

1 Summer solstice orchestrates the  
2 subcontinental-scale synchrony of mast seeding

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18 **High interannual variation in seed production in perennial plants, can**  
19 **be synchronized at subcontinental scales with wide consequences for**  
20 **ecosystem functioning, but how such synchrony is generated is unclear.**  
21 **We investigated the factors contributing to masting synchrony in Euro-**  
22 **pean beech (*Fagus sylvatica*), that extends to the 2000 km geographic**  
23 **range. Maximizing masting synchrony via spatial weather coordination,**  
24 **known as the Moran effect, requires a simultaneous response to weather**  
25 **conditions across distant populations. A celestial cue that occurs si-**  
26 **multaneously across the entire Hemisphere is the longest day (summer**  
27 **solstice). We show that European beech abruptly opens its temperature-**  
28 **sensing window on the solstice, hence widely separated populations all**  
29 **start responding to weather signals in the same week. This celestial**  
30 **"starting gun" generates ecological events with high spatial synchrony**  
31 **across the continent.**

## Introduction

Subcontinental synchronization of interannual variation in seed production by trees means that, in a large-seeding year, a large pulse of resources is made available to wildlife over the majority of the continent by virtue of the synchronous reproduction by millions of trees [1–3]. Such large-seeding years are usually followed by reproductive failures, creating subsequent famine [4, 5]. This synchronized cycle of abundance and shortage triggers far-reaching disruptions in food webs, including rodent outbreaks [6], migration of ungulates and birds [7–11], and spikes in wildlife-borne human diseases [12, 13]. Moreover, in some species tree and leaf growth is reduced in years of high seed production, creating large-scale fluctuations in carbon sequestration [14–16]. The spatial scale of synchrony is a key aspect that amplifies the ecological importance of year-to-year variation in seed production [17–19]. However, several key questions on how plants synchronize masting over such extensive spatial scales remain unanswered.

Masting is synchronous and highly variable reproduction among years by a population of perennial plants [20]. Synchrony often exceeds among populations and decades of investigation have demonstrated that the regional synchronization of masting stems from the Moran effect, i.e. it is driven by spatially synchronized environmental signals [1, 2, 7, 21–25]. A major mechanism that governs the annual allocation of resources to seed production is weather variation [26–28]. These so-called weather cues are employed by plants to maintain synchronized reproductive fluctuations within populations [29]. The mechanisms underpinning weather cues exhibit species-specific variation, with a major example being the influence of summer temperature on the promotion of flower production [30, 31]. Masting plants have evolved to be remarkably responsive to temperature fluctuations [29, 32]. Consequently, substantial flowering effort, or masting, is triggered when the temperature meets some species-specific criteria. For example, one common link is that warm summers are associated with a high subsequent seed crop [33–36]. If individuals and populations collectively respond to the same cue across extensive regions, the spatial scale of masting synchrony aligns with the broad-scale synchronization of weather patterns [2, 21, 23, 24]. Crucially, the highest regional synchrony is achieved if the cue window is temporally conserved [3]. For instance, if the cue window shifts among populations to be earlier in the year in warmer climates, akin to the advancement of bud break or flowering in warmer climate [37, 38], regional synchrony deteriorates [3].

A global comparison of the Moran effect in masting showed that European beech (*Fagus sylvatica*) has stronger spatial synchrony than any other species in Europe (matched only by white spruce, *Picea glauca*, in North America). Both species have very large ranges, including latitudinally (north-south), which makes their ability to maintain significant masting synchrony exceeding 1500 km especially puzzling [2, 3]. New evidence has demonstrated that the timing of the cue window is exceptionally well conserved across the range of European beech [3], but the mechanism facilitating this stability remains unknown.

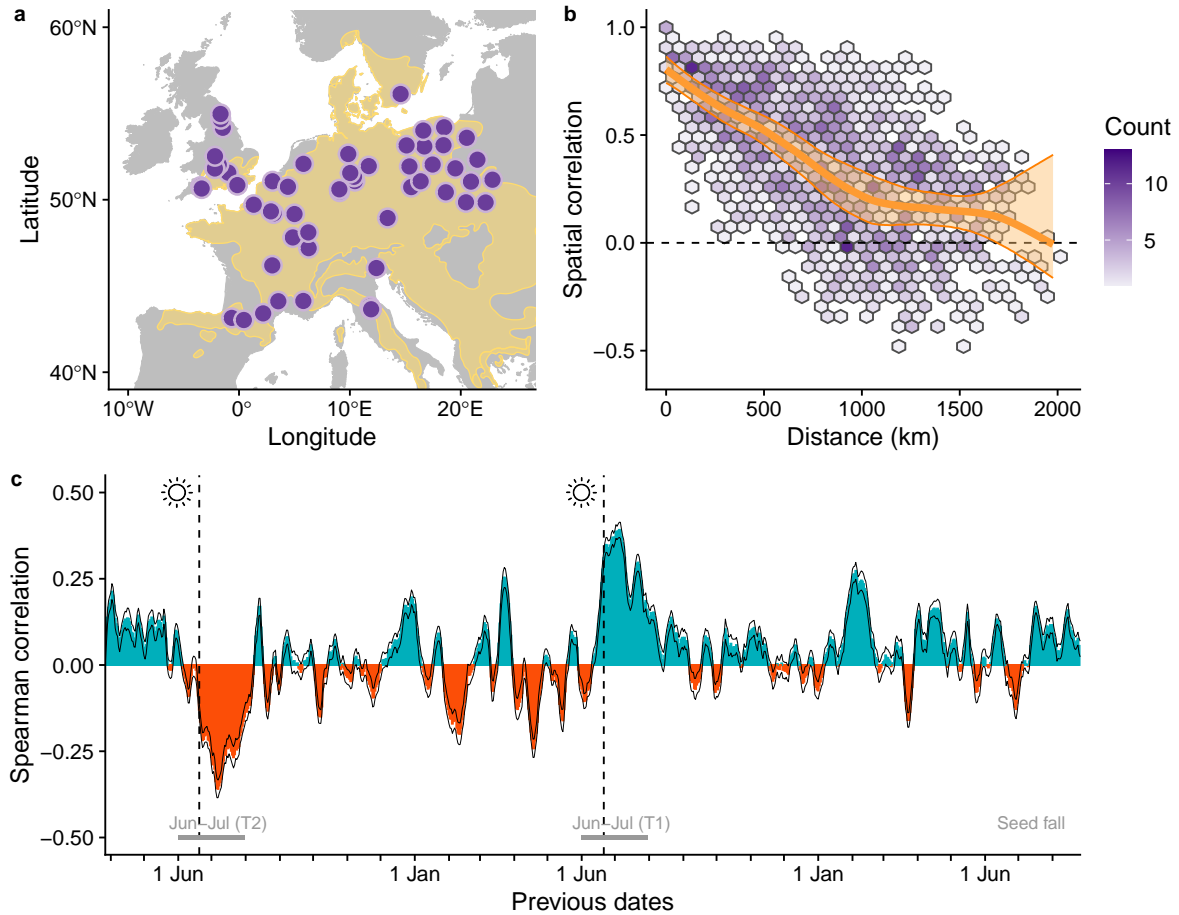
Recently, Zohner *et al.* [39] showed that temperature has opposite effects on leaf senes-

