Linking hydraulic traits to tropical forest function in a size-structured and trait-driven model (TFS v.1-Hydro)

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Abstract. Forest ecosystem models based on heuristic water stress functions poorly predict tropical forest response to drought partly because they do not capture the diversity of hydraulic traits (including variation in tree size) observed in tropical forests. We developed a continuous porous media approach to modeling plant hydraulics in which all parameters of the constitutive equations are biologically interpretable and measurable plant hydraulic traits (e.g., turgor loss point $\pi_{tlp}$, bulk elastic modulus $\varepsilon$, hydraulic capacitance $C_f$, xylem hydraulic conductivity $k_{s,max}$, water potential at 50% loss of conductivity for both xylem ($P_{50,x}$) and stomata ($P_{50,gs}$), and the leaf: sapwood area ratio $A_l:A_s$). We embedded this plant hydraulics model within a trait forest simulator (TFS) that models light environments of individual trees and their upper boundary conditions (transpiration), as well as providing a means for parameterizing variation in hydraulic traits among individuals. We synthesized literature and existing databases to parameterize all hydraulic traits as a function of stem and leaf traits, including wood density (WD), leaf mass per area (LMA), and photosynthetic capacity ($A_{max}$), and evaluated the coupled model (called TFS v.1-Hydro) predictions, against observed diurnal and seasonal variability in stem and leaf water potential as well as stand-scaled sap flux.

Our hydraulic trait synthesis revealed coordination among leaf and xylem hydraulic traits and statistically significant relationships of most hydraulic traits with more easily measured plant traits. Using the most informative empirical trait–trait relationships derived from this synthesis, TFS v.1-Hydro successfully captured individual variation in leaf and stem water potential due to increasing tree size and light environment, with model representation of hydraulic architecture and plant traits exerting primary and secondary controls, respectively, on the fidelity of model predictions. The plant hydraulics model made substantial improvements to simulations of total ecosystem transpiration. Remaining uncertainties and limitations of the trait paradigm for plant hydraulics modeling are highlighted.
1 Introduction

Tropical forests harbor great biodiversity (Myers et al., 2000; ter Steege et al., 2013) and play an important role in regulating regional and global climate (Gash and Nobre, 1997; Silva Dias et al., 2002). However, climate change is inducing changes to the hydrological regime of tropical forests (Feng et al., 2013; Fu et al., 2013; Gloor et al., 2013), with some consensus for a projected increase in drought frequency over the coming century via an intensification of precipitation seasonality (Joetzjer et al., 2013; Boisier et al., 2015), an increase in El Niño events (Cai et al., 2014), and chronically rising atmospheric moisture demand (McDowell and Allen, 2015), even as the directional change of total precipitation remains highly uncertain (Solomon, 2007). Therefore, because of their intrinsic value and strong coupling to the regional and global climate system, it is of paramount importance to have a predictive capability of tropical forest response to changes in water availability (Kumagai and Porporato, 2012; Oliveira et al., 2014; Meir et al., 2015).

Evaluations of many coarse-scale forest ecosystem models in the species-rich tropics indicate a poor predictive ability of these models to simulate tropical forest response to drought (Galbraith et al., 2010; Powell et al., 2013; Joetzjer et al., 2014; Rowland et al., 2015b). While most of these models treat soil water fluxes mechanistically, soil–root and internal plant water fluxes are not mechanistically modeled; consequently, dynamic changes in plant moisture stress in these models are driven by changes in soil water status alone (discussed in Feddes et al., 1978; Egea et al., 2011; Xu et al., 2013; Verhoef and Egea, 2014). Additionally, water stress in many models includes relatively few plant functional type (PFT)-dependent parameters; thus, models are largely ignorant of plant traits that may control response to moisture. In contrast, a mechanistic treatment of plant hydraulics allows water stress to be driven by changes in leaf water status (Sperry and Love, 2015). Coupled with site-specific parameterization, the plant hydraulics approach enables high-fidelity simulation of tropical forest response to moisture (Fish et al., 2006, 2007).

While a range of approaches exist for modeling plant hydraulics at fine scales (i.e., individual trees), all involve an extension of Darcy’s law (Darcy, 1856) from the soil domain to include plants as well. Darcy’s law states that water flux anywhere in the soil–plant continuum is proportional to the product of soil or plant hydraulic conductivity and a gradient in water potential. In order for these models to capture drought response, hydraulic conductivity within the soil–plant continuum must dynamically respond to changes in moisture. Three main approaches are distinguished in terms of how they represent the impact of declines in water potential on tissue water content and xylem hydraulic conductivity. A first class of models is the simplest and simulates moisture sensitivity of soil–root conductance but not xylem hydraulic conductance (Jarvis et al., 1981; Williams et al., 1996; Ogée et al., 2003; Alton et al., 2009; Bonan et al., 2014). This approach has proven useful for modeling the effects experimental drought in tropical forests (Williams et al., 1998; Fisher et al., 2006, 2007), but it remains unclear whether this approach misattributes drought effects occurring within trees to the soil; therefore, a second class of models implements variable xylem conductivity with xylem water potential (Williams et al., 2001; Hickler et al., 2006; Domec et al., 2012; Duursma and Medlyn, 2012; Xu et al., 2016). To simplify computational load, these two approaches do not explicitly track dynamic changes in the volume of plant water storage. Instead, a constant ratio of change in stored water per unit change in water potential, or stem hydraulic capacitance, is assumed, which may overestimate the buffering capacity of tree-stored water under extreme drought conditions when small relative declines in stored water induce very large declines in water potential. An additional consequence of the design of these models is the inability to represent the bidirectional flow of water at the root–soil interface. Reverse flow of water from roots into soil is an important process in root hydraulic distribution (Oliveira et al., 2005), and may also mediate time to desiccation under drought (North and Nobel, 1997).

A convenient way to address these issues is in a third class of models (hereafter the “continuous porous media approach”), which simply extend the modeled mass balance of water from the soil domain into the plant by relating simulated changes in water content to water potential (and vice versa) everywhere within the plant–soil continuum (Edwards et al., 1986, Arnborg et al., 1993; Sperry et al., 1998; Kumagai, 2001; Bohrer et al., 2005; Mackay et al., 2015; Mirfendersesi et al., 2016). While more computationally complex, the continuous porous media approach offers two main advantages in addition to addressing the issue of plant water storage and bidirectional root flow. First, the coupled plant–soil system is represented by a single mass balance equation, such that root water uptake or loss simply emerges from the solution of this equation, and does not need to be ascribed post hoc as is the case in the first two approaches. Second, this approach relies on an explicit description of the relationship between water content and water potential in plant xylem (the pressure–volume, or PV curve), analogous to the water retention curves used in soil physics. As we will show, there is a wealth of information on PV hydraulic traits for leaves and, to a lesser degree, stems in tropical forests. Implementing PV curves in the model greatly increases the scope of data with which the model can be parameterized.

In this paper, we develop a continuous porous media approach intended for application at specific sites in the tropics to explore dynamics of water fluxes from hourly to seasonal timescales and at spatial scales ranging from individual trees to the stand-level scale. This intermediate-scale approach is a model test bed meant to inform implementations of plant hydraulics in coarse-scale forest ecosystem models.

Model parameterization leads us to the challenge of how to represent variation in plant hydraulic traits governing mois-
ture sensitivity and water transport capacity. It has long been recognized that the functional trait diversity of tropical forests mirrors their large species diversity (Corner, 1949; Hallé et al., 1978; Leigh, 1999), and diversity in plant hydraulic traits such as the water potential at turgor loss ($\pi_{\text{tp}}$) and at 50 % loss of conductivity ($P_{\text{50}}$), xylem-specific hydraulic conductivity ($k_{\text{s,max}}$), and the leaf-to-sapwood area ratio ($A_L : A_s$) is no exception (Borchert, 1994; Tobin et al., 1999; Lopez et al., 2005; Meinzer et al., 2008a; Zhu et al., 2013). Interspecific variation in these traits is thought to mediate species differences in survival in both natural and experimentally induced droughts (Nakagawa et al., 2000; Engelbrecht and Kursar, 2003; Kursar et al., 2009; Moser et al., 2014; Meir et al., 2015). Differential drought sensitivity, in turn, is known to explicitly link to species distributions (Chao et al., 2007; Engelbrecht et al., 2007; Baltzer et al., 2008; Condit et al., 2013). Mechanistic plant hydraulics also allows for diversity in tree height, already included in size-structured forest ecosystem models, to act as a hydraulic trait because it influences the path length over which water must traverse to reach the canopy. Larger trees often demonstrate significantly higher vulnerability to drought (Nakagawa et al., 2000; Nepstad et al., 2007; da Costa et al., 2010; Rowland et al., 2015a), a trend demonstrated to be pantropical in scope (Phillips et al., 2010; Bennett et al., 2015). Such a trend highlights that the increasing hydraulic path length associated with tall trees (and perhaps increased radiation loads) is an equally important determinant of drought sensitivity (McDowell and Allen, 2015).

The implication for ecosystem models is that under-representation of diversity in functional traits and tree size in tropical forests is undermining efforts to make accurate projections of tropical forest response to climate. Model parameterization of hydraulic trait diversity should thus provide much-needed model capability to represent a diversity of responses to changes in moisture availability, laying the groundwork for representing trait-mediated differences in survival and subsequent shifts in forest trait composition. Shifts in trait composition are already occurring in some tropical forests (e.g., Enquist and Enquist, 2011; van der Sande et al., 2016), and such shifts (or the diversity of traits alone) have been shown to buffer ecosystems in the face of environmental change, and in some cases, are the difference between predicted complete loss of forest and forest persistence (Fauser et al., 2012; Levine et al., 2016; Sakschewski et al., 2016).

Because a complete representation of trait diversity in models is neither tractable nor desired, the challenge is to identify and represent the dominant dimensions of trait variation within our plant hydraulics model. To date, syntheses of hydraulic traits have typically focused on a limited set of traits in isolation (Bartlett et al., 2012; Chao et al., 2012; Nardini et al., 2014; Anderegg, 2015); also see Mencuccini et al. (2015). In contrast, here we synthesize how a large suite of hydraulic traits represented in our model vary with better quantified dimensions of plant functional trait variation, such as the leaf (Wright et al., 2004) and stem (Chave et al., 2009) economics spectra. We make no a priori assumptions about the coordination (or lack thereof) between leaf and stem economics, given the conflicting evidence (Baraloto et al., 2010; Mendez-Alonzo et al., 2012; Reich, 2014). In addition, we assess the evidence for trade-offs and coordination among hydraulic traits, independent of leaf and stem economic traits.

