

1 Tree diversity increases forest temperature buffering

2 Florian Schnabel^{1,2,*}, Rémy Beugnon^{1,3,4,*}, Bo Yang^{5,*}, Ronny Richter^{1,2}, Nico Eisenhauer^{1,6}, Yuanyuan
3 Huang^{1,6}, Xiaojuan Liu⁷, Christian Wirth^{1,2}, Simone Cesarz^{1,6}, Andreas Fichtner⁸, Maria D. Perles-
4 Garcia^{1,9}, Georg J. A. Hähn^{1,14}, Werner Härdtle⁸, Matthias Kunz^{9,10}, Nadia C. Castro Izaguirre¹¹, Pascal
5 A. Niklaus¹², Goddert von Oheimb⁹, Bernhard Schmid¹³, Stefan Trogisch^{14,1}, Manfred Wendisch³,
6 Keping Ma^{7,†}, Helge Brügelheide^{14,1,†}

7 *These authors contributed equally to this work; †Corresponding authors.

8

9 ¹ German Centre for Integrative Biodiversity Research (iDiv) Halle-Jena-Leipzig, Puschstraße 4 04103
10 Leipzig, Leipzig, Germany

11 ² Systematic Botany and Functional Biodiversity, Leipzig University, Johannisallee 21, Leipzig,
12 Germany

13 ³ Leipzig Institute for Meteorology, Leipzig University, Stephanstraße 3, 04103 Leipzig, Germany

14 ⁴CEFE, Univ Montpellier, CNRS, EPHE, IRD, 1919, route de Mende, F-34293 Montpellier Cedex 5,
15 France

16 ⁵Jiangxi Key Laboratory of Plant Resources and Biodiversity, Jingdezhen University, Fuliang Avenue
17 3#, Jingdezhen 333400, China

18 ⁶Institute of Biology, Leipzig University, Leipzig, Germany

19 ⁷Institute of Botany, Chinese Academy of Sciences, 20 Nanxincun, Xiangshan, Haidian District, Beijing
20 100093, China

21 ⁸Institute of Ecology, Leuphana University of Lüneburg, Universitätsallee 1, 21335 Lüneburg,
22 Lüneburg, Germany

23 ⁹Institute of General Ecology and Environmental Protection, TUD Dresden University of Technology,
24 Pienner Straße 7, 01737 Tharandt, Germany

25 ¹⁰Helmholtz Centre Potsdam - GFZ German Research Centre for Geosciences, Section 1.4 Remote
26 Sensing and Geoinformatics, Telegrafenberg, Potsdam, 14473 Germany

27 ¹¹WSL Swiss Federal Research Institute, Zürcherstrasse 111, 8903 Birmensdorf, Switzerland

28 ¹²Department of Evolutionary Biology and Environmental Studies, University of Zürich,
29 Winterthurerstrasse 190, 8057 Zürich, Switzerland

30 ¹³Department of Geography, Remote Sensing Laboratories, University of Zürich, Winterthurerstrasse
31 190, 8057 Zürich, Switzerland

32 ¹⁴Institute of Biology/Geobotany and Botanical Garden, Martin Luther University Halle-Wittenberg,
33 Große Steinstraße 79/80, 06108 Halle (Saale), Germany

34

35 **Summary paragraph**

36 **Global warming is increasing the frequency and intensity of climate extremes¹. Forests**
37 **may buffer such extreme events by creating their own microclimate below their canopy**
38 **via cooling hot and insulating against cold macroclimate air temperatures^{2,3}. This**
39 **buffering capacity of forests may be increased by tree diversity^{4,5} and may itself maintain**
40 **forest functioning and biodiversity^{6,7}. However, despite its relevance for many ecosystem**
41 **processes, the effect of tree diversity on temperature buffering is largely unexplored.**
42 **Here, we show that tree species richness consistently increases forest temperature**
43 **buffering across daily, monthly, and annual scales over six years. This finding is based on**
44 **data from a large-scale tree diversity experiment covering a species richness gradient of**
45 **1 to 24 tree species. We found that species richness strengthened both components of**
46 **forest temperature buffering^{2,6}: the attenuation of hot and of cold macroclimate air**
47 **temperatures, with the cooling effect being more pronounced. The buffering effect of tree**
48 **species richness was mediated by canopy density and structural diversity, assessed as leaf**
49 **area index and stand structural complexity index⁵, respectively. Safeguarding and**
50 **planting diverse forests⁸ may thus mitigate negative effects of global warming and climate**
51 **extremes on ecosystem functions and communities below the tree canopy.**

52

53 Introduction

54 Global warming and its impacts on the world's forests¹ are largely studied as effects of air
55 temperatures measured outside forests in open-ground conditions (also referred to as
56 macroclimate)⁶. However, this omits that forests can buffer temperature extremes such as hot
57 and cold spells to some extent by creating their own microclimate below their canopy^{2,3,9}, from
58 which other organisms benefit, including sub-canopy trees. Among earth's terrestrial
59 ecosystems, forests are likely the one with the strongest air temperature buffering (hereafter
60 'temperature buffering') capacity owing to their often multi-layered canopies, which provide
61 evapotranspirative cooling and shading, and decrease the mixing of air layers^{3,7}. Temperature
62 buffering occurs when microclimate temperature fluctuations are smaller than fluctuations in
63 macroclimate temperatures⁶. Smaller temperature fluctuations below the canopy can be
64 quantified as a lower temporal variance of temperatures, to which in the following we refer to
65 as microclimate temporal stability¹⁰. The differences between macroclimate (outside forest) and
66 microclimate (inside forest) temperatures are substantial, with global averages of $-4.1 \pm 0.5^{\circ}\text{C}$
67 decreased temperature maxima and $1.1 \pm 0.2^{\circ}\text{C}$ increased temperature minima below the forest
68 canopy². This difference is larger than the average warming of land and ocean temperatures in
69 2001–2020 compared with 1850–1900 (0.8 to 1.1) $^{\circ}\text{C}$ ¹¹.

70

71 The temperature buffering capacity of forests has important consequences for forest functioning
72 and biodiversity above- and belowground, especially in the context of global warming^{6,9,12,13}.
73 For instance, many physiological processes, such as photosynthesis or soil respiration¹⁴, scale
74 exponentially with temperature, which implies that even small temperature increases may have
75 large effects on rates, underlining the importance of temperature buffering. Furthermore,
76 temperature buffering can influence forest biodiversity by slowing shifts in forest community
77 composition towards warm-affinity species (i.e. thermophilization) under global warming^{7,15,16}.

78 However, the reciprocal control of tree diversity on forest temperature buffering remains largely
79 unexplored.

80

81 Simulation studies showed that plant diversity can stabilise climate–vegetation feedbacks¹⁷.

82 Moreover, tree species diversity has been shown to increase tree growth in mixtures^{18–20} and to
83 enhance canopy complexity^{21,22}, resulting in a greater thickness, density and structural diversity
84 of the canopy layer (i.e., the buffering layer). It is thus conceivable that tree species richness
85 may increase the temperature buffering capacity of forests by affecting these forest properties.

86 For instance, mean tree height and the area of foliage per unit ground area (i.e. leaf area index;
87 LAI)²³ as proxies for the thickness and density of the buffering layer, modify the energy
88 exchange at the canopy by influencing its albedo and evapotranspiration, which in turn affects
89 the temperature buffering capacity of the forest⁶. Moreover, structural diversity²⁴ measured as
90 stand structural complexity index (SSCI) from terrestrial laser scans⁵ may reduce the vertical
91 mixing of air masses²⁵ and thereby increase temperature buffering. However, there is very little
92 empirical evidence for tree diversity effects on forest temperature buffering in general, and, in
93 particular, regarding the mechanisms mediating such diversity–microclimate relationships. For
94 example, tree species richness (1- vs 4-species) was shown to increase temperature buffering
95 for some species mixtures⁴, but longer diversity gradients and data from multiple years would
96 be necessary to generalise beyond specific species compositions and macroclimatic conditions
97 as well as to understand the mediators of tree diversity effects on temperature buffering and
98 their temporal dynamics.

