

1 Among tree and habitat differences in the 2 timing and abundance of spring 3 caterpillars

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9

10 Abstract

- 11 1. Climate warming is causing many spring biological events to advance in timing and where
12 the phenology of resource and consumer advance at different rates this can result in
13 trophic asynchrony. While the temperate study system of deciduous tree – caterpillar –
14 insectivorous passerine has been widely studied, little work has examined whether
15 phenological distribution of caterpillars differ among tree taxa and habitats. If such
16 differences exist they have the potential to underpin spatial variation in the trophic
17 asynchrony in this food web.
- 18 2. Our first aim was to identify the effects of host tree versus local woodland composition
19 on caterpillar abundance. Following this, the main aim was to examine the effects of tree
20 taxon on the phenological distribution of caterpillar abundance and the trend in mass of
21 individuals, with guild biomass the product of these two metrics.

- 22 3. We collected data on caterpillar abundance and mass throughout spring from 44 sites
23 with varied woodland compositions across seven years. First, we analysed differences in
24 caterpillar abundance among tree taxa and identified any additional effect of local
25 woodland composition. Second, we explored differences in the phenological distribution
26 of caterpillars among tree taxa, focusing on caterpillar i) abundance, ii) mass and iii)
27 biomass.
- 28 4. We found substantial variation in the caterpillar abundance supported among tree taxa,
29 including evidence that the density of oak foliage within a woodland can increase the
30 abundance of caterpillars found on other trees. Some aspects of the phenological
31 distribution of caterpillars differed among tree taxa, in particular the height of the peak,
32 highest on oak. We show minimal, but significant, variation in timing and duration,
33 whereas we did not find much evidence for variation in the shape of the phenological
34 distribution or mass gain of caterpillars.
- 35 5. We show that the abundance and phenological distribution of caterpillars does differ
36 between deciduous trees and that oak is distinct from most other common taxa.
37 Woodland composition is likely to influence the site-level trend in caterpillar abundance
38 and biomass; contributing to spatial variation in an important component of the
39 woodland ecosystem and an ephemeral resource relied upon by many consumers species.

40 Introduction

41 Climate warming is driving species across multiple taxa and trophic levels to advance
42 phenological traits (Cohen, Lajeunesse, & Rohr, 2018; Parmesan, 2006; Thackeray et al.,
43 2016). The shifts in timing vary between species and trophic levels, with secondary consumers
44 generally responding at a slower rate than producers and primary consumers (Thackeray et
45 al., 2016, 2010). When a consumer is reliant upon an ephemeral resource with which it has
46 become asynchronous this can be detrimental to fitness (Cushing, 1990; Durant, Hjermmann,
47 Ottersen, & Stenseth, 2007; Visser & Both, 2005) and the consumer is described as being
48 ‘mismatched’ (Samplonius et al., 2020).

49 The temperate terrestrial tri-trophic food chain of deciduous tree – caterpillars – cavity
50 nesting insectivorous passerine is perhaps the system in which asynchrony has been studied
51 most (Both, van Asch, Bijlsma, van den Burg, & Visser, 2009; Charmantier et al., 2008; Shutt,
52 Cabello, et al., 2019; Thomas, Blondel, Perret, Lambrechts, & Speakman, 2001; Visser,
53 Holleman, & Gienapp, 2006). This system includes ephemeral resources at two trophic levels
54 making it highly susceptible to trophic mismatch; young palatable leaves which the
55 caterpillars require to grow to pupation (Feeny, 1970; Forkner, Marquis, Lill, & Corff, 2008;
56 van Asch & Visser, 2007), and then the caterpillars are a key food source for many breeding
57 birds (Bańbura, Blondel, de Wilde-Lambrechts, Galan, & Maistre, 1994; Betts, 1955; Sanz,
58 1998; Wilkin, King, & Sheldon, 2009). The temporally peaked distribution of caterpillar
59 abundance/biomass is central to this woodland system (Gibb, 1950; Southwood, Wint,
60 Kennedy, & Greenwood, 2004), with potential implications for breeding consumer species
61 (Buse, Dury, Woodburn, Perrins, & Good, 1999; Reed, Jenouvrier, & Visser, 2013; Visser et al.,

62 2006) and herbivory damage to deciduous trees, possibly influencing tree growth, survival
63 and forest productivity (Kulman, 1971; Marquis & Whelan, 1994; Whitham, Maschinski,
64 Larson, & Paige, 1991; Whittaker & Warrington, 1985). If the phenological distribution differs
65 among individual trees, species, areas within a woodland and more broadly between
66 woodlands, this has the potential to result in a stabilising spatial portfolio effect (Schindler,
67 Armstrong, & Reed, 2015) and buffering of the metapopulation from the negative effects of
68 asynchrony, as some caterpillar consumer (e.g., birds) and resource (i.e., trees) populations
69 are matched while others are mismatched. Extensions of the match-mismatch hypothesis
70 recognise that if availability of the resource is important for consumer fitness then the height
71 and width of the resource peak may matter in addition to its timing (Miller-Rushing et al.
72 2010).

73 The majority of studies within the woodland tri-trophic study system are focused on oak-
74 dominated woodlands or mainly collect data about caterpillar abundance or biomass in
75 relation to oak trees (Burgess et al., 2018; Hinks et al., 2015; Varley, Gradwell, & Hassell, 1974;
76 Visser et al., 2006). Oak (*Quercus*) trees and woodlands are thought to be preferred by some
77 breeding passerines, such as tits (Perrins, 1979), and are described as good resource for the
78 development of caterpillars, at least for the winter moth, *Operophtera brumata*, although
79 statistical comparison among tree taxa has been low powered (Feeny, 1970; Wint, 1983).
80 Oak-dominated woodland, however, is not representative of the habitat present across all
81 deciduous woodlands in these temperate environments. For instance, in Great Britain,
82 although oak trees are the second most common broadleaf tree taxon, they make up just 16%
83 of the area covered by broadleaf woodlands (Stagg & Ward, 2019), and many lepidopteran
84 and passerine species in this system will feed, forage and nest in a variety of woodland

85 habitats (Allan, 1979; Hagemeyer & Blair, 1997; Perrins, 1979; Skinner, 2009). The extent to
86 which inferences about the caterpillar phenological distribution made in relation to oak can
87 be generalised to other habitats is largely unknown (Shutt, Burgess, et al., 2019). Therefore,
88 it is important to consider how the caterpillar peak varies among woodlands with different
89 habitat compositions when aiming to understand the potential for trophic asynchrony across
90 heterogenous temperate deciduous woodlands (Burger et al., 2012).

91 There are three main components that describe the phenological distribution of the spring
92 caterpillar peak; peak date (timing), height (maximum abundance/biomass) and width
93 (duration) (Shutt, Burgess, & Phillimore, 2019). Tree composition of woodland habitats may
94 be an important determinant of spatial variation in the phenological distribution of the
95 caterpillar peak, however relatively few studies have explored this topic. Most relevant work
96 has focused on differences between deciduous, mixed and coniferous woodlands (Burger et
97 al., 2012; Mägi et al., 2009; van Balen, 1973; Veen et al., 2010), with minimal focus on the
98 differences among deciduous woodland compositions and specific tree taxa. Mägi et al.
99 (2009) suggested that the phenological distribution of caterpillars in mixed birch and alder
100 woodland in 2004 was more similar to that of coniferous woodlands than oak; characterised
101 as having a low and wide peak, compared to the high but narrow peak in oak woodlands (van
102 Balen, 1973; Veen et al., 2010; Visser et al., 2006). Birch and coniferous mixed woodlands
103 were also found to host lower abundances of caterpillars (Burger et al., 2012). However, in a
104 spatially replicated study of caterpillars on individual trees in Scotland, birch trees were found
105 to host more caterpillars than many other deciduous species (Shutt, Burgess, & Phillimore,
106 2019). The same study also found that the height of the caterpillar peak differs among four
107 deciduous tree taxa, with the peak on oak trees reaching much greater maximum abundances

108 and some evidence of increased peak heights on willow trees as well (Shutt, Burgess, et al.,
109 2019). Oak may not support the greatest abundance of all deciduous trees however; as in
110 Białowieża Forest, Poland hornbeams, maples and limes were found to receive more
111 herbivory damage than oak (Wesołowski & Rowiński, 2006).

112 There are a number of ways that woodland tree composition may influence the height, width
113 and timing of the peak in caterpillar abundance and biomass. Tree species differ in leaf
114 nutritional quality (Schultz, Nothnagle, & Baldwin, 1982; Yanar, Gömeç, Topkara, Solmaz, &
115 Demir, 2017), which can affect Lepidopteran fecundity (Awmack & Leather, 2002), survival
116 (Wint, 1983; Yanar et al., 2017) and growth (Loader & Damman, 1991). Preferential laying by
117 females on specific hosts (Kakimoto, Fujisaki, & Miyatake, 2003; Thompson & Pellmyr, 1991),
118 or dispersal by larvae (Forkner et al., 2008; van Asch & Visser, 2007) may also amplify
119 differences in abundance among tree taxa, impacting biomass as the product of the
120 abundance and size of caterpillars in the guild. Studies of spring caterpillars report differences
121 in abundance and biomass between deciduous tree species, with oak and willow having the
122 highest peaks (Shutt, Burgess, et al., 2019; Veen et al., 2010), however both studies had quite
123 low power to detect among species differences and the sensitivity of biomass to changes in
124 mass versus abundance is yet to be explored.

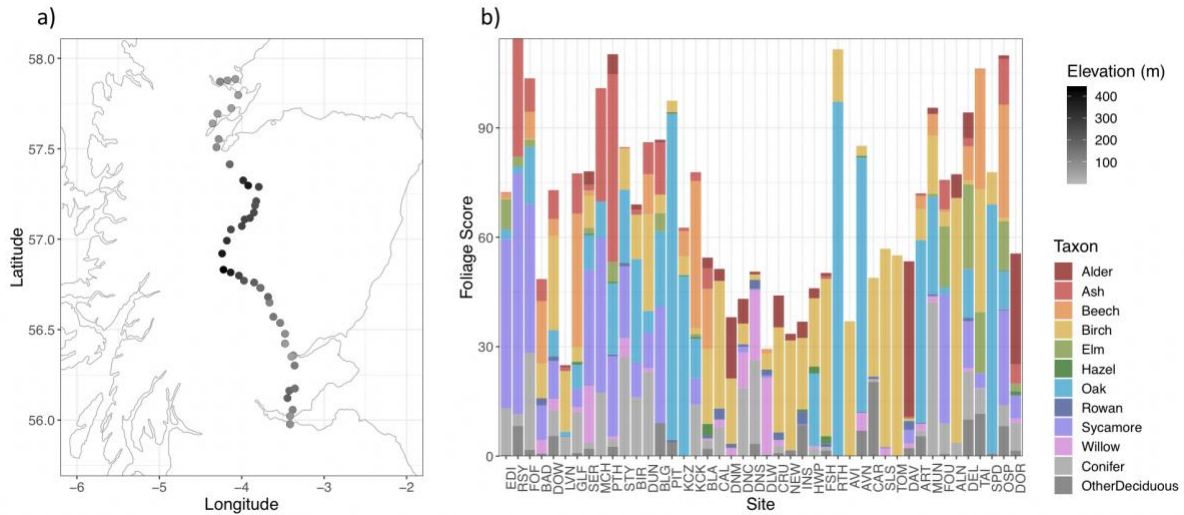
125 The width or duration of the peak may also vary through multiple mechanisms. For example,
126 the diversity of the lepidopteran guild differs between tree species (Kennedy & Southwood,
127 1984), which may impact the peak width if lepidopteran species differ in their hatching
128 phenology or growth rate. In addition, differences in nutritional quality or the phenology of
129 the nutrition provided by different tree taxa (Kakimoto et al., 2003; Loader & Damman, 1991;
130 Schultz et al., 1982) could generate among tree variation in the time caterpillars take to reach

131 pupation. Finally the timing of peak abundance or biomass may vary among tree species as a
132 result of differences in their leafing phenology (Murray, Cannell, & Smith, 1989; Roberts,
133 Tansey, Smithers, & Phillimore, 2015). Such phenological differences among trees could select
134 for adaptation of a caterpillar species' phenology or sort for caterpillar species that differ in
135 their timing. However, the large variation in tree phenology within tree species (Cole &
136 Sheldon, 2017; Murray et al., 1989), and even within individual trees, will reduce the strength
137 of selection to coincide with the phenology of a particular tree species. Shutt, Burgess, et al.,
138 (2019) explored among tree species variation in peak timing in Scotland using three years of
139 data with 6089 records of caterpillar presence/absence (575 caterpillars). They found no
140 difference in peak timing between oak, willow, birch and sycamore, though credible intervals
141 were broad, implying low power to detect a difference of a few days. Now with 7 years of
142 data and 31215 samples we are able to analyse this topic using the abundance (rather than
143 presence/absence) of caterpillars, providing more power to detect any differences.