In what follows, we first describe the continuous porous media hydraulics module that simulates the fast-timescale (hourly) processes of tree water flux and water potential gradients throughout the soil–plant continuum. Second, via a pantropical synthesis of hydraulic traits, we derive empirical relationships for parameterizing hydraulic trait variation in the plant hydraulics model. Third, we demonstrate the model’s capability to simulate different individual-level functional responses to environmental variation (either water or light), arising from hydraulic limitations imposed by tree size or hydraulic traits. Finally, we evaluate the new model in terms of its ability to capture diurnal dynamics of water potential, individual-level variation in leaf water status, and the observed seasonal dynamics of ecosystem-level water use.

## 2 Model description: plant hydraulics and host individual trait-driven forest models

The plant hydraulics model developed here is integrated into a “host” individual tree trait-driven forest model (trait forest simulator, TFS). A TFS simulates every tree in a stand >10 cm diameter at breast height (DBH) driven by plot observations of the tree size distribution, and each tree can possess a unique set of values of the four functional traits: wood density (WD), leaf mass per area (LMA), leaf nitrogen ($N_L$), and leaf phosphorus ($P_L$). The modification of transpiration fluxes in TFS by the plant hydraulics module can be summarized as follows (see Fig. 1): the hydraulics module passes to TFS a nondimensional multiplier (fraction of maximum conductance, $\text{FMC}_{\text{ph}}$) (0,1] for each tree, which is based on each tree’s leaf water potential at the previous time step. TFS uses $\text{FMC}_{\text{ph}}$ within its biophysics calculations to estimate transpiration fluxes for each tree at the current time step, and the transpiration of each tree is passed back to the hydraulics model, which computes changes in water potential and water content throughout the soil–plant continuum (including root water uptake or loss) due to transpiration. The plant hydraulics module also assigns hydraulic traits based on each tree’s size and pre-assigned plant functional traits. Below we describe the components of the plant hydraulics model (Sect. 2.1), and at the end of each subsection highlight the key hydraulic traits for which we seek to understand trait trade-offs and coordination via our empirical synthesis. Section 2.2 gives a description of the host TFS model. The Supplement S1 gives a full technical description of the hydraulics.
module as it is implemented within TFS, and Fig. S1.1 in the Supplement can be referred to as the schematic of our hydraulics module.

2.1 Plant hydraulics model

2.1.1 Overview

In this section we highlight the important developments we made to the model developed by Sperry et al. (1998; hereafter S98). S98 consists of a discretization of the soil–plant continuum as a series of water storage compartments with defined heights, volumes, conducting areas, water retention, and conductivity properties, connected by elements with defined path lengths and conductances. Trees are divided into the four porous medium types of leaf, stem, transporting root, and absorbing root, with the stem being divided into a variable number of compartments and all other types consisting of a single compartment (Fig. S1.1). The soil is radially discretized with a variable number of compartments, or cylindrical “shells” around a characteristic absorbing root (the rhizosphere), and soil hydraulic properties are assumed constant across these compartments and across trees within a given soil type. The present TFS v.1-Hydro scheme does not consider the vertical distribution of soil water or roots. We anticipate this to be a key component for future model development when we incorporate this scheme into host models with variable soil depths.

The fast-timescale dynamics of the hydraulics model is governed by three sets of constitutive relationships: (1) the relationship between water potential and water content, (2) the relationship between hydraulic conductivity and water potential, and (3) the relationship between a stomatal water stress multiplier and leaf water potential. The first two relations are applied to every compartment within the plant–soil continuum and have specific equations for plant and soil porous media types. The soil constitutive equations for the first two relations are given by, respectively, the van Genuchten (1980) and Mualem (1976) formulations. We chose these particular equations for the soil water characteristic and unsaturated hydraulic conductivity because extensive work has parameterized these formulations on tropical soils, which have been noted to have distinct hydraulic prop-
erties when compared to temperate soils of similar texture (Tomasella and Hodnett, 2002). These equations are given in Sect. 3.1.2 and 3.2.2 of Supplement S1. The plant constitutive equations for the first two relations are formulated and described in Sect. 2.1.2 and 2.1.3 below. The third relation for stomatal response to moisture stress is described in Sect. 2.1.5. All parameters of the constitutive relations for plant tissue are biologically interpretable and measurable plant hydraulic traits.

Several linkages are made between tree allometry and hydraulic properties (see Supplement S1 Sect. 4). Leaf, stem, transporting root, and absorbing root water storage compartment volumes derive respectively from the TFS-predicted leaf, stem, coarse-root, and fine-root biomasses using characteristic tissue densities. The heights of these components derived from tree height and rooting depth. The characteristic soil volume over which root uptake occurs is given by half the distance between absorbing roots, which decreases as total community root length (summed across all trees) increases. Total hydraulic conductance between adjacent plant water storage compartments is scaled from xylem hydraulic conductivity using first principles and plant allometric theory (Sect. 2.1.4). The model code only initializes these allometrically dependent hydraulic properties; it does not (yet) implement functions to update them as trees grow. Neglecting the effects of growth has negligible effects on the results presented in this paper but will be necessary for application of the model at timescales longer than 1 year.

The numerical solution (see Supplement S1 Sect. 5) operates at every time step (hourly) and updates water contents and potentials throughout the plant–soil continuum (including root uptake or loss) due to transpiration. It uses a first-order Taylor series expansion of the water content term to linearize the Richards mass balance equation describing the one-dimensional continuous array of plant and soil compartments. This results in a tridiagonal matrix that is solvable without iteration. Following the approach of Siqueira et al. (2008), infiltration and drainage are treated separately from the plant–soil fluxes due to transpiration (Supplement S1 Sect. 6).

### 2.1.2 Tissue water relations

We used PV theory (Tyree and Hammel, 1972; Tyree and Yang, 1990; Bartlett et al., 2012) to describe the constitutive relation between total water potential ($\psi_{\text{tot}}$, MPa) and relative water content (RWC; g H$_2$O g$^{-1}$ H$_2$O at saturation) in the plant compartments (Eq. 1). $\psi_{\text{sat}}$ is the sum of two components: solute potential ($\psi_{\text{sol}}$, MPa) which is negative due to the presence of solutes in living cells, and pressure potential ($\psi_{\text{p}}$, MPa) which is $\geq 0$ due to cell wall turgor (but see Ding et al., 2014). PV theory is usually applied to leaves, but can also apply to sapwood (Chapoton et al., 2006; Scholz et al., 2007; Meinzer et al., 2008b); thus, we apply it here to all plant tissue. Sapwood also stores capillary water (Tyree and Yang, 1990) in its void spaces and embolized conduits. Consequently, this relation is described by three successive dehydration phases representing capillary water (sapwood only), elastic cell drainage (positive turgor), and continued drainage after cells have lost turgor:

$$\psi_{\text{tot}} = \begin{cases} \psi_0 - m_{\text{cap}}(1 - \text{RWC}) & \text{RWC}_\text{ft} \leq \text{RWC} \leq 1 \\ \psi_{\text{sat}}(\text{RWC}) + \psi_{\text{p}}(\text{RWC}) & \text{RWC}_\text{ftp} \leq \text{RWC} \leq \text{RWC}_\text{ft} \\ \psi_{\text{sat}}(\text{RWC}) & \text{RWC} \leq \text{RWC}_\text{ftp} \end{cases}$$  (1)

The first phase (capillary water) is assumed linear, characterized by a slope ($m_{\text{cap}}$) and a saturated water potential ($\psi_0$). The second elastic drainage phase has both solute ($\psi_{\text{sol}}$) and pressure ($\psi_{\text{p}}$) potential changing from the relative water content at which elastic drainage begins (RWC$_{\text{ft}}$) up to the turgor loss point (RWC$_{\text{ftp}}$ and corresponding $\psi = \pi_{\text{ftp}}$), where $\psi_{\text{p}} = 0$. In the final post-turgor loss phase, symplastic (cell water) and xylem water from embolized conduits are expressed up to the point at which $\psi_{\text{tot}}$ approaches $-\infty$ (RWC$_{\text{ft}}$). RWC$_{\text{ftp}}$ is often referred to as the apoplastic fraction (Bartlett et al., 2012), but in light of the considerable amount of water released when vessels embolize in stems (Tyree et al., 1991; Holttta et al., 2009), RWC$_{\text{ftp}}$ is best termed the residual fraction. Leaf PV curves as traditionally interpreted are a special case of Eq. (1) in which there is no capillary water (RWC$_{\text{ft}} = 1$). $\psi_{\text{sol}}$ and $\psi_{\text{p}}$ are respectively given by

$$\psi_{\text{sat}}(\text{RWC}) = \psi_{\text{sat}}(\text{RWC})^{*} = \frac{-|\pi_{\text{ftp}}|}{\text{RWC}^{*}};$$  (2)

$$\psi_{\text{p}}(\text{RWC}) = \psi_{\text{p}}(R^{*}) = |\pi_{\text{ftp}}| - \varepsilon R^{*};$$  (3)

where $\pi_{\text{ftp}}$ (MPa) and $\varepsilon$ (MPa) are osmotic potential at full turgor and bulk elastic modulus, respectively, and RWC$^{*} = (\text{RWC} - \text{RWC}_{\text{ftp}})/(\text{RWC}_{\text{ft}} - \text{RWC}_{\text{ftp}})$ and R$^{*} = (\text{RWC}_{\text{ft}} - \text{RWC}_{\text{ftp}})/(\text{RWC}_{\text{ftp}} - \text{RWC}_{\text{ft}})$ are transformations representing RWC and $R$ (relative water deficit; $1 - \text{RWC}$) of symplastic (cell) water only (Bartlett et al., 2012). The absolute mass ($W$; kg) of water in tissue is given by $W = \rho_{\text{w}} \theta_{\text{sat}} \text{RWC}$, where $\theta_{\text{sat}}$ (m$^3$ m$^{-3}$) is the maximum water content on a per volume basis (or porosity) and $\rho_{\text{w}}$ (kg m$^{-3}$) is the density of water. The constitutive equations used in the model are, in terms of volumetric water content ($\theta$; m$^3$ H$_2$O m$^{-3}$ plant tissue), achievable by using the transformations given above (see Supplement S1). Volumetric capacitance ($C$; kg m$^{-3}$ MPa$^{-1}$) is defined at any point along the three regions as $(\rho_{\text{w}} \theta_{\text{sat}} \text{dRWC})/\text{d}\psi$, which for sapwood is the highest in the capillary region ($\sim 200$–$400$ kg m$^{-3}$ MPa$^{-1}$), intermediate in the elastic region (20–200 kg m$^{-3}$ MPa$^{-1}$) (Tyree and Yang, 1990), and, after an initial increase beyond the turgor loss point, in theory will approach zero as RWC $\to$ RWC$_{\text{ftp}}$. Henceforth and in all figures, we report $C$ at full turgor (C$_{\text{ftp}}$), defined as the change in water mass per unit volume per unit change in water potential over the region where cell turgor ($\psi_{\text{p}}$) $\geq 0$.

$$C_{\text{ft}} = \frac{\rho_{\text{w}} \theta_{\text{sat}} \Delta \text{RWC}}{\Delta \psi}$$

For these hydraulic traits and all others which follow, we denote specific reference to

leaf and xylem (sapwood) tissue with the subscripts \(l\) and \(x\), respectively. Key hydraulic PV traits which we seek to determine as functions of more commonly measured plant traits are \(\varepsilon\), \(\pi\), \(RWC_{l,\text{tip}}\), \(C_{fl}\), and \(RWC_{r}\).

