Dijk et al., 2015; Andreasen et al., 2016] and partly 350 compensates for reduced transpiration in the wet season. In fact, because of this 351 compensation, the wettest parts of the Amazon exhibit weak ET seasonality. On the other 352 hand, most land-surface models exaggerate water stress in the Amazon [Powell et al., 353 2013] and typically exhibit much lower rates of ET and GPP in the dry season, as well as 354 opposite seasonality of net ecosystem exchange, than are observed [de Gonçalves et al., 355 2013; Alemohammad et al., 2016; 2017].

In contrast to the everwet central Amazon, over the Cerrado region of Northeastern Brazil, the seasonal cycles of Rn, precipitation, GPP and ET are much more pronounced, with a marked dry season (Figure 4). The seasonal cycle of GPP tracks precipitation, exhibiting a strong increase during the wet season. Similarly, ET increases sharply in the wet season and then decreases more slowly than precipitation in the dry region (Figure 4). Conversely, net radiation increases sharply during the dry season. This region clearly exhibits a strong water stress response.

- 363 Over the Maritime Continent, rainfall is intense throughout the year and seasonality is 364 modest, with a short peak in November to January (Figure 4). Much of the seasonal cycle
- 365 is attributable to monsoon circulations, which are strongly influenced by topography and





366 the land- and ocean-surface thermal contrast [Chang 2005]; however, the complexity of the topography and the distribution of island land masses leads to strong local variability. 367 368 Additionally, the Madden Julian Oscillation, an important mode of climate variability in 369 in the tropical Indo-Pacific with a lifecycle of 30-90 days, strongly impacts rainfall on 370 intraseasonal timescales [Hidayat and Kizu, 2009]. Convective activity in the region also 371 regulates the East Asian Monsoon [Huang and Sun, 1992]. The region is also influenced 372 by topographic effects and land-see breeze interactions at shorter time scales, and 373 exhibits a strong diurnal cycle in convection [Nitta, 1987; Hamada et al., 2008]. Given 374 the relatively steady annual cycle of precipitation with regular convection, ET and GPP remain relatively steady throughout the entire year, exhibiting minimal seasonality, in 375 376 this light limited environment (Figure 4). The Congo basin exhibits two rainy seasons (Figure 4), with peaks in March-April-May 377 378 and September-October-November, related to seasonal changes in moisture convergence 379 due to the African Easterly jet and Intertropical Convergence Zone (ITCZ) over the

Atlantic [*Washington et al.*, 2013]. Throughout the year, monthly-mean precipitation is much less than that observed over the Amazon or Indonesia. The seasonality of GPP and ET, to a lesser extent, tracks that of precipitation, with substantial decreases during the June to August dry season and even more pronounced reduction during the December to February period. This seasonality in GPP and ET (Figure 4) suggests that the Congo basin should exhibit substantially more water stress during dry seasons compared to the Amazon or Indonesian rainforests (Guan et al. 2015).

387 Integrated over the entire tropical latitudinal band, precipitation is highest in DJF and 388 MAM when the wet season extends over most of the Amazon and adjacent savanna 389 regions (Figure 5). GPP is maximized during the wet season in South America, as GPP is 390 highest in the savanna regions while GPP over the rainforest is effectively seasonally 391 invariant (Figure 7). The seasonal pattern of ET resembles GPP (Figure 7), mostly reflecting the seasonality of water availability in drier, water-limited regions and 392 393 increased radiation in the dry season in the wetter, more energy-limited portions of the 394 Amazon. The seasonal cycle of sensible heat flux (Figure 8) largely follows water stress, 395 especially in the rainforest where radiation remains high throughout the year, with an 396 increase during the dry season. Water stress is further evidenced in the evaporative 397 fraction, EF, the ratio of latent heat flux to latent and sensible heat fluxes (Figure 9).





398 Tropically-averaged EF does not evolve much reflecting seasonal variation in the 399 latitudinal peak in radiation and compensation of decreased canopy interception by 400 transpiration (because of increased net surface radiation) in the dry season. However, in 401 transitional and dry regions to the east, EF exhibits substantial seasonal variation between 402 the wet season, when it peaks, and the dry season. The surface moist static energy flux 403 (assuming sea level elevation) shows variations in SON and JJA but otherwise remains 404 steady across longitudes because of compensation between the increased H and reduced 405 ET. In the dry to wet transition, SON, moist static energy flux exhibits an interesting peak 406 at about -60 longitude (Figure 10) though the combined increase in radiation, due to reduced cloudiness, inducing higher sensible heat flux and maintained high ET rates. 407 408 Over tropical Africa, the precipitation is highest in JJAS during the wet phase of the West African Monsoon, with a secondary maximum in DJF corresponding to the Southern 409 410 African Monsoon (Figure 5). Similarly the latitudinal-averaged GPP and ET increase 411 during the West African Monsoon (Figure 6, Figure 7), accompanied by a strong 412 decrease in sensible heat flux (Figure 8). In DJF the southern African Monsoon displays

413 increased water flux (Figure 7) and photosynthesis tracking the increased rainfall (Figure
414 5). The Congo rainforest clearly exhibits two brief rainy seasons (Figure 4, Figure 9),
415 with peaks in March-April-May and September-October-November (Figure 4) and
416 displays substantial water stress and strong reduction in EF to values below 0.6 during
417 the dry season (Figure 9).

418 3.3 Rainforest water stress

419 One outstanding challenge in modeling tropical land regions is why do most 420 contemporary land-surface models incorrectly represent the wettest rainforest GPP and 421 ET rates, their seasonal cycles, and how they relate to water stress? Capturing this 422 accurately will help better understand the seasonal course of GPP and ET in the tropics.

In the wettest tropical forests, such as the western portion of the Amazon or Indonesia, energy and light limit the rates of ET and GPP. It is thus natural to conclude that soil moisture and water stress have only minor effects in such regions and thus that precipitation variability would not matter much. In fact, there exist sharp vertical gradients in the canopy (as well as at the surface of the soil in the dry season) in terms of light and water availability (along with nutrient allocation) (Figure 3). Understory species receive only a small amount of mostly diffuse light. However, water is not typically





- 430 limiting for low-canopy species. Moreover, because relative humidity is high and VPD is
- 431 low, leading to low stress on understory stomatal and ecosystem conductance [Leuning,
- 432 1995; Leuning et al., 1995; Wang and Leuning, 1998; Medlyn et al., 2011; 2012; Heroult
- 433 *et al.*, 2013].

434 On the other hand, top canopy species receive a large amount of radiation, especially in 435 the dry season, causing sunlit leaf warming and desiccation leading to heat and water 436 stress [Jardine et al., 2014]. Leaf and xylem water status are regulated by the relative 437 demand of sap from transpiration, which depends on incoming radiation, temperature and 438 VPD. It also depends on the supply of sap to the leaves which is controlled by xylem conductivity and reduced by cavitation in the xylem [Martinez-Vilalta et al., 2014; 439 440 Martinez-Vilalta and Garcia-Forner, 2016]. To avoid leaf desiccation and xylem cavitation (formulation of air bubbles blocking the ascent of sap flow from the roots to 441 442 the leaves) stomatal closure is usually observed during peak daytime sunlight hours in 443 rainforest canopy species [Brodribb, 2003; Pons and Welschen, 2003; Zhang et al., 444 2013]. This reduces the drop in leaf and xylem water potential and thus avoids important 445 leaf desiccation or xylem cavitation (Figure 12). This type of behavior with strong 446 stomatal regulation appears to be the norm in the wettest tropical forests [Fisher et al., 447 2006; Konings and Gentine, 2016].

448 In tall canopy species the flow in the xylem from the roots is limited and cannot 449 sufficiently rehydrate the upper xylem and leaves, and it cannot be compensated by the 450 plant internal storage, whereby stomatal shutdown is inevitable to avoid desiccation and 451 xylem cavitation (Figure 12) [Phillips et al., 1997; 2004; Lee et al., 2005; Oliveira et al., 452 2005; Phillips et al., 2008; Scholz et al., 2011; Zeppel et al., 2014; Konings and Gentine, 453 2016]. In summary, water stress in tropical rainforest canopy species is not primarily due 454 to soil water stress but rather to the atmospheric demand and the build up of water stress 455 in the soil-plant continuum. Radiation, temperature and VPD are therefore essential for 456 tropical forests further emphasizing the importance of radiation and light on those forests. 457 Land-surface and ecosystem models, apart from a few exceptions [Xu et al., 2016; Kennedv et al., 2017], do not represent plant hydraulics and typically only rely on an 458 459 empirical reduction of stomatal and ecosystem conductance, and therefore transpiration 460 and GPP, as functions of root-averaged soil moisture or water potential (e.g., [Noilhan 461 and Planton, 1989; Sellers et al., 1996a; 1996b; Ek, 2003; Boulet et al., 2007; Gentine et





462 al., 2007; Ngo-Duc et al., 2007; Stoeckli et al., 2008; Balsamo et al., 2009; Boone et al., 2009a; Bonan et al., 2011; Lawrence et al., 2011; Niu et al., 2011; Bonan et al., 2012; 463 464 Canal et al., 2014; Han et al., 2014; Naudts et al., 2014; De Kauwe et al., 2015; Chaney 465 et al., 2016; Chen et al., 2016; Haverd et al., 2016] among others). The root profile 466 averaging of soil moisture or water potential to define water stress exaggerates the impact 467 of surface drying, as in reality deeper roots may still effectively transport water to the 468 plant xylem even if surface roots experience dry conditions and therefore can maintain 469 overall high rates of GPP and transpiration.

470 The inclusion of plant hydraulics in tall canopy species leads to strong differentiation between leaf (and upper xylem) and soil water potential (Figure 12) during midday, 471 472 especially in the dry season. Indeed, leaf and xylem water potentials substantially drop because of the large transpiration rates through the stomata and because the xylem cannot 473 474 be instantaneously refilled due to the large flow drag in the elongated xylem. As a result, 475 plant hydraulics induce a shutdown of stomata during the day reducing the transpiration 476 rate near peak solar hours, also knownknows as "midday depression," in order to reduce 477 desiccation of the leaf and xylem. In addition, plant hydraulics also induces a natural 478 hydraulic redistribution of water in the root profile reducing dryness in the upper profile 479 in the dry season [Lee et al., 2005; Oliveira et al., 2005; Domec et al., 2010; Prieto and 480 Ryel, 2014; Kennedy et al., 2017], using deep root moisture rather than surface soil 481 moisture when needed, as the water flows down gradient of water potentials. This is 482 fundamentally different from typical parameterizations using average water stress of the 483 root water profile, which are oversensitive to surface water stress, in typical 484 parameterizations [Kennedy et al., 2017]. Both of those effects lead to reduced sensitivity 485 to water stress [Kennedy et al., 2017] and help maintain higher rates of transpiration 486 throughout the entire dry season [Kennedy et al., 2017], whereas typical land surface 487 models overestimate water stress in the dry season [de Gonçalves et al., 2013; 488 Alemohammad et al., 2016; 2017].





# 489 4 Land-atmosphere interactions – local and nonlocal

490 4.1 Local feedback and heterogeneity – shallow clouds (fog and shallow491 convection)

492 We suggest that the most critical land-atmosphere feedbacks in tropical rainforests 493 involve shallow clouds and fog rather than deep convective clouds. Clearly, much of the 494 focus of tropical land-atmosphere interactions has been on feedbacks involving 495 precipitating deep convection, and the impact of heterogeneity on convective rainfall. On 496 the other hand, the coupling of the land surface to radiation has been relatively 497 understudied. Shallow clouds lead to reduced productivity and transpiration [Anber et al., 498 2015a], yet the latter depends on cloud thickness as cumulus (shallow convection) generate more diffuse light and can boost photosynthesis when they are not too thick 499 500 [Ouwersloot et al., 2017]. Fog on the other hand, strongly diminishes the amount of light 501 received by the ecosystems. Fog [Anber et al., 2015a] and shallow clouds [Giangrande et 502 al., 2017] appear to be one of the primary differences between the dry and the wet season 503 (in addition to the preferential occurrence of nighttime mesoscale convective systems in 504 the rainy season, which are not directly relevant for land-atmosphere interactions 505 associated with daytime processes). Low-level cloudiness largely affects the surface 506 incoming radiation by reducing shortwave surface incoming radiation in the wet season, 507 especially in the morning [Anber et al., 2015a; Giangrande et al., 2017], which in turn leads to strong reduction in GPP and ET. These clouds are also tightly connected to 508 509 surface processes and especially the surface energy partitioning. Indeed nighttime fog, which often persists into the early daylight hours, is largely induced by longwave 510 511 temperature cooling, especially in the presence of evening rain in the wet season, which 512 generates dew formation [Anber et al., 2015a]. Shallow clouds are themselves directly 513 forced by surface-generated thermals due to boundary layer processes [de Arellano et al., 514 2014], and they are modified by the sensible and latent heat flux magnitude [de Arellano 515 et al., 2014]. Shallow convection and low-cloud cover are also tightly connected to the 516 seasonality of the forest and to the diurnal cycle [Anber et al., 2015a; Tang et al., 2016; 517 Giangrande et al., 2017].

518 Historically, the study of land-atmosphere interactions in the Tropics, and tropical 519 rainforests in particular, has emphasized effects of heterogeneity, especially due to





520 deforestation, on the generation of deep convection through mesoscale circulations (see 521 [Lawrence and Vandecar, 2015] for a complete review, as well as [Avissar and Pielke, 522 1989; Pielke and Avissar, 1990; Pielke et al., 1991; Dalu et al., 1996; Avissar and Schmidt, 1998; Taylor et al., 2007; 2009; 2011; Rieck et al., 2015; Khanna et al., 2017]). 523 524 The hypothesis behind this is that deforestation reduces EF and surface roughness 525 [Khanna et al., 2017]. The associated increased buoyancy flux over the deforested areas, mostly reflecting a shift toward increased sensible heating, induces mesoscale 526 527 circulations. These circulations enhance cloudiness through local buoyancy fluxes, 528 turbulent kinetic energy generation, and low-level moisture advection from adjacent forested areas, thus providing all the key ingredients for moist convection generation 529 530 [Rieck et al., 2014; 2015]. It seems unlikely however that momentum roughness plays a major role in this high radiation environment [Park et al., 2017], where circulations are 531 532 mostly buoyancy-driven. Instead, the heat and moisture roughness lengths [Park et al., 533 2017] as well as leaf area index and stomatal conductance, which scales the magnitude of 534 the evapotranspiration flux, are the main players, in addition to changes in soil moisture 535 availability, for the circulation.

Induced mesoscale circulations and associated deep convection are clearly observable 536 with remote sensing observations [Khanna et al., 2017] and are more important in the dry 537 538 season [Khanna et al., 2017], when convection is more locally, and regionally, triggered 539 [Anber et al., 2015a; Ghate and Kollias, 2016]. Once precipitation occurs though, cold pools, i.e., density currents induced by ice melt and evaporating rain in downdrafts, 540 541 dominate the surface-induced mesoscale circulation [Rieck et al., 2015], and reduce the 542 surface heterogeneity signal. In the wet season, the relative contribution of local forcing 543 to the total rainfall is small as the bulk of the precipitation is due to mesoscale convective 544 systems or larger-scale systems propagating throughout the basin, less tightly connected 545 to surface and boundary layer processes [Ghate and Kollias, 2016].

Even during the dry season, a large fraction of the Amazon and of Indonesia only experience minimal water stress (Figure 9 and Figure 8) so that increased radiation generates higher rates of photosynthesis (Figure 6) and ET (Figure 7) [*Anber et al.*, 2015a]. As such the radiation feedback of mesoscale-induced clouds may systematically impact clearings and deforested regions (Figure 13) and are more systematic and longer lasting than mesoscale-induced convective rainfall. Fewer studies have studied changes





552 in shallow clouds [Wang et al., 2000; Lawton et al., 2001; Chagnon et al., 2004; Ray et al., 2006; Wang et al., 2009; Pielke et al., 2011; Rieck et al., 2014; Anber et al., 2015a], 553 554 even though the impact of changes in the surface energy partitioning and heterogeneity 555 on low-level clouds is clear and spatially systematic (Figure 1). Given the importance of 556 cloud cover on shortwave radiation and their importance for the differentiation between 557 the dry and wet seasons over wet tropical rainforests we believe that this low-cloud 558 feedback might be quite critical for rainforest ecosystem functioning. Indeed it was 559 pointed out by [Morton et al., 2014; Anber et al., 2015a; Morton and Cook, 2016] that 560 light changes between the dry and wet season due to changes in cloud cover were one of the primary reasons for changes in the seasonality of surface fluxes, in addition to leaf 561 flush out [Lopes et al., 2016; Saleska et al., 2016]. We also note that the shading due to 562 low clouds reduces surface temperature and ecosystem respiration [Mahecha et al., 2010; 563 564 Peterhansel and Maurino, 2011; Thornley, 2011; Hadden and Grelle, 2016; Ballantyne et al., 2017]. So, cloud-induced reductions in respiration can cancel reductions in 565 566 photosynthesis, such that the net effect of cloud shading on net ecosystem exchange is 567 unclear. In an academic study inspired in the thermodynamic characteristics in the Amazonia, [Horn et al., 2015] showed that coupling with the surface leads to a change in 568 the length scales that characterized clouds, and a reduction of the cloud life time. As a 569 570 result, there are larger populations of smaller shallow cumuli.

