

1 **Original article:**

2 Forest management drives evolution of understorey herbs

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Abstract

23 1. Forest management has a strong impact on the forest structure and subsequently on the
24 biotic and abiotic forest understorey environment. Forest understorey herbs can thus be
25 expected to evolutionary respond to management-induced environmental variation
26 (provided sufficient time for adaptation), but this has been little tested to date.

27 2. Here we use a common garden, to test for genetically based variation in phenotypic
28 traits in populations of forest herbs sampled along a forest management intensity
29 gradient. Five different herbaceous species were sampled from 70-100 populations in
30 three regions in Germany and were tested for genetically based variation in flowering
31 start, proportion of flowering ramets, and plant height. Additionally, we investigated
32 the effects of management-induced environmental variation and performed structural
33 equation modelling to study how forest management drives trait differentiation *via* its
34 effects on the microenvironment.

35 3. We found that the studied forest understorey herbs varied genetically in the measured
36 functional and phenological traits among the sampled populations. Forest management
37 likely affected the traits in various directions and strengths depending on the species,
38 either directly through variation in forest structural attributes or indirectly through
39 changes in the microclimatic environment on the forest floor.

40 4. *Synthesis:* We show that forest management can have evolutionary consequences for
41 forest understorey plants. In an applied context, diverse forest management actions
42 within landscapes thus creates heterogeneity that selects for different plant traits and
43 thus helps conserving genetic diversity.

44 **Keywords:** common garden, evolutionary ecology, forest structure, genetic differentiation,
45 intraspecific trait variation, microclimate, structural equation modelling.

47 In our rapidly changing world, land use has been recognized as having the strongest impact on
48 species and genetic diversity (IPCC, 2014; Tobias *et al.*, 2021). The present state of temperate
49 forests in Europe is a product of decisions made by former and current management, rather
50 than natural factors and progression (Spiecker, 2003). Current silvicultural management
51 practices have a profound impact on forest ecosystems, as they can affect forest structure, for
52 instance by altering crown projection area, structural complexity, and tree species composition.
53 Large, dense tree crowns cast shade, buffer against warming and reduce daytime summer
54 temperatures inside forests (De Frenne *et al.*, 2013). Furthermore, light availability is highly
55 dependent on the composition of tree species, which in turn depends on site-specific climatic
56 factors such as water availability (Aussenac, 2000). Existing silvicultural management and
57 conservation actions supporting coppicing, selective tree extractions or other actions can lead
58 to canopy openings, potentially causing nutrient depletion of the soil, a decreasing biomass
59 pool, increasing nitrogen release in the soil, but also warming of the forest floor, likely causing
60 changes in the composition of understorey herb species (De Frenne *et al.*, 2013).

61 Forest understorey herbs play a key role in forest ecosystems (Augusto *et al.*, 2003;
62 Gilliam, 2007; Landuyt *et al.*, 2019), harbouring the highest species diversity among all strata
63 in temperate forests (Gilliam, 2007). In temperate forests, the phenology of early-growing
64 understorey herbs should be attuned to abiotic cues, such as spring temperature, in order to
65 prevent early frost or to finish their reproductive cycle before tree leaf out (Heberling *et al.*,
66 2019). High management intensity for example can result in delayed flowering in understorey
67 herbs (Willems *et al.*, 2021). More specifically, low spring temperatures were associated with
68 intensely managed conifer plantations, with understorey herbs flowering on average two weeks
69 later than in unmanaged forests. However, this study is based on field observations only, and
70 thus cannot infer whether the results are due to phenotypic plasticity or genetic differentiation.

71 Forest management, its influence on the forest floor environment, as well as how it
72 impacts adaptation of understorey herbs, are still underexplored. Soil pH is highly influential
73 on plant phenology and growth, and affects traits such as germination (Gentili *et al.*, 2018),
74 plant height and number of flowers (Jiang *et al.*, 2017). Some aspects of acidic soils can be
75 advantageous for plant growth, as more micronutrients are available to plants compared to
76 neutral-alkaline soils (Loncaric *et al.*, 2008). Studies have found that plants germinate and
77 flower earlier on acidic soils, but species differences in the number of flowers (Gentili *et al.*,
78 2018; Sercu *et al.*, 2021). In general, functional trait responses to forest structural variation and
79 microenvironment depend on the phenology and distribution of the forest understorey herb,
80 and can therefore be highly species-specific (De Frenne *et al.*, 2009; Blondeel *et al.*, 2020;
81 Willems *et al.*, 2021).

82 One response of forest understorey plants to silvicultural management actions and
83 microenvironmental changes, is through the employment of genetically based phenotypic
84 variation to respond appropriately to stress and disturbances (Lemke *et al.*, 2015). Mutations,
85 genetic drift and gene flow are neutral processes affecting genetically based phenotypic
86 variation and population genetic structure, and genetic variation within populations is a
87 necessary prerequisite for adaption to and persistence under changing conditions (Lemke *et al.*,
88 2015). As sessile organisms, plants are vulnerable to strong environmental changes, and
89 especially forest understorey herbs with limited dispersal capabilities may likely be adapted to
90 very specific and local environments. Because forest offer relatively temporally stable
91 environments, we could expect understorey herbs to adapt evolutionary to spatial
92 environmental variation rather than to evolve phenotypic plasticity (Alpert and Simms, 2002;
93 Valladares *et al.*, 2007). However, few studies have looked at genetic adaptation of forest
94 understorey herbs to their local environment (De Frenne *et al.*, 2011; Møller *et al.*, 2022). A
95 useful strategy to investigate evolutionary adaptation is by applying a common garden

96 experiment, because it can reveal genetic differentiation among sampled populations, followed
97 by correlative approaches to test for associations between trait variation and environmental
98 variation at sites of population origins (Turesson, 1922; Scheepens and Stöcklin, 2013).

99 Here we established a common garden using five understorey herbs sampled across a
100 forest management intensity gradient, to (1) investigate whether populations of forest
101 understorey herbs differ genetically in phenotypic variation; (2) If so, how does among-
102 population phenotypic trait differentiation correlate with forest structural variables and
103 microenvironmental variables? And (3) does the forest structure have a direct effect on
104 genetically based phenotypic variation in forest understorey herbs, or indirect effects through
105 the microenvironment?

106

107

Material & methods

108 *Study species*

109 For this study, we chose five perennial forest understorey herbs, four forbs – *Anemone*
110 *nemorosa*, *Galium odoratum*, *Oxalis acetosella* and *Viola reichenbachiana* – and one grass –
111 *Milium effusum* (Fig. 1).

