

1 **Angiosperm speciation speeds up near the poles**

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9 J.I. conceived the study and performed the analysis; J.I. and A.J.T interpreted the
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11

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26 **Abstract**

27 **Recent evidence has questioned whether the Latitudinal Diversity Gradient**
28 **(LDG), whereby species richness increases towards the Equator, results from**
29 **higher rates of speciation in the tropics. Allowing for time heterogeneity in**
30 **speciation rate estimates for over 60,000 angiosperm species, we found that the**
31 **LDG does not arise from variation in speciation rates because lineages do not**
32 **speciate faster in the tropics. These results were consistently retrieved using two**
33 **other methods to test the association between occupancy of tropical habitats and**
34 **speciation rates. Our speciation rate estimates were robust to the effects of both**
35 **undescribed species and missing taxa. Overall, our results show that speciation**
36 **rates follow an opposite pattern to global variation in species richness. Greater**
37 **ecological opportunity in the temperate zones, stemming from less saturated**
38 **communities, higher species turnover or greater environmental change, may**
39 **ultimately explain these results.**

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51 **Introduction**

52 Biodiversity on Earth is very unevenly distributed. The Latitudinal Diversity Gradient
53 (LDG), whereby species richness increases towards the Equator, is the most
54 prominent example of this unevenness. This pervasive pattern has fascinated
55 biologists for more than two centuries (von Humboldt 1808; Wallace 1878; Pianka
56 1966; Hawkins 2001), but its underlying causes remain largely unknown.

57

58 A major class of explanations proposes that elevated rates of speciation generate the
59 higher diversity in the tropics and so primarily create the LDG (Mittelbach *et al.*
60 2007). This increase in speciation is generally attributed to higher environmental
61 energy in the tropics, which in turn can hasten evolution through shorter generation
62 times and higher mutation rates (Dowle *et al.* 2013), and is known as the
63 “evolutionary speed hypothesis” (Rensch 1959; Rohde 1992). Other explanations
64 propose that elevated speciation rates near the Equator can stem from higher chances
65 of allopatric speciation caused by narrower thermal niches (Ghalambor *et al.* 2006), or
66 from stronger biotic interactions (e.g., faster coevolution) – the ‘Red Queen running
67 faster when she is hot’ (Brown 2014). However, while some studies on mammals
68 (Rolland *et al.* 2014), amphibians (Pyron & Wiens 2013) and butterflies (Condamine
69 *et al.* 2012) have shown that tropical lineages speciate faster, other studies have found
70 no latitudinal differences (Rabosky & Huang 2015; Economo *et al.* 2018) or higher
71 rates of speciation in temperate areas (Weir & Schluter 2007), particularly when
72 focusing on recent rates of speciation (Schluter 2016; Schluter & Pennell 2017;
73 Rabosky *et al.* 2018). Recent speciation rates should be less affected by extinction
74 (Nee *et al.* 1994) and more accurate than deep-time estimates (Louca & Pennell
75 2019), even in the absence of paleontological evidence (Marshall 2017). They should

76 also reflect any variable that historically influenced speciation and also varied
77 latitudinally, such as environmental energy (Title & Rabosky 2019).
78
79 Flowering plants are one of the largest eukaryotic radiations with roughly 350,000
80 species and show a marked LDG (Francis & Currie 2003). Despite this pattern,
81 latitudinal differences in macroevolutionary rates in angiosperms have received
82 comparatively little attention at a species-level. Previous efforts have shown that
83 angiosperm diversification is faster in tropical families (Jansson & Davies 2008) and
84 in tropical lineages within a limited number of clades (Svenning *et al.* 2008; Jansson
85 *et al.* 2013). The only species-level analysis found no differences in speciation, but
86 did so by using a method that assumed that all rate differences in the angiosperm Tree
87 of Life (ToL) were due to geographic distribution alone (Antonelli *et al.* 2015). This
88 method ignores any other sources of rate variation and has been shown to be
89 inadequate, especially for large phylogenies where substantial heterogeneity in
90 diversification rates is expected (Caetano *et al.* 2018). One reason for the lack of
91 attention to angiosperms is that their vast diversity makes it difficult to obtain well-
92 sampled phylogenetic and distributional data at a species-level, which is required to
93 estimate speciation rates accurately. Here we tested how speciation rates varied with
94 latitudinal distribution in over 60,000 angiosperm species using different methods that
95 accommodated macroevolutionary rate heterogeneity and incomplete taxon sampling.
96 Our analyses used the largest phylogenetic trees presently available for angiosperms
97 (Zanne *et al.* 2014; Smith & Brown 2018), alongside all geographic distribution data
98 obtainable for these species (Mounce *et al.* 2017) from the Global Biodiversity
99 Information Facility (GBIF). We show that latitudinal variation in recent speciation
100 rates cannot generate the LDG because speciation was not faster in the tropics.

101

102 **Material and Methods**

103 *Phylogenetic datasets and macroevolutionary rate estimates*

104 We estimated speciation rates in the angiosperm radiation with Bayesian Analysis of
105 Macroevolutionary Mixtures (BAMM) v.2.5.0 (Rabosky 2014). BAMM detects
106 heterogeneity in speciation and extinction rates through time and across lineages. The
107 input for BAMM was the most comprehensive time-calibrated phylogenetic tree for
108 seed plants that was built using a hierarchical clustering analysis of Genbank
109 sequence data (Smith & Brown 2018). On average, sister branches in this phylogeny
110 had an overlap of 1792 base pairs, which corresponds to roughly one or two gene
111 regions. We used a modified version of the tree containing GenBank taxa with a
112 backbone provided by Open Tree of Life (i.e., GBOTB) for 72,986 flowering plants.
113 Species names were standardised to The Plant List v.1.1 (Jin & Qian 2019).
114 Following [Igea et al. 2017](#), convergence of such a large BAMM analysis was
115 achieved by dividing the initial tree into clades of less than 6000 species. Doing so
116 resulted in 17 clades (16 monophyletic clades and 1 clade that contained the backbone
117 of the tree plus the unassigned species), which we then used to run independent
118 BAMM speciation/extinction analyses. Within each clade, non-random incomplete
119 taxon sampling was incorporated into the estimation process by calculating the
120 number of species sampled in each family. Prior settings were obtained using the
121 *setBAMMpriors* function in the R package *BAMMtools* (Rabosky et al. 2014) and
122 *expectedNumberOfShifts* was set at 50. Analyses were run for a minimum of 100
123 million generations and $\geq 20\%$ of each run was discarded as a burn-in when necessary.
124 Effective samples sizes of the log-likelihood and the number of rate shifts were both

125 above 200. The event files for all 17 clades were then combined to generate a single
126 BAMM result file that was used for all downstream analyses.

127

128 Although the reliability of BAMM estimates has recently been questioned (Moore *et*
129 *al.* 2016), simulations have shown that accurate estimates of speciation can be
130 obtained even in the absence of fossil data (Rabosky *et al.* 2017; Mitchell *et al.* 2018).
131 Tip-based estimates of speciation like the ones we use here should also be robust to
132 the effect of extinction and are increasingly used when analysing the effect of traits on
133 lineage diversification and summarising geographic patterns of speciation rate
134 variation (Title & Rabosky 2019). BAMM extinction estimates are thought to be more
135 unreliable (Rabosky 2010; Marshall 2017; Mitchell *et al.* 2019), so we do not discuss
136 them here. BAMM estimates of diversification rates, which are calculated as the net
137 outcome of speciation minus extinction (Title & Rabosky 2019), will therefore also be
138 more unreliable than speciation rates. Hence, and following other similar papers
139 (Rabosky *et al.* 2018), we only discuss BAMM estimates of speciation. Furthermore,
140 we also estimated recent speciation rates using the Diversification Rate metric (DR;
141 Jetz *et al.* 2012, Supplementary Note 2), which only considers branch lengths and
142 splitting events and does not accommodate for incomplete sampling. We also focus
143 on estimates of speciation and extinction rates from other methods in subsequent
144 sections.