- In addition to regulating radiative energy balance at the surface, [*Wright et al.*, 2017] have shown that shallow convection transports moisture, provided by plants' transpiration, from the atmospheric boundary layer to the lower troposphere during the late dry season and early dry to wet transition seasons (July-September). This mechanism, referred to as the "shallow convective moisture pump", plays an important role in priming the atmosphere for increasing deep convection (e.g., [*Schiro et al.*, 2016] [*Zhuang et al.*, 2017]), and wet season onset over the Amazon [*Wright et al.*, 2017].
- The results discussed until now omitted the relation between physical processes and the atmospheric composition, and more specifically the role of chemical reactions and aerosol. Over rainforests, the pristine and undisturbed conditions of the atmospheric boundary layer described in the seminal study by [*Garstang and Fitzjarrald*, 1999] are currently undergoing rapid changes due to atmospheric composition modifications. Their direct impact on the radiative and microphysical properties are due to biomass burning





584 and enhancement of concentrations of secondary organic aerosol precursors. Biomass 585 burning in Amazonia leads to increase aerosol optical depth and to abnormal distributions 586 of the heating rate profile. Analyzing systematic experiments performed by large-eddy 587 simulations, [Feingold et al., 2005] studied the processes that lead to the suppression of 588 clouds. Firstly, at the surface there is clear indications that the latent and sensible heat 589 flux are reduced, yielding convective boundary layers characterized by less turbulent 590 intensity and by delays in the morning transition [Barbaro and Arellano, 2014]. Both 591 aspects tend to reduce cloud formations. Secondly, [Barbaro and Arellano, 2014] 592 indicated that the vertical location of the smoke layer is crucial in determining how the cloud characteristics, *i.e* cloud cover, will change. As described by [Feingold et al., 593 594 2005], smoke confined in the well-mixed sub-cloud layer might positively benefit the cloud formation since it distributes the heat uniformly that contributes to enhance 595 596 convection. On the other hand, smoke layers located within the cloud layer tend to stabilize the cloud layer and therefore decrease the possibility of cloud formation. These 597 598 results are very much dependent on the aerosol optical properties defined by their 599 heating, scattering and hygroscopic properties. As a first indicative figure, the mentioned LES study and observations by [Koren et al., 2004] stressed that smoke layers with an 600 aerosol optical depth larger than 0.5 might already lead to cloud suppression by 50%. [Yu 601 602 et al., 2008] have shown observationally that the influence of aerosols on shallow clouds 603 varies with meteorological conditions. When the ambient atmosphere is drier (relative 604 humidity  $\leq 60\%$ ), the aerosol induced cloud burning effect (evaporation of cloud droplets) 605 due to increased absorption of solar radiation by aerosols out-weight the increase of cloud droplets due to aerosol-cloud microphysical effect. The reduced shallow clouds can 606 607 further enhance the surface dryness. In contrast, when the ambient atmosphere is relatively humid (relative humidity ≥60%), the aerosol-cloud microphysical effect out-608 609 weighs the cloud burning effect, leading to an increase of shallow clouds and relative 610 humidity near surface. In so doing, aerosols can amplify the original moisture anomalies 611 near the surface. Aerosols have also shown to increase of the life time of mesoscale 612 convection over Congo and Amazon, due to delay of the precipitation that enhances ice 613 formation and increase lifetime of the mature and decay phase of deep convection 614 [Chakraborty et al., 2016].





615 These modifications are not only related to the direct emission of aerosol, but also to 616 changes in the gas phase chemistry that act as a precursor for the formation of secondary 617 organic aerosol. [Andreae et al., 2002] already described the differences in  $NO_x$  and 618 ozone (O<sub>3</sub>) mixing ratio depending on the Amazonia site. From rather pristine conditions 619 with NO<sub>x</sub> and ozone levels below 0.1 ppb and 20 ppb, to values above 0.1 ppb and 620 maximum levels of  $O_3$  near 50 ppb. Recent field experiments within the Green Ocean 621 Amazon campaign (GoAmazon) (Fuentes et al., 2016; [Martin et al., 2016] corroborate 622 these levels as well as the high levels of the bio-organic compounds, in particular 623 isoprene and monoterpene. Closely related, these changes are accentuated by anthropogenic emissions, i.e. Manaus. The unique distribution of aerosols in Amazonia 624 625 might explain observed differences in deep convection, in particular lighting frequency, between Amazonia, the Maritime continent and the Congo basin [Williams et al. 2004]. 626 627 To represent these chemistry changes and their effect on convection adequately, the 628 dynamic effect that drive processes such as the entrainment of pollutants from the free 629 troposphere need to be taken into account [Vila-Guerau de Arellano et al., 2011]. As a 630 result of this interaction between radiation, the land surface, dynamics and chemical 631 processes, the transition from turbulent clear convective conditions to shallow cloudy convection may be modified in the future. Current efforts in monitoring them and 632 633 improving the parameterizations of convection are under way [Dias et al., 2014]. These 634 efforts should include also in an integrated manner the combined role of dynamics and 635 chemistry to quantify relevant processes like the ventilation of pollutants from the sub-636 cloud layer into the cloud layer, i.e. mass flux parameterizations, under representative Amazon conditions [Ouwersloot et al., 2013]. 637

In addition to affecting cloud microphysics, biomass burning in the tropics significantly 638 639 affects the global carbon budget. For example, in September and October of 2015 fires in 640 the Maritime continent released more terrestrial carbon (11.3 Tg C) than the 641 anthropogenic emissions of the EU (8.9 Tg C) [Huijnen et al., 2016]. The extent of forest 642 fires in this region is tied to El Niño-induced drought conditions, and antecedent SST 643 patterns are closely related to burned area at the global scale, particularly in hotspots concentrated in the tropics [Chen et al., 2016]. Aerosol emissions and biomass burning 644 645 exert a strong control on land-atmosphere coupling of the carbon and water cycles, and 646 the consequences of this coupling is observable globally.





647 1.1. Nonlocal feedback – deep convection and large-scale circulation

Thus far, we have largely viewed land-atmosphere coupling through the lens of local conditions, but how should we modify this view in light of remote influences (see WTG discussion) or coupling between local and larger-scale conditions? Here we illustrate some aspects of how land-atmosphere coupling in the Tropics is impacted by the largerscale.

653 4.1.1 Large-scale coupling, idealized modeling

654 Consider the Lagrangian tendency equation for conservation of atmospheric water 655 vapor, expressed in terms of specific humidity q:

$$656 \quad \frac{dq}{dt} = S(q) \tag{3}$$

where S(q) is the sum of sources and sinks of specific humidity. In the absence of sources and sinks, (3) implies that the specific humidity of a parcel of air is conserved following the atmospheric flow. In what follows, we consider a vertically-integrated form of (3) such that:

661 
$$\left\langle \frac{\partial q}{\partial t} \right\rangle = E - P - \left\langle \mathbf{v}_{\mathbf{H}} \cdot \nabla q \right\rangle - \left\langle \omega \frac{\partial q}{\partial p} \right\rangle$$
 (4)

Here E and P represent, respectively, the surface evapotranspiration source and the 662 precipitation sink of water vapor, while (...) represents a mass-weighted vertical 663 (pressure) integral from the surface (at pressure  $p_s$ ) to the nominal top of the troposphere 664 (at pressure  $p_t$ ), i.e.,  $\langle ... \rangle = \int_{nt}^{ps} ... \frac{dp}{q}$ . The third and fourth terms on the right-hand side 665 (RHS) of (4) are horizontal and vertical moisture advection. Equation (4) is normalized 666 such that  $\langle q \rangle$  has units of mm, thus effectively corresponding to column water vapor, and 667 668 terms on the right hand side are given in units of mm/day. Equation (4) is often used to 669 construct a diagnostic budget of precipitation, or in perturbation form, precipitation anomalies. As a caveat, within the tropics, the dominant large-scale balance in deep 670 671 convecting regions is typically between vertical moisture advection (or equivalently in 672 the vertically-integrated form, moisture convergence) and precipitation, which may limit 673 the utility of (4) in attributing causality.

Using equation (4) as a starting point, [*Lintner and Neelin*, 2007; 2009] constructed a
 framework for estimating *where* spatial transitions between tropical non-precipitating and
 precipitating conditions, referred to as convective margins, should occur. By coupling





the water and energy (surface and atmosphere) equations, and invoking WTG and convective quasi equilibrium assumptions, as well as a zero-surface flux constraint over land, [*Lintner and Neelin*, 2009] derived the following expression for locating the convective margin,  $x_c$ , along a prescribed inflow air-mass trajectory from an initial point over the ocean onto land (see Figure 15 for a schematic overview):

$$682 x_c = L_c \ln\left[\frac{q_c + q_E}{q_0 + q_E}\right] (5)$$

 $L_c$  denotes a length scale defined as  $\frac{v_q Ms}{Mq_p(R_{tog,net}-E)}$  where  $v_q$  is the mean horizontal wind 683 684 field, weighted with respect to the vertical moisture profile. From the WTG temperature equation, and subject to the zero net surface flux constraint over land, the divergent 685 686 component of the large-scale circulation can be related to the net TOA radiative heating, 687  $R_{toa,net}$ . Ms is the dry static stability and  $Mq_p$  the vertical moisture stratification per unit 688 moisture. The moisture values  $q_0$ ,  $q_c$ , and  $q_E$  denote, respectively, the initial inflow air 689 mass moisture, a moisture-related threshold for initiation of deep convection,; and a moisture scale associated with evapotranspiration over the inflow path,  $q_E = \frac{v_q E}{L_c}$ . Because 690 691 of vertical integration, these quantities are column integrated values.

Note that the advantage of coupling the atmospheric moisture equation to the temperature equation is that under the WTG approximation, the divergent component of the flow is itself diagnosed, which can be instructive for identifying mechanisms involved. Also, in equation (5) and the definition of  $q_E$ , we have assumed that  $L_c > 0$  and that convergent low-level flow is signed positive.

697 Evapotranspiration gives rise to two opposing effects on  $x_c$ . First, with increasing E,  $q_E$ 698 should increase, which causes  $x_c$  to decrease, i.e., moistening from evapotranspiration 699 experienced along the inflow path leads to the convective margin being reached closer to 700 the inflow point. Second, as E increases,  $L_c$  increases: this can be understood as the 701 indirect effect of E, acting through reduction of convergence along the flow path, which 702 shifts the onset point for deep convection away from the inflow point (see Figure 15).

Lintner et al. (2013) developed an idealized prototype for diagnosing large-scale landatmosphere coupling constructed from the idealized temperature and moisture equations used in developing the convective margins model described, but further coupled to a simple bucket soil moisture model. From this model, [*Lintner et al.*, 2013] derived an





707 analytic expression for the sensitivity of precipitation to soil moisture variation from 708 which it is possible to infer dependences on key model parameters, such as the timescale 709 for convective adjustment (assumed in the Betts and Miller-type convection scheme 710 applied), cloud-radiative feedback strength, and surface turbulent flux exchange. 711 [Schaefli et al., 2012] developed a conceptually similar model from which an analytic 712 expression for the ratio of evaporated moisture integrated along flow path to precipitation 713 (or recycling ratio) was obtained (Figure 16). We suggest that such idealized model 714 frameworks, which consider tropical land-atmosphere interactions by coupling both water 715 and energy cycles, should continue to be brought to bear on observations as well as more 716 sophisticated regional or global climate or earth system models, as they can be helpful in 717 diagnosing linkages between local and non-local feedbacks.

718 4.1.2 Coupling

[*Green et al.*, 2017] recently developed a method to define the feedback between the
biosphere and atmosphere using multivariate conditional Granger causality (based on
lagged autoregressive vectors). We here use a similar framework using ET from
WECANN and precipitation from GPCP as well as photosynthetically active radiation
from CERES (Figure 18).

724 Most of the feedback between surface ET and precipitation occurs in the spatial 725 transitional, Monsoonal regions, such as the Savanna region of Northeastern Brazil, the 726 Monsoonal region of the Sahel and Southern Africa, as well as India and Northern 727 Australia. In Brazil, these results are consistent with the above-mentioned concept of 728 convective margin and the impact of soil moisture and transpiration rate on the location 729 of the transition between the dry and wet regions. The Sahelian and Southern African 730 Monsoon are also located in regions between very dry (deserts) and humid regions, where 731 surface feedback may be crucial for the penetration of the Monsoonal flow inland 732 [Lintner and Neelin, 2009; Lintner et al., 2015]. Indeed, the biosphere in this region 733 modulates the local climate state: multiple equilibrium states, corresponding to different 734 ecosystem initial conditions, exist under the same external forcing [Wang et al., 2000]. 735 The effect of vegetation on land-atmosphere coupling manifests itself at multiple 736 timescales. At short timescales after precipitation, evaporation is accelerated with 737 intercepted water in the canopy. However, at longer timescales vegetation acts to delay 738 and prolong evaporation of water stored in the root zone. The magnitude and timescale of





these sources of water recycling will vary depending on ecosystem structure, including rooting depth and canopy structure, which may co-evolve with atmospheric conditions at the interannual timescale [*Nicholson*, 2000]. This represents a clear pathway for two-way feedbacks between the land surface and precipitation.

743 We further emphasize that those feedbacks (Figure 18) are likely to also be influenced by 744 non-local conditions, with regional and large-scale changes in ocean to land flow and the 745 in-land distance of penetration influencing local coupling. We note that climate models 746 seem to exhibit soil moisture (and therefore evapotranspiration)- precipitation feedbacks 747 in similar tropical regions, when averaged across models, even though individual model response varies [Koster et al., 2011; Seneviratne, 2013] (one degree pixel and monthly 748 749 time scales). We emphasize that the PAR radiation product is very uncertain in the tropics [Jim nez et al., 2011] as it ultimately relies on a model to obtain surface incoming 750 751 radiation, which might explain the reduced feedback strength. It is also likely that the 752 bulk of the radiative feedbacks are taking place at smaller times scales such as the ones 753 observed with MODIS (Figure 14). This shallow cloud cover is relatively steady spatially 754 and in time, especially in the dry season.

755 4.1.3 Moisture tracking and source attribution

756 A fundamental consideration in the study of the hydrologic cycle over tropical 757 continents is where the moisture for precipitation ultimately derives. As [van der Ent et 758 al., 2010; van der Ent and Tuinenburg, 2017] note, this consideration is not merely of 759 academic interest: indeed, it is quite likely that anthropogenic modification of the land surface has altered terrestrial evapotranspiration (as well as runoff) to impact 760 precipitation. A common approach to moisture source attribution over tropical land 761 762 regions involves deriving air mass histories using Lagrangian trajectories. Such 763 trajectories are obtained by temporally integrating the 3-dimensional wind field to 764 estimate the positions of idealized air mass parcels through time. Trajectories can be 765 computed in either a forward or backward sense: the latter are initialized from an arrival 766 point and integrated backward through time. Combining a Lagrangian back trajectory 767 approach with rainfall and leaf area index data, [Spracklen et al., 2012] quantified the 768 linkage between downstream rainfall amount and upstream air mass exposure to 769 vegetation (Figure 17). Over more than half of the tropical land surface, the Spracklen et 770 al. estimates indicate a twofold increase in downstream rainfall for those air masses





passing over extensive vegetation compared those passing over little upstream vegetation.

772 Based on these estimates and extrapolating current Amazonian deforestation trends in the

773 future, these authors project wet and dry season rainfall decreases of 12 and 21%,

respectively, by the year 2050.

775 Other analyses using air mass histories have demonstrated the significance of terrestrial *E* 776 sources for remote land regions. For example, [*Drumond et al.*, 2014] used the 777 FLEXPART model forced with ERA-Interim reanalysis to estimate E - P along 778 trajectories passing over the La Plata Basin in subtropical South America to establish that 779 much of the moisture entering this region derives from the Amazon Basin to the north 780 and west.

781

#### 782 4.1.4 Seasonality and seasonal transitions

783 One of the outstanding issues in the study of tropical land region climates involves 784 controls on precipitation seasonality, particularly its regional variability. To leading order, the seasonality follows the variation in maximum solar heating, but other factors, 785 786 such as ocean thermal inertia, topography, dynamics and circulation, and moisture 787 transport, as well as the state of the land surface, can exert considerable influence on the timing and amplitude of tropical land region seasonal evolution. Over the Amazon basin, 788 seasonality exhibits marked variation in both latitude and longitude: for example, at 5S, 789 790 the dry-to-west transition proceeds from the central Amazon eastward toward the Atlantic 791 coast [Liebmann and Marengo, 2001]. It is also worth noting a pervasive tendency for 792 the dry-to-wet season transition to occur much more rapidly than the wet-to-dry 793 transition, as evident in tropical monsoon systems including South Asia, West Africa, and 794 South America.

