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Dry corridors opened by fire and low CO₂ in Amazonian rainforest during the Last Glacial Maximum

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1 **Abstract**

2 The dynamics of Amazonian rainforest over long timescales connects closely to
3 its rich biodiversity. While palaeoecological studies have suggested its stabil-
4 ity through the Pleistocene, palaeontological evidence indicates the past exist-
5 ence of major expansions of savanna and grassland. Here we present integrated
6 modeling evidence for a grassier Neotropics during the Last Glacial Maximum
7 (LGM), congruent with palaeoecological and biological studies. Vegetation re-
8 constructions were generated using the Land Processes and eXchanges (LPX)
9 model, driven by model reconstructions of LGM climate, and compared against
10 palynological data. A factorial experiment was performed to quantify the im-
11 pacts of fire and low CO₂ on vegetation and model-data agreement. Fire and
12 low CO₂ both individually and interactively induced widespread expansion of
13 savanna and grassland biomes while improving model-data agreement. The in-
14 teractive effects of fire and low CO₂ induced the greatest ‘savannafication’ of the
15 Neotropics, providing integrated evidence for a number of biogeographically rel-
16 evant open vegetation formations including two dry corridors; paths of savanna
17 and grassland through and around Amazonia that facilitated major dispersal

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18 and evolutionary diversification events. Our results show a bimodality in tree
19 cover that was driven by fire and further enhanced by ‘CO₂ deprivation’, which
20 suggests biome instability in this region of climate space.

21 **Main Text**

22 Biotic diversification in the Neotropics has been hypothesized to rely on broad
23 changes in vegetation, which induce large-scale processes such as dispersal, vi-
24 cariance, and speciation [35]. The Refugia Hypothesis postulates that Ama-
25 zonian rainforest was fragmented into disjunct ‘refugia’ by tracts of savanna
26 during Pleistocene glacial periods. This fragmentation would be sufficient to
27 inhibit gene flow between refugia, allowing speciation. Forests would expand
28 and reconnect during interglacials, leading to range expansion of the newly di-
29 versified taxa [20]. Diversification in the Atlantic forest region in southeastern
30 Brazil has also been attributed to refugia dynamics [9].

31 A number of palaeoecological studies have contradicted the Refugia Hypoth-
32 esis [11][4][5], claiming that Amazonian forests remained intact over Pleistocene
33 climatic fluctuations. These studies are broadly consistent with some past model
34 reconstructions [14][10][36][23], which show Amazonian forest during the LGM
35 to be similar in extent relative to its pre-Industrial status. However, these
36 studies relied on climatic factors only in their reconstructions, performed with-
37 out explicit examination of the effects of non-climatic factors such as fire and
38 CO₂. Furthermore, recent palynological studies have found savanna-like veg-
39 etation during the late Pleistocene within current Amazonian forest [22][18].
40 Re-evaluation of the Lake Pata records has also challenged the notion of stable
41 moist forest even within the basin over glaciations, finding discontinuities and
42 significant compositional changes during the last glacial period [17]. The degree
43 of ‘savannafication’ of the glacial Neotropics is thus largely unknown due to the
44 scarcity of pollen cores that date back to the LGM.

45 Disjunct distributions of species associated with semi-arid biomes also sug-

46 gest the past presence of open vegetation biomes in regions that are currently
47 occupied by closed canopy tropical forest [41][38]. Three past savanna forma-
48 tions, referred to as ‘dry corridors’, have been hypothesized to explain past con-
49 nectivity between the northern and southern savanna regions of South America
50 [8][16][41][48]. The central Amazonian corridor would extend diagonally north-
51 west to southeast from the savannas in the northern Amazonian border to those
52 central Brazil, along an extensive tract of forest that currently experiences a
53 significant degree of seasonality in precipitation [50]. The circum-Amazonian
54 corridor, also known as the Andean corridor, would have existed along the An-
55 des and western border of Amazonian forests, while the coastal corridor has
56 been hypothesized to have existed along the Atlantic coast.

57 Here, we estimated the effects of fire processes and CO_2 on vegetation in the
58 Neotropics during the LGM, addressing the stability of biomes under glacial
59 conditions using a comprehensive process-based model. Emphasis was placed
60 on estimating the isolated and interactive effects of fire and low CO_2 on biome
61 distribution and tree cover. A factorial experiment was conducted to eluci-
62 date these effects, consisting of four scenarios all driven by LGM climate: pre-
63 Industrial CO_2 (280 ppm) without fire processes (*control*), pre-Industrial CO_2
64 (280 ppm) with fire processes (*fire only*), LGM CO_2 (180 ppm) without fire
65 processes (*low CO_2 only*), and LGM CO_2 (180 ppm) with fire processes (*fire*
66 *and low CO_2*).

67 To account for variability among climate models, four distinct model re-
68 constructions of LGM climate were used to drive the Land surface Processes
69 and eXchanges (LPX) [40]. The outputs from all four of the LGM climate
70 reconstructions were then averaged to create an ensemble data set that was
71 also subject to analysis. Each scenario from the five sets of vegetation recon-
72 structions was then compared against palynological data to assess model-data
73 agreement.

74 **Comparing Model Reconstructions to Palynological Data**

75 The activation of fire and low CO₂ resulted in small but consistent improvements
76 in agreement between model reconstructions of LGM vegetation and palynolog-
77 ical data (42 cores). Expansion of open biomes and displacement of forest
78 underlie these improvements. Model-data agreement was further improved by
79 the simultaneous activation of fire and CO₂ relative to control for all five ex-
80 periments, four of which were deemed statistically distinct from the ensemble
81 experiment according to a paired student t-test (fig. 2). Pollen-derived biomes
82 were compared against model reconstructed biomes quantitatively using the Dis-
83 crete Manhattan Metric (DMM) by assessing the net distance between biomes'
84 ecological affinities at each site (methods). The activation of fire (*fire only*) and
85 the imposition of low CO₂ (*low CO₂ only*) in isolation both improved model-
86 data agreement, though fire activation consistently improved DMM scores more
87 than low CO₂. The simultaneous inclusion of fire processes and low CO₂ (*fire*
88 *and low CO₂*) resulted in the lowest average DMM and highest model-data
89 agreement, highlighting the importance of their combined effects.

90 Variations between LGM climate reconstructions were reflected in model-
91 data agreement, though almost all AOGCM runs showed similar relative per-
92 formance within their factorial experiments (fig. 2). The control scenario had
93 lowest data-model agreement, followed by *fire only* and *low CO₂ only* scenarios,
94 with *fire and low CO₂* having the highest relative model-data agreement save
95 for FGOALS-1.0g. Of the twenty model vegetation reconstructions, the ensem-
96 ble *fire-and-low CO₂* run agreed most with palynological data. The degree of
97 testing rigor is limited by the number and locations of suitable pollen cores,
98 which are rare in Amazonia due to poor preservation conditions. Though the
99 distribution of cores is thus far insufficient for a basin-wide test, such compar-
100 isons can effectively identify physical processes and model skill at the local and
101 regional scale.

102 **Fire and Low CO₂ Drove Expansions of Grasslands**

103 Fire and low CO₂ had considerable effects on model reconstructions of veg-
104 etation, potentially outweighing the effects of glacial climate as examined by
105 previous studies [36][23]. Relative to control, wildfire at pre-Industrial CO₂ (*fire*
106 *only* scenario) showed a general shift to biomes associated with increased arid-
107 ity. Tropical moist forest was reduced by 11% relative to control (7.6 million
108 km² vs 8.6 million km²), which was replaced predominantly by savanna (fig. 1
109 a, b). Warm temperate forest, most of which was Atlantic forest, also showed
110 reductions in area and were replaced by sclerophyll woodland. Area occupied
111 by grassland in the *fire only* scenario (4.2 million km²) was 350% greater than
112 that of control (1.2 million km²), replacing savanna in the current caatinga of
113 Northeastern Brazil and woodland and parkland in and adjacent to the Pampas
114 region in the south. In northern Colombia and Venezuela, fire was found to
115 induce desert where savanna and grassland occupied in the control.

116 The effects of low CO₂ (*low CO₂ only* scenario) appear to have a generally
117 similar effect on biome distributions as fire: forest biomes are reduced in extent
118 by intrusions of open vegetation biomes (fig. 1 a, c). Low CO₂ induced shifts
119 in biomes similar to that of the *fire only* scenario. However, unlike fire, low
120 CO₂ replaced entire regions of tropical dry forest with savanna and grassland.
121 Tropical moist forest was reduced by 13% relative to control, from 8.5 million
122 km² to 7.4 million km². Low CO₂ also induced significant expansion of grassland
123 regions.

