

1 **Assembly processes lead to divergent soil fungal communities within and among**
2 **twelve forest ecosystems along a latitudinal gradient**

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41

42

43 **Abstract**

44 Latitudinal gradients provide opportunities to better understand soil fungal community
45 assembly and its relationship with vegetation, climate, soil and ecosystem function. We
46 quantified the relative importance of stochastic and deterministic processes in structuring
47 soil fungal communities using patterns of community dissimilarity observed within and
48 between twelve natural forests. The results revealed that whole fungal communities and
49 communities of arbuscular and ectomycorrhizal fungi consistently exhibited divergent
50 patterns but with less divergence for ectomycorrhizal fungi at most sites. Within those
51 forests, no clear relationships were observed between the degree of divergence within
52 fungal and plant communities. When comparing communities at larger spatial scales,
53 among the twelve forests, we observed distinct separation in all three fungal groups
54 among tropical, subtropical and temperate biomes. Soil fungal β -diversity patterns
55 between forests were greater when comparing forests exhibiting high habitat turnover,
56 with these patterns being driven to a greater extent in each fungal group by temperature,
57 soil pH, soil carbon and plant community composition than by geographic distance. Taken
58 together, although large-scale community turnover could be attributed to specific
59 environmental drivers, strong divergence during community assembly in forest soils at
60 local scales limits the predictability of fungal community assembly outcomes.

61

62 **Keywords:** β -diversity; community assembly; forest; fungi; latitudinal gradient; neutral
63 model

64

65 **Introduction**

66 Understanding community assembly patterns and the processes resulting in these patterns
67 is critical to gaining a more mechanistic understanding of biodiversity maintenance,
68 community stability and ecosystem functioning (Chase, 2010; Nemergut *et al.*, 2013).
69 Stochastic (neutral) and deterministic (niche-based) processes are both important forces in
70 determining the outcomes of community assembly for macro-organisms (Fargione *et al.*,
71 2003; Kraft *et al.*, 2008; Ellwood *et al.*, 2009). Neutral theory assumes that many natural
72 community assembly patterns can be generated solely by considering ecological drift,
73 resulting in communities that are dependent on dispersal (immigration) rather than
74 adaptation to their habitats (Hubbell, 2001; Alonso *et al.*, 2006). Niche-based theory,
75 however, states that deterministic factors, such as biotic interactions, species traits, and
76 environmental variables, modulate the local community. When deterministic factors are of
77 greater importance, local communities assembling under similar environmental conditions
78 should converge upon a common composition, while divergence among local
79 communities should be observed when environmental conditions differ (Chase, 2003;
80 Zhou *et al.*, 2013).

81 Soil microorganisms, especially fungi, have remarkable biodiversity on Earth and
82 play crucial roles in biogeochemical cycling and ecosystem functioning (Philippot *et al.*,
83 2013; Tedersoo *et al.*, 2014). Mycorrhizal fungi make up important functional guilds
84 within soil fungal communities, forming mutualistic associations with 80% of land plants
85 and obtaining carbon from hosts in exchange for mineral nutrients (Smith & Read, 2008).
86 Therefore, these fungi can greatly affect plant productivity, diversity, and ecosystem
87 processes (van der Heijden *et al.*, 1998, 2015; Mayor *et al.*, 2015). Progress in molecular
88 techniques to efficiently characterise fungal communities had resulted in researchers
89 paying increasing attention to understanding the relative contributions of determinism and
90 neutrality for fungal community assembly (*i.e.*, Gao *et al.*, 2015; Powell *et al.*, 2015;
91 Alzarhani *et al.*, 2019). Previous studies have demonstrated that the relative role of
92 deterministic and stochastic processes in structuring soil fungal community assembly is
93 highly dependent on spatial scale (Caruso *et al.*, 2012; Kivlin *et al.*, 2014; Schroter *et al.*,
94 2019). However, the relative importance of stochastic and deterministic processes in
95 structuring fungal communities in forest soils at wide-latitude gradient remains largely
96 unknown.

97 Exploring the fungal community turnover (β -diversity) can lead to insight into the

98 mechanisms generating and maintaining fungal biodiversity in ecosystems (Ettema &
99 Wardle, 2002; Beck *et al.*, 2015). These studies are typically performed at large
100 geographic scales, but with different environmental foci and differing results. Soil fungal
101 communities across Scotland were more strongly affected by deterministic processes in
102 systems with less disturbance (bogs, moors, and semi-natural grasslands) than in more
103 intensive land use systems such as managed grasslands and arable systems (Powell *et al.*,
104 2015). Climatic factors, such as mean annual temperature (MAT) and precipitation (MAP)
105 are often observed to be important predictors of soil fungal β -diversity (Tedersoo *et al.*,
106 2014). Temperature was identified as a key factor influencing soil fungal β -diversity
107 across a large-scale latitudinal gradient in China and USA (*e.g.*, Shi *et al.*, 2014; Zhou *et*
108 *al.*, 2016). Plants contribute organic carbon to fungi, which benefit the plants via
109 decomposition or direct nutrients supply in symbiotic relationships. However, the
110 relationships between soil fungal β -diversity and that of the vegetation communities were
111 rarely considered.

112 Using Illumina Miseq high-throughput sequencing, we examined soil fungal
113 communities within twelve Chinese forests situated throughout eastern China, with the
114 largest distance among forests being more than 4 000 km (Fig. S1). To assess the relative
115 importance of stochastic and deterministic factors for soil fungal community assembly in
116 these forests, we employed a null model approach to generate neutral predictions, which
117 were subsequently compared to the observed β -diversity. This analysis reveals three
118 possible outcomes: the observed β -diversity not differing from the neutral prediction
119 (neutral assembly overwhelms determinism) or the observed β -diversity being larger
120 (divergent assembly) or smaller (convergent assembly) than the neutral prediction. Our
121 first aim was to identify these patterns in soil fungal community assembly and their
122 relationships with that of corresponding plant communities. Second, we estimated the
123 relative importance of habitat turnover based on soil, plant, and climatic parameters as
124 drivers of fungal β -diversity (community variation). We hypothesized that the balance
125 between deterministic and stochastic community assembly processes within individual
126 forests and their relationship with assembly processes occurring within plant communities
127 would differ depending on whether entire fungal communities or specific symbiotic
128 guilds were assessed. Finally, we assessed the relative importance of climate, soil and
129 vegetation properties for driving community turnover among soil fungal communities in
130 forests across the entire gradient.

131

132 **Materials and methods**

133 **Sites and sampling**

134 This study was conducted in 12 permanent forest sites established by the Chinese Forest
135 Biodiversity Monitoring Network (CForBio), ranging in latitude from 21.6° N to 50.8° N
136 in China (Fig. S1). The experimental forests were described previously by Ji *et al.* (2019).
137 Briefly, a total of six forest types include tropical rain forest (TRF), subtropical evergreen
138 broad-leaf forest (SEF), mixed evergreen broad-leaved and deciduous broad-leaved forest
139 (MEDF), warm-temperate deciduous broad-leaved forest (WTDF), temperate
140 broad-leaved Korean pine forest (TPF), and cold temperate monsoon coniferous forest
141 (CTCF) were included in this study. During June to October in 2014, 20 plots (20 m × 20
142 m) each spaced more than 45 m (mean = 247 m) apart were established in each site (Ji *et*
143 *al.*, 2019). In each plot, ten soil cores (3.5 cm diameter, 10 cm depth) were randomly
144 collected, bulked and subsampled to form a composite sample, which resulted in a total of
145 240 composite samples in the entire study. Samples were transported on ice to the
146 laboratory and sieved through a 2-mm sieve to remove roots and debris. One portion of
147 the soil samples was stored at −80°C until DNA extraction; another portion was air-dried
148 for analyses of physicochemical soil properties. In the same quadrats that the soil samples
149 were collected from, plant communities were assessed and characterised (Ji *et al.*, 2019),
150 and species were classified as arbuscular mycorrhizal, ectomycorrhizal, ericoid
151 mycorrhizal and non-mycorrhizal based on the published literature (Wang & Qiu, 2006;
152 Maherali *et al.*, 2016). Soil, plant, and climatic variables in each study site are shown in
153 Table S1, Table S2, and Table S3.

154

155 **Soil and climatic factors**

156 Soil pH, total carbon (TC), total nitrogen (TN), total phosphorus (TP), C:N, and N:P
157 ratios were described in detail by Ji *et al.* (2019). Latitude, longitude, and plant data
158 (basal area, richness and community composition) of the study sites were provided by the
159 CForBio organization (successively obtain from the colleagues as mentioned in the
160 Acknowledgments section around a period in 2015). Mean annual temperature (MAT)
161 and mean annual precipitation (MAP) were obtained from the WorldClim database
162 (www.worldclim.org) with a resolution of 2.5 min (Ji *et al.*, 2019).

163

164 **Molecular analysis**

165 Molecular analyses were performed independently on the 20 soil samples collected from
166 each of the 12 sites. Genomic DNA was extracted from 0.25 g of frozen soil using a
167 PowerSoil DNA isolation kit (MoBio Laboratories, Inc. USA) according to the
168 manufacturer's instructions. The DNA was diluted 10-fold in water used in all
169 downstream PCR experiments. The primers and PCR conditions for amplifying total
170 fungal (including EM fungi – only 1% of sequence reads belonged to AM fungal; ITS)
171 and AM fungi (18S) DNA are shown in Table S4. A barcode was added to the 5' end of
172 the reverse primers (*i.e.*, ITS2 and AMDGR) to identify samples when demultiplexing
173 sequence reads. The PCR products were purified using a PCR product gel purification kit
174 (Axygen, Union City, CA, USA). The yields of purified PCR products were measured
175 using a TBS 380 Fluorescence Spectrophotometer (Promega, USA), and 50 ng of DNA
176 from each of the 240 purified PCR products were pooled and adjusted to 10 ng μL^{-1} . The
177 pooled DNA was subjected to sequencing on an Illumina MiSeq PE250 platform for
178 sequencing using the paired end (2×250 base pair (bp)) option at the Environmental
179 Genome Platform in the Chengdu Institute of Biology, Chinese Academy of Sciences,
180 China.