795 Analyzing multiple observational and reanalysis products, [Fu and Li, 2004] 796 identified a strong influence of surface turbulent fluxes on the dry-to-wet transition and 797 its interannual variability over the Amazon. In particular, their results link earlier wet 798 season onset to wetter conditions in the antecedent dry season: the higher latent fluxes at 799 the end of a wetter dry season encourage weaker convective inhibition (CIN) but enhanced CAPE, both of which are more favorable to wet season rainfall occurrence. 800 801 However, these authors also underscore the participation of the large-scale circulation 802 and its role in establishing a background environment (e.g., moisture convergence) to





support wet season rainfall. Incursion of cold fronts into the southern Amazon may act as
triggers for rapid initiation of wet season onset once the local thermodynamics become
favorable [*Li et al.*, 2006].

806 Recent researches suggest that the land-atmospheric coupling plays a central role in 807 determining the earlier timing of the wet season onset over western and southern 808 Amazonia, relative to that of eastern Amazonia. Both in situ and satellite ecological 809 observations have consistently shown that rainforests increase their photosynthesis, thus 810 evapotranspiration (ET), during late dry season across Amazonia (e.g., [Huete et al., 811 2006; Lopes et al., 2016; Munger et al., 2016; Wehr et al., 2016]). The wet season onset over the southern hemispheric western and southern Amazonia occurs during September 812 to October, about two to three months before the arrival of the Atlantic ITCZ [Fu et al., 813 814 2016]. Using several satellite measurements, including deuterium (HDO) of the 815 atmospheric water vapor and SIF, Wright et al (2017) have shown that such an increase 816 of ET in the late dry season is the primary source of increasing water vapor in the lower 817 troposphere that initiates the increase of deep convection and rainfall over southern 818 Amazonia. In particular, the increase of water vapor with enriched HDO in the boundary 819 layer and free troposphere, follows the increase of photosynthesis during late dry season. The HDO value of the atmospheric moisture is too high to be explained by transport from 820 821 Atlantic Ocean, and is consistent with that from plant transpiration. Such a moistening of 822 the atmosphere starts in western southern Amazonia, the part of Amazonia that is most remote from the Atlantic Ocean with high biomass. It then progresses towards eastern 823 824 southern Amazonia. Thus, during the late dry season this appears to contribute to the 825 timing and spatial variation of the initial moistening of the atmosphere, that ultimately 826 lead to wet season onset over southern Amazonia.

Wet season onset over southern Amazonia has been delaying since the late 1970s [*Marengo et al.*, 2011; *Fu et al.*, 2013]. In addition to the influence of global circulation change, such a change has been attributed to land use. For example, [*Butt et al.*, 2011] have compared long-term rainfall data between deforested and forested areas over part of the southern Amazonia. They observed a significant delay in wet season onset over the deforested areas, consistent with that implied by Wright et al. (2017). In addition, [*Zhang et al.*, 2008; 2009] have shown that biomass burning aerosols, which peak in late dry





834 season, can also weaken and delay dry to wet season transition by stabilizing the 835 atmosphere, reducing clouds and rainfall.

#### 836 5 Discussion - conclusions

837 In this review paper, we have discussed some of the important aspects of land-838 atmosphere interactions pertaining to the tropics. This review article is by no means 839 exhaustive but rather provides insights into some of the important coupled land-840 atmosphere processes at play in the tropics and in rainforest ecosystems in particular.

We have argued that feedbacks between the land surface and precipitation in the tropics are possibly non-local in nature and mostly impact moisture advection from the ocean and the position of deep convection onset. Local rainfall feedback associated with mesoscale heterogeneities appear to be rather small in magnitude, at least compared to the annual-mean rainfall, and not sufficiently spatially systematic to truly affect ecosystem functioning.

847 Moreover, we contend that land surface-cloud feedbacks, especially those involving 848 shallow clouds and fog, are critical in terms of regulating light (direct and diffuse), 849 temperature, and water vapor deficit over tropical forest, but such feedbacks have 850 received relatively little attention. Remote sensing platforms provide useful information 851 for quantifying such feedbacks, but these need to be complemented by ground 852 measurements (especially of photosynthetic rates and respiration). Eddy-covariance 853 measurements may prove difficult to use, as mesoscale circulations alter the homogeneity 854 assumption of eddy-covariance methods.

855 We have also discussed errors and biases in the representation of tropical continental climates in current generation climate and Earth system models. The average soil 856 857 moisture-precipitation feedback strength across earth system models (based on the 858 GLACE experiment) [Koster et al., 2004] tend to exhibit land-precipitation feedbacks in 859 similar transitional regions as the ones observed, which seems to be mostly related to 860 modification of the moisture advection penetration distance from the ocean rather than to local feedbacks. These feedbacks appear to be of relatively minor importance in the core 861 862 of tropical rainforests but are more critical for more marginal rainfall regions (savanna). 863 These regions are of critical importance for the terrestrial global carbon cycle, providing 864 the main terrestrial sink, but might be severely impacted by climate change and droughts





in particular [*Laan Luijkx et al.*, 2015]. Whether the interannual variability in surface  $CO_2$ flux in those regions is a zero-sum game with wet years compensating dry years still is an

 $867 \qquad \text{open question especially in the context of rising CO}_2 \text{ concentration.}$ 

- The core of rainforests seems to be more affected by radiation feedbacks at relatively small spatial scales (~1km), which can be dramatically influenced by land cover and land use change. Projected rates of future deforestation are poorly constrained, especially regionally, though in recent years, the Congo and Indonesia have experienced increasing
- 872 deforestation while the deforestation rate in the Amazon has dropped.

873 Earth system models tend to predict very diverse responses to global warming leading to 874 broad spread in the capacity of rainforests to continue to act as net carbon sinks [Swann et 875 al., 2015] in the future. Indeed, in the Amazon in particular, the models' response varies from becoming much drier to more humid. El Niño events are sometimes thought as a 876 877 proxy of global warming in the tropics [Pradipta et al., 2016] as they warm the free-878 troposphere. Nonetheless for continents the change in the Walker circulation associated 879 with El Niño may strongly differ from the change associated with a more uniform sea 880 surface temperature warming in future climate. In particular, mature El Niño events are 881 associated with strong subsidence over Indonesia, increased ascent off the coast of Peru 882 but reduced precipitation over the Amazon basin and a relatively neutral response over 883 the Congo basin. With SST warming across the tropics, the Maritime continent will most 884 likely become wetter [Byrne and O'Gorman, 2015; Wills et al., 2016]. The fate of the 885 Amazon basin is less clear, as the climate in the region will be impacted by a 886 combination of free tropospheric warming stabilizing the atmosphere to deep convection 887 while warming of the Atlantic enhances the low-level MSE of inflow into the basin. 888 Additionally, warming-induced changes to large-scale circulation such as the intensity or orientation of low-level Atlantic trade winds could impact Amazonian precipitation 889 890 change. Knowledge of the Congo basin remains limited but it appears that the basin will 891 become dryer under the combined effect of increased temperature and reduced 892 precipitation [Greve et al., 2014]. One important question involves how the effect of 893 rising [CO<sub>2</sub>] modifies surface energy flux partitioning though changes in stomatal 894 physiology and modify the regional climate though land-atmosphere interactions 895 [Lemordant, 2016]. 896





- 897 Acknowledgments. This work was supported by Pierre Gentine's new investigator
- 898 grant NNX14AI36G, DOE Early Career grant DE-SC0014203, NSF CAREER and
- 899 GoAmazon DE-SC0011094. We would like to acknowledge high-performance computing
- 900 support from Yellowstone (ark:/85065/d7wd3xhc) provided by NCAR's Computational
- 901 and Information Systems Laboratory, sponsored by the National Science Foundation.

902





## 903 REFERENCES

904

905	Acevedo, O. C., O. L. L. Moraes, R. da Silva, D. R. Fitzjarrald, R. K. Sakai, R. M.
906	Staebler, and M. J. Czikowsky, Inferring nocturnal surface fluxes from vertical
907	profiles of scalars in an Amazon pasture, Global Change Biol, 10(5), 886-894,
908	doi:10.1111/j.1529-8817.2003.00755.x, 2004.

- Alemohammad, S. H., B. Fang, A. G. Konings, F. Aires, J. K. Green, J. Kolassa, D.
  Miralles, C. Prigent, and P. Gentine, Water, Energy, and Carbon with Artificial
  Neural Networks (WECANN): a statistically based estimate of global surface
  turbulent fluxes and gross primary productivity using solar-induced fluorescence, *Biogeosciences*, 14(18), 4101–4124, doi:10.5194/bg-14-4101-2017, 2017.
- Alemohammad, S. H., B. Fang, A. G. Konings, J. K. Green, J. Kolassa, C. Prigent, F.
  Aires, D. Miralles, and P. Gentine, Water, Energy, and Carbon with Artificial Neural
  Networks (WECANN): A statistically-based estimate of global surface turbulent
- 917 fluxes using solar-induced fluorescence, 1–36, doi:10.5194/bg-2016-495, 2016.
- Alexander, J. N., J. R. Peter, and N. K. Ernest, Assimilating solar-induced chlorophyll
  fluorescence into the terrestrial biosphere model BETHY-SCOPE: Model description
  and information content, *geosci-model-dev-discuss.net*, doi:10.5194/gmd-2017-34,
  2017.
- Anber, U., P. Gentine, S. Wang, and A. H. Sobel, Fog and rain in the Amazon, *Proceedings of the National Academy of Sciences*, 112(37), 11473–11477,
  doi:10.1073/pnas.1505077112, 2015.
- Anber, U., S. Wang, and A. Sobel, Effect of Surface Fluxes versus Radiative Heating on
  Tropical Deep Convection, *Journal of Atmospheric Sciences*, 72(9), 3378–3388,
  doi:10.1175/JAS-D-14-0253.1, 2015.
- Andreae, M.O., Artaxo, P., Brandao, C., Carswell, F.E., Ciccioli, P., Da Costa, A.L.,
  Culf, A.D., Esteves, J.L., Gash, J.H.C., Grace, J. and Kabat, P., Biogeochemical
  cycling of carbon, water, energy, trace gases, and aerosols in Amazonia: The LBA-





931	EUSTACH experiments, J Geophys Res-Atmos, 107(D20),
932	doi:10.1029/2001JD000524, 2002.
933	Andreasen, M., K. H. Jensen, D. Desilets, M. Zreda, H. Bogena, and M. C. Looms, Can
934	canopy interception and biomass be inferred from cosmic-ray neutron intensity?
935	Results from neutron transport modeling,, 1-42, doi:10.5194/hess-2016-226, 2016
936	Avissar, R., and C. Nobre, Preface to special issue on the Large-Scale Biosphere-
937	Atmosphere Experiment in Amazonia (LBA), J Geophys Res-Atmos, 107, -,
938	doi:10.1029/2002JD002507, 2002
939	Avissar, R., and R. A. Pielke, A parameterization of heterogeneous land surfaces for
940	atmospheric numerical models and its impact on regional meteorology, Mon Wea
941	<i>Rev.</i> , 1989.
942	Avissar, R., and T. Schmidt, An evaluation of the scale at which ground-surface heat flux
943	patchiness affects the convective boundary layer using large-eddy simulations, $J$
944	Atmos Sci, 55(16), 2666–2689, 1998.
945	Avissar, R., P. Dias, M. Dias, and C. Nobre, The Large-Scale Biosphere-atmosphere
946	Experiment in Amazonia (LBA): Insights and future research needs, J Geophys Res-
947	Atmos, 107(D20), doi:10.1029/2002JD002704, 2002.
948	Baldocchi, D., Falge, E., Gu, L., Olson, R., Hollinger, D., Running, S., Anthoni, P.,
949	Bernhofer, C., Davis, K., Evans, R. and Fuentes, J., FLUXNET: A new tool to study
950	the temporal and spatial variability of ecosystem-scale carbon dioxide, water vapor,
951	and energy flux densities, Bull. Amer. Meteor. Soc., 82(11), 2415-2434, 2001.
952	Ballantyne, A., Smith, W., Anderegg, W., Kauppi, P., Sarmiento, J., Tans, P.,
953	Shevliakova, E., Pan, Y., Poulter, B., Anav, A. and Friedlingstein, P., Accelerating
954	net terrestrial carbon uptake during the warming hiatus due to reduced respiration,
955	Nature Climate Change, 7(2), 148–152, doi:10.1038/nclimate3204, 2017.
956	Balsamo, G., A. Beljaars, K. Scipal, P. Viterbo, B. van den Hurk, M. Hirschi, and A. K.
957	Betts, A Revised Hydrology for the ECMWF Model: Verification from Field Site to





- 958 Terrestrial Water Storage and Impact in the Integrated Forecast System, J
  959 Hydrometeorol, 10(3), 623–643, doi:10.1175/2008JHM1068.1, 2009.
- Barbaro, E., and J. Arellano, Aerosols in the convective boundary layer: Shortwave
  radiation effects on the coupled land-atmosphere system , *J of Climate*,
  doi:10.1002/(ISSN)2169-8996., 2014
- Bechtold, P., N. Semane, P. Lopez, J.-P. Chaboureau, A. Beljaars, and N. Bormann,
  Representing equilibrium and non-equilibrium convection in large-scale models, J *Atmos Sci*, 71(2), 130919100122007–753, doi:10.1175/JAS-D-13-0163.1, 2013.
- Bechtold, P., N. Semane, P. Lopez, J.-P. Chaboureau, A. Beljaars, and N. Bormann,
  Representing Equilibrium and Nonequilibrium Convection in Large-Scale Models, J *Atmos Sci*, 71(2), 734–753, doi:10.1175/JAS-D-13-0163.1, 2014.
- Betts, A. K., and M. A. F. Silva Dias, Progress in understanding land-surface-atmosphere
  coupling from LBA research, *J. Adv. Model. Earth Syst.*, 2, 6,
  doi:10.3894/JAMES.2010.2.6, 2010.
- Betts, A. K., L. V. Gatti, and A. M. Cordova, Transport of ozone to the surface by
  convective downdrafts at night, *J of Climate*, 2002.
- Boisier, J. P., P. Ciais, A. Ducharne, and M. Guimberteau, Projected strengthening of
  Amazonian dry season by constrained climate model simulations, *Nature Climate Change*, 5(7), 656–660, doi:10.1038/nclimate2658, 2015.
- Bonan, G. B., K. W. Oleson, and R. A. Fisher, Reconciling leaf physiological traits and
  canopy flux data: Use of the TRY and FLUXNET databases in the Community Land
  Model version 4, *J of Climate*, 2012.
- Bonan, G. B., P. J. Lawrence, K. W. Oleson, S. Levis, M. Jung, M. Reichstein, D. M.
  Lawrence, and S. C. Swenson, Improving canopy processes in the Community Land
  Model version 4 (CLM4) using global flux fields empirically inferred from
  FLUXNET data, *J Geophys Res-Biogeo*, *116*, 1–22, doi:10.1029/2010JG001593,
  2011.





985 986 987	<ul><li>Boone, A., De Rosnay, P., Balsamo, G., Beljaars, A., Chopin, F., Decharme, B., Delire, C., Ducharne, A., Gascoin, S., Grippa, M. and Guichard, F., The AMMA Land Surface Model Intercomparison Project (ALMIP), <i>Bull. Amer. Meteor. Soc.</i>, 90(12),</li></ul>
988	1865–1880, doi:10.1175/2009BAMS2786.1, 2009.
989	Boone, A., A. Getirana, J. Demarty, and B. Cappelaere, The African Monsoon
990	Multidisciplinary Analyses (AMMA) Land surface Model Intercomparison Project
991	Phase 2 (ALMIP2), GEWEX News, 2009
992	Boulet, G., A. Chehbouni, P. Gentine, B. Duchemin, J. Ezzahar, and R. Hadria,
993	Monitoring water stress using time series of observed to unstressed surface
994	temperature difference, Agr Forest Meteorol, 146, 159-172,
995	doi:10.1016/j.agrformet.2007.05.012, 2007.
996	Böing, S. J., H. J. J. Jonker, A. P. Siebesma, and W. W. Grabowski, Influence of the
997	Subcloud Layer on the Development of a Deep Convective Ensemble, J Atmos Sci,
998	69(9), 2682–2698, doi:10.1175/JAS-D-11-0317.1, 2012.
999	Bretherton, C. S., and S. Park, A New Moist Turbulence Parameterization in the
1000	Community Atmosphere Model, J Climate, 22(12), 3422-3448,
1001	doi:10.1175/2008JCLI2556.1, 2009.
1002	Brodribb, T. J. Stomatal Closure during Leaf Dehydration, Correlation with Other Leaf
1003	Physiological Traits, Plant Physiol, 132(4), 2166-2173, doi:10.1104/pp.103.023879,
1004	2003.
1005	Butt, N., P. A. de Oliveira, and M. H. Costa, Evidence that deforestation affects the onset
1006	of the rainy season in Rondonia, Brazil, Journal of Geophysical Research:
1007	Atmospheres (1984–2012), 116(D11), 407, doi:10.1029/2010JD015174, 2011.
1008	Byrne, M. P., and P. A. O'Gorman, The Response of Precipitation Minus
1009	Evapotranspiration to Climate Warming: Why the "Wet-Get-Wetter, Dry-Get-
1010	Drier" Scaling Does Not Hold over Land, J Climate, doi:10.1175/JCLI-D-15-
1011	0369.s1, 2015.





