Predicting the response of the Amazon rainforest to persistent drought conditions under current and future climates: a major challenge for global land surface models

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Abstract. While a majority of global climate models project drier and longer dry seasons over the Amazon under higher CO2 levels, large uncertainties surround the response of vegetation to persistent droughts in both present-day and future climates. We propose a detailed evaluation of the ability of the ISBAcc (Interaction Soil–Biosphere–Atmosphere Carbon Cycle) land surface model to capture drought effects on both water and carbon budgets, comparing fluxes and stocks at two recent throughfall exclusion (TFE) experiments performed in the Amazon. We also explore the model sensitivity to different water stress functions (WSFs) and to an idealized increase in CO2 concentration and/or temperature. In spite of a reasonable soil moisture simulation, ISBAcc struggles to correctly simulate the vegetation response to TFE whose amplitude and timing is highly sensitive to the WSF. Under higher CO2 concentrations, the increased water-use efficiency (WUE) mitigates the sensitivity of ISBAcc to drought. While one of the proposed WSF formulations improves the response of most ISBAcc fluxes, except respiration, a parameterization of drought-induced tree mortality is missing for an accurate estimate of the vegetation response. Also, a better mechanistic understanding of the forest responses to drought under a warmer climate and higher CO2 concentration is clearly needed.

1 Introduction

The Amazon rainforest biome plays a crucial role in the global climate system regulating the regional energy, water and carbon cycles, and thereby modulating the tropical atmospheric circulation. The forest recycles about 25 to 35 % of the Amazonian precipitation through evapotranspiration (Eltahir and Bras, 1994) and stores about 10 to 15 % of the global above-ground biomass (AGB) (e.g., Potter and Klooster, 1999; Mahli et al., 2006; Beer et al., 2010; Pan et al., 2011).

The vulnerability of the Amazon forest to climate change is of great concern, especially as climate projections based on the fifth phase of the Coupled Model Intercomparison Project (CMIP5) show a between-model consensus towards dryer and longer dry seasons in this region (Fu et al., 2013; Joetzjer et al., 2013). Beyond this model consensus, however, substantial uncertainties in the current assessments given uncertainty in climate feedbacks and climate sensitivity to anthropogenic forcing remain. They arise from many sources including the limited ability of coupled ocean–atmosphere general circulation models (OAGCMs) to capture the present-climate global patterns of temperature and precipitation as well as local vegetation–climate feedbacks (Jupp et al., 2010; Shiohama et al., 2011).
Land surface feedbacks also represent a significant source of uncertainties for climate projections over the Amazon basin (Meir et al., 2006; Friedlingstein et al., 2006; Poulter et al., 2009; Rammig et al., 2010; Galbraith et al., 2010; Booth et al., 2012). This was highlighted by the large spread in the future Amazonian evapotranspiration response to climate change among CMIP5 models (Joetzjer et al., 2013) and the growing evidence that global evapotranspiration has already been perturbed by human activities (Douville et al., 2013). About half of the CMIP5 models are Earth system models (ESMs) that simulate the global carbon cycle and account for direct CO$_2$ effects on plants, such as an increased water-use efficiency (WUE), due to both photosynthesis (i.e., fertilization effect) and stomatal closure responses to increasing atmospheric CO$_2$ concentrations. Given the models’ diversity and limited ability to capture biophysical mechanisms (e.g., Keenan et al., 2013), a process-oriented evaluation of the current-generation land surface models (LSMs) is needed.

The Amazon forest is an ideal setting for evaluating land surface feedbacks in land surface models. The Amazon is projected to experience enhanced dry seasons in most CMIP5 climate scenarios, and possible though uncertain dieback of the Amazon rainforest in some projections (Cox et al., 2000, climate scenarios, and possible though uncertain dieback of projected to experience enhanced dry seasons in most CMIP5 surface feedbacks in land surface models. The Amazon is

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Table 1. ISBA\textsubscript{CC}: notation and main equations for the photosynthesis model.

