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Forest response to increased disturbance in the Central Amazon and comparison to Western Amazonian forests

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Abstract

Uncertainties surrounding vegetation response to increased disturbance rates associated with climate change remains a major global change issue for Amazon forests. Additionally, turnover rates computed as the average of mortality and recruitment rates

- in the Western Amazon basin are doubled when compared to the Central Amazon, and notable gradients currently exist in specific wood density and aboveground biomass (AGB) between these two regions. This study investigates the extent to which the variation in disturbance regimes contributes to these regional gradients. To address these issues, we evaluated disturbance-recovery processes under two scenarios of in-
- ¹⁰ creased disturbance rates in a complex Central Amazon forest using first ZELIG-TROP, a dynamic vegetation gap model which we calibrated using long-term inventory data, and second using the Community Land Model (CLM), a global land surface model that is part of the Community Earth System Model (CESM). Upon doubling the mortality rate in the Central Amazon to mirror the natural disturbance regime in the Western
- ¹⁵ Amazon of ~ 2 % mortality, at steady-state, AGB significantly decreased by 41.9 % and there was no significant difference between the modeled AGB of 104 Mg C ha⁻¹ and empirical AGB from the western Amazon datasets of 107 Mg C ha⁻¹. We confirm that increases in natural disturbance rates in the Central Amazon will result in terrestrial carbon loss associated with higher turnover. However, different processes were re-
- ²⁰ sponsible for the reductions in AGB between the models and empirical datasets. We observed that with increased turnover, the subsequent decrease in wood density drives the reduction in AGB in empirical datasets. However, decrease in stand basal area was the driver of the drop in AGB in ZELIG-TROP, and decreased leaf area index (LAI) was the driver in CLM. Further comparisons found that stem density, specific wood den-
- sity, and basal area growth rates differed between the two Amazonian regions. This suggests that: (1) the variability between regions cannot be entirely explained by the variability in disturbance regime, but rather potentially sensitive to intrinsic environmental factors; or (2) the models are not accurately simulating all forest characteristics in



response to increased disturbances. Last, to help quantify the impacts of increased disturbances on climate and the earth system, we evaluated the fidelity of tree mortality and disturbance in a global land surface model: CLM. For a 100% increase in annual mortality rate, both ZELIG-TROP and CLM were in close agreement with each other and predicted a net carbon loss of 41.9 and 49.9%, respectively, with an insignificant

effect on aboveground net primary productivity (ANPP). Likewise, a 20% increase in mortality every 50 years (i.e. periodic disturbance treatment) resulted in a reciprocal biomass loss of 18.3 and 18.7% in ZELIG-TROP and CLM respectively.

1 Introduction

- One of the largest uncertainties in future terrestrial sources of atmospheric carbon dioxide results from changes to forest disturbance and tree mortality rates, specifically in tropical forests (Cox et al., 2000, 2004; DeFries et al., 2002; Clark, 2007; Pan et al., 2011). There has been evidence that climate change and forest disturbance are linked such that a changing climate can influence the timing, duration, and intensity of dis turbance regimes (Overpeck et al., 1990; Dale et al., 2001; Anderegg et al., 2013). In the tropics, climate change related impacts such as water and heat stress, and increased vulnerability to fires could lead to increased forest dieback (i.e. tree mortality
- notably higher than usual mortality) and increased disturbance rates (Cox et al., 2004; Malhi et al., 2008, 2009; US DOE, 2012). Recent studies have detected that forested
- 20 ecosystems are already responding to climate change related factors such as drought, temperature stress, and associated hydraulic stress, and insect and pathogen infestation, triggering varying levels of forest diebacks (Cao et al., 2009; Allen et al., 2010; Martinez-Vilalta et al., 2012; Anderegg et al., 2013). Increased mortality from droughts, fires, storms, deforestation, and introduced species can decrease carbon sequestra-
- tion, ecosystem productivity and climate stabilization (Millennium Ecosystem Assessment, 2005; Bonan, 2008). These effects could then produce large economic costs, ecological impacts, and lead to climate related positive feedback cycles (Canham and



Marks, 1985; Dale et al., 2001; Laurance and Williamson, 2001). Therefore, due to the connection between climate change and tropical forest disturbance rates, investigations on the consequences of increased disturbance rates on aboveground carbon stocks and fluxes are required.

- The effects of large-scale removal of tropical forest leading to changes in global climate and the global carbon balance have been studied within global general circulation models (GCMs) (Shukla et al., 1990; Henderson-Sellers et al., 1993; Hahmann and Dickinson, 1997; Gedney and Valdes, 2000; Avissar and Werth, 2005; Huntingford et al., 2008). For example, a rapid and complete deforestation of the diverse Ama-200 Basin was predicted to be irreversible (Shukla et al., 1990), losing ~ 180 Gt carbon
- ¹⁰ Zon Basin was predicted to be irreversible (Shukla et al., 1990), losing ~ 180 Gt carbon stock from the tropical forest. In general, these past studies have simulated extreme deforestation, or complete removal of the tropical forest biome, with the goal of evaluating climate impacts (i.e. albedo, evaporation, precipitation, surface boundary conditions). However, instead of sudden and complete removal, gradual increases and spatially
- ¹⁵ heterogeneous patterns of tropical tree mortality due to multiple causes are more likely to occur than complete loss (Fearnside, 2005; Morton et al., 2006). Therefore, it is necessary to improve global climate models in order to simulate dynamic disturbance rates and gradual forest biomass loss in response to increasing mortality rates. This study aims to better understand how shifts in disturbance regimes and background mortality

rates will affect ecosystem processes and carbon cycling dynamics for tropical forests. Disturbance regimes and turnover rates currently vary for different regions of Amazonia (Baker et al., 2004a, b; Lewis et al., 2004; Phillips et al., 2004; Chao et al., 2009), with Central Amazon forests having "slower" turnover rates, and the Western and Southern Amazon forests (which we call "west and south") exhibiting "faster" turnover

rates. This regional variation in turnover rates is connected with differences in carbon stocks, growth rates, specific wood density, and biodiversity. Baker et al. (2004a) investigated the regional-scale comparisons of AGB estimates, concluding that differences in species composition and related specific wood density determine the regional patterns in AGB. Upon including weighting for wood density when estimating AGB,



western Amazon forests were found to have significantly lower AGB than their eastern counterparts. In contrast basal area varied only slightly across their Amazon plot network (27.5 vs. 29.2 m² ha⁻¹, Baker et al., 2004a), making wood density, which is a strong indicator of functional traits (e.g., species light demand; Whitmore, 1998) and patterns of family composition, a strong driver in steady-state AGB variation. It is unclear if these regional variations in forest processes are driven by external disturbance (e.g., increased drought, windstorm, forest fragmentation) or internal influences (e.g., soil quality, phosphorus limitation, species composition) (Phillips et al., 2004; Chao et al., 2009; Quesada et al., 2010; Yang et al., 2014). Investigating the causes that drive variation in tree dynamics in the Amazon, in order to understand consequences 10 for future carbon stocks for each region should still be explored. For example, are the differences in forest structure and function between the two regions a result of the disturbance regime? If the Central Amazon forests were subject to a higher disturbance regime and turnover rates similar to that of the "west and south", would the two regions match in terms of forest dynamics, carbon stocks and fluxes? A goal of this paper is

¹⁵ match in terms of forest dynamics, carbon stocks and fluxes? A goal of this paper is to address these questions, by using model predictions to explore the influence of disturbance regimes on net carbon stocks and fluxes in the Central Amazon, and then compare to observational data from the "west and south" regions of the Amazon.

The effectiveness of climate mitigation strategies, and the consequential negative or positive effect on carbon pools, will be affected by future changes in natural disturbances regimes (IPCC working group III). A recent study by Le Page et al. (2013) found that future natural disturbance rates can greatly alter climate mitigation strategies due to the effect of disturbance on the terrestrial carbon balance. Specifically, by using this economic/energy integrated assessment model, it was found that in order to reach a stringent mitigation target, (3.7 W m⁻² level) the societal, technological, and economic strategies will be substantially more costly- up to 2.5 times when disturbance rates are doubled (Le Page et al., 2013). Due to the strong feedbacks from terres-

trial processes, there is a need to utilize an integrated Earth System Model approach (i.e. iESM; Jones et al., 2013) with biogeochemical, biophysical, and climate model-



ing such as CESM. To improve upon representing terrestrial feedbacks, and evaluate the influence of disturbance on net carbon loss and varying forest dynamics between two regions (central vs. "west and south"), we are using an individual-based, demographic, "gap-model" approach (Botkin et al., 1972; Shugart, 2002) as a "benchmark" to evaluate disturbance and mortality in CLM-CN 4.5 (called CLM for remainder of paper) in order to model responses to disturbances more accurately. We used the dynamic vegetation gap model ZELIG (Cumming and Burton, 1993; Urban et al., 1993) to predict the impact of prolonged high disturbance rates and to serve as a standard for testing the accuracy of the mortality algorithm in CLM. Recently, ZELIG has been updated and modified to simulate a tropical forest in Puerto Rico with a new versatile disturbance rauting (ZELIG, TROP: Holm et al., 2012) making this vegetation dynamic

¹⁰ updated and modified to simulate a tropical forest in Puerto Rico with a new versatile disturbance routine (ZELIG-TROP; Holm et al., 2012), making this vegetation dynamic model a good choice for this study.

Vegetation and carbon response to increased disturbance rates resulting from human induced climate changes must be examined in more detail. Specifically, under-

- standing the future role of the Amazon forest in the global carbon cycle when disturbance is increased, as well as understanding the role of varying disturbance regimes towards contributing to regional variation in forest processes, stocks, and fluxes are both goals of this study. In order to forecast changes in forest carbon sinks and sources in CLM we addressed differences in AGB, ANPP, growth rates, and coarse litter pro-
- ²⁰ duction rates as a result of disturbances. The main research questions of the study are: (1) what are the long-term consequences of increased disturbance rates, imposed by doubling background mortality rates and applying a periodic disturbance treatment, in the Central Amazon? (2) Will the detailed gap model developed to address question number one simulate the central Amazon Forest and accurately estimate rainforest
- stand dynamics? (3) Can the variability in forest dynamics and carbon stocks between the Western and Southern Amazon and the Central Amazon forests be explained by the variability in the natural disturbance regime (i.e. higher mortality rates)? Finally: (4) what are the differences after increasing disturbance rates in ZELIG-TROP vs. CLM for the Central Amazon? We are assuming an independent driver of mortality; therefore



we are not assigning mortality to any particular cause. The final research question will evaluate the accuracy of CLM to predict changes to carbon fluxes due to increased disturbance, a process that is likely to increase with human induced climate change.

2 Methods

5 2.1 Study area and forest inventory plots

The empirical data used for this study were from two permanent transects inventoried from 1996–2006, located in reserves of the National Institute for Amazon Research (Instituto Nacional de Pequisas da Amazonia, INPA) in the Central Amazon in Brazil. The forest inventory transects are approximately 60 km north of Manaus, Brazil, in the Central Amazon where vegetation is old-growth closed-canopy tropical evergreen for-

¹⁰ Central Amazon where vegetation is old-growth closed-canopy tropical evergreen forest. The mean annual precipitation at Manaus was 2110 mm yr⁻¹ with a dry season from July–September, and mean annual temperature was 26.7 °C (Chambers et al., 2004; National Oceanic and Atmospheric Administration, National Climatic Data Center, Asheville, N.C., USA). However, during 2003 to 2004, mean annual precipitation in the study area reached 2739 mm yr⁻¹.

We quantified demographic data such as stem density, diameter at breast height (DBH, cm), and change in diameter for trees > 10 cm DBH from census data from the two transects. This data was used to calculate above-ground biomass (ABG) estimates (Mg C ha⁻¹) and were determined using region-specific allometric equations after har-

vesting 315 trees in the Central Amazon (Chambers et al., 2001; see Eq. 1 below). This data was also used to estimate observed values for above-ground net primary productivity (ANPP, Mg C ha⁻¹ yr⁻¹) after taking into account loss of tree mass due to tree damage (Chambers et al., 2001). Observed mortality rates (% stems yr⁻¹) were based on census intervals ranging from 1 to 5 years on 21 1 ha undisturbed plots located in the Biomass and Nutrient Experiment (BIONTE), and the Biological Dynamics



and Forest Fragments Project (BDFFP), also located in INPA (Chambers et al., 2004). We compared model predictions from ZELIG-TROP to observed field data.

In order to test whether the variability in forest dynamics and carbon stocks between the "west and south" and the Central Amazon forests can be explained by the variability

- in the natural disturbance regime, we used forest inventory data collected and reported in Baker et al. (2004a) and Phillips et al. (2004). We compared the Central Amazon forests (both simulated and observed data) to the observed "west and south" datasets, using inventory data collected from 59 plots as reported in Baker et al. (2004a, b), and from 97 plots as reported in Phillips et al. (2004) with these plots constituting a large part of the RAINFOR Amazon forest inventory network (Malhi et al., 2002). Sites occur
- ¹⁰ part of the RAINFOR Amazon forest inventory network (Malhi et al., 2002). Sites occur across a large range of environmental gradients, such as varying soil types and level of seasonal flooding, however all sites are considered to be mature tropical forests.