- 1012 Campos, J. G., O. C. Acevedo, J. Tota, and A. O. Manzi, On the temporal scale of the 1013 turbulent exchange of carbon dioxide and energy above a tropical rain forest in
- 1014 Amazonia, J Geophys Res, 114(D8), D08124–10, doi:10.1029/2008JD011240., 2009
- 1015 Canal, N., J. C. Calvet, B. Decharme, D. Carrer, S. Lafont, and G. Pigeon, Evaluation of
  1016 root water uptake in the ISBA-A-gs land surface model using agricultural yield
  1017 statistics over France, *Hydrol Earth Syst Sc*, 18(12), 4979–4999, doi:10.5194/hess1018 18-4979-2014-supplement, 2014.
- 1019 Chagnon, F. J. F., R. L. Bras, and J. Wang, Climatic shift in patterns of shallow clouds
  1020 over the Amazon, *Geophys Res Lett*, 31(24), L24212, doi:10.1029/2004GL021188,
  1021 2004.
- 1022 Chaney, N. W., J. D. Herman, M. B. Ek, and E. F. Wood, Deriving Global Parameter
  1023 Estimates for the Noah Land Surface Model using FLUXNET and Machine
  1024 Learning, *J Geophys Res-Atmos*, 1–41, doi:10.1002/2016JD024821, 2016.
- 1025 Chen, Y., Ryder, J., Bastrikov, V., McGrath, M.J., Naudts, K., Otto, J., Ottlé, C., Peylin,
  1026 P., Polcher, J., Valade, A. and Black, A., Evaluating the performance of land surface
  1027 model ORCHIDEE-CAN v1.0 on water and energy flux estimation with a single- and
  1028 multi-layer energy budget scheme, *Geosci Model Dev*, 9(9), 2951–2972,
  1029 doi:10.5194/gmd-9-2951-2016, 2016.
- 1030 Chor, T. L., N. L. Dias, A. Araújo, S. Wolff, E. Zahn, A. Manzi, I. Trebs, M. O. Sá, P. R.
  1031 Teixeira, and M. Sörgel, Flux-variance and flux-gradient relationships in the
  1032 roughness sublayer over the Amazon forest, *Agr Forest Meteorol*, 239, 213–222,
  1033 doi:10.1016/j.agrformet.2017.03.009, 2017.
- Cook, B. I., S. P. Shukla, M. J. Puma, and L. S. Nazarenko, Irrigation as an historical
  climate forcing, *Climate dynamics*, 44(5-6), 1715–1730, doi:10.1007/s00382-0142204-7, 2014.
- 1037 Couvreux, F., Roehrig, R., Rio, C., Lefebvre, M.P., Caian, M., Komori, T., Derbyshire,
  1038 S., Guichard, F., Favot, F., D'Andrea, F. and Bechtold, P., Representation of daytime
  1039 moist convection over the semi-arid Tropics by parametrizations used in climate and





1040	meteorological models, Q J Roy Meteor Soc, 141(691), 2220-2236,
1041	doi:10.1002/qj.2517, 2015.
1042 1043 1044	Couvreux, F., C. Rio, and F. Guichard, Initiation of daytime local convection in a semi- arid region analysed with high-resolution simulations and AMMA observations - Quarterly Journal of the Royal Meteorological Society, 2011
1045	Couvroux F. C. Pio, F. Guichard, M. Lothon, G. Canut, D. Bounial and A. Gounou
1045	Initiation of davtime local convection in a semi-arid region analysed with high-
1040	resolution simulations and AMMA observations O LBoy Mataor Soc. 138(662) 56
1047	71  doi: 10 1002/ci 002 2011
1048	/1, doi:10.1002/qj.905, 2011.
1049	Da Rocha, H. R. et al., Patterns of water and heat flux across a biome gradient from
1050	tropical forest to savanna in Brazil, J Geophys Res, 114, G00B12,
1051	doi:10.1029/2007JG000640, 2009.
1052	Daleu, C. L., S. J. Woolnough, and R. S. Plant (2012), Cloud-Resolving Model
1053	Simulations with One- and Two-Way Couplings via the Weak Temperature Gradient
1054	Approximation, <i>J Atmos Sci</i> , 69(12), 3683–3699, doi:10.1175/JAS-D-12-058.1.
1055	Daleu, C. L., S. J. Woolnough, and R. S. Plant, Transition from suppressed to active
1056	convection modulated by a weak-temperature gradient derived large-scale
1057	circulation, J Atmos Sci, 141023140130007, doi:10.1175/JAS-D-14-0041.1, 2014.
1058	Dalu, G. A., R. A. Pielke, M. Baldi, and X. Zeng, Heat and momentum fluxes induced by
1059	thermal inhomogeneities with and without large-scale flow, Journal of the
1060	atmospheric Science, 1996.
1061	Davidson, E.A., de Araújo, A.C., Artaxo, P., Balch, J.K., Brown, I.F., Bustamante, M.M.,
1062	Coe, M.T., DeFries, R.S., Keller, M., Longo, M. and Munger, J.W., The Amazon
1063	basin in transition, Nature, 481(7381), 321–328, doi:10.1038/nature10717, 2012
1064	de Arellano I V-G H G Ouwersloot D Baldocchi and C M I Jacobs Shallow
1065	cumulus rooted in photosynthesis Geophys Res Lett doi:10.1002/2014GL050270
1065	201 <i>A</i>
1000	2017.




- 1067 De Kauwe, M. G., J. Kala, Y. S. Lin, A. J. Pitman, B. E. Medlyn, R. A. Duursma, G. Abramowitz, Y. P. Wang, and D. G. Miralles, A test of an optimal stomatal conductance scheme within the CABLE land surface model, *Geosci Model Dev*, 8(2), 431–452, doi:10.5194/gmd-8-431-2015, 2015.
  1071 Del Genio, A. D., and J. Wu, The Role of Entrainment in the Diurnal Cycle of
- 1071 Der Genio, A. D., and J. wu, The Kole of Enhamment in the Diumai Cycle of 1072 Continental Convection, *J Climate*, 23(10), 2722-2738, 1073 doi:10.1175/2009JCLI3340.1, 2010.
- 1074 Domec, J.-C., J. S. King, A. Noormets, E. Treasure, M. J. Gavazzi, G. Sun, and S. G.
  1075 McNulty, Hydraulic redistribution of soil water by roots affects whole-stand
  1076 evapotranspiration and net ecosystem carbon exchange, *New Phytologist*, *187*(1),
  1077 171–183, doi:10.1111/j.1469-8137.2010.03245.x, 2010.
- 1078 Doughty, C.E., Metcalfe, D.B., Girardin, C.A.J., Amézquita, F.F., Cabrera, D.G., Huasco,
  1079 W.H., Silva-Espejo, J.E., Araujo-Murakami, A., Da Costa, M.C., Rocha, W. and
  1080 Feldpausch, T.R., Drought impact on forest carbon dynamics and fluxes in
  1081 Amazonia, *Nature*, 519(7541), 78–82, doi:10.1038/nature14213, 2015.
- 1082 Drager, A. J., and S. C. van den Heever, Characterizing convective cold pools, *J. Adv.*1083 *Model. Earth Syst.*, 1–55, doi:10.1002/2016MS000788, 2017.
- Drumond, A., J. Marengo, T. Ambrizzi, R. Nieto, L. Moreira, and L. Gimeno, The role of
  the Amazon Basin moisture in the atmospheric branch of the hydrological cycle: a
  Lagrangian analysis, *Hydrol Earth Syst Sc*, 18(7), 2577–2598, doi:10.5194/hess-182577-2014, 2014.
- Du, S., L. Liu, X. Liu, and J. Hu, Response of Canopy Solar-Induced Chlorophyll
  Fluorescence to the Absorbed Photosynthetically Active Radiation Absorbed by
  Chlorophyll, *Remote Sensing*, 9(9), 911–19, doi:10.3390/rs9090911, 2017.
- 1091 Duveiller, G., and A. Cescatti, Spatially downscaling sun-induced chlorophyll
  1092 fluorescence leads to an improved temporal correlation with gross primary
  1093 productivity, *Remote Sensing of Environment*, 182(C), 72–89,
  1094 doi:10.1016/j.rse.2016.04.027, 2016.





- D'Andrea, F., P. Gentine, and A. K. Betts, Triggering deep convection with a
  probabilistic plume model, *J Atmos Sci*, 71(11), 3881–3901, doi:10.1175/JAS-D-130340.1, 2014.
- 1098 Ek, M. B., Implementation of Noah land surface model advances in the National Centers
  1099 for Environmental Prediction operational mesoscale Eta model, *J Geophys Res*,
  1100 108(D22), 8851, doi:10.1029/2002JD003296, 2003.
- Engerer, N. A., D. J. Stensrud, and M. C. Coniglio, Surface Characteristics of Observed
  Cold Pools, *Mon Wea Rev*, *136*(12), 4839–4849, doi:10.1175/2008MWR2528.1,
  2008.
- Feingold, G., H. L. Jiang, and J. Y. Harrington, On smoke suppression of clouds in
  Amazonia, *Geophys Res Lett*, 32(2), doi:10.1029/2004GL021369, 2005.
- Feng, Z., S. Hagos, A. K. Rowe, C. D. Burleyson, M. N. Martini, and S. P. de Szoeke,
  Mechanisms of convective cloud organization by cold pools over tropical warm
  ocean during the AMIE/DYNAMO field campaign, *J. Adv. Model. Earth Syst.*, 7(2),
  357–381, doi:10.1002/2014MS000384, 2015.
- Fisher, J.B., Malhi, Y., Bonal, D., Da Rocha, H.R., De Araujo, A.C., Gamo, M., Goulden,
  M.L., Hirano, T., Huete, A.R., Kondo, H. and Kumagai, T.O., The land-atmosphere
  water flux in the tropics, *Global Change Biol*, *15*(11), 2694–2714,
  doi:10.1111/j.1365-2486.2008.01813.x, 2009.
- Fisher, R. A., M. Williams, R. L. Do Vale, A. L. Da Costa, and P. Meir, Evidence from
  Amazonian forests is consistent with isohydric control of leaf water potential, *Plant Cell Environ*, 29(2), 151–165, 2006.
- Fitzjarrald, D. R., Turbulent trasnport just above the Amazon, *J Geophys Res-Atmos*, 1–
  1118 13, 2007.
- Frankenberg, C., Fisher, J.B., Worden, J., Badgley, G., Saatchi, S.S., Lee, J.E., Toon,
  G.C., Butz, A., Jung, M., Kuze, A. and Yokota, T., New global observations of the
  terrestrial carbon cycle from GOSAT: Patterns of plant fluorescence with gross





1122	primary productivity, <i>Geophys Res Lett</i> , 38(17), n/a–n/a,						
1125	doi:10.1029/2011GL048/38, 2011.						
1124	Frankenberg, C., C. O'Dell, J. Berry, L. Guanter, J. Joiner, P. Köhler, R. Pollock, and T.						
1125	E. Taylor, Prospects for chlorophyll fluorescence remote sensing from the Orbiting						
1126	Carbon Observatory-2, Remote Sensing of Environment, 147(C), 1-12,						
1127	doi:10.1016/j.rse.2014.02.007, 2014.						
1128	Frankenberg, C., C. O'Dell, L. Guanter, and J. McDuffie, Remote sensing of near-						
1129	infrared chlorophyll fluorescence from space in scattering atmospheres: implications						
1130	for its retrieval and interferences with atmospheric CO2 retrievals, Atmos Meas Tech,						
1131	5(8), 2081–2094, doi:10.5194/amt-5-2081-2012, 2012.						
1132	Fu, R., and W. Li, The influence of the land surface on the transition from dry to wet						
1133	season in Amazonia, Theor Appl Climatol, 78(1-3), 97-110, doi:10.1007/s00704-						
1134	004-0046-7., 2004						
1135	Fu, R., L. Yin, W. Li, and P. A. Arias, Increased dry-season length over southern						
1136	Amazonia in recent decades and its implication for future climate projection, 2013.						
1137	Fu, R., P. A. Arias, and H. Wang, The Connection Between the North and South						
1138	American Monsoons, The Monsoons and Climate Change, 2016						
1139	Garstang, M., and D. R. Fitzjarrald, Observations of surface to atmosphere interactions						
1140	in the tropics., 1999						
1141	Gatti, L.V., Gloor, M., Miller, J.B., Doughty, C.E., Malhi, Y., Domingues, L.G., Basso,						
1142	L.S., Martinewski, A., Correia, C.S.C., Borges, V.F. and Freitas, S., Drought						
1143	sensitivity of Amazonian carbon balance revealed by atmospheric measurements,						
1144	Nature, 506(7486), 76-80, doi:10.1038/nature12957, 2015.						
1145	Gentine, P., A. Garelli, S. B. Park, and J. Nie, Role of surface heat fluxes underneath cold						
1146	pools, Geo Res Letters, 43(2), 874-883, doi:10.1002/2015GL067262, 2016.						





- Gentine, P., A. Holtslag, and F. D'Andrea, Surface and atmospheric controls on the onset
  of moist convection over land, *J Hydrometeorol*, *14*(5), 1443–1462,
  doi:10.1175/JHM-D-12-0137.1, 2013.
- Gentine, P., C. R. Ferguson, and A. A. M. Holtslag, Diagnosing evaporative fraction over
  land from boundary-layer clouds, *J Geophys Res-Atmos*, *118*(15), 8185–8196,
  doi:10.1002/jgrd.50416, 2013.
- Gentine, P., D. Entekhabi, A. Chehbouni, G. Boulet, and B. Duchemin, Analysis of
  evaporative fraction diurnal behaviour, *Agr Forest Meteorol*, *143*, 13–29,
  doi:10.1016/j.agrformet.2006.11.002, 2007.
- 1156 Gerken, T., Ruddell, B.L., Fuentes, J.D., Araújo, A., Brunsell, N.A., Maia, J., Manzi, A., 1157 Mercer, J., dos Santos, R.N., von Randow, C. and Stoy, P.C.,, Investigating the mechanisms responsible for the lack of surface energy balance closure in a central 1158 1159 Amazonian tropical rainforest, Agr Forest Meteorol, 1-0,doi:10.1016/j.agrformet.2017.03.023, 2017. 1160
- Ghate, V. P., and P. Kollias, On the controls of daytime precipitation in the Amazonian
  dry season, *J Hydrometeorol*, JHM–D–16–0101.1–55, doi:10.1175/JHM-D-160101.1, 2016.
- Giangrande, S. E. et al., Cloud Characteristics, Thermodynamic Controls and Radiative
  Impacts During the Observations and Modeling of the Green Ocean Amazon
  (GoAmazon2014/5) Experiment, *Atmos. Chem. Phys. Discuss.*, 1–41,
  doi:10.5194/acp-2017-452, 2017.
- Goulden, M. L., S. D. Miller, and H. R. da Rocha, Diel and seasonal patterns of tropical
  forest CO2 exchange, *Ecological*, 2004
- 1170 Green, J. K., A. G. Konings, S. H. Alemohammad, J. Berry, D. Entekhabi, J. Kolassa, J.-
- 1171 E. Lee, and P. Gentine, Regionally strong feedbacks between the atmosphere and 1172 terrestrial biosphere, *Nat Geosci*, 48, 1–12, doi:10.1038/ngeo2957, 2017.