181

182 **Sequence processing**

183 Clean sequences were obtained from raw sequences after quality control using
184 'Quantitative Insights into Microbial Ecology' (QIIME v.1.7.0, Caporaso *et al.*, 2010).
185 Quality control removed low quality reads with no valid primer sequence or barcode
186 sequence, containing ambiguous bases, or with an average quality score < 20 . For ITS
187 reads, the ITS1 region of each high-quality sequence was extracted using the fungal ITSx
188 software package (Bengtsson-Palme *et al.*, 2013). Chimeric sequences were detected
189 using the 'chimera.uchime' command in Mothur 1.31.2 (Schloss, 2009) and removed
190 from further analysis. The non-chimeric sequences were grouped into different
191 operational taxonomic units (OTUs) at a 97% similarity level based on the UPARSE
192 pipeline using the USEARCH v8.0 after dereplication and singleton exclusion (Edgar,
193 2013).

194 The sequence number per sample was normalized to that sample with the fewest

195 reads (5290 and 1509 for the ITS1 and 18S sequences, respectively) using the
196 ‘sub.sample’ command in Mothur to eliminate the influence of different read numbers on
197 the analysis of fungal communities. Representative sequences from OTUs were selected
198 through the command of ‘get.oturep’ and were identified by a basic local alignment
199 search tool (BLAST) search against the international nucleotide sequence databases
200 collaboration (INSDc) and UNITE database (Kõljalg *et al.*, 2013) for ITS1 sequences and
201 the MaarjAM 18S rRNA gene database (Õpik *et al.*, 2010) for 18S sequences (accessed on
202 Apr. 9th, 2016 and on Sept. 28th, 2016, respectively), using an E value less than $1e^{-50}$ as a
203 threshold. Furthermore, the ITS1 OTUs were assigned to functional guilds following the
204 methods of Tedersoo *et al.* (2014) and using the information in FunGuild (Nguyen *et al.*,
205 2016; accessed on Nov. 24th, 2016). The results of ITS sequences were described in detail
206 by Ji *et al.* (2019). Briefly, of the obtained 14 911 ITS OTUs (1 267 912 reads), a subset
207 of 9 883 ITS OTUs (reads ≥ 3) was extracted to form the total fungal community, and all
208 identified 2 709 EM fungal OTUs (431 138 reads, 34.0% of all fungal sequences) were
209 used for subsequent analyses in the current study. We obtained a total of 4 517 918
210 high-quality 18S rRNA gene sequences, which were clustered into 638 18S OTUs. After
211 normalization, we obtained 592 18S OTUs (357 633 reads) which all matched with
212 known AM fungal sequences from the MaarjAM database. Three (*i.e.*, total, AM and EM)
213 fungal community matrices (Table S5) were used in all analyses, including generation of
214 null models based on the neutral model: total fungi (all ITS1 OTUs), EM fungi (ITS1
215 OTUs assigned to EM taxa) and AM fungi (18S OTUs assigned to Glomeromycota). The
216 DNA sequences of the ITS and 18S rRNA gene amplicons were deposited in the National
217 Center for Biotechnology Information under accession nos. LT986405–LT998319 and
218 MK352490–MK353081, respectively. The raw sequencing data (fastq files) have been
219 submitted to the Environmental Genomic Cloud (<http://egcloud.cib.cn>) with the sample
220 nos. XX to XX under the project number XX.

221

222 **Neutral model simulation and comparison analysis**

223 The approach described in Etienne (2007) was used to estimate immigration and diversity
224 parameters from a neutral model, and these estimates were used to simulate communities
225 assembled under only neutral processes. First, the fungal (total, AM, and EM)
226 sample-OTU data obtained from each site was used to estimate the theta (Θ , diversity
227 index) and immigration rate (I) parameters of the neutral model. The formula was used

228 for multiple samples to estimate neutral parameters using the PARI/GP code included in
229 Etienne (2007). Parameters were estimated for the total, AM, and EM fungal communities
230 separately from each of the 12 forest sites. Using the estimated parameters, the PARI/GP
231 function ‘urn2.gp’ was performed to create 100 communities of matching size. The
232 detailed processes conducted for simulating the neutral community were described by
233 Maaß *et al.* (2014). The β -diversity of a real set of local assemblages (observed fungal
234 communities in the sample-OTU matrices) were compared to the frequency distribution
235 of β -diversities obtained from 100 simulations of the corresponding fungal community
236 under the neutral model.

237 The PARI/GP output files were imported into R (R Core Team, 2015) to calculate
238 pairwise Sørensen community dissimilarities (β -diversity) among all communities within
239 each simulation (Powell *et al.*, 2015). The potential importance of stochasticity was
240 evaluated using standardized effect size (SES) calculated as the difference in observed
241 statistics (median, interquartile range [IQR] and interdecile range [IDR]) associated with
242 dissimilarity distributions and corresponding expectations produced by the neutral models
243 using the following formula: $(\text{estimate}_{\text{obs.}} - \text{mean of estimate}_{\text{exp.}}) / \text{standard deviation of}$
244 $\text{the estimate}_{\text{exp.}}$, where $\text{estimate}_{\text{obs.}}$ denotes the estimate (median/IQR/IDR) associated with
245 the observed distribution of Sørensen dissimilarities and $\text{estimate}_{\text{exp.}}$ denotes the estimate
246 (median/IQR/IDR) associated with the distribution of Sørensen dissimilarities for each of
247 the individual simulated communities. If the SES value was not significantly ($P < 0.05$)
248 different from zero, the community assembly was interpreted as being driven primarily by
249 neutral process; otherwise, the community assembly was regarded as deterministic. In the
250 case of a significant difference in the SES from zero, a positive effect size indicated that
251 the estimate was higher than predicted (divergence), while a negative effect size indicated
252 that the estimate was lower than predicted (convergence) under the neutral hypothesis.
253 The SES values were compared amongst the three fungal groups within each site and the
254 mean values of all 12 sites. The same analyses were conducted for the total plant
255 community, as well as the AM and EM plant communities.

256

257 **Estimation of community and habitat turnover**

258 By applying the approach described in Ranjard *et al.* (2013), we estimated relationships
259 between fungal β -diversity along the latitudinal gradient and in relation to habitat
260 turnover among forests. Briefly, pairwise fungal β -diversities (as Sørensen index) were

261 calculated using the ‘labdsv’ package (Roberts, 2016) in R. To compare with latitude,
262 β -diversity was estimated as the median Sørensen dissimilarity among local communities
263 within each of the 12 forests. To compare with habitat turnover among forests, β -diversity
264 was estimated as the Sørensen dissimilarity between each pair of forests after aggregating
265 each OTU-sample matrix so that each sample represented the sum of OTU frequencies in
266 each of the 12 forests. Habitat turnover (environmental dissimilarity, Ed) was calculated
267 from the Euclidean distance between sites (‘dist’ function in R) based on the site-level soil,
268 plant, and climatic factors using the formula $Ed = 1 - Euc_d / Euc_{max}$, in which Euc_d is the
269 Euclidean distance and Euc_{max} is the maximum distance between sites in the integrated
270 data matrix including soil, plant, and climate properties.

271

272 **Statistical analysis**

273 Multiple comparisons of group means among the 12 forest sites were carried out with
274 Tukey’s honestly significant difference (HSD) test after one-way analysis of variance
275 (ANOVA) or pairwise comparisons after the nonparametric Kruskal–Wallis test where
276 data did not satisfy homogeneity of variance, at $P < 0.05$, which indicated a significant
277 effect of site on environmental factors (soil, plant, and climate), the observed β -diversity,
278 and SES values. Similar analyses of one-way ANOVA followed by multiple comparisons
279 were also conducted for the observed β -diversity amongst different fungal groups and
280 different climate zones. Fungal community compositions were ordinated using nonmetric
281 multidimensional scaling (NMDS) with dissimilarity matrices using the ‘metaMDS’
282 function in the ‘vegan’ package (Oksanen *et al.*, 2013). We used the ‘lmodel2’ package
283 (Legendre, 2011) to perform type II linear regression (ordinary least squares) on the
284 relationships between the observed fungal β -diversity and latitude and habitat turnover
285 (Ed) after estimating these at the site level. To evaluate the effects of climate zone
286 (tropical, subtropical, and temperate) and site (both all data and separate data sets of
287 tropical, subtropical, and temperate) on total, AM, and EM fungal β -diversities,
288 permutational multivariate analysis of variance (PerMANOVA) was conducted based on
289 distance matrices (Sørensen dissimilarity) using the ‘adonis’ function in the ‘vegan’
290 package with 999 permutations. To explore the independent influence of dispersal
291 limitation on the fungal community assembly, partial Mantel tests were further performed
292 to analyze the correlation between fungal communities and geographic distance after
293 excluding the effects of significant abiotic (soil and climate related) and biotic (plant

294 related) factors using the ‘ecodist’ package (Goslee & Urban, 2007). In turn, partial
295 Mantel tests were carried out to explore the relationships between the fungal community
296 and abiotic/biotic factors after excluding the influence of geographic distance. All
297 analyses were conducted in R 3.0.2 (R Core Team, 2015).