124 The combined effect of fire and low CO₂ had the strongest impacts on biome
125 distribution relative to control. Reductions in forest area were dramatic as were
126 expansions of open vegetation biomes, including regions that were unaffected by
127 either factor in isolation. Amazonian forest contracted significantly, particularly
128 the southern margins. Savanna and grassland showed extensive intrusion into
129 the central Amazonian corridor region, with small patches of isolated savanna
130 of various size within the forest. Tropical forest in the *fire and low CO₂* scenario
131 occupied 56% of the area (4.8 million km²) relative to control (8.5 million km²).

132 Warm temperate forest, which represents Atlantic forests, occupied only 15% of
133 the area relative to control. Conversely, grassland expanded by 550% relative
134 to control (6.6 million km² vs. 1.2 million km²), occurring in large patches in
135 the mosaic of open canopy biomes that comprise the majority of the *fire and*
136 *low CO₂* scenario.

137 These trends were consistent among all LGM model reconstructions driven
138 by the four climate reconstructions, though with significant variation. Amazo-
139 nian rainforests remained vast and continuous in scenarios in *fire only* and *low*
140 *CO₂ only* scenarios, consistent with past modeling studies [14][10][36][23]. It is
141 only with both effects simultaneously activated that our reconstructions diverge
142 from past modeling studies.

143 Whittaker plots for model runs were generated to examine the effects of
144 low CO₂ and fire on climate-vegetation relations (fig. 4), showing the location
145 of biomes in climate space. Scenarios without fire tend to have biomes that
146 occupy clearly defined areas of climatic space. The presence of fire tends to
147 obscure these well-defined boundaries. The expansion of grassland is a clear
148 effect of fire, which is well-reflected in its broad and speckled climate-space
149 distribution. Low CO₂ tends to reduce the area of climate-space of all major
150 forest types, allowing savanna to encroach into the climate-space of tropical
151 moist and dry forest and sclerophyll woodland to encroach into that of warm
152 temperate forest. The combined effect of low CO₂ and fire compounds these two
153 effects, resulting in much reduced climate-space of forest biomes while expanding
154 that of grassland (fig. 4 a,d).

155 **Interactive Effects of Fire and Low CO₂**

156 To discern the individual and interactive effects of fire and CO₂ on tree cover,
157 a Stein-Alpert factor separation was performed on the ensemble experiment
158 (methods). Individually, fire and low CO₂ had showed similar effects on tree
159 cover, inducing large reductions in most of Central and South America, while
160 having little to no effect on Amazonia (fig. 5 a,b). Low CO₂ individually had

161 broader effects on tree cover, showing small reductions in the central dry corridor
162 regions of Amazonia and major reductions in southeast of Brazil (fig. 5 a, c).
163 Interactions between fire and low CO₂ resulted in enhanced tree cover loss in
164 several regions, including the northern savanna of Colombia and Venezuela and
165 large patches of Amazonia in central Brazil (fig. 5 a, d).

166 In the control scenario, a spectrum of tree cover occurred over regions of
167 intermediate precipitation (0 - 2000 mm) (fig. 6 a). Intermediate tree cover
168 (30 - 60 %) is present, showing little if any sign of bimodality of tree cover.
169 The activation of low CO₂ tends to reduce the density of very high tree cover
170 (~ 0.95), likely expressing reductions of tree cover even in regions that receive
171 ample precipitation (fig. 6 c). The activation of fire results in low tree cover
172 even in regions of high precipitation, relative to control (fig. 5 b). Fire also has
173 another distinct effect on tree cover: intermediate tree cover is less frequent and
174 the spectrum of tree cover is shaped into a stronger bimodality. The fire-forced
175 bimodality of tree cover is further enhanced by the simultaneous activation of
176 low CO₂ and fire, showing more distinct reductions of intermediate tree cover
177 over a broader range of precipitation (fig. 6 d).

178 **Open Dry Corridors and Forest Stability**

179 A number of hypothesized formations of open vegetation, including two dry
180 corridors, were identified in our model reconstructions (fig. 3). Indications of
181 a central Amazonian corridor appeared in a number of model runs, showing
182 almost full connectivity in the ensemble *fire and low CO₂* scenario, which had
183 the highest agreement with pollen records. In this run, savanna significantly
184 expanded through the southeast margins of Amazonia to connect with the non-
185 forested biomes in the far north. A narrow, continuous tract of savanna and
186 grassland resembling a circum-Amazonian dry corridor was also reconstructed
187 in a number of runs, also most prominent in *fire and low CO₂* scenarios. Moist
188 forest remained robust along the Atlantic coast in northern Brazil, showing no
189 indication of a trans-Amazonian Atlantic corridor at the biome level. However,

190 canopy density and height were slightly reduced according to the ensemble veg-
191 etation reconstruction (fig. ED1, ED2), suggesting a degree of openness in the
192 region. It is notable that the broad savannafication required to open the central
193 Amazonian dry corridor seems contingent on the interactive effects of fire and
194 low CO₂, while both the individual and interactive effects of fire and low CO₂
195 induce opening of the circum-Amazonian corridor.

196 Atlantic forests were found to be present but heavily reduced in size and
197 restricted to the Brazilian coast, consistent with phylogeographic and statistical
198 modeling work of Carnaval and Moritz [9]. The Pernambuco and Bahia refuges
199 were both present in the *fire and low CO₂* reconstruction, separated from one
200 another by tropical savanna.

201 The effects of wildfire and low CO₂ consistently improved model-data agree-
202 ment, suggesting their potential importance in determining vegetation in the
203 LGM Neotropics. Both fire and low CO₂ induced expansions of open vegeta-
204 tion biomes such as savanna and grassland while reducing forested area, partic-
205 ularly in Amazonia. Though reduced in size, Amazonian rain forest remained
206 largely continuous with a stable western core. Thus, we conclude that vege-
207 tation in LGM Neotropics may have been more open than previously thought,
208 with significantly less forest and significantly more mosaics of grassy biomes. A
209 number of areas, including large regions in the southern margins of Amazonia
210 and central corridor location, show marked instability with regard to changes
211 in climate, low CO₂, and fire regime.

212 Past refutations of widespread glacial savannafication were often based on
213 single or a small number of study sites [12][6], and were likely insufficient to con-
214 clusively exclude savanna and grassland expansion at the regional-scale. Moist
215 tropical forest was correctly reconstructed at the Lake Pata site in all twenty
216 LGM vegetation reconstructions performed in our study, though with significant
217 variation in extent and shape of Amazonian rainforest. Existing refutations
218 have not addressed more recent pollen studies in the approximate central corri-
219 dor and Amazonian southern margin, such as the Serra Sul Carajas and Lago

220 do Saci cores, that are interrupted by sedimentary hiatuses around the LGM
221 or suggest savanna during the LGM [22][18]. Past regional modeling studies
222 that suggested that Amazonia was robust during the LGM were performed us-
223 ing equilibrium models, which lacked fire representation and its interactions
224 with low CO₂ [14][30]. Statistical reconstructions of biome distributions during
225 the LGM are based on the assumption that climate solely controls vegetation
226 distribution, and also neglect the effects of low CO₂ and fire [48][13].

227 **Interactive Stressors on Neotropical Trees**

228 Our results are also consistent with two prominent ecological theories: the
229 fire-driven bimodality of tree cover and the expansion of grasses due to low
230 CO₂. Our results also emphasize the oft-neglected but potentially critical role
231 of non-climatic drivers on vegetation cover in palaeoecological contexts. Within
232 the context of the LGM, low CO₂ gave grasses a competitive advantage over
233 trees, compounded by the effects of fire, and worsened by aridity. These pro-
234 cesses would better explain palaeoecological evidence of a largely grassier South
235 America during the LGM, which was overall more vulnerable to intrusions by
236 savanna. Fire was found to be a driving factor in the bimodality of tree cover,
237 with enhanced efficacy at lower levels of CO₂; a potentially significant interac-
238 tion between fire and ‘CO₂ deprivation’. While bimodality of tree cover exists
239 in regions of intermediate precipitation, this is not proof that forest and savanna
240 are alternate stable states, as MAP is too coarse a measure of environmental
241 niche. However, this bimodality of tree cover does indicate instability in this
242 region of climate space, where small changes in growing conditions may lead to
243 large changes in tree cover. In this case, fire may induce a positive feedback
244 in reducing tree cover, where the initial fire-driven proliferation of grasses leads
245 to faster accumulation of dry fuel to drive further fires [49]. Similar results
246 have been derived from simulations of climate change in Africa, where CO₂ and
247 its interactions with fire were found to have major impact on determination of
248 savanna and forest biomes [33]

249 There are a number of broad implications if fire and CO₂ are indeed im-
250 portant determinants of terrestrial vegetation. Even when driven by identical
251 climate scenarios, modifications of CO₂ and fire regime can significantly alter
252 structure and distributions of terrestrial vegetation. Relations between vegeta-
253 tion and climate are sensitive to non-climatic factors and dynamic through time
254 due to fluctuating CO₂. Burbridge et al. suggested the likely role of low CO₂
255 in LGM expansion of dry forest and savanna in Bolivian Amazonia, which is
256 consistent with our findings [4]. We thus recommended greater consideration
257 of non-climatic processes such as fire and CO₂ in palynological reconstructions,
258 particularly in ecotonal regions. Similarly, ecological niche modeling studies
259 which also assume robust, stable relationships between vegetation and climate,
260 may also face this issue if reconstructing vegetation over geologic time. With-
261 out accounting for fire and low CO₂ in glacial periods, stability of forest biomes
262 may be overestimated and vegetation-mediated diversification processes may be
263 obscured.

