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Community dissimilarity of angiosperm trees reveals deep-time diversification across tropical and temperate forests

Kusumoto, Buntarou; Kubota, Yasuhiro; Baselga, Andrés; Gómez-Rodríguez, Carola; Matthews, Tom; Murphy, Daniel J.; Shiono, Takayuki

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8	Running head: Regional patterns of community dissimilarity
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10	Buntarou, Kusumoto ^{1, 2}
11	Yasuhiro, Kubota ³
12	Andrés Baselga ⁴
13	Carola Gómez-Rodríguez ⁵
14	Thomas J. Matthews ^{6,7}
15	Daniel J. Murphy ⁸
16	Takayuki, Shiono ³
17	
18	1. University Museum, University of the Ryukyus, 1 Sembaru, Nishihara, Okinawa, 903-
19	0213 Japan
20	2. Kasuya Research Forest, Faculty of Agriculture, Kyushu University, 394 Tsubakuro,
21	Sasaguri, Fukuoka, 811-2415 Japan (current affiliation)
22	3. Faculty of Science, University of the Ryukyus, 1 Sembaru, Nishihara, Okinawa, 903-
23	0213 Japan
24	4. CRETUS Institute. Department of Zoology, Universidade de Santiago de Compostela,

25	c/ Lope Gómez de Marzoa s/n, 15782 Santiago de Compostela, Spain
26	5. CRETUS Institute. Department of Functional Biology, Universidade de Santiago de
27	Compostela, c/ Lope Gómez de Marzoa s/n, 15782 Santiago de Compostela, Spain
28	6. GEES (School of Geography, Earth and Environmental Sciences) and Birmingham
29	Institute of Forest Research, University of Birmingham, Birmingham, B15 2TT
30	7. CE3C - Centre for Ecology, Evolution and Environmental Changes/Azorean
31	Biodiversity Group and Universidade. dos Açores - Depto de Ciências Agráriase
32	Engenharia do Ambiente, PT-9700-042, Angra do Heroísmo, Açores, Portugal.
33	8. Royal Botanic Gardens Victoria, Melbourne 3004, Australia
34	
35	Correspondence
36	Buntarou, Kusumoto, University Museum, University of the Ryukyus, 1 Sembaru,
37	Nishihara, Okinawa, 903-0213 Japan
38	Email: kusumoto.buntarou@gmail.com
39	
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47 Abstract

48 **Question:** To better understand the influence of deep-time diversification on extant plant 49 communities, we assessed how community dissimilarity increases with spatial and 50 climatic distances at multiple taxonomic ranks (species, genus, family, and order) in 51 angiosperm trees. We tested the prediction that the dissimilarity-to-distance relationship 52 should change across taxonomic ranks depending on the deep-time diversification in 53 different biogeographic regions reflecting geohistories and geographical settings.

54 Location: Global

55 **Methods:** Using a dataset of plot-based surveys across the globe (861 plots), we compiled 56 a community composition matrix comprising 21,455 species, 2,741 genera, 240 families, 57 and 57 orders. We then calculated Sørensen's pairwise dissimilarity (β_{sor}), and its turnover 58 (β_{sim}) and nestedness (β_{sne}) components, among plots within seven biogeographical 59 regions. Finally, we modelled the relationships between the biotic dissimilarities and the 50 spatial/climatic distances at each taxonomic rank, and compared them among regions.

Results: β_{sor} and β_{sim} increased with increasing spatial and climatic distance in all 61 62 biogeographical regions: β_{sim} was dominant in all biogeographical regions in general, while β_{sne} showed relatively high contributions to total dissimilarity in the temperate 63 regions with historically unstable climatic conditions. The β_{sim} -distance curve was more 64 65 saturated at smaller spatial scales in the tropics than in the temperate regions. In general, the curves became flatter at higher taxonomic ranks (order or family), with exception of 66 Africa, North America, and Australia pointing to region-specific geographical constraints. 67 68 Conclusions: Compositional dissimilarity was generally shaped through the abrupt turnover of species along spatial/climatic gradients. The relatively high importance of the 69 nestedness component in the temperate regions suggests that historical dispersal filters 70

71 related to extinction/colonization may play important roles. Region-specific changes in 72 the turnover and nestedness components across taxonomic ranks suggest differential 73 imprints of historical diversification over deep evolutionary time in shaping extant 74 diversity patterns in each biogeographical region.