2.2 Description of ZELIG-TROP

ZELIG-TROP is an individual based gap model developed to simulate tropical forests
 (Holm et al., 2012). It is derived from the gap model ZELIG (Urban, 1990, 2000; Urban et al., 1991, 1993), which is based on the original principles of the JABOWA (Botkin et al., 1972) and FORET forest gap models (Shugart and West, 1977). ZELIG-TROP follows the regeneration, growth, development, and death of each individual tree within dynamic environmental conditions across many plots (400 m² plots, replicated uniquely

- 100 times). Maximum potential tree behaviors (optimal tree establishment, diameter growth, and survival rates) are reduced as a function of light conditions, soil moisture, level of soil fertility resources, and temperature. Specific details on the ZELIG model modifications to create ZELIG-TROP can be found in Holm et al. (2012). Gap models have been used extensively to forecast forest change from varying types and levels of
- ²⁵ disturbances, such as windstorms and hurricanes (O'Brien et al., 1992; Mailly et al., 2000); simulate vegetation dynamics in response to global change (Solomon, 1986; Smith and Urban, 1988; Smith and Tirpak, 1989; Overpeck et al., 1990; Shugart et al., 1992); and explore feedbacks between climate change and vegetation cover (Shuman



et al., 2011; Lutz et al., 2013). ZELIG has been used to simulate forest succession dynamics in many forest types across the globe (O'Brien et al., 1992; Seagle and Liang, 2001; Busing and Solomon, 2004; Larocque et al., 2006; Nakayama, 2008). (Descriptions of the plant mortality algorithm as well as definitions of terms and parameters used in ZELIG-TROP are provided in the Supplement).

2.3 Model parameterization for the Central Amazon

The silvicultural and biological parameters for each of the 90 tropical tree species required for ZELIG-TROP are found in Table 1. The 90 tree species consist of 25 different families, 54 canopy species, 18 emergent species, 12 sub-canopy species, and 6 pioneer species (Table 1). While these tree species do not represent all existing species found in the Central Amazon forest, they represent a diverse array of family types, canopy growth forms, and demographic traits such as growth rates, stress tolerances, and recruitment variations that will produce a robust and reliable result. The majority of the data used to parameterize ZELIG-TROP for the Amazon was derived from a long term (14, 18 years) demographic atudy to estimate tree longevity (Laurance

- ¹⁵ a long-term (14–18 years) demographic study to estimate tree longevity (Laurance et al., 2004) located in Central Amazon. Data was collected on 3159 individual trees from 24 permanent, 1 ha plots which span across an area of 1000 km² (Laurance et al., 2004). Wood density data for the 90 species used in this study were gathered from published sources with sites across South America (Fearnside, 1997; Chave et al., 2006).
- We used results found by Laurance et al. (2004) to determine several parameters; specifically the maximum age of the species (AGEMAX), the maximum diameter at breast height (DBH_{max}, cm), and the growth-rate scaling coefficient (*G*) for ZELIG-TROP. AGEMAX was found by taking the mean of three longevity estimates. DBH_{max} were scaled to match a more accurate representation of maximum DBH in the simu-
- ²⁵ lated field sites (Chambers et al., 2004). Additionally, we used the canopy classification and family traits to infer species-specific rankings related to light, drought, soil nutrient tolerances, and regeneration stocking. Average monthly precipitation (cm) and temperature (°C) required for the environmental parameters in ZELIG-TROP (Table 2) were



based on field data collected from 2002–2004 in the study site (Tribuzy, 2005). Soil field capacity (cm) and soil wilting point (cm) were determined from soil measurements in nearby central Amazon study sites (Laurance et al., 1999).

In order to more accurately simulate the Central Amazonian forest, a few modifications were made to the original ZELIG-TROP model (Holm et al., 2012). First, the allometric equation used to estimate above-ground biomass (Mg C ha⁻¹) was updated to include an equation specific for the Brazilian rainforest in the Central Amazon (Chambers et al., 2001; Eq. 1).

 $\ln(\text{mass}) = \alpha + \beta_1 \ln(\text{DBH}) + \beta_2 [\ln(\text{DBH})]^2 + \beta_3 [\ln(\text{DBH})]^3$

where above-ground biomass (mass) is in kg, α is -0.370, β_1 is 0.333, β_2 is 0.933, and β_3 is -0.122 ($r_{adj}^2 = 0.973$) based upon data collected from 315 harvested trees. Specific wood density is not taken into account in this model.

In model development of the original ZELIG-TROP (modified for a subtropical dry forest), death caused by natural mortality (age-related) was killing tropical trees prema-

- ¹⁵ turely. This was also seen in initial model testing for the wet tropical forest. In contrast to tropical dry forests, individuals in tropical wet forests have a longer life potential and a higher likelihood of reaching their potential size. For example, the Central Amazon is able to support trees > 1000 years old (Chambers et al., 1998, 2001; Laurance et al., 2004), where a dry forest may only be able to support trees to a maximum of 400 years.
- To adjust for this variation, the natural survivorship rate was increased from 1.5% to 6% of trees surviving to their maximum age (Table 1). This was a conservative value, with one study estimating about 15% of species in Central Amazon attaining their maximum ages (Laurance et al., 2004). Lastly, we also modified ZELIG-TROP's mean available light growing factor algorithm, which in part was used to accurately calculate tree height
- and crown interaction effects, as developed in ZELIG-CFS (Larocque et al., 2011). To best portray tree growth and crown development typical of an individual within a tropical canopy, we used an earlier algorithm version developed for ZELIG-CFS. This algorithm was the ratio of available growing light factor (ALGF) to a doubled crown width for each



(1)

individual, thereby adjusting the ALGF relative to horizontal space occupied by the crown and improving the predictive capacities of ZELIG-TROP for the Amazon. This modification thus affected the light extinction on tree growth, allowed more available light from the top to the bottom of the individual-tree crown, and in turn better predicted observed data of basal area growth and abundance of stems per plot.

2.3.1 Calibration methods

ZELIG-TROP simulations for the Central Amazon forest were run for 500 years and replicated on 100 independent plots, each the size of 400 m². All simulations began from bare ground, and results from ZELIG-TROP were averaged over the final 100
 ¹⁰ years of simulation. This was the period when forest dynamics (e.g. stem density, AGB, ANPP) were seen to reach a stable state and represent a mature forest stand. The model was verified by comparing the following five simulated forest attributes (average ± SD) to observed field data from the two inventory transects: (1) total basal area (m² ha⁻¹); (2) total AGB (Mg C ha⁻¹); (3) total stem density (ha⁻¹); (4) leaf area index;
 ¹⁵ and (5) ANPP (Mg C ha⁻¹ yr⁻¹). To test model validity for the Central Amazon forest we report percent difference between the observed and simulated results (Table 3).

2.4 Disturbance treatments

To better understand the long-term consequences of high disturbance in a Central Amazon rainforest, we crafted a simulation that doubled annual background tree mor-

- tality in both ZELIG-TROP and CLM assuming an independent mechanism as the driver of mortality. A description of the Community Land Model (CLM) can be found in the Supplement. Predicting the impacts of increased mortality is critical since other recent studies have found that tree mortality in the Central Amazon has been under-sampled in plot-based approaches, and after analyzing a larger range of gap sizes (in the first studies).
- (including larger gaps), ~ 9.1 to 16.9 % of tree mortality was missing (Chambers et al., 2013). The majority of gaps created in Amazonian rainforests are from windthrow of



canopy trees with a large percentage of gaps having relatively small areas of < 200 m² (Uhl, 1982; Denslow, 1987; Stanford, 1990). However, some windthrow events will create large gaps that then initiate secondary succession processes (Brokaw, 1985; Chambers et al., 2013). Since there can be multiple spatial scales and drivers of tree mortality, we are simulating mortality as a stochastic, independent event within ZELIG-TROP, using the new versatile disturbance routine implemented in Holm et al. (2012). Most mortality events in the Central Amazon occur on individual trees (Chambers et al., 2004, 2013). Therefore, this phenomenon was replicated in the model. Specifically, any

- one tree > 10 cm DBH was randomly selected to die and be removed from the forest canopy on an annual basis at the gap scale, in addition to the existing selection of trees removed by natural senescence. This "high disturbance" treatment for the Central Amazon forests is representative of the current turnover rates in "west and south" (Phillips et al., 2004), thus creating an opportunity to test whether the variability in forest dynamics and carbon stocks between the "west and south" and the Central Amazon
- ¹⁵ forests can be explained by the variability in the natural disturbance regime. Variables compared between the two regions included AGB, wood density (Baker et al., 2004a), recruitment rates, and stem density (Phillips et al., 2004), and stand-level BA growth rates (Lewis et al., 2004).

A second treatment has been applied in order to improve understanding of periodic large-scale disturbance and recovery events. This treatment consisted of removing 20% of stems > 10 cm DBH every 50 years (i.e. periodic treatment). It has recently been noted that patch-scale (400 m²) succession-inducing disturbances exhibit a return frequency of about 50 years within the Central Amazon region (Chambers et al., 2013). Therefore we have set our large-scale disturbance event to repeat four times over a 200

²⁵ year period (every 50 years) after the forest has reached a mature stable state. This treatment was also conducted in both ZELIG-TROP and CLM. An important metric in determining the forest carbon balance as a result of disturbance is the total change in stand biomass over time (Δ AGB, MgCha⁻¹), defined as AGB_{t₂} – AGB_{t₁} over the simulation period.



3 Results

3.1 Model verification results

Results simulated by ZELIG-TROP for the mature Central Amazon tropical forest (predisturbance treatment) were in close range (e.g., within 17%) to empirical data (Table 3), making ZELIG-TROP successful at predicting stand dynamics of a complex 5 tropical forest. Average basal area was 9.7% higher than the observed value (32.96 vs. $30.06 \text{ m}^2 \text{ ha}^{-1}$), average AGB was 5.0% higher (178.38 vs. $169.84 \text{ Mg C ha}^{-1}$), and average leaf area index (LAI) was 1.8% higher (5.8 vs. 5.7). ZELIG-TROP predicted average stem density to be 12.5 % lower (574 vs. 656 stems ha⁻¹), and ANPP was 17.1% lower than observed values reported by Chambers et al. (2001) (5.4 vs. 6.5 Mg C ha⁻¹ yr⁻¹). All simulated values were reported once the forest reached a stable, mature forest. ZELIG-TROP was also successful at accurately predicting stem density and AGB by DBH (cm) size class (Fig. 1a and c). The model over predicted the number of stems in the lowest size class (10-20 cm), by an additional 84 stems per hectare, and in the eighth size class (80–90 cm), but for the remaining size classes 15 values were near to the observed data. Even with these slight over predictions in certain DBH size classes, the model predicted AGB to be within a reasonable range $(8.5 \text{ Mg C ha}^{-1})$ of the observed values $(r^2 = 0.60)$.

ZELIG-TROP was also able to predict a realistic community composition (Fig. 2a). After initiating the model from bare ground, there was a sudden increase in basal area per species, followed by a typical jigsaw pattern of die-offs and growth increases, with the model reaching a steady-state during the last 100 years. The dominant species in terms of basal area, (*Parkia multijuga*), a large, fast-growing emergent species from the Leguminosae family accounted for 17 % of the total basal area in the last 100 years of

simulation. The next four dominant species were all canopy-level species. This was an accurate representation of the forest, as the canopy layer holds large trees and usually the densest area of biodiversity. For example, 63 % of the 90 tree species simulated were categorized as a canopy growth form. However, there was also an even mixture



of emergent, sub-canopy, and pioneer species as dominant and rare species, typical of a diverse Central Amazon forest. There was no one single species that dominated the canopy throughout the course of the simulation. Instead, we saw a diverse species representation (Fig. 2a). During the last 100 years of simulation, emergent species represented 29.6% of the total basal area, sub-canopy species represented 1.7%, and pioneer species represented 5.5% of the total basal area.

Empirical mortality rates (% stems yr^{-1}) from BDFFP and BIONTE data were lognormally distributed averaging $1.02\% \pm 1.72\%$ (Chambers et al., 2004). As estimated by ZELIG-TROP, the no-disturbance annual mortality rates were near to observed val-

¹⁰ ues (1.27 $\% \pm 0.21 \%$) but had a smaller distribution around the mean (Fig. 3). As expected, annual mortality rate doubled (2.66 $\% \pm 0.26 \%$) for the high disturbance treatment.

3.2 Central and Western Amazon disturbance comparisons

3.2.1 AGB, stem density, growth and recruitment rates

Upon increasing the turnover rates of the Central Amazon forest by manually doubling 15 background tree mortality rates to mirror the 2 % yr⁻¹ mortality rates in the "west and south", the two Amazon regions continued to differ in multiple forest structure and functional traits. Stem density, specific wood density, basal area growth rates, and AGB when excluding weighting for wood density in biomass equations from the treatment site did not match the trends observed in the "west and south" plot network. Using 20 a Tukey's multiple comparison procedure following a one-way ANOVA, the empirical dataset showed a significant difference in both wood density and basal area growth rates between the two regions, but no significant difference between the two regions in the model results (Fig. 4). Plus there were slight opposite responses in the wood density and basal area growth rate values predicted by the model compared to the ob-25 served trends (discussed below). Alternatively when comparing stem density there was no significant difference between the two regions in the empirical dataset, but there was



a significant increase between the Central and "west and south" regions in the model results.