264 **Biogeographical Implications**

265 A major consequence of our study was the modeled reconstructions of a num-
266 ber of biologically-significant formations including the central Amazonian and
267 circum-Amazonian dry corridors and Atlantic forest refugia. There is currently
268 no suitable palynological data from the core of the central northwest-southeast
269 savanna corridor, which limits our ability to conclusively determine its past
270 existence and motivates contextualization of our results with biogeographical
271 studies. Our model reconstructions of a central Amazonian dry corridor are con-
272 gruent with molecular genetic studies of the Neotropical rattlesnake (*Crotalus*
273 *durissus*), which occupies seasonal habitats ranging from Mexico to Argentina
274 [41]. Though widespread through the Neotropics, they are absent in Amazonian
275 and Central American moist forest and have generally disjunct distributions.
276 Phylogeographic analysis suggests a progressive, step-wise colonization from
277 Central America to the southeast Brazil traversing the Amazon basin. While

278 divergences in Central American clades were likely to have occurred in the late
279 Miocene or early Pliocene, trans-Amazonian vicariant cladogenesis likely oc-
280 curred more recently in the mid-Pleistocene [50]. This in turn would require
281 a ‘shrunk or fragmented’ Amazonian forest that gives way to seasonal or open
282 vegetation formations that could act as continuous corridors for dispersal.

283 Though the value of establishing the evolutionary history of Neotropical
284 biota within the context of Earth history has been recognized [1][37], actual
285 research aimed at integration has been curbed by the spatial and temporal limi-
286 tations of palaeoecological data [27]. Process-based modeling, particularly when
287 grounded in empirical data, offers extensive and continuous reconstructions of
288 past environments that can justifiably interpolate between data sites.

289 A mechanistic understanding of glacial period savannafications could have
290 important consequences, with conceptual similarities to Haffer’s model of Pleis-
291 tocene diversification [21][19]. During glacial periods, open vegetation corridors
292 would emerge and induce dispersal and range expansion for savanna and grass-
293 land adapted taxa. During interglacials, forest would replace open biomes and
294 dry corridors would close, inducing vicariance (geographical separation of popu-
295 lations) and differentiation. Conversely, a broad central Amazonian non-forested
296 corridor could feasibly act as a barrier and isolate populations of moist forest
297 taxa at opposing sides of the corridor inducing diversification. Closing of cor-
298 ridors during interglacials would then allow the newly diversified moist forest
299 taxa to disperse and expand their ranges. This process could be cyclic, follow-
300 ing Pleistocene climatic fluctuations, and act as a two-way pump or ‘accordion’.
301 This would allow periodic and rapid diversification, given that each phase leads
302 to vicariance of taxa of alternating forest and savanna niches. This is consistent
303 with existing theories of alternating moist and dry habitat corridor and evi-
304 dence of sustained diversification of both moist forest and arid-adapted species
305 during the Pleistocene [45]. It is noteworthy that gallery rainforests along the
306 Amazon river could persist during dry periods and act as a migration corridor
307 for moist forest species, which could compromise the effectiveness of a central

308 dry corridor. While the duration and extent of corridor may not permit full
309 speciation, this process could drive recent intraspecific genetic diversification.
310 This unique confluence of factors involving the geography of the Neotropics and
311 Pleistocene climatic oscillations may have contributed to an accumulation of
312 biological diversity.

313 **Methods**

314 **Model Description and Protocol**

315 Dynamic Global Vegetation Models (DGVMs) simulate spatially and temporally
316 resolved potential vegetation and ecosystem structure (e.g. height, biomass, leaf
317 area index and foliage projective cover) and function (e.g. biogeochemical car-
318 bon and water fluxes and disturbance) via a number of competing Plant Func-
319 tional Types (PFTs). These models, therefore, provide an integrated, process-
320 based way to estimate the impacts of climate changes on terrestrial ecosystems
321 [15][44]. This includes hindcasts of past impacts of climatic change on the
322 Earth’s land surface to complement palaeoecological data, which also serves to
323 ground-truth models over long timescales [32].

324 The DGVM used for this study, Land surfaces Processes and eXchanges
325 (LPX) [40], is a descendant of the widely used Lund-Potsdam-Jena DGVM [43]
326 coupled to the processed based SPread and InTensity of FIRE model (LPJ-
327 SPITFIRE) [47]. PFTs in LPX are initially defined by life form (i.e woody
328 plant or grass). Trees PFTs are then split by climate range (tropical, temperate,
329 boreal) leaf type (broad or needle leaf), and phenological response (evergreen,
330 raingreen or summergreen). Grasses are split by C3 and C4 photosynthetic
331 pathways. LPX uses a photosynthesis-water balance scheme that couples CO₂
332 assimilation with transpiration. Reduced CO₂ concentrations, such as those
333 during the LGM, increases potential water stress on plants by increasing the
334 required stomatal conductance (g_c) for the same photosynthetic rate. The max-
335 imum, unlimited potential stomatal conductance ($g_{c,max}$) is determined by the

336 maximum potential day-time photosynthetic assimilation rate (A_{max}), ambient
337 CO_2 concentration, PFT-specific minimum canopy conductance and, for grasses,
338 photosynthetic pathway. If $g_{c,max}$ results in a maximum transpiration demand
339 (D) that is greater than the supply of water from the soil (S - calculated from
340 soil water content and soil properties), then g_c - and therefore photosynthesis
341 - is reduced so that transpiration is equal to S, as described by Monteith [34].
342 When CO_2 is decreased, $g_{c,max}$ increases even if A_{max} is unchanged, and the
343 value of S that induces water stress is increased - i.e maximum photosynthesis
344 only occurs at higher soil moisture. Photosynthesis is also reduced by a greater
345 amount than with higher CO_2 if S is less than D.

346 In LPX's process-based fire model [40], fire occurrence is a product of igni-
347 tions represented by lightning, the mean probability of ignition calculated from
348 local fuel and atmospheric moisture content, and fire susceptibility based on fuel
349 amount, fuel properties and fuel moisture content. Fires that do start have a
350 rate of spread, flame height and residence time are based on weather conditions
351 and fuel moisture, and calculated using the Rothermel equations [42]. Burnt
352 area is the product of the number of fires and the average spread of fire. PFT
353 mortality in a given area burnt can occur from either cambial damage and/or
354 crown scorching. Cambial damage increases with fire intensity and residence
355 time, but is resisted by the PFTs local bark thickness, and thicker-barked trees
356 surviving longer, more intense fires. Bark thickness is related to PFT height
357 via a PFT-specific allometry relationship. The impact of crown scorching is
358 determined by fire intensity and height in relation to the height of the locally
359 simulated PFTs. Crown scorch mortality increases as flame height increased
360 above canopy height of each PFT. LPX its simulated fire regimes have been ex-
361 tensively ground-truthed against modern data [39][40][24][25][26], which made
362 it a particularly suitable model for the purposes of this study. A more compre-
363 hensive descriptions of LPX and its components can be found in [40][24][47].

364 **LGM Climate Scenarios and Modeling Protocol**

365 Inputs reflecting the LGM climate were derived from four atmosphere-ocean gen-
366 eral circulation model (AOGCM) datasets (MIROC3.2, FGOALS-1.0g, HadCM3M2,
367 CNRM-CM33) produced by the Palaeoclimate Modelling Intercomparison Project
368 Phase II (PMIP2) [2]. Boundary conditions (ice sheets, coastlines, greenhouse
369 gas concentrations, eccentricity, obliquity, and angular precession) for these runs
370 are described in [2]. These datasets were also used to drive previous global-scale
371 reconstructions of LGM vegetation [39][7]. Atmospheric CO₂ was set to 185 ppm
372 in accordance with the PMIP2 protocol for all runs [3].