298

299 **Results**

300 **Divergence dominated fungal and plant community assembly outcomes within** 301 **forests**

302 Distributions of observed community dissimilarities (β -diversity) among the total, AM,
303 and EM fungal communities and among total, AM and EM plant communities are
304 presented as probability densities (Fig. S2). Their distributions indicate that substantial
305 variation was observed among communities for all groups across many sites, with some
306 of the distributions being bimodal (indicating convergence among some samples within a
307 site). SES values for all these fungal groups were consistently greater than zero (Table S6),
308 indicating significant divergence. The highest fungal SES values within a site were
309 generally associated with total or AM fungal communities, and the lowest SES values
310 were always observed for the EM fungal communities except for site Baotianman (BTM,
311 Table S6). Estimates of *Theta* and *I* are provided in Table S7.

312 The results of null model analyses indicated that there was strong evidence for
313 divergent assembly among total, AM, and EM fungal communities in each of the three
314 forest biomes (tropical, subtropical, and temperate zones) based on deviations in observed
315 dispersion (Fig. 1b, c) and/or average dissimilarity (Fig. 1a) from the 100 simulations
316 under the null model. For the plant communities, not all SES estimates associated with
317 average dissimilarity differed from zero (Fig. 1a). However, dispersion among observed
318 communities relative to those simulated assuming was substantial in all cases (Fig. 1b, c),
319 reflecting the large range of dissimilarities observed at most sites (Fig. S2) and providing
320 evidence for divergent assembly.

321 We did not observe any correlations between SESs for fungal and plant communities
322 when comparing responses at individual sites (Table S6) or among biomes (Fig. 1). The
323 most important landscape-level environmental variables for explaining variation in SES
324 values for total fungi were temperature ($P = 0.009$), latitude ($P = 0.012$) and total plant
325 species richness ($P = 0.034$), with plant community composition ($P = 0.053$) and

326 precipitation ($P = 0.051$) being marginally nonsignificant (Table 1). For AM fungi, SES
327 values were significantly correlated with soil C:N ratio ($P = 0.020$) and marginally
328 nonsignificantly correlated with pH ($P = 0.065$) and plant community composition ($P =$
329 0.079) (Table 1). No environmental factors were correlated with EM fungal SES values
330 (Table 1). SES values for plant communities tended to be correlated more closely with
331 plant community variables, including plant species richness (all and AM plants), basal
332 area (AM plants), and community composition (EM plants).

333

334 **Environmental drivers of fungal β -diversity across forests**

335 The mean observed β -diversity of EM fungi was consistently and significantly greater
336 than that of total and AM fungi ($P < 0.0001$), but no significant β -diversity difference was
337 detected among the three plant groups ($P = 0.179$, Table S8). Fungal β -diversity was
338 significantly greater within temperate forests, followed by tropical forests and then
339 subtropical forests ($P < 0.0001$, Table S9). There was no significant difference in plant
340 β -diversity across tropical, subtropical, and temperate forests ($P = 0.764$, Table S9). Both
341 fungal and plant compositions were structured by climate (Figure 2, Table S10) and site
342 (Table S10), regardless of whether whole communities or AM/EM subsets were assessed.
343 We observed larger effects of climate zone and site on AM fungi ($R^2 = 0.280$) than on
344 total ($R^2 = 0.095$) and EM ($R^2 = 0.069$) fungi, based on larger R^2 values in PerMANOVA
345 analyses (Table S10). Among all groups of total, AM, and EM fungi, the effects of site on
346 β -diversities were larger (*i.e.*, greater R^2 values) in the subtropical (R^2 , 0.222, 0.403, 0.172,
347 respectively) zone as compared with in temperate (R^2 , 0.183, 0.247, 0.156, respectively)
348 and tropical (R^2 , 0.086, 0.266, 0.064, respectively) zones (Table S10). Variation explained
349 by climate zone and type was similar when comparing the total, AM, and EM plant
350 groups (Table S10).

351 At the sampling site level, we found that the observed AM fungal β -diversity was
352 positively correlated with latitude ($P < 0.0001$, Fig. 3a), which is the opposite pattern as
353 was observed for AM plant β -diversity ($P = 0.025$, Fig. 3c). The increase in AM fungal
354 β -diversity with latitude was mirrored by a decline in AM fungal richness with latitude (P
355 < 0.001 , Fig. S3). No significant correlation with latitude was observed for β -diversity
356 within the other fungal groups (Fig. 3a) despite total ($P = 0.004$) and EM ($P = 0.023$)
357 plant β -diversity exhibiting negative relationships with latitude (Fig. 3c). We observed the

358 highest β -diversity in soil EM fungal communities followed by total fungal communities
359 and then AM fungal communities (Fig. 3b). β -diversity in all fungal groups increased with
360 greater habitat turnover (*Ed*; *P* values < 0.05 for total and EM fungi, *P* = 0.063 for AM
361 fungi; Fig. 3b). The β -diversity for all plant groups were not significantly associated with
362 habitat turnover, which may be due to the limited overlap in plant composition among
363 plots (Fig. 3d).

364 Finally, we performed partial Mantels test to reveal relationships among geographic
365 distance and individual environmental factors on fungal community composition. Pure
366 geographic distance was not significantly correlated with any fungal community
367 groupings (Table 2). Soil pH and total C were significant variables explaining variation in
368 all fungal groups after accounting for geographic distance, plants and climatic factors,
369 while temperature and plant community composition also explained variation in all fungal
370 groups after accounting for other variables (Table 2). Variation in AM fungal community
371 composition was further explained by variation in AM plant basal area (Table 2).

372

373 **Discussion**

374 **Divergence-dominated fungal community assembly**

375 The relative importance of stochastic and deterministic processes in structuring a
376 microbial community is currently being debated (Zhou & Ning, 2017). Previous studies
377 documented that the assembly of microbial communities depends largely on the
378 deterministic process (niche-based) driven by contemporary environmental changes, such
379 as pH, temperature, precipitation, salinity and nutrients (Fierer & Jackson, 2006;
380 Lozupone & Knight, 2007; Wang *et al.*, 2016; Zhou *et al.*, 2016; Zhang *et al.*, 2019;
381 Zheng *et al.*, 2020). Microbial communities can also be driven by stochastic processes
382 such as geographical separation and dispersal limitation (Wang *et al.*, 2013). Compared
383 with most cases focusing on bacteria from the available microbial community studies, the
384 importance of stochastic processes in generating and maintaining fungal biodiversity is
385 rarely appreciated, and a few studies have addressed the importance of neutrality for these
386 communities in soil and roots (Caruso *et al.*, 2012; Schroter *et al.*, 2019).

387 Here, we observed evidence for assembly processes leading to strong divergence
388 among soil fungal communities within forests each distributed along a large
389 environmental gradient. This was the case for whole fungal communities and for subsets

390 consisting of AM and EM fungi, although divergence was less among communities of the
391 latter group. The degree of divergence at whole fungal communities was greater in more
392 diverse forests in warmer climates at lower latitudes, which is consistent with how each of
393 the forest biomes was ranked in terms of degree of divergence (tropical > subtropical >
394 temperate). Interpreting these differences is challenging due to the possibility of fungal
395 community shifts leading to differences in abundance of functional guilds among these
396 communities, so we mainly discuss the results of the AM and EM fungal community
397 subsets.

398 These results indicating a greater tendency toward divergence are consistent with
399 expectations under two scenarios: (1) environmental heterogeneity (*e.g.*, in soil properties,
400 vegetation and/or microclimate) within forests leads to strong niche-based assembly
401 within soil fungal communities or (2) strong biotic interactions lead to communities
402 diverging more than expected under similar environmental conditions. Powell and
403 Bennett (2016) observed similar results for AM fungal communities sampled in a global
404 survey (Davison *et al.*, 2015). In that analysis, divergence and neutrality were much more
405 frequent outcomes than convergence when trying to compare local communities sampled
406 from similar environments (roots of a single host species collected from multiple plants
407 within a single plot). Here, we observed greater dispersion among AM fungal
408 communities within sites in the higher latitude forests (greater NMDS area of temperate
409 and subtropical than that tropical forest sites; Fig. 2b), but this pattern was confounded by
410 the reduced AM fungal richness at higher latitudes. Similarly, the value associated with
411 the null model approach was revealed in that it detected stronger divergence, on average,
412 in temperate and subtropical forests than in tropical forests (Fig. 1a), although this was
413 independent of other attributes of those plant communities. Veresoglou *et al.* (2019) also
414 reported that latitudinal differences in AM fungal β -diversity were driven to a greater
415 extent by forest types than by latitude. We also observed stronger divergence in soils with
416 higher C-to-N ratios, where antagonistic interactions with decomposer fungi might be
417 more prevalent (Bunn *et al.*, 2019).

418 Less divergence was observed among EM fungal communities in the current study
419 than what was observed for AM fungi and whole fungal communities, which might be
420 attributed to relatively high host specificity for EM fungi compared with AM fungi
421 (Tedersoo *et al.*, 2008; Davison *et al.*, 2015; Wang *et al.*, 2019). If this was the case we
422 might expect a positive relationship between the degree of divergence and variation in

423 EM fungal host communities. However, we did not observe such a relationship. We did
424 not observe much difference in the degree of divergence among the three forest types
425 either. This result may reflect the limitations associated with trying to interpret assembly
426 processes in EM fungal communities from soil samples instead of root samples, where in
427 the latter there might be stronger evidence of niche-based assembly. Regardless, there is
428 some validity to interpreting these data as these samples do reflect the composition of EM
429 fungal communities available to colonise root samples in these environments.

430

431 **Drivers of soil fungal β -diversity in forests at large scales**

432 In addition to a greater prevalence of divergent community assembly for soil fungi within
433 each forest, we observed high levels of divergence among soil fungal communities across
434 the forests. This divergence could be partially explained by climate, vegetation and soil
435 properties. For the most part, the environmental drivers associated with fungal community
436 turnover were consistent among the three fungal groups, with soil C, plant community
437 composition, MAT and soil pH being the most important variables. These results are
438 consistent with other large scale studies of soil fungal diversity (*e.g.*, Tedersoo *et al.*, 2014;
439 Zhou *et al.*, 2016; Guo *et al.*, 2018).