144 In this study we use caterpillar data collected from 44 deciduous Scottish woodland sites that
145 vary in their tree composition, to analyse tree and habitat-driven variations in the
146 phenological distribution of caterpillar abundance and biomass throughout spring. First, we
147 explore differences in the abundance of caterpillars supported by different tree taxa and
148 identify any additional effect of differences in the local woodland density and composition.
149 Second, we examine whether the phenological distribution of the caterpillar guild throughout
150 spring differs among tree taxa for i) abundance and ii) mass, and how variation in each
151 contributes to variation in their product iii) biomass. Throughout we compare each taxon to
152 the average trend and to oak. Together this programme examines the capacity of woodland
153 composition to provide a portfolio effect of herbivory pressure to trees, mismatch buffering

154 for secondary consumers and tests the extent to which oak-dominated study systems are
155 representative of other deciduous woodlands.

156



157

158 Figure 1: a) Map of site locations in Scotland, the darkness of the points indicates the site
159 elevation and b) shows the woodland habitat composition at each site displayed as a foliage
160 score that accounts for tree size, 1 unit of foliage score = 1 small tree (circumference at chest
161 height 0.4-0.99m), sites are ordered by increasing latitude from left to right.

162

163 Methods

164 Study System

165 This study was carried out using 44 woodland sites along a 220 km transect between
166 Edinburgh (55°980 N, 3°400 W) and Dornoch (57°890 N, 4°080 W) (Fig. 1a). This includes the
167 40 sites monitored between 2014-16 in Shutt, Bolton, Cabello, Burgess, & Phillimore (2018)
168 with an additional four sites monitored from 2017 onwards. During spring 2020 the field
169 season was restricted to the 22 most southern sites due to the Covid-19 pandemic. Woodland

170 habitat composition has been surveyed at a 15m radius around each nest box installed at sites
171 (n = 334 since 2017), including all trees with a trunk circumference of ≥ 40 cm at chest height
172 or a 'stand' with ≥ 6 branches within 20cm of each other at the base. Trees were identified to
173 the genus level with the exception of some conifers, for full details of survey methods and
174 site foliage score calculations see Shutt et al. (2018). The foliage scores for each tree taxon
175 at each site are representative of the mean number of 'small' trees of each taxon within a
176 circular area with a 15m radius (Fig. 1b).

177 **Caterpillar Sampling**

178 Caterpillar sampling followed the branch beating methodology described in Shutt, Burgess, &
179 Phillimore (2019). A selection of trees at each site, representative of the site's woodland
180 composition, are monitored for leafing phenology (Shutt, Cabello, et al., 2019); and those that
181 had a branch of minimum length 1m and between 0.5-1.5m above the ground were also
182 beaten. The tree taxa sampled include alder (*Alnus* sp.), ash (*Fraxinus* sp.), beech (*Fagus* sp.),
183 birch (*Betula* sp.), elm (*Ulmus* sp.), hazel (*Corylus* sp.), oak (*Quercus* sp.), rowan (*Sorbus* sp.),
184 sycamore (*Acer* sp.) and willow (*Salix* sp.). A branch, meeting the height and length criteria,
185 was marked and beaten every four days to allow some recolonization between beatings. The
186 trees sampled at each site were divided in two groups and each group was beaten alternately
187 at two days intervals. The branch was held consistent between years unless damaged, broken
188 or dead. The average number of trees sampled at each site in each year was 3, 6, 6, 14, 14,
189 15 and 14 respectively; with a total of 31215 beating samples recorded across the 7 years.
190 Sampling began each year when 45% of the monitored trees across all sites had reached the
191 'first leaf' stage and continued until the end of the blue tit breeding season (Shutt, Burgess,

192 et al., 2019); sampling periods extended between ordinal dates 120-166, 125-175, 130-173,
193 123-167, 129-172, 117-168 and 130-170 for 2014-20 respectively.

194 Branch beating was carried out using a clear plastic rubble sack measuring 76cm x 51cm. The
195 bag was fully extended over the branch's foliage and closed with one hand at the maximum
196 length, keeping the open end facing upwards. The bag was then beaten 30 times at a
197 consistent impact and rate (roughly two per second) to dislodge any free-living invertebrates
198 on the branch. After careful removal from the branch, all caterpillars (larvae of Lepidoptera
199 (>90%), Hymenoptera, Diptera and Coleoptera; Shutt, Burgess, et al., 2019) with an estimated
200 diameter of $\geq 1\text{mm}$ were counted and collected. We chose this threshold due to an
201 assumption that the smallest caterpillars would be of minimal value to avian consumers, and
202 as they are hard to see and collect, excluding them increases the consistency between
203 samples and recorders. From 2017 onwards, the biomass of each sample was recorded to
204 0.01g using a Myweigh Triton T3R-500 Digital Scale. We found that some samples were of
205 insufficient mass ($< 0.02\text{g}$) to show a read on the balance.

206 **Statistical analysis**

207 All analyses used Bayesian generalized linear mixed models (GLMM) in the MCMCglmm
208 package (Hadfield, 2010) in R version 3.5.3 (R Core Team, 2019) and we present full models
209 (Tables S1-5). Poisson GLMMs were used for all models looking at variations in caterpillar
210 abundance and 1000 *a posteriori* simulations under each model were used to confirm that
211 the data were not zero-inflated as compared with model expectation. Gaussian GLMMs were
212 used to model caterpillar mass. Fixed effects parameters and random effects are consistent
213 with a significant effect where the 95% credible intervals (CIs) do not overlap zero. As

214 variances are bounded at zero we infer a random term to be significant where the lower
215 credible interval is removed from 0. We used the mean of the posterior distribution when
216 calculating parameter estimates based on the fixed effects, and the median when random
217 effects were included. All models were run with sufficient iterations to ensure an effective
218 sample size of >1000 for each focal parameter and convergence was assessed by visual
219 inspection of the trace plots. Parameter-expanded priors were used for all models, with the
220 residual variance treated as inverse-wishart with $\nu = 0.002$. Default flat priors were used for
221 the fixed effects.

222 Some structural random terms were included in all models (Table 1). These were site, year
223 (as a factor), each site in each year (site-year), and day at each site in each year (day-site-year)
224 to allow for spatio-temporal differences in caterpillar abundance and mass. We also included
225 the tree ID and recorder of each beating sample. All results comparing caterpillar abundance,
226 mass or biomass among tree taxa are presented as deviations from (i) the fixed effect
227 prediction, which represents an average tree taxon, and (ii) the prediction for oak, described
228 fully in Supplementary Information.

229

230 Table 1: Model compositions. Date refers to ordinal date which was scaled (mean = 146.77,
 231 SD =14.04), year was included as a factor and woodland composition refers to multi-
 232 membership component explained in the text. Int = intercept, abund = abundance, VCV =
 233 variances and covariances.

Model	Abundance decomposition (Table S1)	Habitat abundance (Table S2)	Abundance phenology model (Table S3)	Mass phenology model (Table S4)
Motivation	Decomposition of variance in abundance	Differences in the abundance of caterpillars among tree taxa and effect of woodland density and composition	Differences in phenological distribution of caterpillar abundance among tree taxa	Differences in phenological distribution of caterpillar mass among tree taxa
Response	Caterpillar abundance	Caterpillar abundance	Caterpillar abundance	log(Mean mass per caterpillar)
Fixed effects	Date Date ²	Total foliage score	Date Date ²	Date Date ²
Random terms	Site Tree taxa Tree ID Site-year Day-site-year Year Year-day Recorder	Tree taxa Woodland-composition Site Year Site-year Day-site-year Tree ID Recorder	Int, Date and Date ² VCV across Tree taxa Int, Date and Date ² VCV across Site Int, Date and Date ² VCV across Year Site-year Day-site-year Tree ID Recorder	Int and Date VCV across Tree taxa Int and Date VCV across Site Year Site-year Sqrt(1/abund) VCV across units Day-site-year TreeID Recorder
Family	Poisson	Poisson	Poisson	Censored-Gaussian
Iterations (thin)	2500000 (500)	4500000 (800)	1050000 (100)	5500000 (500)
Burnin	50000	500000	50000	500000
Sample Size	4900	5000	10000	10000

234

235 **Abundance**

236 *Variance decomposition of caterpillar abundance*

237 We decomposed the variance in caterpillar abundance throughout spring to gauge to the
 238 relative contributions of spatial and temporal factors to the total variance observed in our
 239 data (abundance decomposition model, Table 1). The number of caterpillars recorded in each
 240 beating sample was the response variable with date and date² included in the fixed effects to
 241 account for the peaked temporal distribution. Date refers to ordinal date and was scaled in

242 all models (prior to scaling: mean = 146.77, SD = 14.04). The variance explained by date (x)
243 and date² (y) was calculated using the quadratic equation:

244 Equation 1:
$$[\beta_x \quad \beta_y] * \begin{bmatrix} \sigma_x^2 & \sigma_{x,y} \\ \sigma_{x,y} & \sigma_y^2 \end{bmatrix} * [\beta_x \quad \beta_y]^T$$

245 The random terms included each day in each year (day-year) and the tree taxon sampled (tree
246 taxa) in addition to the consistent terms listed above. We calculated the mean percentage of
247 variance on the log scale that is attributable to the date and date² fixed effects and
248 each random term across the posterior distribution.

249 *Habitat variation in caterpillar abundance*

250 We assessed differences in the abundance of caterpillars supported by different tree taxa,
251 the influence of local foliage density and whether the amount of foliage of each tree taxon
252 contributing to local woodland composition, from the habitat surveys, has any additional
253 effect (habitat abundance model, Table 1). To calculate the site-level foliage scores for non-
254 focal tree taxa, we combined all of the coniferous tree taxa into a 'conifer' group and all
255 deciduous tree taxa, that are less common and not sampled in the branch beating, were
256 grouped as 'other deciduous'. Composition foliage scores for each tree taxon at each site
257 were globally mean centred. The 'total' foliage score for each site was then calculated by
258 summing the mean centred foliage scores for each taxon across each site, meaning that the
259 total foliage scores among sites were distributed around a mean of zero. The site level total
260 foliage score was included as a fixed effect to account for any effect of general site foliage
261 density on abundance. We included tree taxon and the woodland composition foliage scores
262 as focal random terms. We assessed the difference in abundance supported by different tree
263 taxa using the posterior distributions of the tree taxon (sampled) random effects. The

264 woodland composition component used a multi-membership approach, whereby the random
265 effect for each tree taxon corresponded to the deviations from the fixed effect slope for total
266 foliage score. The taxon-specific composition slopes intersect at the average amount of any
267 tree taxon present at any site, and random effects were drawn from a distribution with a
268 single variance estimated by the model. This allowed us to test for an effect of the amount of
269 foliage of each tree taxon at a site, regardless of which tree taxon the sample was collected
270 from.

