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1 **Viewpoint:**

2 **Toward a coordinated understanding of hydro-biogeochemical root functions in tropical**
3 **forests for application in vegetation models**

4

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66

67 **Social Media**

68 **On X (previously Twitter): @DanielaCusack and @TropiRoot**

69 **Summary**

70 Tropical forest root characteristics and resource acquisition strategies are underrepresented in
71 vegetation and global models, hampering prediction of forest-climate feedbacks for these carbon-
72 rich ecosystems. Lowland tropical forests often have globally unique combinations of high
73 taxonomic and functional biodiversity, rainfall seasonality, and strongly weathered infertile soils,
74 giving rise to distinct patterns in root traits and functions compared with higher latitude
75 ecosystems. We provide a roadmap for integrating recent advances in our understanding of tropical
76 forest belowground function into vegetation models, focusing on water and nutrient acquisition.
77 We offer comparisons of recent advances in empirical and model understanding of root
78 characteristics that represent important functional processes in tropical forests. We focus on: 1)
79 fine-root strategies for soil resource exploration, 2) coupling and tradeoffs in fine root water versus
80 nutrient acquisition, and 3) above-belowground linkages in plant resource acquisition and use. We
81 suggest avenues for representing these extremely diverse plant communities in computationally
82 manageable and ecologically meaningful groups in models for above- and belowground hydro-
83 nutrient functions. Tropical forests are undergoing warming, shifting rainfall regimes, and
84 exacerbation of soil nutrient scarcity caused by elevated atmospheric CO₂. Accurate model
85 representation of tropical forest functions is crucial for understanding interactions of this biome
86 with climate.

87

88 **Key Words**

89 Fine roots, tropical forests, ecosystem vegetation models, plant functional types, root trait clusters,
90 hydraulics, nutrient acquisition, phosphorus uptake

91 **Introduction**

92 Tropical forests are poorly characterized in vegetation models relative to other ecosystems, and
93 representation of root function lags that of aboveground function (Warren *et al.*, 2015; Bonan &
94 Doney, 2018). Tropical forests have the highest rates of net primary production (NPP) on Earth
95 and contain ~30% of terrestrial carbon (C) stocks (Field *et al.*, 1998; Jobbágy & Jackson, 2000;
96 Hengl *et al.*, 2017), with at least 36% of tropical forest NPP allocated belowground (Aragao *et al.*,
97 2009; Malhi *et al.*, 2011; Huasco *et al.*, 2021). Fine roots are typically considered the absorptive
98 portion of the root structure, which absorb nutrients and water (Guo *et al.*, 2008; McCormack *et*
99 *al.*, 2015, Table 1). These are typically classified as <2 mm diameter and include branching orders
100 1 – 3 (e.g., root tips are first order, e.g., https://youtu.be/q_ICrIL62qg, (Freschet *et al.*, 2021a).
101 Understanding and representing tropical forests' water and nutrient cycling is of particular
102 importance in the context of changing tropical forest rainfall regimes and warming (IPCC 2021)
103 and increased relative nutrient scarcity brought on by accelerated photosynthesis of plants grown
104 under elevated atmospheric carbon dioxide (CO₂) concentrations (i.e., CO₂ fertilization) (Hungate
105 *et al.*, 2003; Fisher *et al.*, 2012; Fleischer *et al.*, 2019).

106

107 Tropical forests in particular are distinct from higher latitude ecosystems across several abiotic
108 and biotic dimensions, giving rise to unique patterns of root traits and functions. The unique
109 aspects of tropical forests include combinations of high plant diversity (Eiserhardt *et al.*, 2017),
110 seasonality dominated by rainfall rather than temperature changes, and the predominance of
111 lowland tropical forests on strongly weathered soils poor in phosphorus (P) and base cations, which
112 represent >50% of tropical forests (Holzman, 2008), and commonly leads to P or multi-nutrient
113 limitation to NPP (Vitousek & Sanford, 1986; Cunha *et al.*, 2022). Tropical forests also have large

114 variation in ecosystem characteristics, including exceptions to the above trends such as
115 monodominant stands of particular species or families (e.g., Dipterocarpaceae) (Janzen, 1974; Hart
116 *et al.*, 1989; Peh *et al.*, 2011), high-fertility soils (e.g., Quesada *et al.*, 2011; Cusack *et al.*, 2018),
117 a lack of marked seasonality in rainfall, and/or strong sunlight seasonality because of changes in
118 cloud cover (Yang *et al.*, 2021). Thus, tropical forests have high alpha and beta diversity (Condit
119 *et al.*, 2002), both for organisms and ecosystem characteristics, which create empirical and
120 modeling challenges for characterizing and condensing species into meaningful groups.

121
122 Large-scale models have often worked well with only rudimentary root system functionality or
123 none at all (Matamala & Stover 2013), but this functionality can break down when models are
124 confronted with global change factors that alter relationships among soil, plants, and atmosphere
125 (e.g., Zaehle *et al.*, 2014). To address these challenges, vegetation models typically group plants
126 according to common characteristics and functions to simplify the diversity in natural ecosystems
127 (Walker *et al.*, 2014; Medlyn *et al.*, 2015; Fer *et al.*, 2021; Kyker-Snowman *et al.*, 2022), using
128 plant functional type (PFT) groupings. These have generally focused on aboveground traits and
129 temperate ecosystems (Wullschleger *et al.*, 2014; Warren *et al.*, 2015). Several leading vegetation
130 models are now increasing the representation of root functions and inclusion of root characteristics
131 as part of PFTs (Table 2). Model comparisons for tropical forests indicate that including P
132 availability, which has been excluded for representation of temperate ecosystems, can improve
133 representation of outcomes like NPP (Fleischer *et al.*, 2019; Yang *et al.*, 2019; Braghiere *et al.*,
134 2022; Nakhavali *et al.*, 2022). The time is now ripe to bring together these areas of model
135 development to improve representation of tropical forests: root functional representation, and
136 inclusion of key resource constraints in tropical forests.

137

138 An alternative to the PFT approach is "trait-flexible" models, in which traits are re-assigned at
139 every generation to new individuals recruiting into the population, rather than being fixed up-front
140 at the beginning of a simulation as in PFT-based models. Hence, such approaches allow for models
141 to dynamically consider the full trait spaces in a more flexible way (Scheiter *et al.*, 2013;
142 Sakschewski *et al.*, 2015). For example, trait-flexible modeling for the Amazon basin provided
143 greater diversity of belowground trait combinations in response to water scarcity than with PFT
144 approaches (Rius *et al.*, 2023), making this approach attractive for application to these high
145 diversity ecosystems where empirical knowledge about trait combinations is limited. However,
146 most vegetation models representing hydro-biogeochemical functions use the PFT approach. Both
147 the PFT and the trait-flexible modeling approaches would benefit from more accurate
148 representation of critical belowground functions in tropical forests, improving outcomes like NPP
149 and responses to global change.

150

151 This Viewpoint provides a roadmap for strengthening our empirical understanding and model
152 representation of the unique root functional characteristics of tropical forests (Fig. 1). We focus on
153 fine roots, including biomass and other traits, with attention to coarse roots (>2 mm diameter)
154 when relevant. We present: **1)** an overview of unique root characteristics in tropical forests in
155 relation to resource acquisition (reviewed in depth in Cusack *et al.*, 2021). **2)** A comparison of our
156 empirical understanding of tropical fine-root function versus root representation in a sampling of
157 leading vegetation models, including the topics: a) soil exploration, b) coordination and tradeoffs
158 in nutrient vs. water acquisition, and c) above-belowground functional linkages for nutrient and
159 water uptake and use. **3)** An assessment of commonly measured tropical root characteristics that

160 are not yet enough understood or are not functionally relevant for model inclusion. Based on this
161 assessment, we call for the development of more balanced above-belowground whole-plant
162 functional types and trait clusters to represent key functions of tropical forests, particularly in
163 relation to P and multi-nutrient acquisition, as well as drought resistance.

164

165 **How are fine-root strategies and functions different in tropical forests?**

166 Tropical forests have distinct belowground characteristics relative to other ecosystems, in part
167 because of the unique resource constraints common in tropical forests. First, tropical evergreen
168 forests have the largest stocks of fine-root biomass globally (Jackson *et al.*, 1996). Fine-root
169 production rates are also higher and turnover times are faster in tropical forests versus other forests
170 (Cusack *et al.*, 2021), following trends for tropical forest NPP. For example, tropical forest fine-
171 root productivity in surface soils averaged $596 \text{ g m}^{-2} \text{ y}^{-1}$ versus $428 \text{ g m}^{-2} \text{ y}^{-1}$ in temperate forests
172 and $311 \text{ g m}^{-2} \text{ y}^{-1}$ in boreal forests, and annual root turnover times averaged 1.4 y^{-1} in tropical
173 forests versus 1.2 y^{-1} in temperate forests and 0.8 y^{-1} in boreal forests (Finer *et al.*, 2011). The large
174 and dynamic stocks of root biomass in tropical forests make them important in the global C cycle,
175 since root turnover provides a principal input to the very large soil C stocks in tropical forests
176 (Rasse *et al.*, 2005). The outsized importance of tropical forests in the global C cycle provides
177 further motivation for accurately understanding tropical forest belowground function and
178 representation in vegetation models.

179

180 Second, tropical forest roots are more diverse than in other ecosystems across several axes. Similar
181 to the high plant species diversity common in tropical lowland forests, these ecosystems have the
182 highest diversity in fine-root morphological traits (Ma *et al.*, 2018, but see Carmona *et al.*, 2021).

183 Fewer plant species have been characterized for root traits in tropical forests compared with other
184 biomes (Fig. 2), in part because of the sheer diversity of co-existing species with entangled root
185 systems per unit area. Still, evidence using global databases suggests that tropical species
186 contribute at least 23% of the unique root trait combinations globally (Guerrero-Ramirez *et al.*,
187 2021). Related to high species diversity, fine-root traits are less phylogenetically constrained
188 within taxonomic levels compared to other ecosystems globally (Valverde-Barrantes *et al.*, 2021;
189 Weemstra *et al.*, 2023), Asefa *et al.*, 2022). An example in these studies is the "magnoliid" type of
190 root (i.e., thick, fleshy roots) that is largely limited to Magnoleaceae in temperate ecosystems, but
191 is found across multiple families in the tropics (e.g., Moraceae, Malvaceae, and Sapotaceae). Root
192 traits can also be diverse over small spatial scales in tropical forests, with high variation in fine-
193 root traits found within and among individuals of the species, as well as among species (in Box 1),
194 even while the large bioregions of the tropics have some separation in root traits (Addo-Danso *et al.*,
195 2020). Overall, tropical forests appear to have greater variation and more unique combinations
196 of root traits, both at species and community scales, compared with temperate ecosystems,
197 presenting a special challenge to vegetation modelers.

198

199 Third, fine-root strategies are organized around different resources in many tropical forests
200 compared with temperate biomes. Specifically, soil moisture variation and P scarcity appear to
201 drive tropical forest root dynamics and traits (reviewed in Dallstream *et al.*, 2023; Cusack *et al.*,
202 2021), rather than temperature fluctuations and N scarcity as in many higher latitude ecosystems.
203 Associations with mycorrhizal symbionts in tropical forests are broadly linked to P and water
204 acquisition and include both arbuscular mycorrhizal fungi (AMF) and ectomycorrhizal fungi
205 (ECM). Rather than the temperate-ecosystem paradigm of AMF promoting fast decomposition

206 and nutrient cycling versus ECM promoting slow nutrient cycling (Cornelissen *et al.*, 2001;
207 Phillips *et al.*, 2013; Read *et al.*, 2017; Averill *et al.*, 2019; but see Weemstra *et al.*, 2016), in
208 tropical forests, both types of mycorrhizal association have been related to fast and slow nutrient
209 cycling (Chuyong *et al.*, 2000; Keller & Phillips, 2019; Weemstra *et al.*, 2020). Also, in contrast
210 to obligate N fixation by actinorhizal N-fixing trees dominant in temperate and boreal biomes,
211 rhizobial N-fixing trees common in tropical forests can down-regulate N fixation (facultative
212 fixation) (Barron *et al.*, 2011; Menge *et al.*, 2014). Thus, tropical forest root symbionts respond to
213 different types of nutrient limitation with distinct strategies compared with root symbionts in
214 higher latitude ecosystems.

215

216 **Integrating multi-functional tropical root representation into vegetation models**

217 We now compare and synthesize current empirical and model understandings of tropical root
218 functions. We organize this section around: 1) root characteristics with strong empirical support
219 for a functional role, and which thus should be prioritized for model integration, versus 2) root
220 traits that are commonly measured but do not yet clearly indicate a root function, or which lack
221 clear relationships to resource availability, and thus are not (yet) suited for model integration. The
222 first part highlights three important functional aspects of roots: a) general soil exploration for
223 resource acquisition, b) coordination and tradeoffs for root nutrient versus water acquisition, and
224 c) above-belowground functional linkages in water and nutrient uptake and use. We consider both
225 the quantity and spatial deployment of roots as well as their activity (Zhang *et al.* 2023). For each
226 of these three areas we describe i.) empirical advances and understanding, ii.) current model
227 representation, and iii.) avenues for model improvement and data needs. We do not advocate that
228 models incorporate all root traits and functions, which would unnecessarily complicate them and

229 increase uncertainty. Rather, we attempt to identify data that is promising for improving functional
230 representation, and model components that are confirmed or at odds with field data following
231 Medlyn *et al.* (2015).

232

233 We summarize empirically measured root characteristics (Table 1) alongside an assessment of root
234 function representation in 15 leading vegetation models (Table 2). These models include examples
235 linked to global Earth System Models (ESMs), demographically resolved vegetation models (e.g.,
236 representing forest age and structure), and individual-based models. We compare how root
237 characteristics are emphasized in empirical and modeling research (Table 3), showing that some
238 functional root characteristics are understudied relative to their representation in models, while
239 other well-characterized tropical root functions are under-developed in models.

240

241 **Root traits strongly linked to tropical forest function – Ripe for models**

242

243 Dynamic soil exploration: *empirical advances*

244

245 Root characteristics like biomass and depth distribution are clearly linked to soil exploration for
246 resources (Fig. 1), with the largest availability of species-level data from the wet tropics for root
247 biomass, production, turnover, and specific root length (SRL, length/mass) (Fig. 2, Guerrero-
248 Ramírez *et al.*, 2021). Higher SRL increases the volume of soil explored per unit of root biomass
249 (McCormack *et al.*, 2015; demonstrated in <https://youtu.be/uHZqG5eKShI>). The most prevalent
250 patterns of allocation to root biomass for soil exploration in tropical forests (recently reviewed by
251 Cusack *et al.*, 2021) are: 1) relatively greater root biomass and root production rates in infertile

252 surface soils versus fertile surface soils, likely for rapid uptake of scarce mineral nutrients released
253 from litter decomposition; 2) relatively greater root biomass in surface soils in wetter versus drier
254 conditions, likely because extreme drying in tropical forests causes surface root death; 3) faster
255 fine-root turnover in wetter versus drier conditions and in fertile versus infertile soils, likely
256 indicating a less conservative plant life strategy when resources are abundant; 4) greater fine-root
257 SRL under resource scarcity, both for dry versus wet conditions and infertile versus fertile soils,
258 likely indicating maximization of soil explored per unit biomass; 5) greater root production rates
259 in the subsoil versus surface soils under dry conditions, likely for deep water acquisition. These
260 comparisons were true both across biogeographic gradients and experimental treatments that
261 varied the availability of rock-derived nutrients like P and potassium (K) (e.g., Wurzburger &
262 Wright, 2015; Cusack & Turner, 2021; Reichert *et al.*, 2022), and across seasonal or drought-
263 induced soil moisture variation (e.g., Kummerow *et al.*, 1990; Janos *et al.*, 2008; Metcalfe *et al.*,
264 2008). These soil exploration patterns of root biomass, production, turnover, and SRL are the best
265 supported by the literature for tropical forests.