### 2.1.3 Embolism vulnerability

We use the inverse polynomial of Manzoni et al. (2013a) for the xylem vulnerability curve, termed here the “fraction of maximum conductivity” for xylem (FMC\(_x\)):

\[
\text{FMC}_x(\psi_x) = \left[1 + \left(\frac{\psi_x}{P_{50,x}}\right)^{a_x}\right]^{-1},
\]

where \(P_{50,x}\) is the water potential at which 50% of maximum conductance is lost and \(a_x\) is a shape parameter (unitless). FMC\(_x\) is defined at compartment nodes and is a nondimensional multiplier bounded on (0,1] limiting the maximum xylem conductance (\(K_{\text{max},i,x}\); kg s\(^{-1}\) MPa\(^{-1}\); FMC\(_x\) = 1 and 0 indicate no and complete xylem embolism) between any two compartments \(i\) and \(i+1\). Section 1 of the Supplement outlines how FMC\(_x\), defined at compartment nodes, limits \(K_{\text{max},i,x}\), which is defined at compartment boundaries. Critically, \(K_{\text{max},i,x}\) derives from plant hydraulic architecture and maximum xylem-specific conductivity (\(k_{s,\text{max},x}\)). Because \(k_{s,\text{max},x}\) relates to the maximum rate at which water can be transported through the xylem, and \(P_{50,x}\) quantifies the xylem water potential at which half of this transport capacity is lost, these can be thought as representing xylem efficiency and safety, respectively. “Safety” refers to a property of the xylem alone and is not to be confused with the hydraulic safety margin (HSM), which is field-observed or modeled minimum leaf water potential relative to \(P_{50,x}\). As with PV traits, we seek to determine to what extent xylem efficiency and safety covary with other plant traits and trade off with each other.

### 2.1.4 Scaling conductance with tree size

Tree size exhibits a first-order control over much of the variation in whole-plant hydraulic conductance (Sperry et al., 2008) since hydraulic path length increases with tree height (\(H\)) (Mencuccini, 2002). For this reason, whole-plant conductance is not a constant parameter in our model. Rather, first principles dictate that, to a first approximation, whole-tree maximum aboveground conductance (\(K_{\text{max,tree,ag}}\); kg s\(^{-1}\) MPa\(^{-1}\)) may be derived as a function of xylem conductivity (\(k_{s,\text{max},x}\); kg m\(^{-1}\) s\(^{-1}\) MPa\(^{-1}\)), sapwood area (\(A_s\); m\(^2\)), and \(H\) (m) as

\[
K_{\text{max,tree,ag}} = \frac{k_{s,\text{max},x}A_s}{H}. \tag{5}
\]

This relation predicts the negative effects of \(H\) (increasing \(H\) means decreasing \(K_{\text{max,tree,ag}}\)). Whole-tree conductance per unit leaf area (\(K_{\text{max,tree,ag}}\)) will determine water status at the level of individual leaves, and thus hydraulic constraints on

leaf-level gas exchange. Dividing through by leaf area (\(A_l\); m\(^2\)) gives

\[
K_{\text{max,tree,ag}} = \frac{k_{s,\text{max},x}A_s}{H A_l}. \tag{6}
\]

In this relation, the leaf sapwood area ratio (\(A_l : A_s\)) emerges as a key plant trait controlling \(K_{\text{max,tree,ag}}\). If \(A_l\) : \(A_s\) decreases with tree height, as has been documented in many tree species (McDowell et al., 2002; also see Calvo-Alvarado et al., 2008), the negative effects of height can be partially overcome. In addition, the near-universal tendency for xylem conduits to decrease in diameter within trees from trunk base to stem tips (Mencuccini et al., 2007; Meinzer et al., 2010; Petit and Anfodillo, 2011; Olson and Rosell, 2013; Olson et al., 2014) also mitigates the negative effects of height according to the Hagen–Poiseuille law. Neglecting the effects of xylem taper may thus overestimate the negative hydraulic effects of increasing path length in size-structured forests. Metabolic scaling theory (MST) makes baseline predictions about the optimal degree of xylem conduit taper in trees subject to the constraint of hydraulic safety (which decreases as conduits get larger) and has been validated against observations of conduit diameter across trees of different heights (West et al., 1999; Savage et al., 2010). We therefore use MST to include the effect of xylem taper on \(K_{\text{max,tree,ag}}\) by modifying Eq. (6) to include a xylem taper term (\(\chi_{\text{tap: notap,ag}}\); unitless) representing the ratio of whole-plant conductance with taper to that without:

\[
K_{\text{max,tree,ag}} = \frac{k_{s,\text{max,petiole}}A_s}{H A_l} \chi_{\text{tap: notap,ag}}, \tag{7}
\]

where \(k_{s,\text{max,petiole}}\) (kg m\(^{-1}\) s\(^{-1}\) MPa\(^{-1}\)) is used to reference \(K_{\text{max,tree,ag}}\) to MST predictions. \(\chi_{\text{tap: notap,ag}}\) is in the range of 23–50 for trees of heights 10–30 m; thus, the benefit of xylem taper for increasing total plant conductance itself increases with tree height. \(\chi_{\text{tap: notap}}\) is assumed constant across individuals in this study, but parameterizing variation in \(A_l\) : \(A_s\) across species is an outcome of this study. The full details of this approach in addition to the treatment of the belowground component of tree conductance (\(K_{\text{max,tree,bg}}\)) are outlined in Sect. 2 of the “Technical description” (Supplement S1).

### 2.1.5 Hydraulic impacts on stomatal conductance

The fraction of maximum stomatal conductance (FMC\(_{gs}\); [0,1]), which is updated at every time step and is used to downregulate non-water stressed stomatal conductance (sensu Jarvis, 1976), is one of two variables passed from the hydraulics module to the host model TFS. As in Mencuccini et al. (2015), the formulation for FMC\(_{gs}\) mimics the loss of xylem conductivity with water potential relation:

\[
g_s = g_{s,\text{max}} \times \text{FMC}_{\text{gs}} = g_{s,\text{max}} \left[1 + \left(\frac{\psi}{P_{50,gs}}\right)^{a_{gs}}\right]^{-1}, \tag{8}
\]
where $P_{50,gs}$ and $a_{gs}$, respectively, represent leaf water potential at 50% stomatal closure and the slope of the curve at $\psi_l = P_{50,gs}$, and $g_{s,max}$ is stomatal conductance in the absence of water supply limitation and comes from the host model’s stomatal conductance scheme (for TFS, we used Medlyn et al., 2011; see Eq. A25 of Fyllas et al., 2014). The second variable passed from the hydraulics module to the TFS model describes the derivative of FMC$_{gs}$ with leaf water content, which is derived from the FMC$_{gs} \sim \psi_l$ (Eq. 8) and leaf pressure–volume (Eqs. 1–3) relations (see Eqs. S34a and S40 in the Supplement S1 for derivation). This variable is used as an additional constraint promoting stability within the numerical solution (see Sect. 5.3 of the Supplement S1 and Supplement S3 Fig. S3.3).

Theory and data suggest that stomata operate in such a way as to prevent catastrophic xylem embolism (Sperry and Love, 2015); this implies that $P_{50,gs} > P_{50,x}$, which is supported by global datasets (Klein, 2014; Manzoni et al., 2013b). We used a 1 : 1 relationship between $P_{50,gs}$ and the water potential at 20% loss of xylem hydraulic conductivity ($P_{20,x}$), a relationship suggested by data from a tropical dry forest (Brodribb et al., 2003). $P_{20,x}$ is easily derived from $P_{50,gs}$ and $a_{gs}$ using Eq. (4). This current approach ignores the continuum in hydraulic safety (Klein, 2014; Martinez-Vilalta et al., 2014; Skelton et al., 2015) and future work needs to identify how such a continuum maps onto other plant traits. $a_{gs}$ is derived from $P_{50,gs}$ using the same relationship derived for $a_x$ and $P_{50,x}$ (Table 2).

### 2.2 Host model TFS

The TFS (Fyllas et al., 2014) is an individual tree model parameterized from plot-level observed tree size distribution and distributions of the four plant traits: WD, LMA, NL, and PL. These traits are assigned to each individual upon initialization in a way that preserves the underlying observed joint distribution of all four traits (be it tight or weak coupling among traits), which are then used to parameterize plant physiological parameters for each tree in a stand (Fig. 1). WD in TFS v.1 drives differential rates of stem respiration and stem C allocation, and LMA, NL, and PL jointly drive differential rates of leaf light-saturated photosynthesis (photosynthetic capacity, $A_{max}$) via parameterization of photosynthetic parameters $V_{c,max}$ and $J_{max}$ (Domingues et al., 2010), in addition to respiration and leaf and stem C allocation. The construction of this model was motivated by a need to diagnose large-scale patterns of productivity and turnover in tropical forests that, among other drivers, implicate a link to soil fertility and, more proximally, plant traits reflective of fertility differences (Baker et al., 2004; Fyllas et al., 2009; Patiño et al., 2012; Quesada et al., 2012; Johnson et al., 2016). Because measurements of stand-level variation in easily measured plant traits are much more abundant than comparative measures of hydraulic traits, we sought to parameterize hydraulic trait variation via empirical relationships with the four TFS traits (see Table 1 and Sect. 3.1 below).

A second key component of TFS is the representation of stand structure and hence individual-level variation in tree crown light environments, which is an additional key determinant of ecosystem productivity (Stark et al., 2012) and, as we will show, individual-level variation in tree hydraulic function and limitation. To estimate the light environment for each tree, TFS uses the perfect plasticity approximation (PPA) of Purves et al. (2007) and Strigul et al. (2008), updated to represent multiple canopy layers typical of tropical forests (Bohmann and Pacala, 2012) as well as basic gap dynamics that allow for a fraction of small trees to receive full sunlight (Fauset et al., unpublished manuscript).

In conclusion, TFS couples individual-level variation in plant traits and light environments with a full canopy biophysics scheme for estimating photosynthesis and transpiration for each tree > 10 cm DBH in a stand. Transpiration is the necessary boundary condition for our new plant hydraulics module (Fig. 1). For further details on TFS biophysics, size structure, trait parameterization, and light competition, see Fyllas et al. (2014) and references therein.

