Short reproductive periods dominate mast seeding across diverse tree species

Katarzyna Kondrat*¹, Jakub Szymkowiak^{1,2}, Andrew Hacket-Pain³, Mitsue Shibata⁴, Takashi
 Saitoh⁵, Jessie Foest¹, Davide Ascoli⁶, Michał Bogdziewicz¹

6

3

7

⁸ ¹Forest Biology Center, Institute of Environmental Biology, Faculty of Biology, Adam Mickiewicz Uni-

9 versity, Uniwersytetu Poznańskiego 6, 61-614 Poznan, Poland

¹⁰ ²Population Ecology Research Unit, Institute of Environmental Biology, Faculty of Biology, Adam Mick-

¹¹ iewicz University, Uniwersytetu Poznańskiego 6, 61-614 Poznan, Poland

¹² ³Department of Geography and Planning, School of Environmental Sciences, University of Liverpool,

13 Liverpool, UK

¹⁴ ⁴Forestry and Forest Products Research Institute, Tsukuba, Japan

¹⁵ ⁵ Field Science Center, Hokkaido University, North-11, West-10, Sapporo 060-0811, Japan

¹⁶ ⁶Department of Agriculture, Forest and Food Sciences, University of Torino, Turin, Italy

17

¹⁸ *Corresponding author: katarzyna.kondrat@amu.edu.pl;

Abstract

Mast seeding, synchronous and highly variable reproduction among perennial plants, profoundly 20 impacts ecosystem dynamics and species interactions. However, the extent of periodicity 21 in mast seeding, defined as cyclical but not strictly regular intervals between reproduction, 22 remains poorly understood, including how it varies across and within species. Here, we used 23 autoregressive analyses on seed production data from 556 populations across 20 tree species, 24 with an average time series length of 19 years (range: 10–62 years), to quantify the prevalence 25 and length of masting periods. We found widespread periodicity, predominantly characterised 26 by a short period of 2–3 years. Although periodicity was common, the signal was often weak, 27 indicating limited explanatory and predictive power. This period length aligns with theoretical 28 predictions that balance the ecological benefits of predator satiation, effective mainly against 29 specialized short-lived insect predators, with costs such as missed reproductive opportunities 30 and resource losses. Extended period lengths (>4 years) were uncommon (2%), suggesting that 31 longer periods may be less ecologically advantageous or subject to specific local conditions. 32 Climate and elevation have limited and species-specific effects on period length and strength, 33 implying local adaptation in cue sensitivity and resource accumulation. Our findings emphasise 34 the adaptive value of short reproductive periods in mast seeding, likely reflecting consistent 35 evolutionary constraints on reproductive timing across diverse ecological conditions. 36

keywords: | masting | masting period | seed production | periodicity | resource pulses
 ³⁸

39 Introduction

Mast seeding, or masting, is synchronous and highly variable reproduction among years by a 40 population of perennial plants (Kelly, 1994; Bogdziewicz et al., 2024). The seed pulses that result 41 from masting influence plant recruitment, demographic processes, and ecosystem dynamics, 42 including effects on seed consumers, their predators, and associated parasites (Ostfeld et al., 43 2000; Hacket-Pain et al., 2022; Seget et al., 2022), as well as nutrient cycling and abundance 44 of mycorrhizal fungi (Müller-Haubold et al., 2015; Michaud et al., 2024). Consequently, 45 understanding whether masting follows predictable periods has long interested ecologists and 46 foresters (Elton, 1924), but periods have been difficult to test largely due to analytical constraints 47 (Bogdziewicz et al., 2023). Exceptions include Allen et al. (2012), who applied ordinal time 48 series analysis and detected a 7-year periodicity in seed production of New Zealand mountain 49 beech (Nothofagus solandri), and Shibata et al. (2020), who used autoregressive models to show 50 that warming reduced the period length in masting of Japanese oak (Quercus crispula) from 3-4 51 years to just 2 years. These studies show that, despite the inherent difficulties, it is possible to 52 detect masting periods using a range of statistical approaches, which have also revealed potential 53

links with climate. However, a broad-scale analysis comparing periodicity across populations
 of multiple species and climates remains limited.

Prior research has mostly centred on a related concept of masting return intervals, which 56 refers to the average time between large seed production events. Return intervals provide a 57 descriptive summary of the spacing between mast years but do not capture the underlying 58 temporal dynamics of reproduction. Return interval studies have often reported longer masting 59 periods, sometimes exceeding 5-10 years, particularly in tropical and temperate tree species. 60 For example, return intervals for three North American oaks varied from 2 to 4 years, (Sork et al., 61 1993), while in tropical forests, (Igarashi et al., 2024) reported 2 to 10 year return intervals in 18 62 dipterocarp species. In four temperate forest species, (Nussbaumer et al., 2016) reported masting 63 frequency (the reciprocal of return interval) ranging from 1.64 to 10 years. Some studies report 64 ranges of typical masting return intervals, noting that return intervals are irregular over time 65 (Övergaard et al., 2007; Broome et al., 2007; Wagner et al., 2010). Moreover, numerous manuals 66 and monographs used by foresters and wildlife managers include tables of species-specific 67 masting return intervals (Burns et al., 1990; Young & Young, 1992), because such information 68 can optimise harvesting schedules, guide forest restoration, and support wildlife conservation 69 dependent on mast resources (Kettle et al., 2010; Köhnke et al., 2020; Bregnard et al., 2021). 70 While the identification of return intervals may be of practical value, such assessments inevitably 71 depend on the selected threshold defining a mast year (Bogdziewicz et al., 2024). Because return 72 intervals depend on arbitrary thresholds for defining mast years, they cannot reliably capture 73 variation in masting behaviour across species or environmental gradients, including elevation. 74 Here, we return to the original concept of masting periods, analysing autoregressive patterns 75 (Bjørnstad et al., 2008; Shibata et al., 2020) in seed production across 556 populations of 20 tree 76 species, offering a broad-scale evaluation of periods across and within species. Understanding 77 these periods provides insight into temporal patterns of seed production, with implications for the 78 evolutionary drivers of masting and broader ecosystem dynamics, such as resource availability 79 and consumer-producer interactions. 80