112 *Anemone nemorosa* (Ranunculaceae) and *O. acetosella* (Oxalidaceae) are both early-
113 flowering, small forbs, reaching heights of 5-25 cm. They rely heavily on clonal spread
114 (Frederiksen and Rasmussen, 2006). By contrast, *G. odoratum* (Rubiaceae) and *V.*
115 *reichenbachiana* (Violaceae) flower later, grow taller, reach heights of 10-30 cm, and rely more
116 on seed dispersal (Frederiksen and Rasmussen, 2006). Lastly, *M. effusum* (Poaceae) is a tall-
117 growing grass, reaching heights of 50-150 cm, and relies heavily on wind-dispersed seeds (De
118 Frenne *et al.*, 2011). All species can spread vegetatively to some degree and vary in
119 colonization rates (Brunet *et al.*, 2012).

120 *Experimental set-up*

121 In 2020, plants were sampled from plots set up within the framework of the Biodiversity
122 Exploratories (www.biodiversity-exploratories.de), a large-scale platform for ecological
123 research (Fischer *et al.*, 2010). The Biodiversity Exploratories contains three regions across
124 Germany: Schwäbische Alb, Hainich-Dün, and Schorfheide-Chorin (Fischer *et al.*, 2010). Each
125 region consists of 50 forest plots, *i.e.*, 150 plots across the three regions.

126 In May 2020, all forest plots were visited. Six genets were sampled for each species (hereafter
127 referred to as individuals), whenever they were occurring on a plot. To avoid sampling
128 genetically identical plants, each individual was sampled with a minimum distance of 10 m to
129 the next sampled plant. Individuals were planted in multitrays (51.5 cm width, 33.5 cm length,
130 5.5 cm deep, 54 pots per tray) filled with potting soil (“CL T torffrei”, Einheitserde, Sinnatal-
131 Altengronau, Germany) for root establishment and growth. In November 2020 all individuals
132 were transferred into 1.5 L pots with potting soil (“Typ T, Struktur 1B”, Hawita, Vechta,
133 Germany). In spring 2021, all pots were relocated to a shading house (Götsch & Fälschle,
134 Alerheim, Germany), under common-garden conditions, situated at the Goethe University
135 Frankfurt (Frankfurt am Main, Germany). All individuals received water *ad libitum* during the
136 whole experiment.

137 *Plant measurements*

138 In total, we sampled 2474 individuals: 606 *Anemone nemorosa*, 545 *Galium odoratum*, 467
139 *Milium effusum*, 430 *Oxalis acetosella*, and 426 *Viola reichenbachiana*, from 70-100
140 populations. However, mortality caused heavy species-dependent losses after transplantation,
141 mainly in February 2021 due to a period of deep frost following a warm spell. One year later,
142 in the growing period of 2022, we tracked flowering phenology, counted ramets and measured
143 vegetative traits on the remaining 1,129 individuals: 126 *A. nemorosa*, 166 *G. odoratum*, 178

144 *M. effusum*, 361 *O. acetosella* and 298 *V. reichenbachiana* (See Table S1 for distribution
145 between populations of sampled individuals).

146 Phenological measurements were taken three times per week (every Monday,
147 Wednesday, and Friday). Since *Anemone nemorosa* is a vernal geophyte, emergence date was
148 recorded only for this study species, while flowering start and flowering end were recorded for
149 all study species. Dates were transformed into Julian days (day of the year; DOY). Flowering
150 peak was calculated using one-hot encoding (Lantz, 2019), transforming the data to be
151 categorical in order to count how many individuals, from each species, were flowering on a
152 given day. To calculate the proportion of flowering ramets, the number of flowering ramets
153 and total number of ramets were counted for *A. nemorosa*, *G. odoratum* and *M. effusum*. To
154 infer the total number of ramets for *O. acetosella*, we applied the formula for the frustum of
155 cone volume measurement:

$$156 \quad \frac{1}{3} \pi * h(R^2 + Rr + r^2) \quad (1)$$

157 with h being the height of the individual, R being the radius of the lower base, and r being the
158 radius of the upper base. We consider that each leaf connects to a rhizome and is therefore
159 potentially independent; all leaves were counted for a subset of 12 pots, based on a subjective
160 stratified sampling of small, intermediate, and large plants, and a regression between the
161 frustum volume (Eq. (1)) and the number of leaves ($R^2 = 0.93$, see Fig. S1) was used to predict
162 the number of leaves in the remaining pots. The proportion of flowering ramets was then
163 calculated as the total number of flowers divided by the predicted number of leaves. For *V.*
164 *reichenbachiana*, major stems were used as a proxy for ramets, and the number of major stems
165 and number of major flowering stems were counted and used to calculate the proportion of
166 flowering stems, hereafter referred to as proportion of flowering ramets. Plant height was

167 measured for *A. nemorosa*, *G. odoratum*, *M. effusum* and *O. acetosella*, but not for *V.*
168 *reichenbachiana* as height is not a meaningful variable in this species.

169 Forest structural attributes and microclimatic variables for each population of origin
170 were extracted from the BExIS2 (www.bexis.uni-jena.de) database, maintained by the
171 Biodiversity Exploratories (Fischer *et al.*, 2010), for the years 2018 and 2019, *i.e.*, previous to
172 the plants being sampled, to catch relatively recent environmental variation leading up to the
173 sampling in 2020. Structural complexity index (SCI), an index quantifying stand structural
174 complexity based on the fractal dimension of cross-sectional polygons (Ehbrecht *et al.*, 2017),
175 crown projection area (m²/ha) of mature trees, and the percentage of conifers based on crown
176 projection, was obtained for all sampled plots. These forest attributes were selected because
177 they characterize stand structure and are expected to have an influence on the biotic and abiotic
178 microclimatic conditions, as well as on light availability (Willems *et al.*, 2021). In addition, we
179 also obtained soil pH and the annual mean spring temperature 10 cm above the forest floor
180 (calculated as the average of March, April, and May of 2018 and 2019) as microenvironmental
181 variables.

182 *Data analyses*

183 All statistical analyses were conducted with R version 4.1.2 (R Core Team, 2021). First, to test
184 for population differences in genetically based phenotypic variation, we ran linear mixed-effect
185 models (LMM) for all traits, using the function “lmer” from the package “lme4” (Bates *et al.*,
186 2007). The measured plant traits were used as the response variables, population, species, and
187 their interaction as the explanatory variable, and region was included as a random factor.