75

Keywords: angiosperm trees; biogeographical regions; beta diversity; dissimilarity
decomposition; dispersal limitation; distance decay; environmental filtering; taxonomic
diversity

79

80 Introduction

81 The similarity in species composition between local biological communities decreases 82 with spatial/environmental distance. This pattern, known as distance decay of similarity or simply distance-decay, is ubiquitous across organisms and biological systems (Nekola 83 & White 1999; Soininen et al. 2007). The decrease of similarity with distance is 84 85 mechanistically driven by dispersal limitation (Hubbell 2001) and niche constraints (Gilbert & Lechowicz 2004), mediated through contemporary environmental gradients 86 and evolutionary diversification. Indeed, there are a number of empirical studies that have 87 applied a distance-decay approach to answer different questions related to community 88 89 assembly processes: niche partitioning along current environmental gradients (Bellier et 90 al. 2014; Trujillo et al. 2019; Cacciatori et al. 2020), historical species sorting through 91 dispersal of organisms (Qian 2009; Saito et al. 2015), dispersal barriers related to geographical constraints (Stuart et al. 2012), differences in dispersal ability among taxa 92 93 (Gómez-Rodríguez & Baselga 2018; Gómez-Rodríguez et al. 2020) and historical habitat stability (Fitzpatrick et al. 2013). A common theme linking these studies is their use of 94

95 the shape of distance decay curves as a measure to infer the role of historical and 96 contemporary factors in generating diversity patterns under study (Morlon et al. 2008; 97 Réjou-Méchain & Hardy 2011; Gómez-Rodríguez & Baselga 2018). Rooted in this 98 approach, we here propose that assessing distance decay patterns at multiple taxonomic 99 ranks (species, genus, family, and order) will allow inferring the influence of deep-time 100 diversification on extant plant communities, and its contrasting relevance in tropical and 101 temperate biogeographic regions.

102

Angiosperm tree floras in tropical and temperate regions are regarded as evolutionary 103 104 source and sink, respectively. Many clades that now comprise the tropical flora have been 105 hypothesized to have originated in western Gondwana, which was part of the 106 supercontinent that contained South America and Africa (Fine & Ree 2006; Christenhusz & Chase 2012), and subsequently diversified, after crossing long-standing geographic 107 108 barriers, among the continents through the Eocene, Oligocene, and Miocene (Hardy et al. 109 2012). In temperate areas, many extant lineages are thought to have originated from the Asian tropical flora (Wen 1999; Donoghue 2008) and afterward have regionally 110 111 diversified in East Asia, Northern America, and Europe in response to the Plio-Pleistocene global cooling (Fine & Ree 2006). Macroscale diversity patterns of 112 angiosperm trees are characterized by evolutionary radiations within disjunct 113 114 families/genera in tropical and temperate forests (Gentry 1988; Donoghue & Smith 2004) 115 and taxon-specific selective dispersal/extinction related to paleoclimate changes in 116 temperate regions (Svenning 2003; Eiserhardt et al. 2015).

117

118 The aforementioned studies suggest that the study of taxonomic diversity across lower

(species) to higher taxonomic ranks (genus, family, and order) could provide a 119 120 fundamental basis for better understanding deep-time diversification related to geohistory 121 including paleoclimates. Indeed, correlations of species richness within a higher taxonomic group (family or order) among continents have been shown to represent a 122 123 consistent biogeographical pattern resulting from diversification at different evolutionary time scales and related to family-specific niche conservatism and global-scale dispersal 124 125 (Ricklefs & Renner 2012; Munoz et al. 2012; Chen et al. 2012). Therefore, we propose that assessing the distance decay curves of tree angiosperm communities at a range of 126 taxonomic ranks, which provide a surrogate for a macroevolutionary hierarchy (Graham 127 128 et al. 2016), should reveal any historical imprint on current spatial diversity patterns 129 (Munoz et al. 2014; Yeh et al. 2019). Specifically, regional differences in deep-time diversification should be reflected in the geographical distribution of higher-rank taxa, 130 131 such as genera, families or orders, through processes such as niche conservatism and 132 dispersal limitation (Kerkhoff et al. 2014; Weiser et al. 2018).

133

134 While similarity indices are commonly used in distance decay studies (Nekola & White 1999), compositional dissimilarity (i.e. 1 - similarity) metrics can describe the equivalent 135 patterns and some of these metrics can be partitioned into turnover and nestedness-136 137 resultant components (Baselga 2010; Legendre 2014; Soininen et al. 2017). The turnover 138 component represents taxonomic replacement that may be caused by species sorting associated with niche differentiation, evolutionary processes such as radiation and 139 allopatric speciation, and/or dispersal limitation (Leibold et al. 2004; Leprieur et al. 2011). 140 141 The nestedness component reflects changes in species richness caused by selective species loss or gain that may be associated with recent vicariant events, e.g. insular 142