From an evolutionary perspective, the period reflects the degree of reproductive delay, 81 shaped by the cost-benefit balance of interannual variation in seed production (Bogdziewicz 82 et al., 2024). Major costs include missed reproductive opportunities, which can reduce popula-83 tion growth rates (Vacchiano et al., 2021), and increased density-dependent seedling mortality 84 associated with concentrating reproduction in large, intermittent events (Visser et al., 2011; 85 Huang et al., 2021; Seget et al., 2022). These constraints select against prolonged periods of 86 reproductive delay (Bogdziewicz et al., 2024), particularly in environments with low produc-87 tivity and high background mortality, where the risks of delaying reproduction may be greater 88 (Waller, 1979). In contrast, two major benefits, known as economies of scale, can favour de-89 layed reproduction (Bogdziewicz et al., 2024). First, alternating between years of low and high 90 seed production allows plants to starve and then overwhelm scarce seed consumers, thereby 91 reducing seed predation rates (Zwolak et al., 2022). This mechanism is particularly effective 92

against consumers with low mobility, high dietary specialization, and short lifespans, such as 93 many insect species (Kelly & Sork, 2002; Zwolak et al., 2022). For example, populations of 94 seed-predating micromoths are often highly vulnerable to even a single year of seed scarcity 95 (Yasaka et al., 2003; Żywiec et al., 2013). Secondly, large and synchronised flowering enhances 96 pollination success by increasing floral density (Kelly et al., 2001; Rapp et al., 2013; Venner 97 et al., 2016). While reproductive delay and interannual variation in flowering do not directly 98 improve pollination rates, they allow resource accumulation that helps populations exceed the 99 flowering threshold needed for efficient pollination (Kelly et al., 2001). Where species or popu-100 lations cannot maintain high flowering effort every year, selection may favour delayed flowering 101 that enables resource build-up (Kelly et al., 2001; Bogdziewicz et al., 2020; Kelly, 2020). 102

From a proximate perspective, periodicity arises from interactions between resource dynam-103 ics and weather cues (Satake & Bjørnstad, 2008; Kelly et al., 2024). Resource budget models 104 propose that plants must accumulate sufficient resources before high seed production occurs 105 (Crone & Rapp, 2014). Then, high seed production occurs when adequate resources align with 106 favourable weather cues, such as - in the case of species inhabiting boreal and temperate regions 107 - warm conditions in preceding years (Bisi et al., 2016; Nussbaumer et al., 2018). Because 108 plant response to the weather cue depends on the levels of accumulated resources, resource 109 dynamics play the role of both promoter and suppressor of reproduction, enabling plants to 110 maintain periodic reproduction despite variability in cue frequency (Monks et al., 2016; Kelly 111 et al., 2024). Specifically, low resource levels can suppress reproduction even when strong cues 112 occur consecutively, thereby preventing successive high-seeding years and reducing the risk of 113 seed overexploitation by consumer populations (Kelly et al., 2000, 2013, 2024). Conversely, 114 high accumulated resource levels enhance plant sensitivity to weather cues, allowing even mod-115 erate cues to trigger large seed production, thus preventing excessively delayed reproductive 116 episodes (Kelly et al., 2024). These processes can stabilise the periodicity of masting, opti-117 mising intervals to maximise fitness benefits. However, environmental conditions influencing 118 resource accumulation and weather cue frequency introduce stochastic variation into these inter-119 vals. Consequently, quantitative assessments of periodicity across multiple species are required 120 to better understand how resource dynamics and cue frequency interact at broader ecological 121 scales. 122

Here, building on earlier methods (Bjørnstad et al., 2008; Shibata et al., 2020), we apply 123 second-order autoregressive models to quantify periodicity and evaluate how period length varies 124 across species and ecological contexts. Additionally, we examine how local climate correlates 125 with period length. To the extent that interaction between resource dynamics and weather cue 126 frequency affects masting patterns, we expected period length to correlate with local climate and 127 elevation. For example, within species, populations inhibiting harsher (e.g. drier and colder) 128 climates or higher elevations may require more time to accumulate sufficient resource levels 129 to trigger large reproductive events, prolonging the period length (Satake & Bjørnstad, 2008; 130 Wion et al., 2020; Foest et al., 2025a) Importantly, our focus on periodicity does not imply 131

that masting events occur at strictly regular or predictable intervals. Aside from the two-year
"alternate bearing" observed in a few species (Garcia *et al.*, 2021), there is little reason to expect
strict periodicity in masting time series. Instead, periodicity arises from interactions between
stochastic weather cues and internal plant resource dynamics, with period length shaped by
selection processes.

¹³⁷ Materials and Methods

Seed production data We obtained data from MASTREE+, a database that records annual, population-level records in perennial plants' reproductive effort (Hacket-Pain *et al.*, 2022; Foest *et al.*, 2024). The species selected for analysis were those that met the following criteria: data were recorded on a continuous scale and included counts of seeds, fruits, or cones; data spanned a minimum of ten distinct locations, each representing either a stand or a patch (excluding regional-scale records); and each time series comprised at least ten years of data.

The final dataset included 20 species, encompassing 556 unique time series (Table S1). Elevation data were included for species with available information and sufficient variation, defined as multiple elevation points where differences in elevation across populations exceeded 100 meters. This threshold was chosen to ensure meaningful differentiation in elevation values, enabling the analysis to capture changes in climatic conditions. The dataset with elevation covered 10 species and 141 populations.

¹⁵⁰ **Climate data** Mean annual temperature (MAT) and mean annual precipitation (MAP) were ¹⁵¹ calculated for each unique location, based on monthly values (1960 to 2020) from the corre-¹⁵² sponding 2.5 minute resolution in the WorldClim dataset (Harris *et al.*, 2020).

Analysis To estimate the period length, we calculated the second-order autoregression (AR2) coefficients that relate current seed production (*t*) to past observations. The coefficients $1 + a_1$ and a_2 quantify the dependence of seed production in year *t* on seed production in the previous one and two years earlier, respectively. In this model, population dynamics are considered periodic if the combinations of $1 + a_1$ and a_2 fall within the parabola or to the left of the $1 + a_1$ = 0 line in the plot (see Fig. 1) (Bjørnstad *et al.*, 2008).

There are two primary ways to generate gradients in period length within the AR2 model (Bjørnstad *et al.*, 2008; Shibata *et al.*, 2020). First, increasing the $1 + a_1$ coefficient, provided the a_2 coefficient remains in the period region, elongates the period length. The period length is determined by the position of points in the parabola diagram (Fig. 1). Points between lines 2 and 2, as well as 2 and 3, indicate a two-year period; between 3 and 4, a three-year period and so on (Bjørnstad *et al.*, 1995, 2008; Cornulier *et al.*, 2013; Ahrestani *et al.*, 2016; Shibata *et al.*, 2020). Secondly, increasing a_2 , while $1 + a_1$ remains negative, shifts points leftward within the periodic region of the parabola diagram (Fig. 1), typically shortening the estimated period

¹⁶⁷ length from 3–4 years to 2 years.