188 Second, to investigate the effect of forest attributes and microclimate from the
189 population of origin on the measured traits in the common garden, we ran LMMs with
190 structural complexity, crown projection area, percentage of conifers, soil pH, spring

191 temperature, and the two-way interaction between species and all the aforementioned as
192 explanatory variables. Region and population were used as random nested structure
193 (Region/Population). LMMs were run using the function “lmer” from the packages “lme4”
194 (Bates *et al.*, 2007). The model test results were obtained by applying the “Anova” function
195 from the package “car” (Fox *et al.*, 2012). Post hoc tests were applied to investigate significant
196 differences among species using the package “lsmeans” (Lenth, 2016).

197 Third, to investigate the direct and indirect effects of the forest attributes and
198 microclimatic variables on the measured plant traits, we performed structural equation
199 modelling (SEM) separately for each trait. Due to the significant species interactions in the
200 LMMs of the second study question, SEMs were done separately for each species. We used
201 the package “piecewiseSEM” (Lefcheck, 2016), which is particularly suitable for handling
202 LMMs with random factors. In each individual SEM, a measured plant trait (flowering start,
203 proportion of flowering ramets or plant height) was included as the response variable, with
204 region and population used as random nested structure. Spring temperature and pH were
205 included as explanatory variables for the plant trait, but also as response variables for structural
206 complexity, crown projection area and percentage of conifers. Additionally, these forest
207 structural attributes were included as direct explanatory variables for the measured plant traits
208 and could also correlate amongst each other. The complete data set varied between SEMs, but
209 each overall path model was evaluated based on a Fischer’s C statistic and on its degrees of
210 freedom and was accepted as an adequate model if the baseline model had a P-value > 0.05.
211 Marginal and conditional R²-values were calculated for each response variable, the former
212 describing the proportion of variance explained by fixed factors alone, the latter describing the
213 variance explained by both fixed and random factors. Despite starting with a model based on
214 *a priori* knowledge of direct and indirect interactions, pathways with a P-value > 0.1 were
215 removed from the final visualization of the SEMs.

216

217

Results

218 *Population differentiation and effect of forest structure and microclimatic variables on plant*

219 *traits*

220 Emergence of *A. nemorosa* showed a marginally significant difference among populations of
221 origin (Table 1). Flowering start and plant height differed significantly among populations of
222 origin across species as well as in the interaction between populations and species (Table 1).
223 The proportion of flowering ramets varied significantly among populations across species
224 (Table 1).

225 Flowering start was, across all species, significantly delayed in plants from sites with
226 higher soil pH. Additionally, the significant interaction between soil pH and species indicated
227 that species differed in their relationship between flowering start and soil pH (Table 2). A post-
228 hoc Tukey test revealed that *V. reichenbachiana* significantly differed from both *O. acetosella*
229 and *M. effusum* (Tukey P-value = <0.001 and 0.018, respectively) and this species – in contrast
230 to the cross-species pattern – showed delayed flowering start in plants from sites with lower
231 soil pH (Fig. S2). Furthermore, the interaction between mean yearly spring temperature and
232 species, showed significant variation among flowering start in species (Table 2), whereas *M.*
233 *effusum* was found to significantly differ from *O. acetosella* (Tukey P-value = 0.040), and stood
234 out from the other species by delaying flowering start in individuals originating from colder
235 sites (Fig. S3).

236 The proportion of flowering ramets showed significant variation among species. A
237 post-hoc Tukey test revealed that *A. nemorosa* significantly differed from *V. reichenbachiana*
238 and *G. odoratum* (Tukey P-values = <0.001 and 0.040, respectively). Similarly, *O. acetosella*
239 significantly differed from *V. reichenbachiana*, *G. odoratum*, and *M. effusum* (Tukey P-values

240 = <0.001, 0.026, and <0.001, respectively), and together with *A. nemorosa*, showed the lowest
241 proportion of flowering ramets (Fig. 1, Fig. S4). At the other end of the scale, *M. effusum*
242 showed a higher proportion of flowering ramets, significantly differing from the rest of the
243 species (Tukey P-value: *G. odoratum* = <0.001), except for *V. reichenbachiana*. *Viola*
244 *reichenbachiana* had the highest proportion of flowering ramets of all species, and significantly
245 differed from all species except *M. effusum* (remaining Tukey P-value to be explained: *G.*
246 *odoratum* = <0.001). Lastly, *G. odoratum* differed significantly from all species and took an
247 intermediate position (Fig. 1, Fig. S4).

248 The significant interaction between soil pH and species for plant height showed that
249 species differed in their relationship between plant height and soil pH (Table 2). Here, *M.*
250 *effusum* differed significantly from *A. nemorosa* and *O. acetosella* (Tukey P-value = 0.019 and
251 0.005, respectively), and increased its height with increasing soil acidity (Fig. S5).

252 *Direct and indirect effects of forest management on understorey herbs*

253 Multiple SEMs revealed that understorey herbs were affected directly by forest structural
254 attributes and indirectly through the microclimate. However, these effects showed different
255 patterns among species. We observed in almost all individual SEMs that our forest structural
256 attributes had strong relationships with each other. Low structural complexity was associated
257 with small crown projection area (found in 100% of SEMs; Fig. 2-4) and with a low percentage
258 of conifers (100%) present at the sites of origin, and a large crown projection area was itself
259 also associated with a high percentage of conifers (57.1%). Although significant paths varied
260 depending on the specific trait or species investigated, we saw a general pattern of soil pH
261 decreasing with percentage of conifers (78.6%) but increasing with structural complexity
262 (14.3%) and crown projection area (35.7%). Furthermore, spring temperature decreased with

263 increasing percentage of conifers (78.6%) and structural complexity (21.4%) but increased with
264 increasing crown projection area (78.6%).

265 Emergence date of *A. nemorosa* was negatively associated with higher soil pH (Fig.
266 2A), meaning that individuals originating from populations with high soil pH emerged earlier.
267 Individuals of *O. acetosella* flowered later if originating from populations with a high soil pH
268 or high spring temperature (Fig. 2D). The flowering start of *G. odoratum*, *V. reichenbachiana*
269 and *M. effusum* in the common garden was not significantly affected by any environmental
270 variables from the sites of origin (Fig. 2BCE).

271 The proportion of flowering ramets decreased with increasing mean spring
272 temperatures in *A. nemorosa* (Fig. 3A), decreased with increasing soil pH in *O. acetosella* and
273 *G. odoratum* (Fig. 3BD), and was not affected by any variables from the sites of origin for *V.*
274 *reichenbachiana* and *M. effusum* (Fig. 3CE).