99

100 Tree diversity effects on microclimate temperatures in forests may change over days, months
101 and years. Compared with open-ground conditions, temperatures within forests are expected to
102 be higher during night-time and winter, and lower during day-time and summer¹³. The
103 underlying reason is that the energy exchange is shifted from the ground surface to the canopy²⁶.

104 Consequently, forest canopies mitigate hot temperatures via evapotranspiration (consumption
105 of latent heat), reflecting or absorbing solar radiation and emitting long-wave radiation, and
106 insulate against cold temperatures via heat retention^{2,3}. However, many more processes may be
107 involved depending on the spatiotemporal scale studied⁶. For example, evapotranspirative
108 cooling effects decrease with decreasing soil water availability⁶, highlighting the potential
109 influence of inter-annual dynamics and extremes in macroclimatic conditions (such as
110 droughts) for temperature buffering. However, the relative importance of tree diversity effects
111 on temperature buffering across temporal scales remains unknown.

112

113 Here, we analyse microclimate measurements conducted within forests of 1 to 24 tree species
114 covering six years (2015–2020) from a large-scale subtropical tree diversity experiment (BEF-
115 China^{18,27}). Assembling the communities with varying species richness randomly from species
116 pools resulted in stands that differ in canopy thickness¹⁸, density²⁸ and structural diversity²².
117 We aim to understand the role of tree species richness and these mediating factors for
118 temperature buffering below forest canopies at different temporal scales (i.e. daily, monthly
119 and yearly). In our subtropical study system, which is characterised by a monsoon climate, high
120 macroclimate temperatures coincide with high water availability for evapotranspirative cooling,
121 which should promote temperature offsets between micro- and macroclimate, particularly for
122 maximum temperatures⁶. Hence, we expect tree species richness effects on microclimate to be
123 most pronounced for the buffering of maximum temperatures. We tested the following
124 hypotheses: H1: tree species richness increases the temperature buffering potential of forest
125 canopies via cooling hot and insulating against cold macroclimate temperatures at daily,
126 monthly, and annual time scales. H2: species richness effects on temperature buffering—
127 measured as microclimate temperature stability—are consistently positive across time scales
128 but strongest when macroclimate temperatures are high. H3: positive tree species richness

129 effects on temperature buffering are mediated by enhanced canopy thickness, density, and
130 structural diversity.

131

132 **Results**

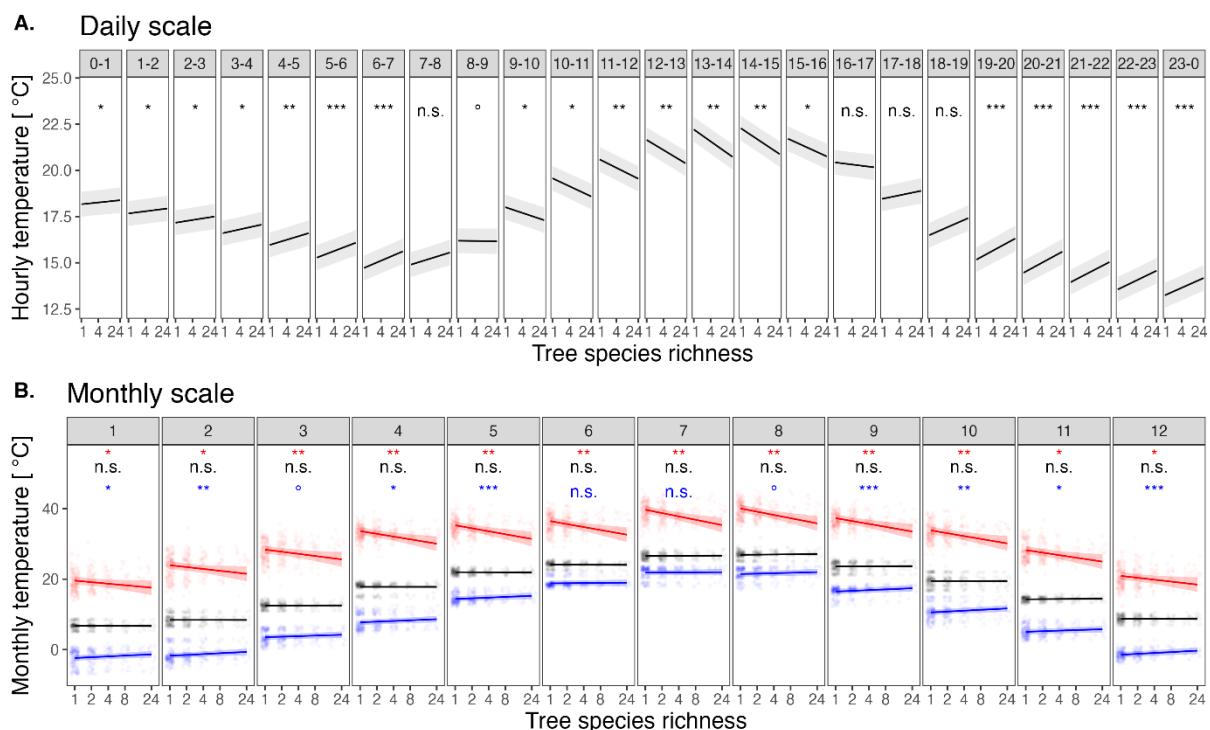
133 On the daily scale, we found below-canopy air temperatures to decrease with tree species
134 richness during daytime, while they increased with species richness during the night (Fig. 1a).
135 Hence, the mode of tree species richness effects on microclimate temperature changed
136 significantly with the diurnal course in macroclimate temperatures from positive (during cold
137 night-time hours) to negative (during hot day-time hours; Fig. 1a: interaction between species
138 richness and hour significant at $P < 0.001$). Mitigating species richness effects on microclimate
139 temperature were strongest at midday peak hours (mean temperature offsets of $-2.5 \pm 0.2^\circ\text{C}$
140 from noon to 3 pm) and positive effects strongest around midnight ($+0.4 \pm 0.04^\circ\text{C}$ from 11 pm
141 to 2 am) between stands with 1 and 24 tree species, respectively.

142

143 On the monthly scale, we examined maximum, minimum and median daily microclimate
144 temperatures across months (Fig. 1b). We found that tree species richness significantly reduced
145 maximum microclimate temperature across all months (January–December, P-value range of
146 slopes 0.002–0.033); this buffering effect was strongest during summer (up to $-4.4^\circ\text{C} \pm 0.6^\circ\text{C}$
147 in 24-species mixtures in July, $P = 0.004$) and during high macroclimate temperatures
148 (Supplement S1). Tree species richness also increased minimum microclimate temperatures in
149 most months (September–May; P-value range of slopes 0.001–0.053); this buffering effect was
150 strongest in winter (up to $+1.1^\circ\text{C} \pm 0.2^\circ\text{C}$ in 24-species mixtures in December, $P < 0.001$), non-
151 significant during summer (June–August; $P > 0.05$), and strongest during low macroclimate
152 temperatures (Supplement S1). We found no significant effect of tree species richness on
153 median monthly temperatures (Fig. 1b; $P > 0.5$ for all months), i.e., species richness only

154 affected temperature extremes. Hence, as hypothesised, tree species richness cooled hot and
155 insulated against cold macroclimate temperatures, which contributed to enhanced temperature
156 buffering in species-rich stands.