440 However, we did observe that AM fungal β -diversity under these forests was lower
441 than the other groups and its relationship with habitat turnover overall was not as strong.
442 It is difficult to draw conclusions from these particular results given the possibility that
443 this is determined, in part or as a whole, by greater variation in the ITS region (used to
444 sample total fungi and EM fungi) than that of the 18S rRNA gene (used to sample AM
445 fungi). That said, when comparing compositional shifts in soil fungal communities among
446 the 12 forests, strengths of relationships with some individual variables tended to be
447 greater for AM fungi than for the other two groups. This difference was particularly large
448 for soil C, which is intriguing given speculation on the role that AM fungal-derived
449 proteins (*i.e.*, glomalin) play in determining the recalcitrance of soil C stocks (Rillig *et al.*,
450 2003). The AM fungal subset was also the only fungal group to exhibit a significant
451 relationship between composition and the basal area of probable host trees. This may
452 reflect the more generalist nature of AM fungal and plant associations, where frequency
453 of host plants may be a greater driver of composition than host identity, than in more
454 specialist EM fungal and plant associations.

455

456 **Conclusions**

457 Here we were able to demonstrate substantial divergence among soil fungal communities,
458 including mycorrhizal fungal communities both within and among forests at a large
459 geographic scale (> 4 000 km). Within forests, fungal communities were consistently
460 more divergent than expected under neutrality and exhibited patterns that appeared to
461 indicate that the relative strength of deterministic assembly processes differed among
462 fungi and plants in these forests. We did observe that the degree of divergence was related,
463 in part, to environmental variation when looking at whole fungal communities or those of
464 AM fungi. However, we were unable to detect patterns to explain the degree of
465 divergence among EM fungal communities despite the inclusion of variables that were
466 related to compositional shifts at large scales. Taken together, although soil fungal
467 community turnover at large scales could be attributed to specific environmental drivers,
468 strong divergence during fungal community assembly in forest soils at local scales limits
469 the predictability of fungal community assembly outcomes.

470

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481

482 **Author Contributions**

483 Y.Z., L.C. and L.-D.G. designed the experiments and analysed the data in consultation
484 with C.G., S.-S.J., H.-W.H., Z.H., J.-Z.H., and J.R.P. The field surveys, soil samplings,
485 and molecular work were conducted by Y.Z., N.-N.J., L.C., Y.-L.W., and C.G. The

486 manuscript was written by Y.Z. and J.R.P. with contributions from all co-authors.

487 Data Availability

488 The data that support the findings of this study are openly available on figshare at

489 <https://www.doi.org/10.6084/m9.figshare.13543046>.

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Figure legends

Figure 1. The co-occurrence (X and Y axes represent plant and fungi, respectively) of standardized effect size (SES) for observed community dissimilarities relative to 100 simulations based on a null model of neutral community assembly. Estimates were calculated in relation to central tendencies, from the median (**a**), and dispersion from the interquartile range (IQR, middle 50% of observations; **b**) and interdecile range (IDR, middle 80% of observations; **c**). Analyses were performed separately for total, arbuscular (AM) and ecto- (EM) mycorrhizal fungi and plants. The median (points) and 95% confidence interval (lines) of the estimates are presented for communities sampled from tropical, subtropical and temperate forest ecosystems. A reduction or increase in the central tendency or dispersion is evidence of community convergence or divergence from a common composition, respectively. Central tendency estimates (**a**) focus on shifts in community composition that tend to occur across all communities while dispersion estimates (**b, c**) focus on extreme shifts between pairs of communities.

Figure 2. Non-metric multidimensional scaling (NMDS) plot showing shifts in community composition for total (**a**), arbuscular mycorrhizal (AM, **b**), and ectomycorrhizal (EM, **c**) fungi, as well as total (**d**), AM (**e**), and EM (**f**) plants, across tropical, subtropical, and temperate forest ecosystems. Solid lines indicate the two-dimensional space that contains all observations within a forest ecosystem type.

Figure 3. Relationships of observed fungal (**a, b**) and plant (**c, d**) community dissimilarity (β -diversity) with latitude (**a, c**) and habitat turnover (environmental dissimilarity, *Ed*; **b, d**). To compare with latitude, β -diversity was estimated as the median Sørensen dissimilarity among local communities within each of the 12 forests. To compare with habitat turnover among forests, β -diversity was estimated as the Sørensen dissimilarity between each pair of forests after aggregating each OTU/species-sample matrix within each forest. Pairwise environmental dissimilarities were calculated based on Euclidean distances among site-level environmental variables. Solid lines indicate that the predicted relationship is significant ($P < 0.05$) or marginally nonsignificant ($0.05 \leq P < 0.10$) and dashed lines indicate that the relationship is not significant ($P \geq 0.10$). Relationships within each group are based on type II linear regression estimated using ordinary least squares and *P*-values are as follows: between total ($P = 0.573$), AM ($P < 0.0001$) and EM ($P = 0.372$) fungal β -diversity and latitude (**a**); between total ($P = 0.034$), AM ($P = 0.063$) and EM ($P = 0.043$) fungal β -diversity and *Ed* (**b**); between total ($P = 0.004$), AM ($P = 0.025$) and EM ($P = 0.023$) plant β -diversity and latitude (**c**); between total ($P = 0.220$), AM ($P = 0.346$) and EM ($P = 0.387$) plant β -diversity and *Ed* (**d**). AM: arbuscular mycorrhizal; EM: ectomycorrhizal.