changes by sea-level rise (Rijsdijk et al. 2014), or a colonization lag after drastic 143 144 environmental changes such as ice age disturbances (Hortal et al. 2011). Therefore, each 145 component is expected to have an independent relationship with geographical and environmental distance (Antão et al. 2019; Bevilacqua & Terlizzi 2020), and their relative 146 147 importance may change depending on historical habitat stability (Baselga et al. 2012). The turnover component can be expected to have a steep slope and an asymptotic 148 149 relationship with spatial distance under strong dispersal limitation, while a flatter 150 relationship would be observed when dispersal limitation is weak or absent (Gómez-151 Rodríguez et al. 2020). The nestedness component is expected to linearly decrease with 152 increasing spatial/environmental distance under strong dispersal limitation, but to be 153 independent of distance under no dispersal limitation (Gianuca et al. 2016; Antão et al. 2019). 154

155

156 Using a global dataset of forest plots (861 plots), we quantified pairwise compositional dissimilarity of angiosperm tree communities at different taxonomic ranks (species, genus, 157 family, and order) in seven biogeographical regions (South American, African, Indo-158 Pacific, Australian, North American, West Eurasian, and East Eurasian). We compared 159 the relationship between compositional dissimilarity and spatial/climatic distance (i.e. 160 161 dissimilarity-to-distance relationship) between the biogeographical regions and between 162 the taxonomic ranks in order to explore the effect of deep-time diversification on the spatial patterns of extant tree communities. Based on the aforementioned reasoning, we 163 164 tested the following predictions: i) taxonomic turnover will be the dominant component 165 of dissimilarity in historically stable regions (Baselga et al. 2012), such as regions that contain tropical areas; ii) in contrast, the nestedness component will be predominant in 166

historically unstable regions (e.g. temperate areas in higher latitudes) that have 167 168 experienced extinction and colonization events in response to the expansion/retreat of ice 169 sheets due to paleo-climate changes (Baselga et al. 2012; Soininen et al. 2017); iii) the relationship between the turnover component and spatial/climatic distance will be very 170 171 steep at short distances, quickly saturating at maximum dissimilarity at the species level due to strong dispersal limitation, while the relationship will become flatter at higher 172 173 taxonomic ranks. This flattening should be more marked in the regions containing tropical 174 areas because of the older evolutionary age of tropical areas, which should have allowed 175 higher-level taxa to spread across wider areas (i.e. lower dispersal limitation); iv) however, 176 the steep relationship between the turnover component and spatial/climatic distances 177 should remain asymptotic even at higher taxonomic ranks (Cowling et al. 2015) if climatic gradients and/or vicariance have been maintained over large time periods; and v) 178 179 the nestedness component should show a negative linear relationship with spatial/climatic 180 distance at the species level in historically unstable regions, but be independent from the 181 distances at higher taxonomic ranks due to less dispersal limitation.

182

183 Materials and Methods

184 Angiosperm tree community data

Community composition data of angiosperm tree species were collated from a series of plot-based surveys across the globe (Ulrich et al. 2016; Kubota et al. 2018). The data were compiled from a literature census using various search engines, including Web-Of-Science (Thomson-Reuters, New York, NY, USA) and Google Scholar (http://scholar.google.com/), and web-based forest plot databases (e.g. Gentry's data; www.wlbcenter.org/gentry data.htm). Our dataset only includes plots where the absolute

191 number of individuals was recorded for all tree species at a given census threshold in 192 individual size (i.e. diameter at breast height). The taxonomic classification (species, 193 genus, family, and order) was standardized following The Plant List (http://www.theplantlist.org/). Unnamed species and morphospecies identified only to 194 195 genus were treated as individual species (we confirmed that excluding these species did not meaningfully affect the results of the dissimilarity-to-distance analyses). We excluded 196 197 naturally/artificially disturbed plots and plots with less than two angiosperm tree species. 198 We also excluded gymnosperms (163 species) from the data. The final dataset comprised 199 861 plots (range = $100-520,000 \text{ m}^2$; Fig. 1). Using all plots, we created community 200 composition matrices for four taxonomic ranks: species (21,455 species), genus (2741 201 genera), family (240 families), and order (57 orders). We then subdivided each matrix 202 into seven biogeographical regions. We defined the biogeographical regions using a 203 modified version of Cox et al.'s floral Kingdoms (South American, African, Indo-Pacific, 204 Australian, and Holarctic; Cox et al. 2001): we subdivided the Holarctic Kingdom into 205 North American, West Eurasian and East Eurasian (Fig. 1) because of their differences in geohistory and paleoclimatic conditions. South American, African, Indo-Pacific, 206 Australian regions include tropical areas, while North American, West Eurasian and East 207 Eurasian are temperate areas. In this study, we avoided a more detailed regionalization 208 (e.g. ecoregions) because of the limited number of plots and their spatially 209 210 inhomogeneous distribution (Fig. 1). More details of the data compilation process are provided in Ulrich et al. (2016) and Kubota et al. (2018). 211

