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Ecosystem heterogeneity determines the ecological resilience of the Amazon to climate change

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Amazon forests, which store ~50% of tropical forest carbon and play a vital role in global water, energy, and carbon cycling, are predicted to experience both longer and more intense dry seasons by the end of the 21st century. However, the climate sensitivity of this ecosystem remains uncertain: several studies have predicted large-scale die-back of the Amazon, whereas several more recent studies predict that the biome will remain largely intact. Combining remote-sensing and ground-based observations with a size- and age-structured terrestrial ecosystem model, we explore the sensitivity and ecological resilience of these forests to changes in climate. We demonstrate that water stress operating at the scale of individual plants, combined with spatial variation in soil texture, explains observed patterns of variation in ecosystem biomass, composition, and dynamics across the region, and strongly influences the ecosystem's resilience to changes in dry season length. Specifically, our analysis suggests that in contrast to existing predictions of either stability or catastrophic biomass loss, the Amazon forest's response to a drying regional climate is likely to be an immediate, graded, heterogeneous transition from high-biomass moist forests to transitional dry forests and woody savannah-like states. Fire, logging, and other anthropogenic disturbances may, however, exacerbate these climate change-induced ecosystem transitions.

Amazon forests | biomass | ecological resilience | climate change | ecosystem heterogeneity

Amazonia consists of 815 million ha of rainforest, transitional forest, and tropical savannahs; stores approximately half of tropical forest carbon (1); and plays a vital role in global water, energy, and carbon cycling (2). Although uncertainties in climate predictions for the region remain large (3), recent analyses imply that significant portions of the basin will experience both longer and more intense dry seasons by the end of the 21st century (3–6). There is particular concern about southern Amazonian forests that experience longer dry seasons than forests in central and western Amazonia (3) and where a trend of increasing dry season length (DSL) and intensity has already been observed (7). Despite the importance of this region for regional and global climate, the climate sensitivity of the Amazon forests remains uncertain: model predictions range from a large-scale die-back of the Amazon (8, 9) to predictions that the biome will remain largely intact, and may even increase in biomass (10–12). Although some of these differences can be attributed to differences in the predicted future climate forcing of the region (13, 14), accurate predictions of how changes in climate will affect Amazonian forests also rely on an accurate characterization of how the ecosystem is affected by a given change in climate forcing. In this study, we examine the

climate sensitivity of the Amazon ecosystem, focusing on the mechanisms underpinning changes in forest dynamics and their implications for the timing and nature of basin-wide shifts in biomass in response to a drying climate.

Variation in forest biomass across the Amazon basin (15–17) has been shown to correlate with DSL (16–18) (Fig. 1), soil texture (16), shifts in stem turnover rate (19), and forest composition (20). In general, high-biomass moist tropical forests occur where DSL, defined here as the number of months in which precipitation is <100 mm (6, 9), is short, and low-biomass, savannah-like ecosystems are primarily found when DSLs are long (Fig. 1A). In addition, a significant relationship is observed between regional-scale spatial heterogeneity in above-ground biomass (AGB > 2 kg of carbon per square meter) and DSL, with drier places having greater spatial heterogeneity: This pattern is seen both at the scale of 1° (Fig. 1C; $r^2 = 0.88$, $P < 0.01$ for remote sensing-based AGB estimates) and at smaller spatial scales (*SI Appendix, section S1*). In other words, in moist areas,

Significance

Understanding how changes in climate will affect terrestrial ecosystems is particularly important in tropical forest regions, which store large amounts of carbon and exert important feedbacks onto regional and global climates. By combining multiple types of observations with a state-of-the-art terrestrial ecosystem model, we demonstrate that the sensitivity of tropical forests to changes in climate is dependent on the length of the dry season and soil type, but also, importantly, on the dynamics of individual-level competition within plant canopies. These interactions result in ecosystems that are more sensitive to changes in climate than has been predicted by traditional models but that transition from one ecosystem type to another in a continuous, non-tipping-point manner.

Author contributions: N.M.L. and P.R.M. designed research; N.M.L., K.Z., M.L., and P.R.M. performed research; M.L. contributed analytic tools; N.M.L., K.Z., and M.L. analyzed data; N.M.L. and P.R.M. wrote the paper; O.L.P. and S.L.L. assisted in the presentation of the RAINFOR results; A.B. provided satellite-derived estimates of regional above-ground biomass used in the analysis; O.L.P., S.L.L., E.A.-D., A.C.S.d.A., R.J.W.B., T.L.E., T.R.F., A.L.M.M., P.N.V., A.P., J.E.S.-E., and Y.M. provided field-based estimates of above-ground biomass used in the analysis.