Fig. 1

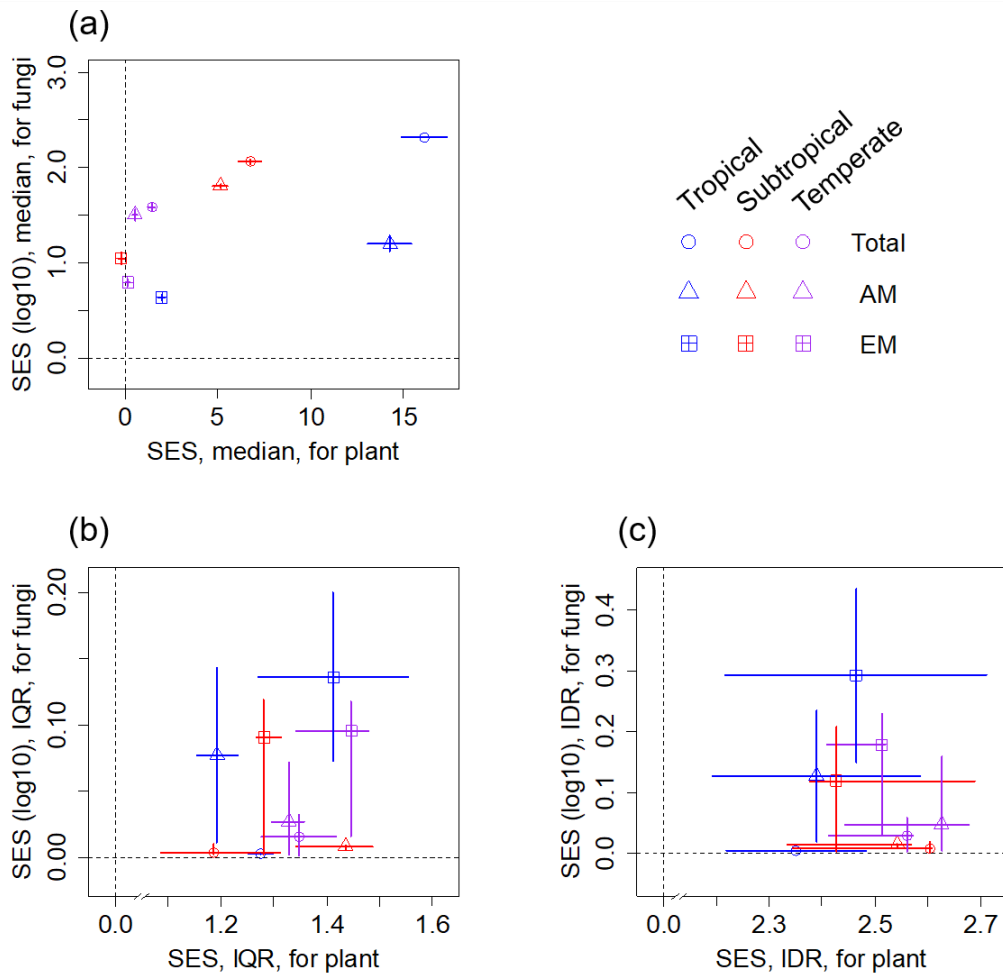


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Fig. 2

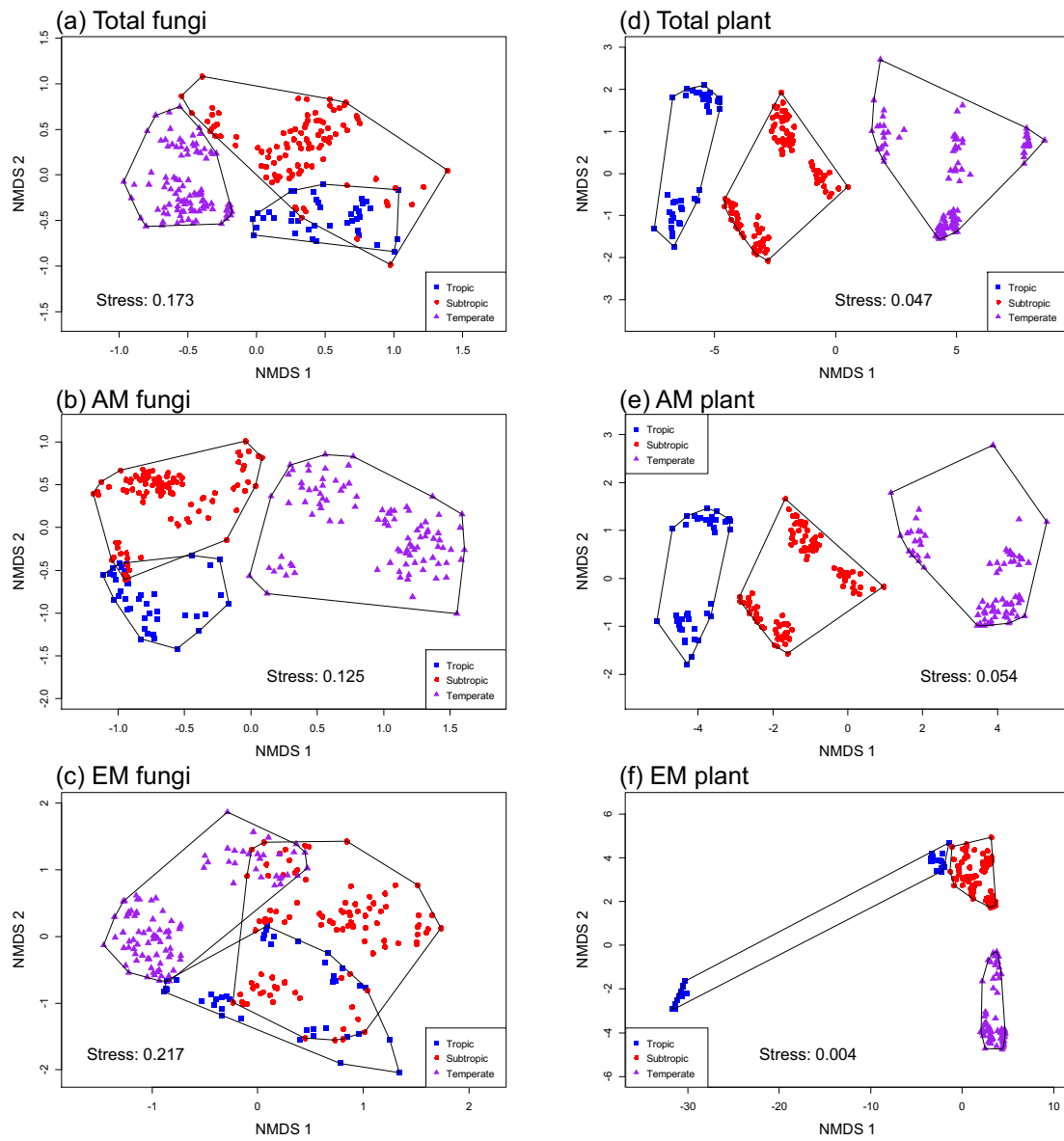


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Fig. 3

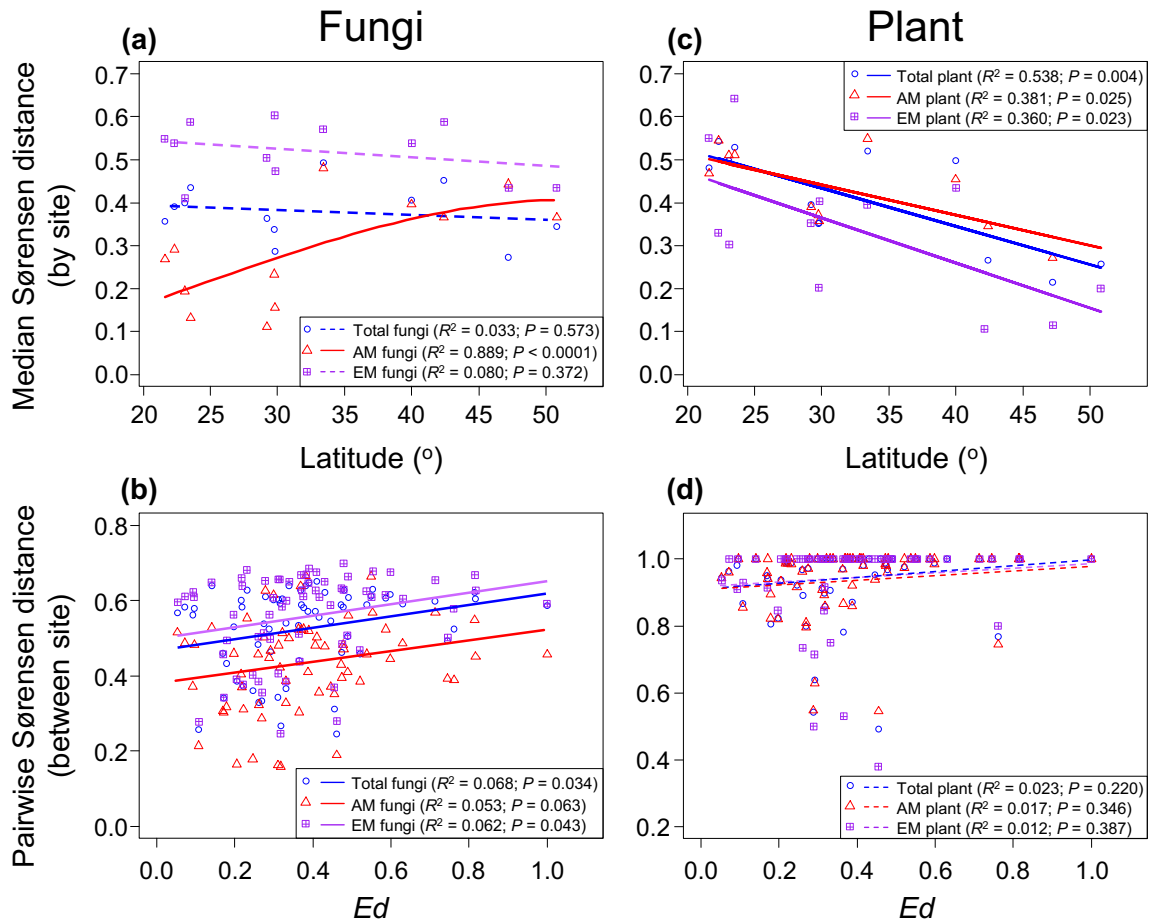


Figure 3. Relationships of observed fungal (**a**, **b**) and plant (**c**, **d**) community dissimilarity (β -diversity) with latitude (**a**, **c**) and habitat turnover (environmental dissimilarity, Ed ; **b**, **d**). To compare with latitude, β -diversity was estimated as the median Sørensen dissimilarity among local communities within each of the 12 forests. To compare with habitat turnover among forests, β -diversity was estimated as the Sørensen dissimilarity between each pair of forests after aggregating each OTU/species-sample matrix within each forest. Pairwise environmental dissimilarities were calculated based on Euclidean distances among site-level environmental variables. Solid lines indicate that the predicted relationship is significant ($P < 0.05$) or marginally nonsignificant ($0.05 \leq P < 0.10$) and dashed lines indicate that the relationship is not significant ($P \geq 0.10$). Relationships within each group are based on type II linear regression estimated using ordinary least squares and P -values are as follows: between total ($P = 0.573$), AM ($P < 0.0001$) and EM ($P = 0.372$) fungal β -diversity and latitude (**a**); between total ($P = 0.034$), AM ($P = 0.063$) and EM ($P = 0.043$) fungal β -diversity and Ed (**b**); between total ($P = 0.004$), AM ($P = 0.025$) and EM ($P = 0.023$) plant β -diversity and latitude (**c**); between total ($P = 0.220$), AM ($P = 0.346$) and EM ($P = 0.387$) plant β -diversity and Ed (**d**). AM: arbuscular mycorrhizal; EM: ectomycorrhizal.

Table 1

Results of Pearson's correlation analyses between site-level environmental variables and the standardized effect sizes (SEs) associated with shifts in median community composition relative to predictions assuming neutral assembly. Separate analyses were performed for total, arbuscular mycorrhizal (AM), and ectomycorrhizal (EM) fungal and plant communities (results for plant communities are presented within parentheses) groups and environmental factors.

Environmental parameters	Total fungi (plant)		AM fungi (plant)		EM fungi (plant)	
	<i>r</i> value	<i>P</i> value	<i>r</i> value	<i>P</i> value	<i>r</i> value	<i>P</i> value
pH	-0.327 (-0.324)	0.300 (0.305)	-0.548 (-0.330)	0.065 (0.321)	-0.432 (0.302)	0.161 (0.429)
TC	-0.269 (-0.120)	0.398 (0.711)	0.227 (-0.235)	0.478 (0.488)	-0.149 (-0.574)	0.645 (0.106)
TN	-0.141 (-0.054)	0.662 (0.867)	-0.165 (-0.232)	0.609 (0.493)	-0.300 (-0.587)	0.343 (0.096)
TP	-0.357 (-0.239)	0.255 (0.454)	-0.336 (-0.293)	0.286 (0.383)	-0.486 (-0.427)	0.109 (0.252)
C:N	-0.358 (-0.284)	0.254 (0.371)	0.659 (-0.246)	0.020 (0.465)	0.165 (-0.254)	0.608 (0.510)
N:P	0.147 (-0.070)	0.649 (0.828)	0.336 (-0.071)	0.286 (0.835)	0.173 (0.111)	0.591 (0.776)
Plant basal area	-0.199 (0.016)	0.536 (0.960)	-0.028 (0.665)	0.931 (0.025)	0.021 (-0.272)	0.949 (0.478)
Plant richness	0.612 (0.563)	0.034 (0.057)	-0.050 (0.670)	0.877 (0.024)	-0.129 (-0.325)	0.688 (0.393)
Plant comm. comp.	0.570 (0.106)	0.053 (0.743)	-0.551 (0.082)	0.079 (0.810)	0.130 (0.696)	0.688 (0.037)
Plant SES	0.483 (/)	0.112 (/)	0.117 (/)	0.732 (/)	0.183 (/)	0.638 (/)
Latitude	-0.693 (-0.464)	0.012 (0.129)	0.238 (-0.533)	0.457 (0.091)	-0.216 (-0.293)	0.500 (0.444)
MAT	0.716 (0.463)	0.009 (0.129)	-0.280 (0.525)	0.378 (0.098)	0.279 (0.330)	0.381 (0.385)
MAP	0.574 (0.287)	0.051 (0.365)	0.090 (0.270)	0.781 (0.422)	0.191 (-0.125)	0.553 (0.749)

Table 2

Results of partial Mantel tests (r and P values) between each fungal community matrix and matrices of environmental parameters associated with geographic distance, soil properties, corresponding plant parameters, and climatic factors.