<table>
<thead>
<tr>
<th>Symbols</th>
<th>Units</th>
<th>Definition</th>
</tr>
</thead>
<tbody>
<tr>
<td>$A_{\text{m}}$</td>
<td>kgCO$_2$ m$^{-2}$s$^{-1}$</td>
<td>photosynthesis rate (light saturated)</td>
</tr>
<tr>
<td>$C_a$</td>
<td>ppmv</td>
<td>atmospheric CO$_2$ concentration</td>
</tr>
<tr>
<td>$C_i$</td>
<td>ppmv</td>
<td>leaf internal CO$_2$ concentration</td>
</tr>
<tr>
<td>$D_a$</td>
<td>g kg$^{-1}$</td>
<td>saturation deficit at the leaf surface</td>
</tr>
<tr>
<td>$D_{\text{max}}$</td>
<td>g kg$^{-1}$</td>
<td>saturation deficit inducing stomatal closure</td>
</tr>
<tr>
<td>$f$</td>
<td>unitless</td>
<td>coupling factor</td>
</tr>
<tr>
<td>$f_0$</td>
<td>unitless</td>
<td>coupling factor at saturating air humidity ($D_a = 0$)</td>
</tr>
<tr>
<td>$f_s$</td>
<td>unitless</td>
<td>coupling factor in well-watered conditions and at saturating air humidity ($D_a = 0$)</td>
</tr>
<tr>
<td>$f_{\text{min}}$</td>
<td>unitless</td>
<td>coupling factor at maximum air humidity deficit ($D_a = D_{\text{max}}$)</td>
</tr>
<tr>
<td>$\Gamma$</td>
<td>ppmv</td>
<td>CO$_2$ concentration compensation point</td>
</tr>
<tr>
<td>$g_m$</td>
<td>mm s$^{-1}$</td>
<td>mesophyll conductance defined as the light saturated rate of photosynthesis (Jacobs, 1994)</td>
</tr>
<tr>
<td>$g_m^*$</td>
<td>mm s$^{-1}$</td>
<td>$g_m$ in well-watered conditions</td>
</tr>
<tr>
<td>$g_s$</td>
<td>mm s$^{-1}$</td>
<td>stomatal conductance</td>
</tr>
</tbody>
</table>

Equations

$$g_m = \frac{A_{\text{m}}}{C_i} ; \text{ at high light intensity and low } C_i \quad \text{(Eq. 1)}$$

$$f = \frac{C_i - \Gamma}{C_i} \quad \text{(Eq. 2)}$$

$$f = f_0 \cdot (1 - \frac{D_a}{D_{\text{max}}}) + f_{\text{min}} \cdot \frac{D_a}{D_{\text{max}}} \quad \text{(Eq. 3)}$$

The ISBA\textsubscript{CC} photosynthesis model relies on the concept of mesophyll conductance ($g_m$), also called internal conductance. As defined by Jacobs (1994), $g_m$ quantifies the slope of the CO$_2$ response curve at high light intensity and low internal CO$_2$ concentration ($C_i$). It can be interpreted as a parameter to model the activity of the Rubisco under these conditions (cf. Table 1, Eq. 1). ISBA\textsubscript{CC} uses a constant unstrressed value of $g_m$ ($g_m^*$) for each vegetation functional type (PFT). ISBA\textsubscript{CC} also defines a ratio $f$ which relates $C_i$ to ambient CO$_2$ ($C_a$) (Table 1, Eq. 2) that decreases linearly with increasing atmospheric humidity deficit (Table 1, Eq. 3). Assimilation is calculated from light, air humidity, $C_a$, the ratio $f$ and stomatal conductance ($g_s$), which measures gas (CO$_2$ and H$_2$O) exchange between the leaves and the atmosphere, is deduced from the assimilation rate. The sensitivity of $g_m$ to the soil water availability is quantified by a WSF, as explained below.

2.1.2 Water stress functions

The WSF is an empirical representation of the effect of soil moisture stress on transpiration and photosynthesis. In ISBA\textsubscript{CC}, soil water content (SWC) affects transpiration and photosynthesis through changes in $g_m$ and/or $f_0$ (Table 1), depending on the PFT and its drought strategy (Table 2). We test the two ISBA\textsubscript{CC} plant strategies (Fig. 1) proposed by Calvet et al. (2004): the drought-avoiding strategy (blue curve) for isohydric plants and the drought-tolerant strategy (purple) of anisohydric plants. One potential model limitation is that these parameterizations were derived from measurements made on saplings of Pinus pinaster and Quercus petraea (Picon et al., 1996), and have not been calibrated for mature trees or tropical species. In addition, we could not find experimental evidence for a direct effect of soil moisture on $C_i$ that would support a function of $f_0 = f$ (soil water index, SWI) (Fig. 1, top right) and ISBA\textsubscript{CC}-simulated photosynthesis and transpiration for tropical rainforests is highly sensitive to $f_0$, because the air is often close to saturation. Therefore, in addition to testing the existing WSF parameterizations, we also tested a linear WSF and the SiB3 (simple biosphere model, version 3) formulation documented in Baker et al. (2008), both applied to $g_m$. These functions assume a constant $f_0$ derived from in situ observations (Table 2, Domingues et al., 2007) and allow for a larger stomatal conductance in line with a higher GPP and a higher evapotranspiration than the existing WSF functions in the model. The linear WSF describes plants that would reduce their stomatal conductance as soon as soil moisture drops below field capacity while the SiB3 WSF describes plants that would wait for drier soils before reducing their stomatal conductance. Despite a fairly similar response of $g_m$ to soil moisture deficit between the linear and the drought-tolerant WSF, and between the SiB3 and drought-avoiding WSF, the linear and SiB3 WSFs induce a stronger response of $g_s$, LE
and GPP to drought (Fig. 1) because \( f_0 \) is not a function of the soil moisture.