- In contrast, the high disturbance treatment did significantly reduce AGB in the Central Amazon to values similar to the "west and south" counterpart, but wood density ⁵ was not included in the biomass allometric equation for the Central Amazon therefore this reduction in AGB was a "false-positive". Specifically, when the Central Amazon was subjected to faster turnover rates there was a significant reduction in AGB (two sample *t* test, $t_{(99, 1.97)} = 108.98$, p < 0.001) and net carbon loss was 74 Mg C ha⁻¹ (from 178 to 104 Mg C ha⁻¹) averaged over the last 100 years of simulation (Fig. 1d) equivalent
- ¹⁰ to a 41.9 % decrease. AGB in the Central Amazon wa impacted the most by the high disturbance treatment. This new value of AGB for the higher disturbed Central Amazon (104 Mg C ha⁻¹) was similar to AGB values in the "west and south" RAINFOR network plots when: (1) using the Chave et al. (2001) biomass equation (107 Mg C ha⁻¹), which had no significant difference between the two regions (two sample *t* test, $t_{(38,2.7)} = 2.29$,
- ¹⁵ considering alpha = 0.01, p = 0.03), and (2) using the Chambers et al. (2001) biomass equation (123 Mg C ha⁻¹) (Fig. 4a). Therefore, similar to the empirical dataset, the demographic vegetation model captured the net carbon loss in biomass associated with higher turnover caused by higher disturbance. However both the Chave et al. (2001) and Chambers et al. (2001) biomass equations included weighting for wood density.
- ²⁰ The significant reduction in stand basal area, and not variation in wood density, was the main driver of decrease in AGB in ZELIG-TROP (Fig. 5e). There was no significant difference in stand basal area between the empirical datasets in the Central and "west and south" plots (p = 0.368), a finding also confirmed by Baker et al. (2004a). While net carbon loss was the expected result, it constitutes a "false positive" resulting ²⁵ from omitting wood density in the model estimate of biomass and from an absence of

significant difference in stand basal area across the Amazonia field network. The high disturbance treatment in the Central Amazon led to a significant increase in stem density by 197 stems from 574 to 771 stems ha⁻¹ (34.3% increase, Fig. 1b.



two sample t test, $t_{(99, 1.97)} = 28.06$, p < 0.001). Compared to the regional gradient

in the RAINFOR network there was no significant difference between the higher disturbed and the Central Amazon empirical dataset (573 stems ha⁻¹ vs. 589 stems ha⁻¹) (two sample *t* test, $t_{(46,2.01)} = 0.84$, p = 0.407, Fig. 4d). ANPP did not significantly alter in the Central Amazon forest under a high disturbance treatment (two sample *t* test, $t_{(99, 1.97)} = 1.54$, p = 0.126), only decreasing ANPP by 0.04 (from 5.39 to 5.35 Mg C ha⁻¹ yr⁻¹, 1.0%, Fig. 5a). Even with increased disturbance events, ANPP did not decrease in the same manner as biomass due to recovery episodes from more frequent thinning and the increase in smaller stems (10 cm DBH size class) in newly opened gaps. When comparing the stand-level BA growth rates (proxy for productivity) there was a significant increase in growth rates as you moved across the RAINFOR

- there was a significant increase in growth rates as you moved across the RAINFOR network from the Central Amazon to the "west and south", but there was no significant difference between the modeled treatments. In fact, an opposite response was seen, and there was a slight decrease as a result of higher disturbance (by 0.21 m² ha⁻¹ yr⁻¹, Fig. 4e or 0.20 Mg C ha⁻¹ yr⁻¹, Fig. 5c). The model might not accurately represent growth rates because prior to applying a higher disturbance regime in the Central
- Amazon, ZELIG-TROP significantly over-estimated the stand-level growth compared to empirical data (3.2 vs. $1.4 \text{ m}^2_1 \text{ ha}^{-1} \text{ yr}^{-1}$).

The recruitment rates (% yr⁻¹) from the treatment site constitute the only variable that matched the "west and south" observational dataset. Under a high disturbance

- treatment in the Central Amazon, as expected, there were subsequent increases in recruitment rate, where recruitment significantly increased from 2.3 to 3.9 % yr⁻¹, constituting a 69.1 % increase above no-disturbance recruitment rates (Table 4, Fig. 6a). Pre-treatment, modeled recruitment rates were 0.9 % yr⁻¹ higher compared to empirical values from the Central Amazon BDFFP plots (Phillips et al., 2004). Recruitment
- ²⁵ and mortality rates are tightly linked (Lieberman et al., 1985), therefore when tree mortality increased, recruitment also significantly increased. In the "west and south" empirical dataset recruitment rates were ~ 79 % higher compared to the Central region (Fig. 4b). However, while turnover rates increased, there was *not* an increase in coarse litter production rate (trunks and large stems > 10 cm diameter, Mg C ha⁻¹ yr⁻¹,



Fig. 6b) compared to the no-disturbance scenario, but rather a significant decrease (two sample *t* test, $t_{(99, 1.97)} = 2.70$, p < 0.01). Under a high disturbance treatment, the production of coarse litter decreased by an average of 0.25 Mg C ha⁻¹ yr⁻¹ (8.3%, Table 4). However it is unclear if this decrease in production of coarse litter is biologically or atmospherically significant.

Once the forest reached a mature stable state (after 500 years) the periodic disturbance treatment was applied, removing 20% of stems in the mature forest every 50 years (for a duration of 200 years). The carbon loss over the 200-year period, including the four large-scale disturbances, was less severe than the high-disturbance treatment, but was still a significant decrease (two sample *t* test, $t_{(99, 1.97)} = 22.73$, p < 0.001). Compared to the no-disturbance scenario, average AGB net carbon loss was 40 Mg C ha⁻¹ (from 178 to 138 Mg C ha⁻¹, 22.7%, Fig. 7c) and ANPP significantly decreased from an average of 5.39 to 5.06 Mg C ha⁻¹ yr⁻¹ (6.1%, two sample *t* test, $t_{(99, 1.97)} = 7.65$, p < 0.001). For the periodic treatment, the decrease in biomass was roughly half the decrease observed in the high-disturbance treatment, however the decrease in ANPP was more severe.

3.2.2 Community composition changes

The individual-based dynamic vegetation model approach was able to explore the longterm changes to community composition and fate of each species with increased dis-

- ²⁰ turbance. A high disturbance treatment lowered the total basal area of the forest, shifted species composition towards a more even canopy structure, and increased the species evenness and diversity (Fig. 2b). The largest basal area reduction occurred in the most common species; specifically the top two emergent species, followed by the most common canopy species. With an increase in disturbance, the species originally occupying
- the largest basal area on the plot, *Parkia multijuga*, decreased by 94.8% in relative difference in basal area compared to all species averaged over the last 100 years. The



next most common emergent species, *Cariniana micrantha*, decreased by 32.6 % with high disturbance, and canopy species filled in as the dominant growth form (Fig. 2b).

Similar to stem density and basal area growth, average wood density differed significantly between the Central Amazon treatment site and the RAINFOR plots from

- ⁵ "west and south" regions (both locations having ~ 2% mortality) (two sample *t* test, $t_{(23,2.07)} = 9.71$, p < 0.001), and displayed opposite patterns (Fig. 4c). Previous studies have found wood density to be higher in the central region (~ 0.68 g cm⁻³), and lower in more disturbed "west and south" (~ 0.57 g cm⁻³) (Baker et al., 2004a). Before implementing the high disturbance treatment average wood density was low for the non-
- ¹⁰ disturbed Central Forest (0.59 g cm⁻³, similar to values of the "west and south"), and with increased disturbances average wood density increased (0.63 g cm⁻³), an opposite response from empirical trends. Taking a closer look at the community composition and representation of species, the emergent canopy class experienced a decrease in basal area, amounting to 7.8% of total basal area, compared to 29.6% prior to high
- disturbances. The three remaining growth forms all increased in basal area. The emergent species had on average the highest wood density (0.72 g cm⁻³), and the pioneer species had on average the lowest wood density (0.52 g cm⁻³). With a decrease in emergent species, it would seem likely that average wood density would decrease, as expected in a forest with higher turnover rates. However the dominant species prior to
- ²⁰ disturbance (the emergent: *Parkia multijuga*), which experienced the largest decrease in basal area, had a very low wood density (0.39 g cm⁻³). In addition, even though the emergent size class decreased, the canopy species (which also had high average wood density of 0.71 g cm⁻³) basal area increased from 63 % to 79.6 %, and the increase in pioneer species from 5.5 % to 5.9 % was not sufficient to lower the total wood density
- ²⁵ of the forest. With higher disturbance rates subcanopy species represented 6.7 % of the total basal area, compared to 1.7 % prior to high disturbances.



3.3 Disturbances and carbon change in CLM-CN 4.5 vs. ZELIG-TROP

After applying a continual disturbance regime within CLM as in ZELIG-TROP, similar patterns in forest biomass in response to disturbance were observed, and both models were in agreement with each other. For example, the relative change in AGB was con-

- sistent (41.9% vs. 49.9% decrease) for ZELIG-TROP and CLM respectively (Fig. 5b). In CLM the aboveground carbon storage pools are not determined using allometric equations, but rather through a carbon allocation framework based off of photosynthesis, total GPP, and respiration (Thornton et al., 2002). Including or excluding specific wood density is not considered in CLM. The models outputs from CLM for the disturbed
- ¹⁰ Central Amazon runs also showed a reduction in AGB similar to the "west and south"; which was also a "false-positive" result. The significant loss of LAI with disturbance was the main driver of reduction in AGB (Fig. 5f). There was a weak non-significant difference in LAI between the empirical datasets in the Central and "west and south" Amazon regions (p = 0.077). Another similarity between the two models was the non-
- significant change in ANPP, however ZELIG-TROP predicted a decrease in ANPP while CLM predicted a slight increase in ANPP (Fig. 5a).

With regards to the periodic disturbance treatment of large-scale disturbance events, CLM also replicated analogous patterns in biomass loss and recovery as seen in ZELIG-TROP (Fig. 7c). In both models, the sudden decrease in biomass as well as re-equilibration during the recovery phase matched. During each pulse disturbance, the forest lost on average 18.3% and 18.7% biomass in ZELIG-TROP and CLM respectively, and gained 16.5% and 15.4% biomass during the recovery phase. Both CLM and ZELIG-TROP predicted that the recovering forest biomass, on average, was less than the amount lost in each large-scale disturbance event, therefore generating a negative total Δ AGB (-0.15 and -0.46 Mg C ha⁻¹ yr⁻¹ for ZELIG-TROP and CLM respectively, Table 4). The negative total Δ AGB was less in ZELIG-TROP, and was likely attributed to ZELIG-TROP predicting growth rates to significantly increase (by 0.20 Mg C ha⁻¹ yr⁻¹, two sample *t* test, $t_{(99, 1, 97)} = 2.14$, *p* < 0.05), most likely due to



the open gaps from disturbance, therefore losses were damped in ZELIG-TROP. In contrast CLM had growth rates that on average decreased, due to the sharp decrease in growth rates following each large-scale disturbance event (Fig. 7b). Both models also showed that each subsequent recovery period was always greater than the previous period, up to a point where re-growth matched the biomass lost in the disturbance event (Fig. 7c).

With regards to the periodic disturbance treatment, there were discrepancies with the response of ANPP to the large-scale forest mortality and recovery events between CLM and ZELIG-TROP. The immediate decrease in ANPP following the large-scale disturbance event was significantly greater in CLM compared to ZELIG-TROP (4.7 vs. 0.6 Mg C ha⁻¹ yr⁻¹, Fig. 7a). The subsequent shape of ANPP during the 50-year recovery was also different between the two models. CLM predicted that within approximately two years after the disturbance, ANPP returned to pre-disturbance levels and stayed relatively constant until the next disturbance. However, ZELIG-TROP did not display a fast return to pre-disturbance levels, but instead predicted a gradual increase in ANPP after each disturbance. Comparing the no-disturbance scenario and the pe-

riodic treatment, both models predicted that overall ANPP significantly decreased with periodic disturbances (two sample *t* test, p < 0.001 and p = 0.002 for ZELIG-TROP and CLM respectively), however the gap model predicted a greater percent difference in average ANPP; a 6.1 % decrease vs. 3.5 % decrease in CLM.

To answer our last research question, what are the differences after increasing disturbance rates in ZELIG-TROP vs. CLM for the Central Amazon, we did find other discrepancies. While the magnitude of change between AGB was similar between the two models, CLM differs greatly from ZELIG-TROP in that it did not captured the inter-

annual variability in carbon stocks, while ZELIG-TROP did (Fig. 5b). Therefore, the demographic forest model captured large fluctuations in annual forest biomass and carbon stocks as a result of either gap dynamics, changes in competition for resources, and/or varying size class and age class structure of the forest. In addition, CLM did not produce pulses of coarse litter in response to tree mortality representative of a hetero-



geneous landscape (Figs. 5d and 7d). While the relative change in AGB was consistent between the two models, there was a large overestimation in the absolute values. With the inclusion of the high disturbance treatment CLM predicted that average AGB net carbon loss was 134 Mg C ha^{-1} (from 269 to 135 Mg C ha^{-1}) vs. 74 Mg C ha^{-1} in ZELIG-TROP.

