

The Resilience of Amazon Tree Cover to Past and Present Drying

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Abstract

The Amazon forest is increasingly vulnerable to dieback and encroachment of grasslands and agricultural fields. Threats to these forested ecosystems include drying, deforestation, and fire, but feedbacks among these make it difficult to determine their relative importance. Here, we reconstruct the central and western Amazon tree cover response to aridity and fire in the mid-Holocene—a time of less intensive human land use and markedly drier conditions than today—to assess the resilience of tree cover to drying and the strength of vegetation-climate feedbacks. We use pollen, charcoal, and speleothem oxygen isotope proxy data to show that Amazon tree cover in the mid-Holocene was resilient to drying in excess of the driest bias-corrected future precipitation projections. Experiments with a dynamic global vegetation model (LPJ-GUESS) suggest tree cover resilience may be owed to weak feedbacks that act to amplify tree cover loss with drying. We also compare these results to observational data and find that, under limited human interference, modern tree cover is likely similarly resilient to mid-Holocene levels of aridification. Our results suggest human-driven fire and deforestation likely

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pose a greater threat to the future of Amazon ecosystems than drying alone.

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1. Introduction

Ecological resilience is a measure of how much disturbance, or forcing, a system can absorb without changing its state [1, 2, 3]. The state of the system can be defined in a number of ways, but usually refers to its feedbacks and/or function [1, 2, 4]. Thus, ecological resilience (hereafter, resilience) is a useful concept for identifying the range of external conditions under which a system’s behavior may vary, but its state does not fundamentally change.

As Amazon vegetation faces increasing ecological stress from climate change [5, 6, 7, 8, 9] and human development [10, 11, 12, 13, 7, 8, 9, 12, 13], the resilience of Amazon vegetation to external forcing becomes a central question surrounding the future of Amazon ecosystems. Amazon resilience can be understood from a wide range of perspectives depending on how the state of the system and its external forcings are defined. For example, some studies explore the resilience of biomass to future climate change, using biomass as the metric defining the state of the system and climate change to define the forcing [14, 15]. Other studies isolate the effect of water availability by examining the resilience of tree cover (the state of the system) to precipitation or drying (the forcing) [16, 17, 18, 19, 20].

Building on this extensive modern work, this study aims to characterize the resilience of central and western Amazon tree cover to drier, mid-Holocene (~ 6 ka) conditions [21, 22, 23, 24, 25, 26, 27, 28], and compare mid-Holocene tree cover resilience to the present. We stress that this approach cannot account for other dimensions of rainforest resilience, like species composition or diversity. Instead, we focus specifically on the resilience of tree cover to drying for three reasons. First, tree cover and precipitation can be reasonably inferred from paleo proxy data [21, 29, 30, 31, 32, 22]. Second, tree cover is a useful metric for comparing two alternative, well-defined states of Amazon vegetation—a high-tree cover forest and a low-tree cover savanna/grassland [18, 33]. Third, some modern data indicate that annual precipitation rates are near a “tipping-point” of 2,000 mm/yr, below which vegetation water-stress is enhanced along with feedbacks that may favor a forest-to-grassland

32 transition [34, 35, 36]. If precipitation is near this tipping-point today, it is
33 possible that the tipping point was crossed during drier times in the past.

34 The 2,000 mm/yr tipping point is ecologically meaningful because, across
35 the tropics, it approximates a transition between a dry-season water surplus
36 ($> 2,000$ mm/yr) and a dry-season water deficit ($< 2,000$ mm/yr) [36]. Be-
37 low this threshold forests take more time to recover from a drought [37],
38 dry-season photosynthesis declines, favoring deciduous and semi-deciduous
39 vegetation [36, 38], and the wet season may grow shorter with a delayed on-
40 set, amplifying drying [39, 40, 41, 42, 43]. Further, both natural and human
41 fires may increase under a dry-season water deficit. Fire correlates positively
42 with the dry-season water deficit and negatively with relative humidity in
43 Amazonia [44, 45] and seasonal aridity makes human-driven fire and defor-
44 estation easier in the dry season [45, 46, 47]. A seasonal water deficit and
45 fire is often thought to favor grasslands in competition against forests be-
46 cause grasses generally recover faster from droughts and fires and they tend
47 to be more flammable, inhibiting some tree growth [48, 49, 50, 51, 37, 45].
48 But could future aridity alone drive widespread forest dieback and grassland
49 expansion in Amazonia?

50 Some models suggest that future climate change can devastate large
51 swaths of Amazon tree cover [52, 53, 54, 5], but recent work has brought
52 models and observations into agreement that Amazon tree cover may remain
53 high despite future drying [35, 19, 20, 13]. Still, models and observations have
54 their own limitations when unraveling the relationship between precipitation
55 and tree cover. Models can simulate the effect of aridity in the absence of hu-
56 man deforestation [9] but the tree cover response to drying is highly sensitive
57 to parameterizations [54, 55, 56, 35]. Observational datasets can be filtered
58 to remove deforested sites [17, 19], but this can introduce biases associated
59 with the spatial pattern of deforestation and does not account for non-local
60 consequences of deforestation [57, 35, 58]. To circumvent these challenges,
61 paleo-archives from times when human land use was less intensive are use-
62 ful for exploring the relationship between precipitation and tree cover while
63 limiting confounding factors.

64 Here, we test the hypothesis that Amazon tree cover is resilient to drying—
65 specifically below the 2,000 mm/yr threshold—using the mid-Holocene as a
66 case study. Existing data suggest that the central and western Amazon
67 Rainforest remained largely intact with high tree cover in the mid-Holocene
68 [59, 29, 21, 22]. However, the vast majority of precipitation proxy records
69 have been interpreted qualitatively, so it is unclear whether high tree cover

70 in the mid-Holocene is owed to high resilience to drying, or whether mid-
71 Holocene proxies record only modest decreases in precipitation. We present
72 a quantitative estimate for mid-Holocene precipitation based on speleothem
73 oxygen isotope gradients and we compile pollen and charcoal proxy data to
74 assess the relationship between precipitation, fire, and tree cover. We then
75 test whether this relationship is supported in a dynamic global vegetation
76 model (LPJ-GUESS v4.0 [60]) and we compare our results to modern pre-
77 cipitation and tree cover data. In doing so, we address two questions: 1)
78 How resilient was mid-Holocene tree cover to drying? And 2) Is Amazon
79 tree cover similarly resilient today?

80 **2. A conceptual model for the tree cover response to external forc-** 81 **ing and internal feedbacks**

82 *2.1. Conceptual model framework*

83 Following previous work, we consider the state of the system in three
84 groups based on the percent of area covered by trees: 1) a high tree cover
85 state (>70%) to represent forests, 2) a low tree cover state (5-20%) to repre-
86 sent savannas or grasslands and 3) a transitional state at intermediate values
87 [17, 18, 33]. We assess how the state of the system (forest, grass/savanna,
88 or transitional) changes as a function of three forcings (precipitation, de-
89 forestation, and anthropogenic fire) and two positive feedbacks that arise
90 from their interactions within the system. The Moisture Recycling Feed-
91 back (MRF) connects precipitation, tree cover, and evapotranspiration and
92 illustrates how drying can decrease tree cover which, in turn, decreases evap-
93 otranspiration leading to further drying [58, 57, 35]. The second feedback
94 includes the MRF loop and adds the effect of fire (MRF + Fire). In this
95 case, a drop in precipitation increases natural fire which decreases tree cover
96 and evapotranspiration, exacerbating the precipitation decline. The system
97 is illustrated in Figure 1.

98 *2.2. Tree cover response scenarios*

99 The response of tree cover to precipitation depends on the relative con-
100 tributions of forcings and feedbacks. Deforestation and anthropogenic fire,
101 independent of any feedbacks, decrease tree cover without modifying pre-
102 cipitation (Fig. 2A, pink arrow). Changes in precipitation, independent of
103 feedbacks, change tree cover when water limits primary productivity, follow-
104 ing the “maximum potential tree cover” curve in Figure 2A.