Points falling outside the triangle or falling inside the top right side of the triangle above the parabola region (i.e. the white space), are considered as non-periodic (Fig. 1).



Figure 1: Parabola plot showing correspondence of combinations of $1 + a_1$ and a_2 coefficients values to period length. White space indicates a lack of period. See Bjørnstad *et al.* (1995, 2008).

We used Bayesian second-order autoregressive models to analyse the relationship between 170 climate and $1 + a_1$ and a_2 coefficients for each species separately. Before modelling, we 171 standardised MAT and MAP by centring and scaling (z-transformation), to place them on 172 comparable scales. Given the correlation between these variables - evidenced by high VIF 173 values in some species — we fitted separate models for MAT and MAP for each species. The 174 model formulas were specified using the bf() function from the brms package (Bürkner, 2021), 175 where the autoregressive coefficients $1 + a_1$ and a_2 were modelled as functions of standardised 176 MAT and MAP. The models were fitted using the brm() function from the brms package (Bürkner, 177 2021), which supports multivariate response modelling and allowed both autoregressive terms 178 to be estimated simultaneously. We used the default priors provided by the package. Each model 179 was run with four MCMC chains, each consisting of 10,000 iterations, including 3,000 warm-up 180 iterations. Model fitting was performed in parallel using 8 CPU cores. All data analysis was 181 performed in R v. 4.2.3 (Team, 2020). 182

Results

Percentage of periodical populations Based on temporal autoregression coefficients, almost all populations (N = 525 out of 556) across the 20 analysed species exhibited periodic masting behaviour, with an overall prevalence of periodicity of approximately ~95%. For eight species, all populations were periodic, including four conifers (*Picea engelmannii, Araucaria araucana, Abies amabilis, Abies alba*) and four deciduous species (*Ouercus douglasii, Ouercus cerris*,

189 *Fagus sylvatica*, *Betula pubescens*) (Fig. 2).

For the species that included both periodic and non-periodic time series, the proportion of periodic populations ranged from 81.8% (*Pinus palustris*) to 96.7% (*Quercus petraea*). Other species with relatively lower percentages of periodic populations included *Alnus incana*, *Quercus robur* (90%), *Pinus sylvestris* (88%), *Fagus crenata* (87%) and *Pinus palustris* (82%) (Fig. 2).



Figure 2: Percentage of periodic populations across species studied (N = 525 periodic populations, 20 species). The x-axis shows the percentage of periodic populations for each species. Whether a time series was categorised as periodic or not is based on whether a_1 and a_2 coefficients fell within the periodic region of the parabola plot (Fig. 1). The minimum time series length for each population was 10 years, with an average of 20 years (Table S1).

The average period length across species was 2.65 (± 0.49 SD) (Fig. 3), and Period length 194 was not correlated with the length of time series (Fig. S1). Note that even long time-series 195 >40 years did not show differences in mean period length, although it was notable that longer 196 period lengths were restricted to relatively short time-series (Figure S1). Among the species 197 analysed, the shortest species-level average period length was observed in *Quercus lobata* (2 ± 0) 198 SD), which displayed a highly consistent period length (all 11 populations had period = 2). 199 In contrast, the longest average period length was recorded in *Quercus cerris* (3.27 ± 0.14) 200 SD), also with relatively low variability. Most species exhibited average period lengths closely 201

centred around the overall mean, corresponding to a two- to three-year period. We found little
evidence of longer period length, with only 10 time-series showing evidence of periods longer
than 4 years, mainly associated with *Pinus species*, *Picea abies* and *Quercus petraea* (Fig. 3). *Quercus petraea* and *Picea abies* included outlier populations with 8-year and 7-year periods,
respectively. Additionally, few populations with 5-year periods were observed in coniferous
species from the *Pinus*, *Picea*, and *Alnus* genera (Fig. 3).



Figure 3: Inter- and intra-specific variation in period length (N = 525 periodic populations, 20 species). Each row represents the distribution of period lengths for populations of each species. Period length for each population was determined based on the position of the population $1 + a_1$ and a_2 coefficients on the parabola plot (Fig. 1). Numbers in brackets indicate the number of studied populations. Each point corresponds to a single population. All presented values are integers, the jitter was added to decrease overlap. Red vertical lines represent the species-level mean period length. Non-periodic populations are excluded from the graph. The $1 + a_1$ and a_2 coefficients for each population are given in Fig. S2.

Period length and climate While some species show an effect of local climate on their 208 autoregressive coefficients ($1+a_1$ and a_2), many do not exhibit a strong or consistent relationship. 209 The diversity of slope directions and significance levels indicate that the effect of climate on 210 these autoregressive coefficients is species-specific. Higher mean annual temperatures (MAT) 211 increased period length in two species, A. araucana and P. abies, as indicated by positive effect 212 of MAT on $1 + a_1$. In turn, higher MAT reduced period length in three species, Q. petraea, P. 213 engelmannii, and A. incana, as indicated by negative effects on $1 + a_1$ (Fig. 4A). Moreover, in 214 five species— Q. petraea, P. edulis, P. albicaulis, F. sylvatica, and B. pubescens—the periods 215 tended to be shorter, with species more likely to exhibit a two-year period in warmer locations, 216 as indicated by positive effects on a₂. In contrast, in *P. engelmannii*, the period was longer in 217 warmer climates (Fig. 4A, Fig.S2). 218

Period length increased with higher mean annual precipitation in four species - P. albicaulis, 219 F. sylvatica, F. crenata, and A. araucana, as indicated by positive effects on $1 + a_1$. In turn, 220 period length increased in drier locations in six species, Q. petraea, Q. lobata, P. palustris, P. 221 engelmannii, P. glauca, and A. incana, as indicated by negative effects on $1+a_1$ (Fig. 4B). Periods 222 became shorter and more likely to transition to a two-year period with increasing moisture in 223 two species, *P. edulis* and *B. pubescens*, as indicated by positive effect on a_2 . In contrast, in 224 F. sylvatica, where precipitation has a negative effect on a_2 , periods lengthened with higher 225 moisture, increasing the likelihood of a shift to a three-year period (Fig. 4B, S2). 226 For the remaining species, Q. cerris, P. sylvestris, A. amabilis, Q. robur, P. ponderosa, 227

Q. douglasii, and *A.alba*, climate variables had no significant effect on the length or strength of
 their periods.



Figure 4: Relationship between autoregressive coefficients $1 + a_1$ and a_2 and local climate (MAT: mean annual temperature, and MAP: mean annual precipitation) (N = 525 populations, 20 species). Panel A shows effects of MAT, while panel B of MAP. Rows represent separate species. Each point represents the slope of the relationship, with error bars indicating the 95% confidence intervals. Coefficients $1 + a_1$ and a_2 reflect the relationship between MAT or MAP and period length. Non-significant effects are indicated by lower opacity.