157



158

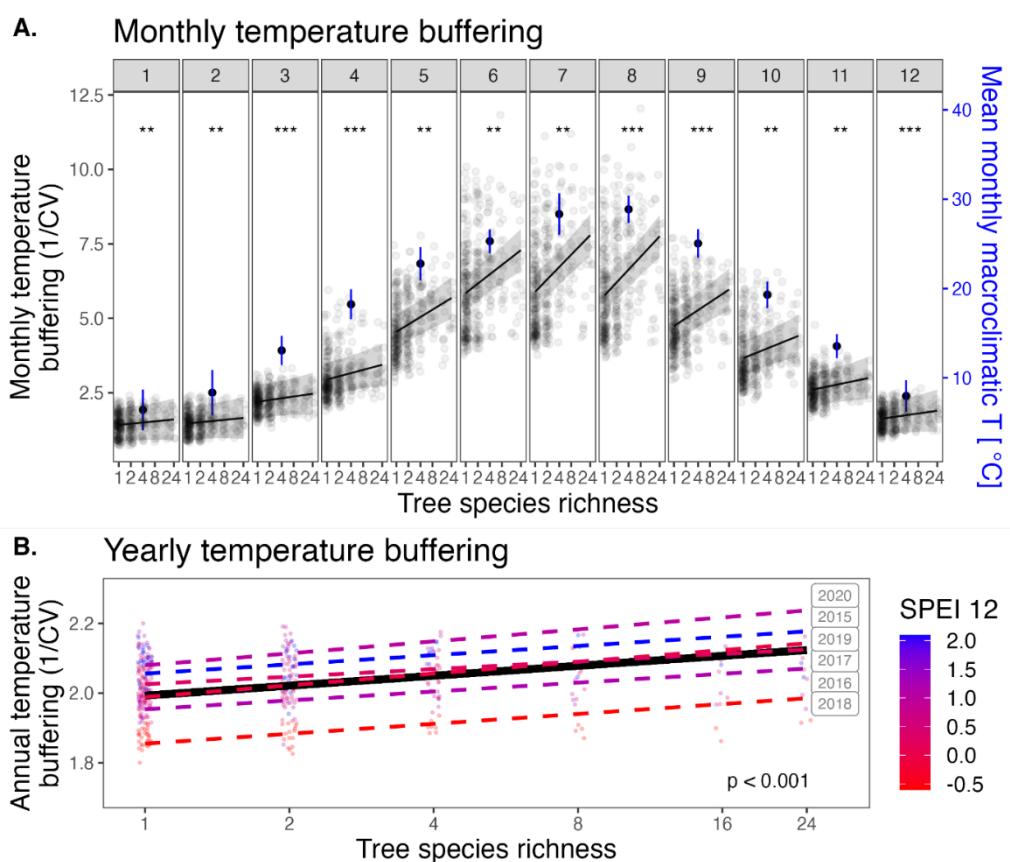
159 Figure 1 Tree species richness effects on microclimate temperature on (A) the daily and (B) the monthly
160 scale. (A) Hourly modulation of microclimate temperatures by tree species richness (n = 63 plots and 4
161 million values). (B) Monthly modulation of maximum (red), median (black) and minimum (blue) daily
162 temperatures per month by tree species richness (n = 63 plots and 4476 values). Lines show predictions
163 of linear mixed-effects models, and shaded bands indicate 95% confidence intervals. Data points in (B)
164 are jittered to enhance visibility. Species richness was log-transformed in all models. See Supplement
165 S2 for complete model outputs. Significance levels: “n.s.”: non-significant, “◦”: p < 0.1, “*”: p < 0.05,
166 “**”: p < 0.01, and “***”: p < 0.001.

167

168 We quantified the temperature buffering capacity of a tree community on monthly and annual
169 time scales as the temporal stability¹⁰ of microclimate temperature, calculated as the inverse of
170 the coefficient of variation (CV) of hourly temperature measurements. This stability metric is
171 commonly used in biodiversity–ecosystem functioning studies to provide insights into the
172 stabilising effects of biodiversity for multiple ecosystem processes and at different levels of
173 organization^{29–31}. We found a consistently positive effect of tree species richness on monthly

174 temperature buffering across the entire year (January–December; $P \leq 0.006$ for all months),
175 which was strongest in summer (June–August; Fig. 2a) and during high macroclimate
176 temperatures (Supplement S1). Tree species richness also had significant positive effects on
177 annual temperature buffering during all years examined ($P < 0.001$; Fig. 2b). Effects of species
178 richness on annual temperature buffering (slope of the species richness–temperature buffering
179 relationship) did not change significantly across years, but the absolute temperature buffering
180 capacity of the examined forest communities changed with macroclimatic conditions.
181 Temperature buffering was significantly lowest during the driest year (i.e. 2018, the year with
182 the lowest SPEI values, $P < 0.001$; Fig. 2b, Supplement S2).

183



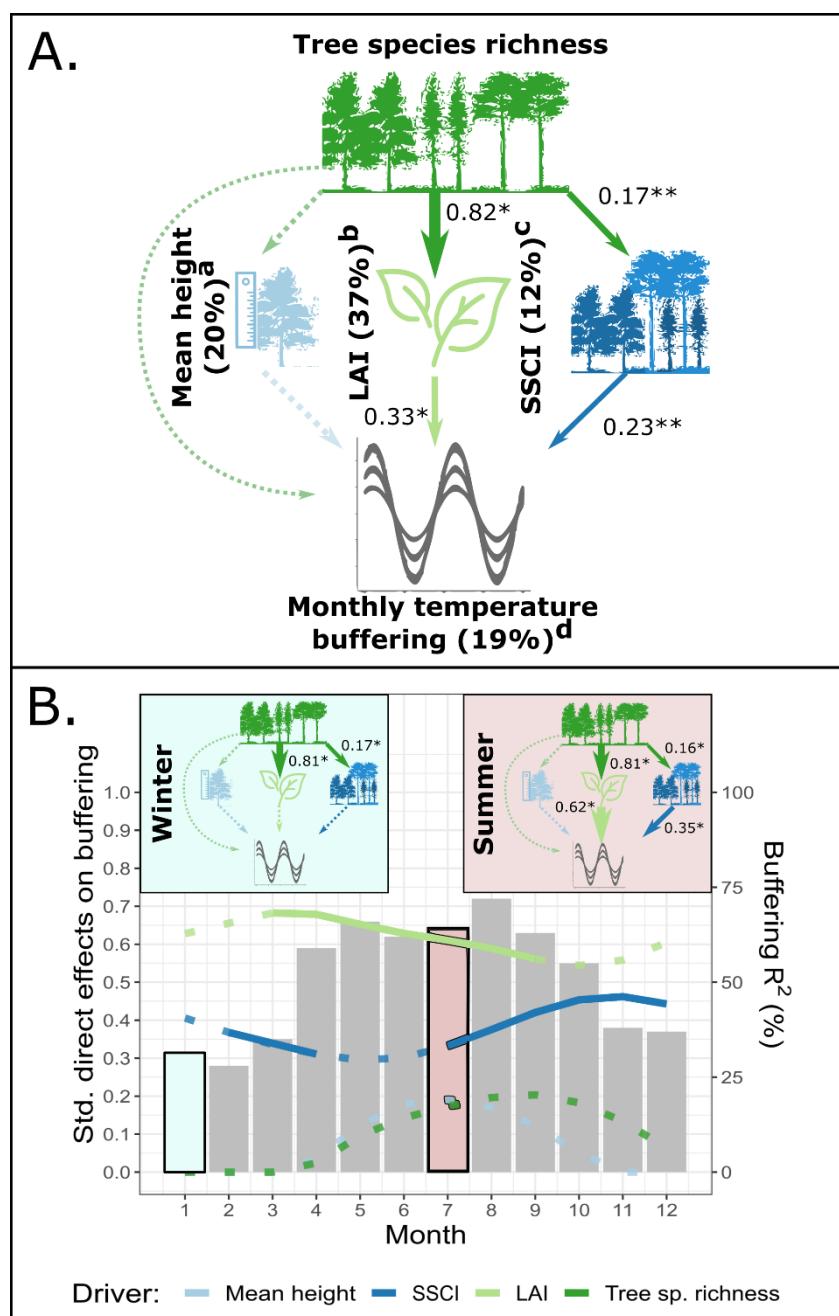
185 Figure 2 Tree species richness effects on temperature buffering on (A) the monthly and (B) the annual
186 scale. (A) Modulation of monthly microclimate temperature stability ($n = 63$ plots and 4476 values) by
187 tree species richness and month of the year. (B) Modulation of annual microclimate temperature stability
188 ($n = 63$ plots and 375 values) by tree species richness and year. In all panels, the lines show predictions
189 of linear mixed-effects models, and data points are jittered to enhance visibility. In (A), p-values refer
190 to the effects of species richness on monthly temperature buffering and solid blue points show mean
191 monthly macroclimate temperatures, while shaded bands and blue whiskers indicate 95% confidence