266

267 Dynamic soil exploration: *model representation*

268

269 Among the 15 models reviewed here (Table 2), root representation was generally implemented as
270 less dynamic in response to moisture or nutrient availability than suggested by the empirical
271 research synthesized above. For example, root turnover was a constant value in the models we
272 assessed. Only two of the models allowed maximum rooting depths to change with tree size (i.e.,
273 size-dependent rooting depth), even though 13 of the models had the capacity to resolve tree size

274 (Table 2). None of the models allowed vertical root depth distributions to respond to changes in
275 soil moisture or nutrient availability.

276

277 Dynamic soil exploration: *avenues for model improvement*

278

279 Enabling individual-, cohort-, or PFT-specific rooting distributions and depths, and related
280 resource partitioning is a forefront for model development, which could build on the vertically
281 variable root allocation scheme of Drewniak (2019). Under this type of representation, different
282 plant groups in the community would have different strategies in accordance with some defined
283 resource strategy, which could include coordination between above- and below-ground traits (see
284 section above). Incorporating belowground resource partitioning would allow for a more holistic
285 differentiation between resource-acquisitive vs. -conservative strategies, as well as contrasting
286 strategies for nutrient vs. water acquisition and drought tolerance (see section above). In addition
287 to variable rooting depth by PFTs, increasing model capacity for root systems and functions to
288 respond dynamically to resource changes is an ongoing challenge for vegetation models (Wang *et*
289 *al.*, 2023). A particular challenge is posed by model structures that are not spatially explicit within
290 grid cells and a given soil layer (Table 2; the gap models reviewed are only spatially explicit
291 aboveground), such that resource partitioning is not possible belowground and resources are shared
292 by all members of the community. Innovative model approaches, which allow for incomplete
293 resource sharing across individuals, cohorts, and/or PFTs while still maintaining mass balance
294 would enable resource-conservative strategies as PFTs to emerge through trait filtering (Scheiter
295 *et al.* 2013). For example, a fraction of the total resource pool could be allocated as PFT-specific
296 (non-shared) and the remainder as shared across the community. Such model developments could

297 be complemented with empirical research, such as species responses to nutrient additions in the
298 field using identification approaches (e.g., DNA barcoding; Jones *et al.*, 2011). This would help
299 assess root exploration patterns and flexibility across species, and could inform the creation of
300 species clustering or PFTs in models.

301

302 Coupled hydro-biogeochemical strategies: *empirical advances*

303

304 There are very few empirical data linking root water and nutrient acquisition strategies in tropical
305 forests, but there have been advances to identify clusters of root traits for nutrient acquisition. This
306 recent work could be built on to include clusters of belowground hydraulic traits (as identified
307 above, e.g., rooting depth, root embolism vulnerability). Much of the nutrient acquisition trait work
308 in tropical forests has been for P, developing clusters of traits, or “syndromes,” targeted at P
309 acquisition. Plant P acquisition strategies include different combinations of root phosphatase
310 production, root branching ratios, SRL, mycorrhizal symbioses, root hair length and density, and
311 organic exudates to promote mineralization by decomposers (Ushio *et al.*, 2015; Weemstra *et al.*,
312 2016; Freschet *et al.*, 2021b, exudate measurement demonstration:
313 <https://www.youtube.com/watch?v=n0CQ0lo7pbs>). A framework grouped these P acquisition
314 strategies into broader root P “syndromes” for tropical forests, identifying sets of root
315 morphological traits and mycorrhizal types that are often found together, and provide unique
316 strategies for P acquisition from mineral and organic forms (Dallstream *et al.*, 2023). For example,
317 one tropical forest study identified clear tradeoffs in P acquisition strategies among tree species,
318 such as high fine-root phosphatase activity versus increased mycorrhizal hyphal length (Zhu *et al.*,
319 2023), although morphological tradeoffs were less clearly linked to P acquisition. We have yet to

320 formulate mathematical response surfaces defining which traits are expressed under what nutrient
321 conditions, which would be most useful for models. Such frameworks could be expanded to
322 include strategies for acquisition of other nutrients and water to develop holistic hydro-
323 biogeochemical functional types.

324

325 Some work in the tropics has explored plant tradeoffs for the acquisition of different nutrients. It
326 was proposed that N fixation and P acquisition are coordinated in P-scarce tropical forests, because
327 phosphatase enzymes are N-rich proteins (Houlton *et al.*, 2008). Studies in Costa Rica (Nasto *et al.*
328 *et al.*, 2014; Soper *et al.*, 2019) and Panama (Nasto *et al.*, 2014; Batterman *et al.*, 2018) found mixed
329 support for a relationship between N fixation and root phosphatase activity, indicating that other
330 P acquisition strategies such as mycorrhizal symbiosis and fine-root production should also be
331 assessed for coordination with N acquisition (Lugli *et al.*, 2020; Allen *et al.*, 2020; Braghieri *et al.*
332 *et al.*, 2022; Reichert *et al.*, 2022). New data presented here from Panama and Singapore demonstrate
333 variation in nutrient uptake rates for different nutrients, with some links to root morphological
334 traits that could be used to further develop resource acquisition syndromes (Box 3, method
335 demonstration (<https://youtu.be/4atZ3E0NrX4>). Because direct nutrient uptake measures at the
336 root system level are destructive and difficult to scale up (e.g., Cornelissen *et al.*, 2001), more
337 work is needed to explore if they can be related to surrogates, such as lab observations linking P
338 uptake rates to root phosphatase activity (Lee, 1988), and root phosphatase relationships with
339 mycorrhizal colonization, root branching ratio (Yaffar *et al.*, 2021), SRL, and other root
340 morphological traits (Lugli *et al.*, 2020; Cabugao *et al.*, 2021; Han *et al.*, 2022, Box 3), as well as
341 responsiveness of these traits to soil P availability (Ushio *et al.*, 2015; Guilbeault-Mayers *et al.*,
342 2020; Cabugao *et al.*, 2021; Lugli *et al.*, 2021).

343

344 Coupled hydro-biogeochemical strategies: *model representation*

345

346 Among the root traits included in the 15 models assessed here (Table 2), water stress or water
347 uptake was represented in 13 models versus only five models that represented nutrient acquisition
348 (N or P), with representation of P dynamics particularly lacking. Similar to the empirical
349 disconnect between nutrient and water acquisition research (discussed above), none of the models
350 explicitly represented coupled hydro-biogeochemical cycling, so we summarize water and nutrient
351 acquisition separately, and generally call for greater coordination of these two areas of model
352 development.

353

354 Overall, 13 of the 15 models represented plant hydraulic traits (Table 2). The most common trait
355 representing plant hydraulic function was maximum rooting depth (in 12 of 15 models, Table 2),
356 which was either a constant (four models) or a PFT characteristic and was not responsive to
357 changes in moisture. The next most common hydraulic parameters were “water stress factor”
358 (related to soil moisture, in 10 models), followed by water uptake rate (six models, Table 2). The
359 model with the broadest representation of plant hydraulic traits was FATES-Hydro (with PARTEH
360 module), which additionally represents root hydraulic resistance, embolism vulnerability, fine-root
361 radius, and permeability. Comparing the models with plant hydraulic traits emphasized by
362 empiricists, root phenology, root hair length and density, and mycorrhizal symbiosis were not used
363 to represent plant hydraulics in the models reviewed here (Table 3).

364

365 Nutrient uptake processes were represented in fewer models compared with plant hydraulics, with
366 only six of the models representing some aspect of nutrient uptake (Table 2). Root exudation of
367 non-structural carbohydrates was linked to priming and nutrient availability in three of the models,
368 and two models had some representation of symbiotic nutrient uptake, including BNF and
369 mycorrhizal nutrient uptake (Table 2). Representation of N acquisition processes was more
370 developed than P acquisition (Table 2). The most common nutrient parameter functionally related
371 to nutrient uptake in the models was the rate of N uptake, which was responsive to changes in soil
372 nutrient availability (five models), followed by the rate of P uptake (four models, Table 2). Some
373 of the models employed constant nutrient uptake parameters based on diffusion and kinetics, and
374 others accounted for chemical interactions of soil nutrients with minerals and soil microorganisms
375 (Thum *et al.*, 2020; Yu *et al.*, 2020). For example, LM4.1-BNF included many parameters for
376 modeling N uptake (Table 2), including passive nutrient uptake (via transpiration stream), active
377 uptake (via a C cost and Michaelis-Menten dynamics), and symbiotic nutrient acquisition.
378 Meanwhile, P uptake was represented only in four of the models using just one parameter (P uptake
379 rate). Some root characteristics that are empirically related to resource acquisition were included
380 in the models, but without nutrient functionality. For example, vertical root biomass distribution
381 was in 12 of the 15 models (Table 2); however, this parameter was a PFT characteristic and not
382 responsive to changes in resource availability. Of the root characteristics commonly related to
383 plant nutrient acquisition by empiricists (Table 1), SRL, root phenology, root hair length and
384 density, root order distribution, root phosphatase and protease enzyme activities were not
385 represented at all or were not directly linked to nutrient acquisition in the models (Table 3). Based
386 on the empirical advances above, more models could consider implementing coordinated strategies
387 for N and P acquisition, together with plant hydraulics.

388

389 Coupled hydro-biogeochemical strategies: *avenues for model improvement*

390

391 Integrating hydraulic and nutrient model components is a forefront for model development. Model
392 development of root dynamics has proceeded on almost entirely independent paths for plant
393 hydraulics versus nutrient acquisition, even within the same model (via separate modules e.g.,
394 FATES, LM, LPJ, Table 2), such that coordinated responses to resource changes and C costs for
395 water versus nutrient acquisition are not represented. And, within these parallel model
396 developments, little attention has been given to the unique characteristics of tropical forests (e.g.,
397 moisture seasonality, drought, and P scarcity). Model advances have: 1) vertically resolved both
398 water and nutrient transport between layers (e.g., ELM-CNP; Yang et al. 2019), 2) represented the
399 C cost of coarse and fine-root allocation across depths (e.g., Sakschewski et al. 2021), and 3)
400 represented water and nutrient foraging functions of roots across depths (Christoffersen *et al.*,
401 2016; Xu *et al.*, 2016; Kennedy *et al.*, 2019; Langan *et al.*, 2017; Joshi *et al.*, 2022, Knox et al.
402 2023). Hydro-biogeochemical model integration would allow better representation of the fast-slow
403 plant lifestyle continuum (Reich 2014) by including trade-offs in nutrient-acquisition (shallow
404 rooted) vs. stable water supply (deep roots). Oliveira et al. (2021) argued that the fast-slow
405 continuum maps onto variation in soil fertility, and the risky-safe hydraulic safety tradeoff occurs
406 across moisture gradients. Hydro-biogeochemical integration would follow in the spirit of
407 allowing ecosystem function and community traits to emerge from competitive ecological
408 interactions (Scheiter *et al.* 2013; Fisher *et al.* 2015). This integration would also enable models
409 to better represent “trait filtering” of plant groups across multiple gradients, such as the sorting of
410 tropical tree species that is observed according to both moisture and P affinities across the Isthmus

411 of Panama (Condit *et al.*, 2013). We argue that the next step in this line of model development to
412 represent tropical forests is to integrate hydraulic and nutrient model components.

413

414 Coupled above-belowground resource strategies: *empirical advances*

415

416 While leaves and fine roots are somewhat analogous as aboveground/belowground resource
417 acquisition plant structures, there is variation in the degree to which analogous traits like specific
418 leaf area (SLA) vs. SRL, and leaf vs. root lifespans correlate across biomes (Withington *et al.*,
419 2006; Jiang *et al.*, 2021).

420

421 Very few studies have focused on above-belowground functional linkages in tropical forests, with
422 most attention to plant hydraulics. For example, maximum rooting depth of different species
423 (usually measured for coarse roots), has been linked to deciduousness in tropical forests,
424 particularly in regions with distinct dry seasons and mixed communities of deciduous, semi-
425 deciduous, and evergreen species (Sobrado & Cuenca, 1979; Sampaio, 1995; Smith-Martin *et al.*,
426 2020). In Amazonian forests designations have been identified for: 1) deep-rooted, evergreen
427 drought avoiders, 2) shallow-rooted, deciduous drought avoiders, and 3) shallow-rooted, evergreen
428 drought tolerators with embolism-resistant vascular systems (Brum *et al.*, 2019; Chitra-Tarak *et*
429 *al.*, 2021). Interestingly, hydraulic above-belowground linkages appear to be strongest under
430 stressful conditions. In the Amazon, only under dry conditions were there linkages among stem
431 embolism vulnerability and rooting depth (e.g., Oliveira *et al.*, 2019; Laughlin *et al.*, 2021), with
432 these linkages lacking in wet conditions. While these hydraulic groupings are helpful, there can
433 also be large variation in maximum rooting depth among coexisting species of similar lifeform

434 and deciduousness, as demonstrated here for a Costa Rican dry forest (Box 2). To explore this,
435 aboveground hydraulic traits could be linked to belowground traits beyond maximum rooting
436 depth, which is very difficult to measure, including overall root biomass depth distributions, vessel
437 diameter, root embolism vulnerability root embolism vulnerability, and seasonal changes in root
438 production (i.e., phenology) (Germon *et al.*, 2020). Data on the embolism resistance of roots is
439 particularly scarce (e.g., Domec *et al.*, 2006), and could be a focus area for future research to link
440 to aboveground hydraulic vulnerability.

441
442 For nutrient above-belowground coordination, a recent global review indicated greater
443 coordination of leaf with root N:P ratios in tropical forests relative to most other biomes, likely
444 related to widespread tropical soil P scarcity and conservation of P in plant tissues (Wang *et al.*,
445 2022). A broad-scale paper linking remotely-sense canopy traits in Panama with soil data found
446 that canopy greenness (a surrogate for NPP) corresponded to variations in soil fertility and toxicity
447 (Fisher *et al.*, 2020). Also, AMF vs. ECM association has been linked to canopy reflectance
448 properties in tropical forests in Hawai'i (as well as in many temperate sites), likely also indicating
449 plant nutrition linkages (Sousa *et al.*, 2021). While these root-canopy linkages are suggestive, we
450 lack more functional measurements of coordinated root and canopy nutrition in tropical forests.

451
452 Coupled above-belowground resource strategies: *model representation*

453
454 Of the 15 vegetation models, several linked aboveground deciduousness with root traits, and most
455 had uneven representation of belowground versus aboveground traits and functions, with an
456 average of ~30 aboveground traits compared to only about eight root traits represented per model

457 (Table 2). Of the models with explicit linkages, Ecosystem Demography model 2 (ED2) included
458 a trait-driven plant hydraulic module that represents drought deciduousness and plant water stress
459 (Medvigy *et al.*, 2009; Medvigy & Moorcroft, 2012). ED2 also uses three PFTs with different
460 rooting depths: a deeper-rooted evergreen PFT, a shallower-rooted deciduous PFT (Xu *et al.*, 2016;
461 Smith-Martin *et al.*, 2020), and a liana PFT with a different rooting depth from trees (Meunier *et al.*,
462 2021). The different rooting depths per PFT are linked to data on deciduous and evergreen
463 phenologies for tropical forests (Xu *et al.*, 2016; Smith-Martin *et al.*, 2020). Similarly, LPJmL4.0-
464 VR has adapted a traditional PFT-based model using deciduousness in the Amazon and defined a
465 spectrum of PFTs from shallow to deep-rooted, which are dependent on tree size, including
466 vertically resolved coarse roots (Sakschewski *et al.*, 2021). These groupings follow the empirical
467 data described above. Overall, above-belowground links in plant hydraulics are still in the early
468 stages of development, but these could form the foundation for more integrative plant function in
469 PFTs or trait clusters for tropical forests, with support from the empirical data.