### 2.2 Site description and observations

Two rainfall exclusion experiments were initiated at the National Forest Tapajós (2.90° S, 54.96° W) and Caxiuanã National Forest (1.72° S, 51.46° W) in 1999 and 2001, respectively. At each site, the experimental design consists of a 1 ha forest undisturbed control (CTL) and TFE plots in a nearby floristically and structurally similar forest plot. In the TFE plot, a portion of throughfall was excluded using large plastic panels below the canopy, approximately 1–2 m above the ground. A 1 m deep trench was dug around each plot to minimize lateral movement of water and roots. Panels were applied 1 year after the beginning of the experiments to assess pre-treatment plot differences. At Tapajós (Caxiuanã), 1999 (2001) was the baseline year, and the TFE experiment lasted from 2000 to 2004 (2002 and remains ongoing). At Tapajós, panels were removed during the dry season (Fig. 2) to reduce their influence on the forest floor through shading and heating. It was estimated that panels increased forest floor temperature by no more than 0.3 °C (Nepstad et al., 2002). At Caxiuanã, panels were not removed because the risk of dry season storms is relatively high. The air temperature below the TFE panels was no different from ambient during the wet season, and varied up to 2 °C warmer during the dry season; soil temperature differences in TFE remained similar to ambient throughout (Metcalfe et al., 2010).

While soils at both sites are highly weathered Oxisols, they differ greatly in texture. Caxiuanã is a sandy soil and presents a stony laterite layer at 3–4 m depth which could hamper the development of deep roots and soil water movement (Fisher et al., 2007), contrasting with the clay rich soil at Tapajós. Caxiuanã also shows a wetter climate (more precipitation and longer wet season) than Tapajós (Fig. 2); the water table depth reached 10 m at Caxiuanã during the wet season (Fisher et al., 2007), but was below 80 m at Tapajós (Nepstad et al., 2002).

Observations from the TFE experiments used to evaluate ISBACC are summarized in Table 3. As a reference we use evapotranspiration outputs from a 1-D model calibrated and validated at Tapajós from Markewitz et al. (2010, Table 5) and GPP estimated at Caxiuanã by Fisher et al. (2007), because there are no suitable direct measurements of water and carbon fluxes. The footprint of flux towers is from 100 to 1000 times that of the experiments (Chen et al., 2008). Both fine-scale model outputs were carefully and successfully validated by the authors using data sets independent from those used to specify the model structure.

### 2.3 Simulations

At both sites, ISBACC was run offline using in situ hourly meteorological measurements made above the forest canopy at nearby weather stations. At Caxiuanã meteorological measurements were available for the entire experimental period (2001–2008), at Tapajós they covered only the years 2002–
Table 2. Description of ISBA_CCC: water stress functions.

<table>
<thead>
<tr>
<th>Name</th>
<th>Soil Wetness Index</th>
<th>Water Stress functions applied to $g_m$</th>
<th>Water Stress functions applied to $f_0$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Avoiding</td>
<td>SWI ≥ 1</td>
<td>$\ln(g_m^<em>) = 4.7 - 7. f_0^</em>$</td>
<td>$f_0 = f_0^* - (f_0^* - f_0^N) \cdot (1 - SWI) / (1 - SWIC)$</td>
</tr>
<tr>
<td></td>
<td>SWIC &lt; SWI &lt; 1</td>
<td>$g_m = g_m^*$</td>
<td>$f_0 = 2.8 - 10 \ln(g_m)$</td>
</tr>
<tr>
<td></td>
<td>SWI ≤ SWIC</td>
<td>$g_m = g_m^* \cdot \frac{SWIC}{SWI}$</td>
<td>$f_0 = 0.74$</td>
</tr>
<tr>
<td>Tolerant</td>
<td>SWI ≥ 1</td>
<td>$\ln(g_m^<em>) = 4.7 - 7. f_0^</em>$</td>
<td>$f_0 = f_0^* - (f_0^* - f_0^N) \cdot (1 - SWI) / (1 - SWIC)$</td>
</tr>
<tr>
<td></td>
<td>SWIC &lt; SWI &lt; 1</td>
<td>$g_m = g_m^* - \frac{(g_m^* - g_m^N) \cdot (1 - SWIC)}{1 - SWI}$</td>
<td>$f_0 = f_0^* - \ln(g_m^*) / 7$</td>
</tr>
<tr>
<td></td>
<td>SWI ≤ SWIC</td>
<td>$g_m = g_m^* \cdot \frac{SWIC}{SWI}$</td>
<td>$f_0 = 0.74$</td>
</tr>
<tr>
<td>Linear</td>
<td></td>
<td>$g_m = SWI \cdot g_m^*$</td>
<td>$f_0 = 0.74$</td>
</tr>
<tr>
<td>SiB3</td>
<td></td>
<td>$g_m = \frac{(1 + \alpha) \cdot SWI}{\alpha + SWI} \cdot g_m^* \cdot \alpha = 0.1$</td>
<td>$f_0 = 0.74$</td>
</tr>
</tbody>
</table>