4 Discussion

4.1 High forest disturbance and long-term impacts

Disturbance is likely to increase in Amazon forests. Since the mid-1970's observed tree mortality and recruitment rates have been increasing in the Amazon (Phillips et al.,

- 2004), and higher than usual mortality rates have also been associated with droughts and strong windstorm events (Nepstad et al., 2007; Chambers et al., 2009; Phillips et al., 2009; Negron-Juarez et al., 2010; Lewis et al., 2011), each of which could increase with human-induced climate change. In addition, reported mortality rates might be underestimated as 9.1–16.9% of tree mortality was missing from plot-based esti-
- ¹⁵ mates in the Amazon, because large gaps in plot-based measurements have previously been excluded (Chambers et al., 2013). We first investigated the impact of continual high disturbance (500 years) in a Central Amazonian forest using a demographic forest model as a benchmark model due to operating at finer scales and having mechanistic mortality algorithms. The elevated disturbance resulted in a decrease in AGB
- ²⁰ by 41.9%, with essentially no change in ANPP (1.0% decrease), and an increase in recruitment rates by 69.1%. As a result of higher proportion of smaller stems (20.7% increase in the 10–30 cm DBH size classes), and decrease in large stems, there was a significant decrease in coarse litter production rate by 8.3%.

We compared empirical data from the higher disturbed "west and south" Amazon plots ("fast dynamics"), to the modeled Central Amazon forest with mirrored tree mortality to evaluate if the models used in this study could predict similar forest dynamics



and characteristics. Only two attributes that are tightly linked with disturbances (i.e. loss in biomass and increase in recruitment) followed the same pattern when shifting from low disturbance to high disturbance. The models were not successful in predicting the shift in basal area growth and specific wood density; forest processes and traits that
 ⁵ are known to have strong gradients across the Amazon Basin (Baker et al., 2004a; Lewis et al., 2004; Phillips et al., 2004). Therefore, results showed that the disturbance regime alone might not explain all of the differences in forest dynamics between the

- two regions, or the models do not accurately capture all disturbance and recovery processes. Furthermore, the net loss in biomass was assumed to be a "false-positive"
 because empirical data has found that variation in wood density drives the reduction in regional-scale AGB (Baker et al., 2004a), while in ZELIG-TROP AGB loss was driven by basal area loss, and in CLM AGB loss was driven by LAI loss. Basal area and LAI were not found to be drivers in AGB loss in observed data. Next we compared the same disturbance scenario in CLM-CN 4.5 and found with regards to AGB response
- to disturbance, CLM performed in a very similar behavior to the gap model. However, CLM did not reproduce the temporal variability in coarse litter inputs, and instead remained constant over time. We also compared the response of large-scale periodic disturbances in the two models, and found that CLM captured similar disturbance and recovery patterns as the gap model.
- It is predicted that disturbances will increase in the future, and this modeling study was unique in that we: (1) showed that the drivers that lead to the net loss in carbon stocks in models are different compared to drivers in empirical datasets, (2) in order to capture regional-scale variation in life history strategies wood density should be included in estimating biomass, (3) predicted that not all differences in tropical forest attributes (e.g., AGB, basal area growth, stem density, and wood density) can be explained by disturbance regimes alone, and also (4) highlighted some inconsistencies
- between a detailed gap model and the global community land surface model used in CESM. It was also unique in that we simulated a *continual* high disturbance rate, in addition to background mortality during each time step. This set it apart from the ma-



jority of disturbance studies that have simulated a one-time total deforestation of the Amazon (Shukla et al., 1990; Henderson-Sellers et al., 1993; Hahmann and Dickinson, 1997; Gedney and Valdes, 2000; Avissar and Werth, 2005). Tropical deforestation and disturbance can occur on multiple and sometimes interacting spatial and tempo-

⁵ ral scales with forest disturbance operating more on a continuum type basis (Asner, 2013; Chambers et al., 2013). By using a dynamic demographic vegetation model, we had the capability to operate at high resolutions and simulate forest succession more accurately.

After applying continual and periodic higher disturbance treatments, we did not ob-¹⁰ serve a continual decrease in forest structure or biomass that lead to a new forest successional trajectory. Instead, we found that the Amazon forest shifted to a new equilibrium state. The outcome of a continual higher disturbance rate generated a stable forest but with less biomass, faster turnover, higher stem density consisting of smaller stems, as well as less emergent species, less ANPP, and less contribution of coarse ¹⁵ litter inputs. From previous studies looking at inventory data we have learned that with increased turnover, there is a change in community composition, less wood density, and when these traits are taken into account there is also less AGB (Baker et al., 2004a). We conclude that including wood density in dynamic vegetation models is

²⁰ disturbances, the interacting affects from CO₂ fertilization should be explored.

4.1.1 CO₂ fertilization and disturbance

Based on observational studies from permanent plots, currently, there is an increase in tree biomass in Amazonian forests by ~ 0.4–0.5 Pt C yr⁻¹ with causal evidence coming from growth fertilization from increasing atmospheric CO₂ (Phillips et al., 1998;

needed. While we have shown that terrestrial biomass will decrease with increased

²⁵ Canadell et al., 2007; Lewis et al., 2009). In a study evaluating the risk of Amazonian forest dieback, Rammig et al. (2010) used rainfall projections from 24 GCMs and a dynamic vegetation model (LPJmL) and predicted that Amazon forest biomass is increasing due to strong CO_2 fertilization effects (3.9 to 6.2 kg C m⁻²), and out ways the



biomass loss due to projected precipitation changes, however larger uncertainties are associated with the effect of CO₂ compared to uncertainties in precipitation. Increasing evidence from an ensemble of updated global climate models are predicting that tropical forests are at a lower risk of forest dieback under climate change, in that they can still retain carbon stocks until 2100 due to fertilization effects of CO₂ (Cox et al., 2013;

Still retain carbon stocks until 2100 due to fertilization effects of CO₂ (Cox et al., 2013; Huntingford et al., 2013), however there is still large uncertainties between models and how tropical forests will respond to interacting effects of increasing CO₂ concentrations, warming temperatures, and changing rainfall patterns (Cox et al., 2013).

Long-term, demographic vegetation models are useful tools at predicting temporal trends related to CO_2 fertilization signals and changes to carbon stocks and fluxes, and the interactions between CO_2 fertilization and varying levels of disturbances are an important next step to evaluate. However, due to the magnitude of forest growth, CO_2 fertilization may not be the only causal factor but instead driven by interacting agents such as biogeography and changing environmental site conditions (Lewis et al.,

¹⁵ 2004; Malhi and Phillips, 2004). Additionally, the role of widespread recovery from past disturbances still needs to be explored as an explanation for biomass accumulation. The interactions between disturbance-recovery, CO₂ fertilization, and/or changes in forest structure might be a more plausible explanation and could have large effects on future forest productivity and growth patterns.

In this study over the period of 100 years there was no significant change in biomass accumulation in both ZELIG-TROP and CLM (Fig. 5b), due to no CO_2 fertilization occurring in the models, and the forest did not act as a carbon sink as predicted by empirical studies across a network of Amazon inventory plots (Phillips et al., 1998, 2004). Upon applying the disturbance treatment, the forest became more stable. With

²⁵ regards to periodic disturbances and sudden tree mortality events both models predicted a negative ΔAGB , -0.15 and -0.46 Mg C ha⁻¹ yr⁻¹ for ZELIG-TROP and CLM respectively, therefore the forest acting as a carbon source (Table 4). CLM predicted a larger decrease in biomass under periodic disturbances, which offsets the current ob-



served biomass accumulation (lower empirical estimates at $0.20-0.39 \text{ Mg C ha}^{-1} \text{ yr}^{-1}$ (Phillips et al., 1998; Chambers and Silver, 2004)).

4.2 Lessons learned from modeling tropical forest disturbance

4.2.1 Model comparison to field data and additional sites

- We found that using a dynamic vegetation gap model that operates at the species 5 level was successful at replicating the Central Amazon forest. ZELIG-TROP has also been validated for the subtropical dry forest of Puerto Rico (Holm et al., 2012), but this is the first application of a dynamic vegetation model of this kind (i.e. gap model) for the Amazon Basin. As a result of using species-specific traits, the values reported by ZELIG-TROP for average basal area, AGB, stem density, LAI, and ANPP were all 10 close to observed values (e.g. ranging from 1.7 to 17.1% difference between ZELIG-TROP and observed field results). Field measurements of AGB from the Central Amazon transects averaged \pm SD: 169 \pm 27.6 Mg C ha⁻¹, and additional field-based measurements from nearby sites in the Central Amazon (FLONA Tapajós plots) range from 132 to 197 Mg C ha⁻¹ (Miller et al., 2003; Keller et al., 2001). ZELIG-TROP pre-15 dicted very similar estimates of AGB: $178 \pm 10.5 \text{ Mg C ha}^{-1}$, therefore model results were within the expected range. From a single-point grid cell, located in the same latitude and longitude coordinates as observational plots, CLM predicted higher levels of AGB (269 Mg C ha⁻¹). In a study comparable to ours, Chambers et al. (2004) found that upon doubling turnover rates in an individual based stand model, forest biomass 20 for a Central Amazon forest decreased by slightly more than 50%. This decrease in forest biomass was similar to the response reported in this study (41.9% and 49.9%). Unlike the Chambers et al. (2004) study, we did not impose an increase in growth rates in the model parameters in conjunction with elevated turnover rates. Instead, annual
- ²⁵ growth rates were determined internally within ZELIG-TROP based on species-specific parameters and environmental conditions.



4.2.2 Growth rates and wood density

Our prediction of average growth rate was higher than field data found in the Central Amazon BDFFP inventory plots (3.1 vs. $1.7 \text{ Mg C ha}^{-1} \text{ yr}^{-1}$, Table 4), but similar to other values found in the Central and Eastern Amazon. For example, using a process-based

- ⁵ model, Hirsch et al. (2004) found above-ground stem growth to be 3.6 Mg C ha⁻¹ yr⁻¹, and field measurements were 2.9 Mg C ha⁻¹ yr⁻¹ at the Seca Floresta site in the Tapajós National Forest (Rice et al., 2004). During the high disturbance treatment, we did not observe an increase in average growth rates compared to the no-disturbance treatment. In fact, there was a slight decrease in annual growth (Table 4, Fig. 4e). This
- ¹⁰ non-significant change in growth rates could have been due to the nonoccurrence of large increases in available light and resources after each additional death, a result of a continual disturbance treatment as opposed to a dramatic disturbance event. Alternatively the Western Amazon plots, counterparts to the high disturbance treatment, did exhibit an increase in growth rates (Fig. 4e). Differences in environmental gradients
- between regions, such as higher total phosphorous, less weathered, and more fertile soils in the Western Amazon (Quesada et al., 2010) could be a stronger controlling factor. In the periodic disturbance treatment, growth and productivity *did* increase directly following each large-scale disturbance (removing 20% of stems). After each pulse disturbance ANPP increased by 14% over the 50-year recovery phase. The change in
- ²⁰ community composition under the high disturbance treatment was also representative of what would be expected (i.e. emergent species decreased by the largest percent in basal area, and canopy and subcanopy species increased), however by not capturing expected changes in wood density the model might be missing some shifts in species composition response to disturbance.
- ²⁵ Wood density is a robust indicator of life history strategies, growth rates and/or successional status of a forest (Whitmore, 1998; Suzuki, 1999; Baker et al., 2004a). In the "west and south" tree species on average have faster growth rates, faster turnover rates, and lower wood density. Upon modeling a Central Amazon forest with distur-



bance rates similar to the "west and south", the higher disturbance did not create a community composition dominated by pioneer species or lower the average wood density, but instead created a forest of less emergent species, more canopy species, and higher wood density. The canopy species that decreased in basal area due to disturbance had an average wood density of $0.67 \,\mathrm{g\,cm^{-3}}$, while canopy species that 5 increased in basal area had an average wood density of 0.70 g cm⁻³, potentially helping to lead to the rise in overall wood density, in addition to the drop in Parkia multijuga discussed in the results section. The growth rate scaling coefficient, G, used in ZELIG-TROP did have a negative relationship with wood density, matching the robust signal observed from inventory data, but was not correlated ($R^2 = 0.13$), leading to another 10 possible explanation of the opposite pattern in wood density shifts with increased disturbance. Wood density is not a main parameterization variable in ZELIG-TROP, and other factors in the gap model (e.g., drought or light tolerances, maximum age, availability of light) could be a stronger driver of community composition shifts over wood 15 density.

It should be noted that wood density is difficult to measure accurately in the field, varies between and within species (Chave et al., 2006), varies within a tree across diameter and from the base of the tree to the top (Nogueira et al., 2005), and the Chambers et al. (2001) AGB model without wood density shows that variation of the data explained by the model is strong ($r^2 = 0.973$). Including wood density in AGB allometric equations is not required, but beneficial for accounting for differences in carbon stocks due to changes in species composition, gradients in soil fertility (Muller-Landau, 2004) as opposed to disturbance regimes, and can be a key variable in greenhouse gas emission mitigation programs.