470

471 For nutrient acquisition, above-belowground coupling in vegetation models is less developed, and
472 most commonly represented as photosynthate (i.e., C) expenditure for the acquisition of soil
473 nutrients based on plant N demand, including representation of physiological limits to nutrient
474 uptake and efficient optimization of C allocation (reviewed in Davies-Barnard *et al.*, 2022). For
475 example, in the representation of nutrient uptake in the Fixation & Uptake of Nutrients (FUN)
476 model, GPP drives nutrient uptake demand and supplies the C for expenditure (Fisher *et al.*, 2010,
477 Allen *et al.*, 2020). The Davies-Barnard *et al.* (2022) review illustrates that C allocation for nutrient
478 uptake represents a significant advance over older representations, such as biological nitrogen
479 fixation (BNF) as a function of evapotranspiration. Key to the C expenditure approach are the

480 concepts of nutrient limitation and photosynthetic downregulation, which occur when there is not
481 enough C to grow new leaves because of high C costs for soil exploration for scarce nutrients.

482

483 Coupled above-belowground resource strategies: *avenues for model improvement*

484

485 Integrating more of the hydraulic function of fine and coarse roots into existing above-
486 belowground hydraulic PFTs is an important next step for model development (Fig. 1). In
487 particular, above-belowground linkages for plant hydraulics could be expanded, including the
488 hydraulic function of coarse roots, which are rare in vegetation models and were in only four of
489 the models we assessed (Table 2). Coarse roots in the models were generally represented as
490 support, biomass storage, and root depth distribution, but they were not directly related to water
491 uptake or transport. Moreover, while models sometimes represent the C cost of fine roots, the C
492 cost of coarse roots is only implicitly embedded within an allocation to stem production. Here we
493 advocate that models explicitly represent the C cost of coarse roots that have a direct link to
494 function. This would allow modeled C assimilated aboveground and allocated to coarse roots
495 (investment cost) to be more directly linked to water uptake, following the approach of
496 Sakschewski *et al.* (2021). With the cost of both fine *and* coarse root production explicitly modeled
497 by soil depth, and the returns of such investment represented in terms of water uptake (see plant
498 hydraulic-enabled models, Table 2), models would be in a position to represent the three-way
499 tradeoff presented by Oliveira *et al.* (2021) among 1) embolism resistance (P50), 2) water table
500 access (deep roots), and 3) water loss control (deciduousness and stomata regulation). An early
501 advance has been made in this direction; the aDGVM2 model has shown how this three-way
502 tradeoff can emerge from variable rooting depth and tradeoffs with P50 and deciduousness

503 (Langan et al. 2017). Given the empirical support for this three-way tradeoff, and recent advances
504 in the modeling of variable rooting depths and plant hydraulics, we argue that this is a well-justified
505 avenue for data-model integration and development using the small but growing availability of
506 data. Focused collection of data on root hydraulics, such as root embolism resistance, would help
507 to clarify above-belowground coordination of this three-way trade-off.

508

509 For above-belowground nutrient coordination in models, there remain outstanding empirical
510 questions—and hypotheses that can be tested in models—of how C allocation and nutrient
511 acquisition interact. For example, what is an accurate tradeoff between C expenditure above versus
512 belowground under nutrient scarcity? To what extent can stoichiometric flexibility of different
513 plant tissues mediate or exacerbate nutrient limitations? How do these individual plant-level
514 processes manifest in larger model grid cells of multiple plants, cohorts, traits, or other plant
515 functional types? Investigations into these types of above-belowground nutrient acquisition
516 questions could then be combined with hydraulic above-belowground linkages to get more coupled
517 hydro-biogeochemical PFTs.

518

519 **Tropical root traits not clearly linked to function -- Not ripe for models**

520 It is important to note a set of root traits that are commonly measured and comprise a large portion
521 of our empirical tropical data (Fig. 2), but which thus far have not been demonstrated to link clearly
522 to root function (Table 1). These traits include: root tissue nutrient content and C:N:P
523 stoichiometry, aspects of root morphology (e.g., root tissue density), and mycorrhizal biomass or
524 colonization rates in the absence of functional characterization. Root nutrient content and
525 morphology have been used as proxies for resource acquisition and symbiotic strategies (Addo-

526 Danso *et al.*, 2018; Bergmann *et al.*, 2020); however, the functional roles of root nutrient content
527 and morphological traits like RDT for resource acquisition are not clear or consistent (Freschet *et*
528 *al.*, 2021b).

529

530 Recently, an expanded global database including root C:N:P and morphology was published as the
531 Global Root Trait (GRooT) database (Guerrero-Ramirez *et al.*, 2021, Fig. 2), which may be useful
532 for further exploration of functional linkages to stoichiometry. To increase the functional utility of
533 this database, these commonly measured traits are being compared and related to smaller data sets
534 for tropical nutrient uptake rates, phosphatase and protease activities. We present an example of
535 this type of exercise using new data, highlighting the difficulty of relating fine-root stoichiometry
536 to functional groupings like N-fixation or mycorrhizal association (Box 4), particularly in the
537 absence of direct measures of N-fixation, such as using $^{15}\text{N}_2$ labeling experiments (e.g.,
538 <https://www.youtube.com/watch?v=7jxM1KZ0f3Q>) or direct measures of mycorrhizal-plant C
539 exchange (e.g., <https://youtu.be/mNq8eQxDCqM>). Given the large availability of root nutrient
540 content data relative to other more functional traits (Fig. 2), it is worth pursuing these comparisons
541 to see if and when we can infer root functionality from stoichiometry, noting that root
542 stoichiometry in models plays an important role for determining nutrient storage and stocks of
543 biomass (Table 2).

544

545 Root morphology has been used as an indicator of nutrient acquisition strategies, but there have
546 been very few direct demonstrations of these relationships. Highlighting the difficulty of using
547 morphological root traits to infer function, fine-root traits (diameter, SRL, root tissue density,
548 branching) for 1,467 Amazonian tree species had no significant association with landscape-scale

549 shifts in bulk soil fertility (Vleminckx *et al.*, 2021). This is in contrast to some aboveground
550 tropical forest traits, like canopy greenness and nutrient content, which often covary with soil
551 fertility and soil texture (Fyllas *et al.*, 2012; Fortunel *et al.*, 2014; Fisher *et al.*, 2020). Root
552 diameter, which is functionally most closely related to water conductivity, has been used as a proxy
553 for AMF colonization rate, even though this relationship has not been consistently demonstrated
554 for tropical forests (Kong *et al.*, 2014; Lugli *et al.*, 2020; Yaffar *et al.*, 2021). We present new data
555 from Panama where some root morphological characteristics were strongly correlated to paired
556 measurements of nutrient uptake for two canopy tree species, and nutrient uptake rates were
557 different among nutrients for one different canopy species. However, these relationships were not
558 apparent in similar new data for two species from Singapore, possibly because the Panama data
559 were characterized according to root order (only root tips – 1st order – used, or roots separated for
560 the first three absorptive root orders for morphology, Box 3). Thus, further exploration of if, how,
561 and under what conditions morphological traits are related to nutrient (and water) uptake is
562 warranted, and there appear to be promising relationships if roots are assessed at a scale relevant
563 to absorptive activity.

564

565 For mycorrhizae, assessments of colonization, presence, or biomass are the most commonly used
566 methods (Sheldrake *et al.*, 2018; Olsson & Lekberg, 2022), but these measures do not necessarily
567 indicate functional activity since fungal biomass can be present but not active. These measures
568 could be improved if they were related to direct measurements of C or nutrient transfers between
569 tree and fungal symbionts, such as ¹³CO₂ pulse labeling of plants and subsequent transfer of ¹³C-
570 enriched C to symbionts (Lekberg *et al.*, 2013; Chaudhary *et al.*, 2022, Kaiser *et al.*, 2015; e.g.,

571 <https://youtu.be/mNq8eQxDCqM>), which would allow a better assessment of the functional value
572 of colonization data.

573

574 Clarifying the utility of these commonly measured fine-root traits for inferring functions in tropical
575 forests would be useful, given the relatively large quantity of fine-root nutrient, morphological,
576 and colonization data. Absent this, empirical research should shift toward root traits more clearly
577 linked to specific root functions, as described above.

578

579 **Achieving data-model integration for a better understanding of tropical root function**

580 We have identified opportunities for improving our understanding of fine-root function in tropical
581 forests, and for integrating key root functions into vegetation models as applied to tropical
582 ecosystems. Our surveys of empirical and modeling approaches to utilizing root data (Table 1-3)
583 demonstrate several broad trends: 1) There are some root characteristics for water acquisition (e.g.,
584 root biomass and maximum rooting depth) that are being implemented in models according to our
585 empirical understanding. 2) There have been numerous recent advances in characterization of root
586 traits and functions in tropical forests, but many of these are missing in vegetation models. 3)
587 Models represent some characteristics that are not easily measured and for which there are few
588 data (e.g., nutrient uptake kinetics, water transport by coarse roots). 4) Functional characterization
589 of fine roots is often different in models versus our empirical understanding (Table 3). For
590 example, SRL is used in some models as a PFT characteristic which is unresponsive to resource
591 changes, yet recent data indicate that only ~50% of variation in SRL might be explained by species
592 differences (Box 1), and SRL can be very responsive to resource changes in tropical forests (see
593 discussion above). 5) There are some root characteristics that are well linked to functions in limited

594 empirical studies, such as phosphatase activity, but which have not yet been sufficiently
595 characterized in tropical forests to implement response functions in vegetation models. 6) Some of
596 the most-measured root traits have not been clearly linked to function, and therefore are not
597 immediately useful for representing resource acquisition processes in models (e.g., root nutrient
598 content and diameter). Overall, there is much work left to be done to bring together empirical and
599 modeling research on tropical forest belowground functions, with a need for greater integration
600 going forward.

601
602 There are existing frameworks for advancing model-data integration and for comparing models
603 with different modalities (Walker *et al.*, 2014; Medlyn *et al.*, 2015; Kyker-Snowman *et al.*, 2022),
604 but the computational cost of increasing model complexity must be justified by improved model
605 performance. More model ensemble experiments for tropical forest biomes would be useful to test
606 the level of improvement achieved by representing expanded root function (e.g., Koven *et al.*,
607 2020; Caldararu *et al.*, 2023, Fleischer *et al.*, 2019). New experiments could also test model-
608 derived hypotheses prior to inclusion of a new process in models. For example, the AmazonFACE
609 experiment (<https://amazonface.unicamp.br/>) will test hypotheses about P dynamics under
610 elevated CO₂ that were developed by using a model inter-comparison (Fleischer *et al.*, 2019).
611 Some key questions that arose from these modeling activities are: will CO₂ enrichment stimulate
612 root phosphatase activity sufficiently to alleviate P limitation to growth (Yang *et al.*, 2019)? And,
613 will including phosphatase production in models improve predictions of tropical forest
614 productivity and responses to elevated CO₂?

615

616 Close interactions between empiricists and modelers over the course of research projects are
617 essential to meet the challenges we have identified in this research agenda. Model-data integration
618 for tropical forests has improved in the past decade, including efforts such as the U.S. Department
619 of Energy Next Generation Ecological Experiments–Tropics (NGEE-Tropics, [https://ngee-
620 tropics.lbl.gov/](https://ngee-tropics.lbl.gov/)), the AmazonFACE, the TropiRoot network
621 (<https://tropiroottrait.github.io/TropiRootTrait/>, described in <https://youtu.be/oT2lgeGDnjI>), and
622 the Landscape Evolution Observatory at Biosphere 2
623 (<https://www.science.org/doi/full/10.1126/science.abj6789>), which bring together field research
624 questions and modeling objectives. Nonetheless, support for these endeavors remains limited. We
625 urge that these collaborations be widespread and supported by funding agencies in order to
626 improve our understanding and prediction of tropical forest function and feedbacks to a changing
627 world.

628

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653

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661

662

663 **Competing interests**

664 None declared

665

666 **Data availability**

667 Data presented in Boxes 1-4 are available in the SI.

668

669

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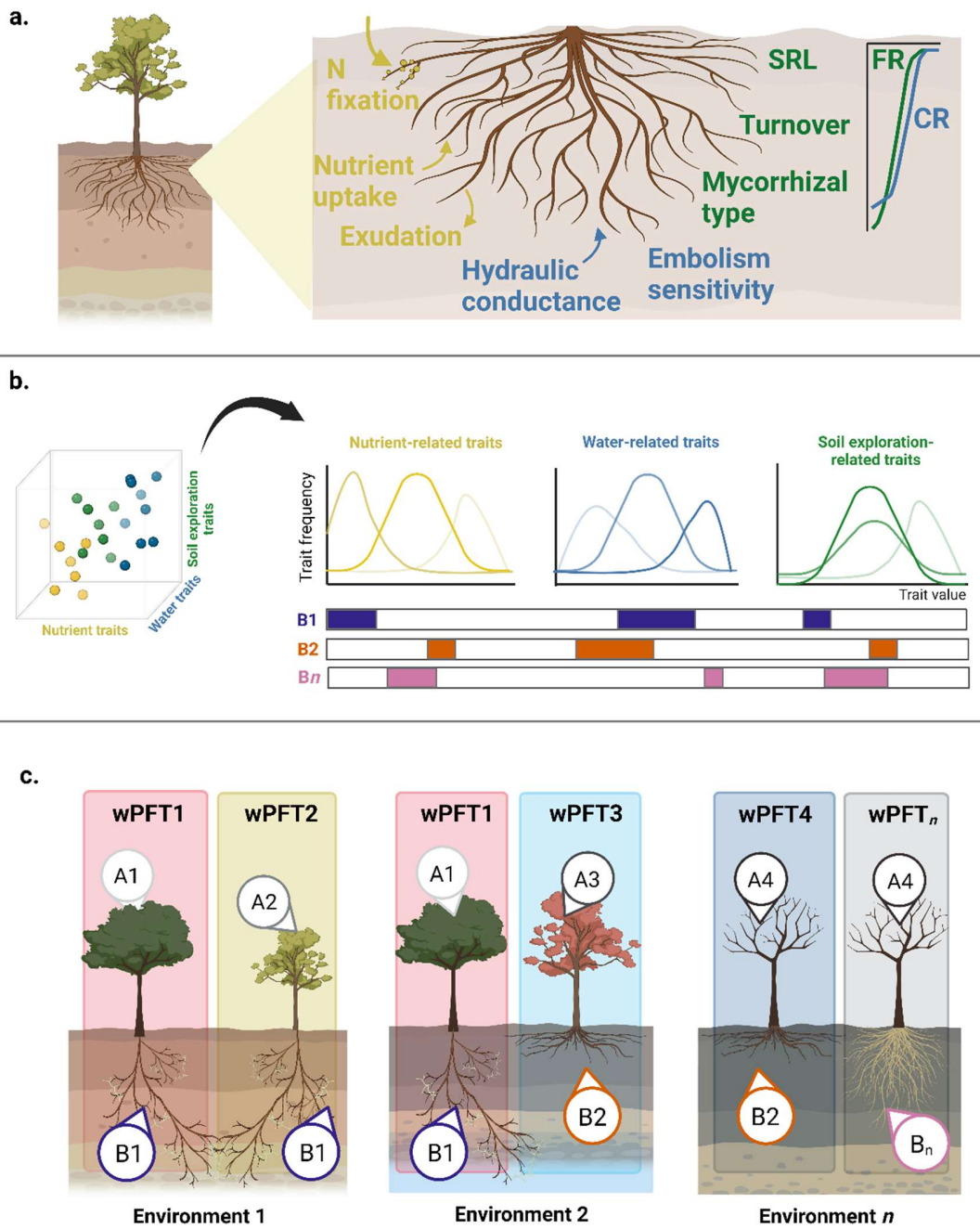
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697 **Figures**