**Symbol Description**
- SWI: Soil Wetness index
- $\Theta$: Soil water content (m$^3$ m$^{-3}$)
- $\Theta_{wc}$: Field capacity (m$^3$ m$^{-3}$)
- $\Theta_{wilt}$: Wilting point (m$^3$ m$^{-3}$)
- SWIC: Critical extractable Soil Wetness Index (drought-avoiding and tolerant)
- $f_0^N$: Minimum value of $f_0$ at SWI = SWIC (drought-avoiding)
- $g_m^N$: Value of $g_m$ at SWI = SWIC in mm s$^{-1}$ (drought-tolerant)
- SWI = $\frac{\Theta_{wilt} - \Theta_{wc}}{\Theta_{wilt} - \Theta}$

Figure 2. Monthly observed precipitation at Tapajós and Caxiuanã. Horizontal black bars indicate when panels were applied for the throughfall exclusion (TFE) experiments.

To cover the entire period of experimentation, we cycled sequentially the available years. ISBA_CCC was run until the slowest soil storage pools of water and carbon had reached equilibrium.

ISBA_CCC explicitly simulates interception of precipitation by the canopy and throughfall as runoff from the leaves. To simulate the experimental treatments at each site, we removed 60% of the throughfall in our model runs. This is consistent with Markewitz et al. (2010) and Sakaguchi et al. (2011) for Tapajós, and similar to the 50% exclusion of incident (above-canopy) rainfall implemented at Caxiuanã (Fisher et al., 2007; Galbraith et al., 2010; Powell et al., 2013). The 60% reduction of throughfall was applied to the entire period at Caxiuanã (2001–2008) and only during the rainy seasons (January to June) from 2000 to 2004 at Tapajós, to mimic the experimental conditions.

At both sites, we imposed the evergreen tropical tree plant functional type. To better represent soil moisture and focus on vegetation response, we constrained ISBA_CCC using the observed texture at each site. The soil texture values used for the simulations are at Caxiuanã 75% sand and 15% clay (Ruivo and Cunha, 2003) and 52% sand and 42% clay at Tapajós following the LBA–Data Model Intercomparison Project (www.climatemodeling.org/lba-mip). To mimic deep Amazonian soils, soil and root depth were fixed at 8 m, even at Caxiuanã, because roots there were found below the laterite layer located at 3–4 m deep (Fisher et al., 2007). Representation of deep soil and roots may avoid the simulation of
Table 3. References and available period for observations used in this study.

<table>
<thead>
<tr>
<th>Variables</th>
<th>Symbol</th>
<th>Units</th>
<th>Tapajós</th>
<th>Caxiuana</th>
</tr>
</thead>
<tbody>
<tr>
<td>Stomatal conductance</td>
<td>gs</td>
<td>mmol CO$_2$ m$^{-2}$ s$^{-1}$</td>
<td>2002–2003 Fisher et al. (2006)</td>
<td></td>
</tr>
<tr>
<td>Gross primary production</td>
<td>GPP</td>
<td>µmol CO$_2$ m$^{-2}$ s$^{-1}$</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

unrealistic responses to drought due to a drying of the upper layers (Baker 2008), although the sensitivity of soil moisture to soil depth may be small in soil diffusion models (Guimberteau et al., 2014). The same soil texture was used for all soil layers because of a lack of soil texture data for deeper depths such as the laterite layer at Caxiuana. To represent the expected increase in bulk density in deeper soil layers, the hydraulic conductivity was assumed to increase exponentially with depth (Decharme et al., 2006).