72 cence phenology before and after the summer solstice, demonstrating that trees can time  
73 their physiology using maximum day length. Similarly, wood formation in trees switches  
74 from a focus on cell division to secondary wall thickening once days start to shorten [40,  
75 41]. The capacity to sense the timing of solstice might explain the remarkable consistency  
76 of the temporal window during which beech trees are receptive to cues triggering masting  
77 events [3, 42]. Right across the European continent, the cues are anchored to June and  
78 July temperatures despite great differences in climate and day length among studied sites  
79 (mean summer annual temperature range across sites: 7.7 – 16.6 °C, maximum day length  
80 15 hr 25 mins to 16 hr 50 mins) [3], and a significant warming trend (1 °C over 40 years)  
81 [42], resulting in substantial variation in the growing season onset and duration. Here, we  
82 show that European beech achieves high synchrony across very large scales by anchoring  
83 the weather cue window to the summer solstice - the longest day of the year that occurs  
84 simultaneously across the whole Northern Hemisphere. The solstice alignment enables  
85 cohesive timekeeping across distant beech populations inhabiting diverse climatic regions.

## 86 **Results**

### 87 **Weather cues in beech are anchored to the solstice**

88 We used a moving window analysis and observed an abrupt rise in the correlation between  
89 European beech seed production and temperatures at the summer solstice, showing that  
90 the longest day coincides with the opening of the cue window. We ran a moving window  
91 correlation between annual seed production and mean temperatures in 61 populations of  
92 European beech sampled across the species' range (Fig. 1a, mean number of annual  
93 observations per site = 35, max = 68), across which masting synchrony reaches 2000  
94 km (Fig. 1b). The correlation coefficients displayed two distinct peaks, consistent among  
95 populations independent of latitude or local climate (Fig. 1c). The first was a negative peak  
96 occurring two years before seed fall, and the second was a positive peak arising one year  
97 before seed fall, both following the summer solstice (Fig. 1c). The correlations' directions  
98 between European beech masting and temperature in the summers before seed fall, one  
99 (T1), and two (T2) years prior align with previous research [33, 42, 44]. Here, however, we  
100 discover that the June-July period is not a mere consequence of physiological processes  
101 operating at that time in the plant (such as resource priming in year T2; cf. [44]), but the cue  
102 anchoring to the solstice. According to our hypothesis, European beech uses the solstice as  
103 a trigger to open the cue window. In support of this, correlation coefficient values between  
104 seed production and temperature rapidly increased after the solstice, compared to before it  
105 (Fig. 2). To test that, we fitted a generalized additive model (GAM), in which a response was  
106 the correlation between annual seed production and mean temperature in the 7-day rolling  
107 window. The predictors were relative day length (with maximum happening at the summer  
108 solstice) and the summer solstice as a categorical variable (before and after the solstice)



**Figure 1:** a) Locations of the 61 time series of annual seed production of European beech used in the study (average N years per series = 38). The yellow area highlights the species range (based on EUFORGEN, [43]). b) Spatial correlation between seed production over years and sites. The orange line represents the non-parametric spatial covariance function, with the shaded area showing 95% bootstrap confidence envelope. Hexes are pairwise Spearman correlations between sites, with the hex color scaled to the number of observations within each hex. Pairwise correlations were calculated for series with at least 5 years of overlap, that we deemed a minimum number of observations to calculate a correlation. c) Mean rolling Spearman correlation between temperature and masting averaged across all 61 sites. The graph shows correlations in two (T2) and one (T1) years before seed production, up until September when seed fall happens. The size of the temperature window is 7 days, with a 1-day step, and correlations are plotted according to the day of the year at the end of each 7-day window. Black dashed lines close to the sun icon indicate the summer solstice (21st June). Correlations are coded blue for positive, and red for negative. The black solid lines represent the standard error of the correlation coefficients across the sites for each window. Correlation for each site separately is reported in Fig S1.