- Greve, P., B. Orlowsky, B. Mueller, J. Sheffield, M. Reichstein, and S. I. Seneviratne,
  Global assessment of trends in wetting and drying over land, *Nat Geosci*, 7(10), 716–
- 1175 721, doi:10.1038/ngeo2247, 2014.
- Guanter, L., Zhang, Y., Jung, M., Joiner, J., Voigt, M., Berry, J.A., Frankenberg, C.,
  Huete, A.R., Zarco-Tejada, P., Lee, J.E. and Moran, Global and time-resolved
  monitoring of crop photosynthesis with chlorophyll fluorescence, *PNAS*, *111*(14),
  E1327–E1333, doi:10.1073/pnas.1320008111, 2014.
- Guichard, F., Petch, J.C., Redelsperger, J.L., Bechtold, P., Chaboureau, J.P., Cheinet, S.,
  Grabowski, W., Grenier, H., Jones, C.G., Köhler, M. and Piriou, J.M., Modelling the
  diurnal cycle of deep precipitating convection over land with cloud-resolving models
  and single-column models, *Q J Roy Meteor Soc*, *130*(604), 3139–3172,
  doi:10.1256/qj.03.145, 2004.
- Guillod, B.P., Orlowsky, B., Miralles, D., Teuling, A.J., Blanken, P.D., Buchmann, N.,
  Ciais, P., Ek, M., Findell, K.L., Gentine, P. and Lintner, B.R., Land-surface controls
  on afternoon precipitation diagnosed from observational data: uncertainties and
  confounding factors, *Atmos. Chem. Phys*, *14*(16), 8343–8367, doi:10.5194/acp-148343-2014-supplement, 2014.
- Guillod, B. P., B. Orlowsky, D. G. Miralles, A. J. Teuling, and S. I. Seneviratne,
  Reconciling spatial and temporal soil moisture effects on afternoon rainfall, *Nat Comms*, *6*, 6443, doi:10.1038/ncomms7443, 2015.
- 1193 H. G. Ouwersloot, M. Sikma, C. M. J. Jacobs, J. V.-G. de Arellano, X. Pedruzo-1194 Bagazgoitia, M. Sikma, and C. C. van Heerwaarden, Direct and Diffuse Radiation in 1195 Shallow Cumulus-Vegetation System: Enhanced Decreased the and 1196 Evapotranspiration Regimes, J Hydrometeorol, 18(6), 1731-1748, doi:10.1175/JHM-1197 D-16-0279.1, 2017.
- Hadden, D., and A. Grelle, Changing temperature response of respiration turns boreal
  forest from carbon sink into carbon source, *Agr Forest Meteorol*, 223, 30–38,
  doi:10.1016/j.agrformet.2016.03.020, 2016.





- 1201 Hamada, J.-I., M. D. Yamanaka, S. Mori, Y. I. Tauhid, and T. Sribimawati, Differences
- 1202 of Rainfall Characteristics between Coastal and Interior Areas of Central Western
- 1203 Sumatera, Indonesia, Journal of the Meteorological Society of Japan. Ser. II, 86(5),
- 1204 593–611, doi:10.2151/jmsj.86.593, 2008.
- Han, X., H.-J. H. Franssen, C. Montzka, and H. Vereecken, Soil moisture and soil
  properties estimation in the Community Land Model with synthetic brightness
  temperature observations, *Water resources Research*, n/a–n/a,
  doi:10.1002/2013WR014586, 2014.
- Haverd, V., M. Cuntz, L. P. Nieradzik, and I. N. Harman, Improved representations of
  coupled soil-canopy processes in the CABLE land surface model, *Geosci. Model Dev. Discuss.*, 1–24, doi:10.5194/gmd-2016-37, 2016.
- Grant, L. D., & van den Heever, S. C., Cold pool dissipation. *Journal of Geophysical Research: Atmospheres*, *121*(3), 1138-1155, 2016.
- Heroult, A., Y. LIN, and A. Bourne, Optimal stomatal conductance in relation to
  photosynthesis in climatically contrasting Eucalyptus species under drought, *Plant Cell and Env*, doi:10.1111/j.1365-3040.2012.02570.x, 2013.
- Hidayat, R., and S. Kizu, Influence of the Madden-Julian Oscillation on Indonesian
  rainfall variability in austral summer, *Int J Climatol*, 23(12), n/a–n/a,
  doi:10.1002/joc.2005, 2009.
- Hilker, T., A. I. Lyapustin, C. J. Tucker, F. G. Hall, R. B. Myneni, Y. Wang, J. Bi, Y.
  Mendes de Moura, and P. J. Sellers, Vegetation dynamics and rainfall sensitivity of
  the Amazon, *Proceedings of the National Academy of Sciences*, *111*(45), 16041–
  16046, doi:10.1073/pnas.1404870111, 2014.
- Hohenegger, C., and C. S. Bretherton, Simulating deep convection with a shallow
  convection scheme, *Atmos. Chem. Phys*, *11*(20), 10389–10406, doi:10.5194/acp-1110389-2011, 2011.





- Horn, G. L., H. G. Ouwersloot, J. V.-G. de Arellano, and M. Sikma, Cloud Shading
  Effects on Characteristic Boundary-Layer Length Scales, *Bound-Lay Meteorol*, *157*(2), 237–263, doi:10.1007/s10546-015-0054-4, 2015.
- Huang, R., and F. Sun, Impacts of the Tropical Western Pacific on the East Asian
  Summer Monsoon, *Journal of the Meteorological Society of Japan. Ser. II*, 70(1B),
  243–256, doi:10.2151/jmsj1965.70.1B 243, 1992.
- 1233 Huete, A. R., K. Didan, Y. E. Shimabukuro, P. Ratana, S. R. Saleska, L. R. Hutyra, W. 1234 Yang, R. R. Nemani, and R. Myneni (2006), Amazon rainforests green-up with 1235 sunlight season, Geophys Res Lett. 33(6), L06405, in dry 1236 doi:10.1029/2005GL025583, 2006.
- Huffman, G. J., R. F. Adler, M. M. Morrissey, D. T. Bolvin, S. Curtis, R. Joyce, B.
  McGavock, and J. Susskind, Global precipitation at one-degree daily resolution from
  multisatellite observations, *J Hydrometeorol*, 2(1), 36–50, doi:10.1175/15257541(2001)002<0036:GPAODD>2.0.CO;2, 2001.
- Huijnen, V., M. J. Wooster, J. W. Kaiser, D. L. A. Gaveau, J. Flemming, M. Parrington,
  A. Inness, D. Murdiyarso, B. Main, and M. van Weele, Fire carbon emissions over
  maritime southeast Asia in 2015 largest since 1997, *Sci. Rep.*, 1–8,
  doi:10.1038/srep26886, 2016.
- Jardine, K., Chambers, J., Alves, E.G., Teixeira, A., Garcia, S., Holm, J., Higuchi, N.,
  Manzi, A., Abrell, L., Fuentes, J.D. and Nielsen, L.K., Dynamic Balancing of
  Isoprene Carbon Sources Reflects Photosynthetic and Photorespiratory Responses to
  Temperature Stress, *Plant Physiol*, *166*(4), 2051–2064, doi:10.1104/pp.114.247494,
  2014.
- Jimenez, C., Prigent, C., Mueller, B., Seneviratne, S.I., McCabe, M.F., Wood, E.F.,
  Rossow, W.B., Balsamo, G., Betts, A.K., Dirmeyer, P.A. and Fisher, J.B., Global
  intercomparison of 12 land surface heat flux estimates, *J Geophys Res*, *116*(D2),
  1147–27, doi:10.1029/2010JD014545, 2011.





- Joiner, J., A. P. Vasilkov, Y. Yoshida, L. A. Corp, and E. M. Middleton, First
  observations of global and seasonal terrestrial chlorophyll fluorescence from space, *Biogeosciences*, 8(3), 637–651, doi:10.5194/bg-8-637-2011, 2011.
- Joiner, J., L. Guanter, R. Lindstrot, M. Voigt, A. P. Vasilkov, E. M. Middleton, K. F.
  Huemmrich, Y. Yoshida, and C. Frankenberg, Global monitoring of terrestrial
  chlorophyll fluorescence from moderate-spectral-resolution near-infrared satellite
  measurements: methodology, simulations, and application to GOME-2, *Atmos Meas Tech*, 6(10), 2803–2823, doi:10.5194/amt-6-2803-2013, 2013.
- Juárez, R. I. N., M. G. Hodnett, R. Fu, M. L. Goulden, and C. von Randow, Control of
  Dry Season Evapotranspiration over the Amazonian Forest as Inferred from
  Observations at a Southern Amazon Forest Site, *J Climate*, 20(12), 2827–2839,
  doi:10.1175/JCLI4184.1, 2007.
- Jung, M., Reichstein, M., Schwalm, C.R., Huntingford, C., Sitch, S., Ahlström, A.,
  Arneth, A., Camps-Valls, G., Ciais, P., Friedlingstein, P. and Gans, F., Compensatory
  water effects link yearly global land CO2 sink changes to temperature, *541*(7638),
  516–520, doi:10.1038/nature20780, 2017.
- Kato, S., N. G. Loeb, F. G. Rose, D. R. Doelling, D. A. Rutan, T. E. Caldwell, L. Yu, and
  R. A. Weller, Surface Irradiances Consistent with CERES-Derived Top-ofAtmosphere Shortwave and Longwave Irradiances, *J Climate*, *26*(9), 2719–2740,
  doi:10.1175/JCLI-D-12-00436.1, 2013.
- Keller, M., Alencar, A., Asner, G.P., Braswell, B., Bustamante, M., Davidson, E.,
  Feldpausch, T., Fernandes, E., Goulden, M., Kabat, P. and Kruijt, B., Ecological
  Research in the Large-Scale Biosphere– Atmosphere Experiment in Amazonia: Early
- 1277 Results, *Ecol Appl*, *14*(sp4), 3–16, 2004.
- Kennedy, D., P. Gentine, R. A. Fisher, and D. M. Lawrence, Implementation of planthydraulics in the Community Land Model, *JAMES*, 2017.
- 1280 Khairoutdinov, M., and D. Randall, High-resolution simulation of shallow-to-deep
  1281 convection transition over land, *J Atmos Sci*, 63(12), 3421–3436, 2006.





- 1282 Khanna, J., D. Medvigy, S. Fueglistaler, and R. Walko, Regional dry-season climate1283 changes due to three decades of Amazonian deforestation, *Nature Climate Change*,
- 1284 7(3), 200–204, doi:10.1038/nclimate3226, 2017.
- Knox, R. G., M. Longo, A. L. S. Swann, K. Zhang, N. M. Levine, P. R. Moorcroft, and
  R. L. Bras, Hydrometeorological effects of historical land-conversion in an
  ecosystem-atmosphere model of Northern South America, *Hydrol Earth Syst Sc*, *19*(1), 241–273, doi:10.5194/hess-19-241-2015, 2015.
- Konings, A. G., and P. Gentine, Global variations in ecosystem-scale isohydricity, *Global Change Biol*, doi:10.1111/gcb.13389, 2016.
- Koren, I., Y. J. Kaufman, L. A. Remer, and J. V. Martins, Measurement of the effect of
  Amazon smoke on inhibition of cloud formation, *Science*, *303*(5662), 1342–1345,
  doi:10.1126/science.1089424, 2004.
- Koster, R.D., Dirmeyer, P.A., Guo, Z., Bonan, G., Chan, E., Cox, P., Gordon, C.T.,
  Kanae, S., Kowalczyk, E., Lawrence, D. and Liu, P., Regions of strong coupling
  between soil moisture and precipitation, *Science*, *305*(5687), 1138–1140, 2004.
- Koster, R.D., Dirmeyer, P.A., Guo, Z., Bonan, G., Chan, E., Cox, P., Gordon, C.T.,
  Kanae, S., Kowalczyk, E., Lawrence, D. and Liu, P., The Second Phase of the Global
  Land–Atmosphere Coupling Experiment: Soil Moisture Contributions to Subseasonal
  Forecast Skill, *J Hydrometeorol*, *12*(5), 805–822, doi:10.1175/2011JHM1365.1,
  2011.
- Köppen, W., *The thermal zones of the Earth according to the duration of hot, moderate and cold periods and of the impact of heat on the organic world.(translated and ...,*Meteorologische Zeitschrift, 1884.
- Krakauer, N. Y., M. J. Puma, B. I. Cook, P. Gentine, and L. Nazarenko, Oceanatmosphere interactions modulate irrigation's climate impacts, 1–16,
  doi:10.5194/esd-2016-23, 2016.





1308	Kuang, Z., Linear response functions of a cumulus ensemble to temperature and moisture						
1309	perturbations and implications for the dynamics of convectively coupled waves, $J$						
1310	Atmos Sci, 2010.						
1011							
1311	Kuang, Z., A Moisture-Stratiform Instability for Convectively Coupled Waves, Journal						
1312	of Atmospheric Sciences, 65(3), 834–854, doi:10.1175/2007JAS2444.1, 2008.						
1313	Laan-Luijkx, I.T., Velde, I.R., Krol, M.C., Gatti, L.V., Domingues, L.G., Correia, C.S.C.,						
1314	Miller, J.B., Gloor, M., Leeuwen, T.T., Kaiser, J.W. and Wiedinmver, C., Response						
1315	of the Amazon carbon balance to the 2010 drought derived with CarbonTracker						
1316	South America Global Biogeochem Cv 20(7) 1002 1108						
1317	doi:10.1002/2014GB005082, 2015.						
1318	Laurent, H., L. Machado, and C. A. Morales, Characteristics of the Amazonian mesoscale						
1319	convective systems observed from satellite and radar during the WETAMC/LBA						
1320	experiment, J of Climate, 2002.						
1321	Lawrence D M K W Oleson and M G Flanner Parameterization improvements and						
1322	functional and structural advances in Vargian 4 of the Community Land Madel						
1322	Incuoral and structural advances in version 4 of the Community Land Model,						
1525							
1324	Lawrence, D., and K. Vandecar, Effects of tropical deforestation on climate and						
1325	agriculture, Nature Climate Change, 5(1), 27–36, doi:10.1038/nclimate2430, 2015.						
1326	Lawton, R. O., U. S. Nair, R. A. Pielke Sr. and R. M. Welch, Climatic Impact of Tropical						
1327	Lowland Deforestation on Nearby Montane Cloud Forests Science 204 584–587						
1328	2001.						
1329	Lebel, T., Cappelaere, B., Galle, S., Hanan, N., Kergoat, L., Levis, S., Vieux, B.,						
1330	Descroix, L., Gosset, M., Mougin, E. and Peugeot, C., AMMA-CATCH studies in						
1331	the Sahelian region of West-Africa: An overview, Journal of Hydrology, 375(1-2),						
1332	3–13, doi:10.1016/j.jhydrol.2009.03.020, 2009.						
1333	Lee, J. E., R. S. Oliveira, T. E. Dawson, and I. Fung, Root functioning modifies seasonal						
1334	climate, P Natl Acad Sci Usa, 102(49), 17576–17581, 2005.						





- Lee, J.-E., J. A. Berry, C. van der Tol, X. Yang, L. Guanter, A. Damm, I. Baker, and C.
  Frankenberg, Simulations of chlorophyll fluorescence incorporated into the
  Community Land Model version 4, *Global Change Biol*, n/a–n/a,
  doi:10.1111/gcb.12948, 2015.
- Lemordant, L., Modification of land-atmosphere interactions by CO<sub>2</sub> effects:
  Implications for summer dryness and heat wave amplitude, *Geo Res Letters*, 43(19),
  10–240–10–248, doi:10.1002/2016GL069896, 2016.
- Leuning, R., A critical appraisal of a combined stomatal-photosynthesis model for C3
  plants, *Plant Cell and Env*, 1995.
- Leuning, R., F. M. Kelliher, D. G. G. PURY, and E. D. Schulze, Leaf nitrogen,
  photosynthesis, conductance and transpiration: scaling from leaves to canopies, *Plant Cell Environ*, 18(10), 1183–1200, doi:10.1111/j.1365-3040.1995.tb00628.x, 1995.
- Levy, M. C., A. Cohn, A. V. Lopes, and S. E. Thompson, Addressing rainfall data
  selection uncertainty using connections between rainfall and streamflow, *Sci. Rep.*,
  7(1), 1–12, doi:10.1038/s41598-017-00128-5, 2017.
- Li, W., R. Fu, and R. E. Dickinson, Rainfall and its seasonality over the Amazon in the
  21st century as assessed by the coupled models for the IPCC AR4, *Journal of Geophysical Research*, 2006.
- Liebmann, B., and J. A. Marengo, Interannual variability of the rainy season and rainfall
  in the Brazilian Amazon basin, *J Climate*, *14*(22), 4308–4318, 2001.
- Lintner, B. R., and J. Chiang, Reorganization of tropical climate during El Nino: A weak
  temperature gradient approach, *J Climate*, *18*(24), 5312–5329, 2005.
- Lintner, B. R., and J. D. Neelin, A prototype for convective margin shifts, *Geophys Res Lett*, 34(5), L05812, doi:10.1029/2006GL027305, 2007.
- Lintner, B. R., and J. D. Neelin, Soil Moisture Impacts on Convective Margins, J *Hydrometeorol*, 10(4), 1026–1039, doi:10.1175/2009JHM1094.1, 2009.