Throughfall exclusion experiments are not fully representative of future climate conditions or atmospheric CO$_2$ concentrations. Besides more severe and persistent dry seasons, atmospheric CO$_2$ concentrations will increase as well as near-surface air temperature and VPD. Therefore, we chose to analyze how the model sensitivity to drought can be affected by increased CO$_2$ concentration and increased temperature. In line with the idealized CMIP5 climate change experiments, we conducted simulations using the same TFE with arbitrary high values of CO$_2$ and temperature: four times the preindustrial CO$_2$ concentration (1080 ppmv), higher temperature (+4°C) and a combination of the two (Table 4). The CO$_2$ concentration and the increase in temperature are constant year round. We did not modify the specific humidity, but a 4°C arbitrary warming lowers the relative humidity and increases the evaporative demand of the atmosphere.
3 Results

3.1 Hydrological response

ISBA\textsubscript{CC} simulates the SWC and its seasonality fairly well between 0 and 3 m (Fig. 3) at both sites for the CTL plots, but the model tends to be too wet during the dry season. The low correlations (around 0.65) between observations and simulations at Tapajós are potentially due to the use of reconstructed forcing data, that were necessary to cover the entire experimental period. Despite a wetter climate (Fig. 2), the simulation at Caxiuanã produces a drier soil, in line with a sandier texture. Due to higher evapotranspiration, the SiB3 and linear WSF reduce the wet bias and improve the seasonality of simulated SWC. When throughfall exclusion is applied to the model, the observed reduction in SWC is also better captured by the linear and the SiB3 WSF (Fig. 3). The SWI remains close to 1 (field capacity) with the drought-avoiding and drought-tolerant WSFs, while it drops below 0.5 with the linear and SiB3 WSFs (Fig. 4). The unstressed transpiration fluxes (at SWI > 1) are lower with the drought-avoiding and drought-tolerant WSFs and the soil moisture is not depleted quickly enough. Therefore, the edaphic water stress is not captured and we expect little impact on the vegetation fluxes. With the linear and SiB3 WSFs, the stomatal conductance is much higher (Fig. 1, bottom left) and soil moisture is depleted much faster by transpiration. The SWI clearly decreases, imposing a strong hydrological stress, mainly with SiB3, as the SWI reaches values close to 0 (the wilting point).

3.2 Vegetation response

3.2.1 Water and carbon budget

To understand the response of ISBA\textsubscript{CC} to drought, we compare the density functions (Fig. 5) of daily SWI, $g_s$, GPP and LE for the dry (August to October) and the wet seasons (February to April). Only the drought-avoiding WSF is plotted because the drought-tolerant WSF showed a very similar behavior. The modeled values of $g_s$, LE and GPP are higher during the dry season than during the wet season in all control simulations, following the higher evaporative demand (Fig. 1) due to higher available energy (fewer clouds) and little soil moisture stress (Fig. 4). The linear and SiB3 WSFs have higher LE and GPP, due to higher stomatal conductance, and a stronger response to drought than using the drought-avoiding and drought-tolerant WSFs. During a drought (dashed lines and shaded areas), the distribution of the SWI is shifted towards lower values with the SiB3 and linear WSFs. With the tolerant (and avoiding) WSF, the sim-
Simulated daily soil wetness index (SWI) with the four water stress function (WSF) at Tapajós (left) and Caxiuanã (right) for both control (CTL) (top) and throughfall exclusion (TFE) for the full experimental period.

Figure 4. Simulated daily soil wetness index (SWI) with the four water stress function (WSF) at Tapajós (left) and Caxiuanã (right) for both control (CTL) (top) and throughfall exclusion (TFE) for the full experimental period.

3.2.2 Autotrophic and heterotrophic respiration

In comparison to ecosystem carbon fluxes derived by Metcalfe et al. (2010) at Caxiuanã, the model overestimates woody tissue respiration and underestimates respiration of leaves and roots. These errors compensate each other and overall the ISBACCC reasonably matches the yearly heterotrophic and autotrophic respiration fluxes (Fig. 8, CTL). This result remains valid over several sites across the Amazon watershed when comparing ISBACCC to the data set compiled by Malhi et al. (2009) (not shown).

In contrast to the observations at Caxiuanã, ISBACCC predicts a decrease of the autotrophic respiration with drought that is not balanced by the increase in vegetation temperature due to the decrease in latent heat production (which reaches a maximum of 2 °C during the driest dry season). Whole ecosystem respiration was observed to increase during the TFE experiments mainly attributable to a temperature corrected enhanced leaf respiration rate per unit leaf area index (LAI) (Metcalfe et al., 2010) as was observed during seasonal drought elsewhere in the Amazon (Miranda et al., 2005). One hypothesis to explain this observation is that the enhanced respiration may supply the supplementary energy demand induced by drought to actively maintain the gradients of the vacuolar solute to keep a minimum turgor (osmotic adjustment) and/or to repair water-stress-induced cell damage (Metcalfe et al., 2010; Atkin and Macherel, 2009, and references within). The majority of ecosystem models, couple autotrophic respiration to assimilation, and implicitly to the LAI which declines during drought. In ISBACCC the heterotrophic respiration is a function of the soil water content, it decreases when in drought, contrary to observations.