Period length and elevation The correlations of elevation on autoregressive coefficients $1+a_1$ 230 and a_2 were weak and largely not significant (Fig. S3). However, four species showed significant 231 effects. In *P. glauca*, $1 + a_1$ and a_2 decreased with elevation, which, given the parameter space 232 covered by this coefficient in that species, indicated a tendency of period length to increase at 233 higher altitudes. Similarly, in F. sylvatica and P. engelmannii, $1 + a_1$ increased with elevation, 234 suggesting a tendency for longer period length. In A. alba, a₂ decreased with elevation, 235 suggesting that period length in that species tends to move from 2-year to 3 to 4-year period as 236 elevation increases (Fig. S3). In other species, the effects of elevation on and $1 + a_1$ and a_2 were 237 not significant (Fig. S3). 238

239 Discussion

Our findings reveal that periodicity in masting is widespread across diverse tree taxa. Specif-240 ically, the 20 species-including angiosperms and gymnosperms from boreal, temperate, and 241 Mediterranean biomes-consistently fell into parameter space that indicated periodicity, with 242 a relatively short mean period length of 2-3 years. We interpret this as evidence that com-243 mon ecological factors, most likely the economies of scale and the inherent costs of delayed 244 reproduction, play key roles in governing masting behaviour across broad phylogenetic and 245 environmental settings. This suggests that the selective pressures favouring reproductive delay 246 converge to a similar outcome across species. 247

The mean period length of 2-3 years fits well with predictions based on economies of 248 scale. Interannual variation in seed production can effectively reduce seed predation by insects, 249 whose short life spans mean that a delay of just 1-2 years is sufficient to lock them into periods 250 of starvation and satiation (Zwolak et al., 2022). Prolonging the delay would increase the 251 costs, such as missed reproductive opportunities (Rees et al., 2002; Visser et al., 2011; Tachiki 252 & Iwasa, 2010), possibly with little additional benefit to reproductive success. In contrast, 253 extending the period length specifically to starve and satiate consumers with longer lifespans, 254 such as generalist vertebrates, is unlikely to be effective. These animals are typically mobile and 255 capable of exploiting alternative resources or migrating to areas of higher seed availability during 256 periods of scarcity, making predator satiation less viable as a strategy against them (Curran & 257 Leighton, 2000; Bogdziewicz et al., 2022; Zwolak et al., 2022). Similarly, while accumulating 258 resources to reach a critical flower density is essential for successful pollination (Kelly et al., 259 2001), extending the delay beyond what is necessary incurs further risks, including losses of 260 accumulated resources due to adverse events like insect outbreaks or drought. Recent findings 261 further suggest that the interaction between resource levels and cue strength allows plants to 262 fine-tune this delay, providing a proximate mechanism to maintain the period length at the 263 desired level (Monks et al., 2016; Ascoli et al., 2017; Kelly et al., 2024). Importantly, reliance 264 on stochastic weather cues remains crucial, as regular periods could enable seed predators to 265 predict and exploit seed pulses, for example, through diapause (Maeto & Ozaki, 2003; Pélisson 266 et al., 2012). 267

Period length is only one among many metrics used to describe mast seeding patterns, 268 alongside more frequently used synchrony among individuals and interannual variation in seed 269 production (CV) (Koenig et al., 2003; Lamontagne & Boutin, 2007; Qiu et al., 2023). Natural 270 selection acts on physiological traits influencing reproductive synchrony and variability at the 271 proximate level, such as sensitivity to environmental cues (Kelly et al., 2013; Bogdziewicz 272 et al., 2020). These physiological traits produce seed production patterns that enhance fitness 273 at the ultimate level, primarily through economies of scale (Kelly, 1994; Pearse et al., 2016; 274 Pesendorfer et al., 2021). Various combinations of reproductive synchrony and individual 275 variability can generate a similar population-level period length, meaning that this pattern arises 276

from group-level dynamics rather than direct selection on period length itself. For example, 277 low synchrony but high variability among individual trees could yield similar periodicity as 278 high synchrony with moderate variability (Koenig et al., 2003), a pattern reflected in our 279 data by the weak correlation between CV and period length (Fig. S4). The specific ecological 280 context, particularly seed predator mobility, generation time, and feeding behaviour, shapes these 281 combinations (Koenig et al., 2003; Bogdziewicz et al., 2021). Relatively immobile predators, 282 like micromoths, may be satiated by seed production from single trees, reducing selective 283 pressure for synchrony (Nilsson & Wastljung, 1987; Satake et al., 2004). In turn, mobile 284 predators might be attracted to large seed crops, potentially selecting against high interannual 285 variability or synchrony (Koenig et al., 2003). Despite these different selective contexts, period 286 lengths frequently converge around 2-3 years, reflecting common costs and benefits associated 287 with masting strategies. 288

Notably, we found little evidence supporting long period length. Only 10 time series 289 exhibited periods exceeding four years, and these were generally derived from relatively short 290 datasets compared to the period lengths identified. This scarcity of long periods contrasts 291 with some reports in the literature indicating longer masting intervals (T Maki, 1952; Wagner 292 et al., 2010). Such discrepancies might arise from masting intervals being identified based on 293 arbitrary thresholds to define mast years (Bogdziewicz et al., 2024). Conversely, our findings 294 of predominantly short periods (2-3 years) align with the results of Qiu et al. (2023), who 295 identified periodicity averaging around three years in 142 species. Furthermore, the detected 296 periods, when present, were often weak (see, e.g., Fig. (S5), suggesting that these inherent 297 periods account for only a modest fraction of total variability in seed production. Consequently, 298 the periodicity identified here might offer limited utility for predicting future seed crops. While 299 our analysis supports the presence of underlying reproductive periods in many species, the 300 weakness and variability of these signals cautions against characterizing masting as strictly 301 periodical. Nevertheless, forecasting models that increasingly link seed production to weather 302 variability (Journé et al., 2023; Wion et al., 2025) could potentially benefit from incorporating 303 patterns indicating a higher likelihood of masting events as more time passes since the last 304 occurrence. 305