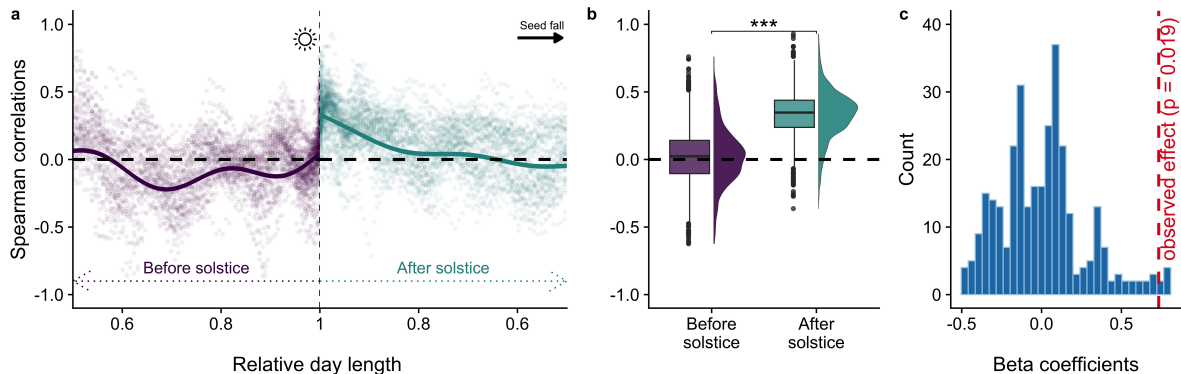


Figure 2: a) Correlations between European beech seed production and temperature at 7-day windows abruptly increase just after the summer solstice. The prediction lines and associated 95% confidence intervals are based on the GAM model with Spearman correlation coefficients as a response and interaction between relative day length (at the particular time window) and the time of the year (before vs. after the summer solstice) as predictors. Each point is the per site ( $N = 61$ ) per window correlation coefficient. b) Correlations shown at before and after the solstice shown at a) binned into 0.95 - 1 (max) day length before and after the solstice to help visualize abrupt change at the same day length at the two sides of the solstice. The asterisks indicate a significant ( $p < 0.0001$ ) difference between boxplots (see Methods). c) Results of a null model testing whether the abrupt increase in the correlation coefficients (between masting and summer temperatures) is highest at the summer solstice or any other day in the year. Histogram shows the distribution of beta coefficients under the null hypothesis i.e. the before-after difference in correlation coefficients in any other day of the year considered instead of the solstice. The red dashed line shows the observed effect i.e., how correlation coefficients “after solstice” differ as compared to “before solstice”. The null model was rejected ( $p = 0.02$ ). All three graphs show the effects for summer T1 (one year before seedfall), effects for summer T2 are presented in Fig. S2.

109 with their interaction. The interaction was significant (Table S1), showing that European  
 110 beech abruptly starts responding to temperature after, but not before, the solstice. The  
 111 predicted correlation between masting and temperature was  $\sim 4$ -fold higher one day after  
 112 the solstice compared to one day before (correlation before the solstice, 0.095; after the  
 113 solstice, 0.37, Fig. 2). Comparing the correlations in bins, defined as 0.95 - 1 (max) day  
 114 length, before (mean correlation = 0.02) and after the solstice (0.35) shows an even greater  
 115 magnitude of difference (Fig. 2b). We rejected the null hypothesis that the highest difference  
 116 in correlations at the solstice would emerge by chance ( $p = 0.02$ ) (Fig. 2c) (see Methods  
 117 for details). Similar anchoring at solstice was observed for the temperature effects in year  
 118 T2 (Fig. S2).

## 119 Discussion

120 More than a mere outcome of the Moran effect, the regional synchrony in European beech  
 121 masting stems from a mechanism that effectively harnesses weather variation for synchro-

122 nization. The negative correlation of European beech masting with temperatures in sum-  
123 mer T2 has been proximally associated with enhanced resource accumulation in cooler  
124 years [44, 45], and the positive correlation in summer T1 with enhanced flower primor-  
125 dia differentiation [30, 33, 44]. Of course, the positive effect of resource accumulation on  
126 seed production does not only appear after the longest day, and flower primordia can be-  
127 gin differentiating in response to high temperatures before the celestial event. Theoretical  
128 predictions suggest that masting can be triggered by any weather cue, yet cues are most  
129 likely to evolve from weather variation that affects reproduction-related processes [20, 29,  
130 46]. This sensitivity of such reproduction-related physiological processes is then reinforced  
131 by natural selection when variable and synchronized reproduction enhances plant fitness,  
132 as documented in a previous study on European beech [47]. However, to ensure ecological  
133 synchrony through weather-based alignment, cue-sensing phenology must be harmonized  
134 across populations. In species in which the cue-sensing is not harmonized, such as *Picea*  
135 *abies* or *Quercus robur*, synchrony deteriorates [3]. Temperatures are highly synchronized  
136 across space, but not necessarily over time. If one population opens its weather cue win-  
137 dow in July, but another in August, they will likely be responding to different temperature  
138 conditions, and their subsequent masting patterns will diverge, reducing synchrony [3]. The  
139 problem is that plants at more northern latitudes have much longer days (by more than an  
140 hour) near midsummer than plants further south, so using absolute day length to open the  
141 weather cue window would result in a spread of cue windows by latitude, and consequently  
142 large-scale asynchrony. A celestial cue that happens simultaneously across the entire con-  
143 tinent, capable of transcending environmental diversity among very distant populations is  
144 the annual maximum day length at the summer solstice. European beech responds to the  
145 solstice to open the cue window on the same day at all latitudes, which creates a high  
146 precision timing of the Moran effect.

147 In addition to summer temperatures, beech masting is correlated with summer (June/July)  
148 precipitation (in T1 and T2), and with spring weather [16, 44]. We thus run a similar, 7-days  
149 moving window analysis for precipitation. Consistent with past work, precipitation in June  
150 and July was positively (in T2) and negatively (in T1) correlated with seed production (Fig.  
151 S3. That effect was, however, not anchored to the solstice. Based on that, we hypoth-  
152 esize that temperature is the primary weather cue for beech masting, while precipitation  
153 during these periods may modulate the strength of temperature effects. That is supported  
154 by generally weaker and less spatially consistent correlations between beech masting and  
155 precipitation compared to temperature effects reported in past studies [44, 48]. Analyses  
156 based on gene expression levels will shed more light on these mechanisms. The effects of  
157 weather during pollination appear generally negligible compared to summer weather [44],  
158 even if important locally [16]. In support, the peaks in correlations during the flowering sea-  
159 son are not distinguishable from fluctuations along the two-year period that precedes seed  
160 fall (Fig. 1, S3).