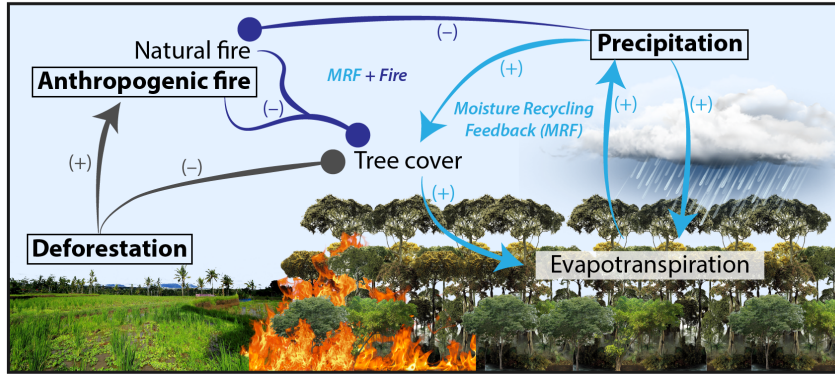


Figure 1: Schematic of forcings (boxes with bold font) and feedbacks (light and dark blue arrows) in our conceptual model. Tree cover defines the state of the model system. Arrows start at a cause and end at an effect with arrowheads denoting a positive relationship and closed circles denoting a negative relationship. The moisture recycling feedback (MRF) loop is shown in light blue arrows, and the contribution of fire (MRF + Fire) in dark blue. Deforestation and anthropogenic fire directly decrease tree cover (no feedback) and can indirectly decrease tree cover through the MRF and MRF + Fire feedbacks. A feedback in this diagram is positive if the sum of closed circles within a given loop is an odd number and negative if the sum is even.

105 Feedbacks in the system are considered stronger when changes in tree
 106 cover are larger for a given change in precipitation (proportional to the area of
 107 wedges in Fig. 2A). Thus, the feedback strength can change if the functional
 108 form of the precipitation-tree cover relationship changes (moving from one
 109 solid line to another in Fig. 2A) or if the initial precipitation rate changes
 110 (moving along a solid line in Fig. 2A). The MRF increases the distance the
 111 system moves along a given precipitation-tree cover curve for a given forcing.
 112 In this case, precipitation is still the only mechanism limiting tree cover in
 113 this feedback loop, so the sensitivity of tree cover to precipitation (the slope of
 114 the curve) remains a single function of precipitation. By contrast, increasing
 115 the sensitivity of fire to precipitation presents a new limit on tree cover and
 116 can push the state of the system to a new curve in precipitation-tree cover
 117 space (Fig. 2A, orange arrow). In our conceptual model the MRF + Fire
 118 loop is the only mechanism for abruptly “tipping” the state of the system
 119 (tree cover) without a proportional change in the forcing (i.e. by moving from
 120 one curve to another). The MRF without fire can lead to an abrupt change
 121 in tree cover but the forcing—precipitation—will change proportionately so
 122 it does not constitute tipping behavior.

123 The implication of our conceptual framework is that multiple stable states—
 124 the condition where more than one ecological regime is stable for a given
 125 precipitation rate—can only be driven by fire. This is supported by fire ex-
 126 clusion experiments around the world where closed forests replace grasslands
 127 without a large change in precipitation [61, 62, 63, 64, 65, 66] and by recent
 128 modeling work showing multiple stable states are absent from a dynamic
 129 global vegetation model unless fire-vegetation feedbacks are included [55].

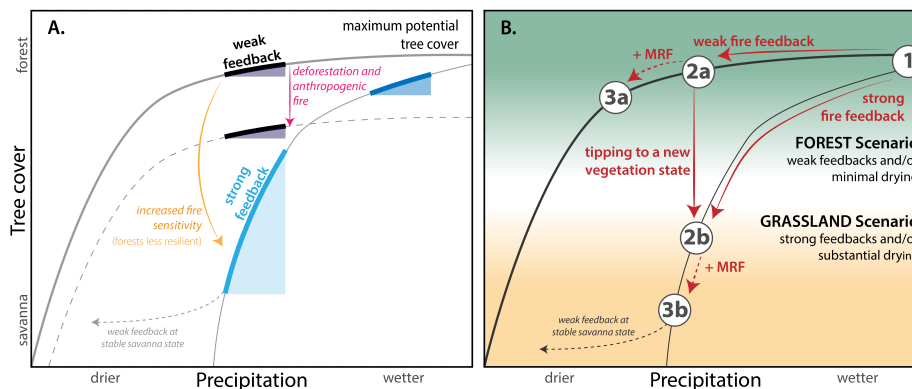


Figure 2: Multiple possible tree cover responses to precipitation based on the relevant forcings and feedbacks. (A) The strength of positive feedbacks is proportional to the area of wedges, and can vary by moving along a curve or switching between curves. (B) Different feedbacks have different effects on how tree cover changes with precipitation. Moving from point 1 to 2a or 2b shows how the fire feedback changes the slope of the tree cover-precipitation relationship (stronger feedback leads to a steeper slope). The MRF increases the distance the system moves along a curve. With the MRF the system can move from point 1 to 3a (3b) with the same forcing otherwise required to move to 2a (2b). If the positive fire feedback strengthens at some precipitation threshold the system can abruptly “tip” from point 2a to 2b. The transition from a forest to grassland thus requires strong, positive feedbacks and/or substantial drying.

130 Figure 2B shows how tree cover could change with time (from points 1
 131 to 3) under different feedback scenarios. Beginning at point 1 with high pre-
 132 cipitation and high tree cover, drying pushes the state of the system along a
 133 defined precipitation-tree cover trajectory from points 1 to 2. If the positive
 134 fire feedback is weak, tree cover will remain near the “maximum potential
 135 tree cover” line, ending at point 2a. If it is strong, the system moves along
 136 a steeper curve to point 2b. In contrast, the MRF increases the magnitude
 137 of change along a precipitation-tree cover curve. Thus, with a strong MRF
 138 the system will move from points 1 to 3 in the same time it would otherwise

139 take to go from points 1 to 2 on its given trajectory (a or b). However,
140 feedback strength is often also a function of precipitation. For example, the
141 fire feedback is likely weak/strong when precipitation rates (and humidity)
142 are high/low, respectively [61, 67, 44]. If the fire feedback abruptly strength-
143 ens at some precipitation threshold, it can “tip” the system from a forest
144 to grassland state. This is depicted in the path from points 2a to 2b where
145 strengthening of the fire feedback at point 2a causes the system to “tip” to
146 point 2b (Fig. 2B).

147 Using this conceptual model with proxy data for fire, vegetation, and
148 precipitation we can infer the strength of the different forcings and feedbacks
149 by reconstructing the history of tree cover and precipitation. To illustrate
150 this, consider a decrease in precipitation from point 1 to points 2a and 2b
151 (Fig. 2B). In order to maintain a forest despite drying, positive feedbacks
152 and anthropogenic forcing must be weak. But a transition to a grassland
153 state indicates that anthropogenic forcing is strong, positive feedbacks are
154 strong, and/or positive feedbacks strengthened within the range of drying
155 causing the system to “tip”. Proxy records can then be used to determine
156 which of these options is most likely by indicating, for example, how sensitive
157 fire is to drying.

158 **3. The Amazon Basin in the mid-Holocene**

159 *3.1. Mid-Holocene proxy records*

160 The mid-Holocene (~ 6 ka) is an ideal period to evaluate the Amazon tree
161 cover response to aridification because it is associated with widespread proxy
162 data indicating one of the driest times of the late Quaternary [21, 68, 25, 69,
163 70]. While the onset of drier conditions varies from the early to mid-Holocene
164 depending on the location, drier conditions generally lasted through, and
165 often peaked at, 6 ka [24, 71, 72, 73, 70, 22]. Drying is especially pronounced
166 in the central, western, and southwestern Amazon and drier conditions are
167 also observed on a continental scale in offshore records of terrestrial runoff
168 [26, 27, 28, 74, 75] (Fig. 3A). This coincides with the desiccation of lakes and
169 fluvial systems in the western Andean plateau [24, 23, 76, 25, 77] and local-to-
170 regional increases in savanna coverage and fire [29, 21] (Fig. 3A), particularly
171 in the southwestern Amazon. While patches of savanna expanded within the
172 forested regions of southwestern Amazon at this time [78, 69], the central
173 and western Amazon domain, or core rainforest region (Fig. 3B, inset),
174 likely remained intact [29, 21, 59, 22].

175 In addition to climate, humans may have also played a role in modifying
176 Amazon vegetation in the mid-Holocene. Human settlements were extensive
177 in the mid-Holocene, spanning much of southwestern Amazonia with evi-
178 dence for cultivation and food production [79, 80, 81, 82, 83]. Some evidence
179 suggests human populations declined in the mid-Holocene [84], but this sig-
180 nal may be related to geomorphic or sampling biases [85]. Despite the broad
181 spatial coverage of human settlement in the mid-Holocene, land use and
182 human-driven vegetation change was more localized and less intensive than
183 today. Many human settlements took advantage of naturally open vegetation
184 rather than creating space through widespread clear-cutting [81, 85]. Fur-
185 ther, at least in the late Holocene the climate-driven expansion of rainforest
186 around human settlements provides evidence for sustainable land manage-
187 ment practices that did not interfere with regional-scale climate-vegetation
188 relationships [86, 81]. Thus, while human occupation was extensively dis-
189 tributed across some parts of the basin during the mid-Holocene, climate
190 most likely remained the primary control on regional tree cover with human
191 land use occurring on a smaller scale than today.