- Lintner, B. R., P. Gentine, K. L. Findell, and G. D. Salvucci, The Budyko andcomplementary relationships in an idealized model of large-scale land-atmosphere
- 1363 coupling, Hydrol Earth Syst Sc, 19(5), 2119–2131, doi:10.5194/hess-19-2119-2015,
- 1364 2015.
- Lintner, B. R., P. Gentine, K. L. Findell, F. D'Andrea, A. H. Sobel, and G. D. Salvucci,
  An idealized prototype for large-scale land-atmosphere coupling, *J Climate*, 26(7),
  2379–2389, doi:10.1175/JCLI-D-11-00561.1, 2013.
- Liu, L., L. Guan, and X. Liu, Directly estimating diurnal changes in GPP for C3 and C4
  crops using far-red sun-induced chlorophyll fluorescence, *Agr Forest Meteorol*, 232,
  1–9, doi:10.1016/j.agrformet.2016.06.014, 2017.
- Lohou, F., F. Said, M. Lothon, P. Durand, and D. Serça, Impact of Boundary-Layer
  Processes on Near-Surface Turbulence Within the West African Monsoon, *Bound-Lay Meteorol*, *136*(1), 1–23, doi:10.1007/s10546-010-9493-0, 2010.
- Lopes, A. P., B. W. Nelson, J. Wu, P. M. L. de Alencastro Graça, J. V. Tavares, N.
  Prohaska, G. A. Martins, and S. R. Saleska, Leaf flush drives dry season green-up of
  the Central Amazon, *Remote Sensing of Environment*, 182(C), 90–98,
  doi:10.1016/j.rse.2016.05.009, 2016.
- Machado, L.A., Silva Dias, M.A., Morales, C., Fisch, G., Vila, D., Albrecht, R.,
  Goodman, S.J., Calheiros, A.J., Biscaro, T., Kummerow, C. and Cohen, J.. The
  CHUVA project: How does convection vary across Brazil?. *Bulletin of the American Meteorological Society*, *95*(9), 1365-1380, 2014.
- Machado, L., and H. Laurent, Diurnal march of the convection observed during TRMMWETAMC/LBA, *J Geo Res: Atmo*, 2002.
- Mahecha, M. D. et al., Global convergence in the temperature sensitivity of respiration at
  ecosystem level, *Science*, *329*(5993), 838–840, doi:10.1126/science.1189587, 2010.





- Marengo, J. A., J. Tomasella, L. M. Alves, W. R. Soares, and D. A. Rodriguez, The
  drought of 2010 in the context of historical droughts in the Amazon region, *Geophys*
- 1388 Res Lett, 38(12), n/a–n/a, doi:10.1029/2011GL047436, 2011.
- Martin, S.T., Artaxo, P., Machado, L.A.T., Manzi, A.O., Souza, R.A.F., Schumacher, C.,
  Wang, J., Andreae, M.O., Barbosa, H.M.J., Fan, J. and Fisch, G., Introduction:
  Observations and Modeling of the Green Ocean Amazon (GoAmazon2014/5), *Atmos. Chem. Phys*, 16(8), 4785–4797, doi:10.5194/acp-16-4785-2016, 2016.
- Martin, S. T. et al., The Green Ocean Amazon Experiment (GoAmazon2014/5) Observes
  Pollution Affecting Gases, Aerosols, Clouds, and Rainfall over the Rain Forest, *Bull. Amer. Meteor. Soc.*, 98(5), 981–997, doi:10.1175/BAMS-D-15-00221.1, 2017.
- Martinez-Vilalta, J., and N. Garcia-Forner, Water potential regulation, stomatal
  behaviour and hydraulic transport under drought: deconstructing the iso/anisohydric
  concept, *Plant Cell Environ*, 40(6), 962–976, doi:10.1111/pce.12846, 2016.
- Martinez-Vilalta, J., R. Poyatos, D. Aguadé, J. Retana, and M. Mencuccini, A new look
  at water transport regulation in plants, *New Phytologist*, doi:10.1111/nph.12912,
  2014.
- Medlyn, B. E., R. A. Duursma, D. Eamus, D. S. Ellsworth, I. C. Prentice, C. V. M.
  Barton, K. Y. Crous, P. De Angelis, M. Freeman, and L. Wingate, Reconciling the
  optimal and empirical approaches to modelling stomatal conductance, *Global Change Biol*, *17*(6), 2134–2144, doi:10.1111/j.1365-2486.2010.02375.x, 2011.
- Medlyn, B. E., R. A. Duursma, D. Eamus, D. S. Ellsworth, I. Colin Prentice, C. V. M.
  Barton, K. Y. Crous, P. Angelis, M. Freeman, and L. Wingate, Reconciling the
  optimal and empirical approaches to modelling stomatal conductance, *Global Change Biol*, 18(11), 3476–3476, doi:10.1111/j.1365-2486.2012.02790.x, 2012.
- Medvigy, D., R. L. Walko, and R. Avissar, Effects of Deforestation on Spatiotemporal
  Distributions of Precipitation in South America, *J Climate*, *24*, 2147–2163, 2011.





- 1412 Miralles, D. G., J. H. Gash, T. R. H. Holmes, R. A. M. de Jeu, and A. J. Dolman, Global
- 1413 canopy interception from satellite observations, J Geophys Res, 115(D16), 237,
  1414 doi:10.1029/2009JD013530, 2010.
- Morton, D. C., and B. D. Cook, Amazon forest structure generates diurnal and seasonal
  variability in light utilization, *Biogeosciences*, *13*(7), 2195–2206, doi:10.5194/bg-132195-2016, 2016.
- Morton, D. C., J. Nagol, C. C. Carabajal, J. Rosette, M. Palace, B. D. Cook, E. F.
  Vermote, D. J. Harding, and P. R. J. North, Amazon forests maintain consistent
  canopy structure and greenness during the dry season, *Nature*, *506*(7487), 221–224,
  doi:10.1038/nature13006, 2014.
- Mueller, B., Seneviratne, S.I., Jimenez, C., Corti, T., Hirschi, M., Balsamo, G., Ciais, P.,
  Dirmeyer, P., Fisher, J.B., Guo, Z. and Jung, M., Evaluation of global observationsbased evapotranspiration datasets and IPCC AR4 simulations, *Geophys Res Lett*, 38,
  -, doi:10.1029/2010GL046230, 2011.
- Munger, J. W., J. B. McManus, D. D. Nelson, M. S. Zahniser, E. A. Davidson, S. C.
  Wofsy, R. Wehr, and S. R. Saleska, Seasonality of temperate forest photosynthesis and daytime respiration, *Nature*, *534*(7609), 680–683, doi:10.1038/nature17966, 2016.
- 1430 Nakicenovic, N., IPCC Special Report on Emissions Scenarios, 2000
- Naudts, K., Ryder, J., McGrath, M.J., Otto, J., Chen, Y., Valade, A., Bellasen, V.,
  Berhongaray, G., Bönisch, G., Campioli, M. and Ghattas, J., A vertically discretised
  canopy description for ORCHIDEE (SVN r2290) and the modifications to the
  energy, water and carbon fluxes, *Geosci. Model Dev. Discuss.*, 7(6), 8565–8647,
  doi:10.5194/gmdd-7-8565-2014, 2014.
- 1436 Ngo-Duc, T., K. Laval, G. Ramillien, J. Polcher, and A. Cazenave, Validation of the land
  1437 water storage simulated by Organising Carbon and Hydrology in Dynamic
  1438 Ecosystems (ORCHIDEE) with Gravity Recovery and Climate Experiment





- 1439 (GRACE) data, *Water resources Research*, 43(4), -, doi:10.1029/2006WR004941,
  1440 2007.
- 1441 Nicholson, S., Land surface processes and Sahel climate, *Rev Geophys*, 38(1), 117–139,
  1442 2000.
- 1443 Nitta, T., Convective Activities in the Tropical Western Pacific and Their Impact on the
  1444 Northern Hemisphere Summer Circulation, *Journal of the Meteorological Society of*1445 *Japan. Ser. II*, 65(3), 373–390, doi:10.2151/jmsj1965.65.3\_373, 1987.
- Naudts, K., Ryder, J., McGrath, M.J., Otto, J., Chen, Y., Valade, A., Bellasen, V.,
  Berhongaray, G., Bönisch, G., Campioli, M. and Ghattas, J., The community Noah
  land surface model with multiparameterization options (Noah-MP): 1. Model
  description and evaluation with local-scale measurements, *J Geophys Res-Atmos*, *116*, -, doi:10.1029/2010JD015139, 2011.
- Noilhan, J., and S. Planton, A simple parameterization of land surface processes formeteorological models, *Mon Wea Rev*, 1989.
- Oleson, K.W., Niu, G.Y., Yang, Z.L., Lawrence, D.M., Thornton, P.E., Lawrence, P.J.,
  Stöckli, R., Dickinson, R.E., Bonan, G.B., Levis, S. and Dai, A., Improvements to the
  Community Land Model and their impact on the hydrological cycle, *J Geophys Res- Biogeo*, 113, -, doi:10.1029/2007JG000563, 2008.
- Oliveira, R. S., T. E. Dawson, S. S. O. Burgess, and D. C. Nepstad, Hydraulic
  redistribution in three Amazonian trees, *Oecologia*, 145(3), 354–363,
  doi:10.1007/s00442-005-0108-2, 2005.
- Ouwersloot, H. G., J. V.-G. de Arellano, B. J. H. van Stratum, M. C. Krol, and J.
  Lelieveld, Quantifying the transport of sub-cloud layer reactants by shallow cumulus
  clouds over the Amazon, *J Geophys Res-Atmos*, n/a–n/a,
  doi:10.1002/2013JD020431, 2013.
- Park, S., and C. S. Bretherton, The University of Washington Shallow Convection andMoist Turbulence Schemes and Their Impact on Climate Simulations with the





- 1466 Community Atmosphere Model, J Climate, 22(12), 3449–3469,
- 1467 doi:10.1175/2008JCLI2557.1, 2009.
- Park, S.-B., S. Boeing, and P. Gentine, Role of shear on shallow convection (in press), J *Atmos Science*, 2018.
- Peterhansel, C., and V. G. Maurino, Photorespiration Redesigned, *Plant Physiol*, 155(1),
  49–55, doi:10.1104/pp.110.165019, 2011.
- Phillips, N. G., R. Oren, J. Licata, and S. Linder, Time series diagnosis of tree hydraulic
  characteristics, *Tree Physiol*, 879–890, 2004.
- Phillips, N. G., T. N. Buckley, and D. T. Tissue, Capacity of Old Trees to Respond to
  Environmental Change, *Journal of Integrative Plant Biology*, *50*(11), 1355–1364,
  doi:10.1111/j.1744-7909.2008.00746.x, 2008.
- Phillips, N., A. Nagchaudhuri, R. Oren, and G. Katul, Time constant for water transport
  in loblolly pine trees estimated from time series of evaporative demand and stem
  sapflow, *Trees*, 11(7), 412–419, 1997.
- Pielke Sr, R.A., Pitman, A., Niyogi, D., Mahmood, R., McAlpine, C., Hossain, F.,
  Goldewijk, K.K., Nair, U., Betts, R., Fall, S. and Reichstein, M., Land use/land
  cover changes and climate: modeling analysis and observational evidence, *Wires Clim Change*, 2(6), 828–850, doi:10.1002/wcc.144, 2011.
- Pielke, R. A., Sr, R. Mahmood, and C. McAlpine, Land's complex role in climate
  change, *Physics Today*, 69(11), 40–46, doi:10.1063/PT.3.3364, 2016.
- Pielke, R., and R. Avissar, Influence of landscape structure on local and regional climate, *Landscape Ecology*, 4(2/3), 133–155, 1990.
- Pielke, R., G. DALU, J. SNOOK, T. LEE, and T. KITTEL, Nonlinear Influence of
  Mesoscale Land-Use on Weather and Climate, *J Climate*, *4*(11), 1053–1069, 1991.
- Pons, T. L., and R. Welschen, Midday depression of net photosynthesis in the tropical
  rainforest tree Eperua grandiflora: contributions of stomatal and internal