161 The observation that seed production in European beech is highly correlated with tem-  
162 peratures around the solstice reinforces the theory that weather cues can serve as a prox-

163 imate mechanism enabling masting plants to synchronize patterns of seed production at  
164 supra-annual and regional scales. The pivotal role of weather variation in driving seed  
165 production in masting plants is not controversial [32]. However, the precise translation of  
166 weather variation into seed production variability remains a topic of debate [29, 49–51].  
167 On one hand, weather conditions may influence the transition from flowers to fruits by, for  
168 example, disrupting pollination success or hindering resource uptake [45, 52, 53]. On the  
169 other hand, gene regulatory networks might integrate various signals, such as temperature  
170 and photoperiod, allowing flowering to occur only when all these cues are received [54].  
171 In the latter scenario, hormones and genes that control flowering exhibit hypersensitivity  
172 to environmental signals, thus granting masting a degree of independence from resource-  
173 and pollen-related mechanisms [29]. The fact that European beech responsiveness to tem-  
174 perature is prominent only as the days begin to shorten suggests that weather's influence  
175 on masting extends beyond mere effects on flower development stages. Rather, it implies  
176 the presence of a strongly conserved regulatory network that enables all plants to respond  
177 uniformly to the cue. Future research should involve monitoring gene expression levels,  
178 such as the FLOWERING LOCUS T, at the fine temporal scales disclosed in this study,  
179 to understand how the negative (two years before seed fall, T2) and positive (in year T1)  
180 correlations with temperature are translated into genetic and hormonal processes [30, 31,  
181 55, 56].

182 Across the entire continent, the summer solstice serves as a celestial "starting gun" that  
183 enables cohesive timekeeping across far-distant beech populations, highlighting how evo-  
184 lution may have capitalized on this astronomical event to maximize ecological synchrony.  
185 A prime fitness benefit of masting lies in predator satiation [20, 32]. The synchronized,  
186 substantial year-to-year variation in seed production effectively starves seed consumers in  
187 low-seeding years, making it easier to satiate these consumer populations in mast years  
188 [57]. High regional synchrony of masting appears to have a selective advantage as it al-  
189 lows to satiate mobile seed consumers, such as highly mobile vertebrates [9, 18], that can  
190 prevent regeneration if mast years are localized [58]. Large-scale synchrony also increases  
191 the likelihood of seed release into disturbed areas [18]. Variation in day length provides a  
192 reliable signal of seasonal shifts and is unaffected by climatic changes in space and time  
193 [59]. To increase masting synchrony under the Moran effect, populations separated by hun-  
194 dreds of kilometers align their cue window with the moment when day length reaches its  
195 longest duration once a year.

196 Anchoring to solstice means that the timing of cue sensing withstands climate variation  
197 and generates a high scale of regional synchrony. However, the climate changes of the  
198 twenty-first century generate novel challenges that may prove the solstice-anchoring sub-  
199 optimal. In European beech, increasing temperatures lead to a decrease in interannual  
200 variation and synchrony of seed production, a process called masting breakdown [60]. The  
201 consequence is inefficient predator satiation and decreased pollination efficiency that to-  
202 gether lead to a collapse in viable seed production [61]. The breakdown follows from higher  
203 summer temperatures and therefore more frequent triggering of reproduction [42]. If sum-

204 mer cues would not be anchored to the solstice, shifting the sensitive periods to earlier in  
205 the season would perhaps compensate for the change in cueing frequency. Second, cli-  
206 mate change leads not only to trends in mean weather conditions but also to changes in  
207 large-scale weather patterns [62]. Long-lasting blocking weather patterns that create ex-  
208 treme conditions over certain regions, appear to change in frequency [62]. For example,  
209 high-pressure areas sandwiched between low-pressure systems called an omega block,  
210 can bring long-lasting heat to the high-pressure area [62, 63]. Transition probabilities to-  
211 wards omega in summer over the Northern Hemisphere increase over time, which may  
212 create repeated periods of masting asynchrony between low- and high-pressure blocks.  
213 Similarly, the among-site variation in correlation values between temperatures and masting  
214 could be associated with omega blocking. The consequences of such disruptions in the  
215 large-scale patterns of synchrony appear as an interesting avenue for future research.

216 Recent discovery indicated that solstice triggers a shift in trees' responsiveness to tem-  
217 perature, enabling plants to anticipate the approaching end of the growing season [39]. Our  
218 study shows that the solstice also serves as a continent-wide "starting gun" to which plants  
219 respond simultaneously from southern France to Sweden, and orchestrates the multiyear-  
220 long process that determines reproductive investment in European beech, generating eco-  
221 logical events marked by the unparalleled spatial synchrony across the continent.

222 **Data availability statement**

223 The data used in this study have been deposited in the Open Science Framework (OSF)  
224 (DOI: [10.17605/OSF.IO/S2CD4](https://doi.org/10.17605/OSF.IO/S2CD4)). The full MASTREE+ dataset is available in [64]. Cli-  
225 mate data have been extracted from E-OBS at [https://cds.climate.copernicus.eu/  
226 cdsapp#!/dataset/insitu-gridded-observations-europe?tab=form](https://cds.climate.copernicus.eu/cdsapp#!/dataset/insitu-gridded-observations-europe?tab=form).

227

228 **Code availability statement**

229 R statistical software v4.3.0 was used in this work [65]. All analyses used published R pack-  
230 ages.

231

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241

242 **Author Contributions Statement**

243 MB conceived the study, MB, VJ, JSz, JF, AHP, DK designed the study. VJ and JSz per-  
244 formed the analysis, MB led the writing of the manuscript. All authors contributed critically  
245 to the interpretation of the analysis and drafts, and gave final approval for publication.

246

247 **Competing Interests Statement**

248 None declared.

249

## Materials and Methods

**Study system and data** European beech (*Fagus sylvatica* L.) is a major forest-forming species in temperate Europe. Beech is a model masting species, with seed production characterized by large interannual variation and synchrony of seed production [16, 22, 66]. Beech masting allows to escape seed predation and increases pollination efficiency [60, 66]. Subsequent cold (two years before seed fall) and hot (one year before seed fall) summers trigger large seed production in European beech [33, 44].

**Seed production and environmental data** Annual observations of seed production of European beech were extracted from MASTREE+, an open-access database of annual records of population-level reproductive effort [64]. For our analysis, we restricted the European beech time series to the continuous observations of seed production that covered more than 14 years, observed after 1952. We excluded pollen-based and ordinal records. We used a conservative, relatively high number of years per time series, to ensure that enough mast years had a chance to happen and optimize the signal-to-noise ratio. The cut-off in 1952 was motivated by limited climatic data availability before that year. That resulted in 61 time series available for the analysis (averaged length time series, 35 years; maximum length time series, 68 years). The number of years per time series is given in Fig. S1. We extracted daily climate data for each site from the corresponding 0.1° grid cell of the E-OBS dataset [67].