192 Our compiled proxy data includes charcoal and pollen records, along with
193 speleothem oxygen isotope data from sites spanning the central and western
194 Amazon (Fig. 3B). We also compare to a variety of proxy types used to
195 reconstruct mid-Holocene vegetation in Smith and Mayle, 2018 [21]. We
196 align our proxy data with the isotope gradient in order to directly compare
197 precipitation, tree cover, and fire, but most available pollen and charcoal
198 records are in the western Amazon and we lack uniform coverage across the
199 central basin. Here, the assorted proxy records of Smith and Mayle [21]
200 help compensate for a paucity of pollen and charcoal data. Some records to
201 the north approach regions where previous studies have interpreted wet mid-
202 Holocene conditions (3A), and this may also bias our results. However, we
203 repeat our analysis with southwestern Amazon pollen and charcoal records,
204 where proxy data reveal more uniform drying, and find that this sampling
205 bias does not affect our conclusions (Supplemental Fig. S1).

206 *3.2. Mid-Holocene climate*

207 Amazon precipitation in the late Quaternary is thought to follow a zonal
208 dipole pattern, with anomalous precipitation shifts between the eastern and
209 western/southwestern basin on precessional and glacial-interglacial timescales
210 [31, 87, 32, 88, 89, 90]. During the mid-Holocene, data suggest the dipole
211 was likely in its “eastern phase” with positive anomalous precipitation over

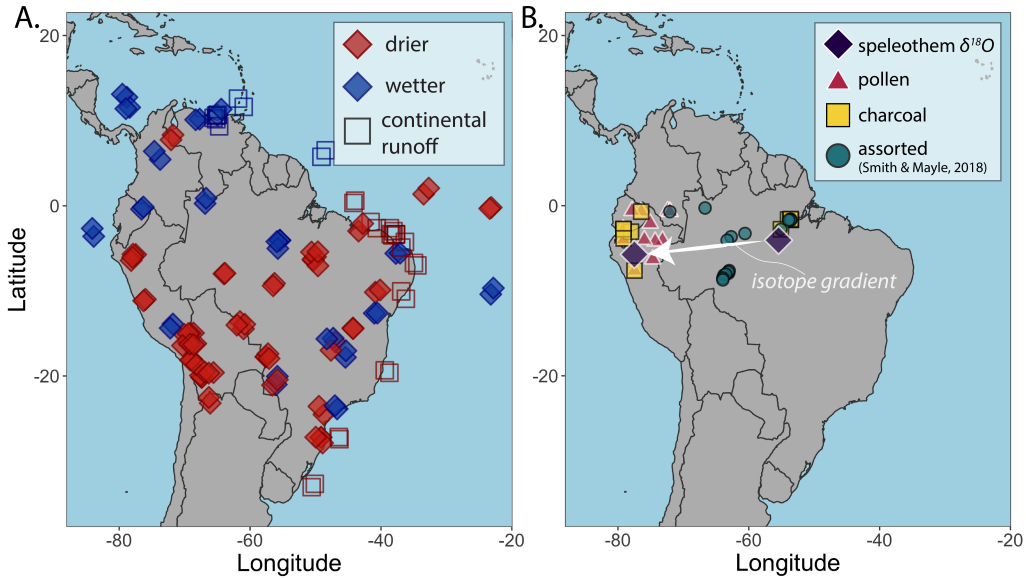


Figure 3: **(A)** Proxy compilation for the mid-Holocene (defined here as 5-7 ka). See Supplementary data for full reference list. Blue points represent a proxy indicating wet or wetter conditions and red points are drier. Diamonds are local/regional proxies and open squares are marine core records of continental runoff. **(B)** Map of the pollen (magenta triangles), charcoal (yellow squares) and speleothem $\delta^{18}O$ (purple diamonds) records used in this study. Teal circles show proxy records of assorted type from Smith and Mayle, 2018 [21] that fall within our domain.

212 eastern Amazonia and NE Brazil [31, 87]. The mechanism for precipitation
 213 dipole transitions remains debated, but it is usually linked to the strength
 214 of the South American Monsoon with the eastern phase occurring during a
 215 weaker monsoon [31, 32, 91, 92].

216 In addition to dynamic changes in the spatial pattern of moisture con-
 217 vergence, Amazon precipitation is also sensitive to how moisture is recycled
 218 across the basin. Today and in the mid-Holocene the central and western
 219 Amazon sits inland from the primary moisture source—the easterly trade
 220 winds that cross the Atlantic Ocean [93, 94]—so precipitation is strongly de-
 221 pendent on upstream moisture recycling. About one-third to half of all pre-
 222 cipitation in this region is derived from evapotranspiration within the basin
 223 today [58, 57, 95, 96], making it particularly sensitive to feedbacks like the
 224 MRF that can promote forest-grassland transitions [42, 97, 58, 57, 35, 98, 47].
 225 If feedbacks like the MRF can promote forest-grassland transitions, the west-

226 ern Amazon is an ideal place to detect this signal.

227 4. Methods

228 4.1. Oxygen isotope gradient and precipitation reconstruction

229 4.1.1. Isotope gradient reconstruction and the importance of the “hydrostat”

230 Oxygen isotopes in calcite cave deposits, or speleothems, can reflect the
231 isotopic composition of past rainfall and provide useful information for re-
232 constructing monsoon dynamics [99, 31, 30, 100, 101, 32, 102]. We use the
233 change in Amazon speleothem $\delta^{18}O$ over space—the isotope gradient ($\Delta\delta^{18}O$;
234 ‰/1,000km)—with a reactive transport model [103, 104, 105, 106] to solve
235 for precipitation rates. This is done by reconstructing past isotope gradients
236 along the South American Monsoon moisture trajectory with $\delta^{18}O$ records
237 from three sites—one in east-central Amazon [30] and two that form a com-
238 posite record in the western Amazon [32, 102] (Fig. 4). We also adopt the
239 $\delta^{18}O$ correction that accounts for differential cave temperatures between the
240 sites previously used in references [30, 107].

241 The isotope gradient is useful because it reflects changes in the regional
242 water balance while limiting the influence of confounding factors. The bal-
243 ance of precipitation (P) and evapotranspiration (ET) across a moisture
244 trajectory sets the isotope gradient because P tends to decrease $\delta^{18}O$ by re-
245 moving moisture from an air mass while ET increases $\delta^{18}O$ by replenishing
246 the air mass [108]. Thus, $\Delta\delta^{18}O$ decreases as P increases relative to ET .
247 Additional effects unrelated to the water balance, like upstream effects and
248 temperature, can influence $\delta^{18}O$ at a single site, but isotope gradients are
249 insensitive to these factors because they do not modify the rainout-recycling
250 balance [103, 104, 109]. As a result, the isotope gradient across Amazonia
251 has long been used as a metric for the balance of P and ET [108].

252 Importantly, the oxygen isotope gradient (along with $\delta^{18}O$ at a given
253 point) is not always sensitive to precipitation rates [106, 110, 103]. As P
254 decreases, the isotope gradient shallows until it asymptotes at the theoret-
255 ical maximum value of zero (meaning there is no change in $\delta^{18}O$ between
256 the upstream and downstream sites). The point where the isotope gradient
257 reaches zero is called the “hydrostat”—the condition where further aridifi-
258 cation will have no affect on $\Delta\delta^{18}O$ values [106, 110, 103]. The hydrostat is
259 reached because the net distillation of moisture is near zero as climatologi-
260 cal rates of P and ET are nearly equal. This can occur when P rates are
261 similar to or below potential evapotranspiration (E_0) rates. Since E_0 is an

262 upper limit on ET , P must exceed ET when $P > E_0$, leading to a decrease
263 in $\delta^{18}O$ and a negative P - $\delta^{18}O$ relationship known as the “amount effect”
264 [111, 103, 112, 113, 114, 115]. Therefore, when $\Delta\delta^{18}O$ is near zero it is possible
265 to constrain the upper bound of precipitation rates—the point where P
266 exceeds potential ET —but not the lower bound.

267 4.1.2. Reconstructing past mean annual rainfall

268 We use the reactive transport model (RTM) [103], which simulates atmo-
269 spheric vapor transport (w), rainout (P), and evapotranspiration (ET), to
270 inversely solve for mid-Holocene rainfall rates from speleothem $\Delta\delta^{18}O$ data.
271 We initialize the model with distributions of the source moisture content
272 (similar to specific humidity), potential evapotranspiration, surface temper-
273 ature, the wind speed profile, evapotranspiration partitioning, a moisture
274 recycling efficiency parameter, and a vapor condensation rate constant. Sam-
275 pling from these input distributions $> 50,000$ times, the model solves for the
276 spatial pattern of precipitation and evapotranspiration as well as the oxygen
277 isotope ratios of these fluxes [103]. See Supplemental Text S1 for further
278 information on model initialization.