373 Several steps were required to prepare these model scenarios for compatibil-
374 ity with LPX. Differences between LGM climate scenarios and a Pre-Industrial
375 Holocene baseline climate variables (‘anomalies’), were superimposed onto a
376 0.5° × 0.5° grid Pre-Industrial climate scenario (Climate Research Unit Version
377 3.0, detrended data from 1900-1950). Climate data with superimposed anoma-
378 lies was then extrapolated onto the exposed continental shelf and removed from
379 regions covered by ice sheets characteristic of the LGM.

380 For all model simulations, ‘spin-ups’ from bare ground were ran for 4000
381 years and main runs for 1380 years. Equilibrium tests were performed on each
382 data set to check for the temporal stability of output variables (fig. S3). If for
383 the majority of grid cells, canopy density (Leaf Area Index - LAI) and foliage
384 projective cover (fpc) showed less than 2% variation relative to the previous
385 time step, equilibrium was taken to have been reached. Model output for the
386 last 138 years (length of base Pre-Industrial data set) was averaged and used
387 for reconstructions of LGM vegetation.

388 **LPX Biome Assignment**

389 Biome assignment was implemented through post-processing of three LPX out-
390 puts: Growing Degree Days (GDD), vegetation height, and foliage projective
391 cover (fpc) (fig. S4). The biome assignment scheme used in this study is iden-
392 tical to that of Prentice et al. [39] and Calvo et al. [7], save a refinement to

393 distinguish seasonal and evergreen forests. A threshold of mean annual GDD
394 (above 5°C) was set to 350°C·days to separate cold biomes from their warm
395 and tropical counterparts (ex. tundra vs desert). Height and fpc are then used
396 to distinguish between bareground, grassland, savanna, and forest biomes. The
397 presence or dominance of PFTs within a grid cell then determines its assign-
398 ment as tropical, boreal or temperate. Forests are classified as either seasonal
399 or evergreen based on the relative proportions of summergreen, raingreen, and
400 evergreen forest pfts.

401 **Model-Pollen Biome Correspondence**

402 The definitions of South American biomes, their correspondence to pollen spec-
403 tra, and LGM biome reconstructions were based on a comprehensive meta-
404 analyses by Mayle et al. [23] and Marchant et. al [28] in addition to the original
405 studies (Extended Data - table 1). Pollen-based biomes and the model-outputs
406 rarely have a simple one-to-one correspondence, given that vegetation model
407 biomes are often developed for global application, while pollen-based biome
408 reconstructions are refined to more subtle regional or local definitions. Model-
409 data comparison was performed using the Discrete Manhattan Metric (DMM)
410 applied to the four fire/CO₂ scenarios for five climate reconstruction inputs (four
411 AOGCM climate reconstructions and one ensemble) for a total of twenty cases.
412 The DMM characterizes biomes through bioclimatic and ecophysiological traits,
413 permitting a quantification scheme to rate the ‘distance’ between two biomes.
414 The scores for each scenario represents the average distance, with smaller values
415 indicating higher average agreement between model and pollen reconstructions.

416 **Discrete Manhattan Metric**

417 **Discrete Manhattan Metric**

418 The ‘Discrete Manhattan Metric’ (DMM) was developed to quantify the distance
419 or ‘closeness’ between biomes, permitting direct comparison between model out-

420 put and pollen records and a measure of overall model-data agreement. Each
421 biome has a set of ecophysiological affinities, represented by a discrete number,
422 resulting in an *affinity matrix* (fig. 4). Given two biomes, each having an affin-
423 ity for each trait represented by a number between zero and one, we find their
424 ‘ecological distance’ by: 1) calculating the difference in their affinity scores for
425 each trait 2) summing the magnitude of these differences 3) normalizing by the
426 total number of traits.

427 The x_{ij} element of affinity matrix, X_{MN} (Extended Data - table 2), is the
428 affinity for the i th biome to the j th trait for N biomes and M traits. The
429 distance, $d(a, b)$ between two biomes (indexed by a and b) is the sum of the
430 differences in traits, normalized by the total number of traits N (eq. 1).

$$d(a, b) = \sum_{j=1}^N |x_{aj} - x_{bj}|/N \quad (1)$$

431 In the case where the biomes are the same, their distance between them is equal
432 to 0. If the biomes are maximally different (ex. tropical rainforest and tundra),
433 their distance is equal to 1. This process would be repeated for every pollen-
434 core site, where the modeled biome reconstruction for that point would be tested
435 against the pollen-based reconstruction. Model performance is thus the mean
436 of the scores for each pollen site, multiplied by two to remain consistent with
437 properties of the continuous Manhattan Metric used in modern benchmarking
438 [24]. A paired student t-test was performed on all 20 model-data comparisons to
439 determine the likelihood of equality between mean scores with a p-value cutoff
440 of 0.05.

441 The correspondence between the LPX biome assignment scheme and pollen-
442 based biomes were based on meta-analysis by Mayle et al. [23], Marchant et
443 al. [28], and the original palaeoecological studies. A number of major pollen-
444 reconstructed biome under examination had natural correspondences with model
445 assigned biomes, while others were more subtle and open to interpretation (Ex-
446 tended Data - table 3). For example, tropical rainforest and dry forest from

447 pollen studies had a natural correspondence to tropical humid forest and tropi-
 448 cal dry forest in LPX, while the various reconstructions of open, non-analogue
 449 vegetation sites were more difficult to categorize in terms of model biomes.
 450 While correspondences may be rudimentary, the impact of errors in catego-
 451 rization would be softened by the design of the DMM, as opposed to a direct
 452 binary metric. Moreover, our study developed and applied a scheme to quantify
 453 model-data agreement between DGVM output and pollen data, which is rarely
 454 attempted despite its importance. Further details and references to original
 455 studies for each core site are located in the ‘core list’ (Extended Data - table 1).

456 **Stein-Alpert Decomposition**

457 A Stein-Alpert decomposition was designed to compute isolated and synergistic
 458 effects of factors within numerical simulations [46] . Though initially developed
 459 for atmospheric models, this factor separation scheme can be adapted for cli-
 460 matic and non-climatic factors within vegetation models. In our decomposition,
 461 f_0 , f_1 , f_2 , and f_{12} are fields composed of tree cover outputs from the ensemble
 462 climate reconstruction for the *control*, *low CO₂ only*, *fire only*, and *fire and low*
 463 *CO₂* (eq. 2-5) scenarios respectively.

$$f_0 : \text{fire off, Pre-Industrial CO}_2 \quad (2)$$

464

$$f_1 : \text{fire off, LGM CO}_2 \quad (3)$$

465

$$f_2 : \text{fire on, PI CO}_2 \quad (4)$$

466

$$f_{12} : \text{fire on, LGM CO}_2 \quad (5)$$

467 Effects from the factors of fire and CO₂ are calculated by addition and subtrac-
 468 tion of the fields. The isolated effect of fire is the difference between the tree

469 cover from the *fire only* and the *control* scenarios (eq. 6). Similarly, the isolated
470 effect of low CO₂ is the difference between the *low CO₂ only* and *control* sce-
471 nario (eq. 7). The simultaneous effect of both fire and low CO₂ is represented
472 by $\langle f_3 \rangle$, which is not a component of a formal Stein-Alpert decomposition
473 (eq. 8). The synergistic effects of fire and CO₂ are computed by subtracting
474 both the tree cover from the *fire only* and *low CO₂ only* scenarios from the *fire*
475 *and low CO₂* scenario while adding the tree cover from the *control* scenario (eq.
476 9).

$$\langle f_1 \rangle = f_1 - f_0 \quad (6)$$

477

$$\langle f_2 \rangle = f_2 - f_0 \quad (7)$$

478

$$\langle f_3 \rangle = f_{12} - f_0 \quad (8)$$

479

$$\langle f_{12} \rangle = f_{12} - (f_1 + f_2) + f_0 \quad (9)$$

480 A logit transformation was performed on the tree cover fields f_i to convert the
481 bounded variable of tree cover (ranging from 0 to 1) to an unbounded variables
482 (ranging from $-\infty$ to ∞) (refer to eq. 10 - 12). For a bounded variable y , which
483 in our case are values of tree cover within each grid cells of a scenario, equation
484 10 to 11 are applied to create an unbounded transformed variable y_2 , which
485 undergoes the arithmetic of the Stein Alpert decomposition. Equation 10 was
486 used to transform the bounds of tree cover from [0,1] to (0,1), as is required for
487 the logit transformation of equation 11. Afterward the arithmetic is performed,
488 equation 12 is used to transform y_2 back to the initial bounded variable y .

$$y_1 = (99y + 0.5)/100 \quad (10)$$

489

$$y_2 = \log(y_1/(1 - y_1)) \quad (11)$$

490

$$y = 2/(1 + e^{-y_2}) - 1.0 \quad (12)$$

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679 **Author Contributions**

680 HS: Project lead, project design, performed experiment, analysis, and writing.

681 DIK: Project design, analysis, post-processing of data, figures, and writing.

682 SJM: Project design, analysis, and writing.

683 MMC: Development of model and execution of experiment.

684 SAC: Project design, analysis, and editing.

685 ICP: Model development, analysis, and writing.

686 **Data Availability**

687 The data and code that support the findings of this study are available here.

688 **Code Availability**

689 LPX is available at the following repository: <https://bitbucket.org/teambcd/lpx>

690 .