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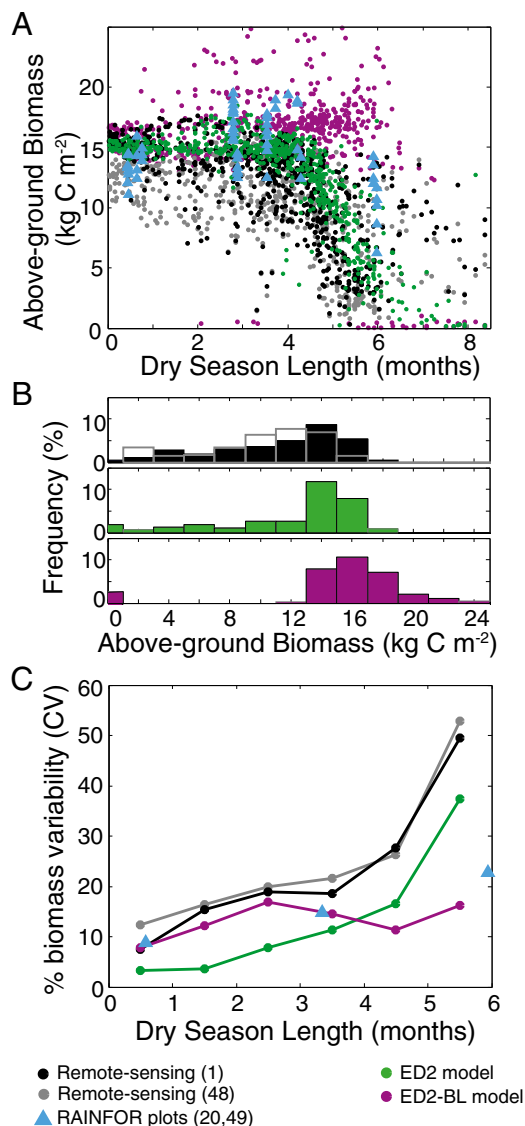


Fig. 1. (A) Change in AGB with DSL for remote sensing-based estimates (black and gray circles), ground-based plot measurements (blue triangles), ED2 model output (green circles), and ED2-BL model output (purple circles). (B) Distribution of AGB in the observations and the two models. (C) Change in the percentage of biomass variability, with the coefficient of variation (CV) defined as $1\sigma/\text{mean}$. Results are for undisturbed primary vegetation forests. Data are from Baccini et al. (1), Saatchi et al. (48), and Baker et al. (20, 49).

where DSL is short, forests have relatively homogeneous levels of AGB, whereas in drier areas, forests are increasingly heterogeneous. As we show below, this observed heterogeneity in response to increasing DSL has important implications for how the structure, composition, and dynamics of Amazon forests will be affected by changes in climate.

The Ecosystem Demography Biosphere (ED2) model, a process-based terrestrial biosphere model that represents individual plant-level dynamics, including competition for light and water (21, 22), was used to investigate the impact of ecosystem heterogeneity on the Amazon forest's ecological resilience to climate perturbations (*SI Appendix, section S3*). Here, the term "ecological resilience" is used to describe the ability of a forest to maintain fundamental characteristics, such as carbon pools, composition, and structure, despite changes in climate (23). ED2 model simulations for the Amazon region, forced with a regional

climate dataset derived from in situ measurements and remote-sensing observations, correctly reproduce the observed pattern of AGB variability as a function of DSL and soil texture (Fig. 1 and *SI Appendix, section S4*). In addition, ED2 model simulations for sites with detailed ground-based soil texture, forest structure, turnover, and composition measurements are also consistent with the observed patterns of variation in these quantities (*SI Appendix, section S4*).

An ensemble of model simulations with varying soil texture was used to investigate the mechanisms that underpin the observed variable response to increasing DSL (*SI Appendix, section S3*). In the model, individual plant productivity is modified by a measure of plant water stress (γ_{WS}) that integrates soil texture, precipitation, and plant transpiration demand such that, as γ_{WS} increases, the plants close their stomata to reduce water loss. In the ED2 ensemble simulations, plot biomass is highly correlated with the average γ_{WS} for the forested sites (defined here as $\text{AGB} > 3$ kg of carbon per square meter) (Fig. 2C; $r^2 = 0.96\text{--}0.99$, $P < 0.01$; *SI Appendix, section S5*). Associated with changes in AGB that occur as water stress increases are correlated changes in the productivity and composition of the plant canopy (*SI Appendix, section S6*).