279 Using the results of our Monte Carlo routine we calculate the mean rain-
280 fall consistent with the observed $\Delta\delta^{18}O$ to produce a mean annual rainfall
281 distribution. During the mid-Holocene, $\Delta\delta^{18}O$ values fall along the hydrostat
282 and the RTM loses sensitivity to the lower bound of rainfall (while the upper
283 bound is still constrained). In order to extract a rainfall distribution in this
284 case we impose a lower bound, thus assuming that any rainfall value below
285 the imposed threshold is unreasonable. To develop a conservative estimate
286 of mid-Holocene aridification we exclude any iteration where mean annual
287 rainfall drops below half the potential evapotranspiration rate, a point at
288 which modern tree cover is mostly below 50% (*e.g.* [36, 19]). We test the
289 sensitivity of our results to this assumption with a range of thresholds (Sup-
290 plemental Fig. S2) and find that this decision influences the mean rainfall
291 value but does not affect our conclusions.

292 Our analysis implicitly assumes that moisture is transported directly from
293 the eastern (upstream) speleothem site to the western sites. While this tra-
294 jectory aligns well with the direction of the easterly trade winds (especially
295 during the wet season) it is reasonable to expect that the effective moisture
296 transport distance between these sites may have differed in the mid-Holocene
297 due to changes in atmospheric circulation patterns. However, because $\Delta\delta^{18}O$
298 is near zero at the mid-Holocene the isotope gradient (and, thus, our results)

299 are not sensitive to changes in assumptions about the effective moisture trans-
300 port distance (Supplemental Fig. S3).

301 4.2. Pollen compilation

302 Twenty-three terrestrial pollen records were extracted from the Neotoma
303 Paleocology Database [116] and the ACER pollen and charcoal database
304 [117] in March, 2020 across our study area (with data spanning 55° - 80°W
305 and 8°S - 0°N; see Fig. 3B). Pollen percentages for all records were calculated
306 based on the sum of terrestrial pollen taxa only (the pollen taxa classified as
307 ‘Tree/Shrub’, ‘Succulent’, ‘Upland herbs’ and ‘Palms’ in Neotoma). Pollen
308 samples were subsequently grouped into 200-year bins, which corresponds
309 with the average resolution of the selected records (197 years between con-
310 secutive samples, on average), to extract a regional signal. The average bin
311 size is 22.4 pollen samples with a range of 2-86 samples. For each bin, we
312 calculate the ratio of arboreal (Tree and Shrubs, Palms) to non-arboreal
313 (Herbs and Succulents) pollen taxa. While the classification of pollen taxa
314 into unique vegetation forms contains some inaccuracies (*i.e.* all the species
315 producing a pollen morphotype may not all belong to the same vegetation
316 form), the associated uncertainties are expected to only have a negligible
317 impact and/or compensate each other in our broad-scale, multi-site recon-
318 struction.

319 4.3. Charcoal compilation

320 Charcoal records are compiled from the Global Charcoal Database (GCD
321 version 2.0) and other published datasets. Charcoal data were analyzed
322 using the paleofire R package software (version 1.1.8) [118]. Eleven charcoal
323 records spanning 53° - 80°W and 8°S - 0°N are included in this analysis to
324 create a regional charcoal curve (Supplemental Data, Fig. 4C). These sites
325 provide an average of regional biomass burning during the Holocene. To
326 facilitate inter-site comparison, the eleven records are pre-treated using a
327 standard protocol [119, 120] for transforming and standardizing individual
328 records that includes: (1) transforming non-influx data (*e.g.* concentration
329 particles cm^{-3}) to influx values (particle cm^{-2}/yr), (2) homogenizing the
330 variance using the Box-Cox transformation, (3) rescaling the values using a
331 minimax transformation to allow comparisons among sites, and (4) rescaling
332 the values to z-scores using a base period of 200 years. Sites are smoothed
333 with a 500-year half width smoothing window and a bootstrap of 200 years
334 [118].

335 *4.4. Dynamic global vegetation modeling*

336 To evaluate the mid-Holocene precipitation-tree cover relationship in-
337 ferred from proxies while accounting for changes in other climate variables
338 we conducted simulations with the second generation dynamic global veg-
339 etation model (DGVM) LPJ-GUESS [60, 121]. The simulated vegetation
340 states emerge as an outcome of simulated vegetation structure, demography,
341 resource competition for light, water and nutrients, and wild fires. Vege-
342 tation is represented as a mixture of plant functional types (PFTs) (11 in
343 this study) [122], distinguished by photosynthetic pathway (C3 or C4), life
344 history strategy (shade tolerance), phenology (evergreen, summergreen or
345 raingreen), growth form (trees or herbaceous plants) and bioclimatic distri-
346 butional limits.

347 LPJ-GUESS incorporates a fire model to link the fire regime and its effects
348 on vegetation dynamics and biogeochemical cycling [123]. Fires are modelled
349 prognostically based on temperature, current fuel load, and moisture. Daily
350 litter moisture is used to estimate the fire season length which, in turn,
351 determines the fraction of a grid cell that is burnt in a year. The relationships
352 between litter moisture, fire season length, and burnt area are calibrated with
353 modern data [123]. Fire return intervals are simulated by the model based on
354 the yearly burnt area fraction. When driven by 20th century climate data, the
355 simulated fire return intervals are in good agreement with observations in the
356 Amazon and globally [123]. Since observational data for model calibration
357 are assumed to be the most representative for natural conditions, human-
358 changed fire regimes and other land use impacts are not explicitly considered,
359 but could still have an effect on the final model processes [123].

360 Climate forcing (temperature, precipitation and radiation) of the offline
361 LPJ-GUESS simulations comes from output of the coupled global climate
362 model, EC-Earth [124]. The mid-Holocene orbital parameters and green-
363 house gas concentration were prescribed for the standard mid-Holocene EC-
364 Earth simulation (MH), while boundary conditions such as a vegetated Sa-
365 hara and reduced dust aerosol (MHgsrd) were further added to a sensitivity
366 simulation to reproduce a climate regime more consistent with proxy re-
367 constructions [125]. The vegetation response in North Africa in these sim-
368 ulations was investigated in detail in a previous study [126]. In total, we
369 performed six simulations. Three simulations—one pre-industrial (PI) and
370 two mid-Holocene (MH, MHgsrd)—were forced with unperturbed climate
371 forcing from EC-Earth and three simulations included further precipitation
372 reductions. Amazon annual precipitation in the unperturbed simulations is

373 2,175, 2,002, and 1,776 mm/year for PI, MH, and MHgsrd. The precipi-
374 tation reduction experiments use MHgsrd boundary conditions with annual
375 precipitation reduced to 1,520, 1,420 and 1,260 mm/year. We compare our
376 precipitation reduction experiments to MHgsrd in Fig. 6, but all results can
377 be found in Supplemental Fig. S4 and S5.

378 Precipitation reductions were applied with a multiplier to scale average
379 precipitation in the proxy region of Fig. 3B to the desired level. All grid
380 cells (inside or outside the proxy domain) were multiplied by the same factor.
381 Therefore, we restrict our analysis to the vegetation patterns within the proxy
382 domain to avoid regions where precipitation scaling is unrealistic and outside
383 the domain of $\Delta\delta^{18}O$ constraints. We also exclude the Andes to avoid a
384 confounding, Andean taxa signal. To understand the role of fire disturbance,
385 the 1,520 and 1,260 mm/year simulations were repeated with the fire module
386 disabled (purple squares in Fig. 7). All LPJ-GUESS simulations were spun-
387 up from bareground and run for 500-years to reach an equilibrium state. We
388 use the last 10 years of output for analysis. Further information about our
389 LPJ-GUESS simulations can be found in the Supplementary Text (S2).

390 *4.5. Modern tree cover and precipitation data*

391 We compile modern data for tree cover, precipitation, and human land
392 use in order to compare past and DGVM-simulated estimates of tree cover
393 and precipitation to modern conditions. Tree cover, precipitation, and land
394 use data were compiled in ref [19]. Tree cover was collected from Landsat
395 remote sensing data at 30m spatial resolution [127] with trees defined as
396 being greater than 5m tall. Mean annual precipitation rates were calculated
397 from the Global Precipitation Climatology Centre (GPCCv7) [128] using the
398 correction of ref [129] for the years 1993-2012. Land use data are derived
399 from ref [130].