1492 1493	conductances, respiration and Rubisco functioning, <i>Tree Physiol</i> , 23(14), 937–947, 2003.					
1494	Poulter, B., Frank, D., Ciais, P., Myneni, R.B., Andela, N., Bi, J., Broquet, G., Canadell,					
1495	J.G., Chevallier, F., Liu, Y.Y. and Running, S.W., Contribution of semi-arid					
1496	ecosystems to interannual variability of the global carbon cycle, Nature, 509(7502),					
1497	600-603, doi:10.1038/nature13376, 2014.					
1498	Powell, T.L., Galbraith, D.R., Christoffersen, B.O., Harper, A., Imbuzeiro, H.M.,					
1499	Rowland, L., Almeida, S., Brando, P.M., da Costa, A.C.L., Costa, M.H. and Levine,					
1500	N.M., Confronting model predictions of carbon fluxes with measurements of					
1501	Amazon forests subjected to experimental drought, New Phytologist, 200(2), 350-					
1502	365, doi:10.1111/nph.12390, 2013.					
1503	Pradipta, P., A. Giannini, P. Gentine, and U. Lall, Resolving Contrasting Regional					
1504	Rainfall Responses to El Niño over Tropical Africa, J Climate, 29(4), 1461-1476,					
1505	doi:10.1175/JCLI-D-15-0071.1, 2016.					
1506	Prieto, I., and R. J. Ryel, Internal hydraulic redistribution prevents the loss of root					
1506 1507	Prieto, I., and R. J. Ryel, Internal hydraulic redistribution prevents the loss of root conductivity during drought, <i>Tree Physiol</i> , 34(1), 39–48,					
1506 1507 1508	Prieto, I., and R. J. Ryel, Internal hydraulic redistribution prevents the loss of root conductivity during drought, <i>Tree Physiol</i> , <i>34</i> (1), 39–48, doi:10.1093/treephys/tpt115, 2014.					
1506 1507 1508 1509	<ul> <li>Prieto, I., and R. J. Ryel, Internal hydraulic redistribution prevents the loss of root conductivity during drought, <i>Tree Physiol</i>, 34(1), 39–48, doi:10.1093/treephys/tpt115, 2014.</li> <li>Ray, D. K., U. S. Nair, R. O. Lawton, R. M. Welch, and R. A. Pielke Sr. Impact of land</li> </ul>					
1506 1507 1508 1509 1510	<ul> <li>Prieto, I., and R. J. Ryel, Internal hydraulic redistribution prevents the loss of root conductivity during drought, <i>Tree Physiol</i>, 34(1), 39–48, doi:10.1093/treephys/tpt115, 2014.</li> <li>Ray, D. K., U. S. Nair, R. O. Lawton, R. M. Welch, and R. A. Pielke Sr. Impact of land use on Costa Rican tropical montane cloud forests: Sensitivity of orographic cloud</li> </ul>					
1506 1507 1508 1509 1510 1511	<ul> <li>Prieto, I., and R. J. Ryel, Internal hydraulic redistribution prevents the loss of root conductivity during drought, <i>Tree Physiol</i>, 34(1), 39–48, doi:10.1093/treephys/tpt115, 2014.</li> <li>Ray, D. K., U. S. Nair, R. O. Lawton, R. M. Welch, and R. A. Pielke Sr. Impact of land use on Costa Rican tropical montane cloud forests: Sensitivity of orographic cloud formation to deforestation in the plains, <i>J Geophys Res</i>, 111(D2), D02108,</li> </ul>					
1506 1507 1508 1509 1510 1511 1512	<ul> <li>Prieto, I., and R. J. Ryel, Internal hydraulic redistribution prevents the loss of root conductivity during drought, <i>Tree Physiol</i>, 34(1), 39–48, doi:10.1093/treephys/tpt115, 2014.</li> <li>Ray, D. K., U. S. Nair, R. O. Lawton, R. M. Welch, and R. A. Pielke Sr. Impact of land use on Costa Rican tropical montane cloud forests: Sensitivity of orographic cloud formation to deforestation in the plains, <i>J Geophys Res</i>, 111(D2), D02108, doi:10.1029/2005JD006096, 2006.</li> </ul>					
1506 1507 1508 1509 1510 1511 1512 1513	<ul> <li>Prieto, I., and R. J. Ryel, Internal hydraulic redistribution prevents the loss of root conductivity during drought, <i>Tree Physiol</i>, 34(1), 39–48, doi:10.1093/treephys/tpt115, 2014.</li> <li>Ray, D. K., U. S. Nair, R. O. Lawton, R. M. Welch, and R. A. Pielke Sr. Impact of land use on Costa Rican tropical montane cloud forests: Sensitivity of orographic cloud formation to deforestation in the plains, <i>J Geophys Res</i>, 111(D2), D02108, doi:10.1029/2005JD006096, 2006.</li> <li>Raymond, D. J., and X. Zeng, Modelling tropical atmospheric convection in the context</li> </ul>					
1506 1507 1508 1509 1510 1511 1512 1513 1514	<ul> <li>Prieto, I., and R. J. Ryel, Internal hydraulic redistribution prevents the loss of root conductivity during drought, <i>Tree Physiol</i>, <i>34</i>(1), 39–48, doi:10.1093/treephys/tpt115, 2014.</li> <li>Ray, D. K., U. S. Nair, R. O. Lawton, R. M. Welch, and R. A. Pielke Sr. Impact of land use on Costa Rican tropical montane cloud forests: Sensitivity of orographic cloud formation to deforestation in the plains, <i>J Geophys Res</i>, <i>111</i>(D2), D02108, doi:10.1029/2005JD006096, 2006.</li> <li>Raymond, D. J., and X. Zeng, Modelling tropical atmospheric convection in the context of the weak temperature gradient approximation, <i>Q J Roy Meteor Soc</i>, <i>131</i>(608),</li> </ul>					
1506 1507 1508 1509 1510 1511 1512 1513 1514 1515	<ul> <li>Prieto, I., and R. J. Ryel, Internal hydraulic redistribution prevents the loss of root conductivity during drought, <i>Tree Physiol</i>, 34(1), 39–48, doi:10.1093/treephys/tpt115, 2014.</li> <li>Ray, D. K., U. S. Nair, R. O. Lawton, R. M. Welch, and R. A. Pielke Sr. Impact of land use on Costa Rican tropical montane cloud forests: Sensitivity of orographic cloud formation to deforestation in the plains, <i>J Geophys Res</i>, 111(D2), D02108, doi:10.1029/2005JD006096, 2006.</li> <li>Raymond, D. J., and X. Zeng, Modelling tropical atmospheric convection in the context of the weak temperature gradient approximation, <i>Q J Roy Meteor Soc</i>, 131(608), 1301–1320, doi:10.1256/qj.03.97, 2005.</li> </ul>					
1506 1507 1508 1509 1510 1511 1512 1513 1514 1515 1516	<ul> <li>Prieto, I., and R. J. Ryel, Internal hydraulic redistribution prevents the loss of root conductivity during drought, <i>Tree Physiol</i>, <i>34</i>(1), 39–48, doi:10.1093/treephys/tpt115, 2014.</li> <li>Ray, D. K., U. S. Nair, R. O. Lawton, R. M. Welch, and R. A. Pielke Sr. Impact of land use on Costa Rican tropical montane cloud forests: Sensitivity of orographic cloud formation to deforestation in the plains, <i>J Geophys Res</i>, <i>111</i>(D2), D02108, doi:10.1029/2005JD006096, 2006.</li> <li>Raymond, D. J., and X. Zeng, Modelling tropical atmospheric convection in the context of the weak temperature gradient approximation, <i>Q J Roy Meteor Soc</i>, <i>131</i>(608), 1301–1320, doi:10.1256/qj.03.97, 2005.</li> <li>Redelsperger, JL., C. D. Thorncroft, A. Diedhiou, T. Lebel, D. J. Parker, and J. Polcher</li> </ul>					
1506 1507 1508 1509 1510 1511 1512 1513 1514 1515 1516 1517	<ul> <li>Prieto, I., and R. J. Ryel, Internal hydraulic redistribution prevents the loss of root conductivity during drought, <i>Tree Physiol</i>, 34(1), 39–48, doi:10.1093/treephys/tpt115, 2014.</li> <li>Ray, D. K., U. S. Nair, R. O. Lawton, R. M. Welch, and R. A. Pielke Sr. Impact of land use on Costa Rican tropical montane cloud forests: Sensitivity of orographic cloud formation to deforestation in the plains, <i>J Geophys Res</i>, 111(D2), D02108, doi:10.1029/2005JD006096, 2006.</li> <li>Raymond, D. J., and X. Zeng, Modelling tropical atmospheric convection in the context of the weak temperature gradient approximation, <i>Q J Roy Meteor Soc</i>, 131(608), 1301–1320, doi:10.1256/qj.03.97, 2005.</li> <li>Redelsperger, JL., C. D. Thorncroft, A. Diedhiou, T. Lebel, D. J. Parker, and J. Polcher African monsoon multidisciplinary analysis - An international research project and</li> </ul>					
1506 1507 1508 1509 1510 1511 1512 1513 1514 1515 1516 1517 1518	<ul> <li>Prieto, I., and R. J. Ryel, Internal hydraulic redistribution prevents the loss of root conductivity during drought, <i>Tree Physiol</i>, 34(1), 39–48, doi:10.1093/treephys/tpt115, 2014.</li> <li>Ray, D. K., U. S. Nair, R. O. Lawton, R. M. Welch, and R. A. Pielke Sr. Impact of land use on Costa Rican tropical montane cloud forests: Sensitivity of orographic cloud formation to deforestation in the plains, <i>J Geophys Res</i>, 111(D2), D02108, doi:10.1029/2005JD006096, 2006.</li> <li>Raymond, D. J., and X. Zeng, Modelling tropical atmospheric convection in the context of the weak temperature gradient approximation, <i>Q J Roy Meteor Soc</i>, 131(608), 1301–1320, doi:10.1256/qj.03.97, 2005.</li> <li>Redelsperger, JL., C. D. Thorncroft, A. Diedhiou, T. Lebel, D. J. Parker, and J. Polcher African monsoon multidisciplinary analysis - An international research project and field campaign, <i>Bull. Amer. Meteor. Soc.</i>, 87(12), 1739–+, doi:10.1175/BAMS-87-</li> </ul>					





1520	Restrepo-Coupe, N., da Rocha, H.R., Hutyra, L.R., da Araujo, A.C., Borma, L.S.,						
1521	Christoffersen, B., Cabral, O.M., de Camargo, P.B., Cardoso, F.L., da Costa, A.C.L.						
1522	and Fitzjarrald, D.R., What drives the seasonality of photosynthesis across the						
1523	Amazon basin? A cross-site analysis of eddy flux tower measurements from the						
1524	Brasil flux network, Agr Forest Meteorol, 182-183, 128-144,						
1525	doi:10.1016/j.agrformet.2013.04.031, 2013.						
1526	Rieck, M., C. Hohenegger, and C. C. van Heerwaarden, The influence of land surface						
1527	heterogeneities on cloud size development, Mon Wea Rev, 140611124045000,						
1528	doi:10.1175/MWR-D-13-00354.1, 2014.						
1529	Rieck, M., C. Hohenegger, and P. Gentine, The effect of moist convection on thermally						
1530	induced mesoscale circulations, Q J Roy Meteor Soc, 141(691), 2418-2428,						
1531	doi:10.1002/qj.2532, 2015.						
1532	Rio, C., F. Hourdin, J. Y. Grandpeix, and J. P. Lafore, Shifting the diurnal cycle of						
1533	parameterized deep convection over land, Geophys Res Lett, 36, -,						
1534	doi:10.1029/2008GL036779, 2009.						
1535	Rochetin, N., F. Couvreux, JY. Grandpeix, and C. Rio, Deep Convection Triggering by						
1536	Boundary Layer Thermals. Part I: LES Analysis and Stochastic Triggering						
1537	Formulation, <i>J Atmos Sci</i> , 71(2), 496–514, doi:10.1175/JAS-D-12-0336.1, 2014.						
1538	Rochetin, N., JY. Grandpeix, C. Rio, and F. Couvreux, Deep Convection Triggering by						
1539	Boundary Layer Thermals. Part II: Stochastic Triggering Parameterization for the						
1540	LMDZ GCM, J Atmos Sci, 71(2), 515–538, doi:10.1175/JAS-D-12-0337.1, 2014.						
1541	Romps, D. M., Numerical tests of the weak pressure gradient approximation, J Atmos Sci,						
1542	69(9), 2846–2856, doi:10.1175/JAS-D-11-0337.1, 2012.						
1543	Romps, D. M., Weak Pressure Gradient Approximation and Its Analytical Solutions, $J$						
1544	Atmos Sci, 69(9), 2835–2845, doi:10.1175/JAS-D-11-0336.1, 2012.						
1545	Roy, S., C. Weaver, D. Nolan, and R. Avissar, A preferred scale for landscape forced						
1546	mesoscale circulations? J Geophys Res-Atmos, 108(D22), 8854,						

1547 doi:10.1029/2002JD003097, 2003.





1548 Saatchi, S., S. Asefi-Najafabady, Y. Malhi, L. E. O. C. Aragao, L. O. Anderson, R. B. 1549 Myneni, and R. Nemani, Persistent effects of a severe drought on Amazonian forest 1550 canopy, PNAS, 110(2), 565-570, doi:10.1073/pnas.1204651110, 2013. 1551 Saleska, S. R., J. Wu, K. Guan, A. C. Araujo, and A. Huete, Dry-season greening of 1552 Amazon forests, Nature, doi:10.1038/nature16457, 2016. 1553 Schaefli, B., R. J. van der Ent, R. Woods, and H. H. G. Savenije, An analytical model for 1554 soil-atmosphere Sc. feedback, Hydrol Earth Syst 16(7), 1863-1878, 1555 doi:10.5194/hess-16-1863-2012, 2012. 1556 Schiro, K. A., J. D. Neelin, D. K. Adams, and B. R. Lintner, Deep Convection and 1557 Column Water Vapor over Tropical Land versus Tropical Ocean: A Comparison 1558 between the Amazon and the Tropical Western Pacific, Journal of Atmospheric Sciences, 73(10), 4043-4063, doi:10.1175/JAS-D-16-0119.1, 2016. 1559 1560 Scholz, F., N. Phillips, and S. Bucci, Hydraulic capacitance: biophysics and functional 1561 significance of internal water sources in relation to tree size, Size- and Age-Related 1562 Changes in Tree Structure, 341–361, 2011. 1563 Scott, R., D. Entekhabi, R. D. Koster, and M. Suarez, Timescales of land surface 1564 evapotranspiration response, J Climate, 10(4), 559-566, 1997. Sellers, P. J., C. J. Tucker, G. J. Collatz, S. O. Los, C. O. Justice, D. A. Dazlich, and D. 1565 1566 A. Randall, A Revised Land Surface Parameterization (SiB2) for Atmospheric GCMS. Part II: The Generation of Global Fields of Terrestrial Biophysical 1567 1568 Parameters from Satellite Data, J Climate, 9(4), 706-737, doi:10.1175/1520-1569 0442(1996)009<0706:ARLSPF>2.0.CO;2, 1996. 1570 Sellers, P., D. A. Randall, G. Collatz, J. Berry, C. Field, D. Dazlich, C. Zhang, G. 1571 Collelo, and L. Bounoua, A Revised Land Surface Parameterization (SiB2) for 1572 Atmospheric GCMS. Part I: Model Formulation, J Climate, 9(4), 676–705, 1996. 1573 Seneviratne, S., Impact of soil moisture-climate feedbacks on CMIP5 projections: First 1574 results from the GLACE-CMIP5 experiment, Geophys Res Lett, 40(19), 5212-5217, 1575 doi:10.1002/grl.50956, 2013.





- 1576 Sentić, S., and S. L. Sessions, Idealized modeling of convective organization with
  1577 Changing Sea surface temperatures using multiple equilibria in weak temperature
  1578 gradient simulations, J. Adv. Model. Earth Syst., 1–65, doi:10.1002/2016MS000873,
- 1579 2017.
- Sobel, A. H., and C. S. Bretherton, Large-scale waves interacting with deep convection in
  idealized mesoscale model simulations, *Tellus Series A-Dynamic Meteorology And Oceanography*, 55(1), 45–60, doi:10.1034/j.1600-0870.2003.201421.x, 2003.
- Sobel, A. H., G. Bellon, and J. Bacmeister, Multiple equilibria in a single-column model
  of the tropical atmosphere, *Geophys Res Lett*, 34(22), L22804,
  doi:10.1029/2007GL031320, 2007.
- Sobel, A. H., J. Nilsson, and L. Polvani, The weak temperature gradient approximation
  and balanced tropical moisture waves, *J Atmos Sci*, 58(23), 3650–3665, 2001.
- Spracklen, D. V., S. R. Arnold, and C. M. Taylor, Observations of increased tropical
  rainfall preceded by air passage over forests, *Nature*, 489(7415), 282–285,
  doi:10.1038/nature11390, 2012.
- 1591 Stevens, B., and S. Bony, What Are Climate Models Missing? *Science*, *340*(6136), 1053–
  1054, doi:10.1126/science.1237554, 2013.
- Stoeckli, R., D. M. Lawrence, G. Y. Niu, K. W. Oleson, P. E. Thornton, Z. L. Yang, G.
  B. Bonan, A. S. Denning, and S. W. Running, Use of FLUXNET in the Community
  Land Model development, *J Geophys Res*, *113*(G1), G01025,
  doi:10.1029/2007JG000562, 2008.
- Sutanto, S. J., J. Wenninger, A. M. J. Coenders-Gerrits, and S. Uhlenbrook, Partitioning
  of evaporation into transpiration, soil evaporation and interception: a comparison
  between isotope measurements and a HYDRUS-1D model, *Hydrol Earth Syst Sc*, *16*(8), 2605–2616, doi:10.5194/hess-16-2605-2012, 2012.
- Swann, A. L. S., M. Longo, R. G. Knox, E. Lee, and P. R. Moorcroft, Future
  deforestation in the Amazon and consequences for South American climate, *Agr Forest Meteorol*, 214-215, 12–24, doi:10.1016/j.agrformet.2015.07.006, 2015.





1604	Tang, S., Xie, S., Zhang, Y., Zhang, M., Schumacher, C., Upton, H., Jensen, M.P.,
1605	Johnson, K.L., Wang, M., Ahlgrimm, M. and Feng, Z., Large-scale vertical velocity,
1606	diabatic heating and drying profiles associated with seasonal and diurnal variations of
1607	convective systems observed in the GoAmazon2014/5 experiment, Atmos. Chem.
1608	Phys, 16(22), 14249-14264, doi:10.5194/acp-16-14249-2016, 2016.
1609	Taylor, C. M., C. E. Birch, D. J. Parker, N. Dixon, F. Guichard, G. Nikulin, and G. M. S.
1610	Lister, New perspectives on land-atmosphere feedbacks from the African Monsoon
1611	Multidisciplinary Analysis, Atmos Sci Lett, 12(1), 38-44, doi:10.1002/asl.336, 2011.
1612	Taylor, C. M., C. E. Birch, D. J. Parker, N. Dixon, F. Guichard, G. Nikulin, and G. M. S.
1613	Lister, Modelling soil moisture - precipitation feedbacks in the Sahel: importance of
1614	spatial scale versus convective parameterization, Geophys Res Lett, 40(23), 6213-
1615	6218, doi:10.1002/2013GL058511, 2013.
1616	Taylor, C. M., D. J. Parker, and P. P. Harris, An observational case study of mesoscale
1617	atmospheric circulations induced by soil moisture, Geophys Res Lett, 34(15),
1618	L15801, doi:10.1029/2007GL030572, 2007.
1619	Taylor, C. M., P. P. Harris, and D. J. Parker, Impact of soil moisture on the development
1620	of a Sahelian mesoscale convective system: a case-study from the AMMA Special
1621	Observing Period, Q J Roy Meteor Soc, 136(S1), 456–470, doi:10.1002/qj.465, 2009.
1622	Thiery, W., E. L. Davin, D. M. Lawrence, A. L. Hirsch, M. Hauser, and S. I. Seneviratne,
1623	Present-day irrigation mitigates heat extremes, J Geophys Res-Atmos, 122(3), 1403-
1624	1422, doi:10.1002/2016JD025740, 2017.
1625	Thornley, J. H. M., Plant growth and respiration re-visited: maintenance respiration
1626	defined - it is an emergent property of, not a separate process within, the system - and
1627	why the respiration : photosynthesis ratio is conservative, Annals of Botany, 108(7),
1628	1365–1380, doi:10.1093/aob/mcr238, 2011.
1629	Thum, T., Zaehle, S., Köhler, P., Aalto, T., Aurela, M., Guanter, L., Kolari, P., Laurila,
1630	T., Lohila, A., Magnani, F. and Tol, C.V.D., Modelling sun-induced fluorescence and





- 1631 photosynthesis with a land surface model at local and regional scales in northern Europe, Biogeosciences, 14(7), 1969–1987, doi:10.5194/bg-14-1969-2017, 2017. 1632 1633 Torri, G., Z. Kuang, and Y. Tian, Mechanisms for convection triggering by cold pools, 1634 Geophys Res Lett, 42(6), 1943-1950, doi:10.1002/2015GL063227, 2015. 1635 van der Ent, R. J., and O. A. Tuinenburg, The residence time of water in the atmosphere revisited, Hydrol Earth Syst Sc, 21(2), 779-790, doi:10.5194/hess-21-779-2017, 1636 1637 2017. 1638 van der Ent, R. J., H. H. G. Savenije, B. Schaefli, and S. C. Steele-Dunne, Origin and fate 1639 of atmospheric moisture over continents, Water resources Research, 46(9), n/a-n/a, doi:10.1029/2010WR009127, 2010. 1640 1641 van Dijk, A.I., Gash, J.H., van Gorsel, E., Blanken, P.D., Cescatti, A., Emmel, C., Gielen, B., Harman, I.N., Kiely, G., Merbold, L. and Montagnani, L., Rainfall interception 1642 1643 and the coupled surface water and energy balance, Agr Forest Meteorol, 214-215, 1644 402-415, doi:10.1016/j.agrformet.2015.09.006, 2015. 1645 Vila-Guerau de Arellano, J., E. G. Patton, T. Karl, K. van den Dries, M. C. Barth, and J. 1646 J. Orlando, The role of boundary layer dynamics on the diurnal evolution of isoprene 1647 and the hydroxyl radical over tropical forests, J Geophys Res-Atmos, 116(D7), 8032, doi:10.1029/2010JD014857, 2011. 1648 1649 Vogel, M. M., R. Orth, F. Cheruy, S. Hagemann, R. Lorenz, B. J. J. M. van den Hurk, and S. I. Seneviratne, Regional amplification of projected changes in extreme 1650 1651 temperatures strongly controlled by soil moisture-temperature feedbacks, Geophys 1652 Res Lett, 44(3), 1511–1519, doi:10.1002/2016GL071235, 2017.
- Wang, J., F. J. F. Chagnon, E. R. Williams, A. K. Betts, N. O. Renno, L. A. T. Machado,
  G. Bisht, R. Knox, and R. L. Bras, Impact of deforestation in the Amazon basin on
  cloud climatology, *PNAS*, *106*(10), 3670–3674, doi:10.1073/pnas.0810156106, 2009.
- Wang, J., R. L. Bras, and E. Eltahir, The impact of observed deforestation on the
  mesoscale distribution of rainfall and clouds in Amazonia, *J Hydrometeorol*, 1(3),
  267–286, 2000.