3.2.3 Biomass carbon stocks

The simulated daily LAI compares reasonably well with the in situ observation at both control sites (Fig. 9). The SiB3 and linear WSFs result in LAIs a little higher than the drought-
tolerant and drought-avoiding WSFs (in line with a higher $g_s$ and GPP; see Fig. 1). At Tapajós, ISBACC underestimates LAI during the first years of the experiment (2000–2002), which might be partly explained by the reconstructed forcing for these years. At Caxiuanã the anomalously low LAI value (4 m² m⁻²) measured in November 2002 is not captured by the model.

ISBACC fails to simulate the observed substantial loss of LAI (from 1 to 2 points, about 20% of leaf area; Meir et al., 2009) during TFE at both sites. With the drought-tolerant and drought-avoiding WSFs, the soil water content remains above field capacity (SWI > 1, Fig. 4) at both sites, and the simulated LAI shows no response to drought. When using the linear or SiB3 WSFs, the loss of LAI remains underestimated at Tapajós, where the SWI remains relatively high compared to Caxiuanã (Fig. 4). At Caxiuanã, the observed LAI in the TFE experiment diverged from the control within 2 years by more than 1 LAI unit. There are no LAI measurements between 2004 and 2007. The model underestimates the early LAI decrease consecutive to TFE in 2003 with all the WSFs. From the end of 2005 to 2007, the SiB3 WSF results in strong and rapid decreases of LAI during the dry seasons followed by rapid recovery during the wet seasons, partly driven by the the strong seasonality of the soil moisture which almost reaches the wilting point during each dry season after 2005 (Fig. 4).

Although there were no LAI observations in 2005 and 2006, it is likely that this 4 point decrease of simulated LAI is too strong, and the speed of the recovery is not realistic. The fast changes in modeled LAI (Fig. 9), showing little memory of previous droughts, are coherent with the model’s hypothesis that the LAI is driven by current assimilation (Gibelin et al., 2006). With the linear WSF, the model’s behavior is
closer to reality because the SWC remains higher and the vegetation shows a smoother response to drought.

Above-ground biomass observations at Caxiuanã show a reduction of stand-level biomass by 20% after 7 years of TFE, mainly due to enhanced tree mortality. The model predicts AGB in the CTL plot with some skill, but the loss of AGB in the TFE is strongly underestimated with the linear and SiB3 WSF, and not captured at all with the original WSF (Fig. 10). This result is not surprising since ISBA_{CC} only represents background turnover rates depending on biomass stocks and fixed turnover times. There is no representation of mortality processes driven by plant physiology or strong climate anomalies.

3.3 Drought response sensitivity to background temperature and CO$_2$

Under a warmer climate (+4 $^\circ$C), the higher evaporative demand increases LE (Fig. 11, top left panels black dots), and the model becomes more sensitive to drought (Fig. 11, top left panels red dots). Conversely, LE is strongly reduced in the high CO$_2$ simulation due to increased WUE, because stomata, as expected, need to be less open, therefore reducing transpiration, for the same CO$_2$ uptake (Woodward, 1987; Lloyd and Farquhar, 2008). Consequently, the model sensitivity to the experimental drought is completely dampened. The SWI remains close to or above 1 even when removing 60% of the incoming throughfall (red dots).

The GPP is barely impacted by the +4 $^\circ$C in the CTL plot, as the temperature is already close to the assimilation optimum temperature, but is limited in the exclusion plot due to the stronger water stress linked with temperature-induced higher evaporation rates. Maximum GPP increases by about 50% under 4 $\times$ CO$_2$ because of the fertilization effect. It remains high in the TFE plot because the soil remains wet due to the reduction of transpiration. Under higher CO$_2$ concentration, the CO$_2$ diffusion into the mesophyll is easier, therefore enhancing the carboxylation rate (Lloyd and Farquhar, 2008, and references within). Merging the two treatments (+4 $^\circ$C and 4 $\times$ CO$_2$), the higher evaporative demand balances the increased WUE and leads the model to simulate a soil moisture deficit. Note that using the SiB3 WSF leads to similar patterns (not shown) indicating that the strong environmental changes imposed here dominate the model’s sensitivity to drought.
4 Discussion