400 **5. Results**

401 *5.1. Oxygen isotope gradient reconstruction*

402 The kernel smoothed oxygen isotope gradient values range from -1.3
403 to 0.1 ‰/1,000 km through the Holocene. The record plateaus and hovers
404 around the hydrostat, the theoretical maximum isotope gradient [103, 106],
405 from $\sim 11 - 5$ ka, before decreasing toward its lowest values in the present
406 (Fig. 4A). The oxygen isotope gradient in modern precipitation across the

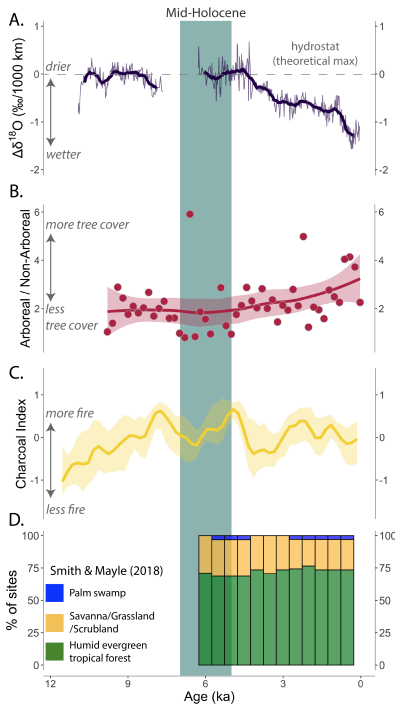


Figure 4: Proxy timeseries data. (A) The oxygen isotope gradient between the two speleothem sites. Darker curve is a kernel smooth through the data. (B) The ratio of arboreal to non-arboreal pollen. Line is a kernel smooth through the data. (C) The charcoal index timeseries (yellow line) with 95% confidence intervals shown by the translucent, yellow ribbon. (D) Percent of sites recording different vegetation types (from compilation of Smith and Mayle, 2018 [21]).

407 Amazon is $\sim -1\text{‰}/1,000\text{ km}$ [108, 103], similar to the gradient from the
 408 most recent speleothem $\delta^{18}\text{O}$ data.

409 While $\Delta\delta^{18}\text{O}$ values are stable near zero for much of the early-to-middle
 410 Holocene, independent proxy evidence from western Amazonia indicates pre-
 411 cipitation was decreasing during this time [24, 23, 68, 131, 77, 76, 132]. Thus,
 412 the $\Delta\delta^{18}\text{O}$ record may not fully reflect the trends in precipitation through the
 413 Holocene due to $\Delta\delta^{18}\text{O}$ being insensitive to precipitation at the hydrostat.

414 5.2. Vegetation in the mid-Holocene

415 In western Amazonia the arboreal to non-arboreal pollen ratio ranges
 416 from 0.8 to 5.9 throughout the Holocene. Kernel smoothing of the data
 417 shows gradual increase in this ratio from $\sim 5\text{ ka}$ to present, mirroring a shift

418 to lower $\Delta\delta^{18}O$ values. There is a weak, negative correlation between the
419 ratio of arboreal to non-arboreal pollen and the oxygen isotope gradient (Sup-
420 plemental Fig. S6). Some of the lowest arboreal/non-arboreal pollen ratios
421 occur in the mid-Holocene, suggesting the possibility of some savanna expan-
422 sion. We note that the arboreal/non-arboreal pollen ratio is not necessarily
423 a reliable indicator of tree cover [133] and therefore we use it only to infer
424 relative trends. Data from Smith and Mayle, 2018 [21] show that, from 6
425 ka to present, between 69 to 76% of sites in our proxy domain record humid
426 evergreen tropical forest conditions, consistent with the rainforest remaining
427 intact [29, 21, 59] (Fig. 4D).

428 5.3. *Fire in the mid-Holocene*

429 Our composite charcoal record shows a progressive increase of the char-
430 coal index from -1 to 0.6 between 12 ka and 10 ka (Fig. 4C). Fire activity
431 declines to a local minimum with a charcoal index of 0 around 6 ka, fol-
432 lowed by a sharp increase to 0.7 around 5 ka. Another local minimum occurs
433 around 3.5 ka, after which charcoal index values remain just above zero until
434 present (Fig. 4C). The charcoal index is not significantly correlated with
435 $\Delta\delta^{18}O$ nor the arboreal/non-arboreal pollen ratio (Supplemental Fig. S6).
436 Despite the lack of correlation, fire may still be responsive to vegetation
437 if it tracks changes in species composition, which may not vary with the
438 arboreal/non-arboreal pollen ratio. Further, in contrast to our charcoal in-
439 dex curve, tropical glacier proxy records from the southwest, outside of our
440 domain, indicate increased fire activity during the mid-Holocene [134, 135].
441 However, it is unclear whether this discrepancy with our composite record is
442 owed to greater aridity in the southwest, a difference in vegetation composi-
443 tion, or something else.

444 5.4. *RTM modeling and rainfall reconstruction*

445 Our RTM results for the pre-industrial show modeled rainfall rates that
446 are consistent with distributions from GPCP data [128] (Fig. 5). Inverting
447 on the isotope gradient from the most recent speleothem $\delta^{18}O$ values yields
448 a domain-mean rainfall estimate of $2,300 \pm 500$ mm/yr, comparable to
449 modern ($2,480 \pm 470$ mm/yr; Fig. 5). This builds on previous work indi-
450 cating that spatial isotope gradients record information about climatological
451 moisture fluxes [108, 109, 103, 104, 106].

452 Our mid-Holocene results indicate substantially drier conditions com-
453 pared to the pre-industrial. In the mid-Holocene we estimate a rainfall rate

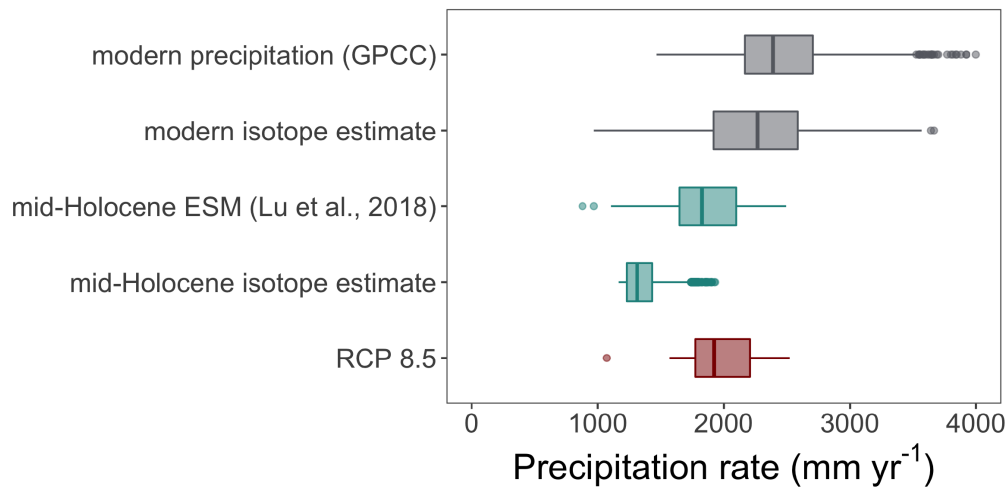


Figure 5: Precipitation distributions (from top to bottom) of: modern precipitation from the proxy domain from GPCCv7 [128], pre-Industrial precipitation estimate from the RTM inversion of the most recent speleothem $\Delta\delta^{18}O$ data, mid-Holocene precipitation in the proxy domain from the Green-Sahara simulations of ref [126], the wettest quarter of precipitation estimates from the RTM inversion on the mid-Holocene hydrostat $\Delta\delta^{18}O$ values, and estimates for future Amazon precipitation in RCP 8.5 from ref [19].

454 of $\sim 1,300$ mm/yr (Fig. 5), about 45% lower than present. This estimate is
 455 based on the wettest 25% of RTM iterations above the imposed lower-bound.
 456 We find no possible solutions exceeding 2,000 mm/yr and the maximum sim-
 457 ulated precipitation rate is $\sim 1,900$ mm/yr. Despite the loss of model leverage
 458 at low precipitation rates, the exercise provides a valuable upper-bound on
 459 mean annual precipitation.

460 Our results demonstrate that mid-Holocene precipitation rates are likely
 461 substantially lower than most end-of-century precipitation estimates from
 462 simulations of unabated carbon emissions (RCP 8.5; Fig. 5). On average,
 463 models predict precipitation by the year 2100 will be ~ 400 mm/yr less than
 464 today with over half the models predicting future rainfall below the 2,000
 465 mm/yr threshold. However, when accounting for model precipitation bias
 466 (usually a dry-bias in Amazonia) the mean change in future precipitation
 467 is much smaller— ~ 26 mm/yr—with no models falling below 2,000 mm/yr.
 468 Since our mid-Holocene precipitation distribution does not exceed $\sim 1,900$
 469 mm/yr, the lowest bias-corrected future estimate is greater than our highest
 470 estimates of mid-Holocene precipitation (Supplemental Fig. S7).