212

Community under-sampling is a potential problem in dissimilarity analyses (Beck et al.
2013), especially in cases such as ours where data are taken from multiple sources that

215 have used different census schemes (e.g. plot area and the size criterion for measuring 216 individuals). Therefore, the plots were screened based on sampling completeness: we 217 estimated sample coverage (SC) based on relative species abundance, which is an unbiased estimate represented by the proportion of all detected individuals (Chao et al. 218 219 2020): SC values are in the range 0 to 1. We filtered the plots at SC \geq 0.9 (n = 661; Fig. 2) and also examined other criteria ($\geq 0.7, 0.8, \text{ and } 0.85$) to test the potential influence of 220 221 arbitrary choices of SC thresholds. SCs filtered out the incompletely sampled plots which 222 cannot be distinguished by plot-areas or census thresholds, allowing us to include the local communities which were equivalently well sampled (Fig. S1 in Appendix S1). 223 224 Differences in plot areas and census thresholds are particularly likely to influence 225 absolute abundance differences among the plots (Baselga 2013). Therefore, we used presence/absence information in the dissimilarity analyses. 226

227

228 Dissimilarity calculation

Using the subset of equivalently well-sampled plots (SC $\ge 0.7, 0.8, 0.85$, or 0.9), we 229 230 calculated pairwise dissimilarity between plots within the same biogeographic region (Fig. 1) for each taxonomic rank (species, genus, family, and order). We followed Baselga's 231 (2010) beta diversity partitioning framework based on Sørensen dissimilarity (β_{sor}), which 232 was decomposed into turnover (β_{sim}) and nestedness-resultant dissimilarity (β_{sne}) 233 components. The compositional dissimilarity of lower taxonomic ranks is inevitably 234 influenced by the dissimilarity of higher taxonomic ranks due to the ranks being 235 236 hierarchically structured: β_{sne} should be higher at higher taxonomic rank, although β_{sim} is predominant at lower (e.g. species) taxonomic rank. Before analyzing dissimilarity-to-237 distance relationships, we assessed whether the influence of deep-time diversification on 238

beta diversity is region-specific or not by evaluating correlations between the dissimilarity matrices at different taxonomic ranks for each dissimilarity component (β_{sor} , β_{sim} , and β_{sne}) in each biogeographical region.

242

243 Spatial and climatic distances

The spatial distance was defined by the great-circular distance between each pair of plots. We downloaded climatic (Bio-1–19) and elevation data at 30-arc second resolution from the WorldClim ver. 2.1 database (Fick and Hijmans 2017; <u>http://www.worldclim.org</u>), overlapped them with the plot coordinates, and assigned the information to each plot. To analyze the effect of climatic distance, we calculated the Euclidean distance in the 20dimensional space between plots using the variables after standardization (i.e. mean = 0 and variance =1).

251

252 Statistical analysis

We modeled the relationship between pairwise compositional dissimilarity (β_{sor} , β_{sim} , and 253 β_{sne}) and spatial/climatic distance (i.e. the dissimilarity-distance curve) in each region and 254 for each taxonomic rank using both negative exponential and power-law functions 255 (Nekola & McGill 2014); these functions were fitted using a generalized linear modelling 256 257 approach with a Gaussian distribution and a log-link function (Millar et al. 2011). Model fit was evaluated using pseudo- r^2 defined as 1 – (model deviance/null deviance) 258 (McFadden 1973). The two functions provided similar fits to the data according to 259 Akaike's Information Criterion (Table S1 in Appendix S1), and thus we only present 260 261 results for the negative exponential model in the main text (see Fig. S2 in Appendix S1 for the results using the power-law model). In the negative exponential model, the 262

263 intercept and slope can be interpreted as the initial dissimilarity (inherent compositional 264 variation among the closest local communities) and the speed of compositional change 265 (or rate of decay), respectively. We tested for differences in the intercepts and slopes of the negative exponential model between biogeographical regions by bootstrapping 266 267 (multiple comparisons among each pair of regions): we computed 1,000 bootstrap samples for each parameter, calculated the difference in parameter values between two 268 269 regions, calculated the proportion of positive and negative differences respectively, and 270 used the smaller of these (i.e. upper or lower tails) proportions as a p-value. We also 271 assessed the influence of SC thresholds (SC = $0.7 \sim 0.9$) on the parameter estimation by 272 evaluating the inter-regional rank correlations for the effect size of parameters between 273 the different SC thresholds. In addition, we fitted a locally estimated scatterplot 274 smoothing curve to visualize changes in the relative importance of the turnover component to overall dissimilarity (β_{sim}/β_{sor}) along the spatial/climatic distance gradients. 275 276

All statistical analyses and graphical works were undertaken using R ver. 3.6.1 (R Core Team 2019) and the following packages: 'betapart' (Baselga & Orme 2012) to calculate and decompose pairwise compositional dissimilarity, 'geosphere' (Hijmans 2019) to calculate spatial distance, 'iNEXT' (Hsieh et al. 2016) for calculating the sampling completeness of each plot, and 'multcompView' (Graves et al. 2019) for multiple comparisons.