The effects of local climate on period length were generally weak or absent, as observed in 306 seven species lacking clear climatic associations. This finding is consistent with recent work 307 showing that intraspecific variation in masting behaviour is often unrelated to climate, with no 308 consistent support for the environmental stress hypothesis across species or masting metrics 309 (Foest *et al.*, 2025a). One possible explanation is that climate ranges within species were too 310 narrow to detect consistent effects. However, most species in our dataset exhibited substantial 311 within-species variation in both MAT and MAP, including those with weak or absent effects 312 (Fig. S6), making this explanation unlikely. This limited effect of climate on period length may 313 instead reflect local adaptation, with the thresholds for climatic cues triggering high-seeding 314 years varying across sites (Kon et al., 2005; Foest et al., 2024, 2025b). For example, species 315

responding to warm temperatures might not exhibit lower cue frequencies at colder sites, as 316 the "warm" threshold may adjust to local climate norms (Foest et al., 2024). Similarly, local 317 adaptation may affect how climate influences resource dynamics, making absolute climatic 318 values less predictive of resource accumulation rates (Piper & Fajardo, 2024). When climate 319 influences period length, temperature and precipitation can function as reproductive cues or 320 vetoes (Kelly et al., 2013; Bogdziewicz et al., 2019). Warmer temperatures may shorten 321 period length by frequently triggering seed production (Shibata et al., 2020), although colder 322 conditions, such as late frosts, could lengthen intervals by inhibiting reproduction (Inouve, 2008; 323 Éliane Schermer et al., 2020). Similarly, precipitation variability affects reproductive dynamics; 324 drought conditions may extend reproductive intervals by limiting resources, whereas excessive 325 precipitation may reduce pollination success, leading to longer resource accumulation periods 326 between mast events (Espelta et al., 2008; Fleurot et al., 2024). Thus, the relationship between 327 climate and period length appears complex and influenced by species-specific adaptations and 328 local environmental contexts. 329

In summary, our analyses highlight the widespread occurrence of periodicity in masting 330 across diverse tree species, predominantly converging toward a 2-3 year period length. This 331 consistent periodicity highlights the adaptive balance between the advantages of predator satia-332 tion, primarily targeting specialised insect seed predators, and the risks associated with prolonged 333 reproductive delays, such as resource loss or missed reproductive opportunities. The limited 334 effect of climatic variables on periodicity across species suggests that local adaptation likely 335 modulates plant sensitivity to weather cues, complicating broad-scale predictions of masting be-336 haviour under changing climates. Interestingly, the rarity of longer masting periods challenges 337 previous assumptions derived from return interval studies, indicating that prolonged periods 338 may be ecologically exceptional rather than typical. 339

340 Acknowledgements

This study was funded by the European Union (ERC, ForestFuture, 101039066). Views and opinions expressed are however those of the authors only and do not necessarily reflect those of the European Union or the European Research Council. Neither the European Union nor the granting authority can be held responsible for them. AHP received funding from Defra through a UKRI Future of UK Treescapes programme Fellowship.

346

347 Author Contributions Statement

KK, MB, JSz, AHP, JF, DA designed the study. KK, Jsz, TS, MS conducted the analysis. KK
and MB co-wrote the first draft of the manuscript. All authors contributed to the interpretation
of the analysis, revised the draft, and gave final approval for publication.

351

Declaration of interests

³⁵³ No competing interests to declare.

354

355 Data availability statement

³⁵⁶ The data supporting the results will be archived in a permanent repository upon acceptance.

357

358 References

- Ahrestani, F.S., Smith, W.K., Hebblewhite, M., Running, S. & Post, E. (2016). Variation in stability of elk and red deer populations with abiotic and biotic factors at the speciesdistribution scale. *Ecology*, 97, 3184–3194.
- Allen, R.B., Mason, N.W.H., Richardson, S.J. & Platt, K.H. (2012). Synchronicity, periodicity
 and bimodality in inter-annual tree seed production along an elevation gradient. *Oikos*, 121,
 367–376.
- Ascoli, D., Vacchiano, G., Turco, M., Conedera, M., Drobyshev, I., Maringer, J. *et al.* (2017).
 Inter-annual and decadal changes in teleconnections drive continental-scale synchronization
 of tree reproduction. *Nature Communications*, 8, 2205.
- Bisi, F., von Hardenberg, J., Bertolino, S., Wauters, L.A., Imperio, S., Preatoni, D.G. *et al.*(2016). Current and future conifer seed production in the Alps: testing weather factors as
 cues behind masting. *European Journal of Forest Research*, 135, 743–754.
- ³⁷¹ Bjørnstad, O.N., Falck, W. & Stenseth, N.C. (1995). A geographic gradient in small rodent
 ³⁷² density fluctuations: a statistical modelling approach. *Proceedings: Biological Sciences*,
 ³⁷³ 262, 127–133.
- Bjørnstad, O.N., Liebhold, A.M. & Johnson, D.M. (2008). Transient synchronization following
 invasion: revisiting Moran's model and a case study. *Population Ecology*, 50, 379–389.
- Bogdziewicz, M., Calama, R., Courbaud, B., Espelta, J.M., Hacket-Pain, A., Journé, V. *et al.* (2023). How to measure mast seeding? *New Phytologist*, 239, 830–838.
- Bogdziewicz, M., Kelly, D., Ascoli, D., Caignard, T., Chianucci, F., Crone, E.E. *et al.* (2024).
 Evolutionary ecology of masting: mechanisms, models, and climate change. *Trends in Ecology & Evolution*, 39, 851–862.
- Bogdziewicz, M., Kelly, D., Tanentzap, A.J., Thomas, P.A., Lageard, J.G.A. & Hacket-Pain, A.
 (2020). Climate change strengthens selection for mast seeding in european beech. *Current Biology*, 30, 3477–3483.e2.
- Bogdziewicz, M., Kuijper, D., Zwolak, R., Churski, M., Jędrzejewska, B., Wysocka-Fijorek,
 E. *et al.* (2022). Emerging infectious disease triggered a trophic cascade and enhanced
 recruitment of a masting tree. *Proceedings of the Royal Society B: Biological Sciences*, 289,
 20212636.
- Bogdziewicz, M., Szymkowiak, J., Tanentzap, A.J., Calama, R., Marino, S., Steele, M.A. *et al.* (2021). Seed predation selects for reproductive variability and synchrony in perennial plants.
 New Phytologist, 229, 2357–2364.

- ³⁹¹ Bogdziewicz, M., Żywiec, M., Espelta, J.M., Fernández-Martinez, M., Calama, R., Ledwoń, M.
- *et al.* (2019). Environmental veto synchronizes mast seeding in four contrasting tree species.
- ³⁹³ *The American Naturalist*, 194, 246–259.

