Interplay between Intertwining of fecundity, sexual and growth selection on <u>the</u> spring phenology along an altitudinal gradient of <u>of</u> European beech (*Fagus sylvatica* L.).

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16 **ABSTRACT**

- 17 Background: Plant phenological traits such as the timing of budburst or flowering may evolve at ecological time scale through response to fecundity and viability selectionfecundity and 18 19 viability selections on phenological traits are difficult to distinguish in plants, where 20 vegetative and reproductive phenologies are closely synchronized. Moreover, interference 21 with sexual selection may arise from assortative mating. This study aims at disentangling 22 investigating how these three components of selection on spring phenology may combine 23 in European beech populations in contrasted environments (high versus low altitude). 24 Methods: we monitored the timing of budburst (TBB) was surveyed in 147 and 192339 adult 25 beech trees in two natural populations at low and high elevation respectively along an altitudinal gradientand estimated their . Male and female individual fecundities were 26 27 estimated using spatially explicit mating models paternity and parentage analyses of 1414 seeds and 473 seedlings, which also allowed mating networks to be inferred. Fecundity 28 selection was infered by regressing fecundities on TBB, while sexual selection was inferred by 29 regressing fecundities on mating opportunities (i.e., TBB mismatch).- The correlation between 30 31 mates for flowering time (i.e., assortative mating) was estimated based on paternity analyses. 32 Morever, TBB and growth was surveyed in 1552 and 17093261 seedlings from 40 families 33 originating from the same populations and grown planted in a common garden, and viability 34 selection was inferred by regressing growth on TBB. 35 Results: assortative mating occurred only at low elevation, where spring phenology was also more spread out. Phenological mismatch reduced male but not female fecundities at both 36 37 plots, indicating sexual selection to maximize mating opportunities. Overall, directionnal 38 fecundity selection on female fitness favored trees with earlier TBB. Sexual selection acted only on male fitness through assortative mating and favored tree with mean TBB value 39 40 (stabilizing selection). In the common garden, early budburst was associated to higher 41 seedling growth. The respective intensity of directionnal and stabilizing selection varied with
- 42 the environment: At low altitude, directional selection for earlier phenology was modulated

43 by strong assortative mating and by an interaction effect between TBB an size on female 44 fecundity whereas at At high altitude elevation, directional selection for earlier phenology was 45 reinforced by selection through male fecundity directional fecundity selection for earlier TBB 46 occurred both through male and female fecundities. At low elevation, directional fecundity selection for earlier spring phenology was mitigated by a positive association between TBB 47 48 and fecundity in the smaller trees. assortative mating occurred only at low elevation, where 49 spring phenology was also more spread out. Phenological mismatch reduced male but not female fecundities at both plots, indicating sexual selection to maximize mating 50 51 opportunities. fecundity selection on female fitness and viability selection on seedlings growth both favor 52 53 early phenology, while sexual selection on male fitness through assortative mating modulates 54 this trend (stabilizing selection). 55 56 57 **Discussion**: This study showed that selection through female fecundity and seedlings growth 58 predominantly selected for earlier TBB, while sexual selection on male fitness through assortative mating modulated this trendsexual selection arising from assortative mating could 59 drive stabilizing selection on TBB through the male function, while selection through female 60 61 fecundity predominantly selects for earlier TBB. This interplay between intertwining of sexual 62 and-fecundity and sexual selection calls for an integrative approach to predict the evolution 63 of spring phenology under a changing climate. 64

Keywords: budburst phenology, selection gradient, assortative mating, Bateman's gradient,
 parentage/paternity analyses, Mixed-Effect Mating Model (MEMM), Fagus sylvatica

67 Introduction

68 Many changes in phenology (i.e., the timing of biological events) were observed during the last decades 69 and attributed to climate change (Parmesan and Yohe 2003). In particular, leafing, flowering and fruiting 70 records advanced significantly in temperate zones (Menzel et al. 2006), consistent with the rise of 71 spring/summer temperatures. Besides the plastic component of these phenological shifts, rapid evolution 72 of phenological traits in response to selection has also been reported (Franks et al. 2007; Hamann et al. 73 2018). However, it is still largely unknown to what extent evolution over a few generations microevolution 74 may contribute to the response of plants' populations to ongoing climate change to climatic variation 75 (Merilä and Hendry 2014). Moreover, in many plants, vegetative phenology (the timing of germination, 76 stem and leaf development) and reproductive phenology (the timing of flowering and fruiting) are tightly 77 synchronized throughout the yearly cycle. Hence, selection on phenological traits is likely to be the complex 78 outcome of viability selection (selection for phenotypes that increase survival), fecundity selection 79 (selection for phenotypes that increase fecundity) and sexual selection (selection arising from competition 80 for mating partners or their gametes) (Figure 1).

- 81 This study aims to account for these different components while investigating selection on vegetative
 82 phenology in a temperate tree species along an altitudinal gradient.
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- 84 in a temperate tree species along an altitudinal gradient.
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88 Figure 1 – A schematic representation of the expected relationships between vegetative/flowering 89 phenology and fitness at individual level. Within a population, individuals with Delayed delayed timing of 90 budburst relative to the population mean (TBB) is are expected to increase have higher survival through 91 frost avoidance, but also to decreaselower fecundity and survival through reduced length of the growing 92 season, and hence reduced reserves. Individuals with -Ddelayed timing of flowering relative to the 93 population mean is are expected to increase have higher fecundity by allowing because more resources to 94 can be accumulated and invested in reproduction, however at the cost of reduced time for seed 95 maturation. Finally, synchronized flowering with the other individuals in the population (i.e., assortative 96 mating) is expected to maximize the number of mates. The combination of these different selection 97 components determines the optimal values of TBB and flowering times, i.e., those maximizing fitness. Note 98 that selection and hence optimal values of phenological traits also vary between environments Moreover, 99 the timing of budburst and flowering are usually closely physiologically synchronized throughout the plant 100 yearly cycle.- Colored boxes represent traits for which we have observations/estimations in this study. "+" 101 (respectively "-") sign indicates an increase (respectively decrease) in the value of the variable under 102 consideration.

103 Most selection studies on the timing of flowering have been conducted in short-lived herbaceous plants 104 (Geber & Griffen, 2003; Munguía-Rosas et al., 2011), while the adaptive value of vegetative phenology has 105 been mainly investigated in long-lived forest trees (Alberto et al. 2013). In both cases, stabilizing selection 106 is the most straightforward expectation within-population, considering life history in temperate 107 ecosystems. In the former case, this is because the fecundity benefits of flowering early (sufficient time for

108 seed maturation) are expected to balance with those of flowering later, as early reproduction usually 109 entails reproducing at a small size with limited resources available for offspring production. Note that this 110 expectation could be different for long-living plants for which resources have been accumulated previous 111 year (Hacket-Pain et al. 2018). Yet, early flowering plants are generally found to be favored (Geber and 112 Griffen 2003; Munguía-Rosas et al. 2011), an apparent paradox for which different explanations have been 113 proposed (Austen et al. (2017). proposed four explanations: (1) selection through other fitness 114 components may counter observed fecundity selection for early flowering; (2) asymmetry in the flowering-115 time-fitness function may make selection for later flowering hard to detect; (3) flowering time and fitness 116 maybe condition-dependent; and (4) selection on flowering duration is largely unaccounted for.

117 Regarding vegetative phenology in long lived plants inhabiting temperate ecosystems, stabilizing 118 selection is expected to be driven by the balance between the benefits of: (1) emerging leaves later and 119 avoiding frost damages on vegetative and reproductive organs, especially in early spring (viability selection) 120 (Augspurger 2013; Bigler and Burgmann 2018); and (2) emerging leaves earlier and maximizing the duration 121 of the growing season, which determines the resource level acquired by photosynthesis (viability and 122 fecundity selection) (Keenan et al. 2014; Richardson et al. 2006). More complex situations may occur when 123 other abiotic or biotic stresses are considered (e.g. early flushing may amplify drought effects, Meier et al. 124 2021). Common-garden experiments generally demonstrate significant genetic differentiation of 125 phenological traits between tree provenances along environmental gradients, suggesting that the 126 differences in climatic conditions led to the evolution of different phenological schedules contributing to 127 populations' local adaptation (Alberto et al. 2013). However, experimental selection studies on tree 128 vegetative phenology remain limited in comparison with those on plant flowering phenology lag behind 129 (but see Bontemps et al. 2017; Alexandre et al. 2020; Westergreen et al. 2023). A recent simulation study 130 with a process-based phenological model accounting both for fecundity and viability selection predicted 131 selection towards earlier TBB across a climatic gradient, and realized TBBs always later than the value 132 conferring highest fitness in different tree species (Gauzere et al., 2020). Moreover, these simulations 133 showed that the strength of this selection was stronger at high than low elevationaltitude, i.e., in the 134 conditions where the growing season is more limiting for the maturation of fruits.