The day length for each location was calculated as

$$\cos \omega_0 = -\tan\left(\frac{\phi \times \pi}{180}\right) \times \tan \delta \quad (1)$$

where  $\omega_0$  is the solar hour angle,  $\phi$  is latitude and  $\delta$  is declination. Declination of the sun can be obtained with

$$\delta = \frac{\pi \times 23.45}{180} \times \sin\left(\frac{2 \times \pi \times (284 + \text{DOY})}{365}\right) \quad (2)$$

where DOY is Day of the Year. Day length in hours was obtained with

$$\text{DL} = \begin{cases} 0 & \omega_0 > 1 \\ 24 & \omega_0 < -1 \\ \frac{24}{\pi \times \text{acos}(\omega_0)} & \text{otherwise} \end{cases} \quad (3)$$

For each site, the relative day length was normalized to the 0 - 1 range.

**Moving window correlation** To determine how correlations between mast seeding and temperature fluctuate at a fine temporal scale, we ran a moving window correlation analysis. Specifically, we ran a moving Spearman correlation between log-transformed annual seed

277 production and mean daily temperature. The window size was set as 7 days, with a 1-day  
 278 step. The moving window size was set relatively small to ensure the detection of fine-  
 279 scale temporal changes in the correlation between masting and temperature. Correlations  
 280 were run over year T2 (two years before seed fall) and T1 (one year before seed fall), as  
 281 we expected the June-July summer cues [33, 44] to be fine-tuned to the summer solstice.  
 282 Moving window correlations were run for each site separately. We computed the standard  
 283 error of the Spearman correlation ( $cor_{se}$ ) as

$$cor_{se} = \sqrt{(1 - cor^2)^2 \times \frac{(1 + cor^2/2)}{(n - 3)}} \quad (4)$$

284 with  $cor$  being the estimated Spearman correlation value and  $n$  the sample size [68].

285 **Cue window and the solstice** To test whether the change in Spearman correlations be-  
 286 tween masting and temperature changes abruptly at solstice we used generalized additive  
 287 models (GAM). We included the correlation coefficients as a response, while the relative day  
 288 length, the summer solstice as a categorical variable (before or after the solstice), and their  
 289 interaction were used as predictors. If the cue window opens at the solstice, we expected  
 290 the interaction to be significant, with the correlation abruptly increasing at the solstice. We  
 291 fitted GAMs with a beta distribution family, by using the `mgcv` package (v1.8-42, [69]), and  
 292 specified the restricted maximum likelihood (REML) method to estimate smoothing param-  
 293 eters. We fitted models using the beta distribution because fitting a model with a Gaussian  
 294 distribution of errors, given our response type, would have been inappropriate and result  
 295 in biased parameter estimates. Because the beta distribution requires the response to be  
 296 bounded between 0 and 1, we rescaled the Spearman correlations to 0-1 range. The scal-  
 297 ing was done for each site separately and is a simple mathematical transformation that does  
 298 not affect the distribution of a variable. We used the inverse of the standard error of corre-  
 299 lation coefficients (i.e.  $1/cor_{se}$ ) as weights in the GAM models, to ensure that observations  
 300 with higher error had lower weight on parameter estimates [70]. The site was included as a  
 301 random intercept. We tested for spatial autocorrelation of model residuals using the `DHARMA`  
 302 package (v0.4.6) [71], and detected none. The test was run separately for correlations in  
 303 one (T1) and two (T2) years before seed fall.

304 As an additional test, we also binned the moving window correlations between masting  
 305 and temperature changes into two categories, i.e. before and after the solstice. For each  
 306 bin, we used a number of days that included 5% of change in day length at each side of  
 307 the solstice (24 days). We tested for the difference between the period (i.e. before/after the  
 308 solstice) using the GAM model with correlations in these bins as a response, while including  
 309 group (before/after the solstice) as a predictor fitted as a parametric term.

310 Furthermore, we used a null model to explicitly test whether the observed increase in the  
 311 values of correlation coefficients between masting and temperature just after solstice could  
 312 emerge by chance also if any other day of the year was considered instead of the solstice.

313 For each day of the year between 25th January and 6th December we compared correlation  
314 coefficients for 24 days before and 24 days after that day using analogous GAM models as  
315 described above. From these models, we extracted the beta coefficients for the “after”  
316 category, showing how correlation coefficients differ as compared to the “before” category.  
317 This procedure resulted in a distribution of beta coefficients under the null hypothesis i.e.  
318 the before-after difference in correlation coefficients if any other day of the year would have  
319 been considered instead of the solstice. Then, to test for the significance of the solstice  
320 effect (i.e. whether the increase in the values of correlation coefficients we observed at the  
321 solstice is more probable than given purely by chance), we calculated the p-value as the  
322 number of times an absolute value of randomized beta coefficients was greater than, or  
323 equal to, the absolute value of a beta coefficient from GAM model comparing correlations  
324 binned around the solstice, divided by the number of tests.