- Wang, S., A. H. Sobel, and Z. Kuang, Cloud-resolving simulation of TOGA-COARE
  using parameterized large-scale dynamics, *J Geophys Res-Atmos*, 118(12), 6290–
- 1661 6301, doi:10.1002/jgrd.50510, 2013.
- Wang, Y. P., and R. Leuning, A two-leaf model for canopy conductance, photosynthesis
  and partitioning of available energy I:, *Agr Forest Meteorol*, *91*(1-2), 89–111,
  doi:10.1016/S0168-1923(98)00061-6, 1998.
- Washington, R., R. James, H. Pearce, W. M. Pokam, and W. Moufouma-Okia, Congo
  Basin rainfall climatology: can we believe the climate models? *Philos T R Soc B*,
  368(1625), 20120296–20120296, doi:10.1098/rstb.2012.0296, 2013.
- Wehr, R., R. Commane, J. W. Munger, J. B. McManus, D. D. Nelson, M. S. Zahniser, S.
  R. Saleska, and S. C. Wofsy, Dynamics of canopy stomatal conductance,
  transpiration, and evaporation in a temperate deciduous forest, validated by carbonyl
  sulfide uptake,, 1–17, doi:10.5194/bg-2016-365, 2016.
- Werth, D., and R. Avissar, The local and global effects of Amazon deforestation, J *Geophys Res-Atmos*, 107(D20), 8087, doi:10.1029/2001JD000717, 2002.
- Wills, R. C., M. P. Byrne, and T. Schneider, Thermodynamic and dynamic controls on
  changes in the zonally anomalous hydrological cycle, *Geo Res Letters*, 43(9), 4640–
  4649, doi:10.1002/2016GL068418, 2016.
- Wright, J. S., R. Fu, J. R. Worden, S. Chakraborty, N. E. Clinton, C. Risi, Y. Sun, and L.
  Yin, Rainforest-initiated wet season onset over the southern Amazon, *PNAS*, *114*(32), 8481–8486, doi:10.1073/pnas.1621516114, 2017.
- Wu, C.-M., B. Stevens, and A. Arakawa, What Controls the Transition from Shallow to
  Deep Convection? *J Atmos Sci*, 66(6), 1793–1806, doi:10.1175/2008JAS2945.1,
  2009.
- 1683 Xu, X., D. Medvigy, J. S. Powers, J. M. Becknell, and K. Guan, Diversity in plant
  hydraulic traits explains seasonal and inter-annual variations of vegetation dynamics
  1685 in seasonally dry tropical forests, *New Phytologist*, 1–16, doi:10.1111/nph.14009,
  2016.





- Yano, J. I., and R. S. Plant, Convective quasi-equilibrium, *Rev Geophys*, 50(4), RG4004,
   doi:10.1029/2011RG000378, 2012.
- Yin, L., R. Fu, E. Shevliakova, and R. E. Dickinson, How well can CMIP5 simulate
  precipitation and its controlling processes over tropical South America? *Climate dynamics*, 41(11-12), 3127–3143, doi:10.1007/s00382-012-1582-y, 2013.
- Yu, H., L. A. Remer, M. Chin, H. Bian, R. G. Kleidman, and T. Diehl, A satellite-based
  assessment of transpacific transport of pollution aerosol, *J Geophys Res-Atmos*, *113*(D14), doi:10.1029/2007JD009349, 2008.
- Zahn, E., N. L. Dias, A. Araújo, L. D. A. Sá, M. Sörgel, I. Trebs, S. Wolff, and A. Manzi,
  Scalar turbulent behavior in the roughness sublayer of an Amazonian forest, *Atmos. Chem. Phys*, *16*(17), 11349–11366, doi:10.5194/acp-16-11349-2016, 2016.
- Zeng, N., and J. D. Neelin, A land-atmosphere interaction theory for the tropical
  deforestation problem, *J Climate*, *12*(2-3), 857–872, 1999.
- Zeng, N., J. Neelin, K. Lau, and C. Tucker, Enhancement of Interdecadal Climate
  Variability in the Sahel by Vegetation Interaction, *Science*, 286(5444), 1537–1540,
  1999.
- 1703 Zeppel, M. J. B., J. D. Lewis, N. G. Phillips, and D. T. Tissue, Consequences of nocturnal 1704 water loss: a synthesis of regulating factors and implications for capacitance, 1705 embolism and use in models, Tree Physiol, 34(10), 1047-1055, 1706 doi:10.1093/treephys/tpu089, 2014.
- Zhang, K., de Almeida Castanho, A.D., Galbraith, D.R., Moghim, S., Levine, N.M., Bras,
  R.L., Coe, M.T., Costa, M.H., Malhi, Y., Longo, M. and Knox, R.G., The fate of
  Amazonian ecosystems over the coming century arising from changes in climate,
  atmospheric CO 2,and land use, *Global Change Biol*, 21(7), 2569–2587,
  doi:10.1111/gcb.12903, 2015.
- Zhang, Y. J., F. C. Meinzer, and J. Qi, Midday stomatal conductance is more related tostem rather than leaf water status in subtropical deciduous and evergreen broadleaf





1714	trees, Plant Cell and Env, 36(1), 149-158, doi:10.1111/j.1365-3040.2012.02563.x,
1715	2013.

- Zhang, Y., R. Fu, H. Yu, R. E. Dickinson, R. N. Juarez, M. Chin, and H. Wang, A
  regional climate model study of how biomass burning aerosol impacts landatmosphere interactions over the Amazon, *Journal of Geophysical Research: Atmospheres (1984–2012)*, *113*(D14), 1042, doi:10.1029/2007JD009449, 2008.
- Zhang, Y., R. Fu, H. Yu, Y. Qian, R. Dickinson, M. A. F. Silva Dias, P. L. da Silva Dias,
  and K. Fernandes, Impact of biomass burning aerosol on the monsoon circulation
  transition over Amazonia, *Geophys Res Lett*, 36(10), 1509,
  doi:10.1029/2009GL037180, 2009.
- Zhuang, Y., R. Fu, and J. A. Marengo, Seasonal variation of shallow-to-deep convection
  transition and its link to the environmental conditions over the Central Amazon, J *Geo Res: Atmo*, doi:10.1002/(ISSN)2169-8996, 2017.
- 1727

1728





- 1730 Table 1. The surface friction velocity, subcloud layer height (where the minimum of
- 1731 virtual potential temperature flux occurs), ratio of subcloud layer height and Obukhov
- 1732 length, ratio of surface friction velocity and Deardorff convective velocity scale, and the
- total number of identified clouds for 12 time instants in each case.

Case	S3	S2	S1	CTL	R1	R2	R3
<i>u</i> <sub>*</sub> <i>u</i> <sub>*</sub> [m s <sup>-1</sup> ]	0.07	0.14	0.21	0.28	0.35	0.42	0.56
$z_i$ [m]	590	590	590	590	590	610	630
$z_i / L$	392. 1	49.0	14.5	6.1	3.1	1.9	0.8
$\frac{u_*}{w_*}$ $u_*/w_*$	0.10	0.20	0.30	0.40	0.50	0.60	0.79
$N_{ m cloud}$	2248	2229	2283	2302	2250	2703	2776

1747 List of Figures





- 1749 Figure 1: Snapshot of cloud cover over the Amazon basin (courtesy NASA, MODIS 1750 visible bands) in the dry season. Small clouds are shallow convective clouds, highlighting 1751 surface Bowen ratio changes between the river and the forest. At the bottom right, the 1752 deep convective cells, does not follow the surface heterogeneity (and is much larger in 1753 scale).Figure 1: Snapshot of cloud cover over the Amazon basin (courtesy NASA, 1754 MODIS visible bands) in the dry season. Small clouds are shallow convective clouds, 1755 highlighting surface Bowen ratio changes between the river and the forest. At the bottom 1756 right, the deep convective cells, does not follow the surface heterogeneity (and is much 1757 larger in scale). Figure 2: Diurnal cycle in local hour of dry (red) and wet (blue) season observations of 1758 1759 precipitation at K34, near Manaus, along with their standard deviation averaged across 1760 years 2010-2014. 1761 Figure 3: Response of tropically-averaged free tropospheric temperature between 700mb 1762 and 200mb to El Niño Southern Oscillation (choosing the ENSO 3.4 index) 1763 Figure 4: Seasonal variations in Evapotranspiration (ET) from WECANN, Precipitation 1764 (Precip) based on GPCP, Net Radiation (Rn) from CERES and Gross Primary Production 1765 (GPP) based on WECANN informed by Solar-Induced Fluorescence (SIF) over the wet part of the Amazon (top left), the Savanna region of Brazil (top right), over Indonesia 1766 1767 (bottom left) and over the Congo basin (bottom right). Figure 5: Seasonality of Precipitation based on GPCP in the tropics in December-1768 1769 January-February (a), March-April-May (b), June-July-August (c), and September-1770 October-November (SON) and its latitudinal average (e). 1771 Figure 6: same as Figure 5 but for Gross Primary Production (GPP) Figure 7: same as Figure 5 for latent heat flux LE 1772 Figure 8: same as Figure 5 for sensible heat flux H 1773
- 1774 Figure 9: same as Figure 5 for evaporative fraction (EF), the ratio of LE to H+LE.
- 1775 Figure 10: same as Figure 5 for sea-level surface moist static energy flux, the sum of
- 1776 sensible heat flux H and latent heat flux
- 1777 Figure 11: Schematic showing the vertical structure of light and water limitations in a
- 1778 tropical forest.





Figure 12: Climatology of the diurnal cycle of leaf water potential and top soil water
potential in the dry and wet seasons in Caxiuana, Brazil simulated by the Community
Land Model (CLM) with plant hydraulics.

Figure 13: Mesoscale heterogeneity impact on cloud generation. a) Typical perspective
regarding the impact of deforestation and clearings generating deep convective clouds

1784 and b) more realistic impact, in terms of mostly a modification of shallow convection

1785 cloud cover, impacting radiation more than precipitation.

- 1786 *Figure 15*: (a) Schematic of the key elements of the convective margins framework as
- applied along an inflow path across northeastern South America. The solid blue and
  black lines are precipitation and vertically-integrated moisture, while dashed blue line
  corresponds to precipitation smeared out by transients. Adapted from Figure 2 of Lintner
  and Neelin (2009). (b) Rainfall longitudinal transects from the Climate Anomaly
  Monitoring System (CAMS) raingauge-derived precipitation data for SeptemberOctober-November for the period 1950-2000 for El Niño (red), La Niña (blue), and all
  (black) years, averaged over 3.75°S-1.75°S. From Figure 4b of Lintner and Neelin
- 1794 (2007).
- 1795 Figure 16: adapted from Schäfli et al. (2012)

Figure 18: Land-atmosphere feedback strength (change in the variance due to the feedback) between Precipitation and ET (top) and Photosynthetically Active Radiation (PAR) (bottom) based on recent metric developed by Green et al. [2017] using a multivariate Granger causality approach.







1801

Figure 1: Snapshot of cloud cover over the Amazon basin (courtesy NASA, MODIS visible bands) in the dry season. Small clouds are shallow convective clouds, highlighting surface Bowen ratio changes between the river and the forest. At the bottom right, the deep convective cells, does not follow the surface heterogeneity (and is much larger in scale).







1806

Figure 2: Diurnal cycle in local hour of dry (red) and wet (blue) season observations of precipitation at K34, near Manaus, along with their standard deviation averaged across years 2010-2014.







1811

1812Figure 3: Response of tropically-averaged free tropospheric temperature between 700mb and1813200mb to El Niño Southern Oscillation (choosing the ENSO 3.4 index) with either no lag (left) or18142-month lag (middle) or 4-month lag (right)

1815



Figure 4: Seasonal variations in Evapotranspiration (ET) from WECANN, Precipitation (Precip) based on GPCP, Net Radiation (Rn) from CERES and Gross Primary Production (GPP) based on WECANN informed by Solar-Induced Fluorescence (SIF) over the wet part of the Amazon (top left), the Savanna region of Brazil (top right), over Indonesia (bottom left) and over





- 1821 the Congo basin (bottom right).
- 1822
- 1823



Figure 5: Seasonality of Precipitation based on GPCP in the tropics in DecemberJanuary-February (a), March-April-May (b), June-July-August (c), and SeptemberOctober-November (SON) and its latitudinal average (e).







1830 Figure 6: same as Figure 5 but for Gross Primary Production (GPP)







1833

1834 Figure 7: same as Figure 5 for latent heat flux LE

1835







- 1837 Figure 8: same as Figure 5 for sensible heat flux H
- 1838

1839



1841 Figure 9: same as Figure 5 for evaporative fraction (EF), the ratio of LE to H+LE.

- 1842
- 1843
- 1844
- 1845







1847 Figure 10: same as Figure 5 for sea-level surface moist static energy flux, the sum of sensible

- 1848 heat flux H and latent heat flux
- 1849
- 1850
- 1851
- 1852








- 1854 Figure 11: Schematic showing the vertical structure of light and water limitations in a tropical1855 forest.
- 1856
- 1857



Figure 12: Climatology of the diurnal cycle of leaf water potential and top soil water
potential in the dry and wet seasons in Caxiuana, Brazil simulated by the Community
Land Model (CLM) with plant hydraulics.



Figure 13: Mesoscale heterogeneity impact on cloud generation. a) Typical perspective regarding the impact of deforestation and clearings generating deep convective clouds and b) more realistic impact, in terms of mostly a modification of shallow convection cloud cover, impacting radiation more than precipitation.





1867

1868



1869

Figure 14: MODIS visible image of the Northwestern Amazon as the basin transition into the wet season. In the dry season surface heterogeneity whether due to rivers, forestdeforested patches or land-ocean contrast are very clear. In the wet season those sharp gradients disappear as cloud cover mostly dominated by deep convection starts organizing at scales independent from the surface heterogeneity.









Figure 15: (a) Schematic of the key elements of the convective margins framework as applied along an inflow path across northeastern South America. The solid blue and black lines are precipitation and vertically-integrated moisture, while dashed blue line corresponds to precipitation smeared out by transients. Adapted from Figure 2 of Lintner and Neelin (2009). (b) Rainfall longitudinal transects from the Climate Anomaly Monitoring System (CAMS) raingauge-derived precipitation data for September-October-November for the period 1950-2000 for El Niño (red), La Niña (blue), and all (black) years, averaged over 3.75°S-1.75°S. From Figure 4b of Lintner and Neelin (2007).

1884 1885



1887 Figure 16: adapted from Schäfli et al. (2012)







1889 Figure 17: 10 day-backtrajectory analysis over several continental regions of the continental1890 tropics, along with LAI, mean TRMM estimated rainfall, and GLDAS ET estimates.

- 1891
- 1892
- 1893
- 1894



1895

Figure 18: Land-atmosphere feedback strength (change in the variance due to the feedback) between Precipitation and ET (top) and Photosynthetically Active Radiation (PAR) (bottom) based on recent metric developed by Green et al. [2017] using a multivariate Granger causality approach.

- 1900
- 1901
- 1902