698 **Fig. 1**



699

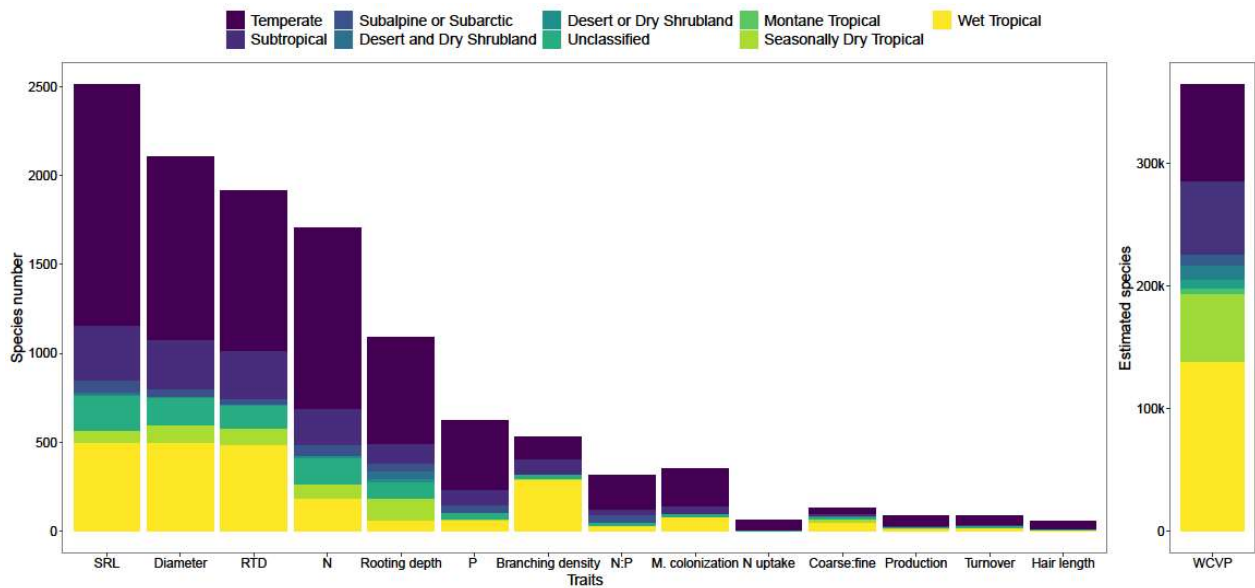
700 **Figure 1.** A conceptual representation depicts the root traits recommended for further tropical
701 forest research and representation in vegetation models as part of PFTs or trait-clusters. The panels
702 include: **(a.)** a graphical depiction of the root system with a subset of suggested priority root traits
703 for the tropics (see also Table 1), **(b.)** multidimensional trait space and trait distributions that could
704 be used to inform more balanced above-belowground whole-plant functional types (here signaled
705 as ω PFTs) for the tropics, and **(c.)** representation of different combinations of belowground trait
706 clusters mixed and matched with aboveground PFTs to test in vegetation models and guide
707 empirical research. Details are as follows: **(a.)** a graphical depiction of root system traits including
708 nutrient uptake traits (in yellow, N-fixation, nutrient uptake rates, carbon exudation), water uptake
709 and drought resistance traits (in blue, hydraulic conductance and embolism sensitivity), and
710 general soil exploration traits (in green, e.g., mycorrhizal type, specific root length (SRL), root
711 turnover). Also shown are hypothetical depth distributions for coarse roots (CR, blue), and fine
712 roots (FR, green, inset). **(b.)** A multidimensional trait space is linked to hypothetical distributions
713 for the root traits depicted in panel (a.), indicating how ranges of the different trait distributions
714 could be selected to form multi-trait belowground functional types (B1, B2, etc.). **(c.)**
715 Belowground and aboveground groupings could be matched to create ω PFTs, and then tested in
716 different combinations in vegetation models to assess improvement in predictions of NPP and
717 other emergent properties of ecosystems depicted in vegetation models.

718

719

720

721 **Figure 2.**



722

723 **Fig. 2.** The total number of species for which fine-root traits data are currently available across
 724 climatic biomes is shown (left) relative to the total number of known species in each biome (right).

725 Despite much higher species numbers present in wet tropical forests, the highest percentage of
 726 available data comes from temperate plant species. Within the tropics, most data are from wet
 727 tropical forests (shown in yellow, “wet tropical”). Overall, SRL has been the most commonly
 728 measured root trait, while traits particularly important in tropical forests like P uptake are virtually
 729 uncharacterized. **Left panel:** the species number in the updated version of the GRooT database
 730 (Guerrero-Ramirez *et al.*, 2021) are shown by trait and biome description (colors); traits included
 731 are specific root length (SRL, m g⁻¹), mean root diameter (Diameter, mm), root tissue density
 732 (RTD, g cm⁻³), root nitrogen concentration (N, mg g⁻¹), maximum rooting depth (Rooting depth,
 733 m), root phosphorus concentration (P, mg g⁻¹), root branching density (Branching density, number
 734 cm⁻¹), root nitrogen to phosphorus ratio (N:P), root mycorrhizal colonization intensity (M.
 735 colonization, %), the net uptake rate of nitrogen (N uptake, μmol g⁻¹ day⁻¹), coarse-to-fine root
 736 mass ratio (Coarse: fine), root hair length (Hair length, μm), root production (Production, g m⁻²

737 year⁻¹), root turnover rate (Turnover, year⁻¹). Data were filtered to include only fine roots for most
738 of the traits, except coarse-to-fine root mass ratio, maximum rooting depth, and root hair length.
739 **Right panel:** estimate total species number by climate biome from the World Checklist of
740 Vascular Plants (WCVP, Govaerts *et al.*, 2021, POWO 2023). Data sources, climate zone
741 descriptions, and processing details are in SI.

742 **Table 1.** Root characteristics and trait functions as understood empirically and represented in models

Root Trait	Units	Function in Nature	Function in Models
<i>Fine Root Function: Soil Exploration for Water & Nutrient Acquisition</i>			
Fine-root biomass	Mg ha ⁻¹	Absorptive tissue	Absorptive tissue
Fine-root productivity	Mg ha ⁻¹ y ⁻¹	Absorptive tissue productivity	Absorptive tissue productivity
Fine-root turnover (inverse of lifespan)	y ⁻¹	Absorptive tissue turnover	Absorptive tissue turnover
Specific root length (SRL)	cm g ⁻¹	Soil volume explored per cost	Conversion factor (fine root biomass to fine root length), Calculate absorptive area as biomass×SRL×2πr
Root growth timing (e.g., phenology/seasonality)	growth or death timing	Align root production & mortality with resource availability	ABSENT
Root hair length	μm	Absorptive tissue	ABSENT
Root hair density	hairs cm ⁻¹	Absorptive tissue	ABSENT

Depth distribution	Distribution parameter(e.g., β)	Distribute absorptive and transportive tissues	Locate absorptive tissue, characteristic of PFTs in some models
Root order distribution (i.e., branching density)	(1+2+3 order) : (4 order)	Absorption per transport	ABSENT (except where vertical distribution of coarse and fine roots are treated separately)
Mycorrhizae	Colonization rate, hyphal length, material transfer rate	Exchange C for water, Ps or other nutrients	Exchange C for N and phosphorus
<i>Fine Root Function: Water Acquisition & Drought Resistance</i>			
Maximum depth	m	Define vertical root domain	Define vertical root domain, characteristic of PFTs in some models

Root hydraulic conductivity	L_p ; $m \text{ sec}^{-1} \text{ MPa}^{-1}$	Water transport	Water transport
P50; pressure at 50% embolism	MPa	Embolism resistance	Embolism resistance
Root radius (or diameter)	mm	Possibly related to water c or AMF colonization, function poorly constrained	Soil-root water conductance
Root membrane permeability	Mass pressure ⁻¹ area ⁻¹ time ⁻¹	Water uptake	Water uptake
Water uptake rate	mg-H ₂ O length ⁻¹ time ⁻¹	Water uptake	Water uptake
<i>Fine-Root Function: Nutrient Acquisition</i>			
Root enzyme activities (e.g., phosphatase, protease)	Degradation rate of	Release organic phosphorus	Release organic phosphorus

	synthetic substrate		
Organic exudate production	C root mass ⁻¹ (or length ⁻¹) time ⁻¹	Release mineral phosphorus	Release mineral phosphorus, present in few models where it responds to nutrient availability
N fixation (nodule biomass and nitrogen fixation rate)	Nodule biomass area ⁻¹ , and fixation rate - mg N ₂ nodule biomass ⁻¹ time ⁻¹	Acquires N from atmosphere and converts to biologically available forms	Exchange C for N, modeled as C cost, maintenance respiration, or nodule turnover time in response to nutrient availability. Or, modeled as a function of evapotranspiration or NPP. Present in few models, often as a characteristic of PFTs
Phosphorus uptake rate	μg P length ⁻¹ (or root mass ⁻¹) time ⁻¹	Realized phosphorus uptake by root or AMF/ECM symbiont	Realized P uptake, present in few models and varies with nutrient availability

Nitrogen uptake rate	$\mu\text{gN length}^{-1}$ (or root mass^{-1}) time^{-1}	Realized nitrogen uptake by root or AMF/ECM symbiont	Realized nitrogen uptake, present in few models and varies with nutrient availability
<i>Traits without a clear relationship to root resource acquisition</i>			
Tissue N concentration	%	Unclear if correlated with function	
Tissue P concentration	%	Unclear if correlated with function	
Tissue N-to-P ratio	Ratio	Stoichiometry	ABSENT
Tissue C-to-N ratio	Ratio	Stoichiometry	Control N demand, present in most models, part of PFTs
Tissue C-to-P ratio	Ratio	Stoichiometry	Control P demand, present in most models, part of PFTs
Root tissue density	g cm^{-1}	Defense, possible relation to AMF colonization rate (volume available for colonization)	ABSENT

<i>Coarse Root Function: Support & Transport</i>			
Coarse root biomass	Mg ha ⁻¹	Support & Water Transport	Track elements in tissues, present in some models as part of PFTs
Coarse root productivity	Mg ha ⁻¹ yr ⁻¹	Support	Track elements in tissues
Coarse root hydraulic resistance	MPa s ⁻¹ kg ⁻¹ H ₂ O	Water transport	Water transport, present in some models

743

744 Table 1. Root characteristics and traits indicated in this Viewpoint as most relevant to tropical forest function are given, grouped by
745 main function, with common units, and specific function as understood empirically (Function in Nature). Functions in vegetation models
746 are then given, followed by categorical description of how these are included in models (details in Table 2).

747

748 **Table 2.** Inclusion of root traits in combination with aboveground traits and parameters in a suite of vegetation models.

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750

751 Table 2. (Table available as Excel file). Root traits as represented in a sample of 15 vegetation models varying in scope (capability of
752 coupling to Earth system models) and sub-grid resolution (trait assignment and spatial resolution). In general, these models predict plant
753 productivity and other emergent properties of ecosystems depending on changing conditions over time. “Spatially explicit” refers to
754 aboveground processes only (e.g., ray tracing and light competition). Root traits (columns) are grouped by main functions (colors). As
755 an indicator of model complexity in above vs. belowground plant processes, a tally of unique aboveground vs belowground root traits
756 for each model is given. Cells in the table denote where a given root trait varies by PFT or individual (p), size (s), is a global constant
757 (c), is dynamic with moisture (m), is dynamic with nutrients (n), or if it is implicitly considered through other means (i). Blank cells
758 denote traits not represented for a given model. Parentheses with numbers indicate multiple sub-traits or parameters associated with a
759 particular trait. Representative citations for each model are given. Hydraulic-enabled models explicitly represent water transport within
760 plants and include models #4, #5, #6, #7, and #12. Nutrient-enabled models explicitly model N and/or P uptake through direct or
761 symbiotic means and include #1, #5, #7, #9, and #10. Root traits associated with C or biomass storage are not considered here. For a
762 complete description of methods used to construct this table, including definitions of terms and acronyms, and the file or table within
763 each citation that was the specific source for trait information, see the SI.

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768 Table 3. A tabulation is presented of the root characteristics present in Table 1 (“1. Empirical
769 research focus”), present in Table 2 (“2. Model focus”), and 3. Common to both models and
770 empirical research, with Caveats when the trait is understood or used differently in empirical work
771 compared with model applications. Note that information is organized in columns such that
772 columns 1, 2 and 3 do not correspond horizontally.

1. Empirical research focus (lacking model representation)	2. Model focus (lacking empirical focus or understanding)	3. Common to models and empirical research	Caveats for commonalities between models and empirical research
Root phenology (seasonality of production/mortality)	Root:Leaf biomass ratio	Root biomass	Models emphasize coarse root biomass as a stock, empirical research emphasizes fine root biomass for resource acquisition
Root hair abundance and length	Water stress factor	Max rooting depth and root depth distributions	Models emphasize for water uptake, empirical research combines with root depth distributions for nutrient uptake
Root order distributions	Fraction of tree hydraulic resistance in roots	Root tissue CN(P)	Stoichiometry is not clearly functionally important in empirical studies. In models this is commonly used for nutrient accounting and to drive nutrient demand, so in neither case is this a functional trait
Root enzyme production (e.g. phosphatase)	Root membrane permeability	Root production and turnover rates	Root turnover rates poorly characterized in tropical empirical data
Root tissue density	C cost of N fixation and C cost of mycorrhizal nutrient acquisition	Fine-root SRL (specific root length)	Used as a PFT trait in some models or as a global constant, in empirical work this is responsive more to resource availability and is not clearly distinct among species

	<p>N fix nodule turnover rates</p>	<p>Water and nutrient uptake rates</p>	<p>Poorly characterized for the tropics</p>
	<p>Maintenance respiration C cost of nodules</p>	<p>Root diameter</p>	<p>Related to uptake and transport in both models and empirical research</p>
	<p>Nutrient uptake rates of AMF vs ECM associations</p>	<p>Root conductance rates and embolism vulnerability (P50)</p>	<p>Very poorly characterized in roots overall, especially in the tropics</p>
		<p>Root organic exudate production</p>	<p>Poorly characterized in tropical empirical data, linked to nutrient uptake</p>
		<p>N fixation rates and nodule biomass</p>	<p>Good empirical understanding of fixation and its function relative to other root traits and represented in most models with improvement needed</p>
		<p>Mycorrhizal type</p>	<p>Type (AMF vs. ECM) related to N uptake rates in models, not supported by tropical data, likely more related to P uptake in tropics but this not in models</p>

773

774 **Box 1. High fine-root trait variation within and among individuals suggests morphological**
775 **trait flexibility within species in tropical forests**

776 New data from two tropical forests indicate large intra-specific and individual-scale variation in
777 morphological traits for absorptive roots (orders 1 – 4), comparing the proportion of variation
778 explained by species, individual trees, individual root segments (multiple per individual), and
779 residual (unexplained) variance. The Panama data include 10 replicate individuals for each of two
780 species, and Puerto Rico data include two to three replicate individuals for each of six species to
781 assess inter- and intra-specific variation, with details given in Notes S1. Overall, root segment
782 within individual contributed a large portion of the variance when there was replication at the
783 individual scale (Panama data). When individuals were not well replicated but more species were
784 measured, individual and species contributed similarly to variance for fine-root morphology
785 (Puerto Rico data). Data are provided as Datasets S2 and S3. These data support recent publications
786 indicating that root traits are less phylogenetically conserved in tropical forests (see main text) and
787 suggest that tropical forest community-scale root characteristics are likely dynamic in response to
788 resource shifts. More work must be done to directly link these commonly-measured fine-root
789 morphological traits to functional root activities like nutrient and water uptake and transfer.

