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Spring frost controls spring tree phenology along elevational gradients on the southeastern Tibetan Plateau

Running head: Spring frost controls spring tree phenology

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25 **Abstract**

26 Temperature is considered to be a main driver of spring phenology, whereas the role of climate
27 extremes (such as spring frosts) has long been neglected. A large elevational gradient of mature
28 forests on the Tibetan Plateau provides a powerful space-for-time ‘natural experiment’ to explore
29 driving forces of spring phenology. Combining 5-yr of in situ phenological observations of Smith
30 fir (*Abies georgei* var. *smithii*) with concurrent air temperature data along two altitudinal
31 gradients on the southeastern Tibetan Plateau, we tested the hypothesis that spring frost was a
32 major factor regulating the timing of spring phenology. Onset of bud swelling and leaf unfolding
33 in the study years occurred \approx 18 or 17 days earlier, respectively, at the lowest (3800 m a.s.l.)
34 elevation relative to upper treelines (4360 or 4380 m a.s.l.). The frequency of freezing events and
35 last freezing date were critical factors in determining the timing of bud swelling along two
36 altitudinal gradients, whereas onset of leaf unfolding was primarily controlled by the onset of
37 bud swelling. This finding provides evidence for detrimental impacts of spring frost on spring
38 phenology, which have been underappreciated in research on phenological sensitivity to climate
39 but should be included in phenology models. It contributes to explain the declining global
40 warming effects on spring phenophases, because climatic extreme events (e.g. spring frosts) tend
41 to increase with warming.

42

43

44 **Introduction**

45 Phenology determines plant survival, growth, and distribution and plays an important role in
46 ecosystem functioning and in the provision of ecosystem services (Chuine & Beaubien, 2001;
47 Inouye, 2008; Forrest & Miller-Rushing, 2010). However, there are uncertainties about the
48 drivers of phenology due to a paucity of a range of intensively-monitored sites (Richardson *et al.*,
49 2013; Gallinat *et al.*, 2015; Piao *et al.*, 2015). Researchers attempting to generate time series long
50 enough to make inferences about climate-driven changes in phenology often accumulate just one
51 data point per year (Miller-Rushing *et al.*, 2010). Furthermore, most tree phenology data are
52 limited to first flowering and leaf unfolding and rarely consider variation in other aspects of tree
53 phenology (e.g., bud swelling) that may respond to climate differently from flowering or leaf
54 unfolding (Miller-Rushing & Primack, 2008).

55 For many species in temperate and cold ecosystems, temperature is the key factor
56 controlling onset of spring phenology (Menzel, 2003; Piao *et al.*, 2011, 2015; Richardson *et al.*,
57 2013; Huang *et al.*, 2014; Laube *et al.*, 2014; Chen *et al.*, 2015; Davis *et al.*, 2015; Fu *et al.*,
58 2015; Ge *et al.*, 2015), although other climatic factors (e.g., precipitation, radiation, and
59 photoperiod) can also play a role (Fu *et al.*, 2014a, b; Davis *et al.*, 2015; Ren *et al.*, 2015; Shen *et*
60 *al.*, 2015a). Climate extremes, in particular, are important for understanding the climatic limits of
61 tree species (Inouye, 2008; Zimmermann *et al.*, 2009), but few studies have explicitly linked
62 spring phenophases to climatic extremes, such as late-season freezing temperatures (Inouye *et al.*,
63 2008; Ernakovich *et al.*, 2014). In spring, plants experience a de-hardening period, during which
64 a certain amount of heat is required to initialize leaf unfolding (Richardson *et al.*, 2013; Fu *et al.*,
65 2014b); during this period, plants are particularly vulnerable to freezing events (Lenz *et al.*,
66 2013). Spring freezing events tend to increase in response to climate warming (Inouye, 2000;

67 Augspurger , 2013; IPCC, 2013), and thus late-spring frosts may play an increasingly critical
68 role in controlling spring phenology in temperate and cold regions (Inouye, 2008; Gu *et al.*, 2008;
69 Wang *et al.*, 2014). Indeed, one recent study showed that spring frost affects the timing of
70 bud-break which, in turn, determines the elevational and latitudinal limits of deciduous
71 broad-leaf tree species in the Alps (Kollas *et al.*, 2014). However, less is known about the
72 impacts of spring frost on spring phenology for conifers, which are the dominant species of many
73 subalpine communities.

74 The Tibetan Plateau hosts mature, natural forests across a broad elevational gradient, which
75 can be used as a space-for-time substitution for longer-term time series data in exploring key
76 drivers of tree phenology. Satellite-based observations and process-based tree-ring growth model
77 have confirmed that advancement of deciduous or semi-deciduous vegetation green-up-date on
78 the Tibetan Plateau over the past three decades is tied closely to spring warming (Piao *et al.*,
79 2011; Zhang *et al.*, 2013; Shen *et al.*, 2015b; Yang *et al.* 2017). However, the impact of spring
80 freezing events on tree phenology is unknown, although spring frosts and freezes occur
81 frequently on the Tibetan Plateau (Shen *et al.*, 2014).

82 We used a five-year dataset of precise, bud-scale measurements of spring phenophase
83 timings to explore climatic drivers of spring phenology for Smith fir (*Abies georgei* var. *smithii*)
84 along two altitudinal gradients. Specifically, we (1) revealed temporal patterns of spring
85 phenophases along two altitudinal gradients; and (2) assessed effects of freezing events and
86 growing degree days on spring phenophases. Given the frequent occurrence of spring freezing
87 events on the southeastern Tibetan Plateau during the period when bud swelling is occurring
88 (Shen *et al.*, 2014), we hypothesized that spring frost would be closely associated with the timing
89 of bud swelling.

90

91 **Materials and methods**

92 *Study region and climate*

93 The study region is situated in the Sygera Mountains (29° 10' – 30° 15' N, 93° 12' – 95° 35'E) on
94 the southeastern Tibetan Plateau. The south Asian monsoon approaches the Sygera Mountains
95 through the valley of Yarlung Zangbo River, resulting in plentiful summer rainfall (Liang *et al.*,
96 2010). Records from the Nyingchi weather station (29° 34' N, 94° 28' E, 3000 m a.s.l.) showed
97 that the mean annual precipitation from 1960 to 2013 was 672 mm, 72% of which occurred from
98 June to September (Liang *et al.*, 2010). July (mean temperature of 15.9 °C) and January (0.6 °C)
99 were the warmest and coldest months, respectively.