145

146 *Latitudinal datasets*

147 We obtained latitudinal data for 236,894 angiosperms from a study that estimated the
148 median, maximum and minimum latitude for each species (Mounce *et al.* 2017) using
149 observations from GBIF. Since our goal was to assess whether speciation rates drive

150 the LDG, we focused on latitudinal differences and did not consider regional and
151 longitudinal differences in diversity, which may be important in some cases (Reyes *et*
152 *al.* 2015; Xing & Ree 2017; Igea & Tanentzap 2019). We then intersected these data
153 with our BAMM dataset ($n = 72,986$) to obtain estimates of speciation rate and
154 latitude for as many angiosperms as possible, hereafter referred to as the ‘full’ dataset
155 ($n = 60,990$). The species names in the GBIF datasets were standardised with The
156 Plant List (TPL) using the *Taxonstand* (Cayuela *et al.* 2012) and *Taxonlookup*
157 (Pennell *et al.* 2016) packages before collation.

158

159 Tropical and temperate species were defined as those where the absolute median
160 latitude was below and above 23.5° , respectively. Species were also grouped into
161 equal-width latitudinal bins according to their median latitude to define latitudinal
162 bands. To ensure that these latitudinal bands were completely tropical or temperate,
163 the width of each band was set at 9.4° (i.e., starting from a band centered in the
164 Equator from -4.7° to 4.7° , then from 4.7° to 14.1° , 14.1° to 23.5° , etc).

165

166 *BAMM-based correlation of tropicality and speciation*

167 We correlated species-specific (i.e., tip-based) speciation rates from BAMM with the
168 latitudinal data using Structured Rate Permutations on Phylogenies (STRAPP)
169 (Rabosky & Huang 2016). STRAPP assesses the significance of the empirical
170 association of macroevolutionary rates and phenotypic traits by comparing it to a null
171 distribution generated by permuting the speciation rates across the phylogeny while
172 maintaining the position of the rate shifts in the phylogenetic tree. Crucially, this test
173 does not require that all rate variation is caused by the focal trait and thus shows a
174 reduced Type I error rate (Rabosky & Huang 2016). We performed all STRAPP two-

175 tailed tests using the *traitdependentBAMM* function of the *BAMMtools* package
176 (Rabosky *et al.* 2014) with 1000 replicates and logging the rates. We assessed
177 whether *i*) tropical and temperate species had different speciation rates using a Mann
178 Whitney U test; *ii*) speciation rates for each of the latitudinal bands were different
179 using the Kruskal-Wallis rank sum statistic; and *iii*) species absolute median latitude
180 was correlated with speciation rate using Spearman's rank correlation coefficient.

181

182 We repeated the analyses discarding sets of species that may bias our results. First, we
183 used only densely sampled species (i.e., with five or more data points in the GBIF
184 dataset). Second, we discarded widespread species (i.e., neither strictly tropical nor
185 strictly temperate) for analyses comparing strict tropical and temperate taxa. Strictly
186 tropical and temperate species were defined as species with median, maximum and
187 minimum latitude all occurring either inside or outside the tropics, respectively. Third,
188 we discarded species with median latitudes in the highest latitudinal bands ($>50^\circ$ and
189 $<-50^\circ$), which contained few species with extreme estimates of λ .

190

191 *Assessing the reliability of the BAMM results*

192 We assessed the consistency of the BAMM results in the face of geographic biases,
193 low sampling and topological uncertainty.

194

195 First, the full dataset had more temperate than tropical species and so was inconsistent
196 with the LDG. Poorly sampled clades may have average longer branches and lower λ ,
197 and so our speciation rate estimates may be affected by latitudinal sampling bias. We
198 therefore tested whether clades with lower sampling fractions were found at lower
199 latitudes and had lower λ rates. We employed two additional strategies to analyse the

200 effect of the disproportionate amount of temperate species in our BAMM analyses.
201 We generated 100 random samples of our full dataset (herein ‘unbiased’ datasets),
202 each with 30,000 species that maintained the proportion of tropical (57.5%) and
203 temperate (42.5%) species present in GBIF. We also generated 100 random samples
204 (herein ‘extreme tropical’ datasets) that assumed all species absent from the GBIF
205 dataset of Mounce *et al.* (2017) were tropical, so the proportions of tropical and
206 temperate species were 70.9% and 29.1%, respectively. We derived these values by
207 estimating that 109,461 species were absent from GBIF by subtracting the number of
208 species in the GBIF dataset from the number of species in The Plant List according to
209 *TaxonLookup* (Pennell *et al.* 2016). Although both the unbiased and extreme tropical
210 datasets were generated to assess the effect of excess temperate species, the extreme
211 tropical dataset also represented the most extreme scenario of undescribed
212 biodiversity. The sizes of these datasets was fixed at 30,000 species so that the
213 tropical species present in each subsample were not excessively redundant when
214 obtained from a total pool of 26,873 species by random draws of 17,130 and 21,273
215 species for the unbiased and extreme tropical datasets, respectively. Using the two
216 datasets, we then assessed whether the full dataset λ estimates were affected by poorer
217 sampling for tropical species. We did so by selecting 10 random replicates of the
218 30,000-species subsampled datasets and rerunning the BAMM speciation analyses as
219 detailed above. We determined whether the relationship of the full with both the
220 unbiased and extreme tropical λ estimates was different for tropical and temperate
221 species. To do this, we predicted the full λ estimates using the subsampled λ
222 estimates, ‘tropicality’ (i.e. a binary variable indicating if a species was exclusively
223 tropical based on our aforementioned definition) and the interaction of the λ estimates
224 and tropicality. We fitted this model separately for each of the 10 replicates in the

225 unbiased and extreme tropical datasets, incorporating the effect of the phylogeny
226 using the *phylolm* package (Tung Ho & Ané 2014). We also tested whether the
227 association between speciation and latitude in the full dataset stemmed from the
228 excessive proportion of temperate species by repeating the STRAPP analyses with the
229 100 replicates in the unbiased and extreme tropical datasets.

230

231 Second, our full dataset contained λ estimates for 60,990 species, which represent
232 17.6% of described angiosperm species (total $n = 346,365$ calculated with
233 *Taxonlookup*). To assess the reliability of λ estimates given the large number of
234 missing taxa, we randomly generated 10 subsamples ('small' datasets) with the same
235 proportion (i.e., 17.6%) of the species in our full dataset ($n = 10,739$). As above, we
236 ran BAMM for each of the 10 subsamples and predicted the full λ estimates using the
237 small λ estimates.

238

239 Third, we analysed a single phylogenetic tree, which may only represent one of many
240 potential hypotheses for the angiosperm ToL. To assess the effect of the unaccounted
241 topological and branch length uncertainty, we tested the relationship between
242 speciation rates and latitude in a smaller dataset of 28,057 species obtained from an
243 independently estimated phylogenetic tree. The tree was estimated with GenBank
244 sequences for seven gene regions with a median per-taxon proportion of missing data
245 of 0.60 (Zanne *et al.* 2014; Qian & Jin 2016). We obtained BAMM λ rates for this
246 dataset as described elsewhere (Igea *et al.* 2017). We then tested the correlation
247 between the BAMM estimates of our full dataset and this reduced dataset. We also
248 used the reduced dataset to determine whether there were differences in BAMM λ
249 rates between tropical and temperate species and among different latitudinal bands;

250 and whether absolute median latitude and speciation rates were associated. All of
251 these analyses were performed as described above.

252

253 *Geography-dependent speciation*

254 We used the Geographic Hidden-State Speciation and Extinction (GeoHiSSE)
255 framework to test for further differences in speciation between tropical and temperate
256 lineages. GeoHiSSE estimates speciation, extinction and transition rates for two
257 geographic areas while incorporating rate heterogeneity that is independent of
258 geography by including unobserved characters known as “hidden states” (Caetano *et*
259 *al.* 2018). Including these hidden states is crucial to create biologically meaningful
260 null models where diversification varies independently of the focal traits. We
261 considered any species occurring strictly below and above 23.5° as tropical and
262 temperate respectively, while species whose range span both >23.5° and <23.5° were
263 classified as widespread. Sampling fractions for each geographic state in the
264 phylogenetic tree were estimated to be 0.12, 0.30, and 0.24 for strictly tropical,
265 strictly temperate, and widespread species, respectively. Following Caetano *et al.*
266 2018, we fitted 12 different models (numbers 1-12 in Caetano *et al.* 2018, Table 1).
267 Some of these were “null models” where no diversification rate variation is caused by
268 geography. Other models included area-dependent and area-independent
269 diversification with two to five hidden states, and some also separated range
270 contraction from lineage extinction (i.e., “+extirpation models”). We then selected the
271 best fitting model using the sample-sized corrected Akaike Information Criterion
272 (AICc). We used the “fast version” of the GeoHiSSE model as implemented in *hisse*
273 v.1.9.6 (Beaulieu & O’Meara 2016) to make analyses feasible in such a large
274 phylogeny. The size of the dataset meant that some of the models took a very long

275 time to fit (2 months on 2.5 GHz processor for a single fit). We could therefore not
276 obtain rate estimates across the nodes and tips using the marginal reconstruction
277 algorithm, which required estimating probabilities for all possible combinations of
278 states at each node in the tree given all combinations at nodes in the rest of the tree. It
279 was also not feasible to model-average the “effective state proportion” (i.e., the
280 estimated proportion of each geographic area and hidden state combination), which
281 would have allowed us to test if a only a small proportion of the lineages were
282 associated with a geographic-dependent diversification process (Caetano *et al.* 2018).