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## Supplementary material

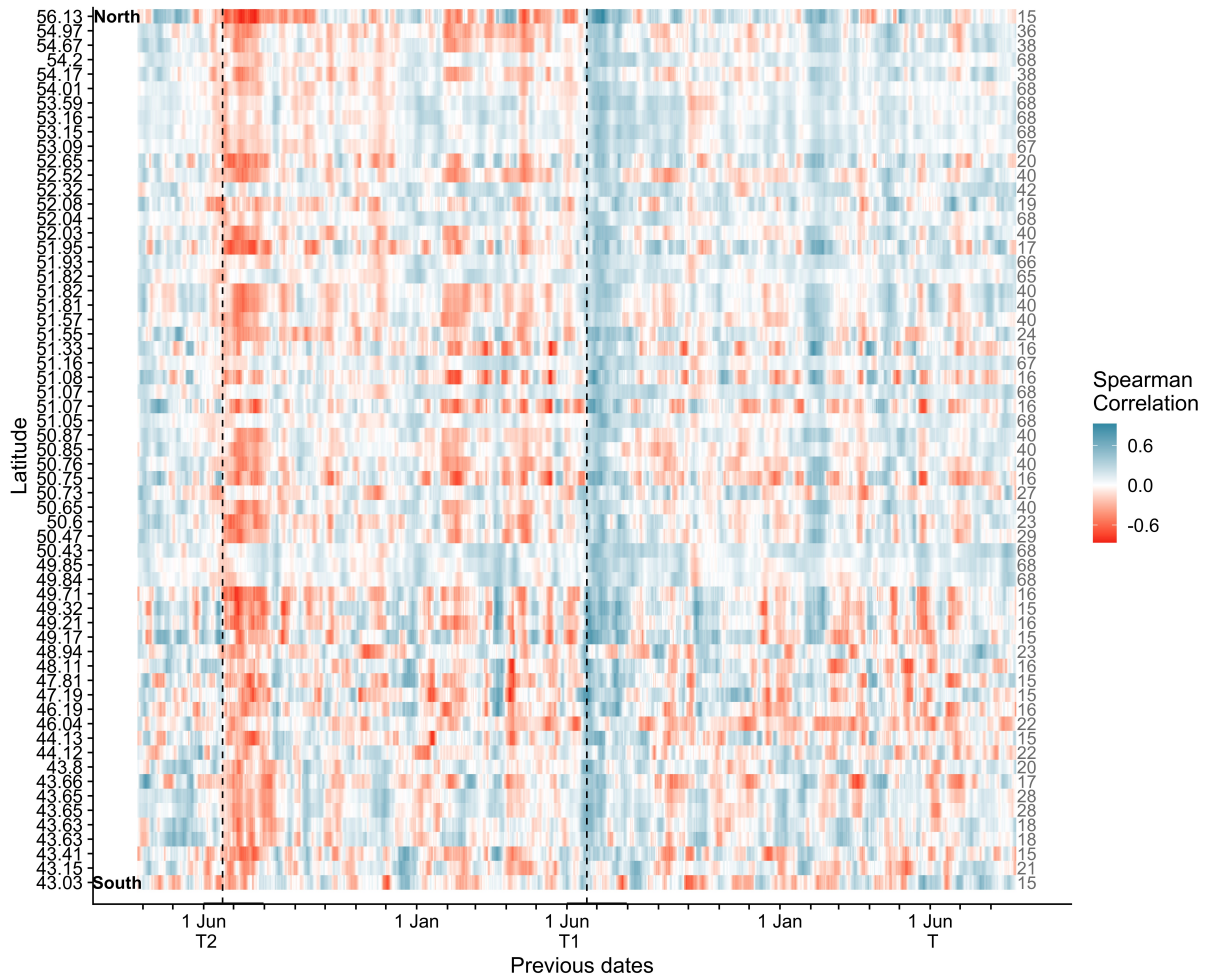
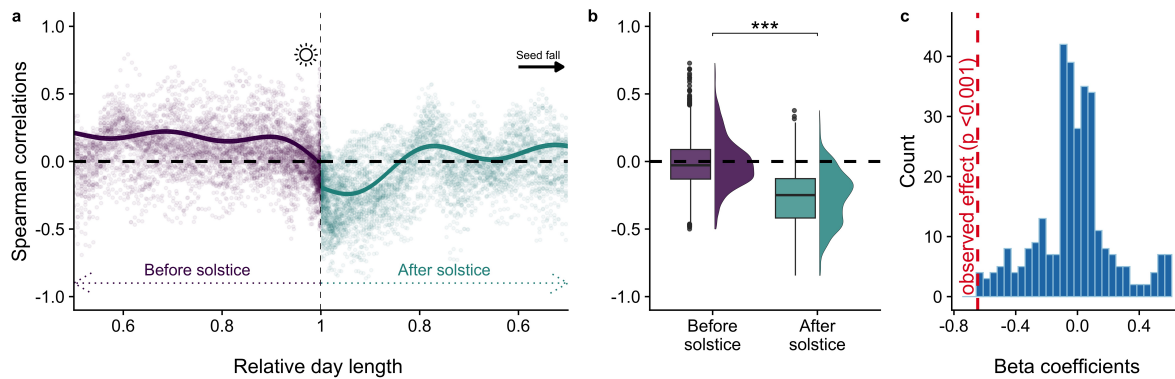
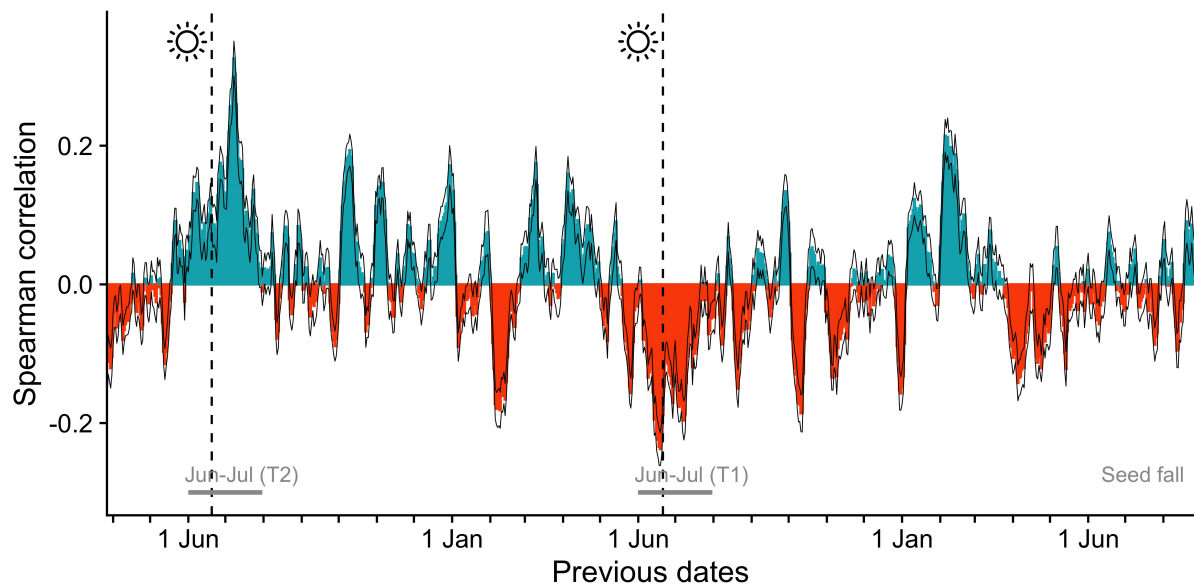