Factors	Total fungi		AM fungi		EM fungi	
	r value	P value	r value	P value	r value	P value
Geographic distance	-0.006	0.692	0.014	0.101	0.0001	0.497
pH	0.073***^b	0.001	0.076***	0.001	0.044***	0.001
TC ^c	0.184***	0.001	0.321***	0.001	0.149***	0.001
TN	-0.034	0.982	0.009	0.253	-0.017	0.923
TP	0.009	0.272	-0.028	0.994	0.001	0.454
C:N	-0.093	1.000	0.025	0.053	-0.076	1.000
N:P	-0.083	1.000	-0.081	1.000	-0.084	1.000
Plant basal area (Tot., AM, EM, respectively)	-0.029	0.961	0.083***	0.001	-0.005	0.598
Plant richness (Tot., AM, EM, respectively)	-0.039	0.998	0.016	0.121	0.019	0.063
Plant community composition	0.270***	0.001	0.213***	0.001	0.199***	0.001
MAT	0.072***	0.001	0.108***	0.001	0.032**	0.008
MAP	-0.092	1.000	-0.176	1.000	-0.075	1.000

The relationships which were found to be significant are indicated in bold, **, $P \leq 0.01$; ***, $P \leq 0.001$. TC, soil total carbon; TN, total nitrogen; TP, total phosphorus; C:N, ratio of TC to TN; N:P, ratio of TN to TP; BSA, plant basal area; MAT, mean annual temperature; MAP, mean annual precipitation. AM, arbuscular mycorrhizal; EM, ectomycorrhizal.

Supplementary Information

Title:

Assembly processes lead to divergent soil fungal communities within and among twelve forest ecosystems along a latitudinal gradient

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Running title: Fungal community assembly in forest soils.

Figure legends

Figure S1. Geographic distribution of sampling sites in a map of China, which was slightly modified with an inset map based on Ji et al. (2019). Temperate forests include Genhe (GH), Liangshui (LS), Changbaishan (CBS), Donglingshan (DLS), and Baotianman (BTM); subtropical forests include Gutianshan (GTS), Badagongshan (BDGS), Tiantongshan (TTS), Heishiding (HSD), and Dinghushan (DHS); tropical forests include Nonggang (NG) and Xishuangbanna (XSBN). The inset color map shows the distribution of 20 quadrats in the 24-ha plot of GTS.

Figure S2. Distributions of Sørensen dissimilarities (β -diversity) of the observed communities of total fungi **(a)**, AM (arbuscular mycorrhizal) fungi **(b)**, EM (ectomycorrhizal) fungi **(c)**, total plant **(d)**, AM plant **(e)**, and EM plant **(f)** across twelve forests.

Figure S3. A significantly relationship was observed between the arbuscular mycorrhizal (AM) fungal diversity (OTU richness) and latitude. The parameters of this strong relationships were: $AIC = 118.7$; $R^2_{adj} = 0.7631$; $F = 18.72$; $P < 0.001$.

Fig. S1

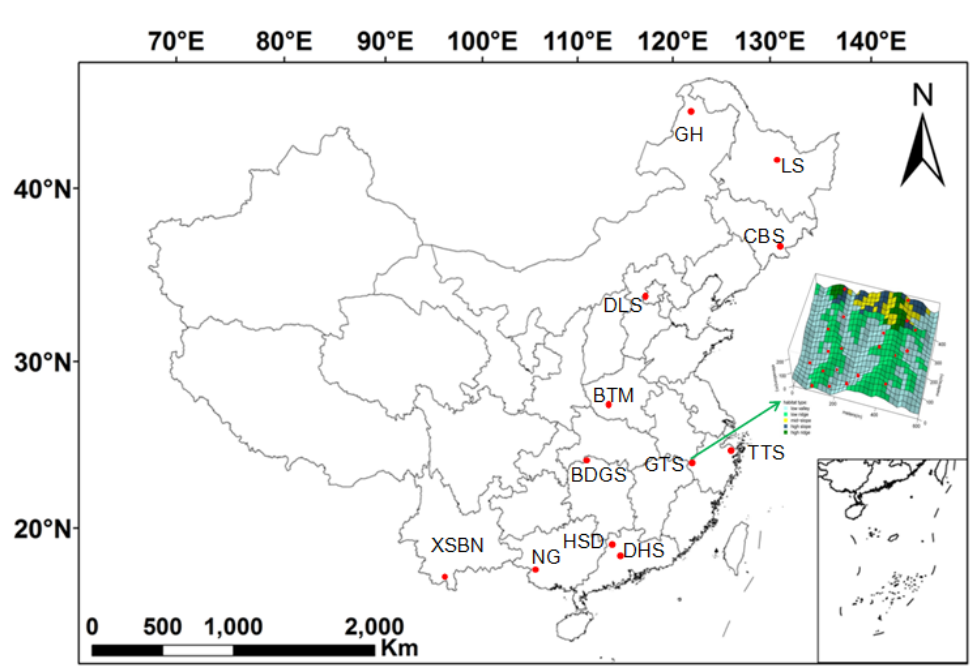


Figure S1. Geographic distribution of sampling sites in a map of China, which was slightly modified with an inset map based on Ji et al. (2019). Temperate forests include Genhe (GH), Liangshui (LS), Changbaishan (CBS), Donglingshan (DLS), and Baotianman (BTM); subtropical forests include Gutianshan (GTS), Badagongshan (BDGS), Tiantongshan (TTS), Heishiding (HSD), and Dinghushan (DHS); tropical forests include Nonggang (NG) and Xishuangbanna (XSBN). The inset color map shows the distribution of 20 quadrats in the 24-ha plot of GTS.

Fig. S2

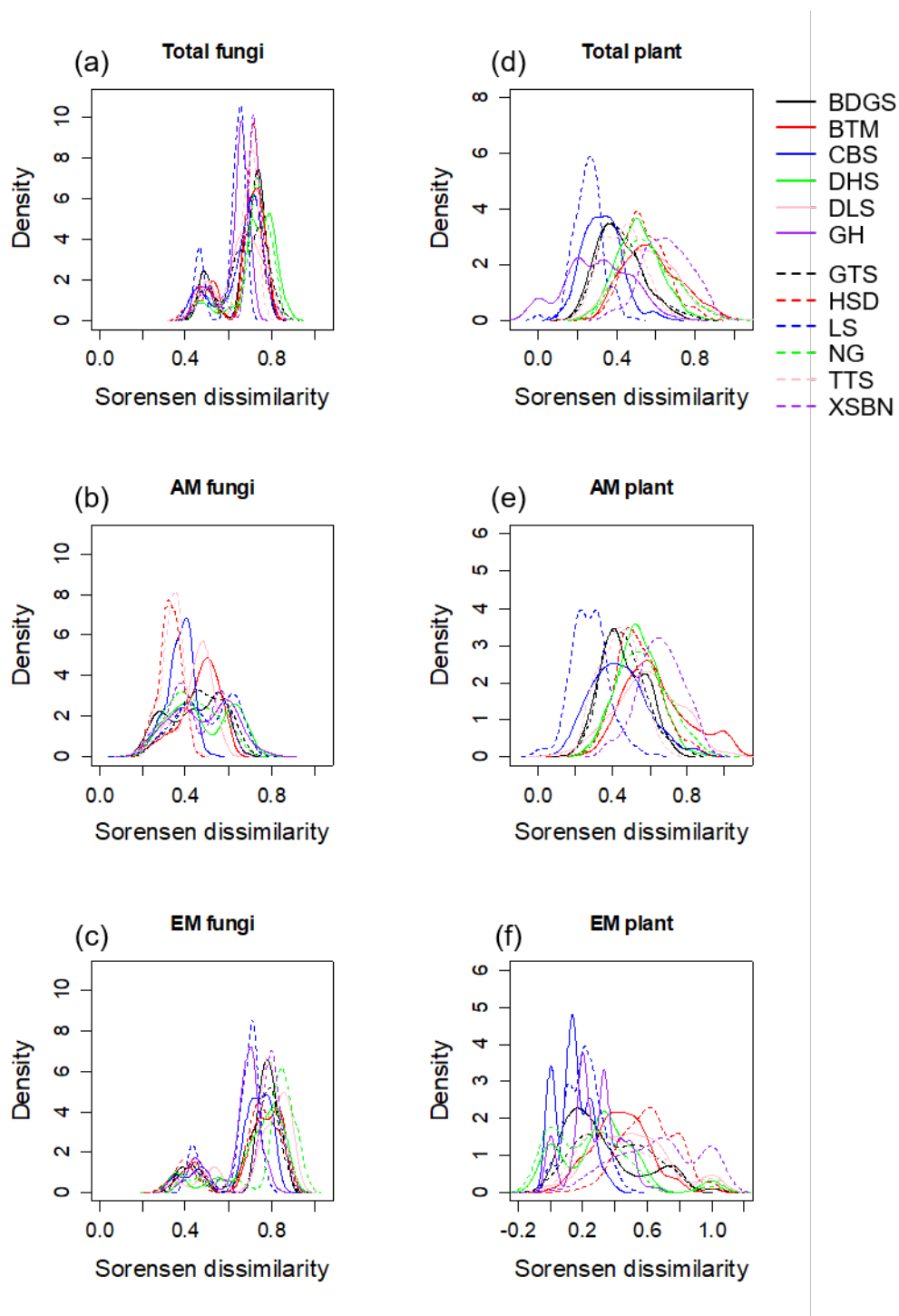


Figure S2. Distributions of Sørensen dissimilarities (β -diversity) of the observed communities of total fungi (a), AM (arbuscular mycorrhizal) fungi (b), EM (ectomycorrhizal) fungi (c), total plant (d), AM plant (e), and EM plant (f) across twelve forests.