135 Compared to fecundity or viability selection, the role of sexual selection on the evolution of phenology 136 remains understudied, even though the existence of sexual selection in plants is now widely acknowledged 137 (Moore and Pannell 2011). Yet, assortative mating for flowering phenology, that is the positive correlation 138 between male and female flowering time across mated pairs, is obligate in plants (Weis et al. 2014). Hence, 139 variation of individual flowering phenologies within the population may result in sexual selection., and may 140 lead to sexual selection, depending on the overlap in individual flowering phenologies within the 141 population. Moreover, phenological assortative mating is by nature density-dependent, as any individual 142 synchronized with the rest of the population will gain opportunities for mating (Weis et al. 2005). Hence, 143 assortative mating is expected to generate a form of stabilizing sexual selection to lead to stabilizing 144 selection-towards an optimal timing of flowering maximizing mating opportunities, depending on the 145 frequency distribution and of late and early phenotypes. Finally, due to anisogany (the higher cost of 146 producing female versus male gametes), male reproductive success is generally expected to be more 147 limited by mating opportunities than than by investment in each gamete, whereas female reproductive 148 success should depend on their ability to produce viable ovules and seeds rather than on the probability 149 of having ovules fertilized (one of Bateman's principles; Bateman 1948; Tonnabel, David, & Pannell, 2019). 150 These contrasting challenges could lead to different patterns of selection on phenology through male and 151 female reproductive functions.

152 Distinguishing fecundity from sexual selection on phenological traits may be particularly challenging, 153 as both jointly act within a single reproduction episode. However, while fecundity selection can occur even 154 under unlimited access to mates, sexual selection involves limited mating opportunities. Hence, the relationship between phenology and fitness (e.g., phenotypic selection analyses, Lande and Arnold 1983) 155 156 is considered to inform about the joined effects of fecundity and sexual selection (i.e., natural selection), 157 while the relationship between phenology-related mating opportunities and fitness (e.g., Bateman's 158 gradient analyses, Bateman 1948) informs about sexual selection on phenology. Selection gradients (the 159 regression coefficients of relative fitness on a trait) are among the widely used metrics to measure selection 160 in natural populations (Kingsolver et al., 2001). They determine the direction and the pattern of the 161 selection and, when regressing multiple traits, have, at least in theory, the ability to distinguish direct

selection on the trait from the indirect selection coming from correlation with other traits (Lande & Arnold,
 1983). Using quadratic regressions instead of simple linear regressions also permits to detect and test
 stabilizing selection (Lande & Arnold, 1983). Finally, standardized selection gradients provide a
 dimensionless measure of selection strength that can be used for comparisons across traits, populations
 or species (Kingsolver et al., 2001).

167 This study takes advantage of the extensive physiological knowledge on a major monoecious tree, the 168 European beech, and of a well-studied altitudinal gradient in South-Eastern France, to estimate different 169 types of selection gradients on phenological traits the timing of budburst (TBB). European beech is an early 170 flushing deciduous species (Davi et al. 2011), sensitive to frost damages (Lenz et al., 2013). We expect 171 intense selection for early phenology at high elevation, due to high constrains on the length of the growing 172 season. Along the studied gradient, vegetative phenology was monitored both in situ and ex situ, in a 173 common garden of maternal progenies (Oddou-Muratorio et al. 2021). Previous studies showed that, in 174 situ, budburst occurs ~9.8 days earlier at the lower elevationaltitude plot compared to the upper 175 elevationaltitude plot (Davi et al. 2011), but that, in the common garden, the lower plot is ~2.1 days late 176 compared to the upper plot (Gauzere et al., 2020). This is a classical counter-gradient pattern where the in 177 situ plastic response of TBB to different temperature accumulation at the two altitudes (Table 1) hides the 178 genetic differentiation revealed in the common garden (Gauzere et al., 2020). Phenotypic selection 179 analyses conducted at the lower plot found that growth and reproductive (seed set) performances could 180 be maximized either by a water-uptake strategy, including early budburst, or by a water-saving strategy, 181 including late budburst (Bontemps et al. 2017). Finally, male and female fecundities were estimated for all 182 the adults in the lower and upper plots through paternity or parentage analysis of germinated seeds and 183 established saplings (Oddou-Muratorio et al., 2018), which -showed that both female and male fecundities 184 increased with tree size and decreased with density and competition in the neighbourhood, the details of 185 these effect varying among plots at different altitude. These fecundity estimates are effective, i.e. they 186 relate to the effective amount of pollen achieving successful pollination, and to the effective number of 187 seeds achieving successful germination and establishment in the population.

188 The specific aim objective of this study was to investigate simultaneously investigate fecundity, sexual 189 and viability selection on spring phenology in both the upper and lower plots along the altitudinal gradient. 190 Our main hypothesis is that interindividual variations in TBB are strongly correlated with interindividual 191 variations in the timing of flowering, making TBB an appropriate trait to study these different components 192 of selection. First, Wwe estimated fecundity selection on phenology by regressing male and female 193 effective fecundity on TBB (both measured in situ). Second, Wwe used paternity analyses to estimate the 194 strength of investigate assortative mating, and we estimated sexual selection by regressing we investigated 195 whether male and female fecundities were affected by mating opportunities, as measured by (sexual 196 selection) by regressing fecundity on TBBthe phenological mismatch within mating neighborhood (also 197 measured in situ). FinallyThird, viability selection was estimated in the common garden we estimated 198 viability selection in the community garden by analyzing the relationship between TBB and seedlings 199 growth, under the hypothesis that vigor (i.e. growth capacity) is positively associated with viability (Collet 200 and Le Moguedec 2007). For these three inferences of fecundity, sexual and viability selection, we relied 201 on the classical metrics of selection gradients (the regression coefficients of relative fitness on a trait, Lande 202 & Arnold, 1983). In addition, we analysed both the upper and lower plots along the altitudinal gradient, as 203 these contrasting environments are expected to result in different selective constraints (Table 1).

Table 1: Climatic context and main expectations regarding selection on phenology at the two studied plots. Climate is synthetised by six variables computed from the long-term daily dataset from 1959 to 2013 described in Davi & Cailleret (2017): the mean annual temperature (tmean, °C), the maximum temperature of July (tmax, °C), the minimum temperature of January (tmin, °C), the sum of growing degree days (GDD, °C), the number of frost days, the water stress level between May and September (mm/m²/day), computed as the difference between ETP and precipitations. See Fig S1 for additionnal details.

Plat			Clima	te			Main constraint	Expectation
PIOL	tmean	tmax	tmin	GDD	NFD	Stress		Expectation

N1 N1-LOW	9	22.3	-0.5	3060.9	14.3	154.7	High water stress	Both early or late budburst may enhance survival and fecundity (Bontemps et al. 2017)
N4N4-HIGH	6.3	18.5	-2.8	2187.1	35.1	69.2	Short growing season	Intense fecundity selection for early phenology

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Methods

211 Studied species and <u>site</u>, sampling design

The European beech is a monoecious, wind-dispersed, predominantly outcrossed tree species (Gauzere, Klein, & Oddou-Muratorio, 2013). Male and female flowers are borne on the same branches and open as the leaves unfold (Nielsen & Schaffalitzky de Muckadell, 1954; Packham et al., 2012), between April and May. Beech is protogynous, i.e. male flowers produce pollen after the peak of receptivity of the stigmas of the same plant (Nielsen and Schaffalitzky de Muckadell 1954).

217 Mont Ventoux is located at the warm and dry southern margin of the European beech distribution, and 218 the climate is typical of low altitude mountains with Mediterranean influences (weather station of Mont 219 Serein, 1 445 m a.s.l., 1993–2006; mean annual temperature of 6.8°C and mean annual rainfall of 1300 220 mm). On the northern face of Mont- Ventoux, the beech forest ranges almost continuously from 750 to 221 1700 m above sea level. This steep elevationaltitude gradient provides almost linear variation in mean 222 temperature and humidity with elevationaltitude (Cailleret and Davi 2011). We studied two plots at 223 opposite positions along an altitudinal gradient, named N1-N1-LOW (1.3 ha; 1,020 m a.s.l.), and N4N4-HIGH 224 (0.8 ha; 1,340 m a.s.l.). N1-LOW is at the lower limit of the altitude range for European beech on Mont 225 Ventoux, while N4-HIGH is at the upper limit for sexual reproduction.