100 Based on an automatic weather station (4390 m a.s.l.) near the treeline on the eastern-facing
101 slopes installed in November 2006, the annual average precipitation at our study sites from 2007
102 to 2013 was 957 mm, 62 % of which fell during the monsoon season (June to September). The
103 warmest and coldest months were July (7.9 ± 0.5 °C) and February (-8.0 ± 1.7 °C), respectively.
104 Snowfall of 50-100 cm usually occurs from November to mid-May.

105

106 *Study species and study sites*

107 Smith fir (*Abies georgei* var. *smithii*) is an evergreen coniferous tree species distributed on the
108 north- or southeast-facing slopes on the southeastern Tibetan Plateau (Liang *et al.*, 2011). It
109 grows along the altitudinal gradient ranging from 3550 to 4400 m a.s.l., with growth primarily
110 constrained by low temperatures (Liang *et al.*, 2010; Li *et al.*, 2013, 2017). We studied Smith fir
111 at eight sites along two altitudinal transects: four sites on a southeast-facing slope (labelled as
112 SE3800, SE4000, SE4200 and SE4360, with the number indicating m a.s.l) and four sites on a

113 north-facing slope (N3800, N4000, N4200, N4380) (Fig. S1). At each site, 10 trees were selected
114 for measurement, except for at site SE4200 where only 6 trees were measured. We used a
115 measuring tape to determine the height of each sampled tree in April 2012, when buds were
116 dormant. We estimated tree age by counting the internodes along the main stem (see also the
117 methods in Liang *et al.*, 2011).

118

119 *Phenology measurements*

120 We made phenological observations on trees weekly between May and September during five
121 consecutive years (2012–2016). From observations of terminal buds we recorded the dates of
122 bud swelling, and leaf unfolding. The onset of bud swelling and leaf unfolding was determined
123 as the dates when trees showed swollen buds or unfolded needles in shoot apices. For all
124 monitored trees, high resolution photographs were taken at each phenological measurement visit
125 with a steel ruler (accuracy of 1 mm) placed behind the shoot apex. In the results, we report
126 “dates” as days elapsed since 1 May each year

127

128 *Temperature data*

129 Air temperature (± 0.2 °C) in each stand was measured hourly with a temperature logger (TidbiT
130 v2 Temp UTBI-001, Onset Computer Corporation, Bourne, MA, USA) that was placed 2 m
131 above the ground under the canopy of a tall mature tree. Loggers were not placed under studied
132 trees, which had small main stems. An epoxy radiation shield designed by the logger
133 manufacturer was covered each sensor to minimize effects of direct sunlight on the
134 measurements.

135 The frequency of spring freezing events was calculated as the number of days with daily
136 minimum temperature $< 0\text{ }^{\circ}\text{C}$ in spring (Shen *et al.*, 2014). The last freezing date was defined as
137 the last spring day when daily minimum temperature $< 0\text{ }^{\circ}\text{C}$ (Schwartz *et al.*, 2006). Safety
138 margins were defined as the number of days between the last freezing date and the onset of bud
139 swelling (Dantec *et al.*, 2015).

140 A winter chilling requirement is considered to be an important factor that determines the
141 onset of spring phenophases (Fu *et al.*, 2015). We first estimated the number of chilling days as
142 the sum of days in the winter when daily temperature $< 0\text{ }^{\circ}\text{C}$ (Yu *et al.*, 2010; Fu *et al.*, 2015).
143 Accumulated daily mean temperature above a certain threshold, *i.e.*, growing degree-days, also
144 has been considered to be an important factor driving the onset of leaf phenology (Fu *et al.*,
145 2014b). As compared to warmer areas, vegetation in colder environment such as Tibetan Plateau
146 requires lower threshold temperature to green up (Piao *et al.*, 2011). Thus a minimum
147 temperature of $0\text{ }^{\circ}\text{C}$ was used as a basis to accumulate degree-days, starting with the date when
148 the mean daily air temperature was $> 0\text{ }^{\circ}\text{C}$ for at least 5 consecutive days from March and
149 continuing until the onset of bud swelling and leaf unfolding.

150 *Data analysis*

151 We used regression tree modelling to investigate the influence of the many explanatory factors
152 hypothesized to be important in controlling timing of bud swelling and leaf unfolding (Table 1).
153 For the onset of bud swelling, we investigated the effects of both warmth-related variables
154 (accumulated growing degree-days, minimum, and maximum temperatures) and frost-related
155 variables (frequency of spring freezing events and last spring freezing date). We evaluated the
156 impacts of three variables (elevation, aspect, and year) on the safety margins. For leaf unfolding,
157 we sought to understand how different measures of spring warmth controlled the onset of this

158 phenophase; as leaf unfolding occurred in June, we did not include spring frost variables in these
159 models because frost was not present at this time of the year. In all models, both aspect and year

160

161 **Table 1** Description of predictor variables used in the regression tree modelling of spring phenology (date of onset of bud swelling and
 162 leaf unfolding). All climate-related variables were computed for each site and measurement year based on temperature logger data.
 163 BS and LF represented bud swelling and leaf unfolding, respectively.

Variable name	Description	In Modeling	
		BS	LF
Degree_days	Cumulative temperatures (above 0°C) from March up to phenophase onset	Yes	Yes
Min_tem	Minimum daily temperature recorded at time of phenophase onset	Yes	Yes
Max_tem	Maximum daily temperature recorded at time of phenophase onset in each year	Yes	Yes
FFE	Number of days/year with daily minimum temperature < 0°C from March to May	Yes	No
LFD	Last freezing date in spring (March to May)	Yes	No
Aspect	The aspect of a given sample plot	Yes	Yes
Year	Year of measurement (2012-2016)	Yes	Yes
Chilling days	Number of days with daily temperature < 0°C in winter	Yes	No

164

165 of measurement also were included to control for the effects of these factors related to sampling
166 design (Table 1). As elevation is a proxy for, and was significantly correlated with, several
167 temperature variables in this study, it was not included in the modelling.