25 4.2.3 CLM 4.5 vs. dynamic vegetation model

Simulating vegetation demography is beneficial to tracking forest demographics, community shifts, competition, and dynamic changes in carbon stocks and fluxes, and should be considered being incorporated into CLM. The version of CLM used here



does not take into account differences between plant size, plant age, or all biotic and abiotic stressors. Using demography typical of a gap model will account for these missing factors, will aid in capturing annual carbon variability as a result of heterogeneous mortality across the landscape, and can help improve global land surface models. The

- ⁵ exact causes and processes leading to plant mortality are difficult to quantify (Franklin et al., 1987; McDowell et al., 2008, 2011), and additional field research is required in this area, especially in the tropics. However, the gap model approach can quantify the contribution from natural death vs. stress related death vs. disturbance related death under no-disturbance and high-disturbance scenarios.
- ¹⁰ The major differences between the gap model ZELIG-TROP and CLM in response to higher disturbance rates was, (1) the average AGB net carbon loss was 74 Mg C ha⁻¹ in ZELIG-TROP vs. 134 Mg C ha⁻¹ in CLM as a result of doubling background mortal-ity, and (2) that the temporal variability in carbon stock and fluxes was not replicated in CLM. While the absolute values in AGB net carbon loss were different between the two
- ¹⁵ models (Fig. 5b), this was due to the fact that ZELIG-TROP was calibrated for a specific location in the Central Amazon and CLM using initial conditions representative of the entire Amazon basin. As a result of this distinction, relative differences should be used as a comparison tool. The two models were consistent in that they both reached new equilibrium steady-states with both continual and periodic disturbances, and therefore
- the relative change in biomass was analogous between ZELIG-TROP and CLM. Temporal variability in carbon stocks and fluxes over time were also absent from the CLM model due to the inexistence of plant demography (i.e. changes in plant size, structure, and age). Regarding the response to periodic disturbances, the major difference between ZELIG-TROP and CLM was the rapid return to pre-disturbance ANPP levels in CLM and the response to periodic disturbance and pre-disturbance and p
- ²⁵ CLM after each large-scale disturbance event, while in ZELIG-TROP the recovery of ANPP was gradual.

With the inclusion of higher disturbance rates, the two models tested here do predict a $\sim 40-50$ % reduction in carbon stocks, however the driver that lead to this decrease are inconsistent with empirical drivers. Additionally, ZELIG-TROP predicted



lower coarse litter production rates, and gains that exceeded losses. CLM predicted higher coarse litter production rates, and losses that exceeded gains (Table 4), but these differences were minimal. However, these differences that we found in gains minus losses between ZELIG-TROP and CLM can lead to inaccurate predictions of carbon response to increasing disturbance rates in integrated assessment models that use CLM. When taking into account the entire Amazon Basin over many years, this discrepancy can significantly affect predictive outcomes when using the global CLM for mitigation strategies.

4.3 Future directions

- ¹⁰ To constrain the future concentration of CO₂ into the atmosphere, current mitigation strategies rely heavily on tropical forests to maintain, or increase, as a carbon sink. In order to accurately develop and impose mitigation strategy targets, the land components of earth system models need to more accurately simulate plant mortality, coarse litter inputs, carbon fluxes, and accelerated growth processes associated with
- ¹⁵ disturbance-recovery events. CLM 4.5 has been the model of focus here, however multiple versions of the Lund–Potsdam–Jena Dynamic Global Vegetation Model (LPJ-DGVM; Sitch et al., 2003), such as LPJmL and LPJ-GUESS-SPITFIRE, are notable models to evaluate changes to forest biomass in the Amazon (Rammig et al., 2010, others), and changes to stand structure, plant mortality, and emissions due to fire (Thon-
- ²⁰ icke et al., 2010). Cramer et al. (2001) showed the varying range and uncertainties in ecosystem response and magnitude of the terrestrial carbon sink as a function of rising CO_2 and climate change using six DGVMs with varying degrees of functionalities. Including transient changes in vegetation structure and also accounting for changes due to elevated disturbance rates requires models to include vegetation dynamics and suc-
- ²⁵ cession processes, biogeochemical processes, and climate and circulation. With the varying degree of capabilities and functionality within vegetation models this study has benchmarked mortality and disturbance processes in CLM and will benefit the iESM project (Integrated Earth System Model; Jones et al., 2013), which combines CLM with



a fully integrated human system component. To evaluate energy market shifts and carbon policy changes, as a first step we will execute the reduction of the terrestrial carbon sink due to increases in mortality from changes in disturbance regimes, within the fully coupled iESM. In addition, the capability of tropical forests to act as a carbon sink ⁵ with and without the inclusion of disturbances needs to be corrected in the models. which if not could either diminish the effect of mitigation policy, or force more stringent changes in energy infrastructure in order to meet the same climate stabilization targets. Ultimately the contributions to iESM will create the capabilities to test the carbon market and energy market responses to changes in forest mortality and increased disturbances in the Amazon and on a global scale.

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References

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Allen, C. D., Macalady, A. K., Chenchouni, H., Bachelet, D., McDowell, N., Vennetier, M., 25 Kitzberger, T., Rigling, A., Breshears, D. D., Hogg, E. H., Gonzalez, P., Fensham, R.,



Zhang, Z., Castro, J., Demidova, N., Lim, J.-H., Allard, G., Running, S. W., Semerci, A., and Cobb, N.: A global overview of drought and heat-induced tree mortality reveals emerging climate change risks for forests, Forest Ecol. Manag., 259, 660–684, 2010.

 Anderegg, W. R. L., Kane, J. M., and Anderegg, L. D. L.: Consequences of widespread tree mortality triggered by drought and temperature stress, Nature Clim. Change, 3, 30–36, 2013.
 Asner, G.: Geography of forest disturbance, P. Natl. Acad. Sci. USA, 110, 3711–3712, 2013.
 Avissar, R. and Werth, D.: Global hydroclimatological teleconnections resulting from tropical

deforestation, J. Hydrometeorol., 6, 134–145, 2005.

Baker, T. R., Phillips, O. L., Malhi, Y., Almeida, S., Arroyo, L., Di Fiore, A., Killeen, T. J., Lau-

rance, S. G., Laurance, W. F., Lewis, S. L., Lloyd, J., Monteagudo, A., Neill, D. A., Patino, S., Pitman, N. C. A., Macedo Silva, J. N., and Vasquez Martinez, R.: Variation in wood density determines spatial patterns in Amazonian forest biomass, Glob. Change Biol., 10, 545–562, 2004a.

Baker, T. R., Phillips, O. L., Malhi, Y., Almeida, S., Arroyo, L., Di Fiore, A., Erwin, T., Higuchi, N.,

Killeen, T. J., Laurance, S. G., Laurance, W. F., Lewis, S. L., Monteagudo, A., Neill, D. A., Nunez Vargas, P., Pitman, N. C. A., Silva, J. N. M., and Vasquez Martinez, R.: Increasing biomass in Amazonian forest plots, Philos. T. R. Soc. B, 359, 353–365, 2004b.

Bonan, G. B.: Forests and climate change: forcings, feedbacks, and the climate benefits of forests, Science, 320, 1444–1449, 2008.

²⁰ Botkin, D. B., Janak, J. F., and Wallis, J. R.: Some ecological consequences of a computer model of forest growth, J. Ecol., 60, 849–872, 1972.

Brokaw, N. V. L.: Gap-phase regeneration in a tropical forest, Ecology, 66, 682–687, 1985.

- Busing, R. T. and Solomon, A. M.: A Comparison of Forest Survey Data with Forest Dynamics Simulators FORCLIM and ZELIG Along Climatic Gradients in Pacific Northwest, Scientific Investigation Report 2004-5078, US Geological Survey, Reston Virginia, USA, 2004.
- Investigation Report 2004-5078, US Geological Survey, Reston Virginia, USA, 2004. Canadell, J. G., Le Quere, C., Raupach, M. R., Field, C. B., Buitenhuis, E. T., Ciais, P., Conway, T. J., Gillett, N. P., Houghton, R. A., and Marland, G.: Contributions to accelerating atmospheric CO₂ growth from economic activity, carbon intensity, and efficiency of natural sinks, P. Natl. Acad. Sci. USA, 104, 18866–18870, 2007.
- ³⁰ Canham, C. D. and Marks, P. L.: The response of woody plants to disturbance: patterns of establishment and growth, in: The Ecology of Natural Disturbances and Patch Dynamics, Academic Press, New York, NY, 197–216, 1985.



- Cao, L., Bala, G., Caldeira, K., Nemani, R., and Ban-Weiss, G.: Climate response to physiological forcing of carbon dioxide simulated by the coupled Community Atmosphere Model (CAM3.1) and Community Land Model (CLM3.0), Geophys. Res. Lett., 36, L10402, doi:10.1029/2009GL037724, 2009.
- ⁵ Chambers, J. Q. and Silver, W. L.: Some aspects of ecophysiological and biogeochemical responses of tropical forests to atmospheric change, Phil. Trans. R. Soc. Lond. B, 359, 463– 476, 2004.
 - Chambers, J. Q., Higuchi, N., and Schimel, J. P.: Ancient trees in Amazonia, Nature, 391, 135–136, 1998.
- ¹⁰ Chambers, J. Q., Schimel, J. P., and Nobre, A. D.: Respiration from coarse wood litter in central Amazon forests, Biogeochemistry, 52, 115–131, 2001.
 - Chambers, J. Q., Higuchi, N., Teixeira, L. M., dos Santos, J., Laurance, S. G., and Trumbore, S. E.: Response of tree biomass and wood litter to disturbance in a Central Amazon forest, Oecologia, 141, 596–614, 2004.
- ¹⁵ Chambers, J. Q., Robertson, A., Carneiro, V., Lima, A., Smith, M.-L., Plourde, L., and Higuchi, N.: Hyperspectral remote detection of niche partitioning among canopy trees driven by blowdown gap disturbances in the central Amazon, Oecologia, 160, 107–117, 2009.
 - Chambers, J. Q., Negron-Juarez, R. I., Marra, D. M., Di Vittorio, A., Tews, J., Roberts, D., Ribeiro, G. H. P. M., Trumbore, S. E., and Higuchi, N.: The steady-state mosaic of disturbance
- and succession across an old-growth Central Amazon forest landscape, P. Natl. Acad. Sci. USA, 110, 3949–3954, 2013.
 - Clark, D. A.: Detecting tropical forests' responses to global climatic an atmospheric change: current challenges and a way forward, Biotropica, 39, 4–19, 2007.
- Chave, J., Riera, B., and Dubois, M. A.: Estimation of biomass in a neotropical forest of French Guiana: spatial and temporal variability, J. Trop. Ecol., 17, 79–96, 2001.
 - Chave, J., Muller-Landau, Baker, T. R., Easdale, T. A., ter Steege, H., and Webb, C. O.: Regional and phylogenetic variation of wood density across 2456 neotropical tree species, Ecol. Appl., 16, 2356–2367, 2006.

Collins, W. D., Bitz, C. M., Blackmon, M. L., Bonan, G. B., Bretherton, C. S., Carton, J. A.,

³⁰ Chang, P., Doney, S. C., Hack, J. J., Henderson, T. B., Kiehl, J. T., Large, W. G., McKenna, D. S., Santer, B. D., and Smith, R. D.: The Community Climate System Model version 3 (CCSM3), J. Climate, 19, 2122–2143, 2006.



- Cox, P. M., Betts, R. A., Jones, C. D., Spall, S. A., and Totterdell, I. J.: Acceleration of global warming due to carbon-cycle feedbacks in a coupled climate model, Nature, 408, 184–187, 2000.
- Cox, P. M., Betts, R. A., Collins, M., Harris, P. P., Huntingford, C., and Jones, C. D.: Amazo-
- nian forest dieback under climate-carbon cycle projections for the 21st century, Theor. Appl. Climatol., 78, 137–156, 2004.
 - Cox, P. M., Pearson, D., Booth, B. B., Friedlingstein, P., Huntingford, C., Jones, C. D., and Luke, C. M.: Sensitivity of tropical carbon to climate change constrained by carbon dioxide variability, Nature, 494, 341–344, 2013.
- ¹⁰ Cramer, W., Bondeau, A., Woodward, F. I., Prentice, I. C., Betts, R. A., Brovkin, V., Cox, P. M., Fisher, V., Foley, J. A., Friend, A. D., Kucharik, C., Lomas, M. R., Ramankutty, N., Sitch, S., Smith, B., White, A., and Young-Molling, C.: Global response of terrestrial ecosystem structure and function to CO₂ and climate change: results from six dynamic global vegetation models, Glob. Change Biol., 7, 357–373, 2001.
- ¹⁵ Cumming, S. G. and Burton, P. J.: A programmable shell and graphics system for forest stand simulation, Environ. Softw., 8, 219–230, 1993.
 - Dale, V. H., Joyce, L. A., McNulty, S., Neilson, R. P., Ayers, M. P., Flannigan, M. D., Hanson, P. J., Irland, L. C., Lugo, A. E., Peterson, C. J., Simberloff, D., Swanson, F. J., Stocks, B. J., and Wotton, B. M.: Climate change and forest disturbances, Bioscience, 51, 723–734, 2001.
- ²⁰ DeFries, R. S., Houghton, R. A., Hansen, M. C., Field, C. B., Skole, D., and Townshend, J.: Carbon emissions from tropical deforestation and regrowth based on satellite observations for the 1980s and 1990s, P. Natl. Acad. Sci. USA, 99, 14256–14261, 2002.
 - Denslow, J. S.: Tropical rainforest gaps and tree species diversity, Annu. Rev. Ecol. Syst., 18, 431–451, 1987.
- Fearnside, P. M.: Deforestation in Brazilian Amazonia: history, rates, and consequences, Conserv. Biol., 19, 680–688, 2005.
 - Franklin, J. F., Shugart, H. H., and Harmon, M. E.: Tree death as an ecological process, Bio-Science, 37, 550–556, 1987.
- Gedney, N. and Valdes, P. J.: The effect of Amazonian deforestation on the Northern Hemisphere circulation and climate, Geophys. Res. Lett., 27, 3053–3056, 2000.
 - Gent, P. R., Danabasoglu, G., Donner, L. J., Holland, M. M., Hunke, E. C., Jayne, S. R., Lawrence, D. M., Neale, R. B., Rasch, P. J., Vertenstein, M., Worley, P. H., Yang, Z.-L., and



Zhang, M.: The Community Climate System Model version 4, J. Climate, 24, 4973–4991, 2011.