Bregnard, C., Rais, O. & Voordouw, M.J. (2021). Masting by beech trees predicts the risk of
 Lyme disease. *Parasites & Vectors*, 14, 168.

- ³⁹⁶ Broome, A., Hendry, S. & Peace, A. (2007). Annual and spatial variation in coning shown by
- ³⁹⁷ the Forest Condition Monitoring programme data for Norway spruce, Sitka spruce and Scots
- ³⁹⁸ pine in Britain. *Forestry: An International Journal of Forest Research*, 80, 17–28.
- Burns, R.M., Honkala, B.H. & Coordinators, T. (1990). Silvics of North America: Volume
 1. Conifers. United States Department of Agriculture (USDA), Forest Service, Agriculture
 Handbook 654.
- ⁴⁰² Bürkner, P.C. (2021). Bayesian Item Response Modeling in R with brms and Stan. *Journal of* ⁴⁰³ Statistical Software, 100, 1–54.
- 404 Cornulier, T., Yoccoz, N.G., Bretagnolle, V., Brommer, J.E., Butet, A., Ecke, F. et al. (2013).
- ⁴⁰⁵ Europe-wide dampening of population cycles in keystone herbivores. *Science*, 340, 63–66.
- Crone, E.E. & Rapp, J.M. (2014). Resource depletion, pollen coupling, and the ecology of mast
 seeding. *Annals of the New York Academy of Sciences*, 1322, 21–34.
- ⁴⁰⁸ Curran, L.M. & Leighton, M. (2000). Vertebrate responses to spatiotemporal variation in seed
 ⁴⁰⁹ production of mast-fruiting Dipterocarpaceae. *Ecological Monographs*, 70, 101–128.
- Elton, C.S. (1924). Periodic fluctuations in the numbers of animals: their causes and effects.
 Journal of Experimental Biology, 2, 119–163.
- ⁴¹² Espelta, J.M., Cortés, P., Molowny-Horas, R., Sánchez-Humanes, B. & Retana, J. (2008).
 ⁴¹³ Masting mediated by summer drought reduces acorn predation in mediterranean oak forests.
 ⁴¹⁴ *Ecology*, 89, 805–817.
- Fleurot, E., Keurinck, L., Boulanger, V., Debias, F., Delpierre, N., Delzon, S. *et al.* (2024).
 Reconciling pollen Limitation Theories: Insights From Temperate Oak Masting. *Ecology Letters*, 27, e70009.
- Foest, J.J., Bogdziewicz, M., Pesendorfer, M.B., Ascoli, D., Cutini, A., Nussbaumer, A. *et al.* (2024). Widespread breakdown in masting in European beech due to rising summer temper atures. *Global Change Biology*, 30, e17307.
- Foest, J.J., Caignard, T., Pearse, I.S., Bogdziewicz, M. & Hacket-Pain, A. (2025a). Intraspecific
 variation in masting across climate gradients is inconsistent with the environmental stress
 hypothesis. *Ecology*, 106, e70076.

- Foest, J.J., Szymkowiak, J., Dyderski, M., Jastrzębowski, S., Fuchs, H., Ratajczak, E. *et al.* (2025b). No refuge at the edge for European beech as climate warming disproportionately
 reduces masting at colder margins. Publisher: EcoEvoRxiv.
- Garcia, G., Re, B., Orians, C. & Crone, E. (2021). By wind or wing: pollination syndromes and
 alternate bearing in horticultural systems. *Philosophical Transactions of the Royal Society B: Biological Sciences*, 376, 20200371.
- Hacket-Pain, A., Foest, J.J., Pearse, I.S., LaMontagne, J.M., Koenig, W.D., Vacchiano, G. et al.
- (2022). MASTREE+: Time-series of plant reproductive effort from six continents. *Global Change Biology*, 28, 3066–3082.
- Harris, I., Osborn, T.J., Jones, P. & Lister, D. (2020). Version 4 of the CRU TS monthly
 high-resolution gridded multivariate climate dataset. *Scientific Data*, 7, 109.
- Huang, L., Jin, C., Zhou, L., Song, K., Qian, S., Lin, D. et al. (2021). Benefit versus cost trade-
- offs of masting across seed-to-seedling transition for a dominant subtropical forest species.
- 437 *Journal of Ecology*, 109, 3087–3098.
- Igarashi, S., Yoshida, S., Kenzo, T., Sakai, S., Nagamasu, H., Hyodo, F. *et al.* (2024). No
 evidence of carbon storage usage for seed production in 18 dipterocarp masting species in a
 tropical rain forest. *Oecologia*, 204, 717–726.
- Inouye, D.W. (2008). Effects of climate change on phenology, frost damage, and floral abundance
 of montane wildflowers. *Ecology*, 89, 353–362.
- Journé, V., Hacket-Pain, A., Oberklammer, I., Pesendorfer, M.B. & Bogdziewicz, M. (2023).
- Forecasting seed production in perennial plants: identifying challenges and charting a path
 forward. *New Phytologist*, 239, 466–476.
- Kelly, D. (1994). The evolutionary ecology of mast seeding. *Trends in Ecology & Evolution*, 9,
 447 465–470.
- Kelly, D. (2020). Nutrient scarcity cannot cause mast seeding. *Nature Plants*, 6, 760–762.
- Kelly, D., Geldenhuis, A., James, A., Penelope Holland, E., Plank, M.J., Brockie, R.E. et al.
- (2013). Of mast and mean: differential-temperature cue makes mast seeding insensitive to
 climate change. *Ecology Letters*, 16, 90–98.
- Kelly, D., Harrison, A.L., Lee, W.G., Payton, I.J., Wilson, P.R. & Schauber, E.M. (2000).
 Predator satiation and extreme mast seeding in 11 species of Chionochloa (Poaceae). *Oikos*,
 90, 477–488.
- Kelly, D., Hart, D.E. & Allen, R.B. (2001). Evaluating the Wind Pollination Benefits of Mast
 Seeding. *Ecology*, 82, 117–126.