283

284 **Results**

The parameters (especially slopes) of the negative exponential models at lower SC thresholds (0.7–0.85) differed from those estimated using SC \geq 0.9, particularly for the climatic distance model (Fig. S3 in Appendix S1). The inter-regional ranking of effect size was consistent for the spatial distance models (i.e. Spearman's rank correlation $\rho =$ 1), whereas change in the order was observed in the climatic distance model ($\rho = 0.75 \sim$ 1.00). Therefore, we only show the results using the most strict criteria (SC \geq 0.9) for all subsequent analyses.

292

Total dissimilarity (β_{sor}), the turnover component (β_{sim}), and the nestedness component (β_{sne}) showed strong correlations between the taxonomic ranks, but with substantial variation in its degree between the regions (Table S2 in Appendix S1), indicating the influence of region-specific deep-time diversification on shaping turnover/nestednessresultant beta diversity.

298

299 Total dissimilarity and spatial distance

 β_{sor} increased with increasing spatial distance between sites, and this finding was consistent within all biogeographical regions (Fig. 3). The negative exponential model provided a relatively good fit to the dissimilarity-distance pattern at the species level (r^2 = 0.34-0.74), but the amount of explained variance generally decreased along taxonomic ranks from genus to order (Fig. 4; $r^2 = 0.08-0.66$ for genera; $r^2 = 0.01-0.55$ for families; $r^2 = <0.01-0.43$ for orders). The intercept and slope of the negative exponential model became smaller at higher taxonomic ranks (Fig. S4 and S5 in Appendix S1).

307

Relationships between turnover and nestedness-resultant components and spatial
distance

310 The spatial patterns of β_{sim} were well characterized by the negative exponential model,

311 especially at the species level (Fig. 3a), and exhibited an asymptotic increase with spatial 312 distance (Fig. 4). β_{sim} was the predominant component of β_{sor} in all geographical regions 313 except for sites near to each other in the West Eurasian region (Fig. 5), but the relative importance of β_{sim} decreased along the taxonomic ranks from genus to order. The intercept 314 315 and slope of the β_{sim} -distance curve were smaller in the temperate regions (North 316 American, West Eurasian, and East Eurasian) where the β_{sim} at species level slowly 317 saturated over the entire geographical extent at the species level (Fig. S4 and S5). In 318 contrast, the intercept and/or slope of the β_{sim} -distance curve was larger in the regions containing tropical areas (South American, African, and Indo-Pacific) where the β_{sim} at 319 320 species level saturated at a smaller geographical extent (Fig. 4). The β_{sim} values became 321 lower, and the shape of the curve became flatter, at higher taxonomic ranks in the South American and Indo-Pacific regions (Fig. 4). In contrast, the slope of the curve was 322 323 relatively steep even at the family and order levels in the African, Australian, and North 324 American regions (Fig. 4 and S5).

325

 β_{sne} was poorly explained by the negative exponential model (Fig. 3a); the relationship between β_{sne} and spatial distance was mostly flat, while a negative linear relationship was found at the species level in the Holarctic regions, especially in West Eurasian (Fig. 4 and S5). At the species level, β_{sne} accounted for a major proportion of β_{sor} only within sites near to each other in the Holarctic regions (Fig. 5). However, the relative importance of β_{sne} increased at higher taxonomic ranks even in the regions containing tropical areas, especially in South American (Fig. S4).

333

334 *Relationships between taxonomic dissimilarity and climatic distance*

In general, the increase in pairwise taxonomic dissimilarity (β_{sor} , β_{sim} , and β_{sne}) with 335 climatic distance was similar to that observed with spatial distance (Fig. S6-S8 in 336 337 Appendix S1). The negative exponential models fitted using climatic distance had slightly better explanatory power than the models fitted using spatial distance for most regions 338 339 and ranks, but provided worse fits at the species and genus levels in the Indo-Pacific, African, West Eurasian , and Australian regions (Fig. 3). β_{sor} and β_{sim} exhibited steep 340 341 slopes and quick asymptotic saturation in all regions at the species level, while the 342 saturation was relatively slower in Australian and the temperate regions compared to the South American, African and Indo-Pacific regions (all containing tropical areas) (Fig. S6). 343 344 The slope of the β_{sim} -distance curve was smaller at higher taxonomic ranks, but it 345 remained relatively high even at the family and order levels in the African and North American regions (Fig. S6 and S8). The relative importance of β_{sim} to β_{sor} increased with 346 the climatic distance, especially in the West Eurasian at the species level, while the pattern 347 348 was less clear at the family and order levels (Fig. 5).