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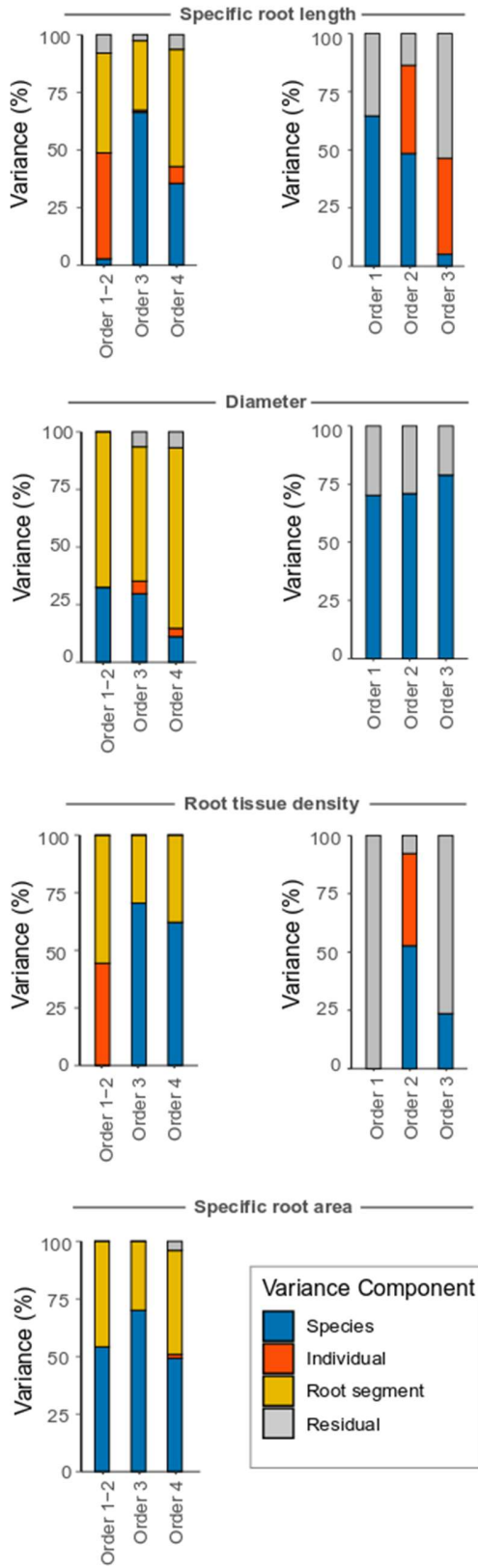
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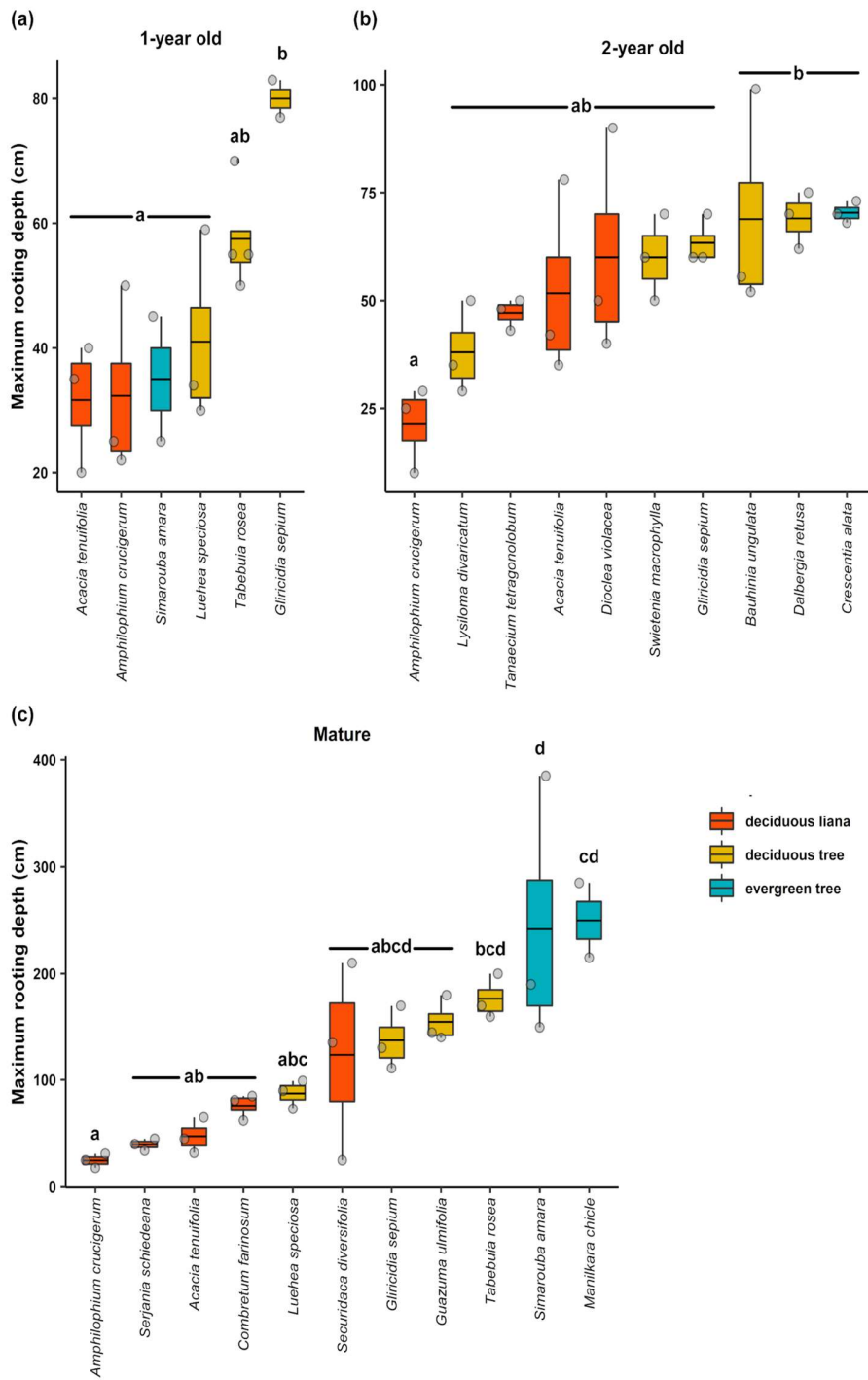
A) Panama

B) Puerto Rico



795 **Box 2: Tropical forest maximum rooting depth linked to life form and deciduousness, but**
796 **much variation remains**

797 Deep roots are particularly important for water uptake and redistribution to support transpiration
798 demands during dry periods (Markesteijn & Poorter, 2009), and aboveground phenology has been
799 linked to rooting depths in dry tropical forests (Smith-Martin *et al.*, 2020). Here, a new analysis of
800 data from a dry tropical forest in Costa Rica show relationships between aboveground life form
801 and rooting depth for juvenile and mature trees. This analysis shows that mature evergreen trees
802 had c. 2x the maximum rooting depth of co-occurring mature deciduous lianas and trees, indicating
803 above-belowground trait coordination. Letters show means separations using Tukey HSD tests,
804 boxes show means and quartiles. Details are in Notes S1 and data are provided as Dataset S4.
805 These patterns were not present in juvenile trees (top panels), suggesting that belowground niche
806 partitioning develops over time. At the same time, there was substantial variation in maximum
807 rooting depth among mature species that were classified as the same functional type using
808 aboveground deciduousness, suggesting that a more refined understanding of belowground
809 hydraulic strategies within these groups could help separate species into more functionally explicit
810 groupings. Such a holistic below-aboveground representation of water acquisition strategies could
811 contribute to improved tropical forest PFTs or trait clusters, which could then be combined with
812 nutrient acquisition types to improve tropical plant representation in vegetation models.



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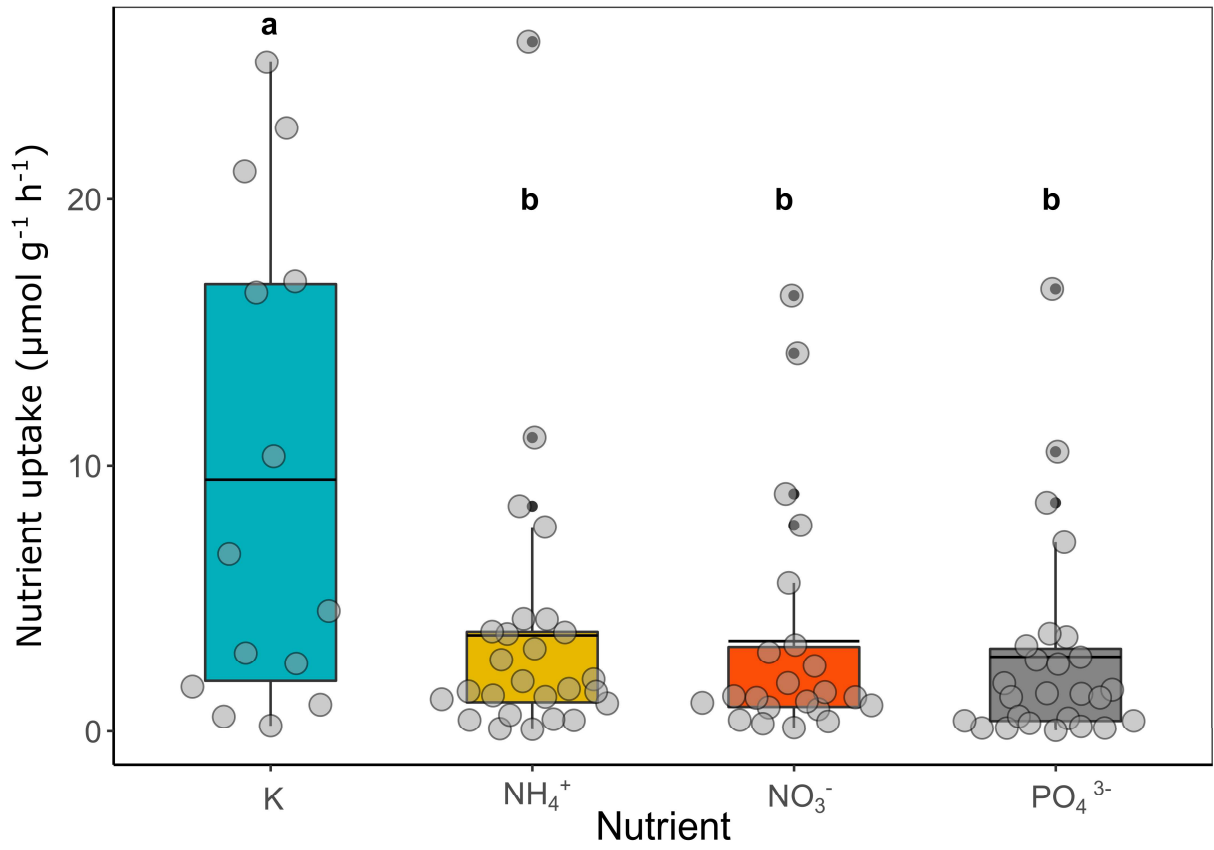
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816 **Box 3: Fine-root nutrient uptake rates and relationships to morphology for tropical trees**

817 Direct measures of nutrient uptake rates by tropical trees are rare and rarely linked to broader
818 nutrient limitation to NPP studies or to fine root morphological root traits. Here, we present new
819 data for fine-root nutrient uptake rates in a well-characterized lowland Panamanian forest and show
820 relationship between uptake rates and fine-root morphology, with similar data available for two
821 tree species in Singapore. Details are in Notes S1 and data are provided as Datasets S5 and S6.
822 Across 33 mature individuals of a relatively abundant Panamanian lowland species *Protium*
823 *picramnioides*, there was significantly greater nutrient uptake rates for potassium (K) versus
824 ammonium (NH₄⁺), nitrate (NO₃⁻), and phosphate (PO₄³⁻). Figure means are shown with quantiles
825 (F_{3,87}=6.78; P=0.022), letters indicate significant differences using Tukey HSD tests. This result
826 supports data from a long-term nutrient fertilization experiment in the same site showing that K
827 addition reduced fine-root biomass, length, RDT, and increased SRL (Wurzburger & Wright,
828 2015), suggesting K limitation to root processes and fine-root dynamic responsiveness to changes
829 in K availability. Data for two other Panamanian species and two species in Singapore also showed
830 variation in uptake rates among nutrients (Notes S1). The Panamanian species had strong
831 correlations between nutrient uptake rates and root morphology, including positive correlations of
832 NO₃⁻ and PO₄³⁻ with SRL ($r^2 = 0.83$ and 0.88 , respectively), negative correlations of uptake with
833 RTD ($r^2 = 0.99$ and 0.71 , respectively), and a negative correlation of PO₄³⁻ with root biomass (r^2
834 = 0.75 , see Notes S1). This result supports the idea in Box 1 that tropical forest fine-root
835 morphology is responsive to changes in nutrient availability, and that morphology is related to
836 nutrient uptake. Methodological details and raw data are in Notes S1 and shown at
837 <https://youtu.be/4atZ3E0NrX4>. Dynamic nutrient uptake rate measurements within and among
838 sites could be related to fine-root morphological characterization, which could help inform

839 dynamic root responses to changing resources in vegetation models, but considerably more data
840 are required.



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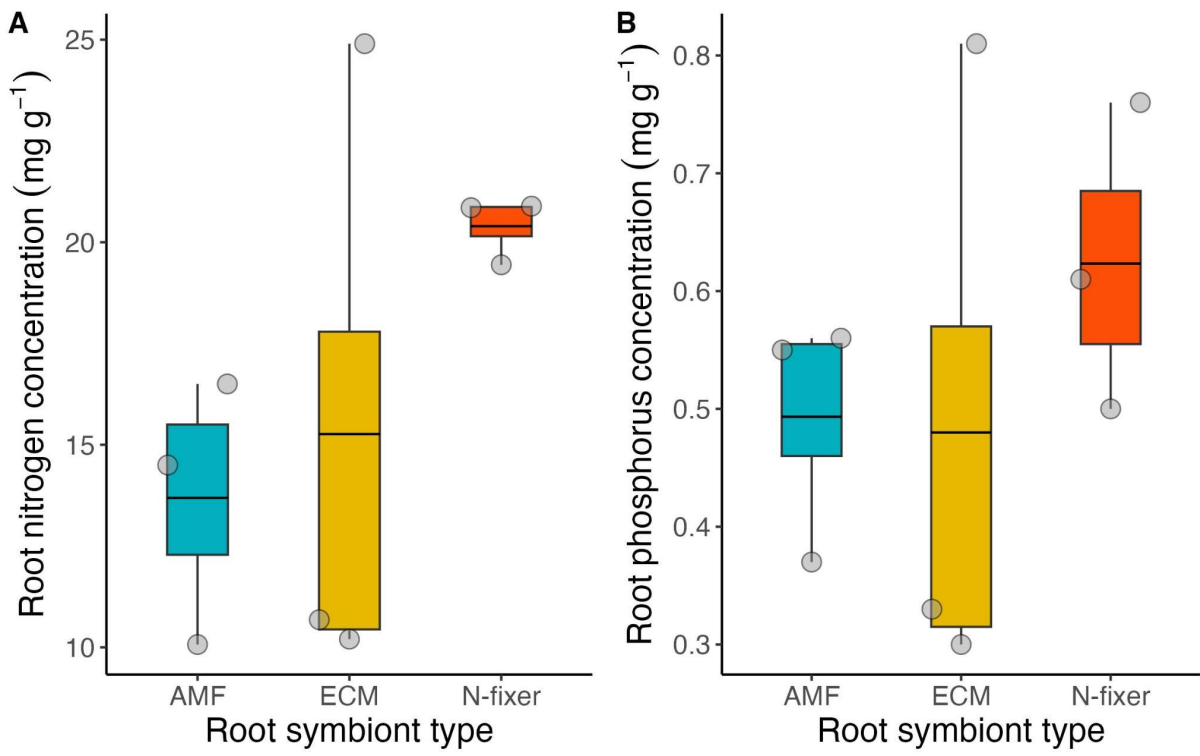
845 **Box 4: Using fine-root stoichiometry as a surrogate for functional traits**

846 Plant tissue stoichiometry could be a relatively easy way to start constraining nutrient acquisition
847 in functional groupings since fine-root C:N:P content is relatively more available than functional
848 trait measurements. However, there remains a knowledge gap across tropical sites linking root
849 stoichiometry directly to nutrient or water acquisition. Fine-root P and N concentrations could
850 reflect fine-root P and N acquisition rates, either directly or via symbiosis. For example, fine-root
851 P is strongly correlated to leaf P concentration (Holdaway *et al.*, 2011), soil inorganic and total P
852 (Holdaway *et al.*, 2011; Schreeg *et al.*, 2014; Freschet *et al.*, 2021b), and soil extractable P (Yaffar
853 *et al.*, 2021). Here we present new data on fine-root N and P content for Panamanian trees with
854 three root symbiont types (arbuscular mycorrhizal (AM), ectomycorrhizal (EM), and N-fixing
855 (Nfix) to explore functional relationships. While N fixers tended to have higher root N content,
856 there were no significant differences in root stoichiometry across these three functional types.
857 Figure shows means and quantiles shown for nine tree species (n = 3 individuals per functional
858 type). Details are in Notes S1 and data for each tree species are provided as Dataset S7.. These
859 data highlight the problems with using root stoichiometry to assess symbiotic activity or nutrient
860 uptake rates without additional measurements. Further investigation to confirm whether root
861 stoichiometry is indicative of tropical plant fine-root functional activity would be useful, since root
862 stoichiometry is one of the most abundant types of tropical root data (Fig. 2).