- Kelly, D. & Sork, V.L. (2002). Mast seeding in perennial plants: why, how, where? *Annual Review of Ecology, Evolution, and Systematics*, 33, 427–447.
- Kelly, D., Szymkowiak, J., Hacket-Pain, A. & Bogdziewicz, M. (2024). Fine-tuning mast
 seeding: as resources accumulate, plants become more sensitive to weather cues. *New Phytologist.*
- Kettle, C.J., Ghazoul, J., Ashton, P.S., Cannon, C.H., Chong, L., Diway, B. *et al.* (2010). Mass
 Fruiting in Borneo: A Missed Opportunity. *Science*, 330, 584–584.
- Koenig, W.D., Kelly, D., Sork, V.L., Duncan, R.P., Elkinton, J.S., Peltonen, M.S. *et al.* (2003).
 Dissecting components of population-level variation in seed production and the evolution of
 masting behavior. *Oikos*, 102, 581–591.
- Kon, H., Noda, T., Terazawa, K., Koyama, H. & Yasaka, M. (2005). Proximate factors causing
 mast seeding in fagus crenata: The effects of resource level and weather cues. *Canadian Journal of Botany*, 83, 1402–1409.
- Köhnke, M.C., Binny, R.N., Holland, E.P. & James, A. (2020). The necessity of tailored control
 of irrupting pest populations driven by pulsed resources. *Theoretical Ecology*, 13, 261–275.
- Lamontagne, J.M. & Boutin, S. (2007). Local-scale synchrony and variability in mast seed production patterns of picea glauca. *Journal of Ecology*, 95, 991–1000.
- ⁴⁷⁴ Maeto, K. & Ozaki, K. (2003). Prolonged diapause of specialist seed-feeders makes predator ⁴⁷⁵ satiation unstable in masting of Quercus crispula. *Oecologia*, 137, 392–398.
- ⁴⁷⁶ Michaud, T.J., Pearse, I.S., Kauserud, H., Andrew, C.J. & Kennedy, P.G. (2024). Mast seeding in
- European beech (Fagus sylvatica L.) is associated with reduced fungal sporocarp production
 and community diversity. *Ecology Letters*, 27, e14460.
- ⁴⁷⁹ Monks, A., Monks, J.M. & Tanentzap, A.J. (2016). Resource limitation underlying multiple
 ⁴⁸⁰ masting models makes mast seeding sensitive to future climate change. *New Phytologist*, 210,
 ⁴⁸¹ 419–430.
- Müller-Haubold, H., Hertel, D. & Leuschner, C. (2015). Climatic drivers of mast fruiting in
 european beech and resulting c and n allocation shifts. *Ecosystems*, 18, 1083–1100.
- Nilsson, S.G. & Wastljung, U. (1987). Seed predation and cross-pollination in mast-seeding
 beech (Fagus sylvatica) patches. *Ecology*, 68, 260–265.
- ⁴⁸⁶ Nussbaumer, A., Waldner, P., Apuhtin, V., Aytar, F., Benham, S., Bussotti, F. et al. (2018).
- Impact of weather cues and resource dynamics on mast occurrence in the main forest tree
 species in Europe. *Forest Ecology and Management*, 429, 336–350.

- ⁴⁸⁹ Nussbaumer, A., Waldner, P., Etzold, S., Gessler, A., Benham, S., Thomsen, I.M. et al. (2016).
- ⁴⁹⁰ Patterns of mast fruiting of common beech, sessile and common oak, Norway spruce and
- ⁴⁹¹ Scots pine in Central and Northern Europe. *Forest Ecology and Management*, 363, 237–251.
- Ostfeld, R.S., Keesing, F., Ostfeld, R.S. & Keesing, F. (2000). Pulsed resources and community
 dynamics of consumers in terrestrial ecosystems. *Trends in Ecology & Evolution*, 15, 232–237.
- Pearse, I.S., Koenig, W.D. & Kelly, D. (2016). Mechanisms of mast seeding: resources, weather,
 cues, and selection. *New Phytologist*, 212, 546–562.
- Pesendorfer, M.B., Ascoli, D., Bogdziewicz, M., Hacket-Pain, A., Pearse, I.S. & Vacchiano,
 G. (2021). The ecology and evolution of synchronized reproduction in long-lived plants.
 Philosophical Transactions of the Royal Society B: Biological Sciences, 376.
- Piper, F.I. & Fajardo, A. (2024). Local adaptation to aridity in a widely distributed angiosperm
 tree species is mediated by seasonal increase of sugars and reduced growth. *Tree Physiology*,
 44, 134–144.
- Pélisson, P.F., Bel-Venner, M.C., Rey, B., Burgevin, L., Martineau, F., Fourel, F. *et al.* (2012).
 Contrasted breeding strategies in four sympatric sibling insect species: when a proovigenic
 and capital breeder copes with a stochastic environment. *Functional Ecology*, 26, 198–206.
- Qiu, T., Aravena, M.C., Ascoli, D., Bergeron, Y., Bogdziewicz, M., Boivin, T. *et al.* (2023).
 Masting is uncommon in trees that depend on mutualist dispersers in the context of global
 climate and fertility gradients. *Nature Plants*, 9, 1044–1056.
- Rapp, J.M., McIntire, E.J.B. & Crone, E.E. (2013). Sex allocation, pollen limitation and masting
 in whitebark pine. *Journal of Ecology*, 101, 1345–1352.
- Rees, M., Kelly, D. & Bjørnstad, O.N. (2002). Snow Tussocks, Chaos, and the Evolution of
 Mast Seeding. *The American Naturalist*, 160, 44–59.
- Satake, A. & Bjørnstad, O.N. (2008). A resource budget model to explain intraspecific variation
 in mast reproductive dynamics. *Ecological Research*, 23, 3–10.
- Satake, A., N. Bjørnstad, O. & Kobro, S. (2004). Masting and trophic cascades: interplay
 between rowan trees, apple fruit moth, and their parasitoid in southern Norway. *Oikos*, 104,
 540–550.
- Éliane Schermer, Bel-Venner, M.C., Gaillard, J.M., Dray, S., Boulanger, V., Roncé, I.L. *et al.* (2020). Flower phenology as a disruptor of the fruiting dynamics in temperate oak species.
 New Phytologist, 225, 1181–1192.