271

272 **Phenological Distributions**

273 *Abundance*

274 To quantify differences in the phenological distribution of caterpillar abundance throughout
275 spring on different tree taxa we allowed each taxon to have a distinct temporal trend over
276 the course of the spring (abundance phenology model, Table 1). The fixed effects included
277 date and date², allowing for a peaked trend over time. The focal random terms were the tree
278 taxa random intercepts and slopes for date and date². In addition to the core random terms,
279 we also included date and date² random slopes for site and year to account for spatio-
280 temporal differences in the peak timing and shape. As adjustments to date and date²
281 coefficients both influence the resulting height and timing of the peak, interpretation of the
282 effects on the phenology of abundance is best understood through calculations of the model
283 predictions for the specific metrics of interest. We used the posterior distributions for the
284 fixed effects and the random effects (intercept, date and date²) for each tree taxon to quantify
285 the timing (peak date) of the maximum peak in abundance, the maximum abundance reached
286 (height), and the width of the peak relative to height (calculated at half the height of the

287 peak)- referred to as peak width- as well as the relative width of the peak among tree taxa,
288 calculated at a consistent height (set at roughly half the height of the lowest curve)- referred
289 to as peak duration. We primarily present peak width as an independent parameter governing
290 the peak shape, but also discuss peak duration (which includes an effect of peak height) due
291 to its relevance to phenological synchrony of consumers.

292 We assessed the adequacy of the quadratic function in describing the shape of the
293 phenological distribution of caterpillar abundance by also modelling the peak including a
294 cubic date term; allowing an asymmetrical trend over time (Table S5). Details of the analysis
295 can be found in Supplementary Information.

296 *Mass*

297 To obtain the mean mass per caterpillar in a sample we divided the mass for each sample by
298 the number of caterpillars weighed. We included a weighting variable ($\sqrt{1/\text{abundance}}$) to
299 account for the number of caterpillars contributing towards each mass measurement. Due to
300 restrictions in measuring small masses in the field, we have uncertainty in the accuracy of
301 measurements $\leq 0.02\text{g}$. All samples meeting this criterion were interval censored to a lower
302 bound of 0.001g, the minimum mass viable for a caterpillar sampled in-line with the
303 methodology (J. C. Weir, unpublished data), and an upper bound of 0.02g divided by the
304 number of caterpillars. The mass measurements were log transformed and modelled as
305 interval-censored Gaussian.

306 The mass model included date and date² in the fixed effects, allowing a curved trend if
307 appropriate (mass phenology model, Table 1). The focal random terms were the tree taxa
308 random intercepts and date slope. In addition to the core random terms, we included a date

309 random slope for sites, allowing for spatial variation. Year was solely included as a random
310 intercept to improve model convergence as there were just 4 years of mass data. To assess
311 differences in the ultimate mass of a caterpillar on different tree taxa during our sampling
312 period, the posterior distributions of the fixed effects and tree taxa random intercepts and
313 slopes were used to calculate the predicted mass at day 168 (16th June), the latest date at
314 which a caterpillar has been sampled from all tree taxa.

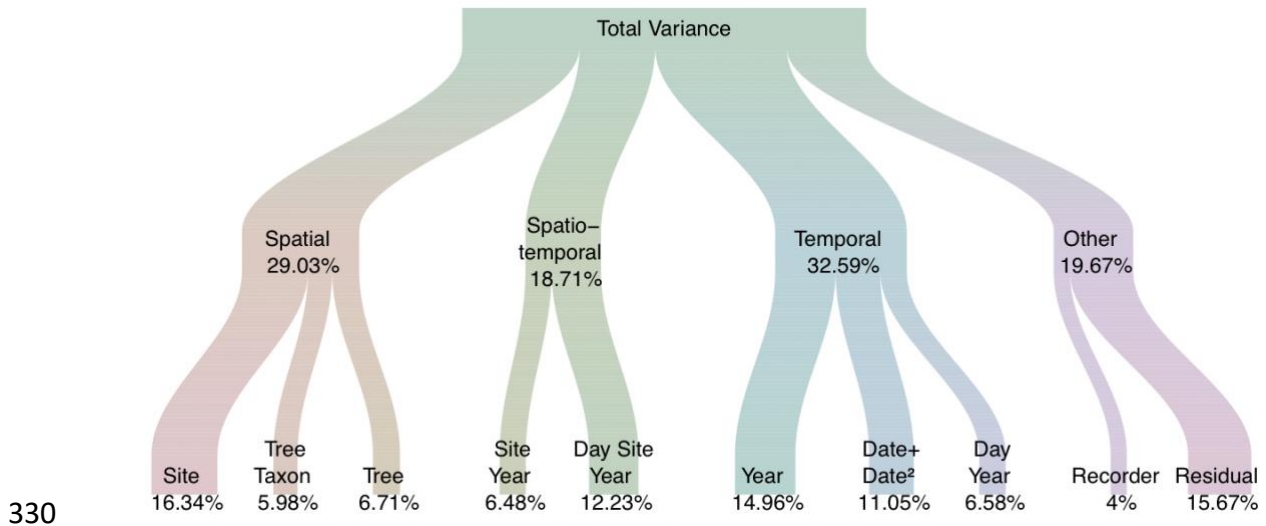
315 *Biomass*

316 We combined the model outputs for the phenological distribution of caterpillar abundance
317 and mass to calculate the distribution of biomass. As abundance and mass were both
318 estimated on the log scale, equations describing the change in caterpillar abundance and
319 mass over time, for each tree taxon and the fixed effect mean trend, can be summed across
320 each iteration of the posterior distribution to convert to a temporal prediction of biomass.
321 Ideally this would be analysed using a bivariate model to allow for covariance between
322 abundance and mass across random terms, however due to the interval censoring required
323 for the mass data this could not be implemented. The coefficients for biomass have been used
324 to calculate the same peak metrics as discussed for the abundance results.

325 Results

326 Within our dataset, 11% of branch beating yielded at least one caterpillar. Where a
327 caterpillar was present, in 68% of cases there was just 1 and in 16% there were 2, with a
328 maximum abundance of 107, recorded during an outbreak at one of our sites during 2019.

329



331 Figure 2: Riverplot depicting the percentage variance composition of caterpillar abundance
 332 estimated from terms in a Poisson GLMM (Table S1). All variables were included as random
 333 terms except for Date+Date² which were continuous variables included as fixed effects
 334 accounting for the peaked shape of the phenological distribution.

335

336 **Abundance**

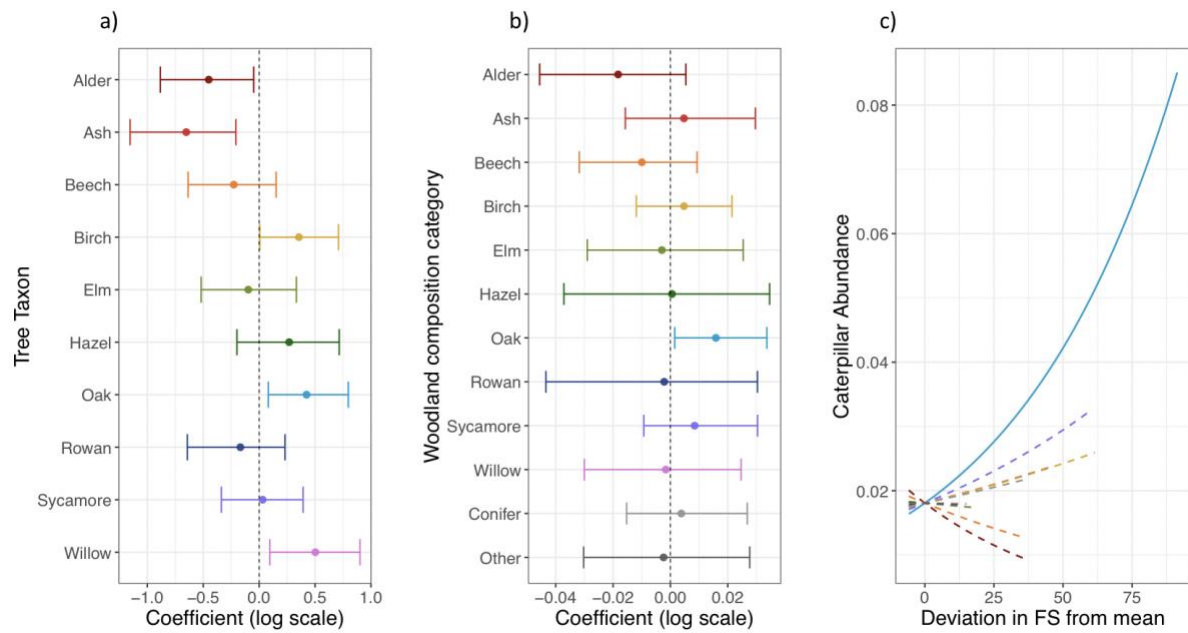
337 *Caterpillar abundance variance decomposition*

338 The posterior distributions of variance (on log scale) for all terms included in the abundance
 339 decomposition model (Table 1, S1), including the calculated variance for the fixed effects,
 340 were removed from zero, indicating that each variable contributes to the variance in
 341 caterpillar abundance observed. The environmental components (all excluding ‘Other’)
 342 explained 80.33% (CIs: 73.76 - 86.78%) of the variance, with temporal and spatial components
 343 explaining similar amounts of 32.59% (CIs: 20.21 - 49.42%) and 29.03% (CIs: 18.02 - 41.10%)
 344 respectively (Fig. 2). The tree taxon sampled explained 5.98% (CIs: 1.24 - 14.08%) of the
 345 variance (Fig. 2).

346 *Habitat variations in caterpillar abundance*

347 We estimated substantial variance in the abundance of caterpillars sampled among tree taxa
348 in the habitat abundance model (Table S2). Alder and ash support significantly fewer
349 caterpillars than average and birch, oak and willow support more (Fig. 3a). Alder hosts 0.64
350 (CIs: 0.41 - 0.95) times as many caterpillars as an average taxon, ash hosts 0.52 (CIs: 0.32 -
351 0.81), birch hosts 1.43 (CIs: 1.01 - 2.03), oak hosts 1.53 (CIs: 1.09 - 2.22) and willow hosts 1.65
352 (CIs: 1.10 - 2.46). We found that alder, ash, beech, elm, rowan and sycamore all support
353 significantly fewer caterpillars than oak (Fig. S1).

354 We found no relationship between the total habitat foliage score and the number of
355 caterpillars recorded and the posterior for the variance in the slope predictions for the
356 woodland composition- tree taxon foliage scores- was not removed from zero (Table S2).
357 However, when we looked at the foliage score taxon random effects we found that an
358 increase in the amount of oak at a site was predicted to significantly increase the abundance
359 of caterpillars found on the branches of any tree at a site (Fig. 3b, Fig. 3c). Exponentiating the
360 coefficient shows that a branch will have 1.02 (CIs: 1.001 - 1.03) times as many caterpillars on
361 it for every additional unit of oak foliage (equivalent to one small tree). Within the woodland
362 compositions across our sites, branches of any tree taxon within mature oak-dominated
363 woodlands can reach 5.19 (CIs: 1.73 – 17.45) times the caterpillar abundance of a woodland
364 without oak, purely attributed to the local habitat composition (Fig. 3c).