168 We used a two-stage modelling approach. First we used random forest analysis (Breiman,
169 2001) to estimate and rank the importance of each explanatory factor in describing variability in
170 the response variables. Second, we used conditional inference trees to gain further insights into
171 the nature of relationships between each response variable and the most important explanatory
172 factors. These two non-parametric, machine-learning methods of analysis allow for the
173 construction of complex, non-linear models with inter-correlated predictor variables (De'ath &
174 Fabricius, 2000; Cutler *et al.*, 2007). The random forest approach averages the outcomes of
175 thousands of bootstrapped regression trees ('forests') to identify those measured explanatory
176 variables that are the best predictor variables. We used the random forest 'variable importance'
177 measure to identify the most influential factors in explaining variation in the response variable
178 and then used partial dependence plots to show the marginal effect of each of these factors (*i.e.*,
179 while holding all of the other explanatory factors at their average values) on the response
180 variable (Cutler *et al.*, 2007). The relative importance of the top-ranked predictor variables was
181 investigated further using conditional inference trees (Hothorn *et al.*, 2006) derived from a
182 recursive partitioning method that generates a set of decision rules describing how variation in
183 the response data is best attributed to each predictor. The conditional inference tree method
184 requires a statistically significant difference ($P < 0.05$), as determined by Monte Carlo simulation,
185 to create a partition in the data; this algorithm minimizes bias and prevents over-fitting and the
186 need for tree pruning (Hothorn *et al.*, 2006). Random forest and conditional inference tree

187 analyses were implemented in R version 3.1.0 using the ‘randomForest’ (Liaw & Wiener, 2002)
188 and ‘party’ (Hothorn *et al.*, 2006) packages, respectively.

189

190 **Results**

191 *Tree size*

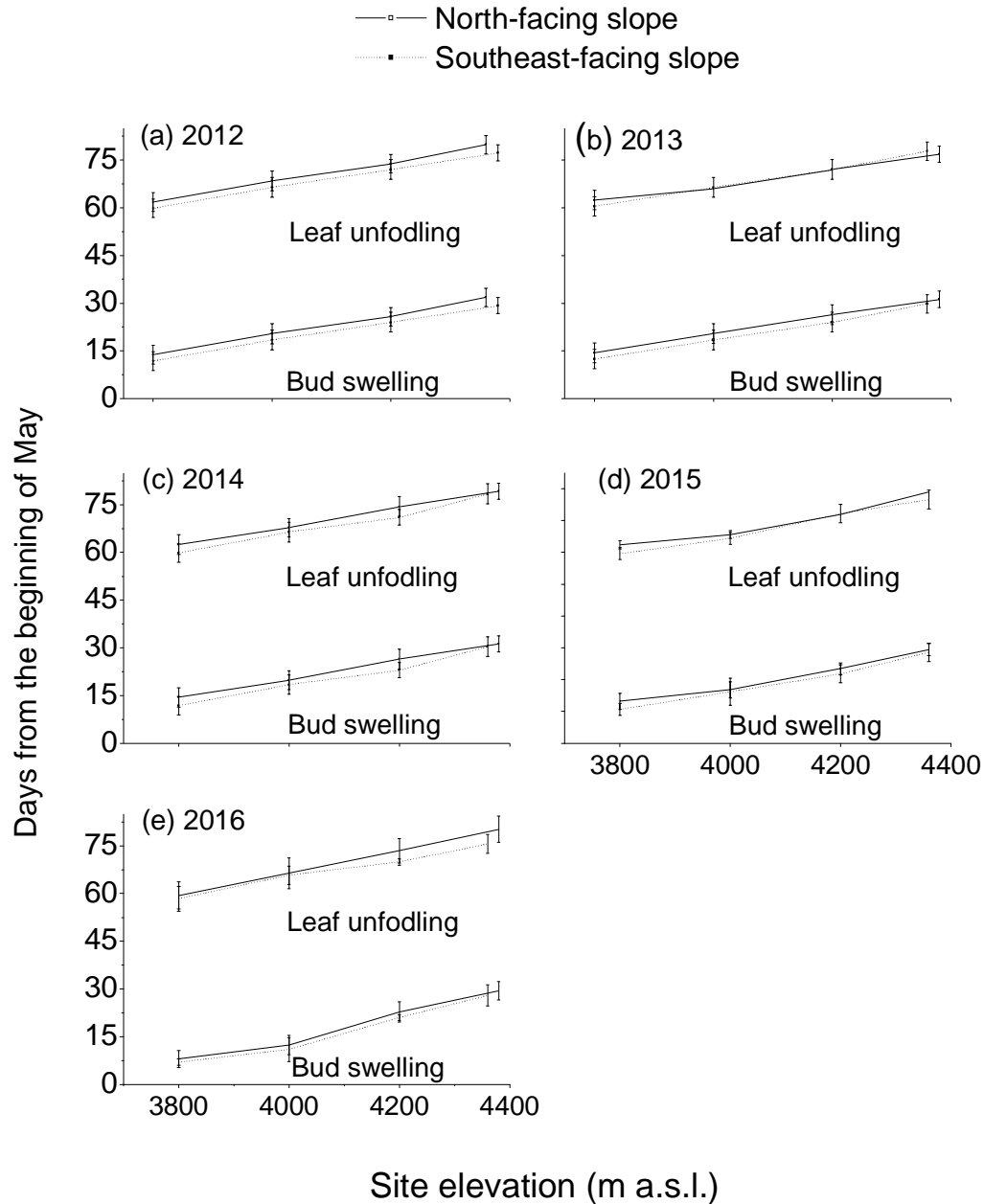
192 On the southeast-facing slope, trees ranged from 0.47 – 1.57 m in height and were estimated to
193 be 13 – 40 years old (Table S1). The trees on the north-facing slope were 0.39 – 1.83 m in height
194 and 10 – 44 years old (Table S1).

195

196 *Among-site variation in phenology*

197 Bud swelling and leaf unfolding began in early May and ended in late May, and occurred on later
198 dates at higher elevations (Fig. 1). On average, the mean lapse rate for the onset of bud swelling
199 was 3.1 ± 0.5 days/100 m in elevation gain; the difference in timing of bud swelling between the
200 lowest and highest sites was 18 ± 3 days. Leaf unfolding began in late June and ended in
201 mid-July; the mean lapse rate for this phenophase was 3.0 ± 0.6 days/100 m, leading to a
202 difference in timing of 17 ± 3 days between the lowest and highest sites. Although the lapse rate
203 and duration of each phenophase did not differ among trees on the north- and southeast-facing
204 sites, the mean dates of bud swelling and leaf unfolding were 2 ± 1 days later for trees growing
205 on the north-facing sites than southeast-facing sites.

206



207

208 **Fig. 1** Variations (means \pm SD) in onset of bud swelling and leaf unfolding with altitude and
209 aspect during four study years.