- Seget, B., Bogdziewicz, M., Holeksa, J., Ledwoń, M., Piechnik, , Milne-Rostkowska, F. *et al.* (2022). Masting increases seedling recruitment near and far: Predator satiation and improved
- dispersal in a fleshy-fruited tree. *Journal of Ecology*, 110, 2321–2331.
- Shibata, M., Masaki, T., Yagihashi, T., Shimada, T. & Saitoh, T. (2020). Decadal changes in
 masting behaviour of oak trees with rising temperature. *Journal of Ecology*, 108, 1088–1100.
- Sork, V.L., Bramble, J. & Sexton, O. (1993). Ecology of mast-fruiting in three species of north
 american deciduous oaks. *Ecology*, 74, 528–541.
- ⁵²⁷ T Maki (1952). Local longleaf seed years. *Journal of Forestry*, pp. 321–322.
- Tachiki, Y. & Iwasa, Y. (2010). Both seedling banks and specialist seed predators promote the evolution of synchronized and intermittent reproduction (masting) in trees. *Journal of Ecology*, 98, 1398–1408.
- Team, R.C. (2020). RA language and environment for statistical computing, R Foundation for
 Statistical. *Computing*.
- Vacchiano, G., Pesendorfer, M.B., Conedera, M., Gratzer, G., Rossi, L. & Ascoli, D. (2021).
 Natural disturbances and masting: from mechanisms to fitness consequences. *Philosophical Transactions of the Royal Society B: Biological Sciences*, 376, 20200384.
- ⁵³⁶ Venner, S., Siberchicot, A., Pélisson, P.F., Schermer, E., Bel-Venner, M.C., Nicolas, M. *et al.* (2016). Fruiting Strategies of Perennial Plants: A Resource Budget Model to Couple Mast
 Seeding to Pollination Efficiency and Resource Allocation Strategies. *The American Natu- ralist*, 188, 66–75.
- Visser, M.D., Jongejans, E., van Breugel, M., Zuidema, P.A., Chen, Y.Y., Rahman Kassim, A.
 et al. (2011). Strict mast fruiting for a tropical dipterocarp tree: a demographic cost–benefit
 analysis of delayed reproduction and seed predation. *Journal of Ecology*, 99, 1033–1044.
- ⁵⁴³ Wagner, S., Collet, C., Madsen, P., Nakashizuka, T., Nyland, R.D. & Sagheb-Talebi, K. (2010).
 ⁵⁴⁴ Beech regeneration research: From ecological to silvicultural aspects. *Forest Ecology and* ⁵⁴⁵ *Management*, 259, 2172–2182.
- Waller, D.M. (1979). Models of mast fruiting in trees. *Journal of Theoretical Biology*, 80,
 223–232.
- ⁵⁴⁸ Wion, A.P., Pearse, I.S., Broxson, M. & Redmond, M.D. (2025). Mast hindcasts reveal pervasive
 ⁶⁴⁹ effects of extreme drought on a foundational conifer species. *New Phytologist*, 246, 450–460.
 ⁵⁵⁰ Publisher: John Wiley & Sons, Ltd.
- ⁵⁵¹ Wion, A.P., Weisberg, P.J., Pearse, I.S. & Redmond, M.D. (2020). Aridity drives spatiotemporal
 ⁵⁵² patterns of masting across the latitudinal range of a dryland conifer. *Ecography*, 43, 569–580.

- Yasaka, M., Terazawa, K., Koyama, H. & Kon, H. (2003). Masting behavior of Fagus crenata
 in northern Japan: spatial synchrony and pre-dispersal seed predation. *Forest Ecology and Management*, 184, 277–284.
- Young, J.A. & Young, C.G. (1992). Seeds of Woody Plants in North America. Dioscorides Press
 Portland.
- ⁵⁵⁸ Zwolak, R., Celebias, P. & Bogdziewicz, M. (2022). Global patterns in the predator satiation
- effect of masting: A meta-analysis. *Proceedings of the National Academy of Sciences of the United States of America*, 119.
- Övergaard, R., Gemmel, P. & Karlsson, M. (2007). Effects of weather conditions on mast year
 frequency in beech (Fagus sylvatica L.) in Sweden. *Forestry: An International Journal of Forest Research*, 80, 555–565.
- ⁵⁶⁴ Żywiec, M., Holeksa, J., Ledwoń, M. & Seget, P. (2013). Reproductive success of individ-
- ⁵⁶⁵ uals with different fruit production patterns. What does it mean for the predator satiation
- ⁵⁶⁶ hypothesis? *Oecologia*, 172, 461–467.

567 Supporting Information

Authors: K. Kondrat, J. Szymkowiak, A. Hacket-Pain, M. Shibata, J. Foest, D. Ascoli, M.
 Bogdziewicz

- 570
- ⁵⁷¹ **Title**: Short reproductive periods dominate mast seeding across diverse tree species
- 572

Species name	Number of populations	Mean time series	Maximal time
		length	series length
Abies alba	17	18	25
Abies amabilis	12	11	12
Alnus incana	10	18	22
Araucaria araucana	15	16	18
Betula pubescens	9	13	17
Fagus crenata	23	17	24
Fagus sylvatica	65	29	43
Picea abies	39	22	40
Picea engelmanii	24	36	41
Picea glauca	42	24	57
Pinus albicaulis	61	22	33
Pinus edulis	40	16	20
Pinus palustris	11	47	62
Pinus ponderosa	57	16	31
Pinus sylvestris	57	21	45
Quercus cerris	11	25	31
Quercus douglasii	10	30	41
Quercus lobata	12	27	41
Quercus petraea	30	13	23
Ouercus robur	11	16	27

 Table S1: Summary of per species sample size in our study.



Figure S1: Relationship between period length and time series length. Each point represents an individual time series. A zero-truncated Poisson generalized linear model (GLM) was fitted to examine the relationship between period length as a response variable and time series length as an explanatory variable. The model showed no significant effect of time series length on period length (p = 0.681). The blue line represents model predictions, with 95% confidence intervals.



Figure S2: Temporal autoregression coefficients across studied species (N = 20). Each dot shows one population.



Figure S3: Relationship between temporal autoregression coefficients $1 + a_1$ and a_2 and elevation (N = 149 populations, 10 species). Each row represents a separate species. Each point represents the coefficient of the linear regression of the relationship, with error bars indicating the 95% confidence intervals. Non-significant coefficients are indicated by lighter colours.



Figure S4: Relationship between period length and CVp (coefficient of variation for each population). Each point represents an individual time series, coloured by time series length. CVp is calculated as the standard deviation of seed production in a population divided by the mean seed production in that population. A Gamma generalized linear model (GLM) with a log link was fitted, including a quadratic term for period length and time series length as explanatory variables, with CVp as the response variable. Period length (p < 0.001) had a significant effect, while time series length was not significant (p = 0.678). The model had a marginal R² of 0.03, indicating that period length explained only a small proportion of the variation in CVp. The red line represents model predictions, with 95% confidence intervals.



Figure S5: Time series of seed production and respective parabola plots for 20 populations of *Pinus* ponderosa Each row contains two populations, with time series graphs (first and third columns) showing interannual variation in seed production and parabola plots (second and fourth columns) showing the period length of given time series based on $1 + a_1$ and a_2 coefficients from AR2 model.



Figure S6: Variation in mean annual temperature (MAT) and mean annual precipitation (MAP) across populations within species. Each bar represents the standard deviation of MAT and MAP values across sites for each species, with species ordered from lowest to highest variation.