- Hahmann, A. N. and Dickinson, R. E.: RCCM2-BATS model over tropical South America: applications to tropical deforestation, J. Climate, 10, 1944–1963, 1997.
- ⁵ Henderson-Sellers, A., Dickinson, R. E., Durbidge, T. B., Kennedy, P. J., McGuffie, K., and Pitman, A. J.: Tropical deforestation: modelling local to regional-scale climatic change, J. Geophys. Res., 98, 7289–7315, 1993.
 - Hirsch, A. I., Little, W. S., Houghton, R. A., Scott, N. A., and White, J. D.: The net carbon flux due to deforestation and forest re-growth in the Brazilian Amazon: analysis using a process-based model, Glob. Change Biol., 10, 908–924, 2004.
- Holm, J. A., Shugart, H. H., Van Bloem, S. J., and Larocque, G. R.: Gap model development, validation, and application to succession of secondary subtropical dry forests of Puerto Rico, Ecol. Model., 233, 70–82, 2012.

10

Huntingford, C., Fisher, R. A., Mercado, L., Booth, B. B. B., Sitch, S., Harris, P. P., Cox, P. M.,

Jones, C. D., Betts, R. A., Malhi, Y., Harris, G. R., Collins, M., and Moorcroft, P.: Towards quantifying uncertainty in predictions of Amazon "dieback", Philos. T. R. Soc. B, 363, 1857– 1864, 2008.

Huntingford, C., Zelazowski, P., Galbraith, D., Mercado, L. M., Sitch, S., Fisher, R., Lomas, M., Walker, A. P., Jones, C. D., Booth, B. B. B., Malhi, Y., Hemming, D., Kay, G., Good, P.,

- Lewis, S. L., Phillips, O. L., Atkin, O. K., Lloyd, J., Gloor, E., Zaragoza-Castells, J., Meir, P., Betts, R., Harris, P. P., Nobre, C., Marango, J., and Cox, P. M.: Simulated resilience of tropical rainforests to CO₂-induced climate change, Nat. Geosci., 6, 268–273, 2013.
 - Jones, A. D., Collins, W. D., Edmonds, J., Torn, M. S., Janetos, A., Calvin, K. V., Thomson, A., Chini, L. P., Mao, J., Shi, X., Thornton, P., Hurtt, G. C., and Wise, M.: Greenhouse gas policies
- influence climate via direct effects of land use change, J. Climate, 26, 3657–3670, 2013. Keane, R. E., Austin, M., Field, C., Huth, A., Lexer, M. J., Peters, D., Solomon, A., and Wyckoff, P.: Tree mortality in gap models: application to climate change, Climatic Change, 51, 509–540, 2001.

Keller, M., Palace, M., and Hurtt. G.: Biomass estimation in the Tapajos National Forest, Brazil.

³⁰ Examination of sampling and allometric uncertainties, Forest Ecol. Manag., 154, 371–382, 2001.





- Larocque, G. R., Archambault, L., and Delisle, C.: Modelling forest succession in two southeastern Canadian mixedwood ecosystem types using the ZELIG model, Ecol. Model., 199, 350–362, 2006.
- Larocque, G. R., Archambault, L., and Delisle, C.: Development of the gap model ZELIG-CFS
- to predict the dynamics of North American mixed forest types with complex structures, Ecol. Model., 222, 2570–2583, 2011.
 - Laurance, W. F. and Williamson, G. B.: Positive feedbacks among forest fragmentation, drought, and climate change in the Amazon, Conserv. Biol., 15, 1529–1535, 2001.
 - Laurance, W. F., Fearnside, P. M., Laurance, S. G., Delamonica, P., Lovejoy, T. E., Rankin-de
- ¹⁰ Merona, J. M., Chambers, J. Q., and Gascon, C.: Relationship between soils and Amazon forest biomass: a landscape-scale study, Forest Ecol. Manag., 118, 127–138, 1999.
 - Laurance, W. F., Nascimento, H. E. M., Laurance, S. G., Condit, R., D'Angelo, S., and Andrade, A.: Inferred longevity of Amazonian rainforest trees based on a long-term demographic study, Forest Ecol. Manag., 190, 131–143, 2004.
- ¹⁵ Lawrence, D. M., Oleson, K. W., Flanner, M. G., Thornton, P. E., Swenson, S. C., Lawrence, P. J., Zeng, X., Yang, Z.-L., Levis, S., Sakaguchi, K., Bonan, G. B., and Slater, A. G.: Parameterization improvements and functional and structural advances in version 4 of the Community Land Model, J. Adv. Model. Earth Syst., 3, M03001, doi:10.1029/2009GL037724, 2011.
- Le Page, Y., Hurtt, G., Thomson, A. M., Bond-Lamberty, B., Patel, P., Wise, M., Calvin, K., Kyle, P., Clarke, L., Edmonds, J., and Janetos, A.: Sensitivity of climate mitigation strategies to natural disturbances, Environ. Res. Lett., 8, 015018, doi:doi:10.1088/1748-9326/8/1/015018, 2013.

Lewis, S. L. Phillips, O. L., Baker, T. R., Lloyd, J., Malhi, Y., Almeida, S., Higuchi, N., Lau-

rance, W. F., Neill, D. A., Silva, J. N. M., Terborgh, J., Torres Lezama, A., Vasquez Martinez, R., Brown, S., Chave, J., Kuebler, C., Nunez Vargas, P., and Vinceti, B.: Concerted changes in tropical forest structure and dynamics: evidence from 50 South American longterm plots, Philos. T. R. Soc. B, 359, 421–436, 2004.

Lewis, S. L., Lopez-Gonzalez, G., Sonke, B., Affum-Baffoe, K., Baker, T. R., Ojo, L. O.,

Phillips, O. L., Reitsma, J. M., White, L., Comiskey, J. A., Djuikouo, M.-N. K., Ewango, C. E. N., Feldpausch, T. R., Hamilton, A. C., Gloor, M., Hart, T., Hladik, A., Lloyd, J., Lovett, J. C., Makana, J.-R., Malhi, Y., Mbago, F. M., Ndangalasi, H. J., Peacock, J., Peh, K. S.-H., Sheil, D., Sunderland, T., Swaine, M. D., Taplin, J., Talyor, D., Thomas, S. C.,

7756

Votere, R., and Woll, H.: Increasing carbon storage in intact African tropical forests, Nature, 457, 1003–1006, 2009. Lewis, S. L., Brando, P. M., Phillips, O. L., van der Heijden, G. M. F., and Nepstad, D.: The 2010

Amazon drought, Science, 331, 554, doi:10.1126/science.1200807, 2011.

⁵ Lieberman, D., Lieberman, M., Peralta, R., and Hartshorn, G. S.: Mortality patterns and stand turnover rates in a wet tropical forest in Costa Rica, J. Ecol., 73, 915–924, 1985.

Lutz, D. A., Shugart, H. H., and White, M. A.: Sensitivity of Russian forest timber harvest and carbon storage to temperature increase, Forestry, 86, 283–293, 2013.

- Mailly, D., Kimmins, J. P., and Busing, R. T.: Disturbance and succession in a coniferous forest of
- northwestern North America: simulation with DRYADES, a spatial gap model, Ecol. Model., 127, 183–205, 2000.

Malhi, Y. and Phillips, O. L.: Tropical forests and global atmospheric change: a synthesis, Philos. T. R. Soc. B, 359, 549–555, 2004.

Malhi, Y., Phillips, O. L., Lloyd, J., Baker, T., Wright, J., Almeida, S., Arroyo, L., Frederiksen, T.,

Grace, J., Higuchi, N., Killeen, T., Laurance, W. F., Leaño, C., Lewis, S., Meir, P., Monteagudo, A., Neill, D., Núñez Vargas, P., Panfil, S. N., Patiño, S., Pitman, N., Quesada, C. A., Rudas-Ll, A., Salomão, R., Saleska, S., Silva, N., Silveira, M., Sombroek, W. G., Valencia, R., Vásquez Martínez, R., Vieira, I. C. G. and Vinceti, B.: An international network to monitor the structure, composition and dynamics of Amazonian forests (RAINFOR), J. Veg. Sci., 13, 439–450, 2002.

Malhi, Y., Timmons Roberts, J., Betts, R. A., Killeen, T. J., Li, W., and Nobre, C. A.: Climate change, deforestation, and the fate of the Amazon, Science, 319, 169–172, 2008.

- Malhi, Y., Aragão, L. E. O. C., Galbraith, D., Huntingford, C., Fisher, R., Zelazowski, P., Sitch, S., McSweeney, C., and Meir, P.: Exploring the likelihood and mechanism of a climate-change-
- induced dieback of the Amazon rainforest, P. Natl. Acad. Sci. USA, 106, 20610–20615, 2009.
 Martínez-Vilalta, J., Lloret, F., and Breshears, D. D.: Drought-induced forest decline: causes, scope and implications, Biol. Letters, 8, 689–691, 2012.

McDowell, N. G.: Mechanisms linking drought, hydraulics, metabolism, and vegetation mortality, Plant Physiol., 155, 1051–1059, 2011.

³⁰ McDowell, N. G., Pockman, W. T., Allen, C. D., Breshears, D. D., Cobb, N., Kolb, T., Plaut, J., Sperry, J., West, A., Williams, D. G., and Yepez, E. A.: Mechanisms of plant survival and mortality during drought: why do some plants survive while others succumb to drought?, New Phytol., 178, 719–739, 2008.



- Millennium Ecosystem Assessment: Ecosystems and Human Well-Being: Full Report, Island Press, Washington, DC, 2005.
- Miller, S. D., Goulden, M. L., Menton, M. C., Da Rocha, H. R., De Freitas, H. C., Silva Figueira, A. M. E., and Dias de Sousa, C. A.: Biometric and micrometeorological measurements of tropical forest carbon balance, Ecol. Appl., 14, S114–S126, 2003.
- Morton, D. C., DeFries, R. S., Shimabukuro, Y. E., Anderson, L. O., Arai, E., del Bon Espirito-Santo, F., Freitas, R., and Morisette, J.: Cropland expansion changes deforestation dynamics in the southern Brazilian Amazon, P. Natl. Acad. Sci. USA, 103, 14637–14641, 2006.
- Muller-Landau, H. C.; Interspecific and intersite variation in wood specific gravity of tropical trees, Biotropica, 36, 20–32, 2004.
- Nakayama, T.: Shrinkage of shrub forest and recovery of mire ecosystem by river restoration in northern Japan, Forest Ecol. Manag., 256, 1927–1938, 2008.
- Negrón-Juárez, R. I., Chambers, J. Q., Guimaraes, G., Zeng, H., Raupp, C. F. M., Marra, D. M., Ribeiro, G. H. P. M., Saatchi, S. S., Nelson, B. W., and Higuchi, N.: Widespread Amazon forest tree mortality from a single cross-basin squall line event, Geophys. Res. Lett., 37,
- torest tree mortality from a single cross-basin squall line event, Geophys. Res. Lett., 3 L16701, doi:10.1029/2010GL043733, 2010.
 - Nepstad, D. C., Tohver, I. M., Ray, D., Moutinho, P., and Cardinot, G.: Mortality of large trees and lianas following experimental drought in an Amazon forest, Ecology, 88, 2259–2269, 2007.
 Nogueira, E. M., Nelson, B. W., and Fearnside, P. M.: Wood density in dense forest in central
- ²⁰ Amazonia, Brazil, Forest Ecol. Manag., 208, 261–286, 2005.

5

10

- O'Brien, S. T., Hayden, B. P., and Shugart, H. H.: Global change, hurricanes and a tropical forest, Climatic Change, 22, 175–190, 1992.
- Oleson, K. W., Lawrence, D. M., Bonan, G. B., Flanner, M. G., Kluzek, E., Lawrence, P. J., Levis, S., Swenson, S. C., and Thornton, P. E.: Technical description of version 4.0 of the
- ²⁵ Community Land Model (CLM), NCAR Tech. Note NCAR/TN-478+STR, 257 pp., 2010. Overpeck, J. T., Rind, D., and Goldberg, R.: Climate-induced changes in forest disturbance and
 - vegetation, Nature, 343, 51–53, 1990. Pan, Y., Birdsey, R. A., Fang, J., Houghton, R., Kauppi, P. E., Kurz, W. A., Phillips, O. L.,
 - Shvidenko, A., Lewis, S. L., Canadell, J. G., Ciais, P., Jackson, R. B., Pacala, S. W.,
- McGuire, A. D., Piao, S., Rautiainen, A., Sitch, S., and Hayes, D.: A large and persistent carbon sink in the world's forests, Science, 333, 988–993, 2011.