226 In 2009, one large masting event occurred, which provided a unique opportunity to collect seeds and 227 monitor regeneration. All potentially reproductive trees were mapped, measured and sampled for genetic 228 analyses (164 at plot N1-N1-LOW and 365 at plot N4N4-HIGH). Mother-trees were chosen among the trees 229 with medium to high seed production, ensuring a minimal distance of 10 m between two mother-trees, 230 and covering the whole plot area. Open-pollinated seeds were collected from 20 mother-trees at each plot 231 (40 families in total for this study, among 60 in total), germinated and sown in the a greenhouse. These 232 open-pollinated seeds first allowed us to estimate patterns of pollen flow and male fecundity (see below). 233 Moreover, Ithe common-garden experiment was arranged in 50 complete blocks (, each block including 234 with two seedlings per family per block, Gauzere et al., 2020) and divided in two contrasted experimental 235 conditions: "watered" (from block 1 to 25) versus "water-stressed" (from block 26 to 50; -(Oddou-236 Muratorio et al. 2021). Briefly, these two two conditions allow us to contrast a situation of non-limiting 237 water availability with a situation of limiting water availability, and to investigate the plastic response of 238 traits to water stress, even though the levels of water stress experienced in the second condition do not 239 match to those experienced in situ. Seedlings of the watered condition were analyzed in a quantitative 240 genetic framework to investigate the within-family, among-families within-plot and among-plots 241 components of the genetic variation at several functionnal traits (Gauzere et al., 2016, 2020). In this study, 242 we took the opportunity to compare the Among these open-pollinated seedlings from plots N1-LOW and 243 N4-HIGH growing in the two experimental conditions. - we genotyped 694 seedlings from plot N1 and 720 244 seedlings from plot N4 (~35.3 seedlings per mother tree).

Additionally<u>Finally</u>, in September 2010, we sampled <u>in situ</u> and genotyped seedlings which originated originating from the same reproduction event in 2009 and germinated <u>in situ</u> in spring 2010 (223 seedlings at plot <u>N1-N1-LOW</u> and 250 seedlings at plot N4). <u>These established seedlings allowed us to estimate</u> <u>patterns of seed flow and female fecundity (see below)</u>. Spring at year 2009 was not colder as an average year considering the mean and minimal temperatures from March to June (Fig. S1). Not late frosts (ie temperatures <-4°C after budburst) were observed in 2009 at any site.

251 Phenology measurement in situ and ex situ (common garden)

In beech, the flowering phenology is hard to follow because (1) it occurs when leaves are spread out,
(2) the succession of the flowering stages is rapid, and (3) the reproductive organs are small. However, as

in oaks (Franjic et al., 2011), reproductive buds open very shortly after leafing (Nielsen and Schaffalitzky de

255 Muckadell 1954). Therefore, we employed budburst phenology as a proxy of reproductive phenology.

256 The budburst was surveyed in situ in spring 2009 on 147 adult trees in population N1N1-LOW, and 192 257 adult trees in N4N4-HIGH. The budburst phenology was characterized using the five stages described by 258 Davi et al. (-2011) and Jean et al. (2023): 1) dormant buds; 2) swelling buds; 3) broken bud scales; 4) 259 emerging leaves; 5) spread out leaves (Fig. S2). The phenological stages of each adult tree were noted on 15 different dates in population N1-N1-LOW (between the 23th of March and the 4th of May 2009), and on 260 261 13 different dates in population N4-N4-HIGH (between the 24th of March and the 5th of May 2009). At each 262 date, individual stage of development was assessed globally for the upper and lower part of the crown, 263 and then average into a single stage value. FirstThen, a phenological score sum (PSS) was computed for 264 each tree as the sum of the phenological stages observed over all of the dates: the higher the PSS at a given 265 date of measurement, the earlier and quicker was leaf unfolding (Bontemps et al. 2017). We also used a 266 linear interpolation to estimate the timing of budburst (TBB) as the date of passage (number of days since 267 1st January) from stage 2 to 3, stage 3 being one the most sensitive stage to frost damages. Finally, we 268 computed the spread of budburst for each adult tree from the temporal sequence of phenological scores, 269 as the number of days where the phenological stage was >2 and ≤ 4 (i.e., the duration of stage 3).

The budburst was also surveyed ex situ for seedlings in the common garden using five stages (Gauzere et al., 2016). The phenological stages were noted on 4 different dates (between the 5th and 26th of April 2011). We also used linear interpolation to estimate TBB as the date of passage (number of days since 1st January) from stage 2 to 3.

274 Fecundity estimation

275 We used the mMale and female fecundities were estimated using the spatially explicit mating model 276 as described in Oddou-Muratorio et al. (2018). Briefly, this model considers mating and dispersal events in 277 a hermaphroditic plant population, and allows individual fecundities to be estimated together with mating 278 system parameters, using genotypes and positions of potential parents and their offspring. It is 279 implemented in a Bayesian framework in the MEMM software. First, the individual male fecundities were 280 estimated with MEMM, jointly with the pollen dispersal kernel, the selfing rate and the pollen migration 281 rate, from the open pollinated seeds with known mother tree. Second, female effective fecundities were 282 estimated jointly with male fecundities, the pollen and seed dispersal kernels, the selfing rate and the 283 pollen and seed migration rates, from one-year established seedlings without any known parent.

284 Remarkably, MEMM estimates of fecundity account for the effect of the relative positions of putative 285 parents and offspring, while at the same basic fecundity, putative parents closer to an offspring would have 286 a higher parentage probability in uncorrected models. Hence, by using MEMM, estimates of fecundity are 287 not sensitive to spatial biases due to sampling design, or edge effects. Moreover, MEMM estimates of 288 fecundities are *effective*: male fecundity is a proxy of the effective amount of pollen achieving successful 289 pollination, and female fecundity is a proxy of the effective number of seeds achieving successful 290 germination and establishment in the population. Therefore, these estimates account for the individual 291 effects (maternal or genetic) that modify the success of mating, including differences in pollen tube growth, 292 seed abortion (for male fecundity) and in seed maturation, germination or early survival during the post-293 dispersal processes (for female fecundity). Effective fecundity provides more realistic estimates of 294 individual plant contribution to the next generation than simpler estimates, such as fruit or seed set. 295 Finally, MEMM estimates of fecundity are relative, and consider uncertainty in parentage reconstruction. 296 Indeed, MEMM does not categorically assign parents to offspring, but rather consider the likelihood of all 297 adults to be the parent of each offspring, accounting for the genotypes of adults and offspring and allowing 298 genotyping errors.

Estimations were performed separately on each plot. The MCMC procedure to estimate individual fecundities and mating system parameters is described in details in Oddou-Muratorio et al. (2018). For this study, we analyzed-used only the <u>fecundities estimated for those</u> adult individuals for which vegetative phenology was monitored, that is: 147 among the 164 adults at plot <u>N1N1-LOW</u>, and 192 among the 365 adults at plot <u>N4N4-HIGH</u>.

304 Fecundity selection analyses on adult trees, *in situ*

To investigate fecundity selection, we used selection gradient analysis (Lande and Arnold 1983), with MEMM estimates of fecundity as the response variable, and TBB as the predictor. Because fecundity variations are shaped by many other factors besides phenology, we included size and competition effects 308 and their interactions using a hierarchical procedure, and selected the most parsimonious model to 309 estimate the effect of phenology.

310 For each sex (male and female) and each plot (N1-N1-LOW and N4N4-HIGH), seven hierarchical models 311 were compared. We first fitted a baseline model M1 including only the predictor of interest (TBB):

312 (1)M1.fec: $\ln(F) = \alpha + \beta TBB + \varepsilon$

313 where F is the fecundity, α is the origin of the regression, β is the directional selection gradient on TBB, 314 and ε is the residual.

315 In a previous study not including phenological traits (Oddou-Muratorio et al. 2018), we showed that 316 both female and male fecundities increased with tree size and decreased with density and competition in 317 the neighborhood. As selection probably simultaneously acts on these different correlated characters 318 (phenology, size, competition), we fitted three models including size or/and competition variables, in 319 addition to TBB:

320 (2)M2.fec: $\ln(F) = \alpha + \beta TBB + \gamma Size + \varepsilon$ 321 M3.fec: $\ln(F) = \alpha + \beta TBB + \delta Compet + \varepsilon$ (3) 322 (4)M4.fec: $\ln(F) = \alpha + \beta TBB + \gamma Size + \delta Compet + \epsilon$

323 Note that several variables were used to measure size and competition (Table 1-2 and Oddou-324 Muratorio et al. 2018). Tree size was assessed by measuring the Diameter at Breast Height (Dbh), but as 325 beech sometimes produces stump shoots resulting in multiple stems, we measured both the maximum 326 Dbh (MaxDbh) and the sum of Dbh (SumDbh) of all the stems produced by a given genotype. Competition 327 on each adult tree was assessed using (1) beech density in a radius of 20 m (ConDens20), (2) a competition 328 index integrating the density and diameter of beech competitors in a radius of 20 m (ConMartin20) and (3) 329 tree stature (a class variable with 3 levels: dominant, codominant, and suppressed). Based on the previous 330 results of Oddou-Muratorio et al. (2018), we chose the most pertinent variable for each sex and plot 331 (Table <u>12</u>).