283

284 *Clade-based analyses*

285 We estimated clade-level measures of speciation across the angiosperm phylogenetic
286 tree. We used a set of 4 million-year-wide time intervals from 0 to 24 million years
287 (myr) to define the ages of the clades. Clades were delimited as the largest non-
288 overlapping monophyletic groups of four or more species where sampling was larger
289 than 30%. This criterion ensures that the probability of retrieving the correct crown
290 age of the clade is above 70% (Rabosky 2016). At least 50% of the species in the
291 clades also had to have latitudinal data. We used RPANDA (R: Phylogenetic
292 ANalyses of DiversificAtion, Morlon *et al.* 2016) to fit six different models of
293 diversification: *i*) constant speciation (λ) and extinction (μ) fixed at 0; *ii*) constant λ
294 and μ ; *iii*) exponential λ and μ fixed at 0; *iv*) exponential λ and constant μ ; *v*) constant
295 λ and exponential μ ; and *vi*) exponential λ and μ . Incomplete sampling within clades
296 was incorporated into the estimates by calculating the proportion of sampled species
297 within the genera in each clade. Speciation rates were estimated using model
298 averaging with AIC weights. The correlation of the clade-level estimates of speciation
299 and the proportion of temperate and tropical species in each clade was assessed using

300 phylogenetic least squares regression as implemented in the R package *nlme* (Pinheiro
301 *et al.* 2014). The lambda parameter was optimised using a maximum likelihood
302 approach. We weighted the regressions by the proportion of GBIF data points in each
303 clade to account for different sampling levels. We also assessed if potential latitudinal
304 differences in sampling proportions could explain our results. We did so by testing if
305 the sampling proportions of largely tropical and largely temperate clades (defined as
306 clades where >50% of the species were tropical and temperate, respectively) were
307 different across the 6 time intervals using Pearson's chi-squared test.

308

309 **Results**

310 We found that tip-specific speciation rates were consistently smaller in the tropics for
311 60,990 species across the whole angiosperm radiation. These results were repeatable
312 with several ways of classifying species as tropical or temperate. Using correlation
313 tests that controlled for phylogenetic pseudoreplication (Rabosky & Huang 2016), we
314 found that tropical species had smaller mean speciation rates (λ) than temperate
315 species (Fig 1a; $\lambda_{\text{temperate}} = 0.73$ species/myr, $\lambda_{\text{tropical}} = 0.65$ species/myr, p-value =
316 0.043). Angiosperms in latitudinal bands closer to the poles (i.e. absolute median
317 latitude >42.3°) speciated at faster rates than in any tropical latitudinal band (Fig. 1b).
318 The absolute median latitude occupied by individual species was also positively
319 correlated with their speciation rate (Fig. 1c; Spearman's $\rho = 0.086$, p-value = 0.009).
320 These results were maintained when we: *i*) analysed only those species that were
321 densely sampled in GBIF ($n = 53,344$ species) and *ii*) discarded the highest latitudinal
322 bands (>50° N and S), which contained fewer species ($n = 60,034$) (Table S1).
323 Speciation rates were also different when comparing strictly tropical and strictly
324 temperate species by discarding species that occurred both inside and outside the

325 tropics ($n = 46,426$), but this difference was not significant (p -value = 0.064; Table
326 S1). .
327
328 Our full dataset contained a disproportionate amount of temperate species and so did
329 not hold a LDG (Fig. S1), but this bias could not explain the consistent association
330 between latitude and speciation rates. First, we generated 100 “unbiased” datasets that
331 maintained the proportion of tropical and temperate species present in GBIF and were
332 consistent with the LDG. Second, we generated another set of 100 “extreme tropical”
333 datasets, where the proportions of temperate and tropical species were 29.1% and
334 70.9%, respectively. Rerunning BAMM for 10 of the 100 unbiased and extreme
335 tropical datasets, we found that the λ estimates were very positively correlated with
336 those from the full dataset (median $\rho = 0.82$, median p -value < 0.001 for both the
337 unbiased and the extreme tropical datasets; Figs 2, S2). This correlation between λ of
338 the full and subsampled datasets did not differ between tropical and temperate
339 species, as the effect size of the interaction between tropicality and the subsampled λ
340 estimates was not different from zero across all the 10 replicates for the unbiased
341 (mean = -0.0005, standard deviation = 0.0017) and the extreme tropical (mean = -
342 0.0015, standard deviation = 0.0024) datasets. As with the full dataset, we also found
343 that speciation rates in the tropics were consistently lower and that λ was positively
344 correlated with absolute median latitude using STRAPP across the 100 replicates of
345 both the unbiased (Fig. 3) and extreme tropical (Fig. S3) datasets. Finally, we
346 confirmed that our results could not be explained by better sampled temperate species
347 having higher λ . Densely sampled clades were only slightly more frequent at higher
348 latitudes (Fig. S4), and they had smaller estimates of λ (Fig. S5), contrary to what we
349 might expect if our results were a sampling artefact.

350

351 We also confirmed that the speciation rate estimates were robust to the low levels of
352 taxon sampling. We compared the BAMM λ estimates from the full dataset with
353 estimates generated by randomly sampling 17.6% of the species in our full dataset 10
354 times and rerunning BAMM. This level of sampling reflected the proportion of
355 angiosperm species included in our full dataset from those described in The Plant List.
356 The resulting λ estimates from the full and the small datasets were positively
357 correlated (median $\rho = 0.67$, median p-value < 0.001). This association was slightly
358 weaker for tropical than for temperate species, but this difference was very small and
359 not statistically significant across the 10 replicates (mean = -0.04, standard deviation
360 = 0.03; Fig. S6).

361

362 Our results were based on a single phylogenetic tree that represented one of many
363 possible topologies in the angiosperm ToL, and so we confirmed that the results were
364 consistent with another independently derived phylogeny (Zanne *et al.* 2014; Qian &
365 Jin 2016). Using BAMM λ estimates and latitudinal information for this 28,057
366 species tree, we found that λ was higher outside the tropics and that speciation rates
367 increased towards the poles (Fig. S7). The BAMM λ estimates of this topology and
368 our full dataset were positively correlated (Spearman's $\rho = 0.46$, p-value < 0.001).

369

370 Elevated speciation rates in the tropics were also rejected as the driving force of the
371 LDG using a method that explicitly tested the influence of geography on species
372 diversification. The best fitting model with GeoHiSSE (AICc weight = 1; Table 1)
373 had five different hidden states that accommodated rate variation across lineages and
374 no differences in turnover and extinction fraction between tropical and temperate

375 regions. This result suggests that most of the variation in diversification rates across
376 the angiosperm radiation is not linked to latitude, though some angiosperm lineages
377 can show range-dependent diversification. We were precluded from exploring this
378 latter process using marginal reconstructions by the sheer size of the dataset (see
379 Methods). The confidence intervals around the maximum-likelihood estimates were
380 too wide to identify any differences in rate estimates, so we do not discuss them
381 further.

382

383 Finally, we found that tropical species did not speciate faster across clades of different
384 ages in the angiosperm phylogeny. Using clades in six 4 million-year-wide time
385 slices, from 0 to 24 million years, where we were confident in speciation rates and
386 GBIF coverage, we fitted models that implemented time-variable and time-constant
387 speciation and extinction. Mirroring the BAMM results, clades with a larger
388 proportion of temperate species had higher rates of speciation in the best fitting model
389 (Fig S8). This result was not totally explained by sampling biases. Temperate clades
390 had significantly denser average sampling only in the 8-12 and 12-16 myr intervals
391 ($\chi^2 = 5.22$, $df = 1$, $p\text{-value} = 0.020$ and $\chi^2 = 12.19$, $df = 1$ $p\text{-value} < 0.001$,
392 respectively), and these differences were small (0.65 vs 0.62 and 0.65 vs 0.61 for
393 temperate vs tropical in each period, respectively).