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870 **References**

871

872 **Addo-Danso SD, Defrenne CE, McCormack ML, Ostonen I, Addo-Danso A, Foli EG, Borden**
873 **KA, Isaac ME, Prescott CE. 2020.** Fine-root morphological trait variation in tropical
874 forest ecosystems: an evidence synthesis. *Plant Ecology* **221**: 1-13.

875 **Addo-Danso SD, Prescott CE, Adu-Bredu S, Duah-Gyamfi A, Moore S, Guy RD, Forrester**
876 **DI, Owusu-Afriyie K, Marshall PL, Malhi Y. 2018.** Fine-root exploitation strategies
877 differ in tropical old growth and logged-over forests in Ghana. *Biotropica* **50**: 606-615.

878 **Allen K, Fisher JB, Phillips RP, Powers JS, Brzostek ER. 2020.** Modeling the carbon cost of
879 plant nitrogen and phosphorus uptake across temperate and tropical forests. *Frontiers in*
880 *Forests and Global Change* **3**.

881 **Andersen KM, Mayor JR, Turner BL. 2017.** Plasticity in nitrogen uptake among plant species
882 with contrasting nutrient acquisition strategies in a tropical forest. *Ecology* **98**: 1388-1398.

883 **Andersen KM, Turner BL. 2013.** Preferences or plasticity in nitrogen acquisition by understory
884 palms in a tropical montane forest. *Journal of Ecology* **101**: 819-825.

885 **Aragao L, Malhi Y, Metcalfe DB, Silva-Espejo JE, Jimenez E, Navarrete D, Almeida S, Costa**
886 **ACL, Salinas N, Phillips OL, et al. 2009.** Above- and belowground net primary
887 productivity across ten Amazonian forests on contrasting soils. *Biogeosciences* **6**: 2759-
888 2778.

889 **Asefa M, Worthy SJ, Cao M, Song XY, Lozano YM, Yang J. 2022.** Above- and belowground
890 plant traits are not consistent in response to drought and competition treatments. *Annals of*
891 *Botany* **130**: 939-950.

892 **Averill C, Bhatnagar JM, Dietze MC, Pearse WD, Kivlin SN. 2019.** Global imprint of
893 mycorrhizal fungi on whole-plant nutrient economics. *Proceedings of the National*
894 *Academy of Sciences of the United States of America* 116: 23163-23168.

895 **Barron, AR, Purves, DW, & Hedin, LO. 2011.** Facultative nitrogen fixation by canopy
896 legumes in a lowland tropical forest. *Oecologia*, **165**: 511-520.

897 **Batterman SA, Hall JS, Turner BL, Hedin LO, Walter JKL, Sheldon P, van Breugel M. 2018.**
898 Phosphatase activity and nitrogen fixation reflect species differences, not nutrient trading
899 or nutrient balance, across tropical rainforest trees. *Ecology Letters* **21**: 1486-1495.

900 **Belda DM, Anthoni P, Warlind D, Olin S, Schurgers G, Tang J, Smith B, Arneth A. 2022.**
901 LPJ-GUESS/LSMv1.0: a next-generation land surface model with high ecological realism.
902 *Geoscientific Model Development* **15**: 6709-6745.

903 **Bergmann J, Weigelt A, van Der Plas F, Laughlin DC, Kuyper TW, Guerrero-Ramirez NR,**
904 **Valverde-Barrantes OJ, Bruelheide H, Freschet GT, Iversen CM, et al. 2020.** The
905 fungal collaboration gradient dominates the root economics space in plants. *Science*
906 *Advances* **6**.

907 **Braghiere RK, Fisher JB, Allen K, Brzostek E, Shi M, Yang X, Ricciuto DM, Fisher RA, Zhu**
908 **Q, Phillips RP. 2022.** Modeling global carbon costs of plant nitrogen and phosphorus
909 acquisition. *Journal of Advances in Modeling Earth Systems* **14**.

910 **Brum M, Vadeboncoeur MA, Ivanov V, Asbjornsen H, Saleska S, Alves LF, Penha D, Dias**
911 **JD, Aragao L, Barros F, et al. 2019.** Hydrological niche segregation defines forest
912 structure and drought tolerance strategies in a seasonal Amazon forest. *Journal of Ecology*
913 **107**: 318-333.

914 **Brzostek ER, Fisher JB, Phillips RP. 2014.** Modeling the carbon cost of plant nitrogen
915 acquisition: Mycorrhizal trade-offs and multipath resistance uptake improve predictions of
916 retranslocation. *Journal of Geophysical Research-Biogeosciences* **119**: 1684-1697.

917 **Cabugao KG, Yaffar D, Stenson N, Childs J, Phillips J, Mayes MA, Yang XJ, Weston DJ,**
918 **Norby RJ. 2021.** Bringing function to structure: Root-soil interactions shaping
919 phosphatase activity throughout a soil profile in Puerto Rico. *Ecology and Evolution* **11**:
920 1150-1164.

921 **Caldararu S, Rolo V, Stocker BD, Gimeno TE, Nair R. 2023.** Ideas and perspectives: Beyond
922 model evaluation – combining experiments and models to advance terrestrial ecosystem
923 science. *Biogeosciences Discuss.* [preprint], <https://doi.org/10.5194/bg-2023-47>.

924 **Carmona CP, Bueno CG, Toussaint A, Trager S, Diaz S, Moora M, Munson AD, Partel M,**
925 **Zobel M, Tamme R. 2021.** Fine-root traits in the global spectrum of plant form and
926 function. *Nature* **597**: 683-+.

927 **Chaudhary VB, Holland EP, Charman-Anderson S, Guzman A, Bell-Dereske L, Cheeke TE,**
928 **Corrales A, Duchicela J, Egan C, Gupta MM, et al. 2022.** What are mycorrhizal traits?
929 *Trends in Ecology & Evolution* **37**: 573-581.

930 **Chave J. 1999.** Study of structural, successional and spatial patterns in tropical rain forests using
931 TROLL, a spatially explicit forest model. *Ecological Modelling* **124**: 233-254.

932 **Chitra-Tarak R, Xu CG, Aguilar S, Anderson-Teixeira KJ, Chambers J, Detto M,**
933 **Faybishenko B, Fisher RA, Knox RG, Koven CD, et al. 2021.** Hydraulically-vulnerable
934 trees survive on deep-water access during droughts in a tropical forest. *New Phytologist*
935 **231**: 1798-1813.

936 **Christoffersen BO, Gloor M, Fauset S, Fyllas NM, Galbraith DR, Baker TR, Kruijt B,**
937 **Rowland L, Fisher RA, Binks OJ, et al. 2016.** Linking hydraulic traits to tropical forest
938 function in a size-structured and trait-driven model (TFS v.1-Hydro). *Geoscientific Model*
939 *Development* **9**: 4227-4255.

940 **Chuyong CB, Newbery DM, Songwe NC. 2000.** Litter nutrients and retranslocation in a
941 **central African rain forest dominated by ectomycorrhizal trees.** *New Phytologist* **148**:
942 **493-510.**

943 **Clark MP, Fan Y, Lawrence DM, Adam JC, Bolster D, Gochis DJ, Hooper RP, Kumar M,**
944 **Leung LR, Mackay DS, et al. 2015.** Improving the representation of hydrologic processes
945 in Earth System Models. *Water Resources Research* **51**: 5929-5956.

946 **Condit R, Pitman N, Leigh EG, Chave J, Terborgh J, Foster RB, Nunez P, Aguilar S,**
947 **Valencia R, Villa G, et al. 2002.** Beta-diversity in tropical forest trees. *Science* **295**: 666-
948 669.

949 **Condit R, Engelbrecht, BMJ, Pino, D, Perez, R, Turner BL 2013.** Species distributions in
950 response to individual soil nutrients and seasonal drought across a community of tropical
951 trees. *Proceedings of the National Academy of Sciences of the United States of America*,
952 **110**(13), 5064-5068.

953 **Cornelissen JHC, Aerts R, Cerabolini B, Werger MJA, van der Heijden MGA. 2001.** Carbon
954 cycling traits of plant species are linked with mycorrhizal strategy. *Oecologia* **129**: 611-
955 619.

956 **Costa FRC, Schietti J, Stark SC, Smith MN. 2023.** The other side of tropical forest drought: do
957 shallow water table regions of Amazonia act as large-scale hydrological refugia from
958 drought? Palavras-ChavePalabras clave. *New Phytologist* **237**: 714-733.

959 **Cunha HFV, Andersen KM, Lugli LF, Santana FD, Aleixo IF, Moraes AM, Garcia S, Di**
960 **Ponzio R, Mendoza EO, Brum B, et al. 2022.** Direct evidence for phosphorus limitation
961 on Amazon forest productivity. *Nature* **608**: 558-+.

962 **Cusack DF, Addo-Danso SD, Agee EA, Andersen KM, Arnaud M, Batterman SA, Brearley**
963 **FQ, Ciochina MI, Cordeiro AL, Dallstream C, et al. 2021.** Tradeoffs and synergies in
964 tropical forest root traits and dynamics for nutrient and water acquisition: Field and
965 modeling advances. *Frontiers in Forests and Global Change* **4**.

966 **Cusack DF, Markesteijn L, Condit R, Lewis OT, Turner BL. 2018.** Soil carbon stocks across
967 tropical forests of Panama regulated by base cation effects on fine roots. *Biogeochemistry*
968 **137**: 253-266.

969 **Cusack DF, Turner BL. 2021.** Fine root and soil organic carbon depth distributions are inversely
970 related across fertility and rainfall gradients in lowland tropical forests. *Ecosystems* **24**:
971 1075-1092.

972 **Davies-Barnard, T., Zaehle, S. and Friedlingstein, P., 2022.** Assessment of the impacts of
973 biological nitrogen fixation structural uncertainty in CMIP6 earth system
974 models. *Biogeosciences*, **19**(14): 3491-3503

975 **Dallstream C, Weemstra M, Soper FM. 2023.** A framework for fine-root trait syndromes:
976 syndrome coexistence may support phosphorus partitioning in tropical forests. *Oikos*.

977 **Daly DC, Fine PVA. 2018.** Generic limits re-visited and an updated sectional classification for
978 Protium (tribe Protieae). Studies in Neotropical Burseraceae XXV. *Brittonia* **70**: 418-426.

979 **de Paula MD, Forrest M, Langan L, Bendix J, Homeier J, Velescu A, Wilcke W, Hickler T.**
980 **2021.** Nutrient cycling drives plant community trait assembly and ecosystem functioning
981 in a tropical mountain biodiversity hotspot. *New Phytologist* **232**: 551-566.

982 **Domec, J.-C., Scholz, F.G., Bucci, S.J., Meinzer, F.C., Goldstein, G., Villalobos-Vega, R.,**
983 **2006.** Diurnal and seasonal variation in root xylem embolism in Neotropical savanna
984 woody species: impact on stomatal control of plant water status. *Plant, Cell & Environment*
985 **29**, 26-35.

986 **Drewniak BA. 2019.** Simulating dynamic roots in the Energy Exascale Earth System Land Model.
987 *Journal of Advances in Modeling Earth Systems* **11**: 338-359.

988 **Dybzinski R, Kelvakis A, McCabe J, Panock S, Anuchitlertchon K, Vasarhelyi L,**
989 **McCormack ML, McNickle GG, Poorter H, Trinder C, Farrior CE, 2019.** How are
990 nitrogen availability, fine-root mass, and nitrogen uptake related empirically? Implications
991 for models and theory. *Global Change Biology* **25**: 885-899.

992 **Eiserhardt WL, Couvreur TLP, Baker WJ. 2017.** Plant phylogeny as a window on the evolution
993 of hyperdiversity in the tropical rainforest biome. *New Phytologist* **214**: 1408-1422.

994 **Espeleta JF, Clark DA. 2007.** Multi-scale variation in fine-root biomass in a tropical rain forest:
995 A seven-year study. *Ecological Monographs* **77**: 377-404.

996 **Fer I, Gardella AK, Shiklomanov AN, Campbell EE, Cowdery EM, De Kauwe MG, Desai A,**
997 **Duveneck MJ, Fisher JB, Haynes KD, et al. 2021.** Beyond ecosystem modeling: A
998 roadmap to community cyberinfrastructure for ecological data-model integration. *Global*
999 *Change Biology* **27**: 13-26.

1000 **Field CB, Behrenfeld MJ, Randerson JT, Falkowski P. 1998.** Primary production of the
1001 biosphere: Integrating terrestrial and oceanic components. *Science* **281**: 237-240.

1002 **Finer L, Ohashi M, Noguchi K, Hirano Y. 2011.** Fine root production and turnover in forest
1003 ecosystems in relation to stand and environmental characteristics. *Forest Ecology and*
1004 *Management* **262**: 2008-2023.

1005 **Fischer R, Bohn F, de Paula MD, Dislich C, Groeneveld J, Gutierrez AG, Kazmierczak M,**
1006 **Knapp N, Lehmann S, Paulick S, et al. 2016.** Lessons learned from applying a forest gap
1007 model to understand ecosystem and carbon dynamics of complex tropical forests.
1008 *Ecological Modelling* **326**: 124-133.

1009 **Fisher, J.B., Sitch, S., Malhi, Y., Fisher, R.A., Huntingford, C., Tan, S.-Y., 2010.** Carbon cost
1010 of plant nitrogen acquisition: A mechanistic, globally-applicable model of plant nitrogen
1011 uptake, retranslocation and fixation. *Global Biogeochemical Cycles* **24**: GB1014.

1012 **Fisher JB, Badgley G, Blyth E. 2012.** Global nutrient limitation in terrestrial vegetation. *Global*
1013 *Biogeochemical Cycles* **26**.

1014 **Fisher JB, Perakalapudi NV, Turner BL, Schimel DS, Cusack DF. 2020.** Competing effects
1015 of soil fertility and toxicity on tropical greening. *Scientific Reports* **10**.

1016 **Fleischer K, Rammig A, De Kauwe MG, Walker AP, Domingues TF, Fuchslueger L, Garcia**
1017 **S, Goll DS, Grandis A, Jiang MK, et al. 2019.** Amazon forest response to CO₂
1018 fertilization dependent on plant phosphorus acquisition. *Nature Geoscience* **12**: 736-+.

1019 **Fortunel C, Paine CET, Fine PVA, Kraft NJB, Baraloto C. 2014.** Environmental factors predict
1020 community functional composition in Amazonian forests. *Journal of Ecology* **102**: 145-
1021 155.

1022 **Freschet GT, Pages L, Iversen CM, Comas LH, Rewald B, Roumet C, Klimesova J,**
1023 **Zadworny M, Poorter H, Postma JA, et al. 2021a.** A starting guide to root ecology:
1024 strengthening ecological concepts and standardising root classification, sampling,
1025 processing and trait measurements. *New Phytologist* **232**: 973-1122.

1026 **Freschet GT, Roumet C, Comas LH, Weemstra M, Bengough AG, Rewald B, Bardgett RD,**
1027 **De Deyn GB, Johnson D, Klimesova J, et al. 2021b.** Root traits as drivers of plant and

1028 ecosystem functioning: current understanding, pitfalls and future research needs. *New*
1029 *Phytologist* **232**: 1123-1158.

1030 **Fyllas NM, Quesada CA, Lloyd J. 2012.** Deriving Plant Functional Types for Amazonian forests
1031 for use in vegetation dynamics models. *Perspectives in Plant Ecology Evolution and*
1032 *Systematics* **14**: 97-110.