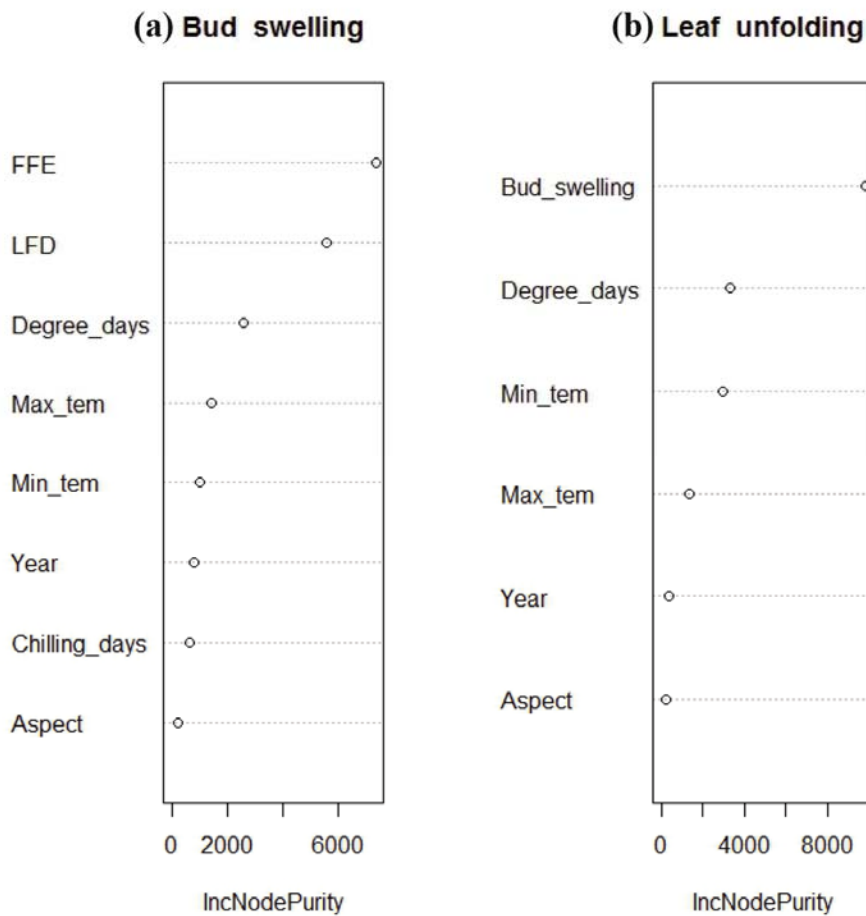
210

211 *Explanatory modelling of phenophase timings*

212 Our random forest models overall explained 84.8% of the variations in the onset of bud swelling

213 and 93.1% of the variation in the onset of leaf unfolding. The frequency of spring freezing events

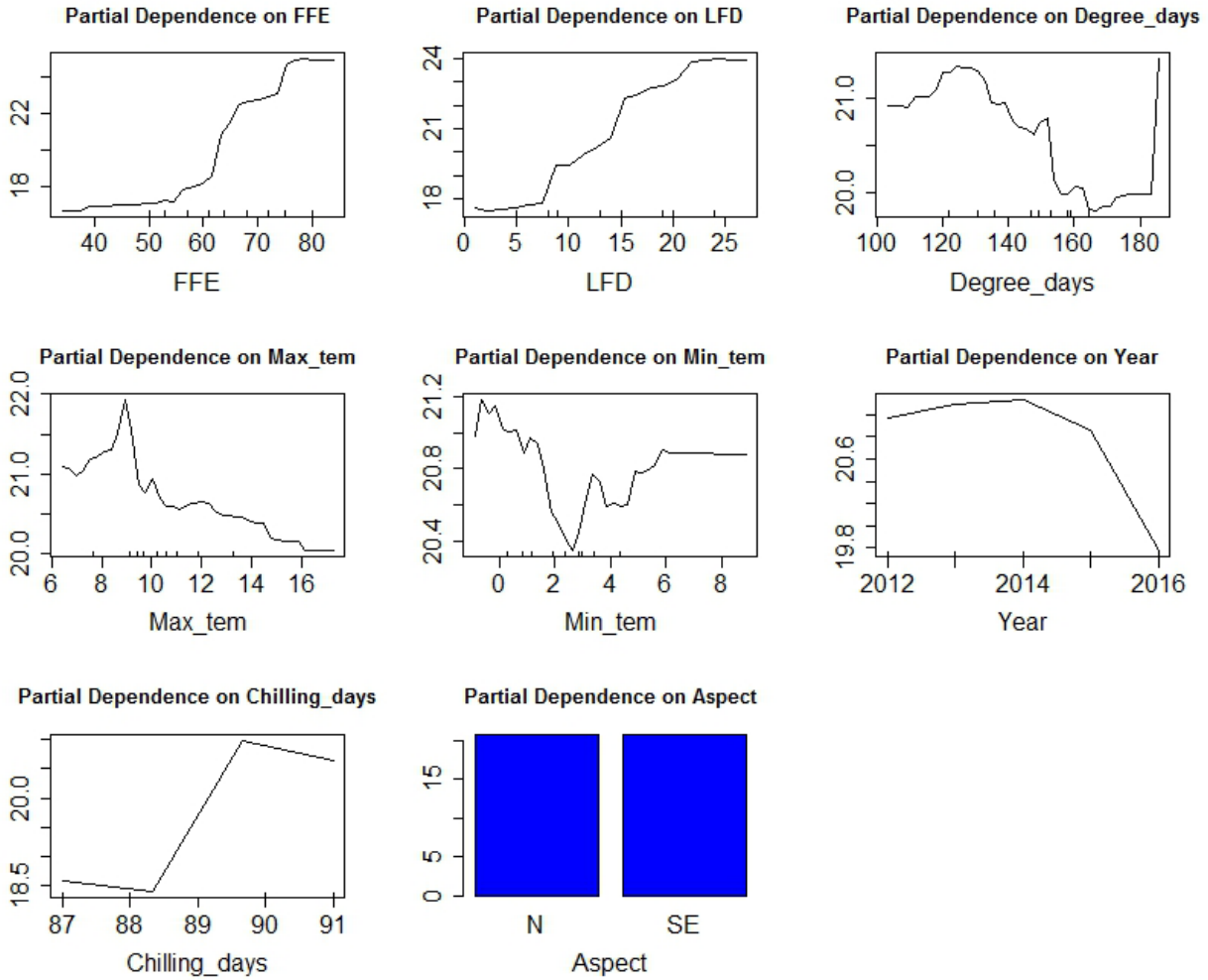
214 and the last frost date were the most important predictors of bud swelling date (Fig. 2a). Of less
215 importance were growing-degree days, maximum temperature, the minimum temperature, the
216 measurement year, chilling days, and finally aspect (Fig. 2a). Mean safety margins were $\geq 7 \pm 5$
217 days and increased significantly with increasing elevation ($r = 0.24$, $P < 0.0001$, $n = 364$) and
218 among years ($r = 0.27$, $P < 0.0001$, $n = 364$), but not with aspect ($r = 0.04$, $P = 0.36$, $n = 364$).
219 For leaf unfolding, onset of bud swelling was the most important predictor; the number of
220 growing degree-days, minimum and maximum temperature ranked considerably lower, but were
221 basically equivalent, in importance, followed by measurement year and aspect (Fig. 2b).