394

395 **Discussion**

396 Using data for over 60,000 flowering plants, our findings rejected the long-standing
397 notion that the greater species diversity of the tropics can be explained by higher
398 recent speciation rates (Mittelbach *et al.* 2007), such as from higher environmental
399 energy (Dowle *et al.* 2013). We instead found higher rates of recent speciation closer

400 to the poles. An explanation for this difference is that temperate biotas have fewer
401 species, so their niche space may be less saturated, thereby increasing opportunities
402 for lineage divergence (Simpson 1953; Schluter 2016). Similarly, reproductive
403 isolation may be elevated at higher latitudes by the greater ecological opportunity
404 generated from recurrent environmental change and climate instability (Cutter & Gray
405 2016). Other explanations may be related to the ecological and life history traits of
406 temperate species. For example, faster speciation rates at higher latitudes could stem
407 from the higher frequency of small-seeded species (Moles *et al.* 2007). Small seed
408 size is positively correlated with angiosperm diversification (Igea *et al.* 2017), and
409 temperate species did have smaller seeds in our full dataset (Supplementary Note 1).
410 Therefore, our results suggest that latitudinal differences in recent speciation rates do
411 not shape the LDG but are shaped by differences in species diversity and traits (Weir
412 & Price 2011; Schluter & Pennell 2017).

413

414 Rejecting the role of speciation rate as the cause of the LDG calls for alternative
415 explanations (Antonelli *et al.* 2015). Many theories have been proposed to explain the
416 LDG, but they can all be divided into four broad categories arising from differences
417 in: *i*) diversification rates; *ii*) dispersal rates; *iii*) time for species accumulation and *iv*)
418 ecological limits to species diversity (Etienne *et al.* 2019). Hypotheses in the different
419 categories often invoke multiple causal mechanisms and so may not be mutually
420 exclusive (Pontarp *et al.* 2019). First, the elevated speciation near the poles could be
421 coupled with higher extinction rates to produce a scenario where temperate zones act
422 as both a “cradle” and “grave” of biodiversity (Cutter & Gray 2016). Net
423 diversification might therefore still be higher in tropical areas despite the absolute
424 rates of speciation and extinction each being smaller than closer to the poles. Our

425 GeoHiSSE analyses do not, however, support this scenario as we found no such
426 differences in turnover (Table 1). Estimates of extinction rate from molecular
427 phylogenies in the absence of fossil data are often unreliable (Marshall 2017; Mitchell
428 *et al.* 2018), which further complicates direct tests of this hypothesis. Second, the
429 tropical conservatism hypothesis proposes that lower net migration from the tropics to
430 temperate zones contributes to the LDG (Wiens & Donoghue 2004), because most
431 lineages originated in the tropics and have been unable to disperse into harsher,
432 temperate environments. For example, in New World woody angiosperms, transitions
433 between tropical and temperate environments are rare and most temperate lineages are
434 relatively young (Kerkhoff *et al.* 2014). This scenario was also not supported with
435 GeoHiSSE as we detected no differences in transition rates between the tropics and
436 the temperate zones and vice versa. Third, if rates of diversification and migration
437 have no latitudinal variation, the greater diversity of the tropics may stem from their
438 larger area through time, i.e. the age-and-area hypothesis (Fine & Ree 2006). For
439 instance, current species richness across the world's 32 bioregions is best predicted by
440 area and productivity over geological time (Jetz & Fine 2012). Area-through-time
441 differences have also been proposed to underlie differences in diversity among
442 tropical regions (Couvreur 2015). Fourth, a whole class of hypotheses posits that there
443 are ecological limits to the number of species that a region can hold, and that these
444 limits vary with latitude (Rabosky 2009). For example, the larger diversity in the
445 tropics could be explained by larger niche spaces and greater niche packing (Pellissier
446 *et al.* 2018). Such a mechanism could interact with differential diversification rates to
447 amplify spatial differences in diversity (Etienne *et al.* 2019). As ecological and
448 macroevolutionary processes can interact, integrating all of them into a mechanistic
449 eco-evolutionary framework can ultimately advance the search for explanations to the

450 LDG (Pontarp *et al.* 2019). Other ecological interactions, like negative conspecific
451 density dependence arising from herbivores and pathogens, have also been proposed
452 to explain the high diversity in the tropics (LaManna *et al.* 2017).

453

454 Recent findings have suggested that rates of species origination are highest in species-
455 poor areas (Quintero & Jetz 2018; Rabosky *et al.* 2018). Our results are consistent
456 with these findings and therefore suggest that latitudinal variation in recent speciation
457 rates is not a major engine of global diversity patterns in one of the largest eukaryotic
458 radiations. Using time-heterogeneous estimates of speciation, we were able to exclude
459 faster rates of evolution, such as arising from more favourable environmental
460 conditions or more intense biotic interactions, as responsible for shaping the
461 angiosperm LDG. Future work must now resolve whether recent rapid speciation
462 nearer the poles is widespread across the ToL and its underlying causes.

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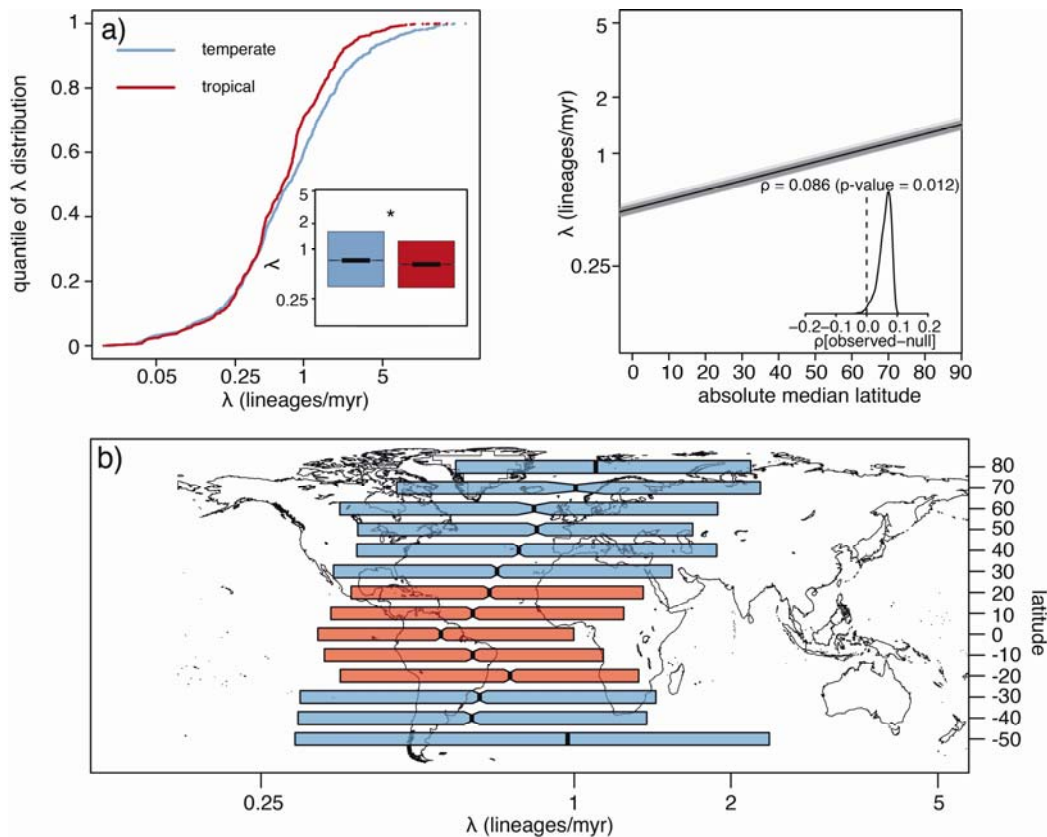
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476 **Figure 1. Tropical species have smaller speciation rates (λ) than temperate**

477 **species. a) Rank-ordered distribution and boxplot (inset) of tip λ inferred with**

478 **BAMM for tropical (red) and temperate (blue) species. * indicates significant**

479 **difference (p-value = 0.043) between the two groups, assessed with a Mann-Whitney**