691 **Competing Interest Statement**

692 The authors declare no competing interests.

⁶⁹³ **Figure Legend**

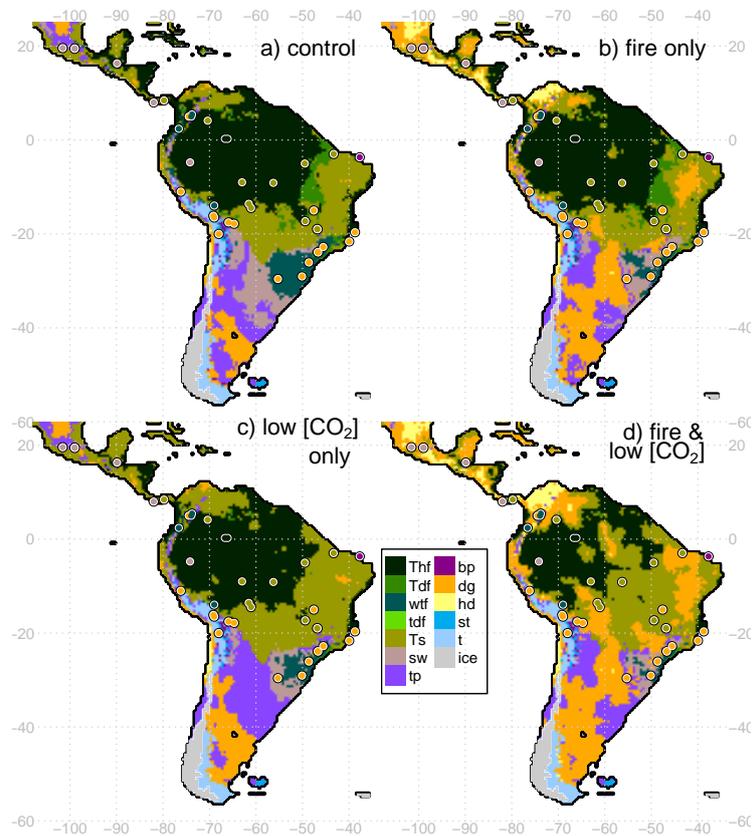


Figure 1: LPX model reconstructions of LGM biome distributions for four scenarios based on the average output from four driving AOGCM LGM climate reconstructions (ensemble experiment). ‘Fire’ indicates the presence of a process-based wildfire representation (b,d). ‘Low CO₂’ indicates scenarios with an LGM CO₂ of 180 ppm (c,d) rather than a Pre-Industrial CO₂ of 280 ppm. Dots represent the locations of pollen cores used to validate model reconstructions while colors indicate the biome reconstructed from the pollen spectra.

Thf = tropical humid forest, Tdf = tropical dry forest, Ts = tropical savanna, sw = sclerophyll woodland, tp = temperate parkland, bp = boreal parkland, dg = dry grass/shrubland, hd = hot desert, st = shrub tundra, t = tundra

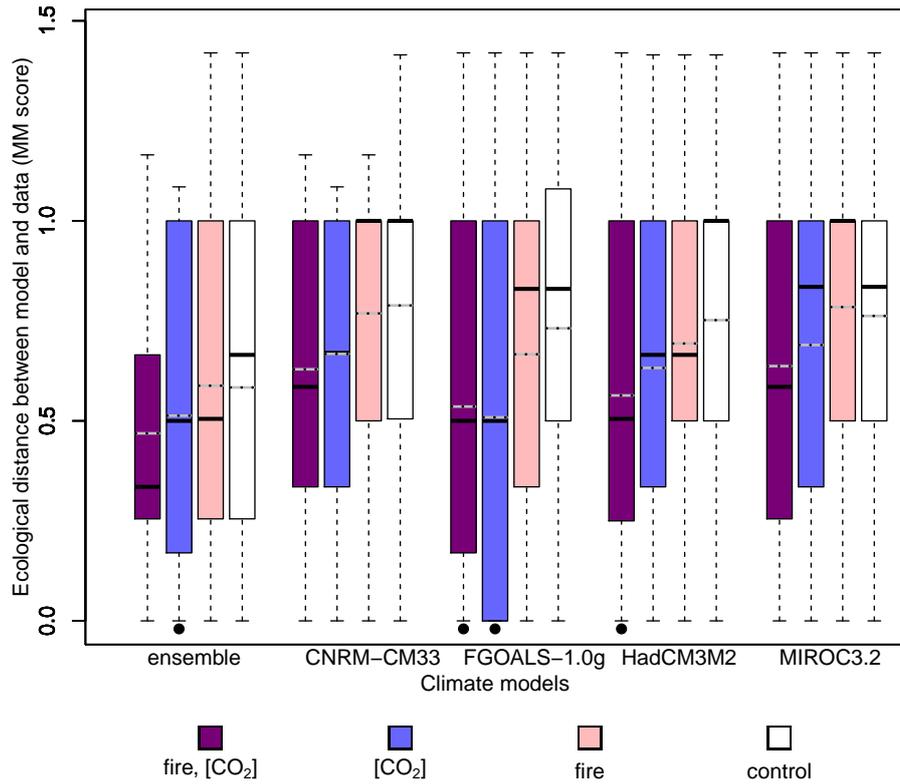


Figure 2: Statistical comparison between model reconstructions and pollen cores. Summary of model-data comparison of scenarios from the fire/low CO_2 factorial experiment for each LGM climate reconstruction, performed by comparing biome reconstructions by LPX with site-based pollen reconstructions of biomes by the Discrete Manhattan Metric (DMM). The DMM a measure of the ecological distance between biome types (methods). Scenarios with solid black bars at bottom are not significantly different from *fire and low CO_2* ensemble scenario, based on a $p > 0.1$ cutoff using a paired T-test over all sites. Dashed bars are the mean of DMM point scores (scenario score) and solid bars are the median of DMM point scores. Boxes denote interquartile range while whiskers denote minimum and maximum DMM point scores with an experiment scenario.

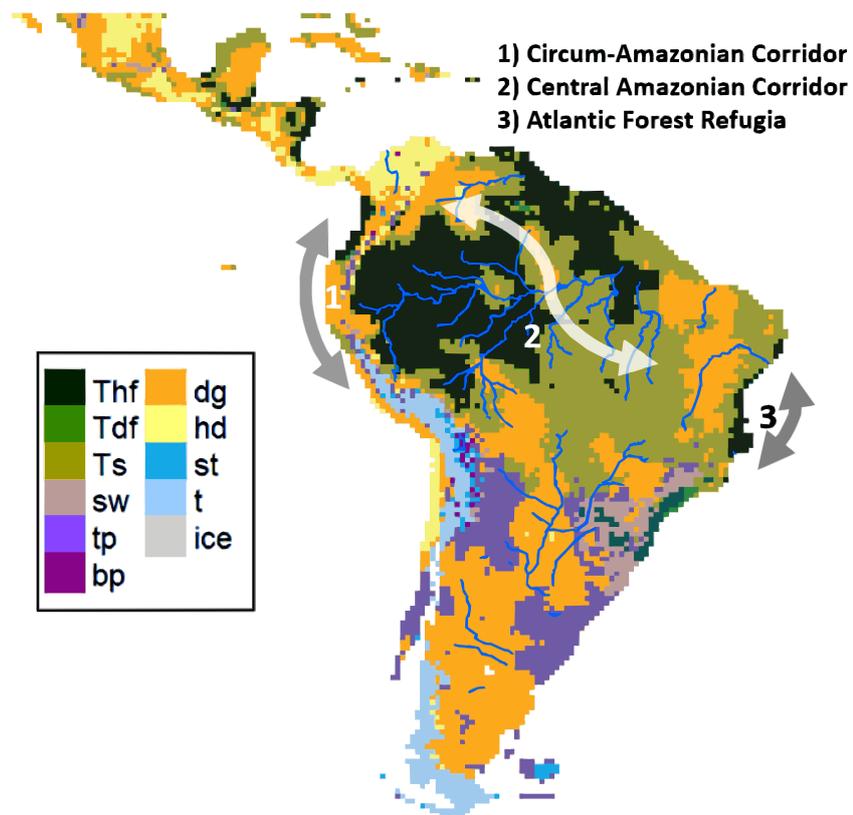


Figure 3: Identification of reconstructed biogeographical formations in the ensemble *fire and low CO₂* scenario. Corridors are hypothesized past open vegetation formations that are theorized to have facilitated dispersal during glacial periods, while refugia are stable forested regions that remained robust against past climatic change.