365

366 Figure 3: a-b) Posterior median and 95% credible intervals for random effects in a Poisson
 367 GLMM (Table S2). Predicted coefficients for a) the log difference in abundance of caterpillars
 368 sampled from each tree taxon and b) the log change in caterpillar abundance on an average
 369 branch with an increase in the amount of foliage of the tree taxon within the local woodland
 370 composition (globally mean centred foliage scores). c) Slope predictions from a Poisson
 371 GLMM (Table S2) analyzing the change in caterpillar abundance with change in the amount
 372 of foliage of each tree taxon present at each site (globally mean centered). The intercept falls
 373 at the mean foliage score (FS) of each taxon at any site and mean total FS across all sites.
 374 Dashed lines indicate an insignificant difference in slope from zero, the solid line indicates a
 375 significant result. All lines represent the range of foliage scores of each taxon present across
 376 all sites.

377

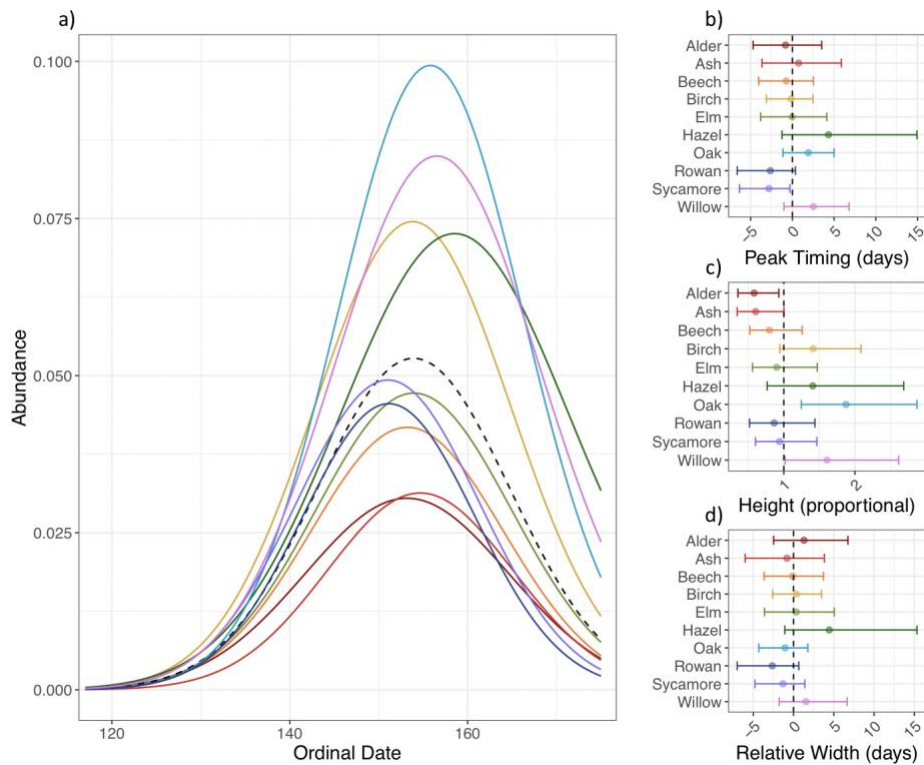
378 Phenological distributions

379 Abundance

380 Annual peaks in the temporal distribution of caterpillar abundance in our data are clear (Fig.
 381 S2), and supported by the significant negative quadratic term from the abundance phenology
 382 model (-0.84, CIs: -1.18 - -0.52, Table S3). We found substantial variance in the tree taxon
 383 intercepts and date slopes effects of the abundance phenology model (Table 1), but not the

384 date² effects (Table S3). The fixed effects indicated the phenological distribution of
385 abundance on an average tree taxon, in an average site and year, and predicted a mean peak
386 date of 154.03 (CIs: 146.96 - 161.31), 3rd June, peak height of 0.06 (CIs: 0.02 - 0.14) caterpillars
387 per branch, peak width of 25.91 days (CIs: 21.14 – 31.47 days) and duration of 40.43 days (CIs:
388 26.08 – 54.79 days), calculated at a threshold of 0.01 caterpillars per branch. In calculating
389 peak width and duration a small proportion (< 1%) of the posterior samples yielded NAs due
390 to some iterations either not predicting a negative quadratic term or predicting a peak height
391 beneath the threshold for peak duration. These iterations were excluded which will make the
392 median and CIs underestimates. The low peak height value was likely due to the large
393 variance among sites and years and 89% of the beating samples recording zero caterpillars.

394 Investigating tree taxon specific peaks showed differences in the phenological distributions
395 of caterpillar abundance (Fig. 4a). The peak timing on sycamore was significantly earlier than
396 for the average taxon (-2.81 days, CIs: -6.33 - -0.29, Fig. 4b). The timing of the caterpillar
397 abundance peak on hazel, oak, rowan and willow trees also showed trends towards deviating
398 from the average taxon by 4.33 (CIs: -1.24 - 14.93), 1.92 (CIs: -1.13 - 5.00), -2.64 (CIs: -6.61 -
399 0.38) and 2.53 (CIs: -1.01 - 6.79) days respectively; however the CIs were not removed from
400 zero (Fig. 4b). The height of the peak in caterpillar abundance was significantly different to an
401 average taxon for alder, oak and willow trees, reaching 0.58 (CIs: 0.35 - 0.93), 1.87 (CIs: 1.25
402 - 2.87) and 1.61 (CIs: 1.02 - 2.61) times the peak height, respectively (Fig. 4c). As a proportional
403 change, CI removal from one was the threshold for significance. The peak on ash and birch
404 also showed trends towards deviating from the average taxon prediction, reaching 0.60 (CIs:
405 0.34 - 1.002) and 1.41 (CIs: 0.94 - 2.09) times the height respectively; however the CIs included
406 one (Fig. 4c).



407

408 Figure 4: Model predictions for the phenological distribution of caterpillar abundance
 409 throughout spring on branches of different tree taxa (Table S3). a) shows the model
 410 predictions for the change in caterpillar abundance over time (ordinal date) for each tree
 411 taxon, the black dashed line depicts the curve prediction from the fixed effects. Plots b-d)
 412 show the difference between each tree taxon and the fixed effect prediction (indicated by a
 413 black dashed line at zero or one) for the b) timing of the peak in abundance, c) proportional
 414 height of the peak (maximum abundance) and d) width of the peak at half of the height. The
 415 median and 95% credible were calculated using the posterior distributions for the fixed
 416 effects and tree taxa random effects and interactions.

417

418 The peak width, representing the relative shape of the peak, did not significantly differ
 419 between any taxon and the trend on an average tree (Fig. 4d). This suggests that the shape
 420 of the peak in caterpillar abundance was relatively consistent among tree taxa. However, the
 421 peak on oak, rowan and sycamore showed non-significant trends towards a narrower, more
 422 strongly peaked, distribution with a width difference of -1.04 (CIs: -4.31 - 1.77), -2.64 (CIs: -
 423 7.00 - 0.65) and -1.32 (CIs: -4.80 - 1.40) days relative to the trend on an average taxon and the

424 peak on hazel and willow showed trends towards a broader peak by 4.44 (CIs: -1.08 - 15.33)

425 and 1.56 (CIs: -1.78 - 6.65) days respectively (Fig. 4d). Due to relatively consistent shape but

426 variable peak height, the peak duration showed more variation among taxa with the peak on

427 willow lasting for a significantly longer duration, by 8.17 days (CIs: 0.33 - 19.01 days), than the

428 average trend (Fig. S3). The abundance peak on ash, birch, hazel and rowan showed non-

429 significant trends towards differing in duration to an average taxon, differing by -7.79 (CIs: -

430 19.12 - 1.63), 4.57 (CIs: -2.16 - 12.07), 11.61 (CIs: -1.42 - 34.41) and -5.82 (CIs: -14.99 - 1.75)

431 days respectively (Fig. S3). Whilst the peak on oak was suggested to have a narrower shape,

432 due to the increased relative peak height the duration was suggested to be longer (4.83 days,

433 CIs: -3.09 – 12.34), albeit non-significantly. The peak on oak was found to be significantly later

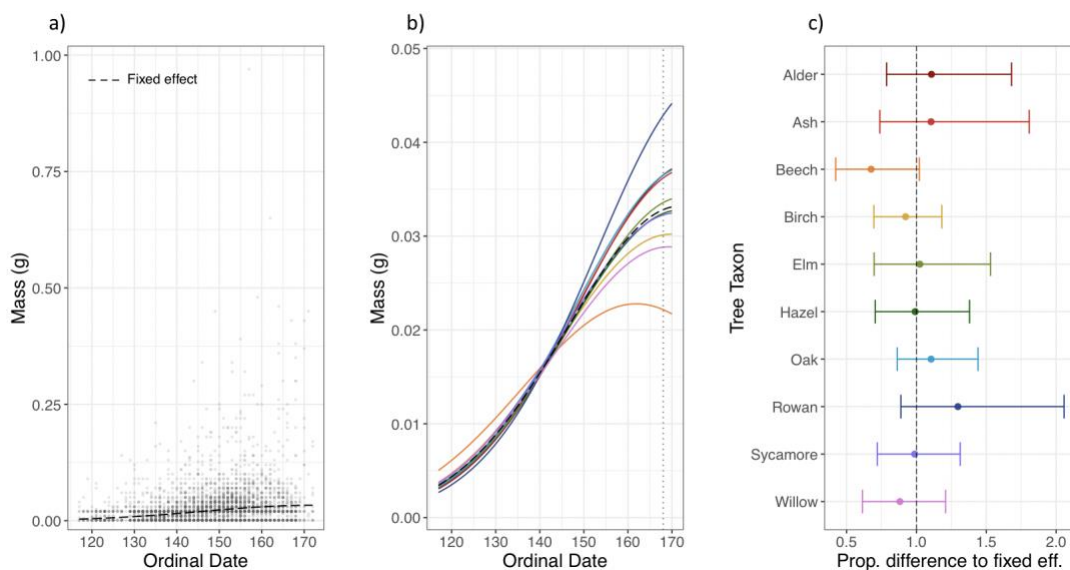
434 than rowan and sycamore, higher than all tree taxa tested except hazel and willow and lasting

435 for a longer duration than peaks on alder, ash, beech, rowan and sycamore (Fig. S4).

436 In the peak asymmetry model (Table S5), we found that although the cubic term was

437 significant (Table S6), the resulting asymmetry in the peak estimate was low (Fig. S8) and

438 therefore the use of a quadratic function adequately described the shape of the peak.



439

440 Figure 5: Slope predictions from a model analyzing the change in caterpillar mass throughout
441 spring (Table S4). a) The black dashed line indicates the change in mass over time from the
442 fixed effects. The points show the mass data, with transparency indicating the frequency of
443 the point. The two bolder lines of points highlight the two levels of interval censored samples
444 making up 4.9% of all samples. b) shows the model predictions for the change in mass over
445 time for caterpillars on each tree taxon with the black dashed line showing the fixed effect
446 trend from plot a). The grey dotted vertical line indicates the day 168 (16th June), the latest
447 date with caterpillar mass data for all tree taxa. c) shows the posterior median and 95%
448 credible intervals for the proportional difference in day 168 mass between each tree taxon
449 and the fixed effect prediction.

450 *Mass*

451 Most caterpillars throughout spring weighed less than 0.1g, although caterpillars of up to
452 0.96g (n=1) have been recorded (Fig. 5a). There was a significant and pronounced trend for
453 mass to increase over the course of the spring and for the rate of increase to decline (Fig. 5b,
454 Table S4). There was no significant variance among tree taxa in the intercept or date slope,
455 however there was slightly more variance estimated in the slope estimates than intercept
456 which is why the predicted trends intersect at a central date (Fig. 5b, Table S4). The day 168
457 mass of caterpillars (in mid-June) predicted from the fixed effects, representing an average
458 tree taxon, was 0.033g (CIs: 0.023 - 0.047g). Whilst the curves and day 168 masses do not
459 differ significantly between any tree taxon and the fixed effect trend (Fig. 5b-c), caterpillars
460 sampled from beech showed a strong trend towards a lower mass, reaching 0.68 (CIs: 0.42 -
461 1.02) of the mass of caterpillars from the average trend. Caterpillars from oak and rowan
462 showed non-significant trends towards a greater mass, reaching 1.11 (CIs: 0.86 – 1.44) and
463 1.30 (CIs: 0.89 - 2.06) times that of the average, respectively (Fig. 5c). In comparison to oak,
464 we found the ultimate mass of caterpillars was significantly less for those sampled from
465 beech, with negative trends also suggested for birch and willow (Fig. S5).