### 3 Methods

#### 3.1 Hydraulic trait synthesis for parameterization of plant hydraulics model

We synthesized the literature and existing trait databases and asked whether hydraulic traits associated with PV curves in leaves ($\delta_{ol,x}$, $\epsilon_{s}$, RWC$_{lp,x}$, and RWC$_{cr,1}$) and sapwood ($\sigma_{s,x}$, $\epsilon_{s}$, RWC$_{lp,x}$, and RWC$_{cr,1}$), xylem conductivity per unit sapwood area ($k_{s,max,x}$) or per unit leaf area ($k_{l,max,x}$), xylem resistance to embolism ($P_{50,x}$), and associated shape parameter ($a_x$) as well as hydraulic architecture ($A_1 : A_s$) were coordinated with the leaf and stem traits employed as input by the TFS model. All statistical analyses were carried out within the R language (R Core Team, 2015).

For leaf PV traits and LMA, we started with the Bartlett et al. (2012) database subset for tropical ecosystems (111 individuals across 14 studies) and added substantial additional

<table>
<thead>
<tr>
<th>Trait</th>
<th>Symbol</th>
<th>Units</th>
</tr>
</thead>
<tbody>
<tr>
<td>Tree height</td>
<td>$H$</td>
<td>m</td>
</tr>
<tr>
<td>Tree diameter at breast height</td>
<td>DBH</td>
<td>cm</td>
</tr>
<tr>
<td>Leaf mass per area</td>
<td>LMA</td>
<td>g m$^{-2}$</td>
</tr>
<tr>
<td>Leaf nitrogen</td>
<td>NL</td>
<td>mg g$^{-1}$</td>
</tr>
<tr>
<td>Leaf phosphorus</td>
<td>PL</td>
<td>mg g$^{-1}$</td>
</tr>
<tr>
<td>Light-saturated photosynthesis rate</td>
<td>$A_{max}$</td>
<td>$\mu$mol m$^{-2}$ s$^{-1}$</td>
</tr>
<tr>
<td>Wood density</td>
<td>WD</td>
<td>g cm$^{-3}$</td>
</tr>
</tbody>
</table>

Table 1. Input tree size and leaf and stem traits used by the TFS model.
### Table 2. Hydraulic traits used by the model. Resultant empirical “ecotransfer” function which estimates hydraulic traits as a function of input traits and its corresponding source.

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Symbol</th>
<th>Units</th>
<th>Empirical ecotransfer function</th>
<th>Source</th>
</tr>
</thead>
<tbody>
<tr>
<td>Leaf-saturated water content</td>
<td>$\theta_{s,l}$</td>
<td>m$^3$ m$^{-3}$</td>
<td>$-2.32 \times 10^4 / $ LMA + 782</td>
<td>$\rho_w \left( \frac{1}{\rho_w} \frac{1}{-0.21 \ln(10^4 / $ LMA) + 1.43} - 1 \right)$</td>
</tr>
<tr>
<td>Leaf osmotic potential at full turgor</td>
<td>$\pi_{o,l}$</td>
<td>MPa</td>
<td>$-0.04 - 1.51$ LMA - 0.0067LMA</td>
<td>This study</td>
</tr>
<tr>
<td>Leaf osmotic potential at turgor loss</td>
<td>$\pi_{tlp,l}$</td>
<td>MPa</td>
<td>$\frac{\varepsilon_x \pi_{o,l}}{\varepsilon_x + \pi_{o,l}}$</td>
<td>Bartlett et al. (2012)</td>
</tr>
<tr>
<td>Leaf bulk elastic modulus</td>
<td>$\varepsilon_1$</td>
<td>MPa</td>
<td>$2.5 + \frac{37.5}{1 + e^{3000 / T + 273}}$</td>
<td>This study</td>
</tr>
<tr>
<td>Leaf relative water content at turgor loss</td>
<td>$\text{RWC}_{tlp,l}$</td>
<td>–</td>
<td>$(\pi_{o,l} (1 - \text{RWC}<em>{tlp,l}) + \varepsilon_1) \frac{\rho_w \varepsilon_x (1 - \text{RWC}</em>{tlp,l}) (\varepsilon_1 + \pi_{o,l})}{\varepsilon_1}$</td>
<td>Bartlett et al. (2012)</td>
</tr>
<tr>
<td>Leaf residual fraction</td>
<td>$\text{RWC}_{l,l}$</td>
<td>–</td>
<td>$0.01 \varepsilon_1 + 0.17$</td>
<td>This study</td>
</tr>
<tr>
<td>Leaf capacitance over RWC$<em>l$ = 1 to RWC$</em>{tlp,l}$</td>
<td>$C_{fl,l}$</td>
<td>g H$_2$O g$^{-1}$ dry weight</td>
<td>$\rho_w \varepsilon_x \pi_{o,x} (1 - \text{RWC}<em>{tlp,x}) (\pi</em>{o,x} \varepsilon_x + \pi_{o,x})$</td>
<td>Bartlett et al. (2012)</td>
</tr>
<tr>
<td>Sapwood-saturated water content</td>
<td>$\theta_{s,x}$</td>
<td>m$^3$ m$^{-3}$</td>
<td>$1 - \frac{WD}{3.74}$</td>
<td>Siau (1984)</td>
</tr>
<tr>
<td>Sapwood RWC at which capillary reserves exhausted</td>
<td>$\text{RWC}_{fl,x}$</td>
<td>–</td>
<td>$1 - 0.72 (1 - \text{RWC}_{tlp,x})$</td>
<td>This study</td>
</tr>
<tr>
<td>Fraction of 1 - $r_{f/2}$ that is capillary in source</td>
<td>$f_{cap}$</td>
<td>–</td>
<td>0.07</td>
<td>This study</td>
</tr>
<tr>
<td>Sapwood osmotic potential at full turgor</td>
<td>$\pi_{o,x}$</td>
<td>MPa</td>
<td>$0.52 - 4.16 \times WD$</td>
<td>This study</td>
</tr>
<tr>
<td>Sapwood osmotic potential at turgor loss</td>
<td>$\pi_{tlp,x}$</td>
<td>MPa</td>
<td>$\frac{\varepsilon_x \pi_{o,x}}{\varepsilon_x + \pi_{o,x}}$</td>
<td>Bartlett et al. (2012)</td>
</tr>
<tr>
<td>Sapwood bulk elastic modulus</td>
<td>$\varepsilon_x$</td>
<td>MPa</td>
<td>$\sqrt{1.02 e^{8.3 WD} - 2.89}$</td>
<td>This study</td>
</tr>
<tr>
<td>Sapwood RWC at turgor loss</td>
<td>$\text{RWC}_{tlp,x}$</td>
<td>–</td>
<td>$1 - \frac{t_{(1 - 0.75 x WD)} (2.74 + 2.01 x WD)}{2.74 + 2.01 x WD}$</td>
<td>This study</td>
</tr>
<tr>
<td>Sapwood residual fraction</td>
<td>$\text{RWC}_{r,x}$</td>
<td>–</td>
<td>$\frac{\varepsilon_x (1 - f_{cap} - \text{RWC}<em>{tlp,x})}{\pi</em>{o,x} (1 - f_{cap})} + 1 - \varepsilon_x f_{cap}$</td>
<td>This study</td>
</tr>
<tr>
<td>Sapwood capacitance over RWC$<em>l$ = 1 to RWC$</em>{tlp,l}$</td>
<td>$C_{fl,x}$</td>
<td>kg m$^{-3}$ MPa$^{-1}$</td>
<td>$\rho_w \varepsilon_x \pi_{o,x} (1 - \text{RWC}<em>{tlp,x}) (\pi</em>{o,x} \varepsilon_x + \pi_{o,x}) (1 - f_{cap}) - \varepsilon_x f_{cap}$</td>
<td>This study</td>
</tr>
<tr>
<td>Xylem water potential at 50% loss of conductivity</td>
<td>$P_{50,x}$</td>
<td>MPa</td>
<td>$- (3.57 WD)^{1.73} - 1.09$</td>
<td>This study</td>
</tr>
<tr>
<td>Slope of xylem vulnerability curve at $P_{50}$</td>
<td>$a_x$</td>
<td>MPa$^{-1}$</td>
<td>54.4 $(-P_{50,x})^{-1.17}$</td>
<td>This study</td>
</tr>
</tbody>
</table>
data (44 individuals across 12 studies) through further literature review, and joined to this dataset species-averaged values of wood density using the Global Wood Density Database (Chave et al., 2009; Zanne et al., 2009). For sapwood PV traits, we generated our own database (33 specimens across 9 studies) by digitizing data points of RWCx vs. \( \psi_s \) conducted on trunk or branch tissue from published figures using WebPlotDigitizer (Rohatgi, 2014), and fit curves to these plots following established methods for extracting PV parameters (Tyree and Hammel, 1972; Sack and Pasquet-Kok, 2011; Bartlett et al., 2012). We found significant differences in the shape of curve and magnitude of capacitance for sapwood PV curves conducted on small trunk cores vs. cut whole terminal branches, with trunk cores yielding capacitance 2–10 times that of branches, suggesting substantial artifact-associated measurements conducted on trunk cores, an issue recently highlighted by Wolfe and Kursar (2015). Using measurements made on the same species using both methods, we developed a correction procedure and applied it to data suffering this bias (Christoffersen, unpublished manuscript), and provide versions of figures without the correction applied in the Supplement S2.

For analyzing relations between xylem efficiency and safety and other traits, we started with the Xylem Functional Traits (XFT) database (Choat et al., 2012; Gleason et al., 2016), subset to tropical regions and added data from 15 additional species not originally present (Mendez-Alonso et al., 2012). We then evaluated the hypothesis that variation in hydraulic transport is explained by WD by prioritizing WD reported in the original publications and adding species-averaged WD data from Zanne et al. (2009) where it was unreported. We also evaluated the hypothesis that rates of hydraulic transport coordinate with leaf-level rates of gas exchange. Because the objective of this paper is to derive hydraulic traits from other non-hydraulic plant traits, we explored the relationship of \( k_{s,max,x} \) regressed on the light-saturated photosynthesis rate \( (A_{max}) \), even though the causality of the relationship could be interpreted in reverse. There is substantial debate surrounding appropriate methods for determining embolism vulnerability (Choat et al., 2010; Cochard et al., 2010; Sperry et al., 2012; Wang et al., 2014) (see Brodribb et al., 2016 for a brief summary). Because there is some consensus that the “gold standard” for \( P_{50,x} \) measurements involves bench dehydration (DH) on long-stem segments (Jansen et al., 2015), we explored trait relationships with and without other measurement methods for \( P_{50,x} \).

For hydraulic architecture, we used the only study of which we were aware for tropical trees (Calvo-Alvarado et al., 2008), which reports independent measurements of individual total tree leaf area and sapwood area across a wide range of tree sizes to explore the relationship of \( A_1 : A_s \) with tree height as well as with LMA, and verified the latter relationship using a much broader dataset of branch-level measurements of \( A_1 : A_s \) conducted across the Amazon basin (Patiño et al., 2012). Via literature survey, we also compiled an independent, extensive dataset of \( A_s \) as it varied with tree DBH (cm) in tropical forests.