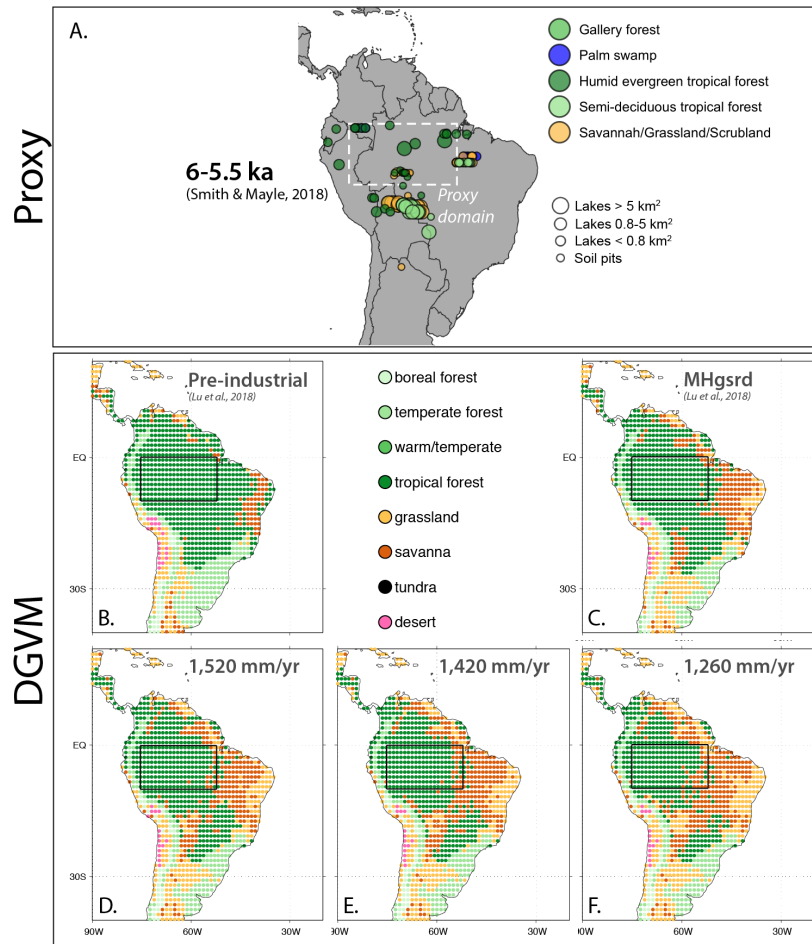


Figure 6: Dynamic global vegetation model results and comparison with comprehensive proxy compilation of ref [21]. (A) Proxy-derived biomes from the 6-5.5 ka timeslice of Smith & Mayle (2018) [21]. Sites with two major biomes are given two, jittered data points. Southwestern proxy region [21] is included for comparison with DGVM. (B-F) Results of DGVM simulations when precipitation is set to the simulations of ref [126] (pre-Industrial and mid-Holocene in B, C) or scaled to 1,520, 1,420, or 1,260 mm/yr within the proxy domain (black boxes) (D-F). Extensive savanna expansion in northeastern Brazil in panels D-F is likely unrealistic because it is outside the proxy domain and does not account for possible wetter conditions in this region in the mid-Holocene [31, 91, 87, 136].

471 5.5. Dynamic global vegetation model response to mid-Holocene rainfall

472 Tree cover in the core rainforest region (central and western Amazonia)
 473 remains intact in our DGVM simulations in a wide range of mean annual

474 precipitation conditions (Fig. 6B-F) [126]. Turning off fire in our simula-
475 tions leads to substantial increases in tree cover outside of the proxy domain
476 (Supplemental Fig. S4) but has little effect within the domain (Fig. 7; Sup-
477 plemental Fig. S8). This suggests that fire is effective at promoting grassland
478 and savanna expansion near the forest fringes (ecotones) in the model, but
479 may not lead to vast tree cover loss in the core forest regions. Most of the
480 savanna and grassland expansion in the southwestern Amazon is inhibited
481 in simulations where fire is turned off, suggesting that drying alone is not
482 sufficient to drive large changes in tree cover within the range of precipitation
483 rates simulated here.

484 Within the proxy domain tree cover remains at forested levels ($> 70\%$)
485 at all precipitation rates whether fire is enabled or not (Fig. 7). Tree cover
486 decreases from $\sim 92\%$ when precipitation is $\sim 2,200$ mm/yr to $\sim 80\%$ when
487 precipitation drops to $\sim 1,300$ mm/yr. Disabling fire leads to an increase in
488 tree cover of $\sim 4 - 5\%$ in the range of precipitation rates we tested. While
489 our precipitation estimates span the range where forests and savanna are
490 thought to reflect alternative stable states (the “bistability range” of 1,300-
491 2,100 mm/yr [17]) we do not observe any marked shift to a savanna vegetation
492 state in our simulations.

493 6. Discussion

494 6.1. Strength of mid-Holocene forcings and feedbacks

495 High tree cover in the mid-Holocene supports the hypothesis that cen-
496 tral and western Amazon tree cover is resilient to crossing below the 2,000
497 mm/yr precipitation threshold, at least down to $\sim 1,300$ mm/yr. Here, we
498 explore why tree cover might be so resilient. We compare the mid-Holocene
499 estimates to the modern precipitation-tree cover relationship to construct an
500 aridification trajectory that defines a tree cover response to drying. Using
501 the proxy records in our conceptual model framework (Section 2), we infer
502 the contributions of forcings and feedbacks from the trajectory. We note that
503 precipitation-tree cover relationships have been theorized to follow a “hys-
504 teresis” pattern where the direction of change (wetting or drying) can deter-
505 mine the tree cover response [16]. This does not invalidate our approach of
506 looking back in time to infer a forward aridification trajectory for two reasons.
507 First, forests prevail in the core of Amazonia in the mid-Holocene through to-
508 day [29, 59, 21] and hysteresis should only arise when the state of the system

509 changes (*i.e.* forest to savanna or grassland). Second tree cover and precipi-
 510 tation were both higher prior to the Holocene [24, 23, 76, 131, 29, 27, 137, 28],
 511 so the modern conditions can also approximate an initial state for Holocene
 512 aridification.

513 Our precipitation constraints and existing vegetation reconstructions [29,
 514 21, 59, 22] are supported by our DGVM simulations showing forest tree cover
 515 is maintained as precipitation reaches mid-Holocene levels (Fig. 7). This
 516 result is consistent with the forest scenario of Figure 2B. In this scenario,
 517 tree cover is near the maximum potential tree cover curve (gray line of Fig.
 518 7) and precipitation is the dominant forcing on the system. High tree cover is
 519 maintained in this scenario as human forcing and the feedbacks that amplify
 520 tree cover loss (like the MRF and MRF + Fire) are not strong enough to
 521 change the state of the system.

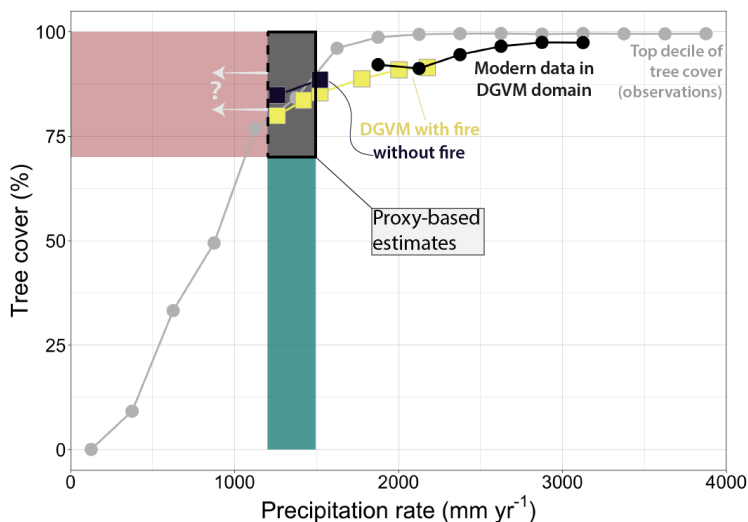


Figure 7: Modeled and proxy-based estimates for fire, precipitation, and tree cover dynamics in the mid-Holocene. Gray circles show the top decile of tree cover observations for each precipitation bin as an estimate of maximum potential tree cover. Black circles show tree cover and precipitation in the proxy domain (see Fig. 6A). Squares show DGVM output with fire (yellow) and without (purple) within the proxy domain. Proxy-based precipitation and tree cover estimates are shown in teal and red rectangles (region of overlap is boxed in black). Overall, observations, models, and proxy data agree that tree cover remains high at $\sim 1,300$ mm/yr of precipitation with no tipping to a grassland state.