The important role that water stress operating at the scale of individual plants plays in generating these responses is illustrated by comparing the native ED2 model predictions with output from a horizontally and vertically averaged version of the model (ED2-BL), analogous to a conventional "big leaf" terrestrial biosphere model that represents the canopy in an aggregated manner (*SI Appendix, section S3*). In the ED2-BL simulations, there is no significant relationship between the spatial heterogeneity of forested sites and DSL (Fig. 1A and C; $r^2 = 0.24$, $P = 0.32$). The absence of individual-level plant dynamics in the ED2-BL model results in a markedly different response to variations in soil texture and DSL than the native model formulation: Biomass initially declines as a function of increasing water stress, but a tipping point is then reached, beyond which the high-biomass forest is no longer stable and is replaced by a low-biomass savannah (Fig. 2). The result is a bimodal distribution of AGB across the basin in the ED2-BL model simulations, in contrast to the continuous distribution seen in the native model formulation and the observations (Fig. 1B). This response mirrors the response seen in other big-leaf-type ecosystem models (9). In native ED2 simulations, when water stress is prevented from influencing plant productivity, DSL and soil texture no longer have an impact on AGB (*SI Appendix, section S5 and Fig. S5*). Taken together, these simulations indicate that the driving mechanism behind the observed heterogeneous response to changes in DSL is the differential performance of individuals within the canopy to declining water availability, and how this response is modulated by soils with different hydrological properties. Specifically, the size and age structure of the ED2 plant canopy results in individuals' differential access to both light and soil water, influencing the dynamics of individual plant growth and mortality (*SI Appendix, section S6*). Due to the nonlinear nature of functions governing plant growth, mortality, and recruitment, this heterogeneity results in a more continuous, graded response to changes in water stress than the big leaf (ED2-BL) formulation (Fig. 2). The consequence of this heterogeneity in plant-level responses to changes in soil moisture is that soil texture is likely to become increasingly important for controlling AGB as DSL increases. Soil fertility gradients also influence Amazonian AGB (16–18); however, as we show in *SI Appendix, section S2*, they do not account for the observed regional-scale pattern of increasing biomass heterogeneity with increasing DSL.

The ED2 biosphere model was used to investigate the expected patterns and time scales of Amazonian ecosystem response to a 1- to 4-month change in DSL over the 21st century (6). Earlier analyses have suggested that by accurately representing the dynamics of individual trees, models such as ED2 that incorporate

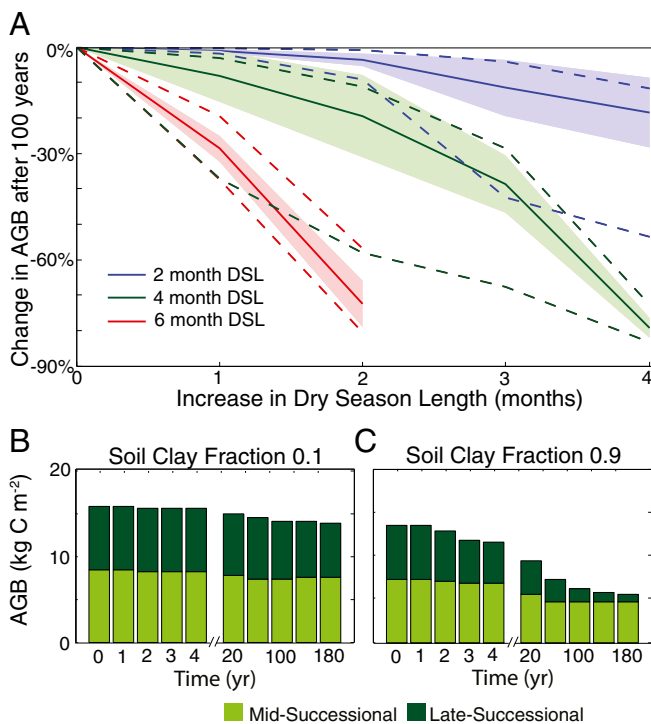


Fig. 3. Predicted response of forest AGB and composition to an increase in DSL. (A) Change in AGB after 100 y as a result of increasing DSL for forests with historic DSLs of 2, 4, and 6 months for the range of soil textures simulated in the ensemble model simulations ($n = 30$). The magnitude of the change in AGB is influenced by soil clay fraction: The mean (solid line), 1 σ deviation (shaded region), and minimum and maximum values (dashed lines) are shown. (B and C) Bar plots illustrating the impact of a 2-month increase in dry season (from 4 to 6 months) on a forest situated on a low clay content soil and a forest situated on a high clay content soil. The color of the bars indicates the contribution of mid- and late-successional trees, illustrating the shift in composition caused by the increase in DSL.