192 intervals. In (B), the P-value refers to the effect of species richness across years. Points and lines are
193 coloured according to their value with deeper red and blue indicating increasing and decreasing drought,
194 respectively, based on annual values of the standardised precipitation evapotranspiration index
195 (SPEI12). Species richness was log-transformed in all models. See Supplement S2 for complete model
196 outputs. Significance levels: “n.s.”: non-significant, “ $^{<0.1}$ ”: $p < 0.1$, “ * ”: $p < 0.05$, “ ** ”: $p < 0.01$, and “ *** ”:
197 $p < 0.001$.

198

199 We used piecewise Structural Equation Models (SEMs; Fig. 3) to examine potential
200 mechanisms that may mediate the observed tree species richness effects on monthly
201 temperature buffering (Fig. 2). Out of a set of potential variables and based on literature-derived
202 hypotheses (Supplement S3), we selected mean tree height, LAI, and SSCI as measures of
203 canopy thickness, density, and structural diversity, respectively. Once controlling for the effect
204 of macroclimate, we found LAI to have the strongest positive effect on temperature buffering
205 (Std. estimate = 0.33, $P = 0.011$), followed by SSCI (Std. estimate = 0.23, $P = 0.002$), while
206 mean tree height had no significant effect on temperature buffering ($P = 0.5$, Fig. 3A). Both
207 LAI and SSCI significantly increased with increasing tree species richness (Std. estimate =
208 0.82, $P = 0.007$ and Std. estimate = 0.17, $P = 0.002$, respectively). Once accounting for these
209 forest properties and their influence on temperature buffering, we found no remaining direct
210 effect of tree species richness on temperature buffering ($P = 0.3$, Fig. 3A). Using tree basal area
211 measured in 2019 (another commonly used proxy for canopy- or stand density^{4,32}) instead of
212 LAI resulted in similar pathways: tree species richness increased basal area (Std. estimate =
213 0.23, $P = 0.049$), which in turn enhanced temperature buffering (Std. estimate = 0.41, $P = 0.011$;
214 Supplement S4). The influence of the different drivers changed over the annual course (Fig.
215 3B): LAI was the strongest driver of temperature buffering during the growing season (March–
216 September), while SSCI mostly affected temperature buffering before and after the growing
217 season.

218



219

220 Figure 3 Structural Equation Models (SEMs) examining potential mediators of tree species richness
 221 effects on monthly temperature buffering. The SEMs test the direct effects of tree species richness and
 222 its indirect effects mediated by mean tree height, leaf area index (LAI), and stand structural complexity
 223 index (SSCI) on monthly temperature buffering while controlling for macroclimate temperatures. The
 224 SEM in (A) was fit to microclimate data of all months, and tree species richness effects on forest
 225 properties (i.e. mean tree height, LAI, and SSCI) were tested on the datasets built for this purpose (a: 32
 226 plots, b: 54 plots, c: 74 plots, d: 27 plots; see methods). All pathways were fit to data from site A measured
 227 in 2019 except for LAI, which was measured in 2014 (see methods and Supplement S3 for details).
 228 Species richness and SSCI were log-transformed in all models. Significant directional relationships
 229 between variables are shown as solid and nonsignificant relationships as dashed arrows. Significant
 230 standardised path coefficients are shown next to each path (*P<0.05, **P<0.01, and ***P<0.001),
 231 and path width is scaled according to coefficient size. The explained variation of variables (marginal
 232 R²) is given in %. The SEM fit the data well (Fisher's C=8.4, df=6, P=0.21). In (B), the same SEM

233 was fit separately for each month to explore temporal trends in the path coefficients. The SEMs in
234 January and July exemplify pathways during winter and summer, respectively. For each month, coloured
235 curves show standardized path coefficients (dashed if non-significant) and bars show the variation in
236 temperature buffering explained by the examined forest properties (marginal R^2); note that the marginal
237 R^2 in (A) is lower than in the monthly models in (B) as it only captures the variation explained by fixed
238 effects, which do not account for the strong variation in temperature buffering between months. See
239 Supplement S4 for complete model outputs.

240

241 **Discussion & Outlook**

242 In a large-scale tree diversity experiment, we observed a consistent increase in forest
243 temperature buffering across daily, monthly, and annual scales with increasing tree species
244 richness. Confirming H1, species-rich forests cooled high and insulated against cold
245 macroclimate temperatures better than species-poor forests. This positive effect had a
246 considerable magnitude with -4.4°C ($\pm 0.6^{\circ}\text{C}$) and $+1.1^{\circ}\text{C}$ ($\pm 0.2^{\circ}\text{C}$) in peak summer and winter
247 for monocultures *vs* 24-species mixtures, respectively. Confirming H2, temperature buffering
248 was thus driven primarily by a reduction of maximum below-canopy temperatures, with this
249 effect being strongest during hot macroclimate conditions (during midday and summer).

250

251 We expected species-rich tree canopies to mainly cool hot temperatures by enhancing
252 evapotranspiration and the reflection or absorption of short-wave and emittance of long-wave
253 radiation^{2,3}. Likewise, tree canopies may insulate against cold temperatures by retaining heat
254 and long-wave radiation, even though many more processes are likely involved⁶. Consistent
255 with our findings, stronger buffering of maximum relative to minimum temperatures
256 predominates across the world's forests^{2,6}. Moreover, next to temperature extremes, droughts
257 will likely threaten the world's forests during the 21st century^{1,33}. We found the lowest absolute
258 temperature buffering in the driest year (2018) of our observation period (Fig. 2), likely due to
259 reduced cooling potentials via evapotranspiration (as a result of the low atmospheric and soil

260 moisture)⁶. However, the buffering role of tree species richness was maintained (Fig. 2),
261 indicating that tree species richness provides an insurance against climate extremes.

262
263 There is ample evidence that forests buffer temperature extremes^{2,3,9} and that species identities
264 matter for temperature buffering^{4,25}, but the role of tree diversity has largely remained hidden.
265 The few former studies on the role of tree species composition for temperature buffering
266 reported predominantly non-significant effects of species richness^{25,32,34}. Positive effects were
267 rare and only found for specific mixtures⁴. It may be that idiosyncrasies of the investigated
268 species prevented the detection of general patterns of species richness in earlier studies or that
269 the level of species richness analysed was too low to detect significant effects. Our experimental
270 design with a long tree species richness gradient ranging from 1 to 24 tree species and
271 randomised extinction scenarios where each richness level was represented by different species
272 compositions and each species occurred at each richness level^{18,27} allowed us to move beyond
273 the effects of specific species compositions while controlling for environmental variation and
274 species identity effects. At the examined scale, this high tree species richness level is common
275 under natural conditions in the region²⁷.