480 **test implemented in STRAPP. b) λ grouped by latitudinal band of each species.**

481 **Notches in boxplots indicate 95% confidence intervals around median, denoted by**

482 **thick vertical lines and boxes span the interquartile ranges. c) Spearman's ρ**

483 **correlation between species absolute median latitude and λ as estimated with**

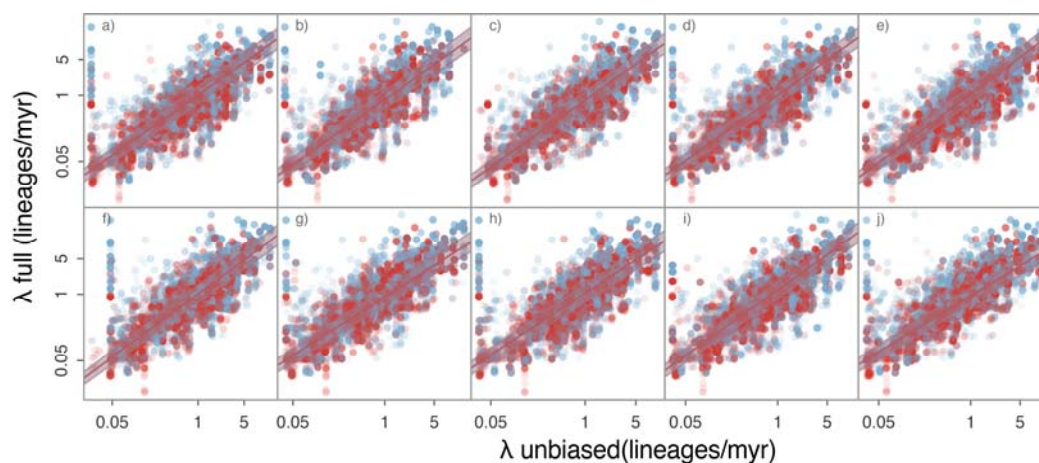
484 **STRAPP. Grey lines are correlations across 1000 samples of the posterior distribution**

485 **estimated with BAMM. Solid black line indicates median correlation \pm 95%**

486 **confidence interval across the posterior distribution. Inset shows the difference**

487 **between the empirical and null correlations estimated with 1000 permutations of**

488 **evolutionary rates across the phylogeny.**



489

490 **Figure 2. BAMM λ estimates were positively correlated in the full and 10**
491 **unbiased datasets (a to j) with no effect of tropicality.** Solid lines are phylogenetic
492 linear regressions predicting λ in the full tree ($n = 60,990$ species) with λ in the
493 unbiased tree ($n = 30,000$ species) in temperate (shown in blue) and tropical (shown
494 in red) species. Shaded areas indicate the 95% confidence intervals.

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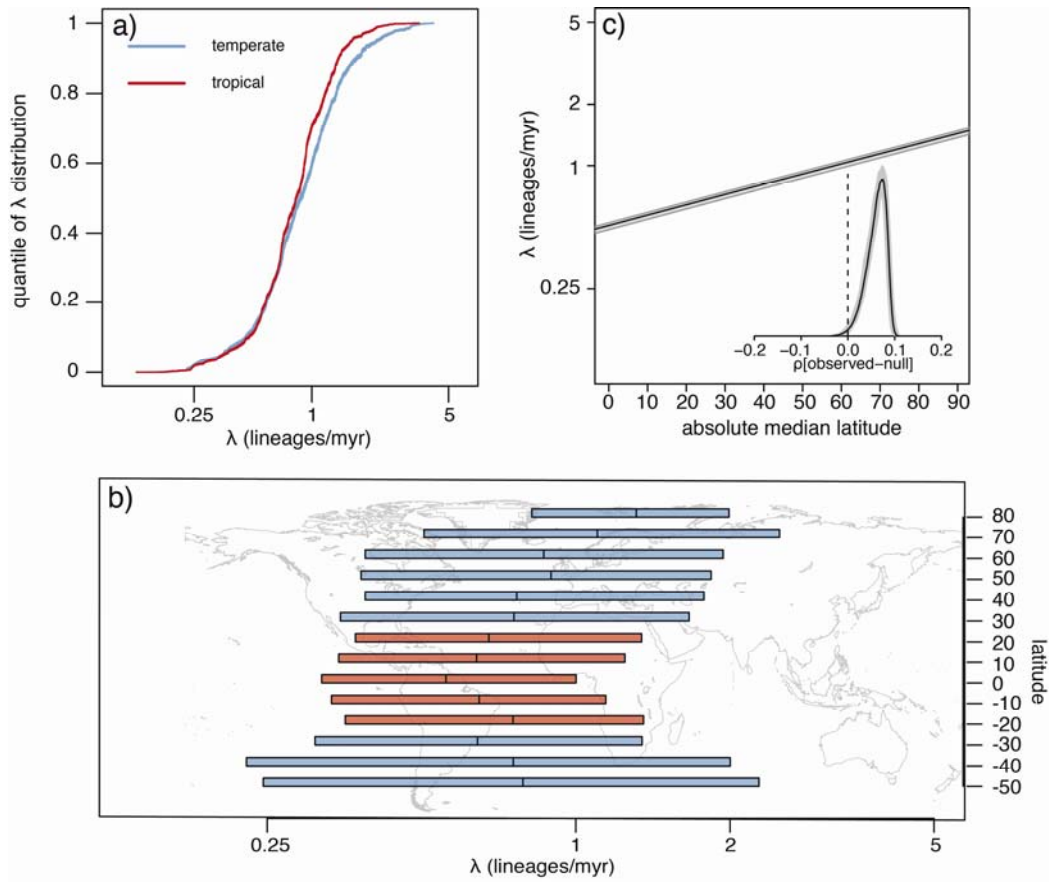
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509 **Figure 3. Temperate species have higher speciation rates (λ) than tropical species**

510 **for 100 subsampled “unbiased” datasets.** Datasets were generated whereby 42.5%

511 and 57.5% of species were temperate and tropical, respectively. **a)** Rank-ordered

512 distribution of λ inferred with BAMM for tropical (red) and temperate (blue) species.

513 **b)** λ grouped by latitudinal band of each species. Boxes show the average interquartile

514 range across the 100 subsampled datasets, and lines show the average medians. **c)**

515 Spearman’s ρ correlation of species absolute median latitude and λ as estimated with

516 STRAPP. Solid black line indicates median correlation \pm 95% confidence interval

517 across the posterior distribution. Inset shows the median (black line) \pm 95%

518 confidence interval for the difference between the empirical and null correlations

519 estimated with 1000 permutations of evolutionary rates across the phylogeny in each

520 of 100 subsampled datasets.

521 **Tables**

522 **Table 1.** Model comparison of 12 GeoHiSSE models with up to five hidden rate
 523 categories. Model numbers and descriptions refer to Caetano *et al* (2018). Null
 524 models have no differences in diversification associated with geography and
 525 diversification parameters are constrained to be equal between areas in the same
 526 hidden state category. Area independent models (CID) have no geographic-dependent
 527 differences in diversification. Full models have no constrained parameters. Extremely
 528 high AICc values for the three simplest models suggest that they were a very poor fit.
 529 Original GeoSSE refers to the formulation in Goldberg *et al.* (2011).

530

model	description	# free parameters	AICc
5	CID-GeoHiSSE, 5 hidden rate classes, null model	13	418090
11	CID-GeoHiSSE+extirpation, 5 hidden rate classes, null model	15	423888
10	GeoHiSSE+extirpation, 2 hidden rate classes, full model	19	425927
9	CID-GeoHiSSE+extirpation, 3 hidden rate classes, null model	11	428085
12	CID-GeoHiSSE+extirpation, 2 hidden rate classes	9	429549
4	GeoHiSSE, 2 rate classes, full model	15	430692
3	CID-GeoHiSSE, 3 hidden rate classes, null model	9	442036
8	GeoSSE+extirpation, full model	9	458691
2	Original GeoSSE, full model	7	470420
1	CID-original GeoSSE	4	1.56E+19
6	CID-GeoHiSSE, 2 hidden rate classes	7	1.56E+19
7	CID-GeoSSE+extirpation	6	1.56E+19

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536 **SUPPLEMENTARY MATERIAL**

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dataset name	$\lambda_{\text{temperate}}$	$\lambda_{\text{tropical}}$	p-value	dataset size
more.GBIF.data	0.702	0.641	0.050	53,344
no.widespread	0.760	0.668	0.064	46,426
no.high.latitude	0.719	0.652	0.044	60,034

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539 **Table S1.** Temperate species have mean higher speciation rates (λ , estimated with
540 STRAPP) than tropical species across different datasets: “more.GBIF.data” excludes
541 species with less than five GBIF data points; “no.widespread” excludes species
542 occurring both in and outside the tropics; “no.high.latitude” excludes species from
543 poorly sampled latitudinal bands with absolute median latitude $\geq 50^\circ$. Units for λ and
544 dataset size are lineages/myr and species, respectively.