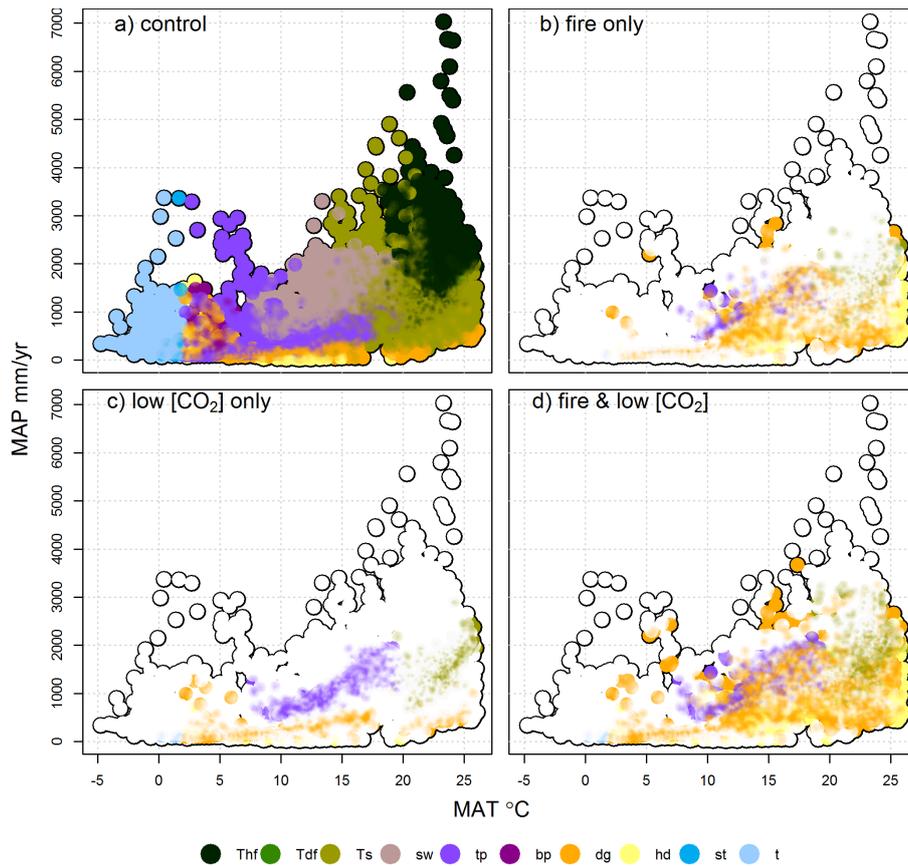


Figure 4: The effects of fire and low CO_2 on climate-vegetation relationships. Whitaker plots showing biomes in the climate space of mean annual temperature (MAT) and mean annual precipitation (MAP) for four scenarios in the fire/ CO_2 factorial experiment. All of biome space is shown for the control scenario (a), while for *fire only* (b), *low CO_2 only* (c), and *fire and low CO_2* (d), points that are identical to control scenario are left white while coloured points represent shifts due to fire and low CO_2 .

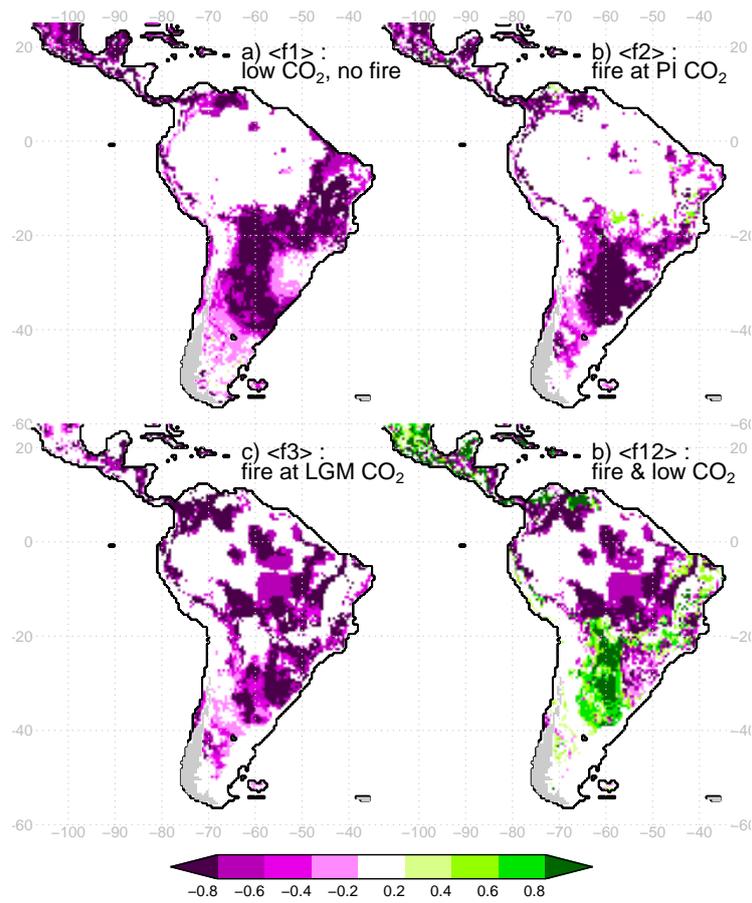


Figure 5: The individual and interactive effects of fire and CO₂ on tree cover. Results from the Stein-Alpert Decomposition showing the responses of tree cover to the a) direct impact of CO₂, b) direct impact of fire c) combined impacts of fire and CO₂, and d) purely interactive impacts of CO₂. Purple regions show losses in tree cover due to a specific factor while green shows increases.

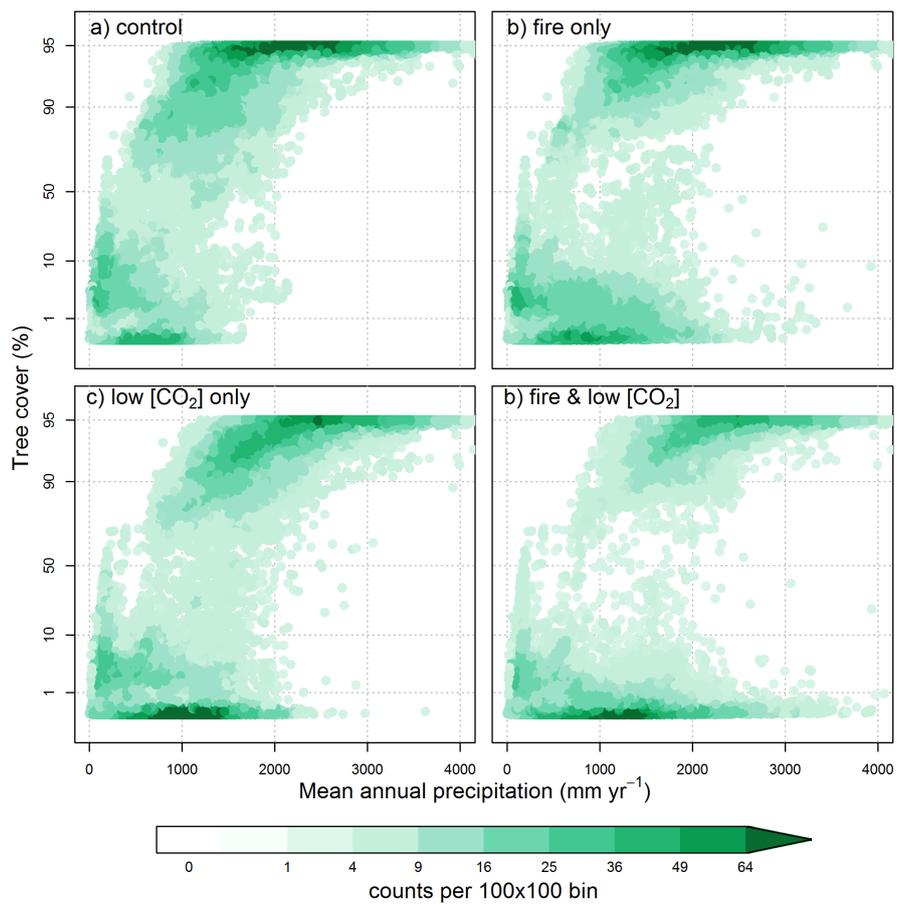


Figure 6: The effects of fire and low CO₂ on the bimodality of tree cover. Fractional tree cover against mean annual precipitation aggregated by runs driven by all four AOGCM LGM reconstructions to understand how fire and CO₂ mediate the moisture-tree cover relationship.