522 Indeed, proxy estimates support, at most, a minor role for human forcing
 523 (deforestation) and fire in determining the state of Amazon tree cover in

524 the central and western basin [138, 81, 80, 29, 86, 85]. Our charcoal index
525 composite (which integrates anthropogenic and natural fire) does not show
526 consistent trends with arboreal pollen or precipitation in the Holocene (Fig.
527 4; Supplemental Fig. S6). While this suggests fire is not strongly responsive
528 to climate, we note the possibility that a long-term link between precipitation
529 (Fig. 4A) and natural fire (Fig. 4C) might be overprinted by opposing trends
530 in anthropogenic fire [84, 29, 139]. In either case, these proxy data suggest
531 that human-driven fire forcing, natural fire feedbacks, or both, are too weak
532 to drive grassland expansion in our proxy domain in the mid-Holocene.

533 Grassland and fire expansion, however, may have occurred near forest-
534 savanna boundaries (or ecotones) in the mid-Holocene. Consistent with our
535 DGVM results, most proxy evidence for grassland and fire expansion is re-
536 stricted to ecotones [29, 21] while central and western Amazon records [21,
537 22] and basin-integrated records [59, 138] show little-to-no change through
538 the mid-Holocene. This distinction between ecotonal and basin-integrated
539 records suggests that fire and savanna expansion in ecotonal regions is a
540 small portion of the basin mean, or their expansion is balanced by decreases
541 in fire and grass coverage elsewhere. Fire plays an important role in ecotone
542 migration in our DGVM simulations but this effect occurs mostly outside
543 our proxy domain where our precipitation scaling is not constrained by proxy
544 data (Supplemental Fig. S4). Within our proxy domain, however, the fire
545 feedback in our DGVM simulations is weak and the tree cover response to
546 drying is similar whether the fire module is enabled or disabled (Fig. 7).
547 Overall, DGVM results are consistent with the finding that mid-Holocene
548 fires were limited to forest-grassland ecotones and did not cause the system
549 to tip to a savanna state. The fire feedback may have been strong in eco-
550 tonal regions, but it was weak in the core forested region where high tree
551 cover persisted.

552 *6.2. A seasonal water deficit in the mid-Holocene*

553 The survival of Amazon forests and the passive nature of feedbacks that
554 promote tree cover loss in our analysis is remarkable when we consider the
555 consequences of mid-Holocene water stress. Today, regions of tropical South
556 America that receive less precipitation than 2,000 mm/yr are vulnerable to
557 seasonal water deficits [36]. Below this threshold more water is lost to evap-
558 otranspiration and runoff in the dry season than can be replenished by wet
559 season water storage. The 2,000 mm/yr threshold should also apply in the
560 mid-Holocene because it is set by the balance of potential evapotranspiration

561 and water storage [36] and potential evapotranspiration scales weakly with
562 temperature in the tropics ($\sim 1.5\%/K$) [140]. Based on our precipitation es-
563 timates, it is likely that the central and western Amazon was affected by a
564 seasonal water deficit in the mid-Holocene. Such a deficit would probably
565 have an outsized effect on evergreen taxa that rely on stored water for dry
566 season photosynthesis, while favoring seasonally dry forest taxa and grasses
567 [36, 38, 37].

568 A seasonal water deficit can also amplify fire occurrence and lengthen
569 the dry season in Amazonia. Fires have been shown to increase expo-
570 nentially with drying [44, 45] but the lack of a correlation between $\Delta\delta^{18}O$
571 and fire across the mid-Holocene suggests that dry season aridity does not
572 increase fire sufficiently to drive widespread forest dieback. Meanwhile, un-
573 der a water surplus in the dry season, transpiration of stored water acts as
574 a “moisture pump” that prompts the onset of deep wet season convection
575 2-3 months before the peak shift in wind patterns associated with southward
576 ITCZ migration [43]. If seasonal drying made it difficult for deep-rooted
577 trees to access water and transpire, the onset of the wet season may have
578 been substantially delayed in the mid-Holocene [39, 40, 41, 43, 42]

579 The consequences of seasonal aridity characterize the 2,000 mm/yr thresh-
580 old as a possible “tipping” point where Amazon forests can be abruptly re-
581 placed by grasslands [37, 17, 18]. We cannot confirm whether the Amazon
582 experienced a delayed onset of the wet season in the mid-Holocene, or whether
583 climate-driven fire occurrence increased. But our results suggest that either
584 (1) Amazon tree cover is resilient to the consequences of a seasonal water
585 deficit or (2) the seasonal water deficit did not increase fire nor dry season
586 length enough to tip the central/western Amazon to a grassland/savanna
587 state.

588 *6.3. Modern tree cover resilience compared to the mid-Holocene*

589 Modern spatial data indicate that when mean annual precipitation falls
590 between 2,100 mm/yr and 1,300 mm/yr forests and savannas are both stable
591 ecological states [17, 18, 33]. This is referred to as the region of “bistability”
592 within which a forest ecosystem can abruptly tip to a savanna state. Our
593 mid-Holocene precipitation estimates suggest that even at the lower bound of
594 this bistability region tree cover did not tip to a savanna state. Does forest
595 survival in this range indicate drying alone is unlikely to lead to tipping
596 today? Or is modern Amazon tree cover less resilient to drying than it was
597 in the mid-Holocene?

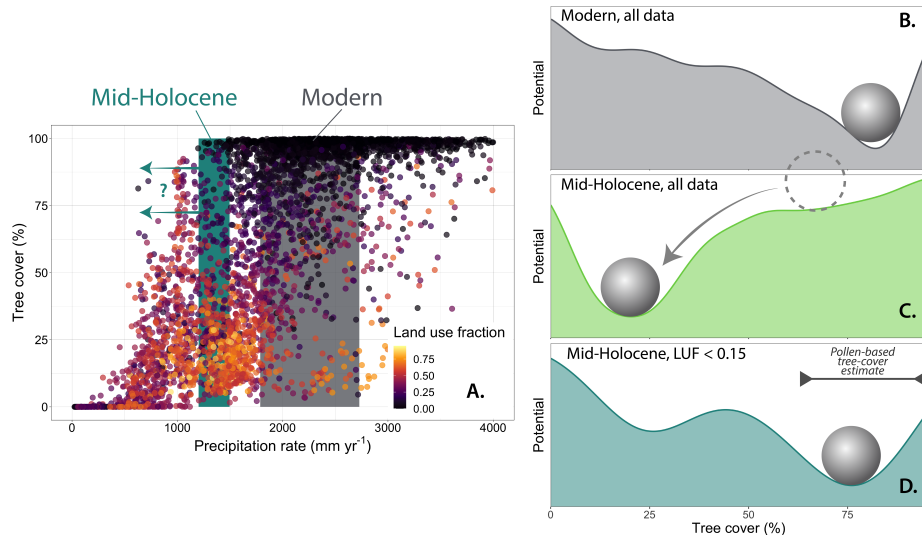


Figure 8: (A) Precipitation and tree cover in tropical South America (data from ref. [19]). Points colored by fraction of land use change in pixel. Isotope-derived modern precipitation rates for central Amazonia denoted by the gray rectangle. Mid-Holocene precipitation estimate shown in teal rectangle. (B) Potential diagram showing tree cover for the range of modern precipitation. Stable state is a high-tree cover forest. (C) Potential diagram for modern tree cover within the range of the mid-Holocene precipitation estimate. Within this range of precipitation, low tree cover savanna is the sole stable state. (D) Potential diagram for modern tree cover with mid-Holocene precipitation when only points with minimal land use fraction ($LUF < 0.15$) are shown, filtering out deforestation). Here, high tree cover forest remains a stable state of the system, consistent with proxy estimates of tree cover.

598 Fire is the only mechanism capable of driving bistability in our concep-
 599 tual model and is required for bistability in some dynamic global vegeta-
 600 tion model simulations [55]. But spatial data can appear bistable due to human-
 601 driven forcing that decreases tree cover independent from precipitation [19].
 602 Bistability in a system is driven by feedbacks, not forcings, so if human-
 603 driven forcing (*i.e.* deforestation and anthropogenic fire) leads to apparent
 604 bistability we cannot characterize the system as bistable. Alternatively, it
 605 is possible that fire-aridity feedbacks are stronger today than in the mid-
 606 Holocene, characterizing the system as bistable and suggesting Amazon tree
 607 cover is less resilient to drying today regardless of human interference.