276
277 Partially confirming H3, we found positive tree species richness effects on temperature
278 buffering to be mediated by enhanced canopy density and structural diversity but not by canopy
279 thickness (Fig. 3). The absence of a remaining direct tree species richness effect after
280 accounting for these forest properties supports the use of the chosen proxies (LAI and SSCI)
281 and suggests that we captured the dominant mechanisms driving temperature buffering. Still,
282 monitoring other potential drivers, such as enhanced transpiration³⁵, will be relevant for
283 comprehensively understanding species richness effects on temperature buffering. Canopy
284 density and structural diversity were already shown to be enhanced by tree species richness in
285 our experiment^{22,28} and elsewhere^{5,19,20}. Likewise, canopy density^{6,32,36} and structural

286 diversity^{5,25,34} were reported to be significant drivers of forest temperature buffering. Moreover,
287 and similar to our findings, structural diversity was more relevant than mere canopy height in
288 this context²⁵. However, these studies did not elucidate the mechanistic links between species
289 richness, canopy density, structural diversity, and temperature buffering. Here, we provide
290 experimental evidence that most species richness effects on temperature buffering are indirect
291 and mediated via diversity-induced changes in these forest properties. This notion is consistent
292 with canopy cover, another proxy for canopy density, mediating species richness effects on
293 minimum and maximum temperatures⁴. Furthermore, our study reveals that drivers of
294 temperature buffering in forests exhibit temporal complementarity, with LAI being most
295 relevant during the peak growing season and SSCI, which captures the structural diversity of
296 canopy elements (stems and branches) during the leaf-off period of the deciduous tree species,
297 taking over outside the growing season.

298
299 The positive effect of tree diversity on temperature buffering we report represents, as also
300 highlighted by advances in grassland research^{37,38}, a previously overlooked biodiversity-
301 ecosystem functioning (BEF) relationship, with potentially far-reaching implications. In
302 contrast to other mechanisms that cause positive BEF relationships in forests, such as biotic
303 interactions between trees, negative density effects, or multitrophic interactions³⁹, which are all
304 species-specific, temperature buffering emerges from the community as a whole. The resulting
305 lower temperature variation in species-rich forests may safeguard ecosystem functions,
306 particularly those that respond non-linearly to temperature¹⁴, against temperature maxima (and
307 minima). This may be especially relevant for functions severely impeded beyond narrow
308 threshold ranges of temperature, such as net photosynthesis rates⁴⁰. Likewise, belowground
309 functioning, including carbon sequestration, decomposition, and nutrient cycling^{13,41,42}, may be
310 enhanced by temperature buffering. As a result, trees in mixtures may grow^{43,44} and regenerate
311 better⁴⁵ in ameliorated microclimates, which may, in turn, enhance temperature buffering via

312 enhancing canopy density (Fig. 3). Moreover, by reducing maximum temperatures (Fig. 1), tree
313 diversity-enhanced temperature buffering may impact forest biodiversity under global warming
314 by reducing the thermophilization of below-canopy communities^{7,15,16}. Finally, forest
315 temperature buffering also alleviates heat stress for humans, and our findings indicate that tree
316 species richness may amplify this effect far stronger than previously reported³².

317

318 Overall, we suggest that preserving and planting diverse forests⁸ is a promising approach to
319 increase the temperature buffering function of forests, thereby protecting ecosystem functions
320 and communities below the tree canopy against global warming. Attempting to maximise LAI
321 and SSCI with only a single or a few tree species would have well-known limitations, as such
322 species-poor forests would have a higher susceptibility to specialist pests and pathogens,
323 droughts and storms⁸. In contrast, species-rich forests are more likely to maintain their buffering
324 capacity in the future⁴, given their higher stability under global change²⁹, while simultaneously
325 providing a broader range of ecosystem services⁸. Despite examining young planted forests (up
326 to 11 years after establishment), we already detected a strong temperature buffering capacity.
327 Our findings thus highlight the benefits of diverse planted forests for large-scale forest
328 restoration initiatives⁴⁶ and urban forests that aim at reducing thermal stress in a warming world.

329

330 **Main references**

331 1. IPCC. *Climate Change 2022: Impacts, Adaptation, and Vulnerability. Contribution of*
332 *Working Group II to the Sixth Assessment Report of the Intergovernmental Panel on*
333 *Climate Change* (Cambridge University Press, Cambridge, UK and New York, NY, USA,
334 2022).

335 2. Frenne, P. de *et al.* Global buffering of temperatures under forest canopies. *Nat. Ecol.*
336 *Evol.* **3**, 744–749; 10.1038/s41559-019-0842-1 (2019).

337 3. Geiger, R., Aron, R. H. & Todhunter, P. *The Climate Near the Ground*. 7th ed. (Rowman
338 & Littlefield, 2009).

339 4. Zhang, S., Landuyt, D., Verheyen, K. & Frenne, P. de. Tree species mixing can amplify
340 microclimate offsets in young forest plantations. *J. Appl. Ecol.* **59**, 1428–1439;
341 10.1111/1365-2664.14158 (2022).

342 5. Ehbrecht, M., Schall, P., Ammer, C. & Seidel, D. Quantifying stand structural complexity
343 and its relationship with forest management, tree species diversity and microclimate.
344 *Agric. For. Meteorol.* **242**, 1–9 (2017).

345 6. Frenne, P. de *et al.* Forest microclimates and climate change: Importance, drivers and
346 future research agenda. *Glob. Chang. Biol.* **27**, 2279–2297; 10.1111/gcb.15569 (2021).

347 7. Zellweger, F. *et al.* Forest microclimate dynamics drive plant responses to warming.
348 *Science* **368**, 772–775; 10.1126/science.aba6880 (2020).

349 8. Messier, C. *et al.* For the sake of resilience and multifunctionality, let's diversify planted
350 forests! *Conserv. Lett.* **15**, e12829; 10.1111/conl.12829 (2021).

351 9. Davis, K. T., Dobrowski, S. Z., Holden, Z. A., Higuera, P. E. & Abatzoglou, J. T.
352 Microclimatic buffering in forests of the future: the role of local water balance. *Ecography*
353 **42**, 1–11; 10.1111/ecog.03836 (2019).

354 10. Tilman, D. The ecological consequences of changes in biodiversity: a search for general
355 principles. *Ecology* **80**, 1455–1474; 10.1890/0012-
356 9658(1999)080[1455:TECOCI]2.0.CO;2 (1999).

357 11. IPCC. *Climate Change 2021: The Physical Science Basis. Contribution of Working Group
358 I to the Sixth Assessment Report of the Intergovernmental Panel on Climate Change*
359 (Cambridge University Press, Cambridge, UK and New York, NY, USA, 2021).

360 12. Naylor, D. *et al.* Soil Microbiomes Under Climate Change and Implications for Carbon
361 Cycling. *Annu. Rev. Environ. Resour.* **45**, 29–59; 10.1146/annurev-environ-012320-
362 082720 (2020).

363 13. Gottschall, F. *et al.* Tree species identity determines wood decomposition via
364 microclimatic effects. *Ecol Evol* **9**, 12113–12127; 10.1002/ece3.5665 (2019).

365 14. Chapin III, F. S., Vitousek, P. M. & Matson, P. A. (eds.). *Principles of Terrestrial
366 Ecosystem Ecology*. 2nd ed. (Springer New York, New York, NY, 2011).