Finally, we standardized the representation of the “biome” category across all databases, and defined the following categories based on the location of the study (not the species’ home range): tropical flooded forest (if identified as such in original publication), tropical wet forest (no months where evaporative demand exceeds precipitation), tropical moist forest (at least 1 month where evaporative demand exceeds precipitation; predominantly evergreen), tropical dry forest (drought-deciduous phenologies make up a substantial fraction of species), tropical savanna (identified as such in original publication), subtropical forest (absolute latitude exceeded 23°), tropical mangrove, and greenhouse. Syntheses of traits were limited to studies conducted on species growing in native (non-greenhouse) environments in tropical (thus excluding subtropical observations) upland (non-flooded) habitats. In some cases, it was also necessary to

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Symbol</th>
<th>Units</th>
<th>Empirical “ecotransfer” function</th>
<th>Source</th>
</tr>
</thead>
<tbody>
<tr>
<td>Maximum xylem conductivity per unit leaf area</td>
<td>( k_{l,max,x} )</td>
<td>kg m(^{-1}) s(^{-1}) MPa(^{-1})</td>
<td>( 0.0021e^{-26.6WD/A_{max}} )</td>
<td>This study</td>
</tr>
<tr>
<td>Maximum xylem conductivity per unit sapwood area</td>
<td>( k_{s,max,x} )</td>
<td>kg m(^{-1}) s(^{-1}) MPa(^{-1})</td>
<td>( k_{l,max,x} / A_l/A_s )</td>
<td>This study</td>
</tr>
<tr>
<td>Leaf to sapwood area ratio</td>
<td>( A_1 : A_s )</td>
<td>m(^2) cm(^{-2})</td>
<td>( 546LMA^{-2.14}H )</td>
<td>This study</td>
</tr>
<tr>
<td>Leaf-to-fine-root absorbing surface area ratio</td>
<td>( A_1 : A_f )</td>
<td>m(^2) m(^{-2})</td>
<td>1</td>
<td>This study</td>
</tr>
</tbody>
</table>
match hydraulic traits for the same species given as multiple (different) records within a database. As a rule, however, we did not average to the species level, and thus some variation included in regressions is intraspecific, albeit small.

3.2 Model setup

We used Princeton downscaled meteorological forcing data (Sheffield et al., 2006), observed soil textural properties, and observed tree size and trait distributions (for WD, LMA, N\textsubscript{t}, and P\textsubscript{L}) in the Caxiuana National Forest of east-central Brazilian Amazonia, a seasonal evergreen forest receiving 2100–2500 mm of rainfall annually (Meir et al., 2008; da Costa et al., 2010; Meir et al., 2015; Rowland et al., 2015a, c), to parameterize and run our model for 1 year. Spin-up was not necessary, as the model was initialized from plot observations as described above, and an initial soil moisture was chosen typical of wet season conditions, which is when the model period began. While actual soil depth at the site is \sim 10 m, we set soil depth in simulations to 4 m, which reflects the effective depth over which the majority of water extraction occurs, based on previous model validation with soil moisture data (Fisher et al., 2007). This setup applied to all model simulations in this paper – both the model experiments and the simulations for comparison with field data.

3.3 Model experiments

We conducted two sets of model experiments designed to explore the impacts of the new plant hydraulics scheme on simulated tree transpiration. In both cases, the TFS model was run with (TFS v.1-Hydro) and without (TFS v.1) the new plant hydraulics scheme. In the first experiment, all trees were assigned identical trait values using plot-mean values of WD, LMA, N\textsubscript{t}, and P\textsubscript{L}. Plant hydraulic traits were identical across individuals in TFS v.1-Hydro. Having eliminated plant hydraulic traits as a source of variation among simulated tree transpiration, we used this model experiment to explore the impact of plant hydraulics on transpiration dynamics as a function of tree height and canopy position. To explore the consequence of trait variation independent of height and canopy position, our second experiment set varied a single trait (P\textsubscript{L}, WD, or LMA), by resampling the observed plot trait distributions of these traits following the algorithm of Taylor and Thompson (1986) (see Fyllas et al., 2014 for a full description of this resampling method as implemented in TFS), keeping all other traits at the plot-mean value. We then explored the dynamics of transpiration, for canopy trees (full sunlight) only, of three large (50–55 cm DBH) trees spanning the range of assigned trait values. In both model experiments, we examined the simulated mean diurnal cycle of transpiration characteristic of the wet and dry season, defined as months where monthly precipitation exceeded or fell below 100 mm, respectively. Months at season boundaries were excluded to maximize seasonal differences.

3.4 Model evaluation

Separate from and in addition to the model experiments, we performed three sets of model evaluation that assessed the model’s ability to capture observed variations in transpiration and water potential at the individual tree level and the stand level. In the first evaluation set, we evaluated diurnal cycles of leaf and stem water potential across multiple individual trees spanning a wide range of sizes and canopy positions, which were also subject to seasonal differences in light and water availability. In this set, all individual trees possessed identical (plot-mean) trait values. We matched simulated trees to those measured by Fisher et al. (2006) on the basis of size (see Table 3) across replicate days in the month matching the month in which observations were made (May for wet season, November for dry season), taking the standard error of the mean as the model error. A total of four individual tree comparisons were made.

The second evaluation set was similar to the first, except we focused on midday simulated leaf water potentials. We conducted four distinct sets of model simulations in which we explored how successive changes to model parameterization impacted model fidelity with the observations. The first model simulation assigned trees identical trait values, equal to the plot mean, and made no accounting of xylem conduit taper (following Eq. 6). The second model simulation was like the first, except this time accounting for xylem conduit taper (following Eq. 7). The third simulation was like the second but adjusted the canopy position of two of the simulated individuals. The fourth simulation was like the third but adjusted the stomatal sensitivity parameter \(P_{30,gs}\) for a single tree to be less sensitive to leaf water potential by changing its value from \(-2.9\) to \(-5.0\) MPa.

In the final set, we evaluated seasonal differences in stand-scaled sap flux. We used model setups in which all individuals possessed identical hydraulic traits and where all individuals possessed different traits according to the plot observed trait distributions in order to assess ecosystem-level consequences of plant trait variation. Each of these setups was run as TFS v.1 and TFS v.1-Hydro. For comparison with observations, simulated individual total transpiration (kg s\(^{-1}\)) was summed across individuals, divided by the plot area, and accumulated for each month of the year. Observed stand-level transpiration was derived following the methods outlined in Fisher et al. (2007).

4 Results

4.1 Hydraulic trait synthesis for parameterization of plant hydraulics model

4.1.1 Hydraulic trait synthesis

All resultant empirical equations concerning traits WD, LMA, or \(A_{\text{max}}\) to hydraulic traits are given in Table 2.
For leaf PV traits (Fig. 2), we found highly significant ($p<0.0001$) but weak negative relationships of $\pi_{o,1}$ (leaf osmotic potential at full turgor) with both WD and LMA ($r^2 = 0.28$ and 0.09, respectively). Because WD and LMA were not significantly correlated in this dataset ($p = 0.48$), we used them as independent predictors of $\pi_{o,1}$ in multiple regression and were able to explain a higher fraction of the variance in $\pi_{o,1}$ ($r^2 = 0.44$; see Table 2). Surprisingly, $\varepsilon_l$ (leaf bulk elastic modulus) demonstrated no significant relationship with LMA but was marginally correlated with WD ($p < 0.05$; $r^2 = 0.10$; Fig. 2c, d). While we found no significant relationship of the residual fraction (RWC$t,l$) with LMA or WD, we did find a significant relationship of RWC$t,l$ with $\varepsilon_{\text{leaf}}$ ($r^2 = 0.32$; $p = 0.002$; Fig. 2e). These relationships imply that leaf capacitance increased with both decreasing WD and LMA, which was corroborated by independent datasets (Fig. 2f).

For bias-corrected sapwood PV traits (Fig. 3), we found that WD well-represented a single axis of variability along which $\pi_{o,x}$, $\varepsilon_x$, and $R_{l,p,x} = 1 - \text{RWC}_{l,p,x}$ were organized ($r^2 = 0.31$ to 0.44), with increasing WD leading to exponentially increasing $\varepsilon_x$ and decreasing $\pi_{o,x}$ and $R_{l,p,x}$. The result was exponentially declining sapwood capacitance ($C_{f,l,x}$) with WD ($r^2 = 0.84$; Fig. 3d). The same trends were generally supported by the uncorrected data (Fig. S2.1), most importantly for $C_{f,0,x}$.

On the full dataset of $P_{50,x}$, we found a significant relationship with WD ($p < 0.0001$), but the explained variance was low ($r^2 = 0.08$) due to a large number of high wood density species with high (less negative) $P_{50,x}$ values. However, when limiting the dataset to measurements made using the bench DH, many values of high (less negative) $P_{50,x}$ at intermediate to high WD were now excluded (the majority of which were obtained using the air injection method) and a stronger negative relationship emerged ($r^2 = 0.34$) (Fig. 4a). In addition, we found that the slope ($\alpha_x$) of the xylem FMC curve decreased significantly ($p < 0.0001$) with increasingly negative $P_{50,x}$ (Fig. 4b), suggesting that threshold responses of embolism to increasing xylem tension were more pronounced in individuals with less resistant xylem.

The relationship of xylem efficiency on a leaf area basis ($k_{l,\max,x}$) with WD, while highly significant ($p < 0.0001$), explained only 7% of the variance (Fig. 5a). However, when examined in relation to reported values of light-saturated leaf photosynthesis rates ($A_{\max}$), a greater fraction of the variance was explained ($r^2 = 0.29$; $p < 0.0001$) (Fig. 5b). Because $A_{\max}$ and WD were uncorrelated (Fig. 5d), we used them as independent predictors (as $A_{\max}/WD$) of $k_{l,\max,x}$, which now explained 42% of the variance (Fig. 5c). Similar
trends were observed for xylem efficiency on a sapwood area basis ($k_{l,max,x}$; Supplement Fig. S2.2).

While the Calvo-Alvarado et al. (2008) study showed no significant common relationship between $A_1 : A_s$ and $H$ or DBH across all individuals studied, it identified increasing $A_1 : A_s$ with $H$ within 4 out of 5 individuals (Fig. 6a). When we took species-averaged $A_1 : A_s$ from the Calvo-Alvarado (2008) study and plotted the interspecific differences as a function of species-averaged LMA, the individuals largely fell on the same regression line as that of Patiño et al. (2012) (Fig. 6b). Therefore, it appears that for tropical forests leaf-level traits set the range of $A_1 : A_s$ by which variation occurs in response to height and/or light availability (Fig. 6c).