275 The height of *O. acetosella* individuals increased with increasing mean spring
276 temperatures (Fig. 4D). Additionally, the plant height of *G. odoratum* was directly affected by
277 forest structure, with individuals growing taller if they originated from populations in more
278 structurally complex forests (Fig. 4B). Neither the plant height of *A. nemorosa* nor that of *M.*
279 *effusum* was significantly affected by any variables from the sites of origin (Fig. 4AC).

280

281 **Discussion**

282 Our common garden study using five forest understorey herbs revealed genetic differentiation
283 among populations in ecologically important traits. Moreover, direct effects of forest structural
284 attributes as well as indirect effects through forest floor microenvironment influenced variation
285 in these traits, although most responses appeared to be species-specific. Taken together, this
286 suggests that forest management influences the evolution of understorey herbs.

287 *Genetically based phenotypic variation in five forest understorey herbs*

288 We showed that understorey herbs harbour significant variation in functional traits depending
289 on their population of origin. As the measurements were obtained from individuals in a
290 common garden rather than *in situ* in the original populations, this phenotypic variation reflects
291 genetic differentiation among populations (Turesson, 1922; Scheepens and Stöcklin, 2013),
292 although we should be aware that epigenetic effects may explain part of this variation (Bossdorf
293 et al. 2008). This genetic differentiation could be the result of adaptation to local conditions,
294 so our next steps was to investigate associations between trait values and environmental values,
295 which would support adaptation as the cause for this genetic variation (Scheepens and Stöcklin,
296 2013). In particular, we investigated associations with forest structural attributes, which are
297 strongly related to forest management decisions, as well as microenvironmental variables that
298 may be influenced by forest management. However, since we did not find overall main effects
299 and since species often differ in the influence of certain environmental factors on plant traits,
300 we will instead focus on the species-specific SEMs revealing contrasting patterns.

301 *The effects of forest structure and microenvironment*

302 Concerning the species-specific SEMs, we will first discuss the general patterns found among
303 the forest structural attributes, before examining the patterns found in the forest floor
304 microenvironment. Lastly, we will dissect each species-specific SEM and discuss how the
305 individual traits are responding to the forest structural attributes and the microenvironment at
306 their sites of origin.

307 General patterns among forest structural attributes were found across all SEMs.
308 Structural complexity was negatively associated with both crown projection area and the
309 percentage of conifers at the sites of origin. Since coniferous forests are often associated with
310 high forest management intensity (Schall and Ammer, 2013; Willems *et al.*, 2021), we often

311 see cohorts of homogenous coniferous plantations leading to low structural complexity
312 (Pommerening, 2004). Similarly, a large crown projection area, caused by coniferous-
313 dominated forests or a monospecific deciduous forest with large crowns, could be the potential
314 cause of the negative association with structural complexity. In our dataset, conifers had larger
315 crown areas compared to deciduous trees, supporting the positive relationship between
316 percentage of conifers and crown projection area (Grote, 2003). However, it has been shown
317 that the crown radii of mixed stands and conifer stands are generally overestimated, while
318 deciduous crowns are underestimated, due to overlapping of crowns in dense and structurally
319 complex stands (Grote, 2003).

320 Concurrent with previous studies (*e.g.*, Willems *et al.* 2021), we found that a higher
321 percentage of conifers decreased mean spring temperatures. Most coniferous trees are
322 evergreen, and hence, the forest floor beneath them is constantly shaded, decreasing the
323 temperature experienced by the understory herb community. Large crown projection areas
324 create a bigger “forest roof”, which not only buffers the temperature on the forest floor and
325 increases mean spring temperatures, but also protects against wind and other mechanistic
326 stresses (De Frenne *et al.*, 2021). Structural complexity creates heterogenous microclimates at
327 fine spatiotemporal scales, and various canopy openings could disrupt the buffering of the
328 “forest roof”, ultimately decreasing spring temperatures (De Frenne *et al.*, 2021).

329 Conifer species are often planted on acidic soils, but they can also contribute to decrease
330 soil pH even further by promoting soil aluminium contents (Augusto *et al.*, 2003). Stands with
331 a high structural complexity usually harbour deciduous trees or a mix of deciduous and
332 coniferous trees. In other words, they usually have a low percentage of conifers, which could
333 explain the positive relationship with soil pH. Large crown projection area can be characterized
334 by dense and overlapping deciduous tree crowns, and deciduous trees are associated with more
335 alkaline soils compared to coniferous stands (Augusto *et al.*, 2003).

336 It is crucial for spring geophytes to have appropriate responses to thermal cues to avoid
337 frost damage on the one hand but also utilize the maximum amount of high light on the other
338 hand in order to successfully finish a reproductive cycle before tree leaf out (Heberling *et al.*,
339 2019). Individuals originating from colder, often coniferous, plots might be adapted to
340 phenological cues at lower temperatures, which persists when they are transplanted to a
341 common garden in Frankfurt, thus having an advance phenological stage in the common garden
342 compared to individuals originating from warmer plots. In line with previous studies (Gentili
343 *et al.*, 2018), *A. nemorosa* individuals originating from more acidic soils emerged earlier in our
344 common garden, establishing the importance of pH as a key factor and soil property. In line
345 with this, the phenological cue for flowering start in *O. acetosella* in our common garden, is
346 later in individuals originating from warmer and more alkaline sites (Sercu *et al.*, 2021), and
347 based on our SEMs these sites mainly consist of deciduous tree species.

348 The proportion of flowering ramets can be seen as a proxy for the amount of resources
349 an individual chooses to allocate to sexual reproduction *versus* vegetative reproduction.
350 *Anemone nemorosa* favours vegetative reproduction during colder spring temperatures and
351 hence conserves valuable resources for warmer periods when investment in flowers is more
352 successful owing to higher pollinator activity (Hegland *et al.*, 2009). Additionally, we saw that
353 both *O. acetosella* and *G. odoratum* reduce their amount of flowering ramets in individuals
354 originating from acidic soils. Previous studies (e.g., Blondeel *et al.*, 2020; Sercu *et al.*, 2021)
355 report similar results, pointing towards the importance of pH on nutrient availability and the
356 potential importance of pH as a driver of trait variation (Gentili *et al.*, 2018).

357 Plant height was the only measured trait that was directly affected by forest structural
358 attributes. The structural complexity of the forest canopy heavily influences light availability,
359 a resource causing competition within the understorey herb community, and these biotic forces
360 could potentially cause increased height in *G. odoratum* (Pau *et al.*, 2011). Additionally,

361 increased spring temperatures allow for greater carbon assimilation (Heberling *et al.*, 2019)
362 and for *O. acetosella* to invest in increased plant height.