349

350 Discussion

Our results reveal substantial differences in patterns of variation in local angiosperm tree 351 communities across different biogeographic regions, with contrasting effects of deep time 352 353 processes of diversification between the biogeographical regions with and without tropical areas. The overall dissimilarity of angiosperm communities between forest plots 354 was found to increase with spatial and climatic distance (a distance decay pattern) in all 355 356 biogeographical regions, and was mainly driven by the turnover component at lower taxonomic ranks (species and genus), although its relative contribution decreased at 357 higher taxonomic ranks (family and order). However, these patterns showed region-358

specific variations. The regions containing tropical areas (South American, African, and
Indo-Pacific) showed steeper increases in total dissimilarity and the turnover component
with spatial/climatic distances compared with the temperate regions (North American,
West Eurasian, and East Eurasian), while Australian region showed intermediate trends
Fig. 4, S5, S6, and S8).

364

The fast increase in dissimilarity in regions with tropical areas was mostly associated with 365 366 the turnover component, indicating a fast compositional replacement along spatial and climatic distance gradients, especially at the species level. This is in line with previous 367 368 studies of tropical forests (Condit et al. 2002; Tuomisto et al. 2003; Pennington et al. 369 2009; Trujillo et al. 2019) that found an important role of environmental filtering and 370 dispersal limitation in generating species turnover. The decreasing compositional 371 dissimilarity with increasing taxonomic rank (from species, genus, family to order) and 372 the flattening of the dissimilarity-distance curves may reflect the deeper evolutionary 373 history of tropical forests (Munoz et al. 2014). Specifically, orders and families of angiosperm trees probably radiated globally across phylogenetic niche space (Hubbell 374 2001) under warmer climates through the Cretaceous to the Paleogene, and then 375 subsequently species and genera within those regions diversified in response to different 376 377 drivers, including geographical isolation and tropical-specific historical habitat stability 378 through the Cenozoic (Fine & Ree 2006). Indeed, our results showed the highest species turnover rates in South American region (Fig. 4), providing a support for the view of the 379 region as an evolutionary "engine" of plant diversity (Antonelli et al. 2015). Moreover, 380 381 the turnover-distance relationships were flatter at the higher taxonomic ranks, suggesting that the persistence or accumulation (dispersal) of old lineages (Coronado et al. 2015) 382

383 plays a role in generating the high degrees of overall dissimilarity across the taxonomic 384 ranks (Pennington et al. 2009). This interpretation is also supported by the higher 385 contribution of the nestedness component at the family and order levels in South American region than in the other regions (Fig. 4 and 5). Meanwhile, some regions (e.g. 386 387 African, Australian, and North American regions), regardless of whether they include tropical areas, exhibited a persistent steep dissimilarity-distance curve in regard to the 388 389 turnover component (Fig. S5), at both the family and order levels. This suggests 390 taxonomic diversification at deeper time scales (Prinzing et al. 2001) and/or the 391 persistence of different and older lineages in isolated sites (Tiffney & Manchester 2001; 392 Tolley et al. 2011) as a result of geohistorically related biogeographical constraints, e.g. 393 elevational gradients in the tropics (Qian & Ricklefs 2016) or climatic refugia (Tiffney & 394 Manchester 2001; Byrne 2008; Tolley et al. 2011).

395

396 In contrast, North American and both West and East Eurasian regions, comprising 397 temperate floras, showed a slower saturation in total dissimilarity and the turnover component with spatial and climatic distances, and a relatively higher contribution of the 398 399 nestedness component in shaping the dissimilarity patterns of angiosperm tree communities than in the remaining regions, all of them containing tropical areas (Figs. 3, 400 401 S4 and S5). In addition, the nestedness component was less dependent on either spatial 402 or climatic distance in these regions. These findings are consistent with the findings of previous studies of the temperate biota (Keil et al. 2012; Lenoir et al. 2012; Fitzpatrick et 403 al. 2013; Soininen et al. 2017; Antão et al. 2019), which suggest that the nestedness 404 405 component reflects the signal of historical processes that become more evident under unstable and harsh environmental conditions in higher latitudes (Baselga et al. 2012). 406

Indeed, temperate angiosperm tree assemblages have been shown to have experienced genus-level local extinction in response to Quaternary glaciations and/or global cooling in the Holarctic regions (Svenning 2003; Eiserhardt et al. 2015; Shiono et al. 2018), supporting the role of historical dispersal filters in shaping the nestedness-resultant dissimilarity of angiosperm tree communities in temperate forests.