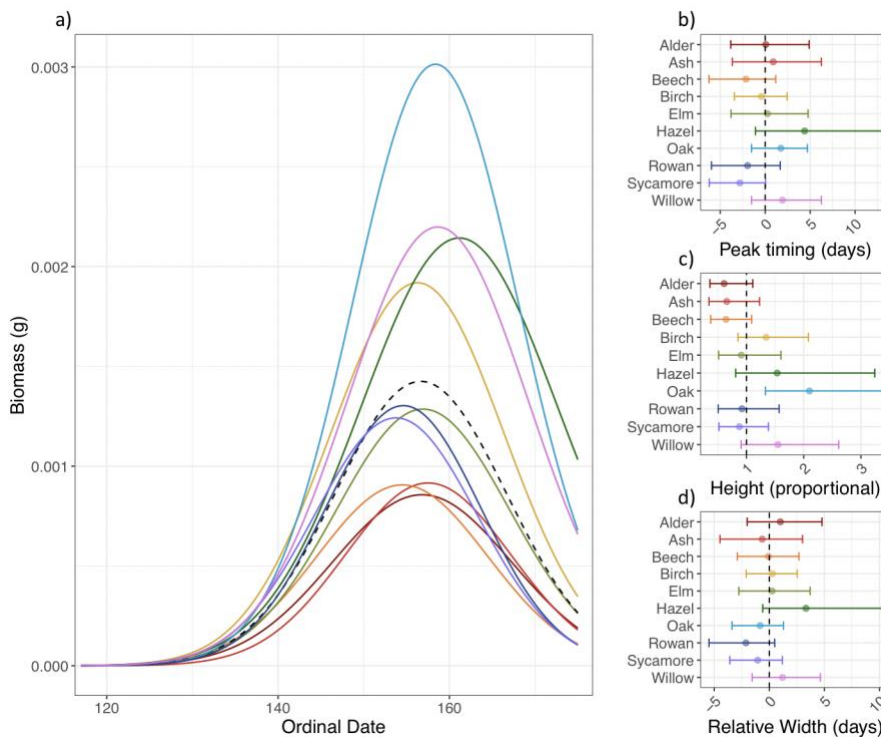
466 *Biomass*

467 The combined fixed effects from the abundance and mass phenology models (Tables 1, S3,
468 S4) predicted a peak date for the distribution of biomass on an average taxon of 156.84 (CIs:
469 150.94 - 163.39), 5th June; two days later than the peak in abundance. The maximal biomass
470 peak height was predicted to be 0.0015g (CIs: 0.0006 - 0.0037g) per branch, a peak width of
471 23.80 days (CIs: 20.00 - 28.13), suggesting a more strongly peaked shape than the distribution
472 of abundance, and a duration of 30.90 days (CIs: 16.68 - 45.45) at a threshold of 0.35mg of
473 caterpillars per branch. Similarly to the abundance peaks, some iterations (<2% in general,
474 but 5-7% for alder, ash and beech peak duration) produced NAs in the calculations of peak
475 width and duration so the median and CIs are underestimates.

476 The most noticeable differences between the phenological distributions of caterpillar
477 abundance and biomass were the increased relative height of the caterpillar peak on oak
478 compared to other taxa and decreased relative height on beech, becoming more similar to
479 the peaks on alder and ash (Fig. 4a, 6a). Unlike in abundance, the biomass peak on sycamore
480 was not significantly earlier than an average taxon, however the CIs fell very close to zero,
481 predicting a peak -2.82 (CIs: -6.22 - 0.04) days earlier (Fig. 6b). Similarly to the abundance
482 peaks; hazel, oak, rowan and willow showed trends towards deviating in timing by 4.39 (CIs:
483 -1.08 - 13.01), 1.74 (CIs: -1.52 - 4.69), -1.96 (CIs: -5.97 - 1.68) and 1.93 (CIs: -1.52 - 6.25) days,
484 respectively, with the addition of beech also now showing trends towards peaking earlier by
485 -2.15 (CIs: -6.25 - 1.17) days (Fig 6b).

486 The height of the peak in caterpillar biomass remained significantly different to an average
487 taxon on oak trees, reaching 2.10 (CIs: 1.33 - 3.37) times the maximum mass (Fig. 6c) and
488 whilst the biomass peak height on alder remained lower than the average trend (0.61, CIs:

489 0.36 – 1.11), the CIs include one. Due to the reduced mass of caterpillars on beech trees, the
 490 biomass peak height was predicted to be 0.64 (CIs: 0.38 - 1.09) times the height predicted for
 491 an average taxon, albeit non-significantly (Fig. 6c). The biomass peak on birch and willow
 492 branches continued to tend towards deviating non-significantly from an average taxon,
 493 reaching 1.34 (CIs: 0.85 - 2.08), and 1.55 (CIs: 0.90 - 2.61) times the height respectively, with
 494 the hazel biomass peak also showing trends towards an increased height of 1.54 (CIs: 0.81 –
 495 3.24) times the average (Fig. 6c).



496

497 Figure 6: Predictions for the phenological distribution of caterpillar biomass throughout
 498 spring on branches of different tree taxa (based on models in Table S3+4). a) shows the model
 499 predictions for the change in caterpillar biomass over time (ordinal date) for each tree taxon,
 500 the black dashed line depicts the curve prediction from the fixed effects. Plots b-d) show the
 501 difference between each tree taxon and the fixed effect prediction (indicated by a black
 502 dashed line at zero or one) for the b) timing of the peak in biomass, c) proportional height of
 503 the peak (maximum biomass) and d) width of the peak at half of the height. The median and
 504 95% credible were calculated using the posterior distributions for the fixed effects and tree
 505 taxa random effects and interactions.

506 The biomass peak width, representing the relative shape of the peak, continued not to differ
507 significantly between any taxon and the average trend (Fig. 6d), however, similarly to in the
508 abundance results, the peaks on hazel, oak, rowan and sycamore showed strong trends
509 towards deviating from average by -3.33 (CIs: -0.60 - 10.24), -0.82 (CIs: -3.39 - 1.29), -2.12
510 (CIs: -5.47 - 0.49) and -1.05 (CIs: -3.59 - 1.19) days respectively (Fig. 6d). Consequently, the
511 peak duration at a consistent height again showed more variation, yet did not significantly
512 differ between any taxon and the average trend (Fig. S6) because there were no distinct
513 variations in the phenological distribution of mass. Nevertheless, the caterpillar biomass peak
514 on beech, hazel, oak and willow showed strong trends towards differing in duration to an
515 average taxon, differing by -6.53 (CIs: -16.89 – 1.57), 10.40 (CIs: -1.72 - 27.50), 7.05 (CIs: -0.87
516 - 15.53) and 7.22 (CIs: -0.53 - 17.48) days respectively (Fig. S6). The biomass peak timing,
517 height and duration on oak differed to other taxa similarly to the peaks in abundance, with
518 the addition of the peak on beech now also falling significantly earlier than on oak (Fig. S7).

519 Discussion

520 We find that several aspects of the phenological distributions of abundance and biomass are
521 sensitive to tree taxon (Fig. 4, 6). The major difference was in the peak height, and like earlier
522 work (Mägi et al., 2009; Shutt, Burgess, et al., 2019; Veen et al., 2010) we find that caterpillars
523 are most abundant on oak and willow, with alder and ash especially poor (Fig. 3a, 4c, 6c). In
524 agreement with Shutt et al. 2019, we find timing differences between tree taxa to be quite
525 slight, though peak caterpillar phenology on sycamore trees falls earlier than on an average
526 taxon, with deviating trends also suggested in multiple other taxa (Fig. 4b, 6b). Whilst the
527 shape of the peak is generally consistent (Fig 4d, 6d), we show the peak in abundance lasts

528 for a significantly longer period of time on willow than average (Fig. S3), with some suggestion
529 that the peak duration also varies among other taxa (Fig S3, S6). We suggest among taxa
530 differences in biomass are more attributable to abundance than mass, as we find no evidence
531 for caterpillar mass gain differing among tree taxa compared to average (Fig. 5), although
532 caterpillars on beech reach a significantly lower mass than those on oak (Fig. S5). In addition,
533 we also found the prevalence of oak within the local woodland composition increases the
534 number of caterpillars hosted by any tree, regardless of taxon (Fig. 3c).

535 Our results strongly support previous work that finds the phenological distribution of
536 caterpillar abundance on oak trees reaches a high abundance but is present for a relatively
537 brief duration (van Balen, 1973; Varley et al., 1974; Veen et al., 2010; Visser et al., 2006). The
538 very high abundance and biomass of caterpillars observed in oak woodlands appears not to
539 be representative of other woodland types that are more common across the UK (Stagg &
540 Ward, 2019), despite its use as a baseline for comparison when the distribution in other
541 woodland compositions has been studied (Mägi et al., 2009). On the contrary, our results
542 suggest the phenological distribution of caterpillars on elm, rowan and sycamore are most
543 representative of an average deciduous tree, particularly in relation to peak height as there
544 are fewer distinctive differences in timing and duration. Generally speaking, oak, hazel, willow
545 and to some extent birch deviate towards hosting larger caterpillar abundances than average,
546 whilst alder, ash and to some extent beech deviate in hosting smaller abundances. These
547 differences in the phenological distribution of spring caterpillar abundance provide evidence
548 for biotic drivers of divergent caterpillar prevalence among tree taxa; though the causes
549 remain to be established.

550 We find that the phenological distribution of the mass of the sampled caterpillar guild
551 increases rapidly in early spring before asymptoting (Fig. 5). Our main results show minimal
552 difference in caterpillar growth rate and the mass of caterpillars at the end of spring when
553 comparing each taxon to an average tree; although caterpillars from beech weighed
554 significantly less than those sampled from oak, with mass also lower on birch and willow (Fig.
555 S5). Our results are consistent with rearing experiments using winter moth (*O. brumata*) that
556 found more successful growth and mass at pupation from feeding on oak compared to hazel
557 and beech (Feeny, 1970; Wint, 1983). Given our findings of greater abundance on oak and
558 these experimental findings, it is perhaps surprising that the caterpillars sampled do not reach
559 a higher mass or asymptote earlier than on other trees. However, as the budburst of oak falls
560 later than many taxa (Cole & Sheldon, 2017), the asymptote in caterpillar mass on oak relative
561 to the duration of food availability may well fall earlier than on other taxa. Our measure of
562 mass will capture various processes including individual growth, and the appearance and
563 disappearance of species from the guild through both varied phenologies among caterpillar
564 species and beta diversity between locations.

565 When we combine the minimal among tree difference in caterpillar mass with abundance
566 estimates we do see some effect on the phenological distribution of biomass, primarily
567 through adjustments to peak height (Fig. 4, 6). The greatest difference in peak dimensions
568 between the distributions of abundance and biomass among tree taxa is the increase in oak
569 peak height, and decrease in beech, relative to the other trees (Fig. 6, S7). As most studies
570 use frass fall or half-fall to quantify the peak in biomass (Charmantier et al., 2008; Hinks et al.,
571 2015; Smith et al., 2011; Visser, Noordwijk, Tinbergen, & Lessells, 1998), the contribution of
572 abundance and mass phenology to the biomass peak timing is unexplored. With just a two-

573 day difference in peak phenology between abundance and mass, our results suggest that the
574 trend in abundance is more prominent than mass in dictating the timing of the biomass peak.
575 Therefore, factors impacting caterpillar abundance differentially between locations are likely
576 to have a greater impact on spatial variation in caterpillar phenology than those impacting
577 mass.