4.1.2 Plant hydraulics parameterization

Figure 7 shows the parameterization of all components of the plant hydraulics model (combining Table 2 and plant hydraulics model constitutive Eqs. 1–4) as a function of the TFS model input traits and tree height ($H$). Figure 7a–c demonstrate the distribution of leaf PV, sapwood PV, and xylem vulnerability curves for a characteristic TFS model simulation in which each tree is assigned four trait values (WD, LMA, $N_{L_1}$, and $P_{L_1}$) according to the observed joint distribution (Taylor and Thompson 1986 resampling algorithm) of these four traits at the focal field site Caxiuana National Forest, which are used to parameterize each tree’s constitutive equations (using Table 2), with each tree plotted as separate lines colored by trait values. This forest is representative of others in the region (Baraloto et al., 2010), showing decoupled leaf and stem economics (i.e., lack of any significant correlation between WD and any of the three leaf traits used as input for TFS). A consequence of this lack of leaf–stem trait coupling and our formulation of $k_{l,max,x}$ as empirically derived primarily from leaf traits (Fig. 5) while $P_{50,x}$ derives from stem traits (Fig. 4a), the model input prescribes only a weak trade-off between $k_{l,max,x}$ and $P_{50,x}$ (Fig. 7c), which is consistent with the data for this relationship (Fig. 8a). In contrast, Fig. 7d–f present “idealized” scenarios for a case in which leaf and stem economics are tightly coupled (specifically, between WD and LMA or WD and $A_{max}$) over a wide range of input values. In this case, the consequence for the relationship between $k_{l,max,x}$ and $P_{50,x}$ is a perfect trade-off (Fig. 7f).

Finally, we represented the dual dependency of $A_1 : A_s$ on both LMA and tree size (Fig. 6) as an $A_1 : A_s - H$ relationship with an LMA-dependent slope (Fig. 7g). For a prescribed leaf biomass allometry function (as ~DBH$^2$; Lescure et al., 1983), this $A_1 : A_s$ formulation predicted LMA as the driver of differences in sapwood area ($A_s$)–DBH allometry, spanning the range of values given by our independent literature compilation across many different studies, though this relationship warrants further validation.

4.1.3 Ability of hydraulics model parameterization to represent observed trade-offs and coordination among hydraulic traits

Based on the pantropical xylem functional traits dataset, evidence for a trade-off between xylem efficiency (as $k_{s,max,x}$) and safety ($P_{30,x}$) was insignificant ($p = 0.14$ on log-transformed data), which was also the case when limiting the database to the DH method ($p = 0.82$; Fig. 8a).

However, when limiting the dataset to tropical dry forests (not savanna), a significant trade-off between xylem efficiency and safety emerged ($p < 0.01; r^2 = 0.17$), which was even stronger ($p < 0.0001; r^2 = 0.44$) when considering two tropical dry forest studies (see circled points in Fig. 8a; Markesteijn et al., 2011a, b; Mendez-Alonzo et al., 2012). Our plant hydraulics model parameterization is able to capture both the lack of a safety–efficiency trade-off, which occurs when stem and leaf economics are decoupled (Fig. 7c), and a strong safety–efficiency trade-off when stem and leaf economics are tightly coupled (Fig. 7f).

When we paired our sapwood PV data with data on xylem hydraulic safety, we found that $C_{lx,x}$ (xylem capacitance) and $P_{50,x}$ demonstrated significant evidence ($p = 0.02$) for a
trade-off in drought avoidance (increasing capacitance) and drought tolerance (increasingly negative $P_{50,x}$) (Fig. 8b). The strength of this relationship became marginally insignificant ($p = 0.05$) when considering $P_{50,x}$ values obtained with the DH method only, but was unaffected by the correction factor that we applied to the sapwood PV curves, remaining significant ($p = 0.02$ and $p = 0.01$ for all data and DH-limited data, respectively; Supplement Fig. S2.4). Because $C_{ft,x}$ and $P_{50,x}$ both derive directly from WD, our plant hydraulics model parameterization also follows this avoidance–tolerance trade-off (cf. thin line Fig. 8b).

Finally, joining of the leaf PV database with the xylem functional traits database demonstrated significant evidence in support of coordination between leaf and xylem drought tolerance, as given by significant relationships between leaf drought tolerance traits $\pi_{o,1}$ and associated turgor loss point $\pi_{tlp,l}$ with xylem drought tolerance traits $P_{88,x}$ and $P_{50,x}$, albeit with $R^2$ values no greater than 0.3 (Fig. 8c–f). This cross-tissue coordination is also preserved by our model parameterization, with WD driving most of the variation in this coordination space and LMA generating residual variation in leaf drought tolerance (thin lines, Fig. 8c–f).

4.2 Model experiments

4.2.1 Impact of plant hydraulics on size and light dependency of transpiration

Figure 9 shows how mean $\pm$ 1 SD diurnal cycles of transpiration change as a function of canopy position and wet/dry season for both TFS v.1 and TFS v.1-Hydro. The large increase in simulated transpiration in the dry season for both TFS v.1 and TFS v.1-Hydro (Fig. 9e–f) is driven by the comparatively large increase in incoming solar radiation, and hence absorbed radiation, due to a reduction in cloud cover...
Figure 6. Tropical synthesis showing the dual dependency of the leaf : sapwood area ratio ($A_l : A_s$) on both leaf mass per area (LMA) and tree height ($H$). (a) $A_l : A_s$ vs. tree height, replotted from Calvo-Alvarado et al. (2008). (b) Species-averaged LMA vs. $A_l : A_s$ from panel (a) overlain on a much broader dataset from the Amazon basin (replotted from Patiño et al., 2012). (c) The slope of the $A_l : A_s$–$H$ relationship in panel (a) vs. LMA. Circled datapoint is *Pentaclethra macroloba*, a compound-leaf species averaging 1600 leaflets/leaf and is excluded from the regression. The dependency of $A_l : A_s$ on LMA in panel (b) is thus implemented in the model via variations in the slope of $A_l : A_s$ with $H$ in panel (c); see Fig. 7g. Asterisk codes for significance are the same as in Fig. 2.

Figure 7. Plant hydraulics model parameterization as a function of plant traits leaf mass per area (LMA), wood density (WD), light-saturated photosynthesis rate ($A_{\text{max}}$), and size ($H$) resulting from the syntheses presented in Figs. 2–6 and summarized in Table 2. (a, b) Leaf and sapwood PV curves, respectively (Eqs. 1–3), which relate tissue water potential to relative water content (RWC). (c) Sapwood xylem vulnerability curves ($FMC_x$; Eq. 4) multiplied by maximum leaf-specific xylem conductivity based on a dataset of the joint distribution of WD and $A_{\text{max}}$ in an eastern Amazonian forest. Note that curves in panels (a) and (c) are also dependent on WD and $A_{\text{max}}$, respectively, in addition to the color scale shown. Panels (d–f) same as panels (a–c) except for an idealized distribution of plant traits in which trait variation occurs over a single axis (LMA ~ WD ~ $A_{\text{max}}$). (g) Leaf : sapwood area ratio ($A_l : A_s$) as a function of LMA and tree height. (h) Sapwood area ($A_s$) as a function of DBH and LMA, overlain with synthesis of independent measurements of $A_s$. Dashed red line in panel (h) represents theoretical maximum for $A_s$ (entirely sapwood).
4.2.2 Impact of plant hydraulics on the trait dependency of transpiration

Figure 10 shows how mean ±1 SE diurnal cycles of transpiration per unit crown area of three large (50–55 cm DBH and 26–27 m height) individuals change as a function of a single plant trait ($P_L$, WD, or LMA) and wet/dry season for both TFS v.1 and TFS v.1-Hydro. In most cases, the seasonal wet to dry increase in transpiration associated with incoming radiation is the same as for the previous model experiment. In the absence of plant hydraulics, leaf nutrients drive large differences in photosynthetic parameters and rates, which, by model design, then drive large differences in $g_{s,max}$ and transpiration rates (Fig. 10a, b). Because our plant hydraulic trait parameterization explicitly links hydraulic transport rates ($k_{s,max}$) to photosynthetic capacity, this trend is preserved in TFS v.1-Hydro, except that all trees experience hydraulic limitation (Fig. 10c, d). In contrast, large canopy trees with differences in WD and LMA in default TFS simulations demonstrated no differences in simulated transpiration dynamics (Fig. 10e, f). In the case of WD, adding plant hydraulics caused hydraulic limitation across all three individuals but surprisingly little to no divergence in simulated dynamics (Fig. 10g, h), despite large differences in plant hydraulic traits, which derive from WD (Table 2). This lack of a difference arises, however, because the effects of decreasing WD have opposing effects on midday leaf water potential via sapwood capacitance and xylem $P_{s,0}$ because of the drought avoidance–tolerance trade-off (Fig. 8b). In the case of LMA, dramatic differences emerged among the three individuals with differing LMAs (209, 95 and 47 g m$^{-2}$) upon the inclusion of plant hydraulics (Fig. 10k, l). This was driven by large differences in total tree aboveground hydraulic conductance ($K_{s,max,tree,ag}$), which arose because of large differences in $A_1: A_s$ that ranged from 0.1–1.8 m$^2$ cm$^{-2}$ (itself arising from the LMA differences; see Fig. 7g, h and Table 2). By Darcy’s law, a reduction in conductance results in a comparative reduction in water flux for a given water potential gradient.

(Fig. 9a–d) (Carswell, 2002; Fisher et al., 2007). At the level of individual tree crowns, canopy position is the dominant control over the amount of absorbed PAR (compare different colored lines in Fig. 9a–d). These trends remain when considering total tree transpiration (Fig. 9i, j, k, l), but variation in crown size adds significant variability in total water fluxes, especially for canopy trees (very small trees can still be in the canopy depending on subplot assignment). The effect of including plant hydraulics in TFS v.1-Hydro was to limit late morning and afternoon transpiration via the depletion of stored water within the canopy and tree stem, which caused midday declines in leaf water potential (see Fig. 10) and induced hydraulic limitation to water flux via the FMC$_{gs}$ term (see Eq. 8). Hydraulic limitation was highest in canopy trees, intermediate in subcanopy trees, and nonexistent in understory trees, but this limitation only occurred during the dry season when incoming radiation was sufficiently high to drive high potential evapotranspiration (compare the effect of including hydraulics in Fig. 9e and g vs. Fig. 9f and h), consistent with other studies (McDowell et al., 2005).
4.3 Model evaluation

TFS v.1-Hydro demonstrated a capability to represent lags in water flux throughout the leaf–stem–soil continuum (Fig. S3.1), with sap flux lagging leaf-level transpiration, and root uptake continuing into the nighttime hours, which is consistent with the diurnal dynamics of soil moisture shown in other ecosystems. In addition, as expected, the model predicted that soil most adjacent to absorbing roots experiences significantly lower rates of volumetric root water uptake as compared to more distal rhizosphere shells (Fig. S3.2), demonstrating the capability for the model to represent rhizosphere hydraulic limitation if soil gets sufficiently dry.