- Phillips, O. L., Malhi, Y., Higuchi, N., Laurance, W. F., Nuñez, P. V., Vásquez, R. M., Laurance, S. G., Ferreira, L. V., Stern, M., Brown, S., and Grace, J.: Changes in the carbon balance of tropical forests: evidence from long-term plots, Science, 282, 439–442, 1998.
- Phillips, O. L., Baker, T. R., Arroyo, L., Higuchi, N., Killeen, T. J., Laurance, W. F., Lewis, S. L., Lloyd, J., Malhi, Y., Monteagudo, A., Neill, D. A., Vargas, P. N., Silva, J. N., Terborgh, J., Martínez, R. V., Alexiades, M., Almeida, S., Brown, S., Chave, J., Comiskey, J. A., Czimczik, C. I., Di Fiore, A., Erwin, T., Kuebler, C., Laurance, S. G., Nascimento, H. E., Olivier, J., Palacios, W., Patiño, S., Pitman, N. C., Quesada, C. A., Saldias, M., Lezama, A. T., and Vinceti, B.: Pattern and process in Amazon tree turnover, 1976–2001, Philos. T. R. Soc. B, 359, 381–407, 2004.
- Phillips, O. L., Aragão, L. E. O. C., Lewis, S. L., Fisher, J. B., Lloyd, J., López-González, G., Malhi, Y., Monteagudo, A., Peacock, J., Quesada, C. A., van der Heijden, G., Almeida, S., Amaral, I., Arroyo, L., Aymard, G., Baker, T. R., Ba'nki, O., Blanc, L., Bonal, D., Brando, P., Chave, J., Alves de Oliveira, A'. C., Dávila Cardozo, N., Czimczik, C. I., Feldpausch, T. R., Freitas, M. A., Gloor, E., Higuchi, N., Jiménez, E., Lloyd, G., Meir, P., Mendoza, C., Morel, A., 15 Neill, D. A., Nepstad, D., Patiño, S., Peñuela, M. C., Prieto, A., Ramrez, F., Schwarz, M., Silva, J., Silveira, M., Sota Thomas, A., ter Steege, H., Stropp, J., Vásquez, R., Zelazowski, P., Alvarez Dávila, E., Andelman, S., Andrade, A., Chao, K., Erwin, T., Di Fiore, A., Honorio, C., Keeling, E., Killeen, H., Laurance, T. J., Peña Cruz, W. F., Pitman, A., Núñez Vargas, N. C. A., Ramrez-Angulo, P., Rudas, H., Salamão, A., Silva, R., Terborgh, N., and Torres-20 Lezama, J. A.: Drought sensitivity of the Amazon rainforest, Science, 323, 1344–1347, 2009. Quesada, C. A., Lloyd, J., Schwarz, M., Patiño, S., Baker, T. R., Czimczik, C., Fyllas, N. M., Martinelli, L., Nardoto, G. B., Schmerler, J., Santos, A. J. B., Hodnett, M. G., Herrera, R., Luizão, F. J., Arneth, A., Lloyd, G., Dezzeo, N., Hilke, I., Kuhlmann, I., Raessler, M., Brand, W. A., Geilmann, H., Moraes Filho, J. O., Carvalho, F. P., Araujo Filho, R. N., 25 Chaves, J. E., Cruz Junior, O. F., Pimentel, T. P., and Paiva, R.: Variations in chemical and physical properties of Amazon forest soils in relation to their genesis, Biogeosciences, 7, 1515–1541, doi:10.5194/bg-7-1515-2010, 2010.
- Rammig, A., Jupp, T., Thonicke, K., Tietjen, B., Heinke, J., Ostberg, S., Lucht, W., Cramer, W., and Cox, P.: Estimating the risk of Amazonian forest dieback, New Phytol., 187, 694–706, 2010.

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- Rice, A. H., Pyle, E. H., Saleska, S. R., Hutyra, L. R., Palace, M., Keller, M., De Camargo, P. B., Portilho, K., Marques, D. F., and Wofsy, S. C.: Carbon balance and vegetation dynamics in an old-growth Amazonian forest, Ecol. Appl., 14, S55–S71, 2004.
- Sanford Jr., R. L.: Fine root biomass under light gap openings in an amazon rain forest, Oecologia, 83, 541–545, 1990.
- Seagle, S. W. and Liang, S.: Application of a forest gap model for prediction of browsing effects on riparian forest succession, Ecol. Model., 144, 213–229, 2001.
- Sitch, S., Smith, B., Prentice, I. C., Arneth, A., Bondeau, A., Cramer, W., Kaplan, J. O., Levis, S., Lucht, W., Sykes, M. T., Thonicke, K., and Venevsky, S.: Evaluation of ecosystem dy- namics,
- ¹⁰ plant geography and terrestrial carbon cycling in the LPJ Dynamic Global Vegetation Model, Glob. Change Biol., 9, 161–185, 2003.

Shugart, H. H.: A Theory of Forest Dynamics, Springer, New York, USA, 1984.

5

25

30

Shugart, H. H.: Forest Gap Models. Vol. 2, The Earth system: biological and ecological dimensions of global environmental change, in: Encyclopedia of Global Environmental Change, edited by: Mooney, H. A. and Canadell, J. C., John Wiley & Sons, 316–323, 2002.

edited by: Mooney, H. A. and Canadell, J. C., John Wiley & Sons, 316–323, 2002. Shugart, H. H. and West, D. C.: Development of an Appalachian deciduous forest succession model and its application to assessment of the impact of the Chestnut Blight, J. Environ. Manage., 5, 161–179, 1977.

Shugart, H. H., Smith, T. M., and Post, W. M.: The potential for application of individual-based

simulation models for assessing the effects of global change, Annu. Rev. Ecol. Syst., 23, 15–38, 1992.

Shukla, J., Nobre, C., and Sellers, P.: Amazon deforestation and climate change, Science, 247, 1322–1325, 1990.

Shuman, J. K., Shugart, H. H., and O'Halloran, T. L.: Sensitivity of Siberian larch forests to climate change, Glob. Change Biol., 17, 2370–2384, 2011.

Smith, J. B. and Tirpak, D. A. (Eds.): The potential effects of global climate change on the US: Appendix D – Forest. Off. Policy, Planning Eval., US Environ. Protection Agency, Washington, DC, 1989.

Smith, T. M. and Urban, D. L.: Scale and the resolution of forest structural pattern, Vegetatio, 74, 143–150, 1988.

Solomon, A. M.: Transient response of forests to CO₂-induced climate change: simulations experiments in eastern North America, Oecologia, 68, 567–579, 1986.

- Suzuki, E.: Diversity in specific gravity and water content of wood among Bornean tropical rainforest trees, Ecol. Res., 14, 211–224, 1999.
- Tribuzy, E. S.: Variacoes da temperature foliar do dossel e o seu efeito na taxa assimilatoria de CO₂ na Amazonia Central, M.S. thesis, Escola Superior de Agricultura "Luiz de Queiroz", Universidade de Sao Paulo, 2005.
- Thonicke, K., Spessa, A., Prentice, I. C., Harrison, S. P., Dong, L., and Carmona-Moreno, C.: The influence of vegetation, fire spread and fire behaviour on biomass burning and trace gas emissions: results from a process-based model, Biogeosciences, 7, 1991–2011, doi:10.5194/bg-7-1991-2010, 2010.
- ¹⁰ Thornton, P. E., Law, B. E., Gholz, H. L., Clark, K. L., Falge, E., Ellsworth, D. S., Goldstein, A. H., Monson, R. K., Hollinger, D., Falk, M., Chen, J., and Sparks, J. P.: Modeling and measuring the effects of disturbance history and climate on carbon and water budgets in evergreen needleleaf forests, Agr. Forest Meteorol., 113, 185–222, 2002.

Thornton, P. E., Lamarque, J.-F., Rosenbloom, N. A., and Mahowald, N. M.: Influence of carbon–nitrogen cycle coupling on land model response to CO₂ fertilization and climate variability, Glob. Biogeochem. Cv., 21, GB4018, doi:10.1029/2006GB002868, 2007.

Uhl, C.: Tree dynamics in a species rich forest tierra firme forest in Amazonia, Venezuela, Acta Cient. Venez., 33, 72–77, 1982.

Urban, D. L.: A Versatile Model to Simulate Forest Pattern: a User's Guide to ZELIG Version 1.

20 0, University of Virginia, Charlottesville, Virginia, 1990.

5

Urban, D. L.: Using model analysis to design monitoring programs for landscape management and impact assessment, Ecol. Appl., 10, 1820–1832, 2000.

Urban, D. L., Bonan, G. B., Smith, T. M., and Shugart, H. H.: Spatial applications of gap models, Forest Ecol. Manag., 42, 95–110, 1991.

²⁵ Urban, D. L., Harmon, M. R., and Halpern, C. B.: Potential response of Pacific northwestern forests to climatic change, effects of stand age and initial composition, Climatic Change, 23, 247–266, 1993.

US DOE: Research Priorities for Tropical Ecosystems Under Climate Change Workshop Report, DOE/SC-0153, US Department of Energy Office of Science, available at: http://science.

³⁰ energy.gov/ber/news-and-resources/, 2012.

Van Daalen, J. C. and Shugart, H. H.: OUTENIQUA – a computer model to simulate succession in the mixed evergreen forests of southern Cape, South Africa, Landscape Ecol., 2, 255–267, 1989.



- Whitmore, T. C.: An Introduction to Tropical Rain Forests, Oxford University Press, NewYork, 1989.
- Yang, X., Thornton, P. E., Ricciuto, D. M., and Post, W. M.: The role of phosphorus dynamics in tropical forests – a modeling study using CLM-CNP, Biogeosciences, 11, 1667–1681, doi:10.5194/bg-11-1667-2014, 2014.

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Table 1. Species-specific allometric and ecological parameters for the 90 tree species used in ZELIG-TROP, representing species found in central Amazonian (Laurance et al., 2004). All species were assigned a probability factor of stress mortality of 0.369, probability factor of natural mortality of 2.813, zone of seed influence of 200, relative seedling establishment rate (RSER) of 0.9, a crown shape value of 4.0, minimum growing degree-day of 5000, and a maximum growing degree-day of 12 229.50.

Species	Growth Form	Age max	DBH max	HT max	G	L	D	Ν	Stock	Wood Density
Anacardium spruceanum	Canopy	175	69.1	3620.4	75.2	2	3	2	0.8	0.46
Aniba canelilla	Canopy	226	37.8	2032.8	38.7	2	3	2	0.5	0.94
Aspidosperma marcgravianum	Emergent	544	90.0	4680.4	30.8	4	3	2	0.5	0.72
Aspidosperma oblongum	Emergent	331	80.0	4173.2	59.5	4	3	2	0.5	0.87
Astronium le-cointei	Canopy	335	50.0	2651.6	34.7	2	3	2	0.5	0.77
Bocageopsis multiflora	Canopy	152	33.1	1794.5	51.3	2	3	2	0.5	0.65
Brosimum acutifolium	Canopy	264	58.3	3072.6	36.2	2	3	2	0.5	0.62
Brosimum guianense	Canopy	477	60.0	3158.8	22.3	2	3	2	0.5	0.89
Brosimum parinarioides	Canopy	483	60.0	3158.8	24.9	2	3	2	0.5	0.62
Brosimum rubescens	Canopy	450	60.0	3158.8	27.1	2	3	2	0.5	0.84
Cariniana micrantha	Emergent	223	80.0	4173.2	76.5	4	3	2	0.5	0.60
Caryocar glabrum	Canopy	527	110.0	5694.8	32.1	2	3	2	0.5	0.71
Casearia arborea	Canopy	91	20.1	1135.1	39.1	2	3	2	0.8	0.57
Casearia sylvestris	Canopy	201	25.5	1409.0	23.7	2	3	2	0.5	0.71
Clarisia racemosa	Canopy	323	80.0	4173.2	44.7	2	3	2	0.5	0.57
Cordia sagotli	Subcanopy	260	26.3	1449.6	14.6	1	3	2	0.8	0.43
Corythophora rimosa	Canopy	235	50.0	2651.6	48.1	2	3	2	0.5	0.81
Couepia longipendula	Canopy	260	46.6	2479.2	37.7	2	3	2	0.5	0.94
Couma macrocarpa	Canopy	233	51.8	2742.9	56.8	2	3	2	0.8	0.50
Couratari stellata	Emergent	592	53.5	2829.1	13.4	4	3	2	0.5	0.63
Dipteryx odorata	Emergent	323	78.4	4092.1	47.7	4	3	2	0.5	0.92
Drypetes variabilis	Subcanopy	252	30.0	1637.2	23.7	1	3	2	0.5	0.73
Duckeodendron cestroides	Emergent	818	140.0	7216.4	18.8	4	3	2	0.5	0.63
Ecclinusa guianensis	Canopy	448	69.7	3650.8	28.5	2	3	2	0.5	0.63
Endopleura uchi	Canopy	223	57.6	3037.1	52.5	2	3	2	0.5	0.79
Eriotheca globosa	Canopy	135	20.1	1135.1	28.3	2	3	2	0.8	0.41
Eschweilera amazoniciformis	Emergent	369	56.1	2961.0	30.5	4	3	2	0.5	0.82
Eschweilera coriacea	Canopy	767	110.0	5694.8	25.7	2	3	2	0.5	0.84
Fusaea longifolia	Subcanopy	413	26.5	1459.7	11.5	1	3	2	0.5	0.74
Glycydendron amazonicum	Canopy	386	44.0	2347.3	23.8	2	3	2	0.5	0.67
Goupia glabra	Emergent	398	100.0	5187.6	44.7	4	3	2	0.5	0.72
Guatteria olivacea	Canopy	54	30.0	1637.2	126.4	2	3	2	0.8	0.47
Gustavia elliptica	Subcanopy	301	24.7	1368.4	16.8	1	3	2	0.5	0.67
Helicostylis tomentosa	Canopy	311	44.7	2382.8	24.0	2	3	2	0.5	0.63
Hevea guianensis	Canopy	288	45.7	2433.5	29.3	2	3	2	0.5	0.55
Inga capitata	Pioneer	162	26.4	1454.6	27.6	3	3	2	0.7	0.60
Inga paraensis	Pioneer	78	40.0	2144.4	95.2	3	3	2	0.7	0.82
Inga splendens	Pioneer	52	38.2	2053.1	157.6	3	3	2	0.7	0.58
Iryanthera juruensis	Subcanopy	569	26.9	1480.0	8.8	1	3	2	0.5	0.66
Iryanthera laevis	Subcanopy	331	27.2	1495.2	15.4	1	3	2	0.5	0.63
Jacaranda copaia	Pioneer	225	30.0	1637.2	21.0	3	3	2	0.8	0.35
Lecythis barnebyi	Subcanopy	336	28.7	1571.3	19.9	1	3	2	0.5	0.82
Lecythis poiteaui	Canopy	747	34.4	1860.4	7.7	2	3	2	0.5	0.80
Lecythis zabucajo	Emergent	628	130.0	6709.2	27.0	4	3	2	0.5	0.86



Table 1. Continued.