694 **Methods-Only References**

695 **References**

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817 **Extended Data**

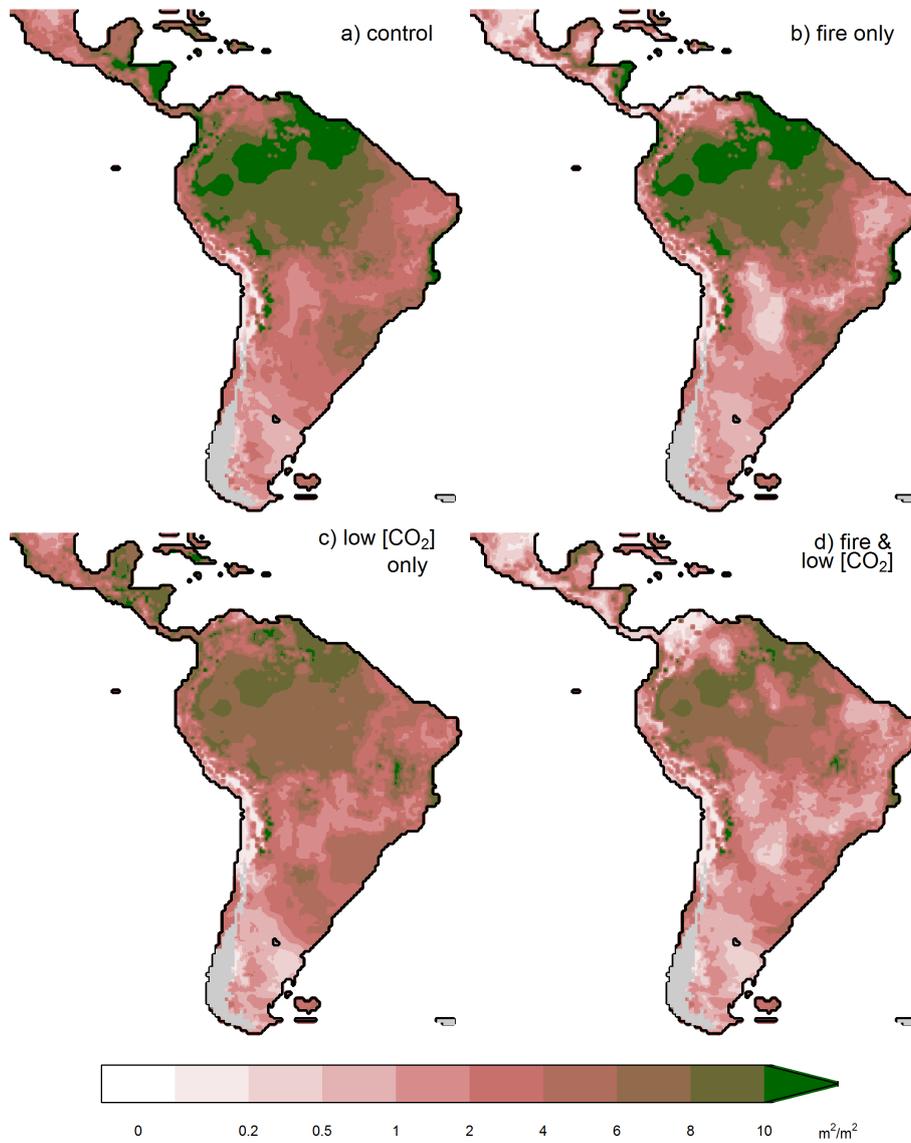


Figure 1: Canopy density (leaf area index) distributions for the ensemble factorial experiment in dimensionless units (m^2/m^2).

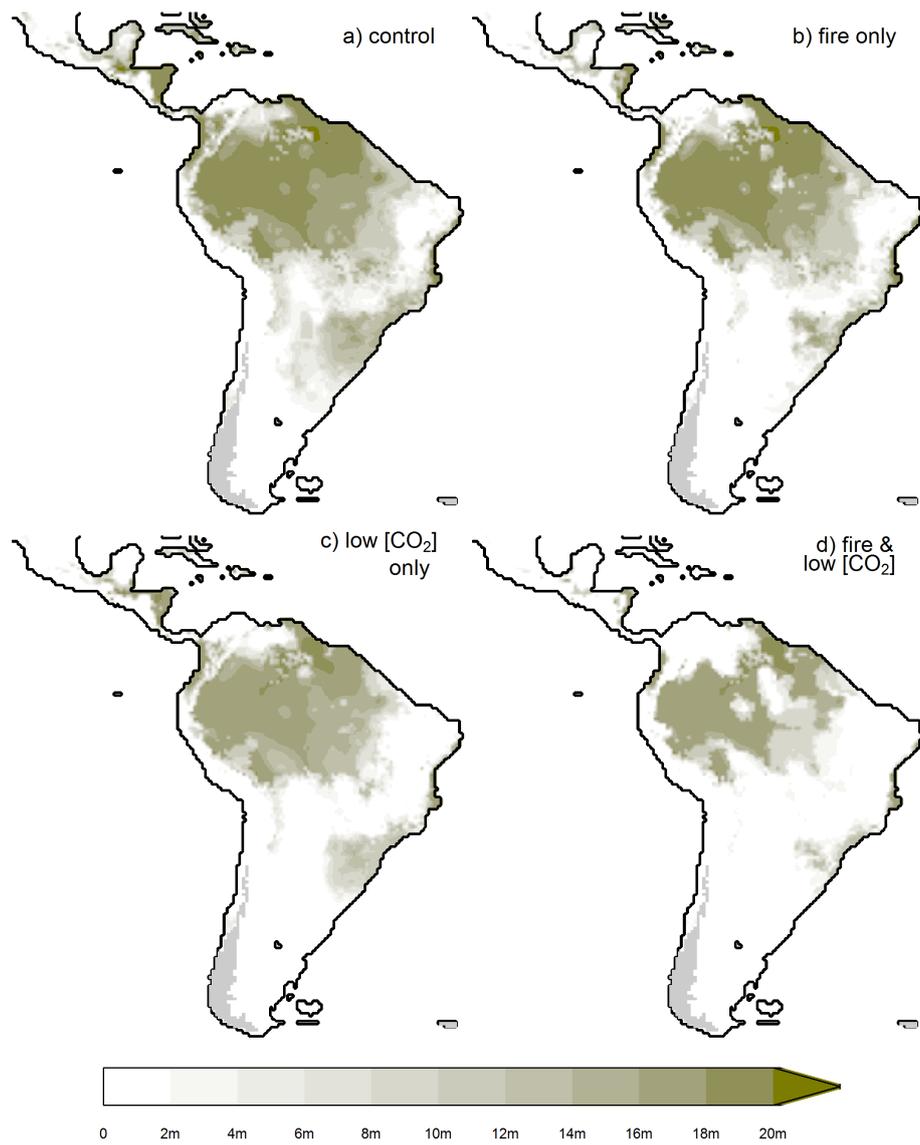


Figure 2: Canopy height (metres) distributions for the ensemble factorial experiment.

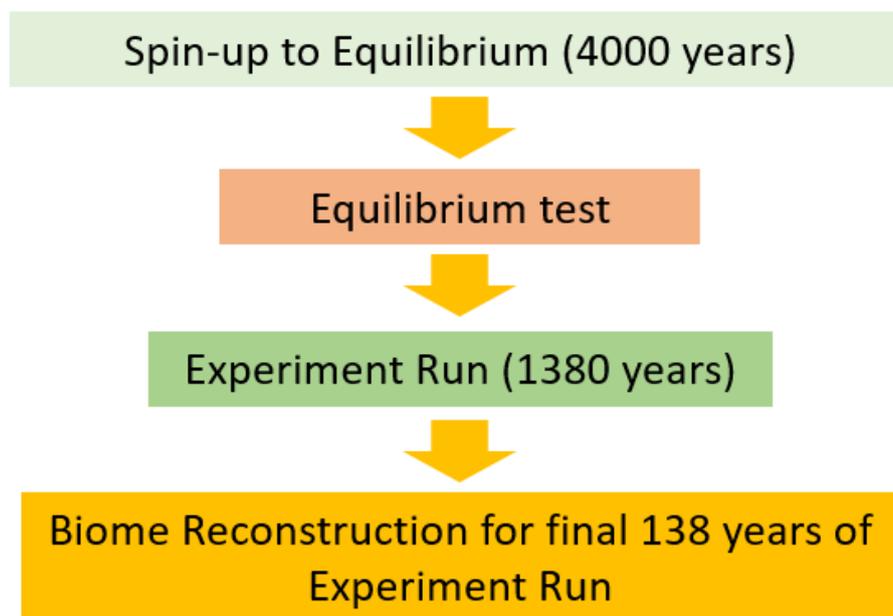


Figure 3: Flow of model protocol from spin-up to biome assignment for each factorial experiment run (LGM climate reconstruction + factorial experiment conditions)