4.3.1 Diurnal dynamics of leaf and stem water potential

Observed diurnal dynamics of leaf and stem water potential ($\psi_{\text{leaf}}$ and $\psi_{\text{stem}}$, respectively) indicated, for all trees regardless of size, predawn ($\sim$ 06:00 am) values in the range of −0.1 to −0.4 MPa, whereas midday water potentials varied largely in concert with tree size (Fig. 11). The model captured the observed diurnal trends in both $\psi_{\text{leaf}}$ and $\psi_{\text{stem}}$ reasonably well, especially given the absence of any model tuning or tissue-level hydraulic trait differences among individuals. This suggests that the first-order control over the variation in $\psi_{\text{leaf}}$ among individuals was primarily due to the variation in radiation interception (both across seasons and within the canopy) and tree size, rather than tissue-level hydraulic traits. See Fig. S3.4 for an alternate version of Fig. 11 showing how root water potentials vary diurnally for these individuals.

Despite large (upwards of −4 MPa) differences in midday and predawn leaf values, observed $\psi_{\text{stem}}$ remained for the most part within −1 MPa of the observed $\psi_{\text{leaf}}$ values (Fig. 11). For the larger trees, observed midday $\psi_{\text{stem}}$ during the dry season was in the range of −2 to −3 MPa. In contrast, modeled $\psi_{\text{stem}}$ rarely went more negative than −1 MPa, suggesting that the model parameterization overestimates sapwood capacitance ($C_{\text{ft,x}}$).

4.3.2 Impact of model parameterization on fidelity of simulated midday leaf water potential

Modeled dynamics of leaf water potential at the Caxiuana site captured observed variation at both diurnal and seasonal timescales (Supplement Fig. S2.6). Throughout each day, reductions in leaf water potential lagged that of the stem and transporting root system, congruent with the buffering role of canopy and stem capacitance, with refilling of distal tissues occurring at night (results not shown). The model also simulated reductions in predawn leaf water potential in the dry
Figure 10. TFS simulated mean (±1 SE) diurnal cycles of transpiration of three large (50–55 cm DBH) individual trees receiving full sunlight but differing in one plant trait, stratified by wet/dry season and TFS version with (v.1-Hydro) and without (v.1) plant hydraulics implemented, at Caxiuana, Brazil. (a–d) Individuals differing in leaf $P$ exhibit different light-saturated photosynthesis rates ($A_{\text{max}}$) and consequently $g_{\text{ss, max}}$ and, in v.1-Hydro, $k_{\text{ss, max}}$. (e–h) Individuals differing in wood density WD exhibit no difference in maximum transpiration rate per unit crown area in v.1, but different rates in v.1-Hydro emerge due to differences in xylem safety ($P_{50,x}$) and, hence, stomatal control ($P_{50,gs}$). (i–l) Individuals differing in leaf mass per area LMA exhibit little difference in v.1 but large differences in v.1-Hydro, due to large differences in the leaf : sapwood area ratio ($A_l : A_s$) and, hence, total aboveground conductance ($K_{\text{max,ag}}$).

Table 3. Properties of simulated and observed trees given in Fig. 11.

<table>
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<th>Figure 11 panels</th>
<th>Tree ID*</th>
<th>Observed DBH (cm)</th>
<th>Simulated DBH (cm)</th>
<th>Simulated canopy layer</th>
<th>Simulated $A_l : A_s$ (m² cm⁻²)</th>
<th>Simulated $K_{\text{max,ag}}$ (kg s⁻¹ MPa⁻¹)</th>
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<tr>
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<td>C2</td>
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<td>2, 3</td>
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<td>0.064b</td>
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<td>e, f</td>
<td>C4</td>
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<td>1, 2</td>
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* As given in Fisher et al. (2006). b Two simulated trees were included in this size class. Value given is the average of the two trees.

Each of three successive model parameterization adjustments was able to make incremental improvements in the fidelity of modeled $\psi_{\text{leaf}}$, and these adjustments are informative for understanding the dominant controls over individual-level differences in $\psi_{\text{leaf}}$. First, when the model’s treatment of xylem conduit taper is turned off, the model in most cases overestimated the negative hydraulic impacts of tree height, with modeled midday $\psi_{\text{leaf}}$ falling below that which was observed (Fig. 12a). Re-instating xylem taper ameliorated the negative effects of tree height for some trees by increasing $K_{\text{max,tree},ag}$ and thus reducing the soil–leaf water potential gradient needed to maintain transpiration (Fig. 12b). Next, adjusting the canopy position of two of the four individual trees from a subcanopy (canopy layer is 2) light environment to an understory (canopy layer is 3) light environment reduced the incoming radiation for these trees and hence their...
transpiration rates and midday $\psi_{\text{leaf}}$ values (Fig. 12c). Finally, we found that the extremely negative midday $\psi_{\text{leaf}}$ values during the dry season of the largest tree (tree C3 of Fisher et al., 2006) could be better simulated by allowing its stomatal regulation to behave in a more anisohydric way (less stomatal regulation of leaf water potential) by setting its $P_{50,gs} = -5.0$ MPa.

### 4.3.3 Fidelity of modeled stand-level transpiration

Without plant hydraulics, TFS v.1 had a large positive bias in simulated dry season stand-level transpiration rates (Fig. 13a, b). Including plant hydraulics (TFS v.1-Hydro) largely eliminated this dry season positive bias when plot-mean trait values were used (Fig. 13a). However, TFS v.1-Hydro tended to underestimate observed transpiration year-round when each tree was assigned different trait values (Fig. 13b) according to the observed plot distribution (according to Fig. 7a–c).

## 5 Discussion

We present here a plant hydraulics model built from the bottom-up. By “bottom-up” we mean that this model is parameterized at the level of plant tissues (leaf, stem, roots) and then scaled to the whole-tree level using established theory about how size and aboveground-branching structures impact whole-tree function. We embedded this model within a size-structured forest simulator (TFS) providing critical input information to the plant hydraulics model on light environments of individual trees and community-level size and trait distributions (collectively these two models comprise TFS v.1-Hydro), but the plant hydraulics scheme is appropriate for inclusion within any demographic dynamic vegetation model. Our plant hydraulics model is also trait driven.
By “trait driven” we mean that individual trees take on distinct plant functional trait values (WD, LMA, N\textsubscript{L}) according to observed plot-level trait distributions and we have parameterized hydraulic traits from each tree’s assigned traits according to a pantropical empirical synthesis. Our empirical synthesis documented widespread correlations between plant hydraulic traits and the leaf and stem economic traits used in our model (Figs. 2–6), and we additionally showed that even where such correlations were weak, the hydraulic trait parameterization for our model respected observed ecological trade-offs and coordination among hydraulic traits (Figs. 7–8) known to be crucial to plant water use and survival. This provides encouragement to the growing research agenda to expand the role of plant functional traits in parameterizing Earth system models (ESMs) (Fisher et al., 2015; Sakschewski et al., 2015). The model also reasonably captures size- and light-related interindividual variation in water potential across individuals (Figs. 11–12) without need for significant model tuning. Finally, the model makes substantial improvements to TFS v.1 in terms of simulated transpiration rates (Fig. 13a, b), highlighting how plant hydraulics mediate the biosphere–atmosphere exchange of carbon and water. In conclusion, TFS v.1-Hydro represents a key and advanced model capability to represent differential performance of individual trees based on hydraulic traits, size and light environments (Figs. 9–10). Future work coupling the present scheme with community dynamics (mortality, growth and recruitment) has the potential to predict shifts in community trait distributions under changing moisture regimes, as has been observed or implied in studies of tropical forest species distributions and community dynamics (Engelbrecht et al., 2007; Fauset et al., 2012).

5.1 Underlying causes and implications of trait–trait relationships

The relationship of leaf π\textsubscript{0,1} (i.e., cell “saltiness”) with WD is robust, and suggests some degree of correspondence between sapwood and leaf osmoregulation at the cellular level, itself potentially a function of soil potassium (K\textsuperscript{+}) availability (Quesada et al., 2012), although there was not significant enough species overlap in our leaf and sapwood databases to directly assess whether π\textsubscript{0,1} tracks π\textsubscript{0,x}. An additional possibility is that species with a tendency to osmoregulate leaf tissue also have sapwood with higher WD and embolism resistance, though recent results challenge this explanation (Binks et al., 2016). The observation that leaf ε\textsubscript{1} (i.e., leaf cell “stiffness”) was coordinated more closely with stem WD rather than LMA (Fig. 2) may be due more to uncertainty in how ε\textsubscript{1} is calculated, either based on symplastic or total relative water content (RWC\textsuperscript{o} or RWC, respectively) (Bartlett et al., 2012), rather than strong evidence for stem over leaf traits governing leaf cell stiffness per se. There may be a mix of both methods used in calculating ε\textsubscript{1} in this dataset due to ambiguity in the original publications (M. Bartlett, personal communication, 2015), so the relationship of ε\textsubscript{1} with WD should be viewed as tentative.

The correspondence of sapwood PV parameters and associated capacitance with WD (Fig. 3) was in some cases particularly tight (r\textsuperscript{2} = 0.84 for sapwood capacitance C\textsubscript{fl,x}). Given the uncertainty associated with the vessel cutting artifact for which we corrected (Supplement S2), the exact relationships with WD should also be considered tentative, but the qualitative nature of WD as a first-order predictor of C\textsubscript{fl,x} should be robust. Nonetheless, the absence of data for sapwood PV traits for WD > 0.7 g cm\textsuperscript{-3} highlights a real data need, which is particularly apparent for some regions of the tropics, such as eastern and northeastern Amazonia, which have a large abundance of high wood density species (Baker et al., 2004; Quesada et al., 2012). Second-order predictors for sapwood capacitance have been identified in separate studies, which we have not explicitly dealt with here. Recently, Wolfe and Kursar (2015) identified a similar trend of C\textsubscript{fl,x} with WD, but additionally identified deciduousness as a modulator of C\textsubscript{fl,x}, with evergreen species tending to have lower C\textsubscript{fl,x} at a given WD.

The increase of xylem efficiency (k\textsubscript{s,max,x}) with A\textsubscript{max} we highlighted in our synthesis (Fig. 5) has already been identified in individual studies (Brodribb et al., 2002; Santiago et
Our synthesis of hydraulic traits suggests that a trade-off between drought avoidance (as given by $C_{\text{fl,x}}$) and tolerance ($P_{50,x}$) is more prominent than between xylem efficiency ($k_{s,max,x}$) and safety ($P_{50,x}$), though more data are needed to corroborate the limited data suggesting the former (Fig. 5). Partially responsible for this weak safety–efficiency trade-off is methodological variation in terms of how $k_{s,max,x}$ is measured (Gleason et al., 2016). However, there is undoubtedly a real component to this weak trade-off, as inter-vessel pit membrane thickness, which does not necessarily correlate with hydraulic conductivity, is a strong anatomical predictor of $P_{50,x}$ (Li et al., 2016). Perhaps of greater significance, however, is understanding patterns associated with the presence or absence of the safety–efficiency trade-off in particular bioclimatic regions. The correspondence of this trade-off with tropical dry forests where stem economic traits closely track leaf economic traits (Marksteijn et al., 2011a, b; Mendez-Alonzo et al., 2012), in contrast to their wetter counterparts (Baraloto et al., 2010), suggests that the hydraulic safety–efficiency trade-off may be an underlying mechanism for the extent to which stem and leaf trait coupling is observed.