578 The impact of oak foliage density, increasing caterpillar abundance throughout the local area
579 (Fig. 3c), suggests that oak-dominated woodlands may be unique in spring caterpillar
580 prevalence when compared to other woodland compositions. Whether the increased
581 abundance throughout areas with increasing oak is driven by increased survival rates or
582 successful pupation remains unknown, however increased fecundity is unlikely (Honěk, 1993)
583 to have a wide-spread impact as caterpillars from oak show no significant increase in mass
584 compared to those on an average tree. Fecundity may, however, influence the differences in
585 abundance between oak and specific taxa, such as beech, as the divergence in mass is more
586 substantial; however this difference could be driven by differing lepidopteran guild
587 compositions. The increased abundance on oak also appears not be driven by the general
588 density of foliage, as the site foliage density effect was close to zero and non-significant (Table
589 S1). The tall and broad crown of a mature oak tree may make dispersal of caterpillars to other
590 trees more successful than from other typically smaller tree taxa such as willow and birch. An
591 implication of these findings is that the common practice of monitoring caterpillar prevalence
592 purely on oak (Burgess et al., 2018; Hinks et al., 2015; Visser et al., 2006) may miss spatial-
593 heterogeneity in caterpillar abundance both between locations and within larger sites.
594 However, the increased caterpillar abundance on oak could make these habitats the most
595 important to consumers; acting as a source that contributes to the repopulation of areas with

596 lower consumer breeding success, as the prevalence of oak within an area has been shown
597 to increase fledging success (Shutt et al., 2018).

598 The tree composition within a woodland seems likely to influence mismatch buffering
599 through the variation it drives in dimensions of the site-level caterpillar peak. Our study
600 suggests the habitat-driven buffering potential may be stronger than previously suggested
601 (Shutt, Burgess, et al., 2019), as we have identified further differences in the caterpillar peak
602 among tree taxa, particularly in peak height but also in timing and duration. The differences
603 in peak height between tree taxa could impact mismatch between different areas within a
604 similar region or more broadly between woodlands and geographical locations as the tree
605 composition present is likely to influence the maximum caterpillar density reached,
606 potentially altering the negative consequences of trophic asynchrony for the consumer. The
607 peak on willow lasts for eight days longer than average, providing a food source for over a
608 third of the time many consumers spend in the nest. The peaks on alder, ash, beech, rowan
609 and sycamore were all shorter than on oak (Fig. S4), implying that the peak may vary
610 substantially in duration between an oak-dominated woodland and one with a mixed
611 composition. There is also potential for within site peak duration to be influenced by tree
612 composition through deviations in peak phenology, as the peak falls three days earlier on
613 sycamore than on average tree taxa (Fig. 4) and three-five days earlier on beech, rowan and
614 sycamore than on oak (Fig. S4).

615 Habitat composition is manipulatable by foresters and land managers and here we examine
616 the implications of our findings in this context. First, in general increased defoliation
617 negatively impacts on tree health and productivity (Kulman, 1971; Marquis & Whelan, 1994;
618 Whitham et al., 1991; Whittaker & Warrington, 1985). Defoliation may be minimised if there

619 is a lower density of oak, thereby preventing the additive effect it has on the abundance of
620 caterpillars overall. However, with the exception of oak, caterpillar abundance appeared to
621 be insensitive to the amount of other tree taxa present. The second implication relates to the
622 conservation of consumer populations for whom more resource is expected to be beneficial,
623 though the importance of resource abundance versus resource timing relative to breeding is
624 relatively underexplored (but see Naef-Daenzer & Keller, 1999; Ramakers, Gienapp, & Visser,
625 2019). The high density of prey in oak woodlands is thought to be a driver of preference for
626 this habitat by some breeding passerines (Perrins, 1979). However, mixed woodlands
627 containing trees with more varied caterpillar peak timings may extend the period of
628 caterpillar availability and total abundance across spring, also potentially of benefit to
629 consumer species. In addition our study reveals substantial site effects in the phenological
630 distribution of caterpillar abundance (Table S3), and some site effect on mass (Table S4),
631 which implies that factors in addition to habitat play an important role (Fig. 2).

632 We have shown new detailed insights into habitat-driven variations in caterpillar abundance
633 and the phenological distribution of caterpillar abundance and biomass among tree taxa. Our
634 results highlight the spatio-temporal differences in caterpillar phenology, abundance and
635 trophic interactions between different woodland compositions. We show that the
636 phenological distribution of spring invertebrates in oak-dominated study systems are not
637 representative of the distributions present across the broader range of deciduous woodland
638 compositions also at risk of temperature-mediated trophic mismatch. Our findings have
639 applications in woodland management and reforestation for the conservation of woodland
640 lepidopteran and insectivore communities.

641

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831 Supplementary Information

832 We found that alder, ash, beech, elm, rowan and sycamore all support significantly fewer
833 caterpillars than oak (Fig. S1), hosting 0.42 (CIs: 0.29 - 0.58), 0.34 (CIs: 0.22 - 0.54), 0.52 (CIs:
834 0.39 - 0.69), 0.60 (CIs: 0.41 - 0.84), 0.55 (CIs: 0.39 - 0.78) and 0.67 (CIs: 0.53 - 0.84) times as
835 many caterpillars respectively.

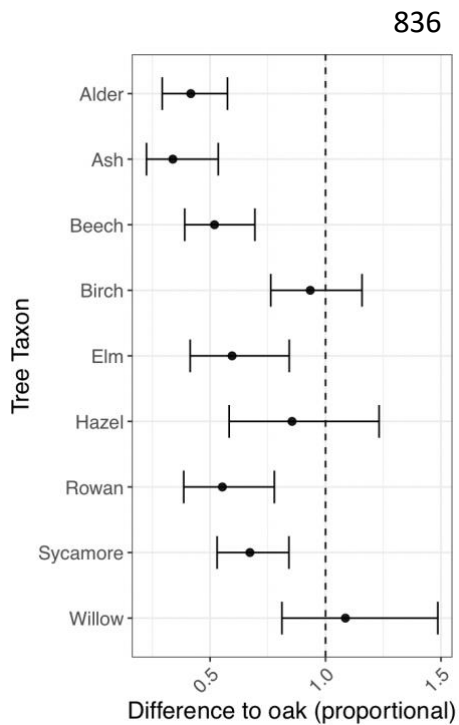


Figure S1: Posterior median and 95% credible intervals of the model prediction for the proportional difference in caterpillar abundance between each tree taxon and oak. Calculated from tree taxa random effects in a Poisson GLMM (Table S1). Black dashed line at 1.0 indicates no difference to oak/significance threshold.

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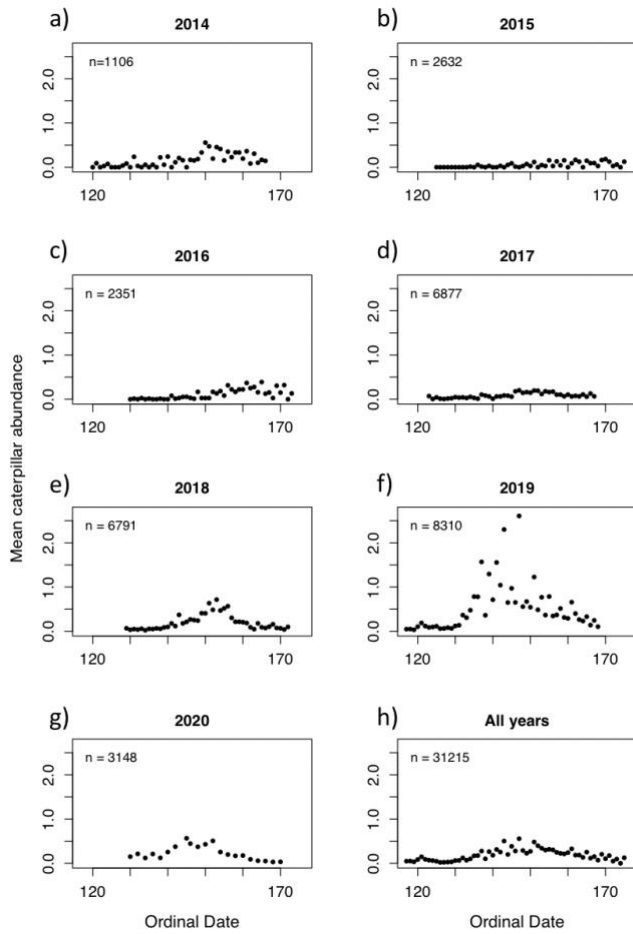


Figure S2: Plots a-g show the mean abundance of caterpillars sampled on each day of each year and plot h shows the mean abundance on each day across all years, calculated from the mean at each site on each day, n indicates the total number of beating samples within each year or the total for all years.

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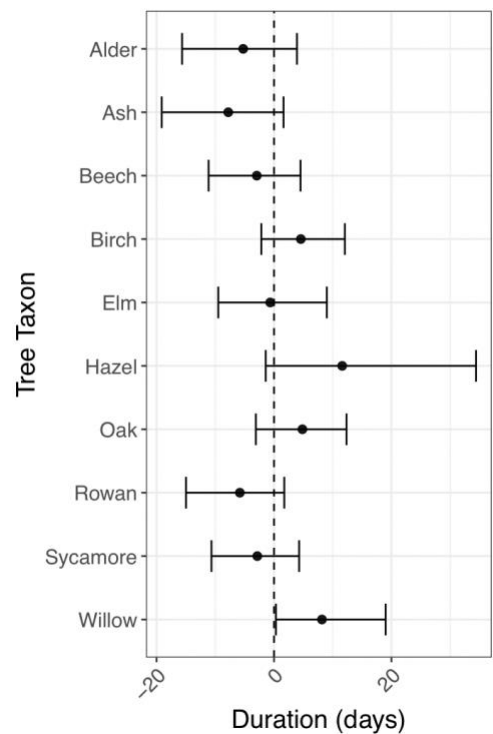
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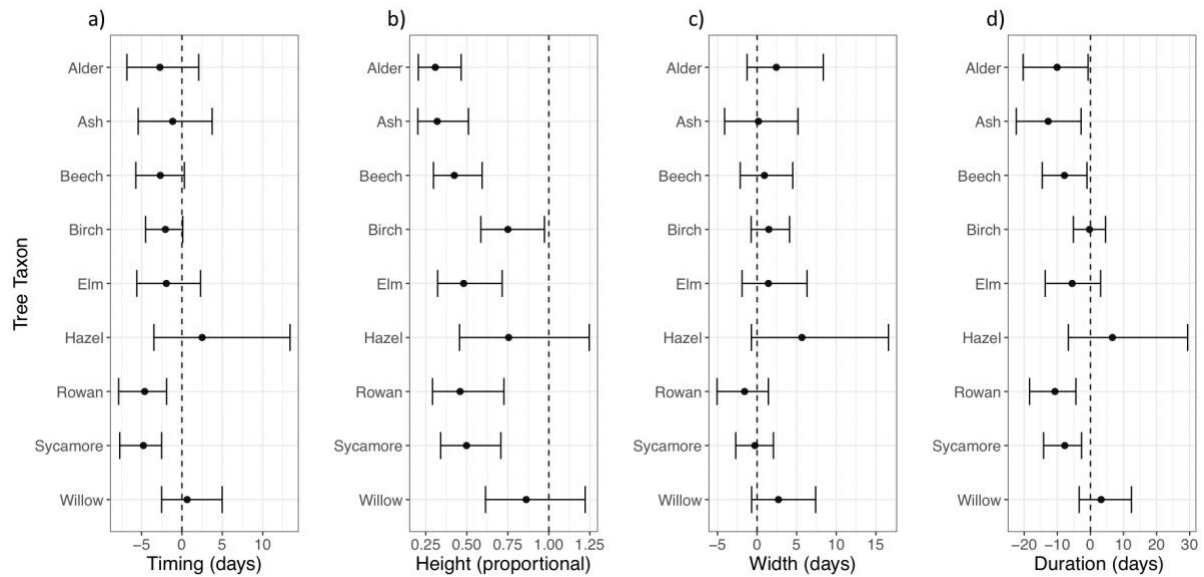
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875 Figure S3: Posterior median and 95% credible intervals for
 876 model prediction of the difference in peak duration (at an
 877 abundance of 0.01 caterpillars) for each taxon compared
 878 to the fixed effect trend. Calculated from the posterior
 879 distributions for the fixed effects and tree taxa random
 880 effects and interactions in a Poisson GLMM (Table S4).