412

413 In general, the negative exponential models fitted using spatial- and climatic-distance had 414 comparable explanatory power and were similar in terms of the shape of the dissimilarity-415 distance curves (Fig. 3, 4, and S6), suggesting that environmental filtering and/or 416 dispersal limitation have important roles in driving compositional turnover (Trujillo et al. 417 2019). However, a relatively lower explanatory power for the climatic distance model (i.e. it explained 11% less variance than the spatial distance model) was observed in the Indo-418 Pacific at the species level (Fig. 3). This suggests that taxonomic turnover in this region 419 420 is likely to be driven by dispersal limitation and/or other geographical factors, such as the 421 spatial separation between islands and continental landmasses, and variation in island 422 sizes (Ibanez et al. 2018). Moreover, the Indo-Pacific tropical forests are phylogenetically 423 similar to the East Eurasian temperate forests (Kubota et al. 2018), suggesting the regional divergence of the temperate flora originating from the Asian tropics (out-of-Asia 424 hypothesis; Donoghue 2008). Therefore, in the East Eurasian region, the highest 425 426 contribution of the turnover component compared with the other Holarctic regions (Fig. 4-5, S4-S5) may also be promoted by *in situ* diversification of angiosperm trees through 427 428 geographical vicariance related to high insularity and highly dissected topography (Xiang 429 et al. 2004; Kubota et al. 2014).

431 One potential caveat of analyzing assemblage dissimilarity between local plots is that 432 species occurrence data obtained in vegetation plots might potentially suffer from 433 sampling incompleteness (Beck et al. 2013). To deal with sampling bias, we computed sampling completeness using species relative abundance in individual plots (Chao et al. 434 435 2020) and assessed the spatial patterns of taxonomic dissimilarity by only analyzing nearly completely sampled plots (sampling completeness \geq 90%). Loosening of the SC 436 threshold down to 70% did not alter the general dissimilarity trends (Fig. S3), but it did 437 438 cause a slight reduction in the slope of the dissimilarity-spatial distance curve in some regions. This suggests that including incomplete plots in which common species are likely 439 440 to be well sampled but rare species are likely to be missed may overestimate similarity 441 among local communities. Another potential bias is related to the shortfall of taxonomic knowledge, especially at higher taxonomic levels (family or order). Indeed, the taxonomic 442 443 resolution of lineages differs among clades and regions, and in particular, is poorly 444 resolved for the tropics (Laffan 2018). For example, in Malesia, it is estimated that only 445 29% (of approximately 45,000) vascular plant species have been comprehensively treated 446 taxonomically in the Flora Malesiana, and while there are additional taxonomic 447 publications and treatments for this region, these are fragmented and overall the flora very much remains incompletely known and described (Middleton et al 2019). From the 448 449 viewpoint of filling gaps in our knowledge of plant biodiversity, further taxonomic and 450 systematic studies are needed to better understand the relative role of evolutionary events at different time scales in shaping the taxonomic dissimilarity of woody angiosperms 451 452 globally.

453

454 Concluding remarks

455 As with many macroecological patterns, the increase in dissimilarity with distance can be studied at multiple spatial scales (Nekola & White 1999; Wang et al. 2011; Fitzpatrick et 456 457 al. 2013; Olivier & van Aarde 2014; Kasel et al. 2017; Chun & Lee 2017; Trujillo et al. 2019). There have been many studies focused on the dissimilarity-distance pattern of 458 459 vegetation that measure beta diversity at different spatial extents, from local scales (Morlon et al. 2008; Wang et al. 2011; Wang et al. 2018) through to regional (Condit et 460 461 al. 2002; Tuomisto et al. 2003) and global-scales (Fitzpatrick et al. 2013; König et al. 462 2017). Despite these previous studies of beta diversity at local, regional and global scales, there are few examples of studies that use local community data to analyze large-scale 463 464 dissimilarity patterns (but see Myers et al. 2013; Kubota et al. 2018). In addition, beta 465 diversity patterns at multiple taxonomic scales were unexplored though it is potentially 466 informative to understand the imprints of deep-time diversification in extant diversity 467 patterns. The present study contributes to filling this knowledge gap by showing how 468 pairwise taxonomic dissimilarity and its components (calculated within biogeographical 469 regions) at different taxonomic ranks change across biogeographical regions through the analysis of local tree communities across the globe. Our findings of taxonomic 470 dissimilarity among angiosperm tree communities, which showed region-specific 471 variations in the dissimilarity-to-distance relationships across taxonomic ranks, reveal the 472 473 geographical pattern of diversification that is mechanistically driven by niche assembly 474 at higher taxonomic ranks (Ricklefs & Renner 2012), and global/regional-scale dispersal limitation (Hubbell 2001). 475