367 15. Frey, S. J. K. *et al.* Spatial models reveal the microclimatic buffering capacity of old-
368 growth forests. *Sci. Adv.* **2**, e1501392; 10.1126/sciadv.1501392 (2016).

369 16. Frenne, P. de *et al.* Microclimate moderates plant responses to macroclimate warming.
370 *PNAS* **110**, 18561–18565; 10.1073/pnas.1311190110 (2013).

371 17. Claussen, M., Bathiany, S., Brovkin, V. & Kleinen, T. Simulated climate–vegetation
372 interaction in semi-arid regions affected by plant diversity. *Nat. Geosci.* **6**, 954–958;
373 10.1038/ngeo1962 (2013).

374 18. Huang, Y. *et al.* Impacts of species richness on productivity in a large-scale subtropical
375 forest experiment. *Science* **362**, 80–83; 10.1126/science.aat6405 (2018).

376 19. Schnabel, F. *et al.* Drivers of productivity and its temporal stability in a tropical tree
377 diversity experiment. *Glob. Chang. Biol.* **25**, 4257–4272; 10.1111/gcb.14792 (2019).

378 20. Barrufol, M. *et al.* Biodiversity promotes tree growth during succession in subtropical
379 forest. *PLoS ONE* **8**, e81246; 10.1371/journal.pone.0081246 (2013).

380 21. Kunz, M. *et al.* Neighbour species richness and local structural variability modulate
381 aboveground allocation patterns and crown morphology of individual trees. *Ecol. Lett.* **22**,
382 2130–2140; 10.1111/ele.13400 (2019).

383 22. Perles-Garcia, M. D., Kunz, M., Fichtner, A., Härdtle, W. & Oheimb, G. Tree species
384 richness promotes an early increase of stand structural complexity in young subtropical
385 plantations. *J. Appl. Ecol.* **58**, 2305–2314; 10.1111/1365-2664.13973 (2021).

386 23. Gates, D. M. & Hanks, R. J. Plant Factors Affecting Evapotranspiration. In *Irrigation of*
387 *Agricultural Lands*, edited by R. M. Hagan, H. R. Haise & T. W. Edminster (American
388 Society of Agronomy, Madison, WI, USA, 1967), pp. 506–521.

389 24. McElhinny, C., Gibbons, P., Brack, C. & Bauhus, J. Forest and woodland stand structural
390 complexity: Its definition and measurement. *For. Ecol. Manage.* **218**, 1–24;
391 10.1016/j.foreco.2005.08.034 (2005).

392 25. Ehbrecht, M., Schall, P., Ammer, C., Fischer, M. & Seidel, D. Effects of structural
393 heterogeneity on the diurnal temperature range in temperate forest ecosystems. *For. Ecol.*
394 *Manage.* **432**, 860–867; 10.1016/j.foreco.2018.10.008 (2019).

395 26. Stuenzi, S. M. *et al.* Variability of the surface energy balance in permafrost-underlain
396 boreal forest. *Biogeosciences* **18**, 343–365; 10.5194/bg-18-343-2021 (2021).

397 27. Bruelheide, H. *et al.* Designing forest biodiversity experiments: general considerations
398 illustrated by a new large experiment in subtropical China. *Methods Ecol. Evol.* **5**, 74–89;
399 10.1111/2041-210X.12126 (2014).

400 28. Peng, S., Schmid, B., Haase, J. & Niklaus, P. A. Leaf area increases with species richness
401 in young experimental stands of subtropical trees. *J. Plant Ecol.* **10**, 128–135;
402 10.1093/jpe/rtw016 (2017).

403 29. Schnabel, F. *et al.* Species richness stabilizes productivity via asynchrony and drought-
404 tolerance diversity in a large-scale tree biodiversity experiment. *Sci. Adv.* **7**;
405 10.1126/sciadv.abk1643 (2021).

406 30. Craven, D. *et al.* Multiple facets of biodiversity drive the diversity–stability relationship.

407 *Nat. Ecol. Evol.* **2**, 1579–1587; 10.1038/s41559-018-0647-7 (2018).

408 31. Isbell, F. *et al.* Biodiversity increases the resistance of ecosystem productivity to climate

409 extremes. *Nature* **526**, 574–577; 10.1038/nature15374 (2015).

410 32. Gillerot, L. *et al.* Forest structure and composition alleviate human thermal stress. *Glob.*

411 *Chang. Biol.*; 10.1111/gcb.16419 (2022).

412 33. Hartmann, H. *et al.* Climate Change Risks to Global Forest Health: Emergence of

413 Unexpected Events of Elevated Tree Mortality Worldwide. *Annu. Rev. Plant Biol.* **73**,

414 673–702; 10.1146/annurev-arplant-102820-012804 (2022).

415 34. Donfack, L. S. *et al.* Microclimate and land surface temperature in a biodiversity enriched

416 oil palm plantation. *For. Ecol. Manage.* **497**, 119480; 10.1016/j.foreco.2021.119480

417 (2021).

418 35. Kunert, N., Schwendenmann, L., Potvin, C. & Hölscher, D. Tree diversity enhances tree

419 transpiration in a Panamanian forest plantation. *J. Appl. Ecol.* **49**, 135–144;

420 10.1111/j.1365-2664.2011.02065.x (2012).

421 36. Zellweger, F. *et al.* Seasonal drivers of understorey temperature buffering in temperate

422 deciduous forests across Europe. *Global Ecol. Biogeogr.* **28**, 1774–1786;

423 10.1111/geb.12991 (2019).

424 37. Wright, A. J., Barry, K. E., Lortie, C. J. & Callaway, R. M. Biodiversity and ecosystem

425 functioning: Have our experiments and indices been underestimating the role of

426 facilitation? *J. Ecol.* **109**, 1962–1968; 10.1111/1365-2745.13665 (2021).

427 38. Huang, Y. *et al.* Plant diversity stabilizes soil temperature. *bioRxiv*;

428 10.1101/2023.03.13.532451 (2023).

429 39. Trogisch, S. *et al.* The significance of tree-tree interactions for forest ecosystem
430 functioning. *Basic Appl. Ecol.* **55**, 33–52; 10.1016/j.baae.2021.02.003 (2021).

431 40. Hüve, K., Bichele, I., Rasulov, B. & Niinemets, U. When it is too hot for photosynthesis:
432 heat-induced instability of photosynthesis in relation to respiratory burst, cell permeability
433 changes and H₂O₂ formation. *Plant, cell & environment* **34**, 113–126; 10.1111/j.1365-
434 3040.2010.02229.x (2011).

435 41. Seidelmann, K. N., Scherer-Lorenzen, M. & Niklaus, P. A. Direct vs. Microclimate-
436 Driven Effects of Tree Species Diversity on Litter Decomposition in Young Subtropical
437 Forest Stands. *PLoS ONE* **11**, e0160569; 10.1371/journal.pone.0160569 (2016).

438 42. Beugnon, R. *et al.* Abiotic and biotic drivers of tree trait effects on soil microbial biomass
439 and soil carbon concentration. *Ecol. Monogr.* **93**; 10.1002/ecm.1563 (2023).

440 43. Fichtner, A. *et al.* Neighbourhood diversity mitigates drought impacts on tree growth. *J.*
441 *Ecol.* **108**, 865–875; 10.1111/1365-2745.13353 (2020).

442 44. Schnabel, F. *et al.* Neighbourhood species richness and drought-tolerance traits modulate
443 tree growth and δ13C responses to drought. *bioRxiv*; 10.1101/2022.11.22.517351 (2022).