332 Table 12: Variables included in the phenotypic fecundity selection analyses. The size and competition 333 variables were identified as best predictors (BestPred) of female fecundity (F_{Ω}) and male fecundity (F_{Ω}) in 334 the previous study of Oddou-Muratorio et al. (2018).

Category	Variable name	Variable definition	BestPred	Range at N1	Range at N4
Size	MaxDbh (cm)	Maximum diameter of the clonal copies	F♀	10.1-45.4	9.2-28.3
	SumDbh (cm)	Sum of diameters of the clonal copies	Fð	11.6-231.6	10.2-127.1
Competition	ConDens20	Conspecific local density in a radius of 20 m	F ♀	2-50	6-115
	TotMartin20	Total competition index in a radius of 20 m	F♀ and F♂	8.7-100	43.0-127.4
	Stature	dominant, codominant, or suppressed	Fð	44, 51, 52, resp.	35, 66, 91, resp.

335 * this variable was retained as the estimator of phenological mismatch in the fecundity and sexual selection analyses of this study. 336

337 Finally, we also fitted three models including two-way interaction terms between TBB and 338 size/competition covariates, in order to account for possible changes in the relationship between TBB and 339 fecundity depending on size or competition:

340		(5)	M5.fec: ln(F) = $\alpha + \beta TBB + \gamma Size + \delta Compet + \kappa TBB \times Size + \varepsilon$
341		(6)	M6.fec: ln(F) = $\alpha + \beta TBB + \gamma Size + \delta Compet + \lambda TBB \times Compet + \varepsilon$
342	(7)	M7.fe	c: $ln(F) = \alpha + \beta TBB + \gamma Size + \delta Compet + \kappa TBB \times Size + \lambda TBB \times Compet + \varepsilon$

343 We compared these seven hierarchical models based on the Akaike information criterion (Akaike 1987) 344 corrected for small sample size (AICc, Burnham and Anderson 2002), and we selected the most

- parsimonious model, denoted BestFec-Model in the following. <u>Significance of the effects (AIC and *p*-values)</u>
 was assessed with the function *drop1* of the R package *stats*.
- 347 A quadratic effect of TBB can also be included to estimate stabilizing selection through a bell-shaped 348 response function (Lande and Arnold 1983). This was done for by adding an additional term γ TBB² in the 349 BestFec-Model selected above.

350 Variable transformation and model fitting

351 Note that MEMM estimates of fecundity are relative, as required for selection gradient estimation 352 (Lande and Arnold 1983). Moreover, the fecundities were log-transformed to approach Gaussian 353 distribution and to account for the higher variance associated to higher fecundities. Besides, all the 354 predictor variables (including TBB and PMis) were scaled to mean zero and unit variance. Such 355 transformation of the predictor variables allows improving the interpretability and comparability of the 356 estimated regression coefficients, especially when interactions are present (Schielzeth 2010). Once the 357 best model selected, we estimated the standardized selection gradients on fecundity by fitting this selected 358 best model without log-transformation of fecundity.

All models were fitted using the *Im* function implemented in R-base. Model comparisons was performed using the *aictab* function of the R package 'AlCcmodavg' (Mazerolle 2020). For the best models, the residuals were visually inspected through a plot of residuals vs predicted. All the analyses are available as online supplementary material (file SelectionAnalyses_adult.html at https://doi.org/10.57745/ZVPNXX).

363 Mating opportunities and assortative mating estimation

Direct observation of mating or pollination events being impossible in this anemophilous species, we 364 365 used phenological mismatch as a proxy of mating opportunities. We computed the sum of phenological 366 mismatches between each adult tree *i* and its neighbors in a radius R as: $|PMis_i|_s = \sum_{i \text{ in } R} |TBB_i - TBB_i|$, with R-values of 20, 50, 75 or 100 m. Note that similar |PMis|s values can be obtained either with a low 367 368 density and large asynchrony or with a high density and low asynchrony. We also computed a mean 369 phenological mismatch |PMis|_m, weighted by density. We hypothesize that the greater the phenological 370 mismatch, the lower the opportunities of mating. Note also that using absolute mismatches implicitly 371 assumes a symmetric effect asynchrony (earlier and later trees plays the same role for mating 372 opportunities).

We estimated the strength of assortative mating as the correlation in vegetative phenologies between mates. We used the genetic data of maternal progenies (seedlings of the common garden) and adults trees to run paternity analyses and identify mating pairs, i.e. the most likely father siring a known mother. We used the genotypes of all the sampled adult trees in situ (147 trees at plot N1-LOW, and 192 trees at plot N4-HIGH) and of 1414 seedlings growing in the common garden (694 seedlings from plot N1-LOW and 720 seedlings from plot N4-HIGH, for an average ~35.3 seedlings per mother tree.

379 The genotypes of seeldings and adults were scored at a combination of 13 microsatellite loci (Oddou-380 Muratorio et al. 2018). The number of alleles observed in each cohort was greater than 106. Combine 381 across all 13 loci, the exclusion probability of a non-father was > 0.9999 at both plots. Paternity assignments 382 were conducted using the maximum-likelihood procedure implemented in the software CERVUS v.3.0.7 383 (Marshall et al. 1998; Kalinowski et al. 2007). Likelihood scores, based on allele frequencies in the 384 experimental population, were calculated for each seed /potential father couple. To determine whether 385 the paternity of each offspring could be assigned to the father with the highest likelihood, we used the 386 difference in likelihood scores (Δ LOD) between the two most likely pollen donors. The critical value (Δ C) of 387 ΔLOD below which paternity/parentage could not be assigned at 80% was determined using a distribution 388 of Δ obtained from 5 000 simulated mating events. This distribution was generated using the following 389 simulation parameters: 1% of genotyping error and no unsampled parents. Indeed, considering 390 simultaneoulsy the risk of genotyping error and the subsampling of the breeding male population may 391 inflate the lack of power in detecting the true father although it was sampled (type II error rate)-(Oddou-392 Muratorio et al, 2003).

393 Sexual selection analyses on adult trees, *in situ*

To investigate sexual selection, we used Bateman gradient analysis (Bateman 1948; Tonnabel et al. 2019), with a proxy of mating opportunities as predictor (here, phenological mismatch), and MEMM estimates of fecundity as the response variable.

397 <u>We followed the same strategy and methods as described above for fecundity selection. We followed</u> 398 the same strategy as for fecundity selection. For each sex (male and female) and plot (N1-N1-LOW and 399 N4N4-HIGH), we fitted seven models as described by equations (1) to (7), but replacing TBB by PMis, the 400 phenological mismatch between each tree and its neighbors in a 20 m radius. For instance, for the first 401 model:

402 (8) $M1.sex: \ln(F) = \alpha + \beta PMis + \varepsilon$

We compared the seven models based on the AICc, and selected the most parsimonious model,denoted BestSexSel-Model in the following

405 We finally fitted a compound model, derived from the BestFec-Model but adding PMis as predictor, 406 and we compared the BestFec-Model, the BestSex-Model, and the compound model.

407 Seedlings growth measurements

408 We measured seedlings diameter (D2011start, D2011end) and height (H2011start, H2011end) in the 409 common garden on two dates (respectively April 2011, and September 2011). This allowed us to estimate 410 diameter growth in 2011 as GrowthD = D2011_{end} - D2011_{start} and height growth in 2011 as H2011_{end} -411 H2011_{start}. In total, growth was measured in 2011 for 1552 seedlings originating from 20 families at plot 412 N1N1-LOW, and for 1709 seedlings originating from 20 families at plot N4N4-HIGH. Note that the trial was 413 divided in two contrasted experimental conditions: These seedlings were grown either in "watered" 414 condition (from block 1 to 25, 1652 seedlings) versus or in "water-stressed" condition (from block 26 to 415 50, 1609 seedlings).

416 Growth selection analyses on seedlings, in the common garden

As the common garden was designed to minimize seedlings mortality, we focused on growth as a performance trait related to viability, which is particularly expected when competition is homogeneous among seedlings (Collet and Le Moguedec 2007). We used the following mixed model to investigate the effect of phenology and plot on annual growth in diameter and height (respectively *GrowthD* and *GrowthH*) during year 2011:

422 (9) $M1_{viabSel}$: GrowthD or GrowthH = (*TBB in Plot*) + D2011orH2011 + Family + Block

423 Where D2011 (respectively H2011) is the initial diameter (respectively height) in spring 2011, 424 introduced to account for difference of vigor among seedlings. We tested for the effect of TBB nested 425 within plot (N1-N1-LOW or N4) to account for the fact that the effect of TBB on growth may differ among 426 plots (knowing moreover that TBB is on average higher for seedlings at plot N1-N1-LOW as compared to 427 plot N4, Gauzere et al. 2020a). Family (the maternal family of the seedlings) and Block (the trial unit to 428 which seedlings belongs to) were introduced as random factors to remove undesirable variation in growth 429 related respectively to genetic variation for phenology and microenvironmental effects (e.g., half the bocks 430 received a water-stress treatment).