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546 **Supplementary Note 1**

547 Smaller seeded species have been shown to speciate faster than large seeded species
548 (Igea *et al.* 2017). We assessed whether seed size differed between temperate and
549 tropical lineages as it has previously been shown to do so (Moles *et al.* 2007), and so
550 latitudinal variation in seed size could explain the differences in speciation between
551 tropical and temperate species. We obtained seed size measurements for 13,178
552 species in our full dataset from a previous study (Igea *et al.* 2017) and we found that
553 seed size was larger in tropical species (mean seed size_{temperate} = 0.0018 g; mean seed
554 size_{tropical} = 0.0125 g; $t = -28.239$; $df = 4300.7$; $p\text{-value} < 0.001$). This difference
555 remained when phylogeny was considered (phylANOVA: $t = 33.933$; $p\text{ value} =$
556 0.001; significance assessed with 1,000 random simulations using phytools (Revell
557 2012))

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559 **Supplementary Note 2**

560 BMM is a model-based approach to estimate tip speciation rates. To compare the
561 BMM results with a non-model based approach (i.e., that only relies on branch
562 lengths and splitting events), we calculated the Diversification Rate metric (DR; Jetz
563 *et al.* 2012) for the species in our full dataset ($n = 60,990$). This measure of recent
564 speciation rate incorporates the number of nodes and the internode distances
565 separating a species from the root and gives greater weight to branches closer to the
566 present (Title & Rabosky 2019). We then compared the DR values of tropical and
567 temperate species and assessed the variation of DR across latitudinal bands. Mirroring
568 the BMM results, we found that tropical species had smaller values of DR and that
569 latitudinal bands closer to the poles had higher values of DR (Fig. S9).

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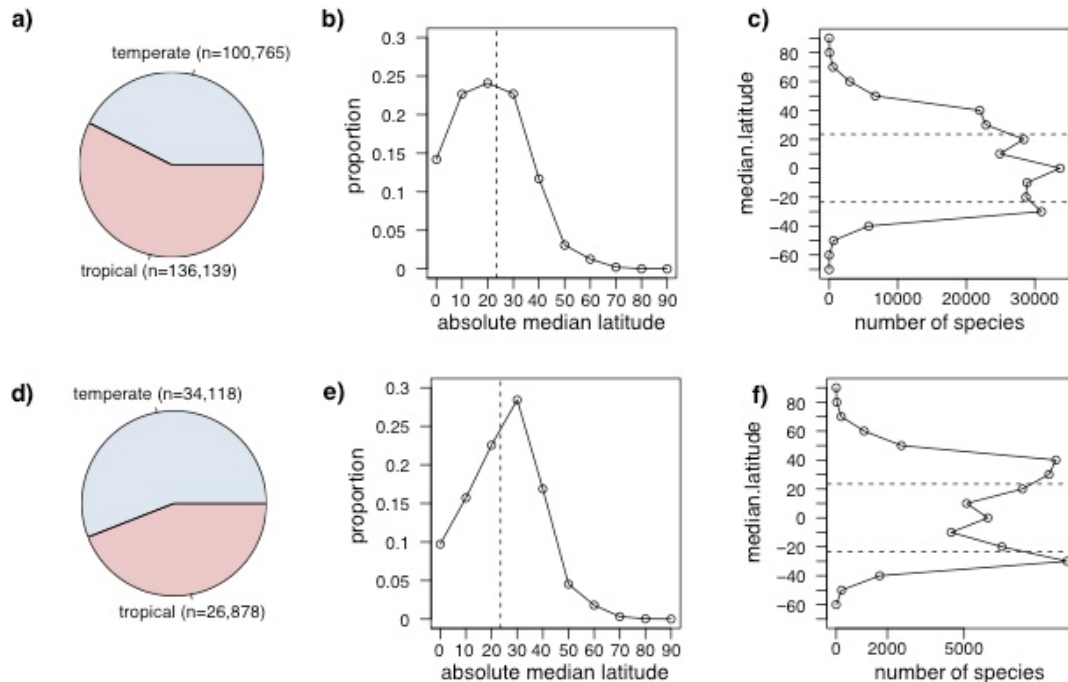
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578 **Figure S1.** The full dataset had a disproportionate number of temperate species. **a)**
579 Proportion of tropical and temperate species in the GBIF dataset (n = 236,894); **b)**
580 proportion of species in each latitudinal band in the GBIF dataset; **c)** number of
581 species in each latitudinal band in the GBIF dataset; **d)** proportion of tropical and
582 temperate species in the full dataset (n = 60,990); **e)** proportion of species in each
583 latitudinal band in the full dataset; **f)** number of species in each latitudinal band in the
584 full dataset; in **b), c), e)** and **f)**, the dotted vertical line denotes the 23.5° threshold
585 between tropical and temperate zones.

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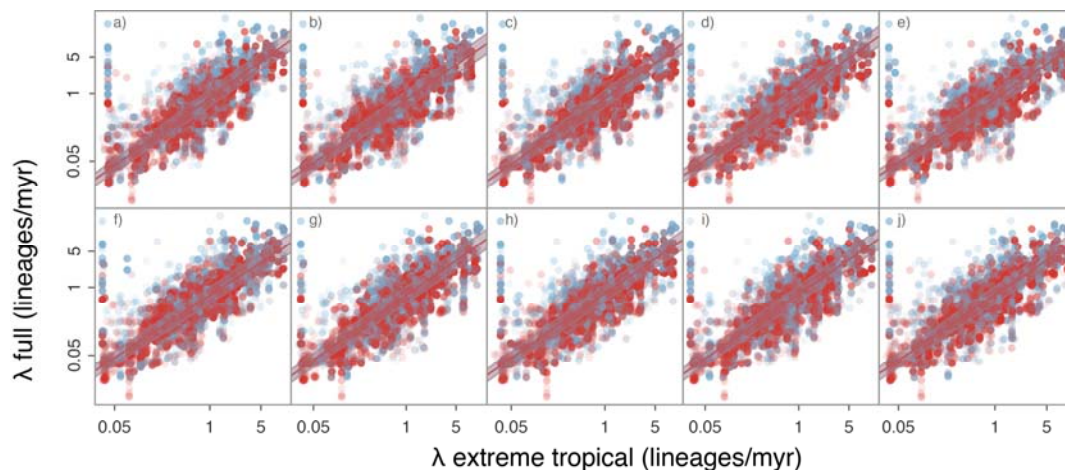
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Figure 2. BAMM λ estimates were positively correlated in the full and 10

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extreme tropical datasets (a to j) with no effect of tropicity on this correlation.

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Solid lines are phylogenetic linear regressions predicting λ in the full tree ($n = 60,990$

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species) with λ in the extreme tropical tree ($n = 30,000$ species) in temperate (blue)

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and tropical (red) species. Shaded areas indicate the 95% confidence intervals.