444 45. Dobrowski, S. Z. *et al.* Forest structure and species traits mediate projected recruitment
445 declines in western US tree species. *Global Ecol. Biogeogr.* **24**, 917–927;
446 10.1111/geb.12302 (2015).

447 46. Brancalion, P. H. S. *et al.* Global restoration opportunities in tropical rainforest
448 landscapes. *Sci. Adv.* **5**, eaav3223; 10.1126/sciadv.aav3223 (2019).

449

450 **Methods**

451 **Study site and experimental design**

452 We used data from a large-scale tree biodiversity experiment, the Biodiversity–Ecosystem
453 Functioning China Experiment (BEF-China experiment)^{18,27}, located in Xingangshan, Dexing,
454 Jiangxi (29°08'–29°11' N, 117°90'–117°93' E). The experiment was established at two sites, A
455 and B, which were planted in 2009 and 2010, respectively. Each site covers approximately 20
456 ha in size. The site's climate is governed by the subtropical monsoon, with cold and dry winters
457 and hot and humid summers. The mean annual temperature and precipitation are 16.7°C and
458 1821 mm (mean from 1971–2000)⁴⁷. Inter-annual changes in climate-induced water availability
459 are strong and driven primarily by changes in precipitation and only to a lower degree by
460 changes in temperature^{18,29}. The native forests of the study region harbour a high tree species
461 richness and are dominated by broadleaf tree species²⁷. Based on a total pool of 40 native
462 evergreen and deciduous broadleaf tree species, we created manipulated species richness
463 gradients of 1 to 24 coexisting species (Supplement S5). Overall, 226,400 individual trees were
464 planted on 566 plots, with each plot featuring a size of 25.8 × 25.8 m² (1/15 ha) and 400 trees.
465 To increase generality and statistical power, tree species were allocated to different extinction
466 scenarios following a broken-stick design with partly overlapping species pools per extinction
467 scenario^{18,27}. Here, we used data from extinction scenarios to which species were randomly
468 allocated, specifically the 64 Very Intensively Studied Plots (VIPs) of the BEF-China
469 experiment²⁷.

470

471 **Micro- and macroclimate measurements**

472 The microclimate air temperature was recorded hourly over six years (January 2015–December
473 2020) across the VIP plots (32 at each site) using temperature loggers (HOBO Pro v2, U23-
474 001) covered by a radiative shield and installed at 1 m height in the centre of the plots (see

475 Supplement S5). Data were controlled and cleaned to remove unrealistic data due to logger
476 malfunction (e.g., temperature outliers or time series divergent dynamics; Supplement S5).
477 Plots with incomplete monthly records were excluded from the monthly analyses and
478 incomplete yearly records were excluded from yearly analyses (1 plot of the 64 plots was
479 removed in all analyses; Supplement S5). Macroclimate data—minimum, average, and
480 maximum monthly temperature (°C), monthly precipitation sum (mm) and monthly potential
481 evapotranspiration (mm) sum—were retrieved from the high-resolution gridded dataset of the
482 Climatic Research Unit (CRU) Time-Series (TS) version 4.06⁴⁸ with a 0.5° (latitude/longitude)
483 resolution, which is based on interpolated climate station observations. To explore if diversity–
484 microclimate relationships were influenced by water availability, we further calculated the
485 Standardised Precipitation-Evapotranspiration Index (SPEI)⁴⁹ based on these precipitation and
486 evapotranspiration data with the SPEI package⁵⁰. The SPEI is a commonly used drought index
487 that captures the climatic water balance (precipitation minus potential evapotranspiration) at
488 different time lengths from a single month (SPEI1) to an entire year (SPEI12; January–
489 December). SPEIs below –1 and above 1 can be considered exceptionally dry or wet compared
490 to the average conditions during a climate reference period⁵¹ (here 1901–2019).

491

492 **Temperature buffering and stability**

493 Using the hourly microclimate temperature measurements, we calculated different measures
494 describing temperature extremes and temperature buffering. We calculated monthly minimum,
495 median and maximum microclimate temperatures per plot. Minimum and maximum monthly
496 temperatures were calculated by taking the median of the 95% lowest and highest temperatures,
497 respectively. We quantified temperature buffering on monthly and annual time scales as the
498 temporal stability (S)¹⁰ of microclimate temperature, calculated as:

499

500
$$S = \frac{\mu}{\sigma}$$

501

502 Where μ and σ are the mean and standard deviation of hourly temperature measurements per
503 month or year, hereafter referred to as monthly or annual temperature stability.

504

505 **Assessment of microclimate drivers**

506 We assembled a range of variables describing canopy thickness, density, and structural diversity
507 from former studies and tree inventories in the BEF-China experiment. Out of these potential
508 variables, we selected the ones with the highest relevance for temperature buffering according
509 to literature-derived hypotheses (focussing on the ones most successfully used as predictors of
510 temperature buffering in other studies; Supplement S3), and compared correlations between
511 variables (Supplement S4). Specifically, we selected mean tree height, leaf area index (LAI),
512 and Stand Structural Complexity Index (SSCI) to describe canopy thickness, density, and
513 structural diversity, respectively. Tree height was measured as the mean height of the central 6
514 \times 6 trees in each plot to avoid edge effects, as described in Huang et al.¹⁸. LAI was measured
515 using digital hemispheric photography at five positions within each plot in August by Peng et
516 al.²⁸, and SSCI by a single terrestrial laser scan at the centre of each plot under leaf-off
517 conditions of the deciduous tree species (February–March) as described by Perles-Garcia et
518 al.²² (see these studies for details on the respective methods and datasets). For all forest property
519 variables, we used data collected at site A of the BEF-China experiment in 2019 (where we had
520 the best data coverage), except for LAI, measured in 2014.

521

522 **Statistical analyses**

523 We used linear mixed-effects models (LMMs) to test for the effects of tree species richness on
524 microclimate temperatures and temperature buffering across time scales and VIP plots (n = 63

525 plots, tree species richness ranging from 1–24 species). We tested for species richness effects
526 on hourly temperatures and on minimum, median and maximum monthly temperatures using
527 LMMs in which species richness in interaction with hour or month were considered fixed
528 effects (Supplement S2). Similarly, we tested for species richness effects on monthly and annual
529 temperature stability using LMMs in which species richness in interaction with month or
530 calendar year were considered fixed effects (Supplement S2). We accounted for the
531 experimental design of our study through a nested random effect structure of plots nested within
532 the experimental site (A or B) and for temporal autocorrelation by using a first order
533 autocorrelation structure (corCAR1) for time covariates (days, months or years). Additionally,
534 we explored how diversity effects, i.e. the slopes of the regressions between species richness
535 and monthly minimum, median and maximum microclimate temperatures and monthly
536 temperature buffering (Figs. 1–2), depended on macroclimate conditions (monthly minimum,
537 average and maximum temperatures and SPEI values) (Supplement S1). At the annual scale,
538 we tested if temperature stability was related to annual climatic water balances by replacing
539 calendar years by annual SPEI values in the respective LMM (Supplement S2).