Figure S1: Moving window Spearman correlations between temperature and masting across 61 sites distributed over the European beech range. Black dashed lines indicate the summer solstice (21 June) in each year (one and two years prior to seed fall). Each line represents one site sorted by its latitude (43.03° - 56.13°). The number on the right side of the plot (grey color) indicates site-level sample size (monitoring duration in years). The observations are from the period 1952 - 2019.



**Figure S2:** a) Correlations between European beech seed production and temperature at 7-day windows abruptly decrease just after the solstice in year T2. The purple color shows correlations and GAM model prediction for days before the summer solstice and green after the solstice. The prediction lines are based on the GAM model with Spearman correlation coefficients as a response and interaction between relative day length (at the particular time window) and the summer solstice as predictors. Each point is the per site per window correlation coefficient. The predicted correlation at maximum day length is at -0.013 before the solstice and -0.19 after the solstice. The model is summarized in Table S1. b) Correlations before and after the solstice binned into 0.95-1 (maximum) day length before and after the solstice. The \*\*\* indicates significance difference between bins tested with GAM ( $p < 0.0001$ ). The median correlation before the solstice is -0.028 and the median correlation after the solstice is -0.25. c) Results of a null model testing whether the abrupt increase in the correlation coefficients (between masting and summer temperatures) is highest at the summer solstice or any other day in the year. Histogram shows the distribution of beta coefficients under the null hypothesis i.e. the before-after difference in correlation coefficients in any other day of the year considered instead of the solstice. The red dashed line shows the observed effect i.e., how correlation coefficients “after solstice” differ as compared to “before solstice”. The null model was rejected ( $p < 0.001$ ). All three graphs show the effects for summer T2 (two years before seedfall), effects for summer T1 are presented in Fig. 2.



**Figure S3:** Mean rolling Spearman correlations between precipitation and masting averaged across all 61 sites. The graph shows correlations in two (T2) and one (T1) years before seed production, up until September when seed fall happens. The size of the precipitation window is 7 days, with a 1-day step, and correlations are plotted according to the day of the year at the end of each 7-day window. Black dashed lines close to the sun icon indicate the summer solstice (21st June). Correlations are coded blue for positive, and red for negative. The black solid lines represent the standard error of the correlation coefficients across the sites for each window. Correlations for each site separately are reported in Figure S4.