363 Why do we not see any influence of forest or microenvironmental variables on the traits
364 of *M. effusum* and *V. reichenbachiana*? Despite showing evidence for genetic differentiation,
365 some species are not responding to the microenvironmental variables from their sites of origin.
366 It could very well be that we picked the wrong environmental variables that could explain it,
367 or that the variation is neutral with respect to any environmental variables.

368 Previous work by Willems *et al.* (2021) was carried out on the same forest plots, but as
369 the results were based on field observations alone, no conclusion about the genetic basis could
370 be made. With our complimentary work, we now established that the effects of forest
371 management found under natural conditions (Willems *et al.*, 2021) is a result of genetic
372 differentiation and thus evolution at small spatial scales. Since our plants originated from
373 within three regions in Germany, we are not comparing forests from across a large spatial range
374 subject to macroclimatic variation. Instead, all observed genetic variation occurs within these
375 three regions, despite gene flow counteracting population differentiation, especially in the
376 species with higher dispersal abilities (notably *M. effusum*). An important implication is that
377 all observed genetic variation can thus also be conserved at this spatial scale. By applying
378 diverse forest management actions across regional scales, we create heterogeneity that likely
379 selects for different plant traits and thus helps conserving genetic diversity.

380 *Conclusion*

381 In conclusion, forest understorey herbs showed genetic differentiation in ecologically
382 important traits among the sampled populations. Our results suggests that forest management,
383 which strongly determines the forest structural variables included in this study (Ehbrecht *et al.*,
384 2017), drives the evolution of forest understorey herbs either directly or indirectly through its

385 influence on the microenvironment. Given that forest understorey herbs are important for
386 biodiversity and ecosystem services, it is imperative that future management considers this.
387 Diversifying management practices can affect intraspecific genetic diversity and thus can play
388 a role in the conservation of genetic diversity of understorey herbs.

389

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406

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412

413 **Data availability statement**

414 This work is based on data elaborated by the HerbAdapt and Forest Structure (core) projects
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416 publicly available in the Biodiversity Exploratories Information System
417 (<http://doi.org/10.17616/R32P9Q>, <https://www.bexis.uni-jena.de/ddm/data/Showdata/22766>,
418 and <https://www.bexis.uni-jena.de/ddm/data/Showdata/31455>).

419

420 **Conflict of interests**

421 The authors have no conflict of interest to declare.

422

423 **Author contributions**

424 CM, PDF and JFS designed the study. CM conducted the field sampling and the common
425 garden experiment with help from MMS. CM and PDF analysed the data with input from TV,
426 KV and JFS. CM wrote the first draft of the manuscript with all co-authors contributing to
427 revisions. Acquisition of the financial support of the project leading to this publication by PDF
428 and JFS.

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528 **Table 1:** Results of mixed-effects models of emergence (for *Anemone nemorosa* only), flowering start, plant height and the proportion of flowering ramets in
529 five different forest understorey herbs explained by population, species, and their interaction as fixed factors, and region as random variable. Chi-square (X^2) test
530 values, degrees of freedom (df), and P-values (P) are given. Significant P-values are indicated in bold followed by * (<0.05), ** (<0.01) and *** (<0.001), and
531 marginally significant values are indicated in bold followed by ·.

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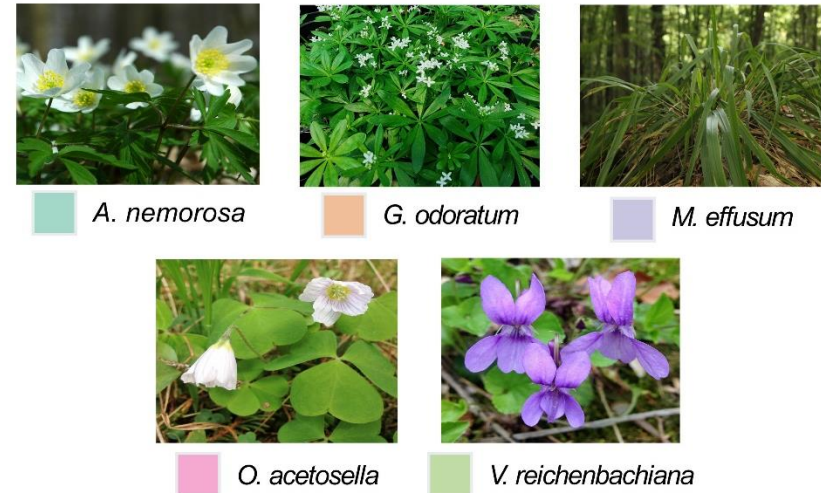
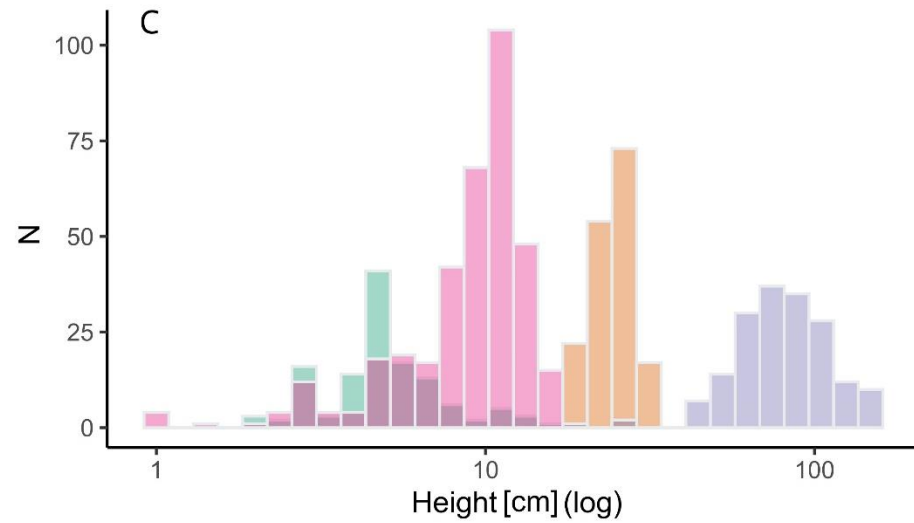
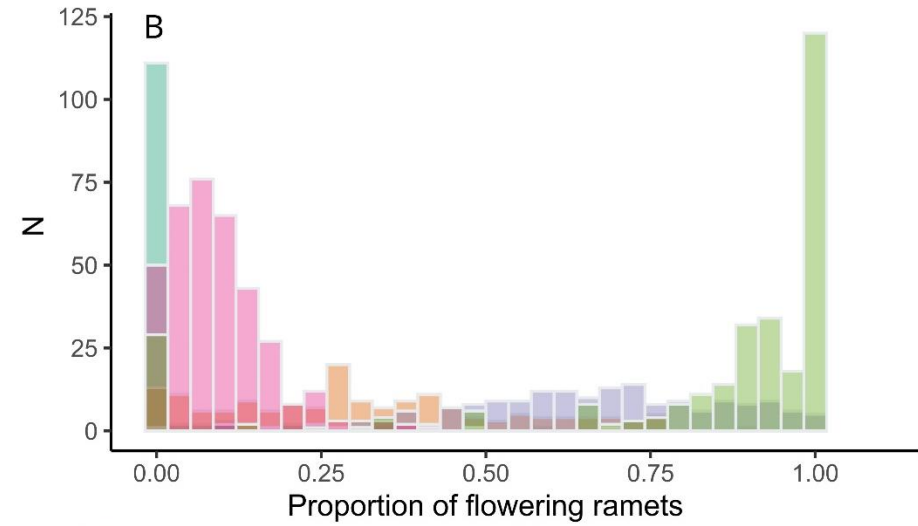
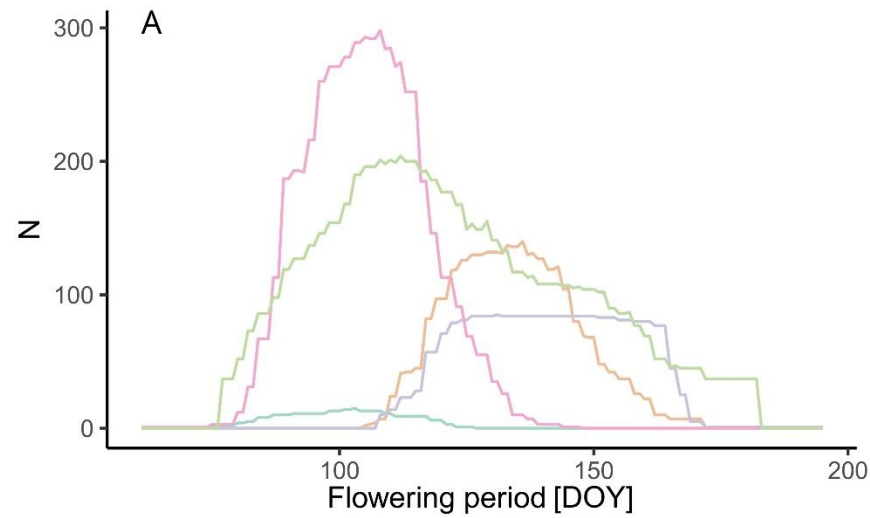
	Emergence			Flowering start			Proportion of flowering ramets			Plant height		
	X^2	df	P	X^2	df	P	X^2	df	P	X^2	df	P
Population	85.40	67	0.064 ·	160.0	116	0.004 **	162.5	127	0.018 *	258.3	123	<0.001 ***
Species				0.946	4	0.918	3.839	4	0.428	3.135	3	0.371
Population * Species				220.7	164	0.002 **	216.5	0.439	0.439	214.6	148	<0.001 ***