222

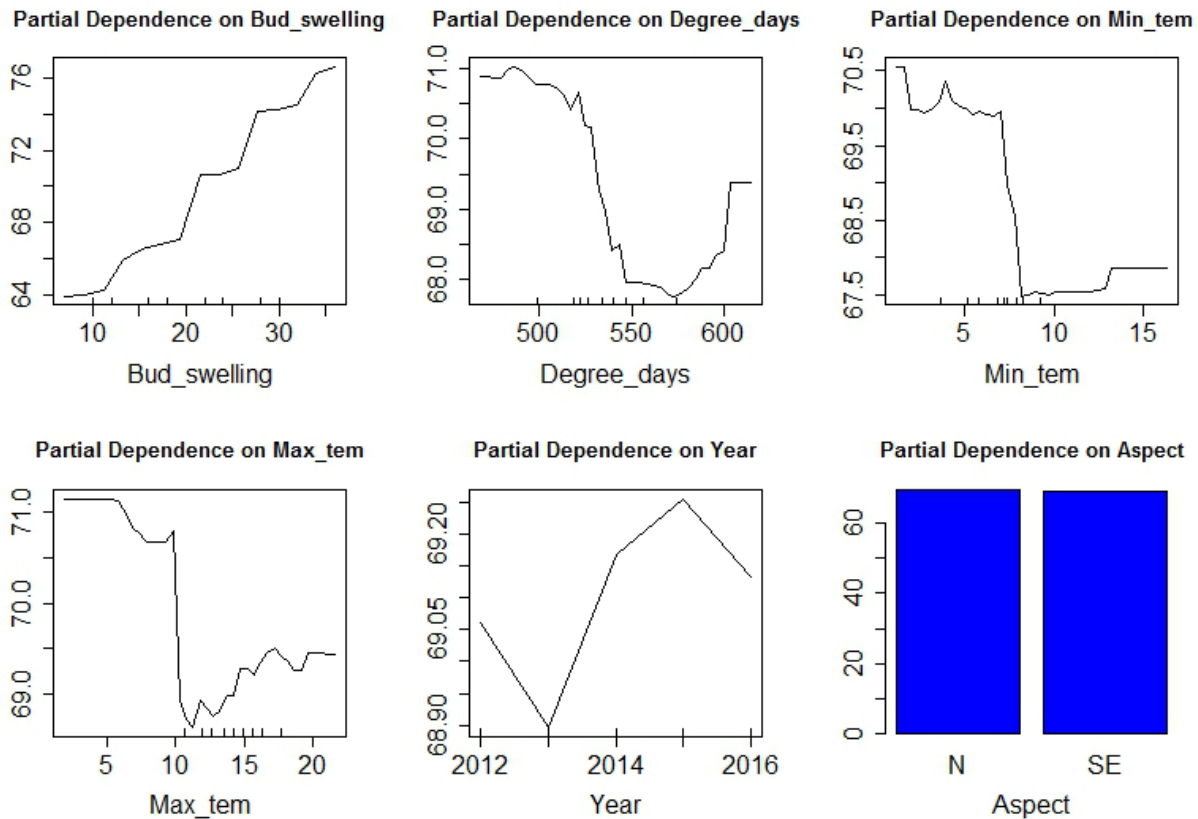
223 **Fig. 2** Ranked, relative importance of variables included in random forest models explaining
224 variation in the onset dates of (a) bud swelling and (b) leaf unfolding.

225 **(a)**



226

227 (b)



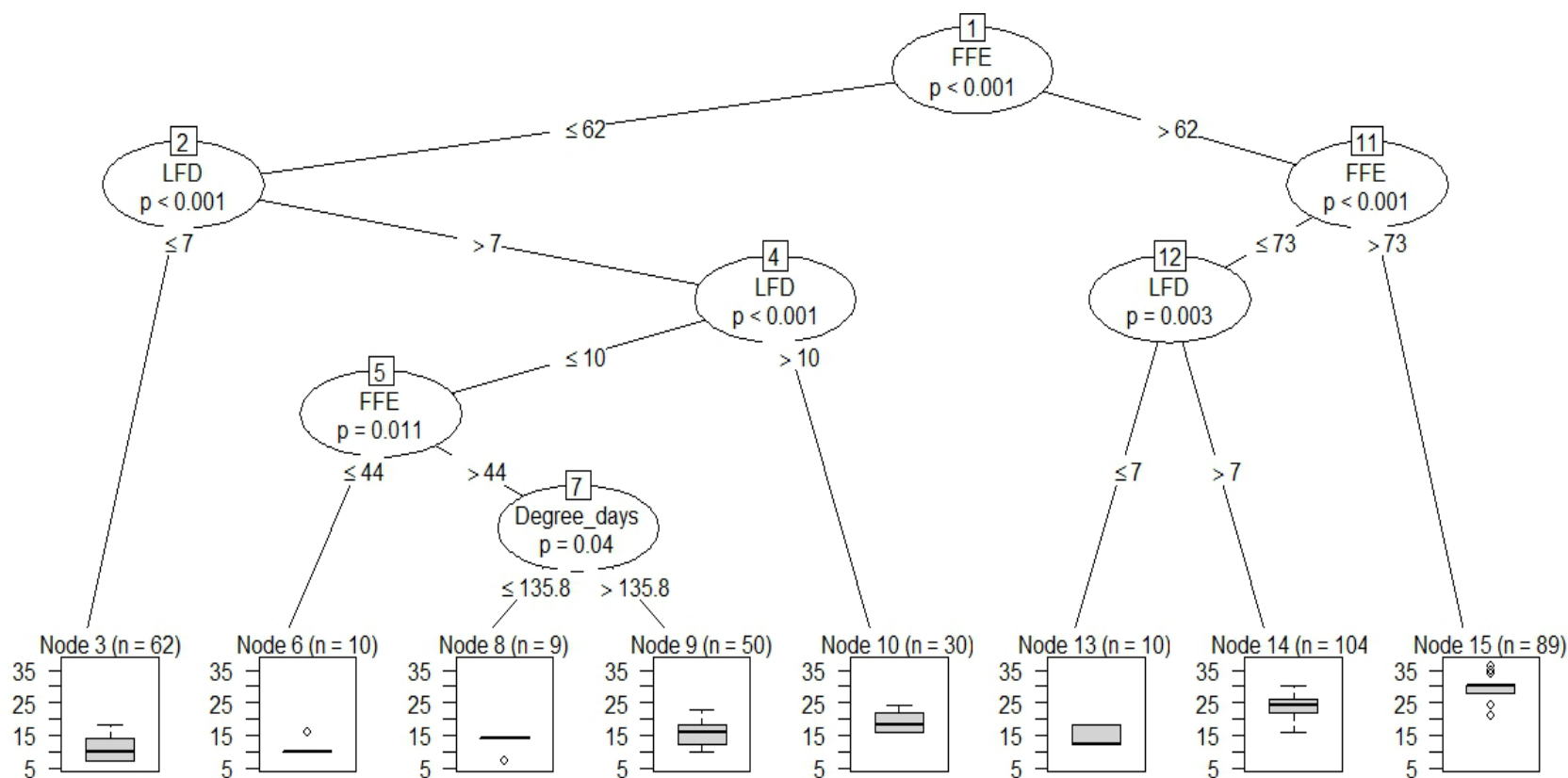
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229 **Fig. 3** Partial dependence plots, based on results from the random forest analysis, showing the
230 mean marginal influence of predictor variables on the onset date of (a) bud swelling and (b) leaf
231 unfolding. Each plot represents the effect of one predictor variable on the response, while
232 holding the other predictor variables constant at their mean values.