4.1 Water stress functions

The parameterization of the drought-avoiding and drought-tolerant strategies originally implemented in ISBA\textsubscript{CC} is not effective at simulating gas exchange fluxes when running the model over the Amazon forest, even when the soil moisture is not limiting. This conclusion is very likely to remain valid for other tropical forests, further studies need to assess their validity at global scale. Also, even if the original WSF were meant to represent isohydic and anisohydric drought responses, their performances are not consistent with physiological observations as there is little difference in modeled transpiration between both strategies due to a \( f_0 \) compensation effect. The linear WSF is more suitable for ISBA\textsubscript{CC} but, as the WSF is applied to \( g_m \) and not to \( g_s \), the response to drought of \( g_s \) is nonlinear (Fig. 1). The SiB3 WSF responds too strongly to drought.

The difference in timing and amplitude of the vegetation response to drought when using the linear and SiB3 WSFs illustrate the model sensitivity to the chosen WSF. The WSF parameterization is also likely to be site dependent thus increasing the modeling challenge. The use of different WSF formulations in different land surface models (Egea et al., 2011; Zhou et al., 2013) reflects our inability to define the general behavior(s) for multi-species biomes in which the physiological processes are not yet fully understood. The use of hydrodynamic models that do not include empirical soil moisture response functions, but instead predict drought-induced stomatal closure from the simulation of hydraulic potential, in the continuum soil–plant–atmosphere, has demonstrated some promising results (Williams et al., 2001; Fisher et al., 2006, 2007; Zeppel et al., 2008; McDowell et al., 2013).

4.2 Respiration

Despite measurement uncertainties, leaf respiration at Caxiuana increases significantly with drought (Metcalfe et al., 2010), a process not captured by ISBA\textsubscript{CC}. Other LSMs exhibit the same behavior as shown in a multi-model comparison against the TFE data (Powell et al., 2013). Although a decrease in leaf dark respiration is usually observed when photosynthetic capacity declines under drought, increases in leaf dark respiration have been observed elsewhere (Metcalfe et al., 2010; Atkin and Macherel, 2009), including Amazonia during seasonal drought (Miranda et al., 2005). Powell et al. (2013) asked if we are missing a critical physiological process to accurately compute the plant carbon balance during drought. Even if changes in respiration might be smaller than the decrease in carbon assimilation when in drought (Atkin and Macherel, 2009), resolving this problem via fur-
Figure 7 – model outputs as proxy (Markewitz et al. 2010). Annual differences (exclusion plot minus control plot) in simulated GPP at Tapajós (left) and Caxiuanã (right).

Figure 8. Annual ecosystem (R_ECO), heterotrophic (R_H), autotrophic (R_A), leaf (R_L), root (R_R), and soil (R_S) respirations for the fourth year of the experiment for the control plot (top) and the difference between the exclusion and control plot (bottom) at Tapajós (left) and Caxiuanã (right).

4.3 Mortality

Mortality is a complex process, highly nonlinear in both time and space (Allen et al., 2010; Fisher et al., 2010; McDowell et al., 2011), and is represented by a wide array of algorithms in commonly used LSMs (McDowell et al., 2013). The inability to simulate drought-induced tree mortality is expected from a compartment carbon model such as ISBACC that has no deterministic climate–mortality relationship. This is also a concern for LSMs linking mortality to negative carbon balance through the carbon starvation hypothesis (da Costa et al., 2010; Powell et al., 2013; McDowell et al., 2013). Also, ISBACC, like most LSMs, does not account for the water column pressure within the xylem; the drought-induced cavitation process cannot be represented. Given recent evidence for drought-induced tree mortality (da Costa et al., 2010; Allen et al., 2010), the ability to simulate climate and drought-induced mortality in LSMs is vital to assess the resilience of the Amazon forest under climate change and to estimate vegetation feedbacks. Moreover, increased mortality risk during drought is associated with other processes like fire or insect outbreaks.

The detailed longitudinal data sets and the control over soil moisture that the throughfall exclusion experiments offer yield insights into ecological processes and dynamics are crucial for validating the processes represented by LSMs.

At Caxiuanã baseline mortality rates in the experiment were strongly consistent with data from multiple nearby monitoring plots. In general, however, applying the results of these experiments to larger-scale models will introduce uncertainty. For example, the observed decrease in biomass in the 1 ha throughfall exclusion plots was due to a few large trees that died first (Nepstad et al., 2007; Meir et al., 2009). Therefore, a combination of data sources seems the most effective way forward to constrain biomass and its sensitivity to climate within LSMs. For example, data from long-term inventory plots, such as those from French Guyana since 1991, the RAINFOR data sets in Amazonia (Phillips et al., 2009) or trait-based mortality model outputs (e.g., Aubry-Kientz et al., 2013), should be used with the detailed results from the throughfall exclusion experiments.