- 431 We also tested another mixed model:
- 432 (10) $M2_{viabSel}$: GrowthD/H = (TBB in Plot × Treatment) + D/H2011 + Family + Block

where the treatment (watered vs water-stress) was specified as a fixed effect, in order to investigatewhether the effect of TBB on growth may differ among plots and among treatments.

These models were fitted with the function *lmer* in *lme4* package <u>(Bates et al. 2015)</u>. All the analyses are available as online supplementary material (file GrowthSelectionAnalysis_seedlings.html<u>at</u>).

Results

438 **Preliminary examination of interindividual Interindividual** variations in phenology and fecundity

439 The timing of budburst (TBB) was observed to spread over 17 days at plot N1-LOW, with a mean TBB 440 on April 20th (Fig. S3). At plot N4-HIGH, TBB was observed to spread over 13 days, with a mean TBB on May 441 4thTBB ranged from 103 to 121 (over 17 days) at plot N1 (with a mean TBB on the 20th of April) and from 442 117 to 130 (over 13 days) at plot N4 (with a mean TBB on the 4th of May; Fig. S3). Plot N4-HIGH showed aA 443 smaller inter-trees variance of TBB-was observed in Plot N4, with an important-significant proportion of 444 trees with the same TBB of 124. Larger trees at plot N1-LOW At plot N1, larger trees had an earlier budburst 445 (corr_{TBB-circ} = -0.15, p-value=0.007), while there was no significant relationship between size and TBB at plot 446 N4-N4-HIGH (corr_{TBB-circ} = -0.02, p-value=0.12). The within-tree-spread of budburst within a tree was higher 447 at plot N1-N1-LOW (mean spread = 4.8 days) than at plot N4-N4-HIGH (mean spread=2.9 days, Fig. S4A). 448 Trees with a later budburst also showed a higher spread of budburst at plot N1 (corr_{TBB-spread} = 0.57, p-449 value< 10⁻³) but not at plot N4-N4-HIGH (corr_{TBB-spread} = 0.09, p-value=0.20; Fig. S4B).

450 Male fecundities, as estimated by MEMM, followed a strongly L-shaped distribution (Fig. S5A). At plot 451 N1-LOW, male fecundities (MF) ranged from 0.013 to 10.5 (median = 0.33) and 97 trees (66%) exhibited a 452 non-negligible male fecundity. At plot N4-HIGH, male fecundities ranged from 2.10^{-3} to 16.6 (median = 453 0.016) and 69 trees (36%) exhibited a non-negligible male fecundity. Female fecundities, as estimated by 454 MEMMseedlings, also followed a strongly L-shaped distribution (Fig. S5B). At plot N1-LOW, female fecundities ranged from 3.10⁻³ to 13.3 (median = 0.014), and 55 trees (37%) exhibited a non-negligible 455 456 female fecundity. At plot N4-HIGH, female fecundities ranged from 3.10⁻⁴ to 25.8 (median = 0.005), and 30 457 trees (16%) exhibited a non-negligible female fecundity.

The distribution of male fecundities estimated by MEMM was strongly L-shaped (Fig. S5A). At plot N1,
 male fecundities (MF) varied from 0.013 to 10.5 (median = 0.33) and 97 trees (66%) had a non-negligible
 male fecundity. At plot N4, male fecundities varied from 2.10⁻³ to 16.6 with MEMM (median = 0.016) and
 69 trees (36%) had a non-negligible male fecundity.

The distribution of female fecundities estimated by MEMMseedlings was also strongly L-shaped (Fig. S5B). Female fecundities varied from 3.10⁻³ to 13.3 at plot N1 (median = 0.014), and from 3.10⁻⁴ to 25.8 at plot N4 (median = 0.005). The number of trees with a non-negligible female fecundity was 55 (37%) at plot 465
 N1, and 30 (16%) at plot N4.

466 Fecundity selection analyses for TBB, based on adult trees in situ Fecundity selection on TBB

467 The study demonstrated that earlier budburst had a positive effect on female fecundity. This was 468 observed in all trees at high altitude and in larger trees at low altitude (Table 3). The best model for female 469 fecundity at plot N1-LOW included TBB, size, competition and their interactions. Delayed TBB had a 470 negative impact on female fecundity in the larger trees, but a positive impact on smaller trees, as illustrated 471 by figure 2A. Delayed TBB had a significant negative effect on female fecundity for the more competed 472 trees (Fig. 2B). The directional selection gradient estimated for TBB was marginally significant ($\beta_{\text{TBB}} = -0.37$, 473 p = 0.07; Table 3), indicating a positive effect of earlier TBB on female fecundity for a tree with average 474 DBH and average competition.

475 We selected the most parsimonious model among the seven that were fitted (Table S2), and visually 476 inspected the residuals (Fig. S11). Then, considering this best model, we evaluated the strength and 477 direction of directional fecundity selection on TBB through the effect of TBB on fecundity (Table 2). For 478 female fecundity at plot N1, the best model included TBB, size, competition and interactions among TBB 479 on the one hand and size and competition variables on the other hand. Delayed TBB significantly decreased 480 female fecundity for the larger trees, but tended to increase it for smaller trees (Fig. 3A). Delayed TBB 481 significantly decreased female fecundity for the more competed trees (Fig. 3B). Overall, we estimated a 482 marginally significant directional selection gradient for earlier TBB (β_{TBB} = -0.37, p = 0.07; Table 2), i.e. for 483 a tree with average DBH and average competition. 484 The best model for female fecundity at plot N4-HIGH included TBB, size, and competition, as well as

the interaction between TBB and size, although For female fecundity at plot N4, the best model included
 the interaction between TBB and size, although For female fecundity at plot N4, the best model included
 TBB, size, competition and interaction among TBB and size, but this interaction term was not significant.
 As the second-best model, without the competition term, performed nearly as well as nearly as the best
 one (ΔAICc = 0.56), we favored parsimony and kept it. Our results suggest that delayed TBB and increased

489 competition may decrease female fecundity (β_{TBB} =-0.34, p=0.052; $\beta_{ConMartin20}$ =-0.43, p=0.05), while female 490 fecundity increased with tree size (β_{MaxDbh} =0.78, p<0.001). Delayed TBB and increased competition tend to

491 decrease female fecundity (β_{TBB} =-0.34, p=0.052; $\beta_{\text{conMartin20}}$ =-0.43, p=0.05), while female fecundity 402 increased with tree size (β_{TBB} =-0.78, p=0.001)

492 increased with tree size ($\beta_{MaxDbh}=0.78$, p<0.001).

493 **Table 2**<u>Table 3</u>. -Fecundity selection on female (*F*₂) and male (*F*₃) fecundities at plots <u>N1N1-LOW</u> and

494 N4N4-HIGH. Fecundity selection on phenology was assessed through the effect of TBB on fecundity,
 495 accounting for joint effect of size and competition (see variable names in Table 1Table 1). We selected the
 496 most parsimonious regression model for each sex and plot (Table S2) to estimate the effect's coefficient
 497 and Sum Of Squares (SOSq) associated with each term. Significance of the effect (AIC and *p*-values) was

498 assessed with the R function "drop1". F_{\bigcirc} and F_{\bigcirc} were log-transformed.