540
541 To examine the mechanisms that may mediate tree species richness effects on temperature
542 buffering, we used Structural Equation Models (SEMs). The hypothesis-driven SEMs were
543 informed by previous work, including from the herein-examined experiment (see Supplement
544 S3 for the conceptual model and the literature-derived hypotheses). Specifically, we examined
545 if canopy thickness, density, and structural diversity, captured by mean tree height, LAI and
546 SSCI, respectively, mediate tree species richness effects on temperature buffering. We
547 accounted for potential correlations between these forest properties through including partial,
548 non-directional correlations between them. We controlled for monthly variations in
549 macroclimate temperatures by dividing monthly temperature buffering values through monthly
550 macroclimate temperature values. Species richness effects on temperature buffering changed

551 over the year (Fig. 2a). Hence, to capture potential temporal changes in the strength of the
552 examined drivers, we fit separated SEMs for each month. In contrast, the strength of species
553 richness effects on temperature buffering did not change significantly between years (Fig. 2b),
554 which allowed us to focus on the year for which we had the most data on forest properties. We
555 thus explored how species richness affected temperature buffering via canopy thickness, density
556 and structural diversity in 2019 at site A, where we had measurements of all forest properties
557 (except for LAI which was measured in 2014) and where temperature buffering was close to
558 the mean response across years (Fig. 2b). Using stand-level basal area measured in 2019 instead
559 of LAI resulted in similar pathways (Supplement S4). To remain consistent with prior
560 knowledge on relationships between species richness and the examined forest properties in our
561 experiment, we fitted direct pathways between species richness and LAI and SSCI using the
562 datasets and model structures from the original studies (refs^{22,28}). Therefore, the tree species
563 richness–forest properties models were fitted on larger plot sets ($n = 32, 54$, and 74 plots for
564 mean tree height, LAI, and SSCI, respectively) than the forest properties–temperature buffering
565 models fit for the plots for which we had microclimate data and data on all examined forest
566 properties ($n = 27$ plots, tree species richness ranging from 1 – 16 species). To prevent pseudo-
567 replication caused by measuring tree height, LAI and SSCI on an annual basis, relationships
568 between tree species richness and these forest properties were fitted using yearly datasets
569 instead of monthly ones. In the tree species richness–LAI model, we included terms correcting
570 for very large residual effects resulting from the presence of few specific species in the
571 examined tree communities following Schmid et al. 2017⁵² as detailed in Peng et al.²⁸. We
572 assessed global model fit via Fisher's C statistic ($P > 0.05$) and the independence of variables
573 with tests of direct separation ($P < 0.05$ for violation of independence) and posteriori, included
574 partial, non-directional correlations between non-independent variables⁵³ (Supplement S4).

575

576 All data handling and statistical analyses were performed using the R statistical software
577 version 4.1.3. Explanatory variables in the SEMs were centred and divided by one standard
578 deviation using the ‘scale’ function, to avoid any model-fit deviation due to scale differences
579 between variables. Tree species richness was log2-transformed in all models. LMMs and
580 individual SEM pathways were fit with the nlme package⁵⁴ and SEMs with the piecewiseSEM
581 package⁵³. Model assumptions (i.e. normality, independence and homogeneity of variance, and
582 independence of explanatory variables) were tested with the ‘check_model’ function in the
583 performance package⁵⁵.

584

585 **Data availability**

586 The datasets generated and analysed in the study will be made publicly available upon
587 publication via the BEF-China project repository, at <http://data.botanik.uni-halle.de/bef-china>

588

589 **Code availability**

590 All R scripts used for this study can be found in our GitHub repository, at
591 [https://github.com/remybeugnon/Schnabel-Beugnon-Yang-et-al_tree-diversity-temperature-
592 buffering](https://github.com/remybeugnon/Schnabel-Beugnon-Yang-et-al_tree-diversity-temperature-buffering)

593

594 **Method references**

595 47. Yang, X. *et al.* Establishment success in a forest biodiversity and ecosystem functioning
596 experiment in subtropical China (BEF-China). *Eur. J. Forest. Res.* **132**, 593–606;
597 10.1007/s10342-013-0696-z (2013).

598 48. Harris, I., Osborn, T. J., Jones, P. & Lister, D. Version 4 of the CRU TS monthly high-
599 resolution gridded multivariate climate dataset. *Scientific data* **7**, 109; 10.1038/s41597-
600 020-0453-3 (2020).

601 49. Vicente-Serrano, S. M., Beguería, S. & López-Moreno, J. I. A Multiscalar Drought Index
602 Sensitive to Global Warming. The Standardized Precipitation Evapotranspiration Index. *J.*
603 *Clim.* **23**, 1696–1718; 10.1175/2009JCLI2909.1 (2010).

604 50. Beguería, S. & Vicente-Serrano, S. M. *SPEI: Calculation of the Standardised*
605 *Precipitation-Evapotranspiration Index.* (2017).

606 51. McKee, T. B., Doesken, N. J. & Kleist, J. The relationship of drought frequency and
607 duration to time scales. In *Proceedings of the 8th Conference on Applied Climatology*
608 (1993), Vol. 17, pp. 179–183.

609 52. Schmid, B., Baruffol, M., Wang, Z. & Niklaus, P. A. A guide to analyzing biodiversity
610 experiments. *J. Plant Ecol.* **10**, 91–110; 10.1093/jpe/rtw107 (2017).

611 53. Lefcheck, J. S. piecewiseSEM: Piecewise structural equation modelling in r for ecology,
612 evolution, and systematics. *Methods Ecol. Evol.* **7**, 573–579; 10.1111/2041-210X.12512
613 (2016).

614 54. Pinheiro, J., Bates, D., DebRoy, S., Sarkar, D. & R Core Team. *nlme. Linear and*
615 *Nonlinear Mixed Effects Models* (2020).

616 55. Lüdecke, D., Ben-Shachar, M., Patil, I., Waggoner, P. & Makowski, D. performance: An R
617 Package for Assessment, Comparison and Testing of Statistical Models. *J. Open Source*
618 *Softw.* **6**, 3139; 10.21105/joss.03139 (2021).

619

620 **Acknowledgements**

621 We thank local workers for their help in the field. This research was supported by the Deutsche
622 Forschungsgemeinschaft (DFG, German Research Foundation; grant DFG FOR 891), the
623 International Research Training Group TreeDì jointly funded by the DFG (grant
624 319936945/GRK2324) and the University of Chinese Academy of Sciences (UCAS). We are

625 grateful for the support of iDiv funded by the DFG (grant DFG-FZT 118, 202548816). F.S.
626 acknowledges support by a TreeDì start-up grant. R.B. acknowledges funding by the Saxon
627 State Ministry for Science, Culture and Tourism (SMWK; grant 3-7304/35/6-2021/48880),
628 N.E. funding by the DFG (grant Ei 862/29-1), H.Y. and N.E. funding by the DFG (grant FOR
629 5000), and B.S. support by the University Research Priority Program Global Change and
630 Biodiversity of the University of Zurich.

631

632 **Author Contributions**

633 H.B., K.M., B.Y., W.H., P.A.N, G.v.O., B.S. and C.W. designed the experiment; F.S., R.B. and
634 R.R. conceived the study; B.Y., H.B., F.S., R.B., X.L., A.F., M.D.P.G., G.H., W.H., M.K.,
635 N.C.C.I., P.A.N., G.v.O. and S.T. measured and/or compiled data; F.S., R.B., R.R., B.Y., N.E.,
636 Y.H., X.L., C.W., and H.B. developed and refined the analysis concept; R.B. analysed the data
637 with support by F.S. and G.H; F.S., R.B., R.R., B.Y., S.C., N.E., Y.H., C.W. and H.B interpreted
638 the data; R.B. created figures; F.S. wrote the manuscript with support by R.B; F.S., R.B., B.Y.,
639 R.R., N.E., Y.H., C.W., S.C., A.F., M.D.P.G., G.H., W.H., M.K., X.L., N.C.C.I., P.A.N.,
640 G.v.O., B.S., S.T., M.W., K.M. and H.B. contributed substantially to revisions of drafts.

641

642 The authors declare no competing interests

643 Supplementary Information is available for this paper at: Suppl.S1–S5

644 Correspondence and requests for materials should be addressed to Helge Bruelheide
645 (helge.bruelheide@botanik.uni-halle.de) and Keping Ma (kpma@ibcas.ac.cn).