Fig. S3

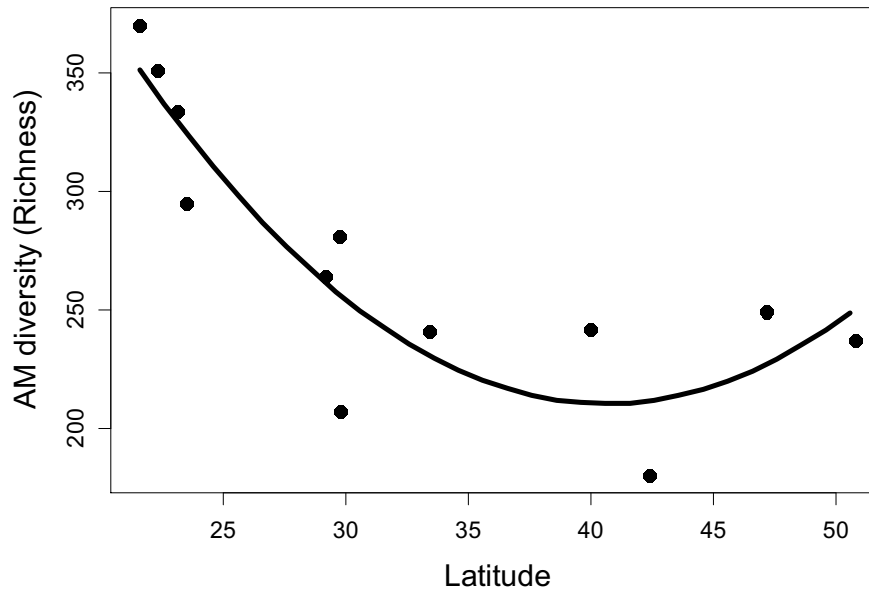


Figure S3. A significantly relationship was observed between the arbuscular mycorrhizal (AM) fungal diversity (OTU richness) and latitude. The parameters of this strong relationships were: $AIC = 118.7$; $R^2_{adj} = 0.7631$; $F = 18.72$; $P < 0.001$.

Table S1.

The basic soil, plant, and climatic information of 12 forest sites (content supplement based on Table 1 in Zheng *et al.* 2020).

Site	Soil factors						Plant factors			Climatic factors		
	pH	TC ^b (g kg ⁻¹)	TN (g kg ⁻¹)	TP (g kg ⁻¹)	C:N	N:P	Forest types ^c	Basal area (×1000 cm ²)	Richness	Climate zone	MAT (°C)	MAP (mm)
BDGS	4.21±0.04f ^a	11.39±0.422a	0.796±0.019a	0.778±0.046cd	14.28±0.329def	1.080±0.058c	SEF	20.48±1.243ab	45.35±1.390a	Subtropic	11.5	2105
BTM	4.88±0.07cd	6.233±0.508cd	0.444±0.041d	0.536±0.050de	14.44±0.384de	0.900±0.117cd	WTDF	15.52±0.876bcd	24.10±2.531de	Temperate	15.1	886
CBS	5.48±0.05b	7.950±0.529bc	0.652±0.044abc	1.162±0.053ab	12.26±0.188fg	0.553±0.022d	TPF	17.71±1.169ab	11.40±0.600f	Temperate	3.6	700
DHS	3.78±0.01g	4.078±0.248def	0.290±0.015e	0.156±0.007f	13.98±0.272defg	1.894±0.102b	SEF	10.58±0.645cde	26.65±2.045cd	Subtropic	20.9	1929
DLS	6.71±0.05a	7.104±0.581c	0.536±0.042bcd	0.838±0.052c	13.26±0.213defg	0.633±0.023cd	WTDF	9.770±0.521de	10.10±0.680fg	Temperate	4.8	550
GH	4.93±0.07c	11.93±1.364a	0.507±0.051cd	0.921±0.047bc	23.50±0.830a	0.572±0.070d	CTCF	14.43±1.692bcde	3.150±0.274g	Temperate	-5.3	450
GTS	4.68±0.04de	4.034±0.254def	0.208±0.016e	0.136±0.018f	19.79±0.450b	1.948±0.224b	SEF	17.10±1.326bc	36.55±2.222b	Subtropic	15.3	1964
HSD	4.12±0.03f	3.056±0.188ef	0.231±0.010e	0.078±0.003f	13.08±0.353efg	3.014±0.114a	SEF	16.30±1.328bcd	45.80±2.022a	Subtropic	19.6	1744
LS	5.64±0.08b	11.80±0.561a	0.703±0.038a	1.026±0.060abc	17.24±0.799c	0.702±0.031cd	TPF	24.07±3.426a	17.10±0.695ef	Temperate	-0.3	676
NG	6.54±0.05a	5.327±0.174cde	0.549±0.014bcd	1.210±0.166a	9.696±0.136h	0.575±0.059d	TRF	8.552±0.429e	32.60±1.645bc	Tropic	22	1500
TTS	3.78±0.05g	10.37±0.964ab	0.667±0.047ab	0.446±0.022e	15.25±0.367cd	1.545±0.121b	MEDF	14.58±0.833bcde	30.90±1.174bcd	Subtropic	16.2	1375
XSBN	4.51±0.04e	2.285±0.099f	0.187±0.005e	0.252±0.013ef	12.19±0.278g	0.782±0.049cd	TRF	15.13±1.343bcde	47.65±2.833a	Tropic	21.8	1493

^a Values (mean ± SE) in the same column without shared letters denote significant difference at $P < 0.05$, as indicated by Tukey's HSD test. ^b TC, total carbon content; TN, total nitrogen content; TP, total phosphorus content; C:N, ratio of TC to TN; N:P, ratio of TN to TP; MAT, mean annual temperature; MAP, mean annual precipitation. ^c Forest types: TRF, tropical rain forests (XSBN and NG); SEF, subtropical evergreen broad-leaf forests (DHS, HSD, GTS, and BDGS); MEDF, mixed evergreen broad-leaved and deciduous broad-leaved forest (TTS); WTDF, warm-temperate deciduous broad-leaved forests (BTM and DLS); TPF, temperate broad-leaved Korean pine forests (CBS and LS); CTCF, cold temperate monsoon coniferous forest (GH).

Table S2.

The plant community composition data (available at the link below).

<https://www.doi.org/10.6084/m9.figshare.13543046>

Table S3.

The raw data of environmental factors including soil, plant basal area and richness, climatic parameters, and sampling time information (available at the link below).

<https://www.doi.org/10.6084/m9.figshare.13543046>

Table S4.

Primer sets and PCR conditions used to amplify fungal DNA prior to Illumina sequencing.

Target	Nested PCR	Primer	Sequence (5'-3')	Thermal condition	Reference	
Fungi (incl. EM fungi)	First PCR	ITS1F	CTTGGTCATTTAGAG GAAGTAA	95°C for 5 min, 30 cycles of at 94°C for 50 s, 52°C for 1 min, 68°C for 1 min.	White et al., 1990 Gardes & Bruns, 1993	
		ITS4	TCCTCCGCTTATTGA TATGC			
	Second PCR	ITS5	GGAAGTAAAAGTCG TAACAAGG			95°C for 3 min, 30 cycles of at 94°C for 50 s, 52°C for 1 min, 68°C for 1 min.
		ITS2	GCTGCGTTCTTCATC GATGC			
AM fungi	First PCR	GeoA2	CCAGTAGTCATATGC TTGTCTC	95°C for 5 min, 30 cycles of at 94°C for 1 min, 58°C for 50 s, 68°C for 1 min.	Schwarzott & Schüßler, 2001 Lee et al., 2008	
		AML2	GAACCCAAACACTT TGGTTTCC			
	Second PCR	NS31	TTGGAGGGCAAGTC TGGTGCC			95°C for 3 min, 30 cycles of at 94°C for 1 min, 58°C for 50 s, 68°C for 1 min.
		AMDGR	CCCAACTATCCCTAT TAATCAT			

Note: AM, arbuscular mycorrhizal; EM, ectomycorrhizal.

Table S5.

The fungal community composition data (available at the link below).

<https://www.doi.org/10.6084/m9.figshare.13543046>

Table S6.

The values of standardized effect size (SES) associated with median community dissimilarities for the observed total, arbuscular mycorrhizal (AM), and ectomycorrhizal (EM) fungal and plant (in parentheses). β -diversities are presented as the mean of the estimate relative to those of 100 simulated communities for each of the 12 locations.