881





882

883 Figure S4: Posterior median and 95% credible intervals for (Table S4) predictions of the
 884 difference between each tree taxon and oak for different properties of the phenological
 885 distribution of caterpillar abundance in spring, calculated from the posterior distributions for
 886 the fixed effects and tree taxa random effects and interactions. Plots show a) the difference
 887 in the timing of the peak in abundance, b) the proportional difference in the height of the
 888 peak (maximum abundance), c) the difference in the peak width at half of the peak height
 889 and d) the difference in the duration of the peak (at a set abundance of 0.01 caterpillars).
 890

891 When comparing the abundance peak metrics of each taxon to the peak on oak we saw a
 892 greater number of significant differences than compared to the average trend. The peak on
 893 rowan and sycamore trees was significantly earlier than on oak by -4.61 (CIs: -7.84 - -1.90)
 894 and -4.78 (CIs: -7.70 - -2.52) days respectively (Fig. S4). The peak on alder, ash, beech, birch,
 895 elm, rowan and sycamore were all found to have a significantly lower height, reaching 0.31
 896 (CIs: 0.20 - 0.47), 0.32 (CIs: 0.20 - 0.51), 0.42 (CIs: 0.30 - 0.59), 0.75 (CIs: 0.59 - 0.97), 0.48 (CIs:
 897 0.32 - 0.72), 0.46 (CIs: 0.29 - 0.73) and 0.50 (CIs: 0.34 - 0.71) times the height of the peak on
 898 oak trees, respectively. The width of the peak at half the height on birch, hazel and willow
 899 showed non-significant trends towards being broader than on oak by 1.48 (CIs: -0.75 - 4.10),
 900 5.66 (CIs: -0.72 - 16.69) and 2.69 (CIs: -0.69 - 7.40) days respectively. The peaks on alder, ash,

901 beech, rowan and sycamore last for a significantly shorter duration by -10.07 (CIs: -20.34 - -
902 0.71), -12.76 (CIs: -22.40 - -2.82), -7.84 (CIs: -14.56 - -1.04), -10.72 (CIs: -18.39 - -4.38) and -
903 7.75 (CIs: -14.15 - -2.98) days respectively (Fig. S4).

904

905 When comparing the day 168 mass of caterpillars on each tree taxa to those from oak we
906 found caterpillars sampled from beech weighed significantly less at 0.60 (0.38 - 0.97) times
907 the mass of those sampled from oak (Fig. S5). Birch and willow also show non-significant
908 trends towards a lower mass, reaching 0.83 (CIs: 0.65 - 1.03) and 0.79 (CIs: 0.55 - 1.08) times
909 the mass respectively (Fig. S5).

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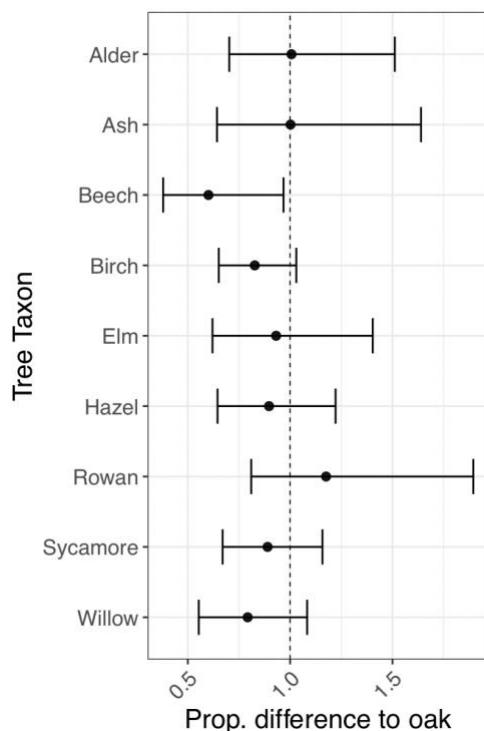


Figure S5: Posterior median and 95% credible intervals for model predictions of the proportional difference in caterpillar mass between each tree taxon and caterpillars sampled from oak (Table S5). Calculated for day 168 (16th June) for the average year and site, the latest date with caterpillar mass data for all tree taxa, using the posterior distributions for all fixed effect parameters and tree taxa random effects and interactions in a censored-gaussian GLMM.

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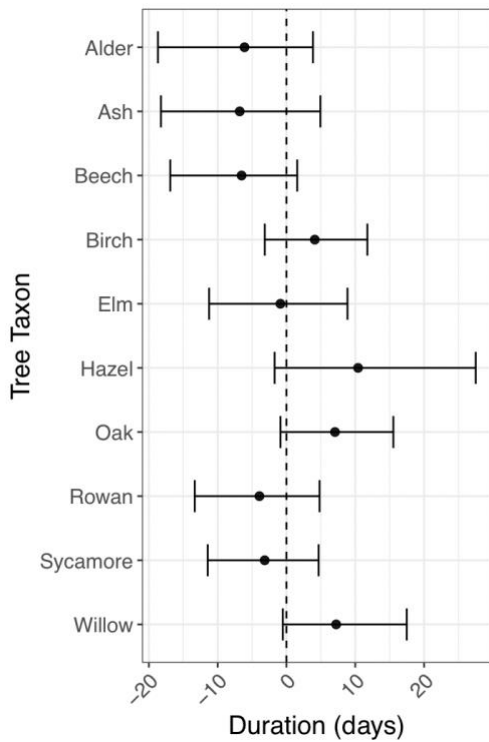


Figure S6: Posterior median and 95% credible intervals for model prediction of the difference in peak duration (at a biomass of 0.00035g of caterpillars) for each taxon compared to the fixed effect trend (Table S4+5). Calculated from the posterior distributions for the fixed effects and tree taxa random effects and interactions in a bivariate censored-gaussian (mass) and Poisson (abundance) GLMM.

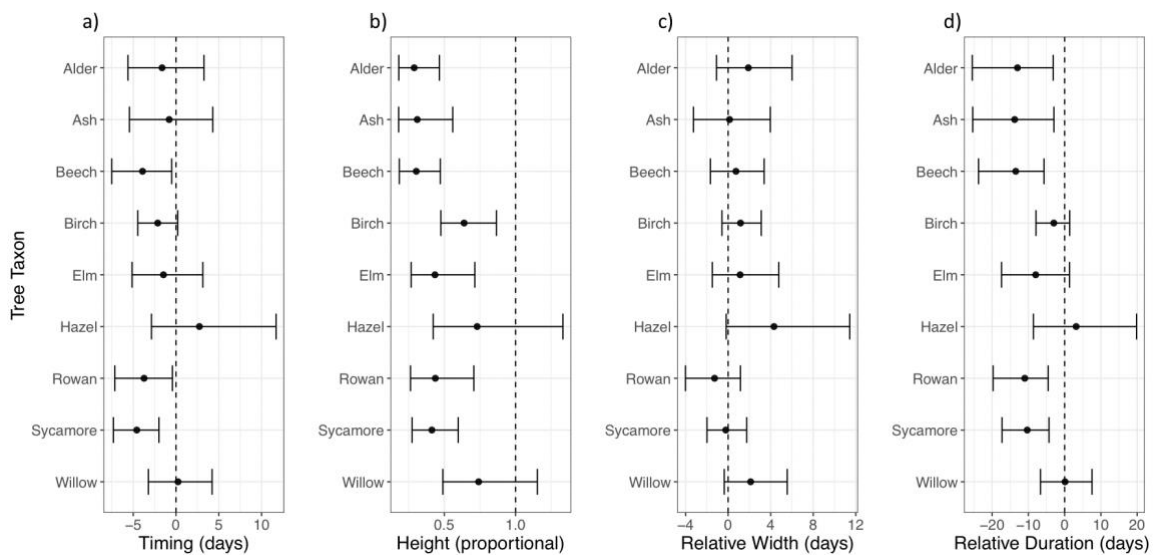
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941 As with the distributions of caterpillar abundance among tree taxa, there were multiple
 942 differences between each taxon and oak in the biomass peak metrics (Fig. S7). The peak on
 943 beech, rowan and sycamore trees was significantly earlier than on oak by -3.91 (CIs: -7.51 - -
 944 0.51), -3.73 (CIs: -7.15 - 0.42) and -4.58 (CIs: -7.31 - -1.99) days respectively (Fig. S7), and the
 945 peak timing on birch trees also showed a strong trend towards falling -2.14 (CIs: -4.46 - 0.20)
 946 days earlier. The biomass peaks on alder, ash, beech, birch, elm, rowan and sycamore were
 947 found to have a significantly lower height, reaching 0.29 (CIs: 0.18 - 0.47), 0.31 (CIs: 0.18 -
 948 0.56), 0.31 (CIs: 0.18 - 0.47), 0.64 (CIs: 0.48 - 0.87), 0.43 (CIs: 0.27 - 0.71), 0.44 (CIs: 0.26 - 0.71)
 949 and 0.41 (CIs: 0.27 - 0.60) times the height of the peak on oak trees, respectively. The width
 950 of the peak at half the height on birch, hazel and willow showed non-significant trends
 951 towards being broader than on oak by 1.67 (CIs: -0.58 - 3.11), 4.31 (CIs: -0.18 - 11.43) and 2.12

952 (CIs: -0.37 - 5.56) days respectively. The peak duration on alder, ash, beech, rowan and
 953 sycamore lasted for a significantly shorter duration by -13.01(CIs: -25.47 - -3.16), -13.78 (CIs:
 954 -25.36 - -2.69), -13.48 (CIs: -23.74 - -5.69), -11.00 (CIs: -19.73 - -4.53) and -10.32 (CIs: -17.27 -
 955 -4.32) days respectively, and peaks on birch and elm also show strong non-significant trends
 956 towards a shorter duration by -2.98 (CIs: -7.91 - 1.39) and -7.98 (CIs: -17.40 - 1.34) days (Fig.
 957 S7).

958



959

960 Figure S7: Posterior median and 95% credible intervals for combined censored-gaussian
 961 (mass) and Poisson (abundance) GLMM (Table S4+5) predictions of the difference between
 962 each tree taxon and oak for different properties of the phenological distribution of caterpillar
 963 biomass in spring, calculated from the posterior distributions for the fixed effects and tree
 964 taxa random effects and interactions. Plots show a) the difference in the timing of the peak
 965 in biomass, b) the proportional difference in the height of the peak (maximum biomass), c)
 966 the difference in the peak width at half of the peak height and d) the difference in the duration
 967 of the peak (at a set biomass of 0.00035g of caterpillars).