1033 **Germon A, Laclau JP, Robin A, Jourdan C. 2020.** Tamm Review: Deep fine roots in forest
1034 ecosystems: Why dig deeper? *Forest Ecology and Management* **466**.

1035 **Guerrero-Ramirez NR, Mommer L, Freschet GT, Iversen CM, McCormack ML, Kattge J,**
1036 **Poorter H, van der Plas F, Bergmann J, Kuyper TW, et al. 2021.** Global root traits
1037 (GRooT) database. *Global Ecology and Biogeography* **30**: 25-37.

1038 **Guilbeault-Mayers X, Turner BL, Laliberte E. 2020.** Greater root phosphatase activity of
1039 tropical trees at low phosphorus despite strong variation among species. *Ecology* **101**.

1040 **Guo D, Xia M, Wei X, Chang W, Liu Y, Wang Z. 2008.** Anatomical traits associated with
1041 absorption and mycorrhizal colonization are linked to root branch order in twenty-
1042 three Chinese temperate tree species. *New Phytologist* **180**: 673-683.

1043 **Han MG, Chen Y, Li R, Yu M, Fu LC, Li SF, Su JR, Zhu B. 2022.** Root phosphatase activity
1044 aligns with the collaboration gradient of the root economics space. *New Phytologist* **234**:
1045 837-849.

1046 **Hart TB, Hart JA, Murphy PG. 1989.** Monodominant and species-rich forests of the humid
1047 tropics - causes for their co-occurrence. *American Naturalist* **133**: 613-633.

1048 **Haverd V, Smith B, Nieradzik L, Briggs PR, Woodgate W, Trudinger CM, Canadell JG,**
1049 **Cuntz M. 2018.** A new version of the CABLE land surface model (Subversion revision
1050 r4601) incorporating land use and land cover change, woody vegetation demography, and

1051 a novel optimisation-based approach to plant coordination of photosynthesis. *Geoscientific*
1052 *Model Development* **11**: 2995-3026.

1053 **Hengl T, de Jesus JM, Heuvelink GBM, Gonzalez MR, Kilibarda M, Blagotic A, Shangguan**
1054 **W, Wright MN, Geng XY, Bauer-Marschallinger B, et al. 2017.** SoilGrids250m: Global
1055 gridded soil information based on machine learning. *PLOS ONE* **12**.

1056 **Holbrook NM, Whitbeck JL, Mooney HA 1995.** Drought responses of neotropical dry forest
1057 trees. In: Medina E, Mooney HA, Bullock SH eds. *Seasonally Dry Tropical Forests*.
1058 Cambridge: Cambridge University Press, 243-276.

1059 **Holdaway RJ, Richardson SJ, Dickie IA, Peltzer DA, Coomes DA. 2011.** Species- and
1060 community-level patterns in fine root traits along a 120,000-year soil chronosequence in
1061 temperate rain forest. *Journal of Ecology* **99**: 954-963.

1062 **Houlton BZ, Wang YP, Vitousek PM, Field CB. 2008.** A unifying framework for dinitrogen
1063 fixation in the terrestrial biosphere. *Nature* **454**: 327-U334.

1064 **Huasco WH, Riutta T, Girardin CAJ, Pacha FH, Vilca BPL, Moore S, Rifai SW, del Aguila-**
1065 **Pasquel J, Murakami AA, Freitag R, et al. 2021.** Fine root dynamics across pantropical
1066 rainforest ecosystems. *Global Change Biology* **27**: 3657-3680.

1067 **Hungate BA, Dukes JS, Shaw MR, Luo YQ, Field CB. 2003.** Nitrogen and climate change.
1068 *Science* **302**: 1512-1513.

1069 **Jackson RB, Canadell J, Ehleringer JR, Mooney HA, Sala OE, Schulze ED. 1996.** A global
1070 analysis of root distributions for terrestrial biomes. *Oecologia* **108**: 389-411.

1071 **Janos DP, Scott J, Bowman D. 2008.** Temporal and spatial variation of fine roots in a northern
1072 Australian *Eucalyptus tetrodonta* savanna. *Journal of Tropical Ecology* **24**: 177-188.

1073 **Janzen DH. 1974.** Tropical black water rivers animals and mast fruiting by the Dipterocarpaceae.
1074 *Biotropica* **6**: 69-103.

1075 **Jiang XY, Jia X, Gao SJ, Jiang Y, Wei NN, Han C, Zha TS, Liu P, Tian Y, Qin SG. 2021.**
1076 Plant nutrient contents rather than physical traits are coordinated between leaves and roots
1077 in a desert shrubland. *Frontiers in Plant Science* **12**.

1078 **Jobbágy EG, Jackson RB. 2000.** The vertical distribution of soil organic carbon and its relation
1079 to climate and vegetation. *Ecological Applications* **10**: 423-436.

1080 **Jones FA, Erickson DL, Bernal MA, Bermingham E, Kress WJ, Herre EA, Muller-Landau
1081 HC, Turner BL. 2011.** The roots of diversity: Below ground species richness and rooting
1082 distributions in a tropical forest revealed by DNA barcodes and inverse modeling. *Plos
1083 One* **6**.

1084 **Joshi J, Stocker BD, Hofhansl F, Zhou SX, Dieckmann U, Prentice IC. 2022.** Towards a
1085 unified theory of plant photosynthesis and hydraulics. *Nature Plants* **8**: 1304-+.

1086 **Kaiser C, Kilburn MR, Clode PL, Fuchslueger L, Koranda M, Cliff JB, Solaiman ZM,
1087 Murphy DV. 2015.** Exploring the transfer of recent plant photosynthates to soil microbes:
1088 mycorrhizal pathway vs direct root exudation. *New Phytologist* **205**: 1537-1551.

1089 **Keller AB, Phillips RP. 2019.** Leaf litter decay rates differ between mycorrhizal groups in
1090 temperate, but not tropical, forests. *New Phytologist* **222**: 556-564.

1091 **Kennedy D, Swenson S, Oleson KW, Lawrence DM, Fisher R, da Costa ACL, Gentine P.
1092 2019.** Implementing plant hydraulics in the Community Land Model, Version 5. *Journal
1093 of Advances in Modeling Earth Systems* **11**: 485-513.

1094 **Knox RG, Koven CD, Riley WJ, Walker AP, Wright SJ, Holm JA, Wei X, Fisher RA, Zhu**
1095 **Q, Tang J, et al. 2023.** Nutrient dynamics in a coupled terrestrial biosphere and land model
1096 (ELM-FATES). [preprint], <https://doi.org/10.22541/essoar.167810418.80767445/v1>

1097 **Kong DL, Ma CG, Zhang Q, Li L, Chen XY, Zeng H, Guo DL. 2014.** Leading dimensions in
1098 absorptive root trait variation across 96 subtropical forest species. *New Phytologist* **203**:
1099 863-872.

1100 **Kou-Giesbrecht S, Malyshev S, Cano IM, Pacala SW, Shevliakova E, Bytnerowicz TA,**
1101 **Menge DNL. 2021.** A novel representation of biological nitrogen fixation and competitive
1102 dynamics between nitrogen-fixing and non-fixing plants in a land model (GFDL LM4.1-
1103 BNF). *Biogeosciences* **18**: 4143-4183.

1104 **Koven CD, Knox RG, Fisher RA, Chambers JQ, Christoffersen BO, Davies SJ, Detto M,**
1105 **Dietze MC, Faybishenko B, Holm J, et al. 2020.** Benchmarking and parameter sensitivity
1106 of physiological and vegetation dynamics using the Functionally Assembled Terrestrial
1107 Ecosystem Simulator (FATES) at Barro Colorado Island, Panama. *Biogeosciences* **17**:
1108 3017-3044.

1109 **Kummerow J, Castillanos J, Maas M, Larigauderie A. 1990.** Production of fine roots and the
1110 seasonality of their growth in a Mexican deciduous dry forest. *Vegetatio* **90**: 73-80.

1111 **Kyker-Snowman E, Lombardozzi DL, Bonan GB, Cheng SJ, Dukes JS, Frey SD, Jacobs EM,**
1112 **McNellis R, Rady JM, Smith NG, et al. 2022.** Increasing the spatial and temporal impact
1113 of ecological research: A roadmap for integrating a novel terrestrial process into an Earth
1114 system model. *Global Change Biology* **28**: 665-684.

- 1115 **Langan L, Higgins SI, Scheiter S. 2017.** Climate-biomes, pedo-biomes or pyro-biomes: which
1116 world view explains the tropical forest-savanna boundary in South America? *Journal of*
1117 *Biogeography* **44**: 2319-2330.
- 1118 **Laughlin DC, Mommer L, Sabatini FM, Bruelheide H, Kuyper TW, McCormack ML,**
1119 **Bergmann J, Freschet GT, Guerrero-Ramirez NR, Iversen CM, et al. 2021.** Root traits
1120 explain plant species distributions along climatic gradients yet challenge the nature of
1121 ecological trade-offs. *Nature Ecology & Evolution* **5**: 1123-+.
- 1122 **Lee RB. 1988.** Phosphate influx and extracellular phosphatase activity in barley roots and rose
1123 cells. *New Phytologist* **109**: 141-148.
- 1124 **Lekberg Y, Rosendahl S, Michelsen A, Olsson PA. 2013.** Seasonal carbon allocation to
1125 arbuscular mycorrhizal fungi assessed by microscopic examination, stable isotope probing
1126 and fatty acid analysis. *Plant and Soil* **368**: 547-555.
- 1127 **Longo M, Knox RG, Medvigy DM, Levine NM, Dietze MC, Kim Y, Swann ALS, Zhang K,**
1128 **Rollinson CR, Bras RL, et al. 2019.** The biophysics, ecology, and biogeochemistry of
1129 functionally diverse, vertically and horizontally heterogeneous ecosystems: the Ecosystem
1130 Demography model, version 2.2-Part 1: Model description. *Geoscientific Model*
1131 *Development* **12**: 4309-4346.
- 1132 **Lugli LF, Andersen KM, Aragao L, Cordeiro AL, Cunha HKV, Fuchslueger L, Meir P,**
1133 **Mercado LM, Oblitas E, Quesada CA, et al. 2020.** Multiple phosphorus acquisition
1134 strategies adopted by fine roots in low-fertility soils in Central Amazonia. *Plant and Soil*
1135 **450**: 49-63.
- 1136 **Lugli LF, Rosa JS, Andersen KM, Di Ponzio R, Almeida RV, Pires M, Cordeiro AL, Cunha**
1137 **HFV, Martins NP, Assis RL, et al. 2021.** Rapid responses of root traits and productivity

1138 to phosphorus and cation additions in a tropical lowland forest in Amazonia. *New*
1139 *Phytologist* **230**: 116-128.

1140 **Ma ZQ, Guo DL, Xu XL, Lu MZ, Bardgett RD, Eissenstat DM, McCormack ML, Hedin LO.**
1141 **2018.** Evolutionary history resolves global organization of root functional traits. *Nature*
1142 **555**: 94-+.

1143 **Malhi Y, Doughty C, Galbraith D. 2011.** The allocation of ecosystem net primary productivity
1144 in tropical forests. *Philosophical Transactions of the Royal Society B-Biological Sciences*
1145 **366**: 3225-3245.

1146 **Marechaux I, Chave J. 2017.** An individual-based forest model to jointly simulate carbon and
1147 tree diversity in Amazonia: description and applications. *Ecological Monographs* **87**: 632-
1148 664.

1149 **Markestijn L, Poorter L. 2009.** Seedling root morphology and biomass allocation of 62 tropical
1150 tree species in relation to drought- and shade-tolerance. *Journal of Ecology* **97**: 311-325.

1151 **Matamala R, Stover DB. 2013.** Introduction to a virtual special issue: Modeling the hidden half
1152 - the root of our problem. *New Phytologist* **200**: 939-942.

1153 **McCormack ML, Dickie IA, Eissenstat DM, Fahey TJ, Fernandez CW, Guo DL, Helmisaari**
1154 **HS, Hobbie EA, Iversen CM, Jackson RB, et al. 2015.** Redefining fine roots improves
1155 understanding of belowground contributions to terrestrial biosphere processes. *New*
1156 *Phytologist* **207**: 505-518.

1157 **McMurtrie RE, Nasholm T. 2018.** Quantifying the contribution of mass flow to nitrogen
1158 acquisition by an individual plant root. *New Phytologist* **218**: 119-130.

1159 **Medlyn BE, Zaehle S, De Kauwe MG, Walker AP, Dietze MC, Hanson PJ, Hickler T, Jain**
1160 **AK, Luo YQ, Parton W, et al. 2015.** Using ecosystem experiments to improve vegetation
1161 models. *Nature Climate Change* **5**: 528-534.

1162 **Medvigy D, Moorcroft PR. 2012.** Predicting ecosystem dynamics at regional scales: an
1163 evaluation of a terrestrial biosphere model for the forests of northeastern North America.
1164 *Philosophical Transactions of the Royal Society B-Biological Sciences* **367**: 222-235.

1165 **Medvigy D, Wofsy SC, Munger JW, Hollinger DY, Moorcroft PR. 2009.** Mechanistic scaling
1166 of ecosystem function and dynamics in space and time: Ecosystem Demography model
1167 version 2. *Journal of Geophysical Research: Biogeosciences* **114**.

1168 **Menge DNL, Lichstein JW, Angeles-Perez G. 2014.** Nitrogen fixation strategies can explain the
1169 latitudinal shift in nitrogen-fixing tree abundance. *Ecology* **95**: 2236-2245.

1170 **Metcalfe DB, Meir P, Aragao L, da Costa ACL, Braga AP, Goncalves PHL, Silva JD, de**
1171 **Almeida SS, Dawson LA, Malhi Y, et al. 2008.** The effects of water availability on root
1172 growth and morphology in an Amazon rainforest. *Plant and Soil* **311**: 189-199.

1173 **Nakhavali MA, Mercado LM, Hartley IP, Sitch S, Cunha FV, di Ponzio R, Lugli LF,**
1174 **Quesada CA, Andersen KM, Chadburn SE, et al. 2022.** Representation of the
1175 phosphorus cycle in the Joint UK Land Environment Simulator (vn5.5_JULES-CNP).
1176 *Geoscientific Model Development* **15**: 5241-5269.

1177 **Nasto MK, Alvarez-Clare S, Lekberg Y, Sullivan BW, Townsend AR, Cleveland CC. 2014.**
1178 Interactions among nitrogen fixation and soil phosphorus acquisition strategies in lowland
1179 tropical rain forests. *Ecology Letters* **17**: 1282-1289.

1180 **Ng JC. 2022.** Unearthing in-situ nutrient uptake rates: an assessment of underlying assumptions
1181 used for their prediction. Final Year Project (FYP), Nanyang Technological University,
1182 Singapore. <https://hdl.handle.net/10356/156728>

1183 **Oliveira RS, Costa FRC, van Baalen E, de Jonge A, Bittencourt PR, Almanza Y, Barros FD,**
1184 **Cordoba EC, Fagundes MV, Garcia S, et al. 2019.** Embolism resistance drives the
1185 distribution of Amazonian rainforest tree species along hydro-topographic gradients. *New*
1186 *Phytologist* **221**: 1457-1465.

1187 **Oliveira RS, Eller CB, Barros FD, Hirota M, Brum M, Bittencourt P. 2021.** Linking plant
1188 hydraulics and the fast-slow continuum to understand resilience to drought in tropical
1189 ecosystems. *New Phytologist* **230**: 904-923.

1190 **Olsson PA, Lekberg Y. 2022.** A critical review of the use of lipid signature molecules for the
1191 quantification of arbuscular mycorrhiza fungi. *Soil Biology & Biochemistry* **166**.

1192 **Peh KSH, Lewis SL, Lloyd J. 2011.** Mechanisms of monodominance in diverse tropical tree-
1193 dominated systems. *Journal of Ecology* **99**: 891-898.

1194 **Phillips RP, Brzostek E, Midgley MG. 2013.** The mycorrhizal-associated nutrient economy: a
1195 new framework for predicting carbon-nutrient couplings in temperate forests. *New*
1196 *Phytologist* **199**: 41-51.