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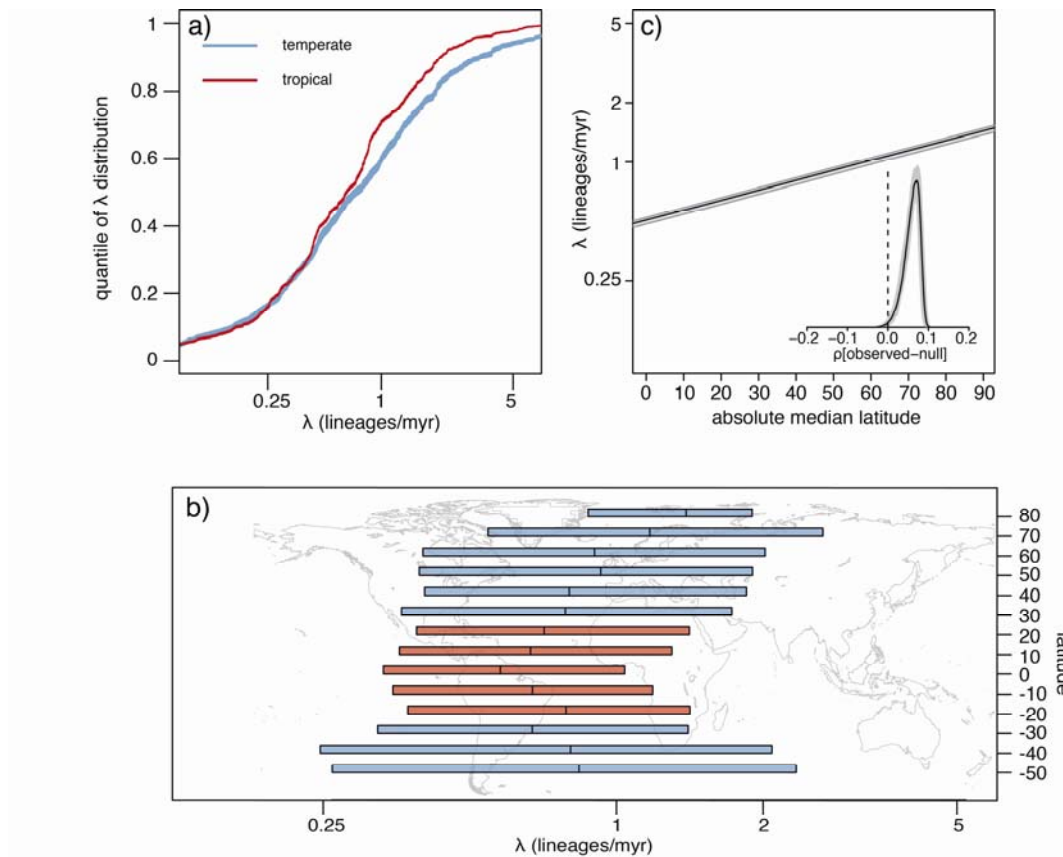
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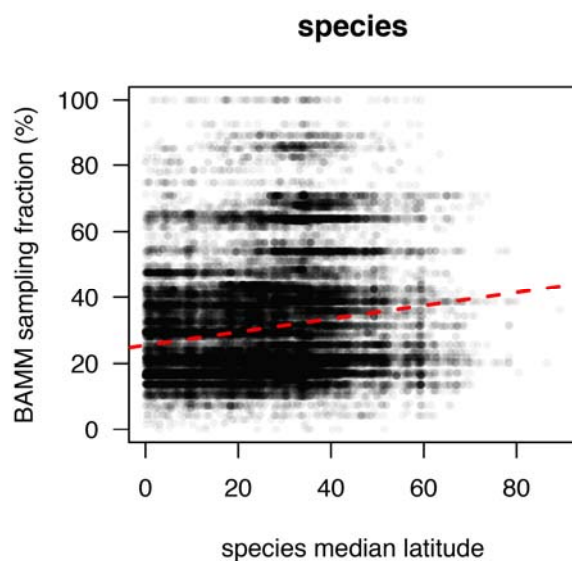
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621 **Figure S3. Temperate species have higher speciation rates (λ) than tropical**
622 **species for 100 subsampled “extreme tropical” datasets with 29.1% temperate**
623 **and 70.9% tropical species. a) Rank ordered distribution of λ inferred with BAMM**
624 **for tropical (red) and temperate (blue) species. b) λ grouped by latitudinal band of**
625 **each species. Boxes and lines are presented as in Fig. S4. c) Spearman’s ρ correlation**
626 **of species absolute median latitude and λ as estimated with STRAPP. . Solid black**
627 **line indicates median correlation \pm 95% confidence interval across the posterior**
628 **distribution. The inset shows the median (black line) \pm 95% confidence interval for**
629 **the difference between the empirical and null correlations estimated with 1000**
630 **permutations of evolutionary rates across the phylogeny in each of 100 subsampled**
631 **datasets.**

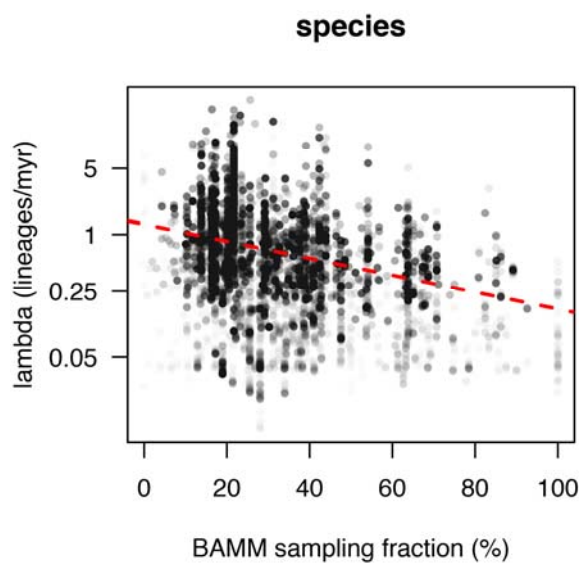
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638 **Figure S4.** Relationship between the species absolute median latitude and the family-
639 level sampling fraction used in the BMM analyses. The red line is the slope of the
640 linear regression (slope = 0.201, p-value < 0.0001).

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670 **Figure S5.** More densely sampled clades have smaller estimates of speciation rate (λ).

671 The dotted line is the slope of the linear regression of the $\log(\lambda)$ and the BMM

672 sampling fraction (slope = -0.027 , p-value < 0.0001).

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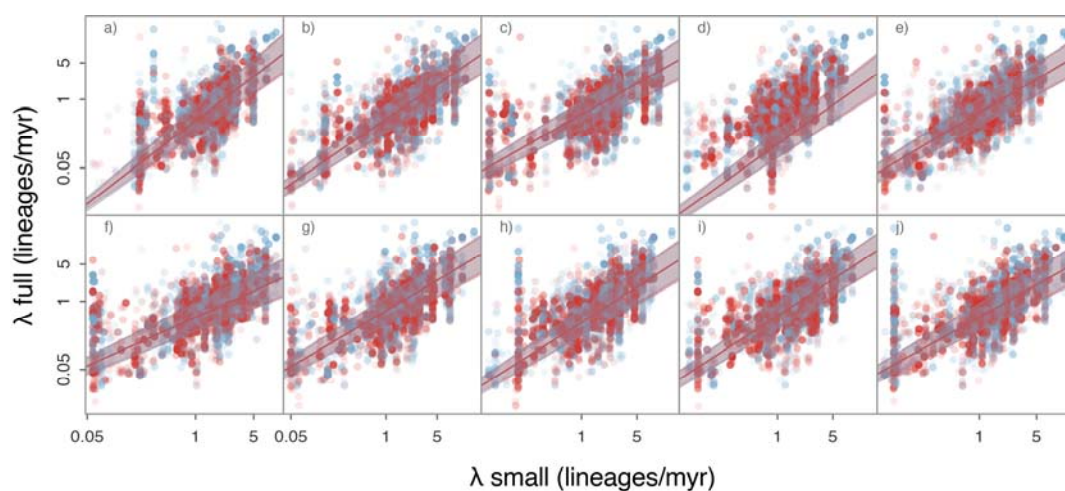
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Figure S6. BAMM λ estimates are positively correlated in the full and 10 small

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datasets and tropicality has no effect on this correlation. **a) - j)** Solid lines are

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phylogenetic linear regressions predicting λ in the full tree ($n = 60,990$ species) with λ

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in the small tree ($n = 10,739$ species) in temperate (shown in blue) and tropical

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(shown in red) species. Shaded areas indicate the 95% confidence intervals.