Term	Coefficient	SofSq	AIC	p-value
<i>F</i> ♀, plot N1<u>N1-LOW</u>	Adjusted R ² =0.1414; p-valu	ie<0.001 ; AIC _{ref} =262	2.85	
ТВВ	-0.37	18.91	264.24	0.072
MaxDbh	0.78	61.06	271.53	0.001
ConMartin20	0.06	0.36	260.91	0.802
TBB:MaxDbh	-0.88	54.02	270.34	0.003
TBB:ConMartin20	-0.55	32.28	266.59	0.019
<i>F</i> ♀, plot N4<u>N4-HIGH</u>	Adjusted R ² =0.145; p-value	e<0.001 ; AIC _{ref} =370.	43	
ТВВ	-0.34	21.97	371.09	0.071
MaxDbh	0.78	87.06	380.67	0.000
ConDens20	-0.37	19.26	370.68	0.091
<i>F</i> ♂, plot N1<u>N1-LOW</u>	Adjusted R ² =0.119; p-value	e<0.001; AIC _{ref} =112.0)1	
ТВВ	-0.08	0.88	110.44	0.519
SumDbh	0.56	45.31	130.54	<0.001
<i>F</i> ♂, plot N4<u>N4-HIGH</u>	Adjusted R ² =0.176; p-value	e<0.001; AIC _{ref} =336.8	30	
ТВВ	-0.41	31.78	340.50	0.019
SumDbh	0.46	33.64	340.83	0.015
Stature	Dom : 0.31	96.33	349.60	<0.001
	Cod : 0.74			
	Suppr : -1.05			

499In contrast, earlier budburst only increased male fecundity at high altitude. The best model for male500fecundity at plot N1-LOW showed a marked increase in male fecundity with tree size (β_{SumDbh} =0.56,501p<0.001), but no significant effect of TBB. At plot N4-HIGH (Fig. 3B), delayed TBB decreased male fecundity</td>502(β_{TBB} =-0.41, p=0.019). Male fecundity also increased with tree size (β_{SumDbh} =0.46, p=0.015), and depended503on tree stature, with higher fecundity for codominant and dominant trees. Table S2 displays the results of504all the fitted models of fecundity selection while Fig. S6 shows the residuals of the best models

505 In the best model for male fecundity at plot N1, the effect of TBB was not significant, while male 506 fecundity markedly increased with tree size (β_{sumDbh} =0.56, p<0.001). By contrast, the effect of TBB was 507 significant in the best model for male fecundity at plot N4 (Fig. 4B), and delayed TBB decreased male 508 fecundity (β_{TBB} =-0.41, p=0.019). Male fecundity also increased with tree size (β_{sumDbh} =0.46, p=0.015), and 509 depended on tree stature, with higher fecundity for codominant and dominant trees.



510

Figure 3-2 – Interaction effects between TBB and size (A) or competition (B) on female fecundity at low
 elevationaltitude (plot N1N1-LOW). Predicted regression lines are plotted for three values of each
 moderator variable, corresponding to +/- 1 standard deviation from the mean. Confidence interval at
 95% are shown around each regression line. Dots are the observed values



515

Figure 4-<u>3</u> – Relationship between TBB and female fecundity (A) or male fecundity (B) at high
 elevationaltitude (plot N4N4-HIGH). The lines are the predictions with their 95% confidence intervals,
 and triangles are the observed values.

519 <u>Estimation of Aa</u>ssortative mating and phenological mismatch



Figure 2-4 – Join distribution of parent pairs' phenological score (PSS) under random mating (A), and in
 realized mating events (B) at plot N1. A. The density of the data cloud was computed under the
 hypothesis that each tree mates once as male and once as female with all possible trees. B. Paternity
 analyses of seeds sampled on mother-tree allowed to identify mates' pairs. See also Fig. <u>S6S7</u>.

532 The phenological mismatch with neighbors, measured by |PMis|, and |PMis|, was slightly higher at 533 plot N1 as compared to N4 (Fig. S7, S8): for instance, in a 20 m radius, $\mu_{IPMisIm}$ = 4.52 days at plot N1 while 534 HIPMISTOR = 2.12 days at plot N4. The phenological mismatch was more variable at plot N4 than at plot N1: 535 for instance, in a 20 m radius, cv_{IPMisIs} = 1.05 at plot N4 versus cv_{IPMisIs} = 0.54 at plot N1. For the analyses 536 that follow, we selected | PMis|s within a 20 m radius, referred to as PMis hereafter, as the most accurate estimator of phenological mismatch due to its high variation (Table S1). At plot N1-LOW, PMis decreased 537 538 as phenological spread increased (corr_{PMis-spread} = -0.23, p-value=0.004). Conversely, the opposite trend was 539 observed at plot N4-HIGH (corr_{PMis-spread} = 0.12, p-value=0.09, Fig. S8A). The relationship between TBB and 540 PMis was found to be quadratic, with a TBB value that minimized PMis (Fig. S8B). This result is expected if 541 TBB is not strongly spatially structured (Fig S9). For the following analyses, we selected |PMis|s in a radius 542 of 20 m as the best estimator of phenological mismatch (i.e., the estimator with the highest variation, Table 543 S1) and we denote it PMis in the following for the sake of simplicity. At plot N1, PMis decreased with 544 increasing phenological spread (corr_{PMis spread} = -0.23, p-value=0.004) while a reverse trend occurred at plot 545 N4 (corr_{PMis spread} = 0.12, p-value=0.09, Fig. S9A). Finally, the relationship between TBB and PMis was 546 quadratic, with a TBB value minimizing PMis (Fig. S9B), which is expected if the TBB variable is not strongly 547 spatially structured (Fig S10). The distributions of |PMis|s and |PMis|m at each plot can be seen on Figures 548 S10 and S11.

549 Sexual selection analyses for phenological mismatch, based on adult trees in situ.

550 Sexual selection

558

551 Only male and not female fecundities-fecundity variation were was significantly affected by PMis (Fig. 552 5). In the best models, male fecundity decreased with increasing PMis both at plots N1N1-LOW ($\beta_{PMis}=-$ 553 0.44, p<0.001) and N4-N4-HIGH ($\beta_{PMis}=-0.45$, p<0.001). Despite similar selection gradient values at both 554 plots, the distribution of observed values of TBB and fecundity suggest stronger sexual selection at plot N1, 555 in line with the stronger signal of assortative mating. Besides, male fecundity increased with SumDbh at 556 both plots, and for tree with codominant stature at plot N4. The results of all the fitted models of sexual 557 selection are shown in Table S3, and the residuals of the best models in Fig. S12.

559 Finally, wWe also tested compound best models, where both TBB and PMis were included as factors in 560 the best model for fecundity selection. Only for male fecundity at plot N4-N4-HIGH did the compound 561 model outcompeted the BestFec and BestSex models (Table S4). The effect of TBB in the compound model 562 $(\beta_{TBB}=-0.34, p=0.045)$ was very similar to that of TBB in the BestFec model, showing that sexual selection 563 does not impact does not affect the estimate of fecundity selection. For Regarding female fecundity, the 564 effect of PMis was not significant in the compound model and the effect of TBB did not differ from that of 565 the BestFec models, showing that sexual selection does not impact does not affect the estimate of 566 fecundity selection. For male fecundity at plot N1N1-LOW, the effect of TBB was not significant in the 567 compound model and the effect of PMis was not different from that did not differ from that of the BestSex 568 model.

569

Table 34: Sexual selection on female (F_{\uparrow} **) and male (** F_{\Diamond} **) fecundities at plots N1N1-LOW and N4N4-HIGH.** Sexual selection on phenology was assessed through the effect of phenological mismatch (PMis) on fecundity accounting for joint effect of size and competition. We selected the most parsimonious regression model for each sex and plot (Table S3) to estimate the effect's coefficient and Sum Of Squares (SOSq) associated with each term.-Significance of the effect (AIC and *p*-values) was assessed with the **R** function "*drop1*". F_{\Diamond} and F_{\Diamond} were log-transformed.

Term	Effect	SofSq	AIC	p-val			
<i>F</i> ္, plot N1	Adjusted R ² = 0.088; p-value	e<0.001; AIC _{ref} =268.8	36				
PMis	0.13	2.34	267.25	0.536			
MaxDbh	0.85	95.81	282.07	<0.001			
<i>F</i> ္, plot N4	Adjusted R ² = 0.117; p-value	e<0.001; AIC _{ref} =373.9	91				
PMis	-0.2908	16.119	374.27	0.128			
MaxDbh	0.9382	167.782	395.14	<0.001			
F♂, plot N1	Adjusted R ² =0.199; p-value	Adjusted R ² =0.199; p-value<0.001; AIC _{ref} =98.04					
PMis	-0.4436	28.305	110.44	<0.001			
SumDbh	0.6072	53.031	121.963	<0.001			
F♂, plot N4	Adjusted R ² =0.181; p-value	<0.001 AIC _{ref} =335.73	3				
PMis	-0.45	37.64	340.50	0.010			
SumDbh	0.42	28.28	338.84	0.026			
Stature	Dom : 0.31	117.71	352.18	<0.001			
	Cod : 0.86						
	Suppr : -1.17						

576

Plot 🗲 Low elev. (N1) 📥 High elev. (N4)



Figure 5 – Relationship between phenological mismatch and male fecundity (Bateman's gradient). The higher the phenological mismatch, the lower the opportunities for mating. The phenological mismatch, PMis, was estimated as the sum of absolute difference in TBB between a tree and each of its neighbors in a 20 m radius. Symbols represent observed values (square: plot N1N1-LOW; triangle: plot N4N4-HIGH) and lines are the prediction of the best sexual selection model.