608 We test whether Amazon tree cover is as resilient to drying today as in
 609 the mid-Holocene by comparing our mid-Holocene reconstructions to mod-

ern data with and without accounting for land use (human-driven forcing).
Figure 8A shows precipitation and tree cover across tropical South America today with data points colored by the land use fraction in each pixel (data from ref. [19]). Using these data, we can construct potential diagrams to identify the stable states of the system under different precipitation conditions. Potential diagrams are normalized, inverted PDFs (probability density function) of tree cover data for a given range of precipitation. The wells in a potential diagram reflect stable attractors, or stable states, and the ball represents the actual state of the system. For the modern precipitation range—defined by the RTM inversion of modern $\Delta\delta^{18}O$ data (Fig. 5)—the potential diagram shows a deep potential well at high tree cover, the stable state of the system (Fig. 8B).

Under mid-Holocene precipitation the high tree cover well shallows, tipping the ball to the new, low tree cover stable state (Fig. 8C). While this conflicts with evidence for high tree cover in the mid-Holocene [59, 138, 29, 21, 22], suggesting tree cover is less resilient to drying today, it is mostly an artifact of the intensification of modern human land use. When all pixels where the land use fraction exceeds 15% are removed, high tree cover forests remain the stable state for modern vegetation at low, mid-Holocene precipitation rates (Fig. 8E). This brings the modern precipitation-tree cover data in agreement with our mid-Holocene reconstructions, suggesting that in the absence of human disturbance, Amazon tree cover today is probably as resilient to drying as in the mid-Holocene.

6.4. Implications for the future of Amazon tree cover

Despite the vast uncertainty associated with model predictions and modern observations, some agreement surrounding the future of Amazon vegetation is emerging [20, 19]. Malhi et al., (2009) show that correcting for precipitation biases in models is crucial for predicting future vegetation assuming the modern spatial distribution of vegetation holds over time. When doing so, grasslands are an unlikely future state of Amazon vegetation [20]. More recently, Ahlström et al., (2017) demonstrate that additional corrections for land use, along with precipitation bias-corrections, bring models and observations into agreement that Amazon tree cover is resilient to drying. While our analysis cannot inform how biodiversity or vegetation species composition responds to drying, our results confirm that tree cover can remain high even when precipitation appears to fall well below future projections (*e.g.* Fig.

646 5). In the absence of human interference, our analysis could not identify
647 feedbacks that invalidate this conclusion.

648 Drying alone, at least within the precipitation range studied here, does
649 not pose a major threat to Amazon tree cover within our study region, but
650 land use does. For example, impacts of modern human deforestation on
651 Amazon water cycling are already measurable. Deforested watersheds show
652 increased runoff, balancing the loss of evapotranspiration [97, 141, 142, 143,
653 144, 145]. While deforestation reduces tree cover without directly influencing
654 precipitation (Fig. 1) the loss in moisture recycling may lead to a decrease
655 in precipitation, especially in the western Amazon [58, 57]. Drying can also
656 allow for more deforestation by increasing the effectiveness of methods that
657 use fire [46, 45], setting the stage for a positive feedback between deforestation
658 and aridity [47].

659 Temperature increases may also threaten Amazon vegetation. While we
660 do not address the vegetation response to temperature (mid-Holocene tem-
661 peratures were similar to pre-Industrial; Supplemental Fig. S10), some mod-
662 eling work suggests the modern Amazon is near a “thermal threshold” where
663 warming has a much larger effect on vegetation than cooling [146]. Future
664 work reconstructing precipitation and vegetation in warmer and colder pa-
665 leoclimate states will help account for the role of temperature in climate-
666 vegetation relationships.

667 The geometry of savanna expansion and fire is another key difference that
668 separates the mid-Holocene from modern forest loss. In the mid-Holocene
669 most grassland expansion and fire was limited to forest-grassland interfaces
670 [29, 21]. The core forested region of the Amazon Basin remained intact
671 and not fragmented. By contrast, deforestation today extends far into the
672 Amazon Basin, allowing human-driven activity to fragment forested regions
673 and exert ecological stress from the inside-out. This is especially dangerous
674 because human activity deep in the basin interior brings fires closer to core,
675 annually inundated floodplain regions, which may be particularly vulnerable
676 to forest loss and grassland expansion due to their low local tree cover and
677 longer ecological recovery times [147]. Taken together, this suggests human
678 activity can strengthen the fire feedback, perhaps making tree cover more
679 sensitive to drying than it has been in the past.

680 7. Conclusion

681 We attribute the persistence of high central and western Amazon tree
682 cover in the mid-Holocene to resilience to drying rather than to nominal arid-
683 ification [21, 22]. Our results suggest Amazon tree cover is highly resilient to
684 the magnitude of drying predicted in the worst-case-scenario, bias-corrected
685 future climate projections (RCP8.5), although forest species composition may
686 likely change. When human land use is locally, but not regionally intensive,
687 mid-Holocene climate-vegetation patterns reveal that Amazon forests are not
688 strongly sensitive to precipitation change (Fig. 4) and high tree cover persists
689 likely because natural fires and the MRF were not strong enough feedbacks
690 to drive widespread dieback. An important implication of this finding is,
691 without human interference, feedbacks that amplify forest loss (particularly
692 fire feedbacks) are too weak to tip Amazon forests to a grassland state, at
693 least under mid-Holocene conditions. This resilience likely holds in seasonally
694 dry conditions, inferred for the mid-Holocene, when fires are more common
695 [44, 45], evergreen trees are water-stressed [36], and forests take longer to
696 recover from drought [37].

697 The bulk of modern data indicate mid-Holocene levels of drying should
698 yield a savanna-state, but accounting for land use brings observations, models
699 [19], and paleoclimate reconstructions into agreement that Amazon tree cover
700 is resilient to crossing below the 2,000 mm/yr precipitation threshold. We
701 argue that, in the absence of extensive anthropogenic deforestation and fire,
702 modern tree cover is just as resilient to drying as it was in the mid-Holocene.
703 Even under the most drastic projections of future aridification, human-driven
704 deforestation, not drying, will play the larger role in the evolution of Amazon
705 tree cover.

706 Our proxy-model approach allows for a direct comparison between past
707 and modern conditions and leaves room for future work to investigate other
708 forms of resilience. For example, it is unclear how mid-Holocene biodiversi-
709 ty, ecosystem heterogeneity, and rainfall variability affected past resilience
710 [14, 15, 16, 148]. Did these factors act to maintain Amazon forests despite
711 growing water stress? Additionally, the Amazon’s resilience to mid-Holocene
712 drying fundamentally depends on our definition of the forest’s function. We
713 use tree cover as a proxy for the system’s function because it is easier to
714 compare past and present states. But ecosystem function may be defined by
715 other characteristics like biodiversity or species composition and the resilience
716 of these to mid-Holocene drying remains unknown. Future mid-Holocene

717 paleoecological work will provide a more comprehensive perspective to the
718 Amazon ecosystem resilience to drying.

719 We agree that ongoing anthropogenic fire and deforestation pose a serious
720 threat to the survival of the rainforest [7, 10, 11]. Without human interfe-
721 rence, feedbacks that can tip a forested system to a grassland may be too
722 weak to do so in the range of future precipitation projections. But fire and
723 deforestation leave a notable imprint in the Amazon Basin by extending far
724 into the interior of forested regions and providing mechanisms to strengthen
725 the feedbacks capable of causing widespread forest loss [47]. The resilience
726 of Amazon vegetation to drying when human interference is minimal implies
727 that, despite global climate change, curbing fire and deforestation will have
728 the most immediate and long-lasting effect on the preservation of Amazon
729 tree cover.

730 **Data Availability**

731 All proxy and DGVM data are available in the supplemental material
732 of this manuscript. The following data can also be downloaded from their
733 original sources. Charcoal data can be downloaded from the Global Char-
734 coal Database (GCD version 2.0) using the paleofire package in R ([https://
735 gpwg.paleofire.org/](https://gpwg.paleofire.org/)). Pollen data can be downloaded from the Neotoma
736 database (<https://www.neotomadb.org/>) and the ACER pollen and char-
737 coal database (see supplement of ref. [117]). The Tigre Perdido isotope
738 record [102] (part of the western composite site) can be downloaded from
739 the SISAL database (siteID: 25). The Diamante record [32] (also part of the
740 western composite) was provided by H. Cheng. The Paraíso record can be
741 found in the supplement of ref. [30]. The time-interpolated and temperature-
742 corrected $\delta^{18}O$ and $\Delta\delta^{18}O$ data are in the supplement of this text. The reac-
743 tive transport model code can be found in the supplement of ref. [103] or at
744 https://github.com/tykukla/Vapor_Transport_Model_KuklaEtAl2019.

745 **Competing Interests Statement**

746 The authors declare no competing interests.

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