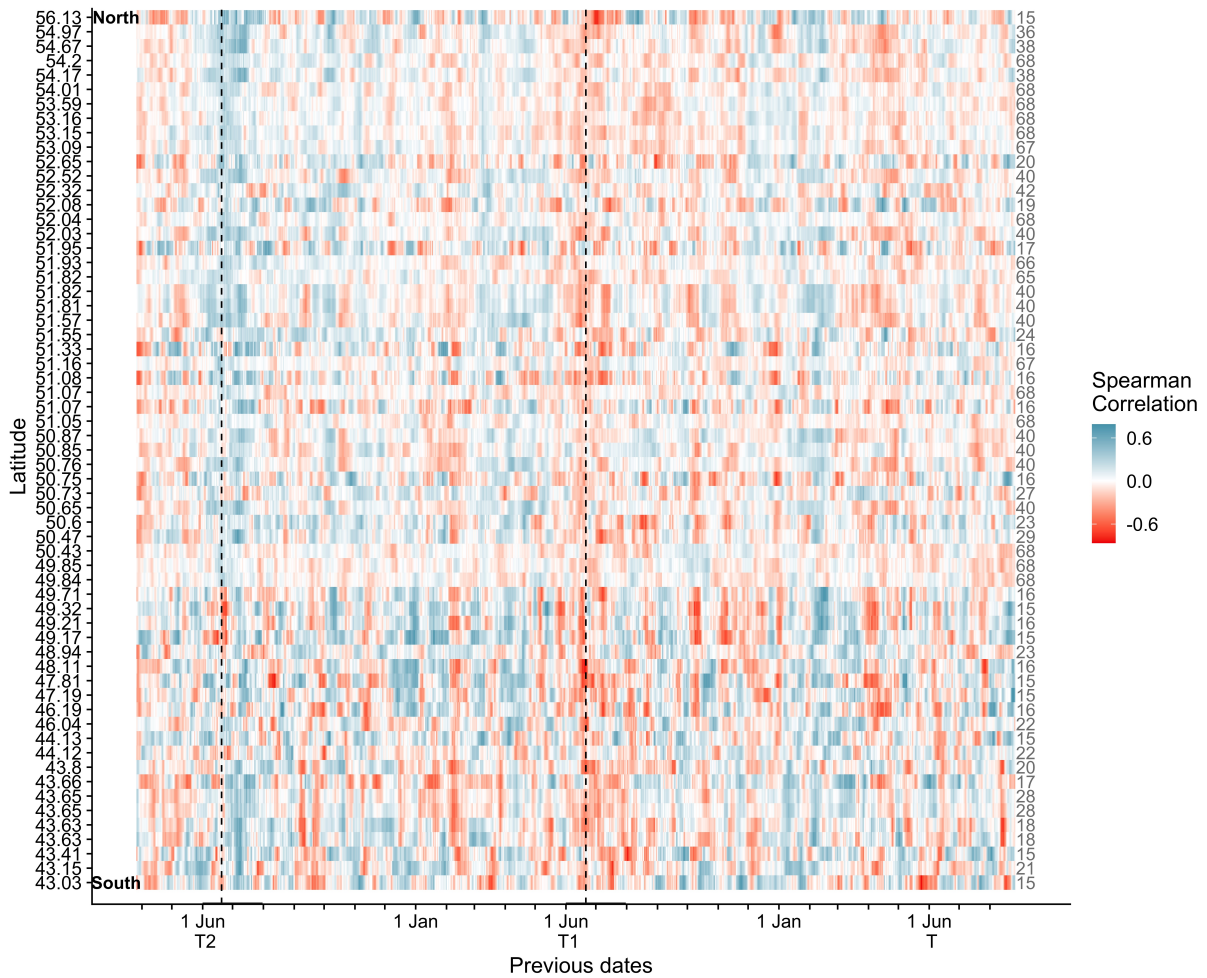
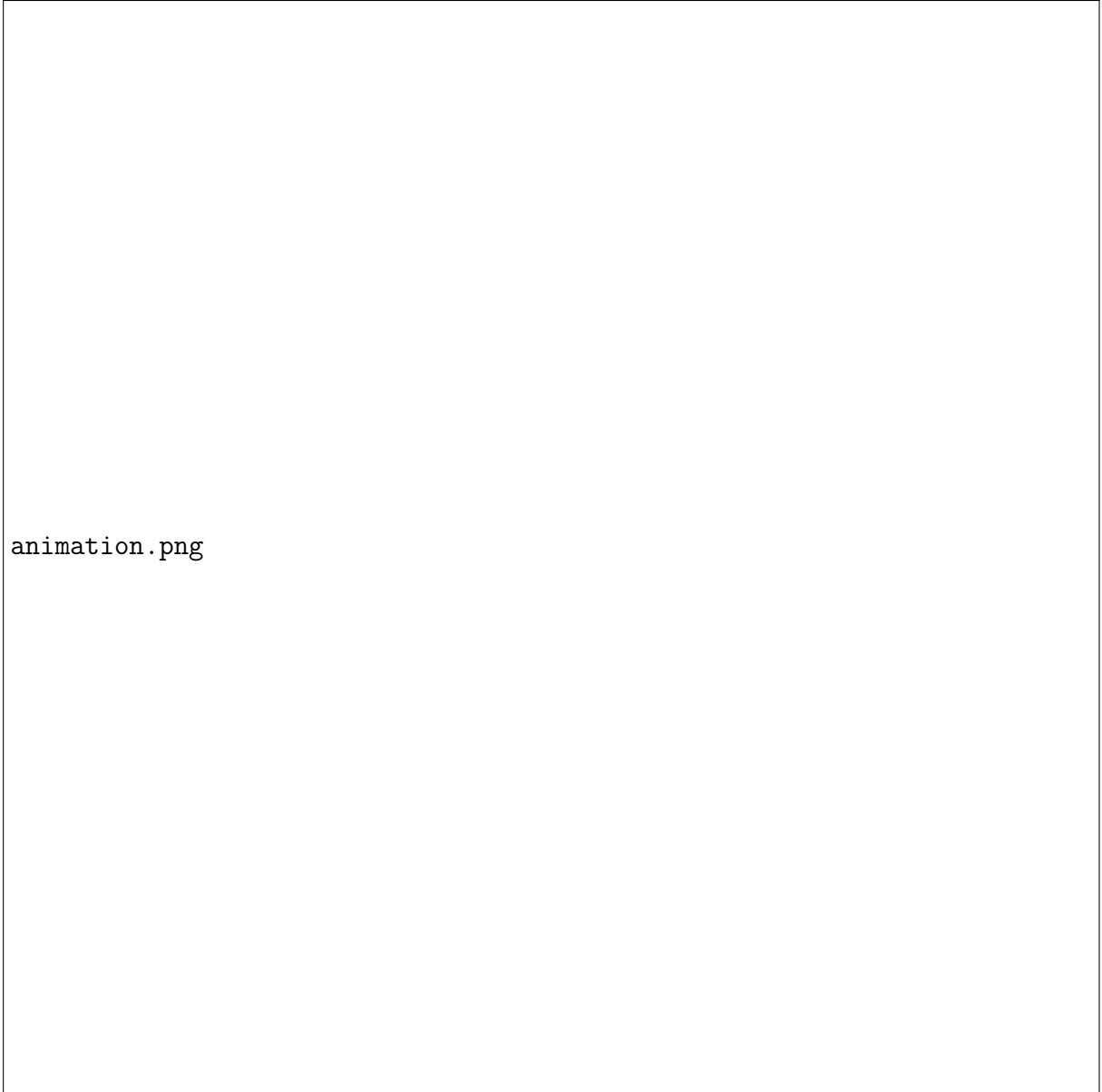


Figure S4: Moving window Spearman correlation between precipitation and masting across 61 sites distributed over the European beech range. Black dashed lines indicate the summer solstice (21 June) in each year (one and two years prior to seed fall). Each line represents one site sorted by its latitude (43.03° - 56.13°). The number on the right side of the plot (grey color) indicates site-level sample size (monitoring duration in years). The observations are from the period 1952 - 2019.



animation.png

Video: The animation showing the scale of synchrony is provided as a video file. Video shows annual seed production for each site between the years 1995 and 2019, scaled within each site to values between 0 and 1 (to facilitate among-site comparisons). 1995 is shown as the first year as the spatial coverage at that period is highest. Point size is scaled to the size of annual seed production, while the color shows the within-year synchrony among sites. That synchrony is calculated as the inverse coefficient of variation.

Table S1: Summary of generalized additive models (GAM) smooth terms between Spearman correlation coefficient and relative day length for one (T1) and two (T2) years prior seed fall in European beech. Effective degrees of freedom (EDF) represent the complexity of the smooth. Chi-square values indicate the test of significance of smooth terms.

| Smooth term     | EDF   | Chi-squared | p        |
|-----------------|-------|-------------|----------|
| A) Year T1      |       |             |          |
| Before solstice | 8.99  | 12.045      | < 0.0001 |
| After solstice  | 8.95  | 30.095      | < 0.0001 |
| Site            | 59.83 | 24.134      | < 0.0001 |
| B) Year T2      |       |             |          |
| Before solstice | 8.99  | 7.794       | < 0.0001 |
| After solstice  | 8.99  | 26.403      | < 0.0001 |
| Site            | 59.46 | 5.999       | < 0.0001 |