4.4 Drought response changes under different background conditions

Increases in CO₂ and temperature are modeled to have antagonistic effects on soil moisture through evapotranspiration because the WUE increases under higher CO₂ concentration (reducing transpiration) while higher temperature will enhance transpiration through a higher vapor pressure deficit. The simulated ET is highly reduced when imposing a high CO₂ concentration and the sensitivity of ISBACC to TFE is completely dampened. Unfortunately, there are no direct observations of the response to elevated CO₂ in tropical forests with which to constrain the reduced transpiration effect im-
Figure 9. Times series of the daily leaf area index (LAI) for the control plot (top) and the difference between the exclusion and control plot (bottom) at Tapajós (left) and Caxiuanã (right).

Figure 10. Times series of the yearly above-ground biomass (AGB) for the control plot (top) and the difference between the exclusion and control plot (bottom) at Tapajós (left) and Caxiuanã (right).

plemented in ISBA_{CC} (and in other LSMs). There is some evidence for an recent increase in WUE due to CO₂-induced stomatal closure, both from flux towers (Keenan et al., 2013) and inferred from increasing global runoff (Gedney et al., 2006; Betts et al., 2007), but these results are disputed. Projections of the vegetation–climate interactions in the Amazon are highly sensitive to the response of the stomatal closure to a CO₂ enrichment (Cowling et al., 2008; Good et al., 2013).
If, as recently suggested in Keenan et al. (2013), LSMs tend to underestimate $\text{CO}_2$-induced stomatal closure, it is likely that increasing WUE will partly offset future droughts and mitigate the expected drier and longer dry season (Fu et al., 2013). Therefore, the stomatal response, which regulates the water exchange within the soil–plant–atmosphere continuum, is fundamental to modeling the vegetation response to climate change (Berry et al., 2010). On the other hand, less ET reduces the water flux towards the atmosphere, the local evaporative cooling and might reduce precipitation through vegetation–atmosphere feedbacks. Numerous global climate model simulations of deforestation in the Amazon showed that regional precipitation is expected to decrease because of the combined influences of increased albedo, decreased surface roughness and decreased water recycling that accompany deforestation (e.g., Dickinson and Henderson-Sellers, 1988; Malhi et al., 2008; Coe et al., 2009). Though increased WUE does not affect albedo or surface roughness like deforestation, it will affect the entire basin, not just the deforested areas.

5 Conclusions

Accurate representation of soil moisture and its seasonal dynamics is a prerequisite to simulate drought impacts on vegetation. Despite reasonable representation of the land surface hydrology, the land surface model ISBA$_{CC}$ fails to correctly simulate the vegetation response to the two Amazon rainfall exclusion experiments. First, a sensitivity analysis based on four WSFs showed that the amplitude and timing of the vegetation response of ISBA$_{CC}$ to drought is quite sensitive to the chosen WSF. The drought-avoiding and drought-tolerant strategies originally implemented in ISBA$_{CC}$ are not suitable for the Amazon forest on account of a $g_\text{s}$ significantly lower than that observed. Of the functions we tested, the simplest linear function performs best.

While at Caxiuanã, the measured autotrophic respiration tends to be higher in the TFE than in the CTL plot, ISBA$_{CC}$ simulates an opposite trend. The observed loss of AGB, hiding the drought enhanced tree mortality, was greatly underestimated or even not captured by the model, as it does not represent mortality driven by strong climate anomalies. In the CTL plots, the representation of the vegetation (in terms of fluxes and stocks) is quite well simulated by ISBA$_{CC}$ but, physiological processes are missing to correctly estimate the vegetation response in case of drought, as with most LSMs (Powell et al., 2013). We also showed that the vegetation response to higher $\text{CO}_2$ and warmer temperature greatly affects its response to drought. As discussed and shown with other models, this can have a great impacts on the estimate of the Amazon response to climate change and the vegetation feedbacks in climate projections.

Land surface models are designed to conduct investigations of processes with large spatial and temporal scales, and therefore the vegetation representation necessarily includes many empirical approximations and coarse abstractions of reality. The definition of a generic drought response for Amazonian forests is evidently a difficult undertaking, particularly when given evidence of the functional diversity of these forests in hydraulic functioning alone (Fisher et al., 2006; Baraloto et al., 2009). The introduction of more complex mechanistic models of drought stress removes the requirement to generate these empirical functions, but implies significantly higher model complexity and requirements for model specification using data that are difficult to acquire.
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