533

534 **Table 2.** Results of mixed-effects models of emergence (for *Anemone nemorosa* only), flowering start, plant height and the proportion of flowering ramets in five
535 different forest understorey herbs explained by species, environmental variables, and their interactions with species as fixed factors, and with region and
536 populations included as nested random variables. For continuous variables, upwards arrows (↑) indicate positive estimates and downwards arrows (↓) indicate
537 negative estimates. Chi-square (X^2) test values, degrees of freedom (df), and P-values (P) are given. Significant P-values are indicated in bold followed by *
538 (<0.05), ** (<0.01) and *** (<0.001), and marginally significant values are indicated in bold followed by ·.

539

	Emergence			Flowering start			Proportion of flowering ramets			Plant height					
	X^2	df	P	X^2	df	P	X^2	df	P	X^2	df	P			
Species					4	0.997	212.78	4	<0.001 ***	1.792	3	0.617			
pH	↓	5.663	1	0.017 *	↓	1	0.021 *	↓	2.032	1	0.154	↓	1.633	1	0.201
Spring temperature	↓	1.210	1	0.271	↓	1	0.428	↓	0.416	1	0.519	↓	1.254	1	0.263
Conifer	↓	0.113	1	0.737	↓	1	0.952	↓	2.237	1	0.135	↓	1.373	1	0.241
Structural complexity	↓	0.646	1	0.422	↓	1	0.799	↓	0.535	1	0.465	↓	0.977	1	0.323
Crown projection area	↑	0.061	1	0.805	↓	1	0.331	↓	0.653	1	0.419	↓	0.020	1	0.888
pH * Species					4	<0.001 ***	1.243	4	0.871		13.35	3	0.004 **		
Spring temperature * Species					4	0.047 *	3.440	4	0.487		3.495	3	0.321		
Conifer * Species					4	0.770	8.825	4	0.066 ·		7.146	3	0.067 ·		
Structural complexity * Species					4	0.954	0.781	4	0.941		5.667	3	0.129		
Crown projection area * Species					4	0.958	1.448	4	0.836		1.377	3	0.711		