968

969 **Model output tables**

970 Table S1: Poisson GLMM for variance composition of caterpillar abundance. Date refers to
 971 ordinal date (scaled: mean = 146.7727, SD =14.04083).

	Coefficient/Variance (Mean/mode and CI)	Effective sample size
<u>Fixed Terms</u>		
Intercept	-3.319 (-4.111 - -2.533)	4900
Date (scaled)	0.48 (0.388 - 0.57)	4655
Date ² (scaled)	-0.442 (-0.528 - -0.355)	4482
<u>Random Terms</u>		
Site	0.667 (0.36 - 1.075)	1844
Tree ID	0.273 (0.222 - 0.34)	4900
Tree Taxa	0.146 (0.048 - 0.608)	1657
Site Day	0.491 (0.424 - 0.598)	4148
Day	0.272 (0.195 - 0.364)	4539
Site Year	0.258 (0.184 - 0.359)	3392
Year	0.278 (0.077 - 1.885)	1111
Recorder	0.103 (0.048 - 0.33)	2814
Residual	0.641 (0.563 - 0.739)	3885

972

973 Table S2: Poisson GLMM analysing differences in caterpillar abundance between the different
 974 tree taxa sampled, the effect of site foliage density and the effect of the amount of foliage of
 975 each tree taxon within the local woodland composition, included using multi-membership.

	Coefficient/Variance (Mean/mode and CI)	Effective sample size
<u>Fixed Terms</u>		
Intercept	-4.02 (-4.892 - -3.192)	4707
Total Foliage Score	0.001 (-0.019 - 0.018)	3842
<u>Random Terms</u>		
Sampled Tree Taxa	0.126 (0.038 - 0.594)	3146
Habitat Composition	0 (0 - 0.001)	1063
Site	0.388 (0.194 - 0.696)	3547
Year	0.288 (0.062 - 1.796)	816
Site Year	0.232 (0.158 - 0.336)	5000
Tree ID	0.269 (0.221 - 0.338)	5000
Site Day	1.032 (0.916 - 1.166)	5000
Recorder	0.166 (0.069 - 0.484)	3030
Residual	0.643 (0.57 - 0.746)	4589

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977

978 Table S3: Poisson GLMM for analysing variations among tree taxa in the phenological
 979 distribution of caterpillar abundance throughout spring. Date refers to ordinal date (scaled:
 980 mean = 146.7727, SD =14.04083).

	Coefficient/Variance (Mean/mode and CI)	Effective sample size
<u>Fixed Terms</u>		
Intercept	-3.164 (-4.191 - -2.225)	10000
Date (scaled)	0.864 (0.042 - 1.743)	8866
Date ² (scaled)	-0.841 (-1.177 - -0.523)	7069
<u>Random Terms</u>		
TreeTaxa- Intercept var	0.154 (0.036 - 0.626)	1340
TreeTaxa- Intercept:Date slope covar	0.02 (-0.097 - 0.154)	8865
TreeTaxa- Intercept:Date ² slope covar	0.002 (-0.102 - 0.081)	8750
TreeTaxa- Date slope var	0.046 (0.011 - 0.212)	3123
TreeTaxa- Date slope:Date ² slope covar	0 (-0.04 - 0.068)	8719
TreeTaxa- Date ² slope var	0.02 (0 - 0.129)	2586
Site- Intercept var	0.872 (0.553 - 1.548)	6973
Site- Intercept:Date slope covar	-0.104 (-0.337 - 0.056)	6315
Site- Intercept:Date ² slope covar	-0.229 (-0.437 - -0.124)	5703
Site- Date slope var	0.243 (0.145 - 0.425)	3166
Site- Date slope:Date ² slope covar	0.018 (-0.041 - 0.092)	6168
Site- Date ² slope var	0.105 (0.057 - 0.186)	5096
Year- Intercept var	0.472 (0.159 - 3.22)	473
Year- Intercept:Date slope covar	-0.293 (-1.975 - 0.367)	7155
Year- Intercept:Date ² slope covar	0.038 (-0.269 - 0.491)	5163
Year- Date slope var	0.501 (0.171 - 2.851)	578
Year- Date slope:Date ² slope covar	-0.052 (-0.627 - 0.149)	5671
Year- Date ² slope var	0.047 (0.005 - 0.339)	1302
Site-Year	0.259 (0.197 - 0.37)	5472
Recorder	0.146 (0.064 - 0.355)	2616
Site-Day	0.266 (0.204 - 0.333)	4355
Tree ID	0.263 (0.212 - 0.328)	6000
Residual	0.64 (0.555 - 0.727)	2203

981
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 984

985 Table S4: Censored-gaussian GLMM for analysing variations among tree taxa in the
 986 phenological distribution of log caterpillar mass throughout spring. Date refers to ordinal date
 987 (scaled: mean = 146.7727, SD =14.04083).

	Coefficient/Variance (Mean/mode and CI)	Effective sample size
<u>Fixed Terms</u>		
Intercept	-3.884 (-4.131 - -3.651)	10446
Date scaled	0.533 (0.38 - 0.687)	3491
Date ² scaled	-0.147 (-0.207 - -0.088)	1549
<u>Random Terms</u>		
TreeTaxa- Intercept var	0 (0 - 0.026)	2113
TreeTaxa- Intercept:Date slope covar	0 (-0.006 - 0.024)	3714
TreeTaxa- Date slope var	0.001 (0 - 0.086)	1011
Site- Intercept var	0.029 (0 - 0.078)	948
Site- Intercept:Date slope covar	-0.009 (-0.04 - 0.016)	2207
Site- Date slope var	0.044 (0.015 - 0.098)	1652
Year	0 (0 - 0.183)	1063
Site-Year	0.02 (0 - 0.052)	824
Recorder	0.006 (0 - 0.04)	2522
Site-Day	0.128 (0.087 - 0.184)	2677
Tree ID	0.041 (0.021 - 0.071)	2614
Weighting	0.97 (0.863 - 1.069)	1388
Residual	0.023 (0.002 - 0.055)	671

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991 Asymmetry in the abundance peak

992 **Methods**

993 It is already established that there is a peaked phenological distribution of caterpillar
994 abundance throughout spring in temperate deciduous woodlands (Gibb, 1950; Shutt,
995 Burgess, et al., 2019; Southwood et al., 2004; van Balen, 1973), though the shape of peak has
996 not been examined. We assessed the adequacy of the quadratic function in describing the
997 shape of the phenological distribution of caterpillar abundance by also modelling the peak
998 including a cubic date term; allowing an asymmetrical trend over time. The extent of
999 asymmetry was quantified by comparing the percentage of the peak width predicted to fall
1000 before and after the vertex of the curve.

1001

1002 The peak asymmetry model included the caterpillar abundance per sample as the response
1003 variable with an intercept, date, date² and date³ in the fixed effects, allowing an asymmetrical
1004 peak (Table S5). Date refers to ordinal date and was scaled (prior to scaling: mean = 146.77,
1005 SD =14.04). The random terms allowed for each site-year to have a separate intercept, date
1006 and date² slopes and for covariance among these terms. Separate site and year random
1007 intercepts and slopes were not included to aid model convergence. We do not anticipate this
1008 will pose an issue, as our aim was solely to identify the general phenological trend. We
1009 calculated the extent of asymmetry at each quartile of the peak height across the posterior
1010 distribution because the ratio of duration to the left and right of the peak is not necessarily
1011 constant.

1012 Table S5: Model composition for the peak asymmetry model. Date refers to ordinal date
 1013 which was scaled (mean = 146.77, SD =14.04). Int = intercept and VCV = variances and
 1014 covariances.

Model	Peak asymmetry (Table S6)
Motivation	Evidence of asymmetry in peak shape
Response	Caterpillar abundance
Fixed effects	Date Date ² Date ³
Random terms	Int, Date and Date ² VCV across Site-year Day-site-year Tree ID Recorder
Family	Poisson
Iterations (thin)	2000000 (1000)
Burnin	50000
Sample Size	1950

1015

1016 **Results**

1017 Annual peaks in the temporal distribution of caterpillar abundance in our data are clear (Fig.
 1018 S2), and supported by the significant date² term (-0.666, CIs: -0.769 - -0.565, Table S6) in the
 1019 peak asymmetry model. The cubic parameter was significant and negative (-0.193, CIs: -0.256
 1020 - -0.124), indicating a negatively skewed peak (Table S6), however the resulting asymmetry in
 1021 the curve is quite small (Fig. S8). Each quartile shows 52-57% of the peak duration to the left
 1022 and 43-48% to the right. On the basis of the quite minor asymmetry we conclude that
 1023 inclusion of the quadratic date term, without cubic, is sufficient.

1024

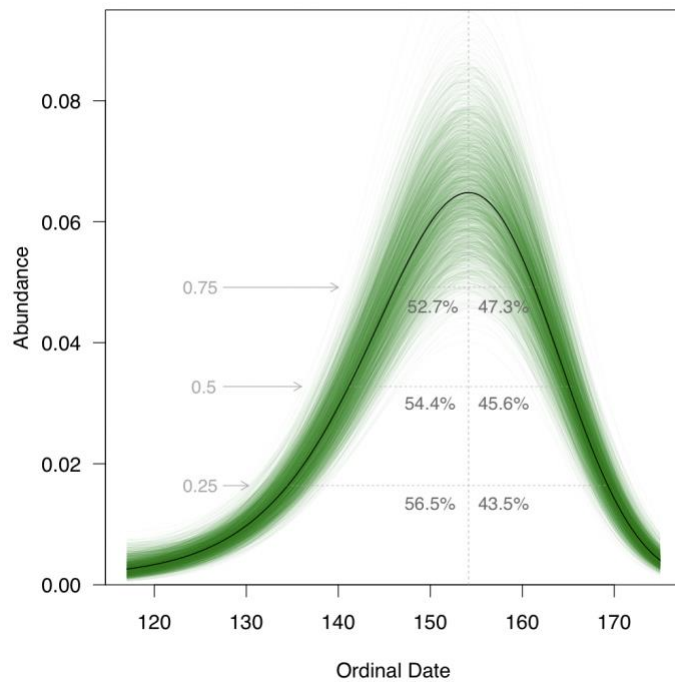


Figure S8: Posterior distribution for Poisson GLMM allowing a cubic (asymmetrical) relationship between caterpillar abundance and ordinal date (Table S3). Percentage of peak width falling before and after the peak date at 0.25, 0.5 and 0.75 of the peak height are shown in grey.

1036

1037 Table S6: Poisson GLMM using a cubic function to analyse possible skew in the distribution of
 1038 caterpillar abundance across ordinal date (scaled: mean = 146.7727, SD =14.04083).

	Coefficient/Variance (Mean/mode and CI)	Effective sample size
<u>Fixed Terms</u>		
Intercept	-2.977 (-3.244 - -2.699)	1950
Date (scaled)	0.862 (0.703 - 1.024)	1684
Date ² (scaled)	-0.666 (-0.769 - -0.565)	1506
Date ³ (scaled)	-0.193 (-0.256 - -0.124)	1589
<u>Random Terms</u>		
SiteYear- Intercept var	1.6 (1.3 - 2.059)	1950
SiteYear- Intercept:Date slope covar	-0.278 (-0.473 - -0.085)	2078
SiteYear- Intercept:Date ² slope covar	-0.425 (-0.583 - -0.303)	1512
SiteYear- Date slope var	0.643 (0.489 - 0.851)	1659
SiteYear- Date slope:Date ² slope covar	0.074 (-0.024 - 0.153)	1734
SiteYear- Date ² slope var	0.202 (0.13 - 0.27)	2105
Recorder	0.164 (0.065 - 0.382)	1690
Site Day	0.232 (0.175 - 0.3)	1950
Tree ID	0.335 (0.274 - 0.402)	1950
Residual	0.632 (0.568 - 0.735)	1826

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