233 Partial dependence plots indicated that the three response variables often were related
234 non-linearly to the predictor variables (Fig. 3), which in turn interacted in complex ways (Fig. 4).
235 Bud swelling was positively and essentially linearly related to both the frequency of spring
236 freezing events and the last freezing date, negatively related to the sum of growing degree-days
237 and minimum temperature (Fig. 3a), and non-linearly related to the other variables (Fig. 3a).
238 Conditional inference tree modelling suggested that whether bud swelling occurred at later or

239

240 (a)



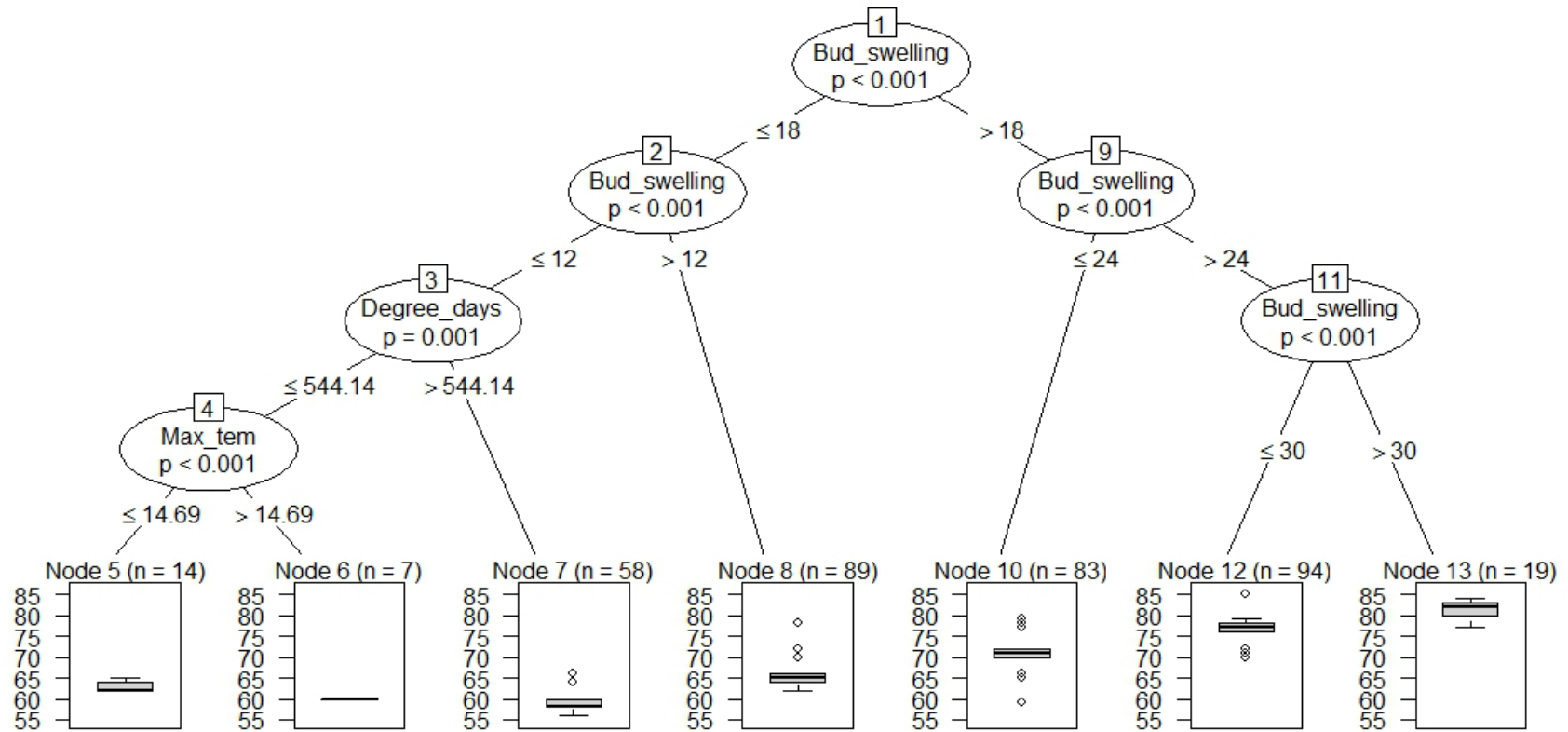
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245 (b)



246

247 **Fig. 4** Conditional inference trees explaining variation in the onset date of: (a) bud swelling and (b) leaf unfolding based on sets of
248 predictor variables (see Table 1). The trees show pathways of how the response data were recursively partitioned based on predictor
249 variables. The observations associated with each terminal node are the result of these partitionings. P values at each node are from a
250 Monte Carlo randomization test; in order for a split to occur $P < 0.05$.