Zone	Site	SES value ^a			
		Total fungi (plant)	AM fungi (plant)	EM fungi (plant)	Comparison (plant)
Subtropic	BDGS	227.2 (8.513) ^b	27.87 (5.660)	5.373 (-1.621)	H-M-L (h-m-l) ^c
Temperate	BTM	295.9 (2.787)	5.592 (1.816)	36.18 (-0.052)	H-L-M (h-m-l)
Temperate	CBS	7.881 (0.023)	23.35 (0.152)	1.090 (-1.210)	M-H-L (m-h-l)
Subtropic	DHS	260.2 (1.557)	95.31 (1.139)	61.30 (n.a.)	H-M-L (h-l-n.a.)
Temperate	DLS	34.38 (0.345)	7.183 (0.364)	4.680 (1.812)	H-M-L (l-m-h)
Temperate	GH	19.27 (-0.245)	197.4 (n.a.)	6.519 (-0.506)	M-H-L (h-n.a.-l)
Subtropic	GTS	49.28 (2.063)	87.38 (2.320)	5.739 (0.048)	M-H-L (m-h-l)
Subtropic	HSD	53.85 (1.959)	67.43 (3.141)	3.232 (0.846)	M-H-L (m-h-l)
Temperate	LS	56.00 (4.164)	185.8 (-0.337)	6.412 (0.516)	M-H-L (h-l-m)
Tropic	NG	217.0 (7.319)	4.617 (5.976)	2.859 (n.a.)	H-M-L (h-l-n.a.)
Subtropic	TTS	131.8 (19.56)	68.66 (13.32)	7.148 (n.a.)	H-M-L (h-l-n.a.)
Tropic	XSBN	197.7 (24.93)	55.34 (22.55)	7.174 (1.954)	H-M-L (h-m-l)
Fungi	ANOVA ^d	Df = 2; $F = 7.922$; $P = 0.002$			
	Mean ^e	129.2±30.17A	68.83±18.89AB	12.31±5.177B	H-M-L
Plant	Kruskal-Wallis test ^d	$\chi^2 = 8.530$, $P = 0.014$			
	Mean ^e	6.082±2.339A	5.099±2.105AB	0.199±0.410B	h-m-l

^a The SES values were calculated according to the formula as '(observed estimate–mean of the simulated estimates) / standard deviation of the simulated estimates' using the Sørensen dissimilarities of total, AM, and EM fungal and plant community data.

^b Consistently positive effect size (SES value > 0) indicates that the estimate is higher than predicted (divergence) under the neutral hypothesis.

^c Letters of 'H (h)', 'M (m)', and 'L (l)' denote the SES values is high, medium, and low, respectively, following the order of total, AM, and EM fungi and plants in the same row.

^d The outcomes of one-way ANOVA or nonparametric Kruskal-Wallis test analyses on the effect of site on the SES values amongst different fungal and plant groups.

^e Values (mean ± SE) in that row without shared capital letters denote significant difference amongst fungal and plant groups at $P < 0.05$, as indicated by Tukey's HSD test.

Table S7.

The estimated parameters associated with the neutral model of biodiversity fit at the level of Chinese forest soil for total, AM, and EM fungal and plant (within parentheses) communities.

Site	Theta (Θ)			I (median)		
	Total fungi (plant)	AM fungi (plant) ^a	EM fungi (plant)	Total fungi (plant)	AM fungi (plant)	EM fungi (plant)
BDGS	16.72 (22.83)	13.76 (18.35)	12.57 (3.39)	4.12E+12 (128)	672 (110.3)	18 (3.40)
BTM	20.40 (16.51)	22.70 (15.38)	7.52 (4.14)	3.64E+12 (29)	26 (20.9)	21377 (3.91)
CBS	50.06 (6.19)	12.33 (4.66)	29.57 (1.68)	89 (15.5)	191 (6.80)	13 (6.72)
DHS	12.85 (23.42)	9.08 (21.84)	4.29 (n.a.)	5.11E+12 (30.5)	4537 (24.1)	2.6E+12 (n.a.)
DLS	34.85 (8.59)	19.29 (4.58)	14.24 (1.91)	501 (4.53)	46 (3.10)	26.5 (6.41)
GH	35.56 (1.80)	6.46 (n.a.) ^b	13.93 (1.31)	275 (1.46)	2.5E+12 (n.a.)	46.5 (1.37)
GTS	25.08 (24.16)	6.86 (16.57)	12.39 (3.13)	8189 (40.25)	3E+12 (33.9)	24 (3.47)
HSD	32.31 (47.12)	8.97 (35.20)	16.76 (9.91)	868 (48.44)	3346 (51.6)	14 (4.46)
LS	31.29 (6.65)	8.31 (7.07)	15.46 (1.50)	1016 (56.23)	10764 (13.8)	51 (54.5)
NG	16.72 (21.13)	20.46 (20.79)	16.98 (n.a.)	3.80E+12 (40.59)	108 (37.5)	23.5 (n.a.)
TTS	16.26 (12.72)	10.37 (10.34)	10.88 (n.a.)	4.69E+12 (491)	1727 (313)	48.5 (n.a.)
XSBN	20.40 (19.14)	11.78 (16.98)	12.82 (3.03)	3.1E+12 (5.4E+11)	2251 (4.55E+11)	26 (4.02)

^a AM, arbuscular mycorrhizal; EM, ectomycorrhizal.

^b n.a., Result was not available because of the limited plant community data in these sites.

Table S8.

The observed β -diversity (Sørensen index) of the total, arbuscular mycorrhizal fungi (AM), and ectomycorrhizal (EM) fungal and plant (within parentheses) communities at each of 12 forest sites.

Sequence	Sites	Total fungi (plant)	AM fungi (plant)	EM fungi (plant)
1	BDGS	0.338 (0.352)	0.233 (0.371)	0.603 (0.203)
2	BTM	0.493 (0.520)	0.480 (0.552)	0.571 (0.396)
3	CBS	0.453 (0.266)	0.365 (0.346)	0.588 (0.107)
4	DHS	0.399 (0.499)	0.194 (0.510)	0.410 (0.303)
5	DLS	0.405 (0.499)	0.396 (0.454)	0.538 (0.434)
6	GH	0.345 (0.258)	0.365 (n.a.)	0.435 (0.201)
7	GTS	0.364 (0.396)	0.112 (0.390)	0.505 (0.352)
8	HSD	0.434 (0.530)	0.132 (0.512)	0.588 (0.642)
9	LS	0.272 (0.215)	0.443 (0.272)	0.434 (0.115)
10	NG	0.390 (0.543)	0.291 (0.545)	0.539 (0.330)
11	TTS	0.287 (0.354)	0.156 (0.357)	0.474 (0.403)
12	XSBN	0.357 (0.482)	0.269 (0.468)	0.548 (0.549)
Fungi	ANOVA ^a	Df = 2; $F = 20.62$; $P < 0.0001$		
	Mean	0.378±0.019B ^b	0.286±0.036C	0.519±0.019A
Plant	ANOVA ^a	Df = 2; $F = 1.815$; $P = 0.179$		
	Mean	0.410±0.034A	0.434±0.028A	0.336±0.047A

^a The outcomes of one-way ANOVA analyses on the effect of site on the observed β -diversity amongst different fungal and plant groups.

^b Values (mean \pm SE) in that row without shared capital letters denote significant difference in observed β -diversity amongst fungal and plant groups at $P < 0.05$, as indicated by Tukey's HSD test.

Table S9.

The observed β -diversity (Sørensen index) of the total, arbuscular mycorrhizal (AM), and ectomycorrhizal (EM) fungal and plant communities across tropic, subtropic, and temperate climate zones.

Type		Tropic		Subtropic		Temperate	
		Mean ^a	Median ^a	Mean	Median	Mean	Median
Fungi	Total	0.712	0.734	0.463	0.437	0.770	0.818
	AM	0.754	0.774	0.526	0.574	0.792	0.826
	EM	0.716	0.744	0.514	0.533	0.765	0.807
	ANOVA ^b	Df = 2; $F = 115$; $P < 0.0001$					
	Mean	0.739 \pm 0.010B ^c		0.508 \pm 0.020C		0.796 \pm 0.010A	
Plant	Total	0.800	0.930	0.828	0.921	0.824	0.961
	AM	0.801	0.927	0.830	0.922	0.799	0.875
	EM	0.757	0.990	0.854	1.000	0.783	1.000
	ANOVA ^b	Df = 2; $F = 0.281$; $P = 0.764$					
	Mean	0.786 \pm 0.015A ^c		0.837 \pm 0.008A		0.802 \pm 0.012A	

^a The values were shown as the mean (median) of the Sørensen index within each climate zone for each plant group.

^b The outcomes of one-way ANOVA analyses on the fungal and plant observed β -diversity amongst three different climate zones.

^c Values (mean \pm SE) in that row with shared capital letters denote non-significant difference amongst three climate zones at $P < 0.05$, as indicated by Tukey's HSD test.

Table S10.

Detailed PerMANOVA outcomes using all data focusing on the effects (*i.e.*, predictors) of climate zone ($n = 240$) and site ($n = 240$) and re-analyzed to reveal the site effect for each three climate zones of tropic ($n = 40$), subtropic ($n = 100$), and temperate regions ($n = 100$). The analyses were conducted for total, arbuscular mycorrhizal (AM), and ectomycorrhizal (EM) fungi and plant communities individually.

Effect	Df	Total fungi (plant)		AM fungi (plant) ^a		EM fungi (plant) ^a	
		R^2	P value	R^2	P value	R^2	P value
Climate zone	2	0.095 (0.143)	0.001 (0.001)	0.280 (0.157)	0.001 (0.001)	0.069 (0.130)	0.001 (0.001)
Site (All data)	11	0.262 (0.572)	0.001 (0.001)	0.490 (0.540)	0.001 (0.001)	0.207 (0.598)	0.001 (0.001)
Site (Tropic)	1	0.086 (0.359)	0.001 (0.001)	0.266 (0.361)	0.001 (0.001)	0.064 (0.327)	0.003 (0.001)
Site (Subtropic)	4	0.222 (0.492)	0.001 (0.001)	0.403 (0.480)	0.001 (0.001)	0.172 (0.528)	0.001 (0.001)
Site (Temperate)	4	0.183 (0.562)	0.001 (0.001)	0.247 (0.465)	0.001 (0.001)	0.156 (0.620)	0.001 (0.001)