577 Estimation of stabilizing selection and standardized selection gradients on phenology

578 Stabilizing selection and selection gradients on phenology

579 We found no evidence of stabilizing selection on TBB through a significant effect of TBB² <u>onn</u>either on 580 female nor on male fecundity <u>(Online Appendix 1)</u>. However, the significant effect of PMis on male 581 fecundity at plot <u>N1N1-LOW</u> and <u>N4-N4-HIGH illustrates a form of stabilizing selection on TBB, as lower</u> 582 <u>PMis are obtained for average TBB due to the quadratic relationship between PMis and TBB (Figure 583 <u>S8B)</u>.results de facto in a stabilizing selection on TBB since lower PMis are obtained for average TBB (e.g. 584 quadratic relationship between PMis and TBB, Figure S8B).</u>

Standardized selection gradients (Table S5) <u>show-suggest</u> that selection for earlier TBB through female fecundity <u>was-is slightly</u> higher at plot <u>N4-N4- high</u> (β_{TBB} '=-0.43) than at plot <u>N1N1-LOW</u> (β_{TBB} '=-0.24), allthough these differences are not significant due to large standard errors. Selection for earlier TBB through male fecundity at plot <u>N4-N4- high</u> was within the same order of magnitude (β_{TBB} '=-0.30) than through female fecundity. Finally, the directional selection for reduced phenological mismatch with neighbors was <u>slightly</u> higher at plot <u>N1N1-LOW</u> ($\beta_{|PM|}$ '=-0.40) than at plot <u>N4-N4- high</u> ($\beta_{|PM|}$ '=-0.16), although these differences are also not significant.

592 Growth selection analyses for TBB, based seedlings in the common garden

593 Impact of phenology on seedlings growth

Growth selection analyses revealed a significant effect of TBB on seedlings growth (Table 45): both diameter and height growth significantly decreased with delayed budburst (increasing TBB). Moreover, growth increased with increasing initial size, and growth was reduced for seedlings originating from plot N4-N4-HIGH compared to those from plot N1N1-LOW. As expected, the variance in growth was significantly structured by block and family (Table S6). A more detailed analysis showed an the expected strong negative effect of water stress on growth. Moreover, the negative effect of delayed budburst on growth (although albeit-much lower than that of treatment) was higher in the water-stress treatment (Table S7).

601Table 45: Selection on seedling growth in diameter (Dgrowth) and height (Hgrowth). Selection on602phenology was assessed through the effect of TBB on seedling growth, accounting for the effects of plot603(N1N1-LOW or N4N4-HIGH), initial size (initD or initH), and common garden design (with Block and Family604included as random effects). The global significance of each fixed term was assessed based on the Sum Of605Squares (SoSq) and the F-value (F-test), while the effect of TBB within each plot was assessed based on the606t-value (Student test). See Table S7 for a more complex model including treatment.

607 A- Diameter growth

Term	npar	SoSq	F-value	p-value	Effect	St. error	t-value	p-value
Plot	1	2.86	4.12	0.042	-0.345	0.387	-0.89	
initD	1	228.76	330.18	<0.001	0.291	0.016	18.66	
Plot:TBB11	2	15.80	11.40	<0.001	N1: -0.015	0.006	-2.80	0.005
					N4: -0.021	0.005	-3.94	<0.001

608

609 *B-* Height growth

Term	npar	SoSq	F-value	p-value	Effect	St. error	t-value	p-value
Plot	1	67549	11.83	0.001	-76.940	35.256	-2.18	
initH	1	17753	3.11	0.078	0.046	0.020	2.33	
Plot :TBB11	2	53894	4.72	0.009	N1: -0.188	0.511	-0.37	0.712
					N4: -1.545	0.504	-3.07	0.002

610

Discussion

611 In this study, we estimated fecundity, sexual and viability selection by combining field and common 612 garden data with parentage analyses in order to better understand the selection regime on spring 613 phenology in European beech. Our main results were that fecundity selection on female fitness and viability 614 selection on seedlings growth both favor early phenology, while sexual selection on male fitness through 615 assortative mating modulates this trend (stabilizing selection). Furthermore, this study confirmed that 616 environmental differences (here, altitudinal differences) can also have a major impact on the potential for 617 contemporary evolution. This study confirms the interest of in situ phenotypic selection analyses to better 618 understand the evolutionnary potential of tree populations (see also Bontemps et al. 2017; Alexandre et 619 al. 2020; Westergreen et al. 2023).

620 Earlier budburst increases female fecundity and seedling growth, with contrasted effects of drought 621 stress

622 We showed that earlier budburst increased the female fecundity at both elevations (of all trees at high 623 elevation, and of the larger trees at low elevation) and the male fecundity only at high elevation. This-Our 624 findings that earlier budburst increased female fecundity of adult trees in situ and seedlings growth in the 625 common garden is consistent with the pervasive phenotypic selection for early reproductive phenology 626 documented in plants (Geber and Griffen 2003; Munguía-Rosas et al. 2011; Austen and Weis 2015).-These 627 findings also contradict the expectation of stabilizing selection on vegetative and reproductive phenology, 628 driven by the balance between the benefits of avoiding frost damages on the one hand and maximizing the 629 duration of the growing season on the other hand . Hence, we seem to face a similar paradox to the one 630 observed for flowering phenology in short-lived plants, and for which Austen et al. (2017) already proposed 631 four explanations: (1) selection through other fitness components may counter observed fecundity 632 selection for early flowering; (2) asymmetry in the flowering-time-fitness function may make selection for 633 later flowering hard to detect; (3) flowering time and fitness maybe condition-dependent; and (4) selection 634 on flowering duration is largely unaccounted for. Before detailing how this study shed ligth on mechanisms 635 related to explanation (1) (see the second paragraph of this discussion), we can first add to this list a fifth 636 possible explanation related to temporally fluctuating selection in long-lived plants. Indeed, as we 637 estimated selection during a single reproductive episode, we can not exclude that other selection patterns 638 may be observed in different years, due to year-specific climatic conditions. A review already suggested 639 that changes in selection direction across years are common in vertebrates (Siepielski, Dibattista, & 640 Carlson, 2009). In our case in particular, we may not have been in favorable conditions to observe selection 641 for later budburst driven by late frosts, as they did not occur in the year when we sampled seeds and 642 seedlings for this study. Selection for later budburst through late frosts damages can be expected as a 643 strong selection force (Westergren et al. 2023), as late frosts can strongly reduce the photosynthetic 644 capacities of adult trees and hence their seeds' development and maturation and/or reducing seedling 645 survival; however, this selection could occur only occasionally in balance of other selections that apply 646 every year with a more moderate intensity. Finally, it should be noted that among the possible sources of 647 selection generally not accounted for in classical phenotypic selection analyses are those related to 648 interspecific interactions. For instance, in this multispecies ecosystem at ecological margin between 649 Mediterranean and mountainous climates, interspecific competition could participate to shape patterns of 650 selection, and reinforce the trend for beech to flush earlier than competing species (Palacio-Lopez et al., 651 2020). 652 We nonetheless found one evidence opposing the general pattern of fecundity selection for earlier TBB

653 through female fecundity. At low elevationaltitude, directional fecundity selection for early budburst was 654 found only for the larger trees, while the smaller ones tend to show the reverse pattern, or at least, no gain 655 in fecundity with early budburst. Larger trees also had an earlier budburst, resulting in a consistent signal 656 of directionnal selection for TBB and for this size category. These Such contrasted selection gradients on 657 TBB among neighboring trees suggest that different ecological strategies exist within the same drought-658 prone population, likely due to some micro-environmental heterogeneity. These two strategies This pattern 659 can be related to the "growth-stress survival" trade-off (Grime 1977; Grubb 1998), whereby slower 660 development (small trees) and delayed budburst can be viewed as a drought tolerance strategy. Indeed This 661 is consitent with, a previous study at the same low elevational titude plot, found that where trees displaying late budburst were also associated with low size, low leaf water content and other traits (e.g., high leaf mass per area) symptomatic of a water-saving strategy while trees displaying early budburst were also associated with large size, high leaf water content and other traits (e.g. low water use efficiency) symptomatic of a water-uptake strategy (Bontemps et al. 2017). By contrast with this "growth-stress survival" trade-off for adult trees facing variable levels water stress in situ, we found the reverse trend in the common garden, where the positive relationship between

668 Finally, we found selection for earlier spring phenology using growth measurements from a common 669 garden experiment, with seedling showings early budburst also having and higher seedlings growth. This 670 association was slightly stronger in the drought-stress treatment of the experiment., which is at odd with 671 the above-described "growth-stress survival" trade-off. These last_results for seedlings in the common 672 garden-can be interpreted as the fact that an early budburst allows seedlings to start photosynthesizing 673 when the conditions are the most optimal for growth (i.e., before drought) and can be view as a "drought-674 escape" strategy. Taken together, Oour findings hence suggest that the patterns of selection on phenology 675 may change across ontogenic developmentstages of life history (reviewed in Schluter et al. 1991). Vitasse 676 (2013) had for instance already showed that the earlier ontogenic stage of seedlings in the understory 677 explain their earlier leaf emergence. Here, we suggest that adaptive response to drougth may differ $_{7}$ 678 eventually due to different challenges in terms of survival (and fecundity) for juvenile versus mature trees. 679 (Vitasse 2013).