251 earlier dates was largely controlled by the frequency of spring freezing events, with locations
252 experiencing more than 62 freezing days in spring having the latest bud swelling dates; for
253 earlier-onset locations (less than 62 freezing days), other factors such as last freezing date, the
254 sum of growing degree days interacted to explain variability in earlier bud swelling timing (Fig.
255 4a).

256 The timing of leaf unfolding onset was positively and linearly related to onset of bud
257 swelling, and had a negative, *s*-shaped relationship with growing degree days, with a switch from
258 later to earlier timing of leaf unfolding past a critical value for the sum of growing degree-days
259 that fell within the range of 530 – 540 accumulated °C (Fig 4b); a similar critical minimum
260 temperatures of 7 – 7.5 °C was also evident (Fig. 3b). The relationship between leaf unfolding
261 date and maximum temperature was more complex (Fig. 3b) likely due to interaction effects with
262 the sum of growing degree days, maximum and minimum temperature in explaining finer
263 variation in this phenophase (Fig. 4b).

264

265 **Discussion**

266 A number of climatic variables can influence the onset and duration of plant phenophases, but
267 research to date has tended to focus on temperature means rather than variance or extremes
268 (Inouye, 2008; Wang et al., 2014), and on later phenophases such as flowering and leaf
269 emergence. In contrast, the data we present here illustrates that timing of extreme events such as
270 late spring frost and freezes, can strongly affect two phenophases, onset of bud swelling and leaf
271 unfolding. The onset of bud swelling is of particular importance, because it is a prerequisite to
272 the other important and well-studied phenophases. Furthermore, we took advantage of five years

273 of data across a steep elevational gradient to provide additional information on climatic control
274 of bud swelling and leaf unfolding.

275 The date of onset of bud swelling increased with elevation in all five years of this study. This
276 result can be attributed directly to effects of temperature, as tree growth of Smith fir at high
277 elevations is known to be limited by temperature (Liang *et al.*, 2010; Wang *et al.*, 2012).
278 However, at high altitudes, trees are frequently exposed to large diurnal temperature fluctuations
279 in spring (Ernakovich *et al.*, 2014) and meristematic tissues are especially vulnerable to damage
280 from spring freezing (Gu *et al.*, 2008). Once development starts in spring, freezing resistance is
281 irreversibly lost and trees cannot re-acclimate to low temperatures (Lenz *et al.*, 2013). In
282 particular, the freezing resistance of trees decreases quickly as temperature increases during the
283 de-hardening period (Lenz *et al.*, 2013). In some cases, abnormally warm weather followed by
284 sudden cold waves (particularly freezing events) in early to mid-spring can have disastrous
285 impacts on plants effects (Gu *et al.*, 2008). Thus, adaptations for avoidance of spring freezing
286 damage are critical for the survival and subsequent development of many tree species in
287 temperate and cold regions (Kollas *et al.*, 2014). In our study, the frequency of freezing events
288 and the date of the last hard freeze were the critical factors in regulating the timing of bud
289 swelling. The normal date in late May of the last freeze that we observed matched that seen in
290 the timing of snow melt at treeline in another study (Liu *et al.*, 2011). Safety margin for spring
291 frost increased significantly with elevation, suggesting the strong directional selection due to late
292 spring frosts and freezes (Dantec *et al.*, 2015). The negative effects of spring freezes on the
293 survival of Smith fir seedlings (tree age \leq 5yrs) had also been reported on the southeastern
294 Tibetan Plateau (Shen *et al.*, 2014). It is likely that Smith fir escapes from the spring freezing
295 injury at the upper treelines by delaying bud swelling until very late spring (Wang *et al.*, 2014).

296 Our results in combination with those from these other studies together support our
297 hypothesis that frost-avoidance occurs in early phenophases of Smith fir, and our data provide
298 the first empirical evidence that onset of spring phenology on the Tibetan Plateau is controlled
299 primarily by spring frost, rather than the sum of growing degree-days.

300 Overall, these results are in line with phenological studies of deciduous broad-leaf trees in
301 the eastern and western Alps (Lenz *et al.*, 2013; Kollas *et al.*, 2014) and Mount Fuji in Japan
302 (Gansert, 2002), but different from what has been observed in warmer temperate forests where
303 accumulation of growing degree-days is related more closely to phenological events (Peñuelas &
304 Filella, 2001; Parmesan & Yohe, 2003; Wang *et al.*, 2011; Dai *et al.*, 2014). Our results partly
305 explain the declining global warming effects on spring phenophases (Shen *et al.*, 2013), because
306 climatic extreme events (e.g. spring frosts) tend to increase with warming (Inouye, 2000;
307 Augspurger, 2013; IPCC, 2013). Although the accumulation of chilling days has been shown to
308 regulate the responses of spring phenology to climatic warming in Europe (Fu *et al.*, 2015), the
309 impact of chilling on the timing of bud swelling was negligible in our study. In addition,
310 green-up date for grassland on the Tibetan Plateau is statistically related with mean minimum
311 temperature during the pre-season in both arid and wet regions (Shen *et al.*, 2016). Comparisons
312 between the findings of Shen *et al.* (2016) and our results suggest that spring phenology of
313 different alpine plant function types might respond to different temperature variables.

314 The onset of leaf unfolding from the lowest to highest sites occurred in the warmest period
315 of the growing season (from late June to mid-July) when freezing events were absent, and was
316 driven more by the onset of bud swelling. This result suggested that earlier bud swelling
317 translated into earlier leaf unfolding. Presumably, the annual growth cycle of trees forms an
318 integrated system, where one phenophase can affect or regulate the subsequent phase (Fu *et al.*,

319 2014a). Such carryover effects have also been reported in temperate and boreal forests (see
320 review in Fu *et al.*, 2014a). Unexpectedly, the sum of growing degree-days played a secondary
321 role in controlling onset of leaf unfolding. These results are different from other studies
322 conducted in temperate, boreal, and some alpine forests where accumulation of degree-days is
323 the major determinant of the onset of leaf unfolding (Peñuelas & Filella, 2001; Parmesan & Yohe,
324 2003; Wang *et al.*, 2011; Dai *et al.*, 2014).