1197 **POWO. 2023.** Plants of the World Online. Facilitated by the Royal Botanic Gardens, Kew.
1198 Published on the Internet; <http://www.plantsoftheworldonline.org/>

1199 **Quesada CA, Lloyd J, Anderson LO, Fyllas NM, Schwarz M, Czimczik CI. 2011.** Soils of
1200 Amazonia with particular reference to the RAINFOR sites. *Biogeosciences* **8**: 1415-1440.

1201 **Rasse DP, Rumpel C, Dignac MF. 2005.** Is soil carbon mostly root carbon? Mechanisms for a
1202 specific stabilisation. *Plant and Soil* **269**: 341-356.

1203 **Read QD, Henning JA, Sanders NJ. 2017.** Intraspecific variation in traits reduces ability of trait-
1204 based models to predict community structure. *Journal of Vegetation Science* **28**: 1070-
1205 1081.

1206 **Reich, P.B., 2014.** The world-wide “fast–slow” plant economics spectrum: a traits manifesto.
1207 *Journal of Ecology* **102**: 275-301

1208 **Reichert T, Rammig A, Fuchslueger L, Lugli LF, Quesada CA, Fleischer K. 2022.** Plant
1209 phosphorus-use and -acquisition strategies in Amazonia. *New Phytologist* **234**: 1126-1143.

1210 **Rius BF, Darela JP, Fleischer K, Hofhansl F, Blanco CC, Rammig A, Domingues TF, Lapola**
1211 **DM. 2023.** Higher functional diversity improves modeling of Amazon forest carbon
1212 storage. *Ecological Modelling* **481**.

1213 **Russo SE, Kochsiek A, Olney J, Thompson L, Miller AE, Tan S. 2013.** Nitrogen uptake
1214 strategies of edaphically specialized Bornean tree species. *Plant Ecology* **214**: 1405-1416.

1215 **Sakschewski B, von Bloh W, Boit A, Rammig A, Kattge J, Poorter L, Penuelas J, Thonicke**
1216 **K. 2015.** Leaf and stem economics spectra drive diversity of functional plant traits in a
1217 dynamic global vegetation model. *Global Change Biology* **21**: 2711-2725.

1218 **Sakschewski B, von Bloh W, Druke M, Sorensson AA, Ruscica R, Langerwisch F, Billing M,**
1219 **Bereswill S, Hirota M, Oliveira RS, et al. 2021.** Variable tree rooting strategies are key
1220 for modelling the distribution, productivity and evapotranspiration of tropical evergreen
1221 forests. *Biogeosciences* **18**: 4091-4116.

1222 **Sampaio EVSB 1995.** Overview of the Brazilian caatinga. In: Medina E, Mooney HA, Bullock
1223 SH eds. *Seasonally Dry Tropical Forests*. Cambridge: Cambridge: Cambridge University
1224 Press, 35-63.

1225 **Schaphoff S, von Bloh W, Rammig A, Thonicke K, Biemans H, Forkel M, Gerten D, Heinke**
1226 **J, Jagermeyr J, Knauer J, et al. 2018.** LPJmL4-a dynamic global vegetation model with
1227 managed land - Part 1: Model description. *Geoscientific Model Development* **11**: 1343-
1228 1375.

1229 **Scheiter S, Langan L, Higgins SI. 2013.** Next-generation dynamic global vegetation models:
1230 learning from community ecology. *New Phytologist* **198**: 957-969.

1231 **Schreeg LA, Santiago LS, Wright SJ, Turner BL. 2014.** Stem, root, and older leaf N:P ratios
1232 are more responsive indicators of soil nutrient availability than new foliage. *Ecology* **95**:
1233 2062-2068.

1234 **Seland Ø, Bentsen M, Olivie D, Toniazzo T, Gjermundsen A, Graff LS, Debernard JB, Gupta**
1235 **AK, He YC, Kirkevåg A, et al. 2020.** Overview of the Norwegian Earth System Model
1236 (NorESM2) and key climate response of CMIP6 DECK, historical, and scenario
1237 simulations. *Geoscientific Model Development* **13**: 6165-6200.

1238 **Sheldrake M, Rosenstock NP, Mangan S, Revillini D, Sayer EJ, Olsson PA, Verbruggen E,**
1239 **Tanner EVJ, Turner BL, Wright SJ. 2018.** Responses of arbuscular mycorrhizal fungi
1240 to long-term inorganic and organic nutrient addition in a lowland tropical forest. *Isme*
1241 *Journal* **12**: 2433-2445.

1242 **Shi M, Fisher JB, Brzostek ER, Phillips RP. 2016.** Carbon cost of plant nitrogen acquisition:
1243 global carbon cycle impact from an improved plant nitrogen cycle in the Community Land
1244 Model. *Global Change Biology* **22**: 1299-1314.

1245 **Smith-Martin CM, Muscarella R, Ankori-Karlinsky R, Delzon S, Farrar SL, Salva-Sauri M,**
1246 **Thompson J, Zimmerman JK, Uriarte M. 2022.** Hurricanes increase tropical forest
1247 vulnerability to drought. *New Phytologist* **235**: 1005-1017.

1248 **Smith-Martin CM, Xu XT, Medvigy D, Schnitzer SA, Powers JS. 2020.** Allometric scaling
1249 laws linking biomass and rooting depth vary across ontogeny and functional groups in
1250 tropical dry forest lianas and trees. *New Phytologist* **226**: 714-726.

1251 **Smith B, Warlind D, Arneth A, Hickler T, Leadley P, Siltberg J, Zaehle S. 2014.** Implications
1252 of incorporating N cycling and N limitations on primary production in an individual-based
1253 dynamic vegetation model. *Biogeosciences* **11**: 2027-2054.

1254 **Sobrado MA, Cuenca G. 1979.** Aspects of water-use of deciduous and evergreen species in a
1255 tropical dry forest of Venezuela. *Acta Cientifica Venezolana* **30**: 302-308.

1256 **Soper FM, Nasto MK, Osborne BB, Cleveland CC. 2019.** Nitrogen fixation and foliar nitrogen
1257 do not predict phosphorus acquisition strategies in tropical trees. *Journal of Ecology* **107**:
1258 118-126.

1259 **Sousa D, Fisher JB, Galvan FR, Pavlick RP, Cordelli S, Giambelluca TW, Giardina CP,
1260 Gilbert GS, Imran-Narahari F, Litton CM, et al. 2021.** Tree canopies reflect
1261 mycorrhizal composition. *Geophysical Research Letters* **48**: 1-9.

1262 **Thonicke K, Billing M, von Bloh W, Sakschewski B, Niinemets U, Penuelas J, Cornelissen
1263 JHC, Onoda Y, van Bodegom P, Schaepman ME, et al. 2020.** Simulating functional
1264 diversity of European natural forests along climatic gradients. *Journal of Biogeography*
1265 **47**: 1069-1085.

1266 **Thornton PE, Lamarque JF, Rosenbloom NA, Mahowald NM. 2007.** Influence of carbon-
1267 nitrogen cycle coupling on land model response to CO₂ fertilization and climate variability.
1268 *Global Biogeochemical Cycles* **21**.

1269 **Thum T, Nabel J, Tsuruta A, Aalto T, Dlugokencky EJ, Liski J, Luijkx IT, Markkanen T,
1270 Pongratz J, Yoshida Y, et al. 2020.** Evaluating two soil carbon models within the global

1271 land surface model JSBACH using surface and spaceborne observations of atmospheric
1272 CO₂. *Biogeosciences* **17**: 5721-5743.

1273 **Ushio M, Fujiki Y, Hidaka A, Kitayama K. 2015.** Linkage of root physiology and morphology
1274 as an adaptation to soil phosphorus impoverishment in tropical montane forests. *Functional*
1275 *Ecology* **29**: 1235-1245.

1276 **Valverde-Barrantes OJ, Authier L, Schimann H, Baraloto C. 2021.** Root anatomy helps to
1277 reconcile observed root trait syndromes in tropical tree species. *American Journal of*
1278 *Botany* **108**: 744-755.

1279 **Vitousek PM, Sanford RL. 1986.** Nutrient cycling in moist tropical forest. *Annual Review of*
1280 *Ecology and Systematics* **17**: 137-167.

1281 **Vleminckx J, Fortunel C, Valverde-Barrantes O, Paine CET, Engel J, Petronelli P, Dourdain**
1282 **AK, Guevara J, Beroujon S, Baraloto C. 2021.** Resolving whole-plant economics from
1283 leaf, stem and root traits of 1467 Amazonian tree species. *Oikos* **130**: 1193-1208.

1284 **Walker AP, Hanson PJ, De Kauwe MG, Medlyn BE, Zaehle S, Asao S, Dietze M, Hickler T,**
1285 **Huntingford C, Iversen CM, et al. 2014.** Comprehensive ecosystem model-data synthesis
1286 using multiple data sets at two temperate forest free-air CO₂ enrichment experiments:
1287 Model performance at ambient CO₂ concentration. *Journal of Geophysical Research:*
1288 *Biogeosciences* **119**: 937-964.

1289 **Wang B, McCormack ML, Ricciuto DM, Yang XJ, Iversen CM. 2023.** Embracing fine-root
1290 system complexity in terrestrial ecosystem modeling. *Global Change Biology* **29**: 2871–
1291 2885.

1292 **Wang ZQ, Gong HY, Sardans J, Zhou QP, Deng JM, Niklas KJ, Hu HF, Li YL, Ma ZQ,**
1293 **Mipam TD, et al. 2022.** Divergent nitrogen and phosphorus allocation strategies in

1294 terrestrial plant leaves and fine roots: A global meta-analysis. *Journal of Ecology* **110**:
1295 2745-2758.

1296 **Warren JM, Hanson PJ, Iversen CM, Kumar J, Walker AP, Wullschleger SD. 2015.** Root
1297 structural and functional dynamics in terrestrial biosphere models - evaluation and
1298 recommendations. *New Phytologist* **205**: 59-78.

1299 **Weemstra M, Mommer L, Visser EJW, van Ruijven J, Kuypers TW, Mohren GMJ, Sterck**
1300 **FJ. 2016.** Towards a multidimensional root trait framework: a tree root review. *New*
1301 *Phytologist* **211**: 1159-1169.

1302 **Weemstra M, Peay KG, Davies SJ, Mohamad M, Itoh A, Tan S, Russo SE. 2020.** Lithological
1303 constraints on resource economies shape the mycorrhizal composition of a Bornean
1304 rain forest. *New Phytologist* **228**: 253-268.

1305 **Weemstra M, Valverde-Barrantes O, Fortunel C, Oblitas Mendoza E, Prata E, Vásquez**
1306 **Pilco M, Vicentini A, Vleminckx J, Baraloto C. 2023.** Weak phylogenetic and habitat
1307 effects on root trait variation of 218 Neotropical tree species. *Frontiers in Forests and*
1308 *Global Change* **6**: 1187127.

1309 **Weng ES, Malyshev S, Lichstein JW, Farrior CE, Dybzinski R, Zhang T, Shevliakova E,**
1310 **Pacala SW. 2015.** Scaling from individual trees to forests in an Earth system modeling
1311 framework using a mathematically tractable model of height-structured competition.
1312 *Biogeosciences* **12**: 2655-2694.

1313 **Withington JM, Reich PB, Oleksyn J, Eissenstat DM. 2006.** Comparisons of structure and life
1314 span in roots and leaves among temperate trees. *Ecological Monographs* **76**: 381-397.

1315 **Wullschleger SD, Epstein HE, Box EO, Euskirchen ES, Goswami S, Iversen CM, Kattge J,**
1316 **Norby RJ, van Bodegom PM, Xu XF. 2014.** Plant functional types in Earth system

1317 models: past experiences and future directions for application of dynamic vegetation
1318 models in high-latitude ecosystems. *Annals of Botany* **114**: 1-16.

1319 **Wurzburger N, Wright SJ. 2015.** Fine-root responses to fertilization reveal multiple nutrient
1320 limitation in a lowland tropical forest. *Ecology* **96**: 2137-2146.

1321 **Xu C, Christoffersen B, Robbins Z, Knox R, Fisher RA, Chitra-Tarak R, Slot M, Solander**
1322 **K, Kueppers L, Koven C, et al. 2023.** Quantification of hydraulic trait control on plant
1323 hydrodynamics and risk of hydraulic failure within a demographic structured vegetation
1324 model in a tropical forest (FATES-HYDRO V1.0). *EGUsphere* **2023**: 1-32.

1325 **Xu XT, Medvigy D, Powers JS, Becknell JM, Guan KY. 2016.** Diversity in plant hydraulic
1326 traits explains seasonal and inter-annual variations of vegetation dynamics in seasonally
1327 dry tropical forests. *New Phytologist* **212**: 80-95.

1328 **Yaffar D, Defrenne CE, Cabugao KG, Kivlin SN, Childs J, Carvajal N, Norby RJ. 2021.**
1329 Trade-offs in phosphorus acquisition strategies of five common tree species in a tropical
1330 forest of Puerto Rico. *Frontiers in Forests and Global Change* **4**.

1331 **Yang XJ, Ricciuto DM, Thornton PE, Shi MY, Xu M, Hoffman F, Norby RJ. 2019.** The effects
1332 of phosphorus cycle dynamics on carbon sources and sinks in the Amazon region: A
1333 Modeling Study Using ELM v1. *Journal of Geophysical Research-Biogeosciences* **124**:
1334 3686-3698.

1335 **Yang XQ, Wu JP, Chen XZ, Ciais P, Maignan F, Yuan WP, Piao SL, Yang S, Gong FX, Su**
1336 **YX, et al. 2021.** A comprehensive framework for seasonal controls of leaf abscission and
1337 productivity in evergreen broadleaved tropical and subtropical forests. *Innovation* **2**.

1338 **Yu L, Ahrens B, Wutzler T, Schrumpf M, Zaehle S. 2020.** Jena Soil Model (JSM v1.0; revision
1339 1934): a microbial soil organic carbon model integrated with nitrogen and phosphorus
1340 processes. *Geosci. Model Dev.* **13**: 783-803.

1341 **Zaehle S, Medlyn BE, De Kauwe MG, Walker AP, Dietze MC, Hickler T, Luo YQ, Wang**
1342 **YP, El-Masri B, Thornton P, et al. 2014.** Evaluation of 11 terrestrial carbon-nitrogen
1343 cycle models against observations from two temperate free-air CO₂ enrichment studies.
1344 *New Phytologist* **202**: 803-822.

1345 **Zhang Y, Cao J, Lu M, Kardol P, Wang J, Fan G, Kong D. 2023.** The origin of bi-
1346 dimensionality in plant root traits. *Trends in Ecology & Evolution* (in press).
1347 <https://doi.org/10.1016/j.tree.2023.09.002>

1348 **Zhu LQ, Huang RZ, Wang JP, Huang GM, Guan HZ, Lin LJ, Yang MJ, Li YY, Zou XH.**
1349 **2023.** Litter, root, and mycorrhiza input affected soil microbial community structure in
1350 schima superba pure forest in subtropical China. *Diversity-Basel* **15**.

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1353 **Supporting Information**

Filename	Description
SI_NotesS1_NewPhytViewpoint_Cusack_Final.pdf	Notes S1 contains methodological details and results for main text and boxes
Dataset S2_Box1data_RootTraits_Panama.csv	Dataset S2 provides raw data for Panama roots used in Box 1
Dataset S3_Box1data_RootTraits_PuertoRico.csv	Dataset S3 provides raw data for Puerto Rico roots used in Box 1
Dataset S4_Box2data_MaximumRootingDepthData_CostaRica.csv	Dataset S4 provides raw data for Costa Rica roots used in Box 2
Dataset S5_Box3data_uptake_Panama.csv	Dataset S5 provides raw data for Panama roots used in Box 3
Dataset S6_Box3data_uptake_Singapore.csv	Dataset S6 provides raw data for Singapore roots referenced in Box 3 and in Notes S1
Dataset S7_Box4data_Panama_stoich.csv	Dataset S7 provides raw data for Panama root stoichiometry used in Box 4

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