476

To conclude, our results generally supported our five predictions. First, taxonomic
turnover increased faster with spatial/climatic distance in those biogeographical regions

479 encompassing the tropics, i.e., in those areas where climatic conditions have been more 480 stable historically, compared to the temperate regions. Second, in general, the turnover 481 component decreased and its relationship with spatial/climatic distance became flatter at higher taxonomic ranks (order or family); this may reflect the evolutionary histories of 482 483 angiosperm trees associated with region-specific geohistories in the tropics and extratropics. However, and third, we also found relatively steep turnover patterns with 484 485 spatial/climatic distances in African, North American, and Australian regions at family 486 and/or order levels, which may be related to region-specific geographical constraints. 487 Fourth, the nestedness component was generally smaller than the turnover component 488 and almost independent from spatial/climatic distance in the regions containing tropical 489 areas at the species level. However, and fifth, the nestedness component comprised a 490 relatively larger proportion of overall dissimilarity in the Holarctic regions, which are often more historically unstable regions. In sum, the relationship between pairwise 491 492 dissimilarity and distance for angiosperm tree communities at species, genus, family, and 493 order levels illustrates the importance of geographical filters associated with historical and contemporary factors, in shaping regional beta diversity patterns of angiosperm trees. 494

495

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500

501 Data availability statement

502	The	relevant	data	used	in	this	study	is	available	from	Zenodo	(doi:
503	10.52	281/zenodo	.45724	04).								

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771	Supporting information
772	Appendix S1. Supplementary tables and figures
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774	

775 Figure legends

Fig. 1 The global distribution of forest plots across seven biogeographical regions: South
American, African, Indo-Pacific, Australian, North American, West Eurasian, and East
Eurasian. Plots were colored by sampling completeness (SC).

779

Fig. 2 Histograms for sampling completeness evaluated as sampling coverage (SC) per community in seven biogeographical regions (South American, African, Indo-Pacific, Australian, North American, West Eurasian, and East Eurasian). The equivalently wellsampled plots (SC \geq 0.9) were used in the dissimilarity-distance analyses.

784

Fig. 3 Pseudo r^2 of negative exponential models for the relationships between taxonomic dissimilarity and (a) geographical and (b) climatic distance per taxonomic rank [species (SP), genus (GN), family (FM) and order (OR)] in each biogeographical region: South American (SA), African (AF), Indo-Pacific (IP), Australian (AU), North American (NA), West Eurasian (WE) and East Eurasian (EE). Total dissimilarity matrices (β_{sor}) were decomposed into turnover (β_{sim}) and nestedness (β_{sne}) components. Dashed line represents 5% for a visual guide.

792

Fig. 4 The relationship between pairwise dissimilarity and spatial distance (dissimilarity -to-distance relationship) as fitted by a negative exponential model from presenceabsence composition data at the species, genus, family and order levels, in each biogeographical region: South American (SA), African (AF), Indo-Pacific (IP), Australian (AU), North American (NA), West Eurasian (WE) and East Eurasian (EE). Total dissimilarity matrices (β_{sor}) were decomposed into the turnover (β_{sim}) and nestedness-resultant (β_{sne}) components.

800

801	Fig. 5 Changes in the relative importance of the turnover component to total dissimilarity
802	$(\beta_{sim}/\beta_{sor})$ along geographical (left) and climatic (right) distance per taxonomic rank
803	[species (SP), genus (GN), family (FM) and order (OR)] in each biogeographical region:
804	South American (SA), African (AF), Indo-Pacific (IP), Australian (AU), North American
805	(NA), West Eurasian (WE) and East Eurasian (EE). Climatic distance is calculated as the
806	Euclidian distance between sites based on 19 bioclim variables and elevation. Locally
807	estimated scatterplot smoothing curves (LOESS) are shown.
808	

809

Fig. 1

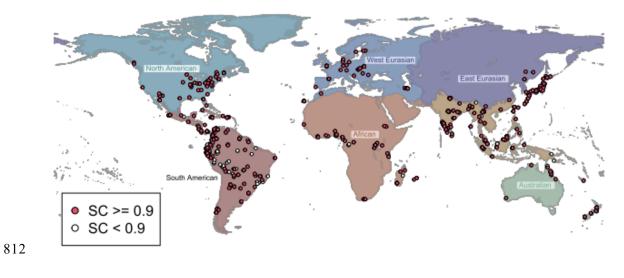


Fig. 2