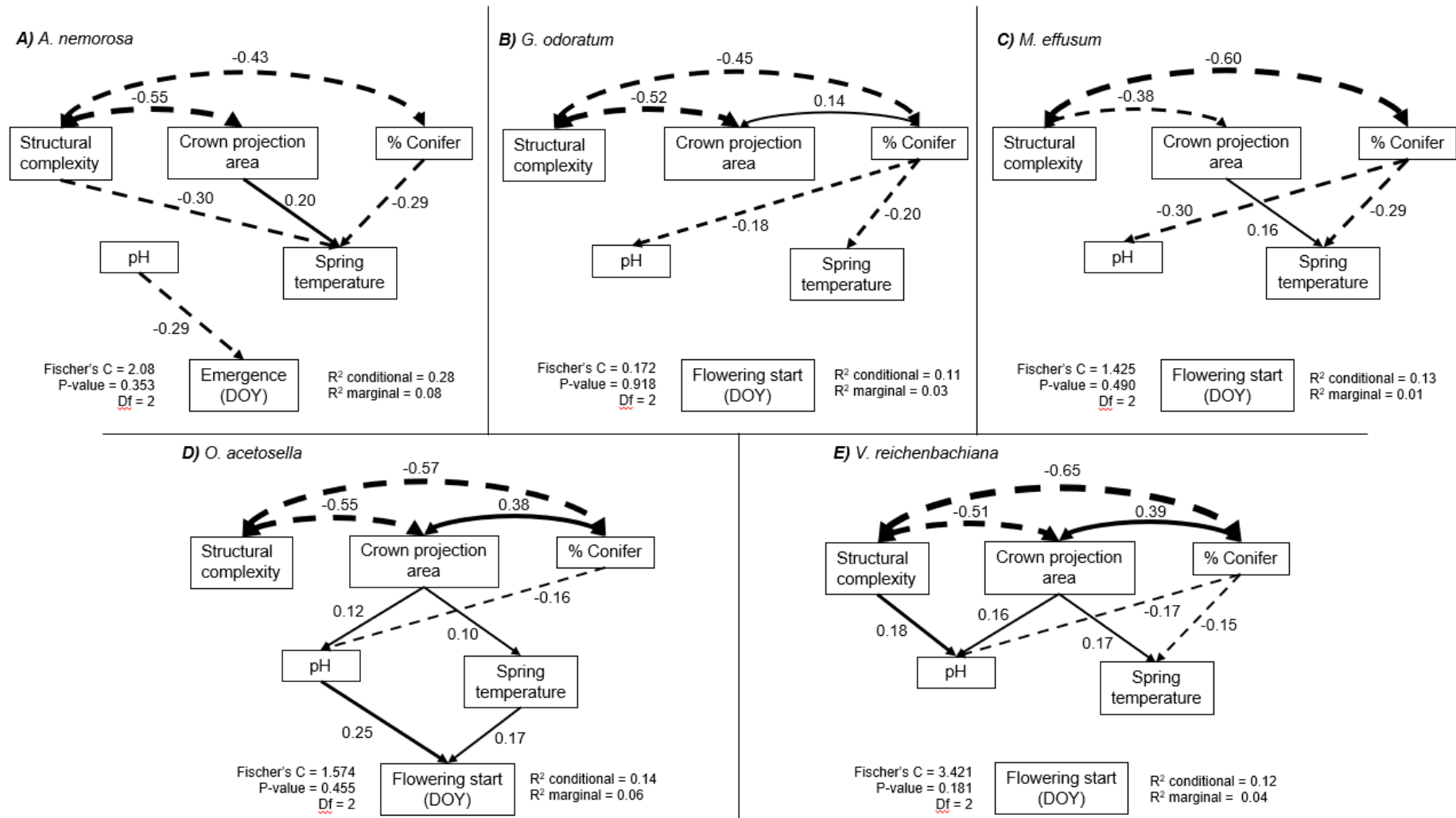
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Figure 1. Trait distribution and number of observations (N) for (A) flowering period [DOY], (B) proportion of flowering ramets, and (C) plant height [cm] (x-axis logged) for *Anemone nemorosa*, *Galium odoratum*, *Milium effusum*, *Oxalis acetosella*, and *Viola reichenbachiana*. Photo credits: Franziska Willems, Pieter De Frenne & Charlotte Møller.