Although regional patterns of Amazonian AGB are complex, reflecting the impact of multiple factors, our results suggest that plant-level responses to soil texture heterogeneity and changes in DSL are important in explaining the observed basin-wide pattern of variation in Amazonian AGB, providing a mechanistic explanation for the observed correlations between DSL, AGB, and changes in stand structure and composition (16, 17). These conclusions may also apply to African and Asian tropical forests; however, important differences exist in the future climate predictions for these regions (38) and their soil edaphic and nutrient characteristics and historical fire regimes (39–41).

The response of forests to changes in their climate forcing is an emergent ecosystem-level response that is ultimately driven by individual trees responding to changes in their local environments. Nonlinearities in the performance of individual plants, such as their rates of photosynthetic assimilation and mortality, as environmental conditions change imply that terrestrial biosphere models need to represent these differential responses of individuals to capture emergent ecosystem properties accurately (42). This analysis demonstrates that the conventional approach of modeling average plants in average environments within climato-

logical grid cells underestimates the direct, near-term response of tropical forests to climatological change but overestimates the direct impacts of larger scale changes in forcing. Consequently, accurate predictions for the timing and nature of forest responses to changes in climate require consideration of how climate and soils affect the performance of individuals within plant canopies. As we have shown here, models that incorporate plant-level dynamics are able to characterize extant patterns of variation in the structure, composition, and dynamics of Amazonian ecosystems more accurately, and accounting for these patterns has important implications for the sensitivity and ecological resilience of Amazon forests to different levels of climatological perturbation.

Methods

The ED2 and ED2-BL model is an integrated terrestrial biosphere model that incorporates land-surface biophysics, vegetation dynamics, hydrology, and soil carbon biogeochemistry, and it uses a size- and age-structured system of partial differential equations to approximate the individual-level dynamics of plant canopies (21, 22, 43). The horizontally and vertically averaged ED2-BL model represents exactly the same biophysical and biogeochemical processes as the ED2 model, but the size- and age-structured canopy is replaced with a horizontally and vertically averaged canopy akin to those canopies used by conventional terrestrial biosphere models. Additional information on the model formulation is provided in *SI Appendix, section S3*.

The ED2 and ED2-BL models were run for the entire Amazon basin forced with a rescaled National Centers for Environmental Prediction (NCEP) reanalysis product (44) and observation-based soil maps (45, 46) at 1° resolution, and with increasing atmospheric CO₂ (47) (*SI Appendix, section S3*). The model results were compared against remote-sensing estimates (1, 48) aggregated to the same resolution as the model simulations. Plot-based observations were made on the scale of 0.4–10 ha (20), and were compared against model simulations forced with site-specific inputs (*SI Appendix, Table S4*).

The water stress factor (γ_{WS}) was used in both the ED2 and ED2-BL models to scale photosynthesis in response to water stress. The γ_{WS} was calculated for each individual (i) as:

$$\gamma_{WS}^i = 1 - \frac{K \int_0^{RD} (\theta(z) - \theta_{WP}) dz C_{root}^i}{K \int_0^{RD} (\theta(z) - \theta_{WP}) dz C_{root}^i + T_{max}^i} \quad [1]$$

where γ_{WS} ranges from 0 (unstressed) to 1 (stressed). T_{max}^i is the maximum transpiration (kg of water per year) for individual i , C_{root}^i is the root biomass (kg of carbon) for individual i , $\theta(z)$ is the soil moisture (kg of water per cubic meter) at soil depth z , K is the root conductance (m²·kg of carbon per year), and θ_{WP} is the soil wilting point (kg of water per cubic meter). The available soil water (kg of water per square meter), $\theta(z) - \theta_{WP}$, is integrated over the rooting depth (RD) of the individual.

Spatial heterogeneity (σ/μ) was calculated over 1-month DSL intervals for model simulations and remote-sensing based estimates. These calculations were done at 1-ha resolution for plot-based observations and model simulations for these locations, at 500-m and 1-km resolution for remote-sensing based estimates, and at 1° resolution for the regional model simulations and remote-sensing based estimates. Due to the relatively low number of plots, the spatial heterogeneity of the plot-based observations was calculated for three DSL categories: 0–2 months, 2–5 months, and 5–8 months.

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