325 Last, our data provide evidence that warming may promote an advance in the timing of
326 spring phenophase. The elevational transect that we studied represents a temperature gradient of
327 ca. 3.8 °C, or a lapse rate of -0.66 °C/100 m (Liang *et al.*, 2011). Therefore if the regional
328 temperature warms by ≈ 1 °C, spring phenology could advance by 4.5 days, similar to the
329 advancement of 4.6 days °C found in a meta-analysis of temperate plants around the world
330 (Wolkovich *et al.*, 2012). It appears that the temperature sensitivity of bud swelling and leaf
331 unfolding phenology is somewhat greater than the sensitivity of flowering phenology which was
332 found to be 3.3 days/ 1°C warming (Miller-Rushing & Primack, 2008). Earlier spring
333 phenologies that accompany a warming climate also may help to explain some observed upward
334 shifts in alpine treelines, The 1.2 – 1.5 °C warming observed on the Tibetan Plateau in the last
335 100 years has been paralleled by an up to 80-m upslope shift in treeline if species interactions do
336 not constrain these shifts (Liang *et al.*, 2016).

337

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345

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- 473

474 **Supporting information captions**

475 **Table S1** *Tree height and age of Smith fir individuals at four elevations along two altitudinal*
476 *transects. Note that the height and age of each sampled tree was determined in April 2012, when*
477 *buds were dormant.*

478

479 **Fig. S1** *Location of the study sites, the meteorological station at Nyingchi (3,000 m a.s.l.) in the*
480 *southeastern Tibet. SE and N and the number represent the slopes and the site elevation (m a.s.l.),*
481 *respectively. Inset (upper right corner) indicates the position of the study region on the Tibetan*
482 *Plateau.*

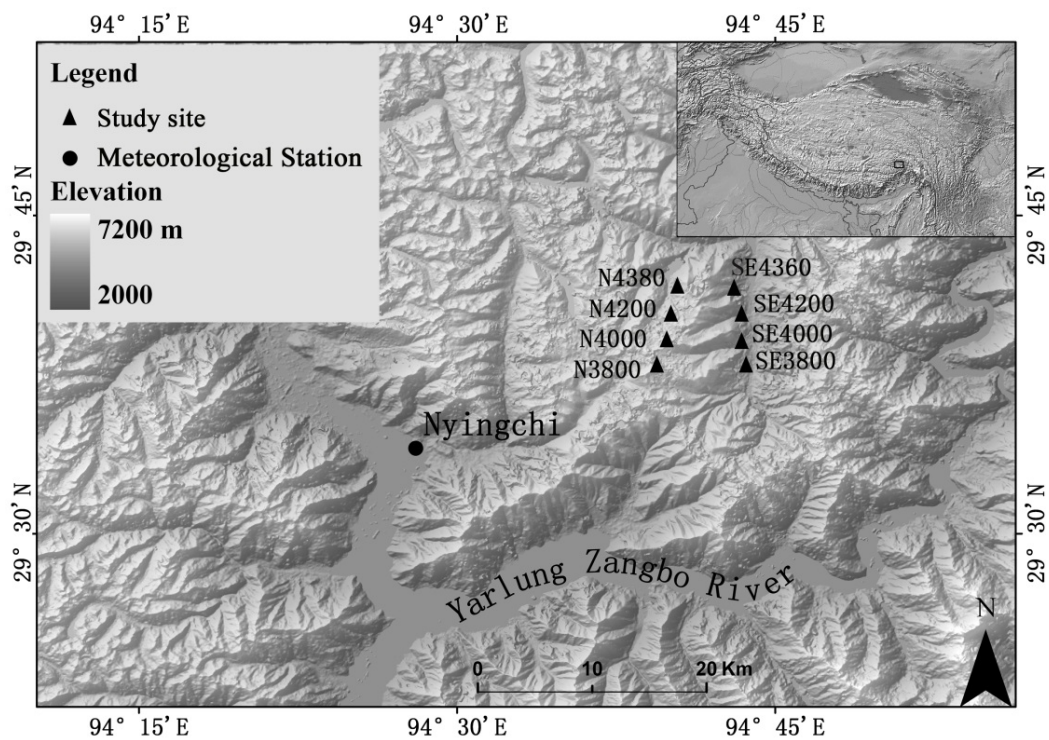
483 **Supporting information**

484 **Table S1**

Transect 1	Code of trees	Tree height (cm)	Age (years)
SE4360	SE4360-1	47	26
	SE4360-2	133	40
	SE4360-3	119	28
	SE4360-4	48	18
	SE4360-5	108	34
	SE4360-6	98	31
	SE4360-7	73	26
	SE4360-8	53	20
	SE4360-9	93	28
	SE4360-10	99	24
SE4200	SE4200-1	77	16
	SE4200-2	120	21
	SE4200-3	130	20
	SE4200-4	79	13
	SE4200-5	69	20
	SE4200-6	48	13
SE4000	SE4000-1	79	18
	SE4000-2	68	21
	SE4000-3	138	16
	SE4000-4	133	16
	SE4000-5	102	18
	SE4000-6	133	18
	SE4000-7	84	22
	SE4000-8	94	14
	SE4000-9	80	14
	SE4000-10	84	19
SE3800	SE3800-1	154	20
	SE3800-2	124	18
	SE3800-3	118	18
	SE3800-4	98	18
	SE3800-5	70	14
	SE3800-6	149	18
	SE3800-7	141	18
	SE3800-8	157	17
	SE3800-9	74	17
	SE3800-10	155	15

Transect 2	Code of trees	Tree height (cm)	Age (years)
N4380	N4380-1	85	25
	N4380-2	105	24
	N4380-3	125	31
	N4380-4	130	34
	N4380-5	120	26
	N4380-6	120	29
	N4380-7	165	27
	N4380-8	90	23
	N4380-9	170	28
	N4380-10	155	21
N4200	N4200-1	145	33
	N4200-2	157	36
	N4200-3	165	39
	N4200-4	120	32
	N4200-5	140	38
	N4200-6	130	39
	N4200-7	110	38
	N4200-8	110	28
	N4200-9	115	27
	N4200-10	130	44
N4000	N4000-1	45	10
	N4000-2	145	16
	N4000-3	70	10
	N4000-4	170	24
	N4000-5	113	19
	N4000-6	144	18
	N4000-7	183	22
	N4000-8	72	13
	N4000-9	70	13
	N4000-10	73	11
N3800	N3800-1	150	17
	N3800-2	115	20
	N3800-3	130	18
	N3800-4	165	24
	N3800-5	62	21
	N3800-6	180	26
	N3800-7	150	34
	N3800-8	80	29
	N3800-9	55	11
	N3800-10	39	10

485



486

487 **Figure S1**

488