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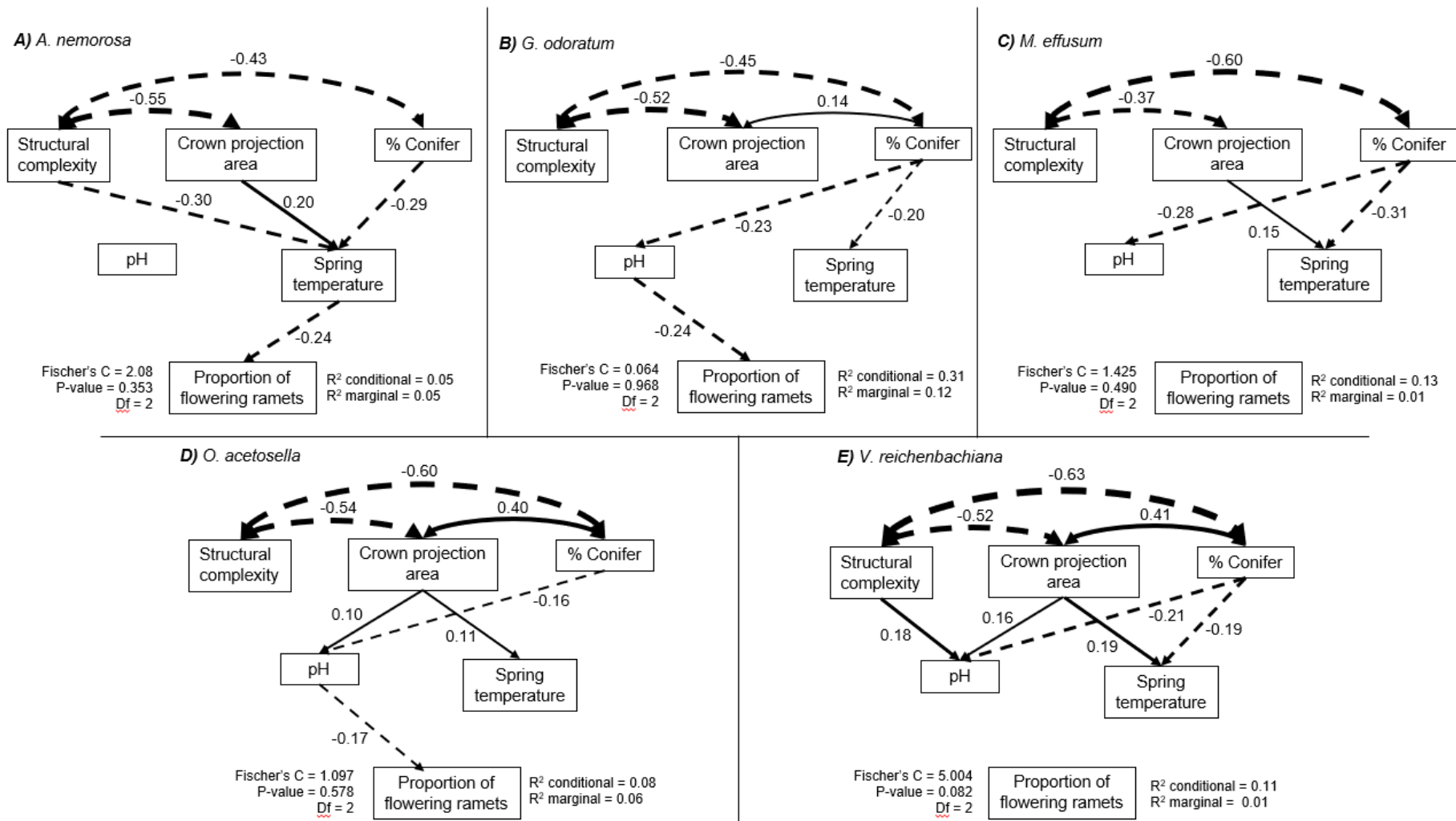
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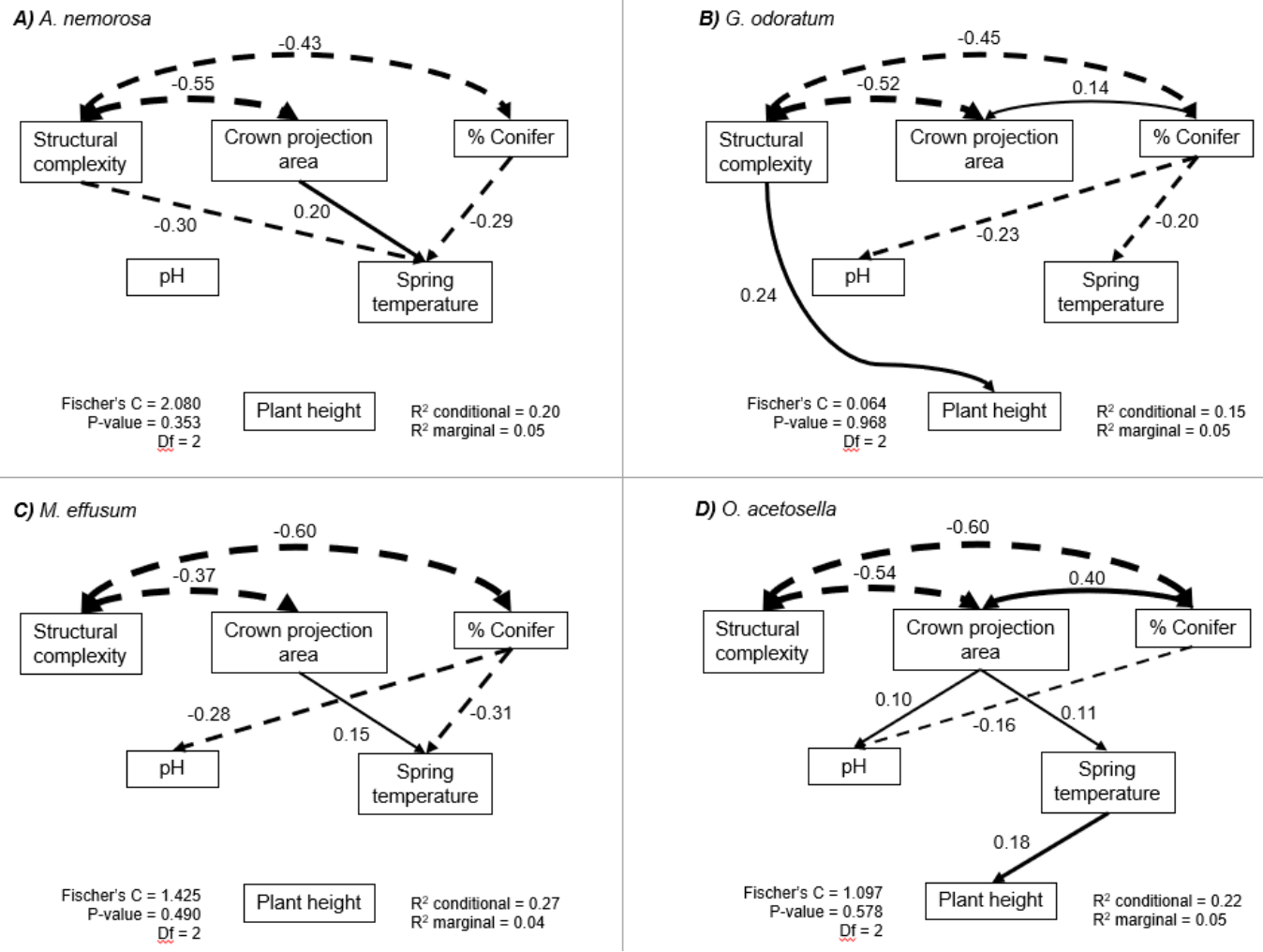
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Figure 2. Effects of forest structural attributes and microenvironment of the sites of origin on phenological traits in the common garden. Shown are the results of the piecewise structural equation model (SEM) for direct and indirect relationships among forest characteristics and microclimatic variables, and the flowering start (DOY) for (A) *Anemone nemorosa*, (B) *Galium odoratum*, (C) *Milium effusum*, (D) *Oxalis acetosella*, and (E) *Viola reichenbachiana*. Arrows represent unidirectional and bidirectional relationships among variables; only marginally or more significant paths ($P < 0.1$) are shown. Solid lines are positive relationships, dashed lines are negative ones. The line thickness is proportional to the magnitude of the standardized estimates, plotted near each line. The R^2 values are given for each SEM, along with the overall model fit to the data as Fischer's C, its associated P-value and degrees of freedom (Df).



551

552 **Figure 3. Effects of forest structural attributes and microclimate of the site of origin on flowering proportions in the common garden.** Shown are the
 553 results of the piecewise structural equation model (SEM) for direct and indirect relationships among forest characteristics and microclimatic variables, and the
 554 proportions of flowering ramets for (A) *Anemone nemorosa*, (B) *Galium odoratum*, (C) *Milium effusum*, (D) *Oxalis acetosella*, and (E) *Viola reichenbachiana*.
 555 Arrows represent unidirectional and bidirectional relationships among variables; only marginally or more significant paths ($P < 0.1$) are shown. Solid lines are
 556 positive relationships, dashed lines are negative ones. The line thickness is proportional to the magnitude of the standardized estimates, plotted near each line.
 557 The R^2 values are given for each SEM, along with the overall model fit to the data as Fischer's C, its associated P-value and degrees of freedom (Df).



558

559 **Figure 4. Effects of forest structural attributes and microclimate of the site of origin on plant height in the common garden.** Shown are the results of the
 560 piecewise structural equation model (SEM) for direct and indirect relationships among forest characteristics and microclimatic variables, and plant height for (A)
 561 *Anemone nemorosa*, (B) *Galium odoratum*, (C) *Milium effusum*, and (D) *Oxalis acetosella*. Arrows represent unidirectional and bidirectional relationships among
 562 variables; only marginally or more significant paths ($P < 0.1$) are shown. Solid lines are positive relationships, dashed lines are negative ones. The line thickness
 563 is proportional to the magnitude of the standardized estimates, plotted near each line. The R^2 values are given for each SEM, along with the overall model fit to
 564 the data as Fischer's C, its associated P-value and degrees of freedom (Df).