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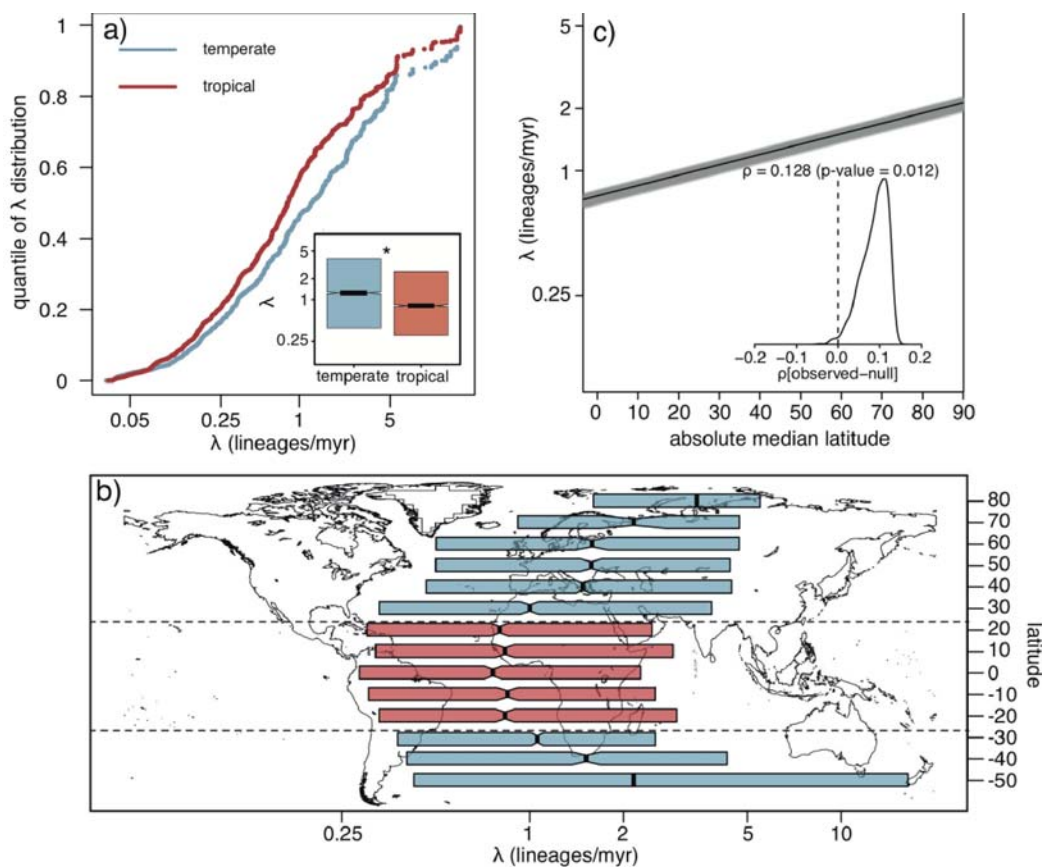
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708 **Figure S7.** Tropical species have smaller speciation rates (λ) than temperate species

709 with the Zanne phylogeny ($n = 28,057$ species). a) Rank ordered distribution and

710 boxplot (inset) of λ inferred with BAMM for tropical (red) and temperate (blue)

711 species; b) λ grouped by latitudinal band of each species; and c) Spearman's ρ

712 correlation between species absolute median latitude and λ as estimated with

713 STRAPP. Lines, boxes and symbols as in Fig. 1.

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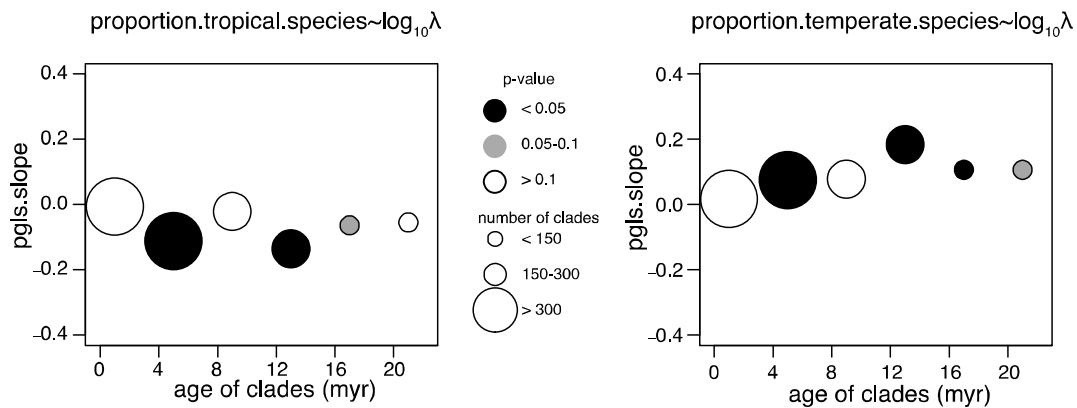
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Figure S8. Temperate clades have higher speciation rates in the clade-based analysis.

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Correlation of a) the proportion of tropical and b) temperate species in each clade with

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the corresponding speciation rate (λ) estimated with RPANDA. The correlation

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coefficient is the phylogenetic generalised least squares (PGLS) slope. The size of the

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points is scaled to the number of clades in each time interval and their colours show

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the statistical significance of the slope.

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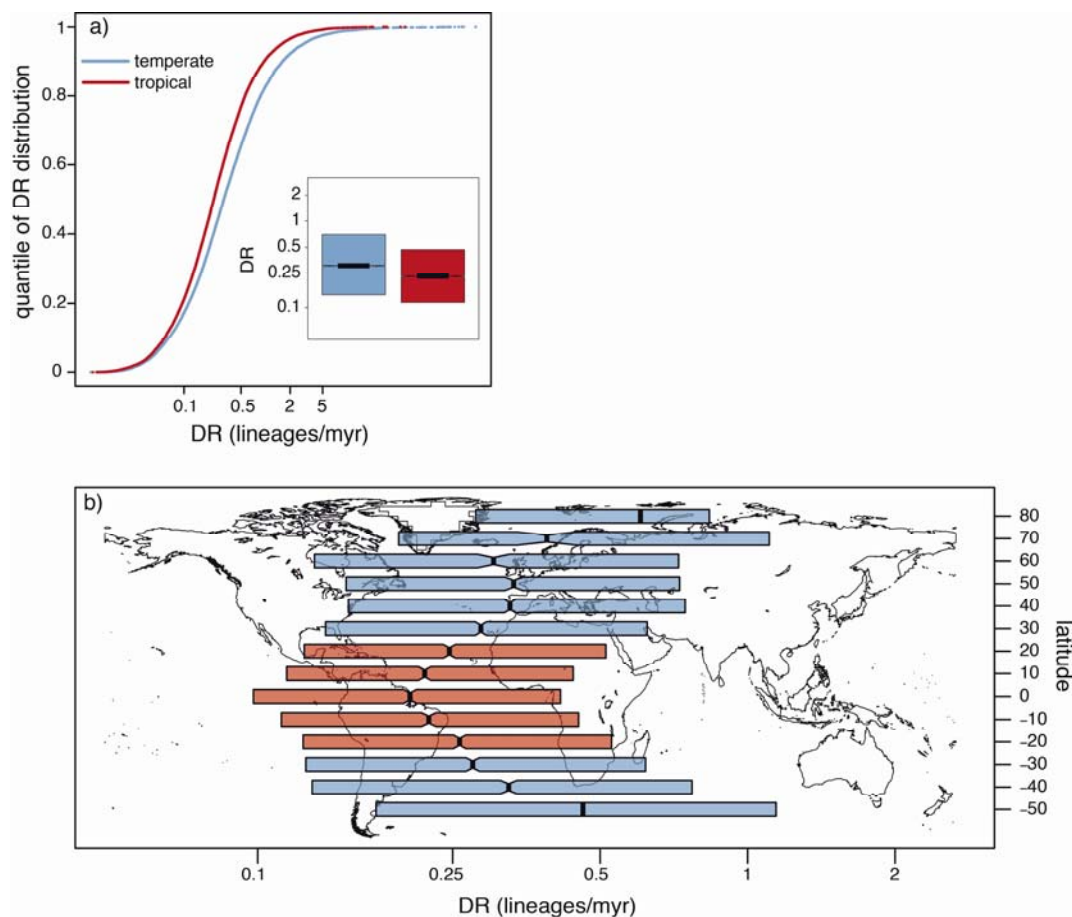
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Figure S9. Tropical species have smaller DR values than temperate species. a)

758 Rank-ordered distribution and boxplot (inset) of DR for tropical (red) and temperate

759 (blue) species. b) DR grouped by latitudinal band of each species. Notches in

760 boxplots indicate 95% confidence intervals around median, denoted by thick vertical

761 lines and boxes span the interquartile ranges.

762

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767 anonymous reviewers for comments that improved the manuscript.

768

769 **Data accessibility:**

770 Scripts and data to reproduce the analyses will be deposited upon acceptance in

771 Figshare

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773

774 **Declaration of interests**

775 The authors declare no competing interests.

776

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