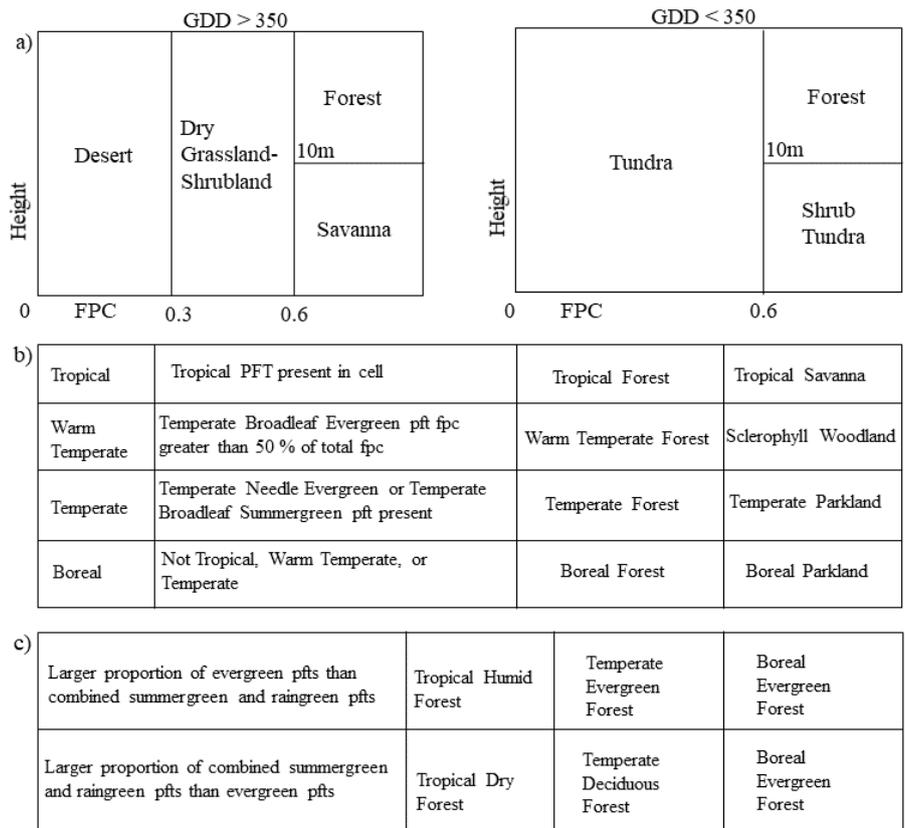


Figure 4: Diagram representing the biome assignment scheme. a) Division of cold and warm-hot biomes according to GDD and general organization of biomes according by fpc and height. b) Classification into more specific biomes by presence and dominance of pfts. c) Further classification of forests into seasonal and evergreen categories based on pft proportions.

Site Name	Latitude	Longitude	Country	Pollen Biome	Reference
Lake Patsucuarco	19.6	-101.58	Mexico	9	[35]
Chalco Lake	19.5	-99	Mexico	9	[12]
Lake Texcoco	19.4	-99	Mexico	9	[13]
Lake Quexil	16.3	-89.9	Guatemala	9	[21]
El Valle	8.43	-79.8	Panama	8	[28]
La Chonta	8	-82	Costa Rica	9	[17]
Fuquene II	5.45	-73.77	Colombia	3	[33]
Agua Blanca	5	-74.45	Colombia	12	[14]
Herrera	5	-73.9	Colombia	3	[34]
El Pinal	4.1	-70.4	Colombia	8	[5]
Timbio	2.4	-76.6	Colombia	3	[36]
Lagoa Das Patas	0.26	-66.7	Brazil	1	[26]
Lake Pata	0.26	-66.1	Brazil	1	[5]
Lagoa de Caco	-2.97	-43.3	Brazil	8	[19][17]
GeoB 3104-1	-3.67	-37.7	Ocean	11	[8]
Ciudad Universitaria	-4.75	-74.2	Colombia	3	[31]
Serra Sul Carajas	-5	-49.5	Brazil	8	[15]
Katira	-9	-63	Brazil	8	[32]
Lago do Saci	-9.1	-56.3	Brazil	8	[18]
Laguna Junin	-11	-76.2	Peru	12	[16]
Laguna Bella Vista	-13.6	-61.56	Bolivia	3	[22]
Consuelo	-13.95	-68.9	Peru	8	[30]
Chaplin	-14.5	-61.1	Bolivia	12	[4]
Aguas Emendadas	-15	-47.6	Brazil	12	[2]
Titicaca	-16.1	-69.2	Bolivia/Peru	12	[24]
Lake Huinamamarca	-16.5	-69	Bolivia	12	[25]
Crominia	-17.3	-49.4	Brazil	8	[29]
Wasa Mayu	-17.54	-65.9	Bolivia	12	[14]
Siberia	-17.8	-64.7	Bolivia	12	[25]
Salitre	-19	-46.8	Brazil	12	[18]
Serra Negra	-18.95	-46.85	Brazil	8	[26]
GeoB 3229-2	-19.63	-38.7	Brazil	12	[8]
Salar de Uyuni	-20	-68	Bolivia	12	[11]
Catas Altas	-20.1	-43.4	Brazil	12	[4]
GeoB 3202-1	-21.6	-39.9	Brazil	12	[7]
Morro de Itapeva	-22.8	-45.5	Brazil	12	[3]
Colonia	-23.9	-46.7	Brazil	12	[20]
Curucutu	-23.9	-46.7	Brazil	12	[27]
Volta Velha	-26.1	-48.6	Brazil	12	[6]
Cambara Sol	-29.1	-50.1	Brazil	12	[9]
Sao Francisco	-29.6	-55.3	Brazil	12	[10]

Table 1: List of original palynological studies used in conjunction with meta-analyses by Marchant et al. [28] and Mayle et al.[23] for $18\,000 \pm 1000$ ^{14}C yr BP. 1:tropical humid forest, 2:tropical dry forest, 3: warm temperate forest, 4: temperate evergreen forest, 5: temperate deciduous forest, 6: boreal evergreen forest, 7: boreal deciduous forest, 8: tropical savanna, 9: sclerophyll woodland, 10: temperate parkland, 11: boreal parkland, 12: dry grass/shrubland 13: hot desert, 14: shrub tundra, 15: tundra

Biome	Dense	Sparse	Tall	Short	Hot	Cold	Seasonal	Evergreen
Thf	1	0	1	0	1	0	0	1
Tdf	1	0	1	0	1	0	1	0
wtf	1	0	1	0	2/3	1/3	1/2	1/2
tef	1	0	1	0	1/3	2/3	0	1
tdf	1	0	1	0	1/3	2/3	1	0
bef	1	0	1	0	0	1	0	1
bdf	1	0	1	0	0	1	1	0
Ts	2/3	1/3	0	1	1	0	2/3	1/3
sw	2/3	1/3	0	1	2/3	1/3	1/2	1/2
tp	2/3	1/3	0	1	1/3	2/3	1/2	1/2
bp	2/3	1/3	0	1	0	1	1/2	1/2
g	1/3	2/3	0	1	1	0	1/2	1/2
d	0	1	0	1	1	0	1/2	1/2
st	1/3	2/3	0	1	0	1	1/2	1/2
t	0	1	0	1	0	1	1/2	1/2

Table 2: Affinity matrix for LPX biomes to compute ‘distance’ between biomes in trait space. Thf = Tropical humid forest, Tdf = Tropical dry forest, wtf = warm temperate forest, tef = temperate evergreen forest, tdf = temperate deciduous forest, bef = boreal evergreen forest, bdf = boreal deciduous forest, Ts = Tropical savanna, sw = sclerophyll woodland, tp = temperate parkland, bp = boreal parkland, g = dry grass/shrubland, d = desert, st = shrub tundra, t = tundra

Pollen Reconstructed Biomes	Model Assigned Biomes
Tropical Rainforest	Tropical Humid Forest
Tropical Seasonal Forest	Tropical Dry Forest
Cerrado	Tropical Savanna
Caatinga Steppe	Dry Grass-Shrublands
Desert	Hot Desert

Table 3: Correspondence legend between pollen reconstructed and model assigned biomes.