Species	Growth Form	Age max	DBH max	HT max	G	L	D	Ν	Stock	Wood Density
Licania apetala	Canopy	199	38.4	2063.3	37.8	2	3	2	0.5	0.76
Licania oblongifolia	Canopy	196	54.2	2864.6	65.7	2	3	2	0.5	0.88
Licania octandra	Subcanopy	339	35.0	1890.8	21.7	1	3	2	0.5	0.81
Licania cannella	Canopy	359	56.5	2981.3	29.0	2	3	2	0.5	0.79
Macrolobium angustifolium	Canopy	335	40.0	2144.4	27.7	2	3	2	0.5	0.68
Manilkara bidentata	Emergent	773	90.0	4680.4	20.6	4	3	2	0.5	0.87
Manilkara huberi	Emergent	349	100.0	5187.6	55.9	4	3	2	0.5	0.93
Maquira sclerophylla	Emergent	420	60.0	3158.8	24.0	4	3	2	0.5	0.53
Mezilaurus itauba	Canopy	684	44.0	2347.3	12.9	2	3	2	0.5	0.74
Micropholis guyanensis	Canopy	248	55.5	2930.6	45.9	2	3	2	0.5	0.66
Micropholis venulosa	Canopy	491	60.0	3158.8	22.9	2	3	2	0.5	0.67
Minquartia guianensis	Emergent	490	70.0	3666.0	30.4	4	3	2	0.5	0.77
Myrciaria floribunda	Subcanopy	490	29.1	1591.6	11.7	1	3	2	0.5	0.77
Davlija strukiju sa	Canopy	195	29.9	1632.1	33.0	2	3	2	0.5	0.61
Parkia multijuga	Emergent	206	119.0	0101.3	101.7	4	3	2	0.8	0.39
Peitogyne paniculata	Canopy	251	40.0	2144.4	28.0	2	3	2	0.5	0.80
Pourouma bicolor	Pioneer	48	29.8	1027.1	124.6	3	3	2	0.8	0.38
Pourouma guianensis	Pioneer	58	31.3	1703.2	01.0	3	3	2	0.8	0.38
Pouteria ambeianinolia	Carlopy	290	36.0	2043.0	21.0	2	3	2	0.5	0.70
Pouteria caimito	Capopy	240	/0.0	2306.7	36.4	2	3	2	0.5	0.78
Pouteria curraniifolia	Canopy	240	43.2	2300.7	25.4	2	2	2	0.5	1 10
Pouteria duianensis	Canopy	720	80.0	4173.2	17.5	2	3	2	0.5	0.94
Pouteria macrophylla	Canopy	387	29.6	1616.9	13.2	2	3	2	0.5	0.86
Pouteria manaosensis	Canopy	981	50.0	2651.6	8.4	2	3	2	0.5	0.64
Pouteria multiflora	Canopy	547	35.5	1916.2	9.5	2	3	2	0.5	0.75
Pouteria oppositifolia	Canopy	277	35.8	1931.4	21.7	2	3	2	0.5	0.65
Pouteria venosa	Canopy	702	45.8	2438.6	10.0	2	3	2	0.5	0.92
Protium altsonii	Emergent	238	70.0	3666.0	56.4	4	3	2	0.5	0.68
Protium decandrum	Canopy	158	32.8	1779.2	40.3	2	3	2	0.5	0.52
Protium heptaphyllum	Canopy	96	26.2	1444.5	60.0	2	3	2	0.8	0.62
Protium tenuifolium	Canopy	170	38.2	2053.1	49.1	2	3	2	0.5	0.57
Qualea paraensis	Emergent	379	70.0	3666.0	31.9	4	3	2	0.5	0.67
Scleronema micranthum	Emergent	353	90.0	4680.4	50.3	4	3	2	0.5	0.60
Sloanea guianensis	Subcanopy	179	28.5	1561.1	26.8	1	3	2	0.5	0.82
Swartzia corrugata	Subcanopy	407	21.1	1185.8	7.7	1	3	2	0.5	1.06
Swartzia recurva	Canopy	177	38.4	2063.3	45.5	2	3	2	0.5	0.97
Swartzia ulei	Canopy	293	50.0	2651.6	39.1	2	3	2	0.5	1.00
Tachigali paniculata	Canopy	91	27.7	1520.6	60.1	2	3	2	0.8	0.56
Tapirira guianensis	Canopy	54	41.6	2225.6	188.0	2	3	2	0.8	0.45
Tetragastris panamensis	Canopy	320	38.4	2063.3	25.1	2	3	2	0.5	0.72
Vantanea parviflora	Canopy	205	69.6	3645.7	65.1	2	3	2	0.5	0.84
Virola calophylla	Subcanopy	293	30.8	1677.8	18.6	3	2	2	0.8	0.51
Virola multinervia	Canopy	373	32.0	1738.7	14.0	2	3	2	0.8	0.45
Virola sebifera	Canopy	161	30.2	1647.4	44.4	2	3	2	0.8	0.46
Vochysia obidensis	Canopy	92	47.4	2519.7	109.1	2	3	2	0.8	0.50

Key: Age max, maximum age for the species (yr); DBH max, maximum diameter at breast height (cm); HT max, maximum height (m); G, growth rate scaling coefficient; Light (L), Drought (D), Nutrient (N): light/shade tolerance class, maximum drought tolerance class, and soil nutrient tolerance class; Stock, regeneration stocking, wood density; (full parameter explanation found in original ZELIG paper: Urban, 1990).



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Table 2. Environmental parameters used in ZELIG-TROP for the central Amazon basin. Values reported in a range were monthly low and high averages.

Lat./Long./ Alt. (m)	Plot Area (m ²)	Mean monthly temperature (°C)	Mean monthly precipitation (cm)	Soil field capacity (cm) ^a	Soil wilting point (cm) ^a	Relative direct and diffuse solar radiation (%)
-2.3/-60.0/ 100.0	400.0	25.18–27.47	8.01–45.16	52.0	32.9	0.6/0.4

^a Lawrence et al. (1999).

Table 3. Averages (and standard deviations) of five forest attributes for the observed values recorded from sites near Manaus, Brazil, averaged over 5 ha, and the modeled ZELIG-TROP results. ZELIG-TROP results are averaged for the final 100 years, after an initial spin up of 400 years. The remaining values correspond to the percent differences between the observed and simulated values, and the minimum and maximum range of a ZELIG-TROP simulation.

	Avg. Basal Area (m ² ha ⁻¹)	Avg. Biomass (Mg C ha ⁻¹)	Avg. Stem Density (ha ⁻¹)	Avg. LAI (Mg C ha ⁻¹ yr ⁻¹)	Avg. ANPP
Empirical Data	30.06 (6.61)	169.84 (27.60)	656 (22)	5.7 (0.50)	6.5
ZELIG-TROP	32.96 (1.22)	178.38 (10.53)	574 (70)	5.8 (0.24)	5.4 (0.22)
Percent Diff. (%)	9.66	5.03	-12.49	1.75	-17.08
ZELIG-TROP min./max.	31.14/35.97	167.97/189.26	472/688	5.26/6.48	5.08/5.92



Table 4. Comparison of empirical data and stand model data from Chambers et al. (2004) unless otherwise noted, ZELIG-TROP pre- and post-disturbance treatments, and CLM preand post-disturbance treatments for the pool of carbon in live trees, and the annual flux of carbon from stem growth, coarse litter production rates from mortality, ANPP; and recruitment rate of stems, mean DBH, and average Δ AGB.

Positive = sink	Live Trees (MgCha ⁻¹)	Growth (Mg C ha ⁻¹ yr ⁻¹)	Coarse Litter (MgCha ⁻¹ yr ⁻¹)	$\begin{array}{c} \text{ANPP} \\ \text{(MgCha}^{-1} \text{ yr}^{-1} \text{)} \end{array}$	Recruitment (% yr ⁻¹)	Mean DBH (cm)	AGB change (MgCha ⁻¹ yr ⁻¹)
4	((a - a 5			(go.i.u j.)
Empirical	156	1.70	-2.10	6.50°	1.38°	21.1	NA
Stand Model ⁴	160	1.60	-1.70	6.60	NA	20.4	NA
ZELIG-TROP ¹	178	3.09	-3.03	5.39	2.33	22.3	0.02
ZELIG-TROP ²	104	2.89	-2.78	5.35	3.94	18.3	0.01
ZELIG-TROP ³	138	3.29	-3.49	5.06	3.41	26.9	-0.15
CLM-CN ¹	269	4.88	-4.82	7.81	NA	NA	0.04
CLM-CN ²	135	4.91	-4.93	7.83	NA	NA	0.00
CLM-CN ³	230	4.71	-4.95	7.54	NA	NA	-0.46
ZELIG Diff. ^{1,2}	-74	-0.20	0.25	-0.04	1.61	-4.0	0.01
ZELIG Diff. ^{1,3}	-40	0.20	-0.46	-0.33	1.08	4.6	-0.17
CLM Diff. ^{1,2}	-134	0.03	-0.11	0.02	NA	NA	-0.04
CLM Diff. ^{1,3}	-39	-0.17	-0.15	-0.27	NA	NA	-0.50

¹ = No Disturbance,

² = High Disturbance,

³ = Periodic Disturbance,

⁴ Chambers et al. (2004),

⁵ Chambers et al. (2001),

⁶ Phillips et al. (2004).





Figure 1. Comparison between observed field data from "transects" in Central Amazon, ZELIG-TROP model data from no-disturbance scenario, and ZELIG-TROP model data from highdisturbance treatment. (A) Average stem density (stems ha^{-1}) and SD by DBH (cm) size class, (B) stem density simulated over 500 years, (C) average above-ground biomass (Mg ha^{-1}) and SD by DBH (cm) size class, and (D) above-ground biomass simulated over 500 years. Average results and *t* test between two model results taken once the model reached a steady-state, or the final 100 years of simulation.











Figure 3. Comparison of relative frequency of annual mortality rates (% stems yr⁻¹) from observed data, ZELIG-TROP no-disturbance, and ZELIG-TROP high-disturbance model data after the disturbance treatment. (Observed data: Chambers et al., 2004.)





Figure 4. Comparison between "central and east" Amazon ("slow dynamics") and "west and south" Amazon ("fast dynamics") between the empirical (RAINFOR dataset) and modeled ZELIG-TROP results for average (**A**) above-ground biomass (AGB, Mg C ha⁻¹ yr⁻¹) with the observed dataset either including or not including wood density in the Chambers et al. (2001) allometric equation, (**B**) recruitment rate (% yr⁻¹), (**C**) average wood density (g cm⁻³), (**D**) stem density (stems ha⁻¹), and (**E**) stand-level basal area (BA) growth rate (m² ha⁻¹ yr⁻¹), with 95% CIs bars included. Different lower case letters represent significantly different values using Tukey's multiple comparison, following a one-way ANOVA.





Figure 5. CLM-CN model evaluation and comparisons to ZELIG-TROP for a no-disturbance scenario and a high disturbance treatment: **(A)** ANPP, **(B)** above-ground biomass, **(C)** stem growth, **(D)** coarse litter production rates, all measured in Mg C ha⁻¹, and **(E)** basal area from ZELIG-TROP and observed data reported by Baker et al. (2004a), and **(F)** leaf area index (LAI) from CLM-CN4.5 and observed data reported by McWilliams et al. (1993) and Malhi et al. (2013). Statistical significance test in all panels are two-sample Student's *t* test between the no-disturbance and high disturbance treatments, separately for each model.











Figure 7. CLM-CN model evaluation and comparisons to ZELIG-TROP for a periodic disturbance treatment: **(A)** ANPP, **(B)** stem growth, **(C)** aboveground biomass (AGB), and **(D)** coarse litter production rates, all measured in Mg C ha⁻¹. Statistical significance test in all panels are two-sample Student's *t* test between the no-disturbance and high disturbance treatments, separately for each model.