5.2 Model limitations

Our analysis reveals some limitations of our particular approach to trait-driven plant hydraulics modeling, namely that not all trait–trait relationships are of sufficient predictive accuracy to enable simulations with low uncertainty (most notably leaf drought tolerance traits and the leaf : sapwood area ratio; see Figs. 2 and 6). This is a likely outcome of the reality that emergent properties of biological systems arise from the combination of multiple traits and their associated trade-offs; thus, no single trait is likely to explain the majority of an emergent process (Mencuccini et al., 2015). More work identifying general plant trait–hydraulic trait relationships is justified, but should consider incorporation of multiple traits for improved realism and accuracy, and should emphasize understanding how trees as an integrated system function.

One consequence of using the empirical trait–trait correlations given here is to either under- or overestimate the range of any given hydraulic trait in a simulated community of individuals, depending on the hydraulic trait of interest. For example, in our model experiment in which we varied a single trait (Fig. 10), variation in xylem-specific hydraulic conductivity ($k_{s,max,x}$; Fig. 10c, d), which was derived from the full range of observed $P_L$ (and, hence, modeled $A_{max}$) at this site, was limited to only 1.1–2.9 kg m$^{-1}$ MPa$^{-1}$ s$^{-1}$, which is only $\sim$15% of the range of observed $k_{s,max,x}$ values for tropical moist forests (cf. green points, Fig. S3). Indeed it seems unlikely that the observed range of $k_{s,max,x}$ at this site would be this small. On the other hand, the range of $A_1 : A_s$ predicted by LMA and tree height (Fig. 6 and ?g) is likely overestimated; we found that individuals with $A_1 : A_s$ greater than the maximum observed $A_1 : A_s$ (1.8 m$^2$ cm$^{-2}$) for whole trees in the Calvo-Alvarado et al. (2008) study were unviable (whole-tree conductance was insufficient for any appreciable tree transpiration to occur). Indeed, our result showing that a slight negative bias results when all individuals receive distinct trait values according to the community trait distribution vs. a single community-mean value (Fig. 13a vs. 13b) follows as a consequence. Furthermore, in contrast to our simple implementation of hydraulic architecture, $A_1 : A_s$ may vary substantially throughout a tree’s crown, at different branching levels, or different light environments (see Schuldt et al., 2011, as an example), given the deviations from self-similar, volume-filling branching and area-preserving $A_s$ that have been observed (Whitehead et al., 1984; Bentley et al., 2013; Smith et al., 2014).

5.3 Practical implications

Because TFS v.1-Hydro parameterized the majority of tree hydraulic traits at the scale of plant tissues and built the size-scaling into the model (rather than specifying hydraulic properties at the whole-tree level), it has great potential for being used as a tool in data assimilation and other inverse approaches when in situ field data on sap flow, water potential, or water content data are available at one or multiple points within individual trees. This may aid in deciphering unobserved or unobservable processes, such as diagnosing where within the soil–plant continuum hydraulic limitation or embolism may be occurring (branches, stems, or fine roots) or how much stored water is used for daily transpiration.

A second benefit to its method of construction is that the constitutive equations of the model (i.e., the relationships between water potential and water content, and between hydraulic conductivity and water potential) use formulations from plant physiology (such as PV theory) and, as a consequence, have parameters (Table 2) that are empirically mea-
surable and biologically interpretable. Hence, as more of these hydraulics data become available, this model can readily incorporate this information by either substituting the empirical trait relationships with measured quantities or updating the empirical hydraulic trait relationships.

Finally, the “fraction of maximum stomatal conductance” ($FMC_{ps}$) encapsulates the hydraulic limitation of each tree and is implemented in a similar fashion to traditional soil-based “beta” approaches (Powell et al., 2013). While this approach might seem simplistic relative to the comparative rigor of the hydraulics scheme, we note that (1) the present function (Eq. 8) is based on leaf water potential, a new prognostic, non-steady-state metric of tree water status that integrates individual-level differences in hydraulic traits, size, and microenvironment (e.g., light) in addition to soil water potential; (2) there is no consensus as to the exact mechanism (direct or indirect) by which stomata respond to leaf water potential—recent theoretical developments on stomatal function are largely restricted to the effects of temperature and humidity alone (Buckley, 2005; Peak and Mott, 2011; Nikinmaa et al., 2013); and (3) this approach simplifies the interface between the hydraulics module and the host model.

5.4 Future directions

In light of the limitations to our empirical approach for parameterizing hydraulic traits via more commonly measured traits, one potential alternative for parameterizing community-level hydraulic trait variation is to do so independent of other traits employed by the model. One approach is to use the Taylor and Thompson (1986) resampling algorithm to resample individuals or species with values in multiple trait databases, which were used in this paper, but sparse database overlap may limit the utility of this approach. Alternatively, region-specific distributions of individual hydraulic traits from the databases presented here could be coupled with empirical relationships among hydraulic traits as given by our synthesis of coordination and trade-offs among hydraulic traits (Fig. 8). This analysis suggested that two independent hydraulic trait spectra may be sufficient: one safety–efficiency or avoidance–tolerance trait spectrum and potentially a stomatal hydraulic safety spectrum. More work is needed to clarify these hydraulic coordination and trade-off surfaces and, in particular, to quantify how much residual variation is real or due to measurement biases.

From an empirical perspective, explaining the remaining 16–95% of unexplained variance in hydraulic traits shown here should be of high priority. We suggest some additional traits that show promise. The volumetric fractions of airspace, water, and solid material in leaves and stems (sensu Roderick et al., 1999; Roderick and Berry, 2001) are one such trait. For leaves, these volumetric fractions have been shown to explain much more variance in leaf PV traits than LMA and WD (Sack et al., 2003; Bouche et al., 2015; Buckley, 2015; Scoffoni, 2015). While not as routine of a measurement in comparison to LMA, these measurements are not as intensive as the full PV curve. Second, identifying the strongest wood anatomical correlates of xylem vulnerability to embolism ($P_{s0,x}$) should also be of high priority, such as the thickness of the intervessel pit membranes (Choat et al., 2008; Jansen et al., 2009), as well as improving understanding of the air-seeding mechanism (Schenk et al., 2015; Jansen and Schenk, 2015) Third, and perhaps most important for whole-tree hydraulics, is more data on hydraulic architecture. Tuning hydraulic architecture in our dynamic model proved to be a key “knob” for matching observations. Additional measurements of $A_{s} : A_{x}$, xylem taper, and crown-branching patterns should not be conducted in isolation, rather in tandem with each other and with other leaf and stem economic traits (in particular LMA, as suggested by Fig. 6b) to isolate physical constraints on structure and the dynamics of whole-tree transpiration (Smith et al., 2014). The paired relationship of $A_{s} : A_{x}$ with xylem taper as integrative measures of whole-tree hydraulic regulation is virtually nonexistent for tropical forests, although colocated measurements of these two traits exist (Zach et al., 2010; Horna et al., 2011).

For modeling hydraulic impacts on stomatal closure, we favored the use of a “stomatal vulnerability function”, but this is one possibility among many. Specifically, multiple studies have found that branch $\psi$ rarely falls below that at which stem capacitance begins to decrease precipitously (Meinzer et al., 2008b; Wolfe and Kursar, 2015), and stomatal regulation has also been observed to track leaf turgor loss points (Brodribb et al., 2003). Whether stomatal closure in response to water stress in tropical forests forms an entirely distinct “safety margin” axis (Martinez-Vilalta et al., 2014; Skelton et al., 2015) or maps on to other hydraulic traits remains to be elucidated. A recent review highlights the interacting roles of both hydraulic supply and vapor pressure deficit in controlling stomatal responses, which we have not yet accounted for (Sperry and Love, 2015). We also did not include into our model variability in leaf-level hydraulic conductance, or comparative values for root traits. While terminal branches and leaves account for a majority of total plant resistance due in part to xylem taper, roots may be more vulnerable to embolism for the same reason and therefore represent a weak link in drought-induced mortality; these should be topics of future work.

6 Conclusions

We present a plant hydraulics model based on biologically interpretable and measurable plant hydraulic traits and rooted in established plant physiology theory. It is capable of scaling tissue-level hydraulic traits to whole-tree hydraulic function. Embedding it within an individual tree trait-driven model allowed us to explore how individual-level variation in hydraulic traits (including tree size) interacts with light envi-
environments to drive differences in hydraulic function across individuals. Addition of plant hydraulics made substantial improvements to modeled transpiration fluxes during periods of high incoming radiation. Our synthesis of data for the tropics on a wide range of hydraulic traits needed to parameterize the model allowed us to develop empirical relationships among commonly measured plant traits and less common hydraulic traits. These empirical relationships are analogous to pedotransfer functions (Clapp and Hornberger, 1978; Cosby et al., 1984) developed for predicting soil hydraulic properties as a function of soil textural “traits”. While they should be refined as more data become available, they are an important first step towards representing ecological dimensions of hydraulic trait variation in process ecosystem models. Critically, this individual- and trait-driven framework provides a test bed for identifying both critical processes and functional traits needed for inclusion in coarse-scale Earth system models (ESMs). Likely some degree of simplification of the present approach will be required upon implementation in ESMs; nonetheless we expect that inclusion of trait-driven plant hydraulics schemes will lead to reduced uncertainty in the future state of tropical forests under climate change.

7 Code availability

The JAVA source code for TFS v.1-Hydro can be obtained from the corresponding author upon request.

8 Data availability

The Supplement S4 describes supplementary data for this paper. Leaf PV data (Figs. 2 and 8) that were newly extracted from publications are available as .csv data files as described under the heading “Leaf PV database”. Leaf PV data originating from pre-existing leaf PV databases are available by accessing the original articles cited under the heading “Leaf PV database”. Sapwood PV data (Figs. 3, S2.1, and S2.3) are available as .csv data files as described under the heading “Sapwood PV database”. Sapwood area data (Fig. 7h) are available as .csv data files as described under the heading “Sapwood Area database”. Xylem hydraulic trait data (Figs. 4, 5, 8, and S2.2) that were newly extracted from publications are available as .csv data files as described under the heading “Xylem Functional Traits Database”. Xylem hydraulic trait data originating from the TRY archive can be accessed from this archive (http://www.try-db.org); the references of data that are used in analyses are given under the heading “Xylem Functional Traits Database”. Sapwood area and sapwood PV databases are given in Christoffersen (2016a) and Christoffersen (2016b), respectively.

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