680 Mating opportunities limit male fecundity, and drive stabilizing selection on TBB

681 Another main finding of this study in line with the first explanation put forward by Austen et al. (2017) 682 is that stabilizing selection on male mating success through assortative mating can modulate fecundity 683 selection for earlier phenology. To begin with, I this study is among the first ones to show and estimate 684 assortative mating on spring phenology in a tree species. Moreover, and consistently with Bateman's 685 principle, we found that increasing phenological mismatch with neighbors, as a proxy of decreasing mate 686 opportunities, affected male but not female fecundity. Hence, variation in phenology among trees within 687 stand create opportunities for sexual selection, and could can drive stabilizing selection on TBB through 688 the male function. Such stabilizing selection was already observed in a pollen-limited population of Quercus 689 lobata, where trees that flowered early or late set fewer acorns than trees that flowered at the population's 690 peak (Koenig et al., 2012). Our study generalizes this result in cases where pollen is not limiting fruit outset.

691 Assortative mating has important evolutionary and ecological consequences (Jiang et al. 2013), and 692 assortative mating for phenological traits in particular can markedly affect the evolutionary response to 693 climate change (Godineau, Ronce, & Devaux, 2021; Soularue & Kremer, 2014; Whittet et al., 2017). 694 However, the standard measure of assortative mating based on the observation of individual synchronicity 695 of flowering schedules (Weis et al. 2005, 2014) is hardly applicable in forest trees. Hence, the potential 696 assortative mating for phenological traits had been mostly investigated between tree populations, by 697 measuring the difference in the timing of pollen shedding among oak populations along temperature clines 698 (Whittet et al. 2017) or by inferring the latitudinal origin of pollen in pine open-pollinated progenies grown 699 in common gardens (Nilsson 1995). Another common approach is to estimate mating system parameters 700 using genetic markers; such studies proposed assortative (respectively disassortative) mating as a general 701 mechanism driving higher (respectively lower) relatedness between mated individuals than expected by 702 chance (Hardy et al., 2019; Ismail & Kokko, 2020; Monthe et al., 2017). Here, we applied the approach 703 widely used in animal species to quantify assortative mating (e.g., Jiang et al. 2013): we first used paternity 704 analyses to infer mated pairs *a posteriori*, and secondly, we computed the correlation of spring phenology 705 across members of mated pairs. This is one of the rare studies to our knowledge to evaluate effective 706 assortative mating for spring phenology in a tree species, by combining budburst phenology data and 707 marker-based paternity analyses (see also Gérard et al. 2006; Lagache et al. 2014; Larue et al. 2022). Our 708 approach showed significant assortative mating for spring phenology at the lower plot, where budburst 709 spread over 17 days. The correlation in vegetative phenology between mating pairs was moderate (ρ =0.19) 710 as compared to the range reported in the literature (e.g. 0.05–0.63 within the same old-field community, 711 Weis et al. 2014). At the upper plot, it is likely because the guicker development of leaf unfolding was 712 quicker (with a which spread over-only 13 days range) may explain why that assortative mating was absent, 713 or too weak to be not detected, all though we cannot rule out that other factors, such higher canopy density 714 at higher altitude, constraint mating.

715 As we expected, because of assortative mating, we found that a timing of flowering synchronized with 716 close neighbors maximizes mating success through the male function, but does not significantly affect the 717 female fecundity. Indirectly, this favors intermediate timing of bud burst since in absence of a strong spatial 718 structure, males with intermediate TBB are those most synchronized with their neighbors. To our 719 knowledge, this study is the first to test and validate the Bateman's principle in a tree, likely due to the 720 difficulty of estimating the number of mates in these species producing a large number of offspring. We 721 used the phenological mismatch as a proxy of (potential) mating opportunities rather than the mating 722 success that could have been estimated based on paternity analyses (Tonnabel et al. 2019), because our 723 sampling design, with only 35.3 seeds per mother tree, may underestimate the contribution of rare fathers. 724 The effect of phenological mismatch could be related to stabilizing selection on TBB, since the phenological 725 mismatch is a quadratic function of TBB. However, and surprisingly, we did not retrieve the expected 726 consequence of a significant quadratic relation between effective fecundity and TBB. This could be due to 727 different abilities to detect significant linear coefficients (from the slope of the regression line) as compared 728 to quadratic coefficients (from the curvature of the fitness surface).

729 The observed effect of mating opportunities on MEMM-estimates of fecundity is counter-intuitive, as 730 these estimates are claimed to be effective estimates of basic fecundity (Oddou-Muratorio et al. 2018). 731 This is likely because the effect of the phenological mismatch is not included in the MEMM model we used; 732 thus, any effect of phenological mismatch on individual reproductive success will be retrieved into the 733 estimate of individual fecundity. In the same way that MEMM models the effect of the relative positions 734 of putative parents and offspring on fecundity through the pollen dispersal kernel (spatial assortative 735 mating), we could also model in MEMM the effect of phenological mismatch on fecundity (temporal 736 assortative mating, as done in Gérard et al. 2006; Gleiser et al. 2018; Larue et al. 2022). Thereby, the 737 estimated fecundity would no longer depend on the mating opportunities. This option would be interesting 738 to include in future developments of MEMM.

Altitudinal variation of selection on spring phenology and overall evolutionary potential of the studied
 beech population Implications of the observed selection on spring phenology for beech adaptation to
 climate variations

742 A main result of this study is that early phenology is associated to high reproductive outputs and high 743 seedling growth in the studied European beech population. As hypothetised by Austen et al. (2017), we 744 nonetheless found two evidence of selection countering this observed fecundity and growth selection for 745 early phenology: first, stabilizing selection through assortative mating for male mating success, and second, 746 selection for later TBB through female fecundity as a drought-tolerance strategy. Moreover, as we 747 estimated selection during a single reproductive episode, we can not exclude that other selection patterns 748 may be observed in different years. In particular, we may not have been in favorable conditions to observe 749 selection for later budburst driven by late frosts, as they did not occur in the year when we sampled seeds 750 and seedlings for this study. Selection for later budburst through late frosts damages can be expected as a 751 strong selection force (by reducing the photosynthetic capacities of adult trees and hence their seeds' 752 development and maturation and/or reducing seedling survival) but occurring only occasionally in balance 753 of other selections that apply every year with a more moderate intensity. Alternatively, it is also possible 754 that other selective forces, non-acconted for in this study or in general, may contribute to the observed 755 general pattern of selection for earlier TBB. For instance, in this multispecies ecosystem at ecological 756 margin between Mediterranean and mountainous climates, interspecific competition could participate to 757 shape patterns of selection, and reinforce the trend for beech to flush earlier than competing species 758 (Palacio-Lopez et al., 2020).

759 Although selection gradients for each component of selection (female fecundity, male fecundity, sexual 760 selection on male fecundity) did not differ significantly among altitudes, this study highlighted a number 761 of qualitative indications that selection for earlier phenology (i.e., precocity) is stronger overall at high 762 altitude than at low altitude in the population studied. First, selection for precocity through female 763 fecundity was reinforced by selection for precocity through male fecundity only at high altitude. Second, 764 selection for precocity through female fecundity was modulated by the interaction effect between size and 765 TBB only at low altitude. Third, assortative mating, the fuel for sexual stabilizing selection through male 766 mating success, was stronger at low altitude. Stronger selection for earlier phenology at high latitude is 767 consistent with the physiological expectation that the length of the growing season strongly constrains the

768 level of resources acquired through photosynthesis (Keenan et al. 2014; Richardson et al. 2006). It is also 769 consistent with the simulation study of Gauzere et al. (2020a) showing that selection for earlier budburst 770 is stronger under conditions that are more limiting to reproductive development, i.e., in cold environments. 771 Moreover, the selection strength on female fecundity was higher at high than at low elevation, and the 772 additional selection on male fecundity in plot N4 suggested even stronger 'global' fecundity selection at 773 high than low elevation. These results are consistent with the physiological expectation that the length of 774 the growing season strongly constraint the resource level acquired by photosynthesis, particularly at high 775 elevation. They also support the simulation study of Gauzere et al. (2020a) showing that selection for 776 earlier budburst dates is stronger in conditions more limiting for reproductive development, i.e., here in 777 cold environments.

778 On a quantitative point of view, the standardized directional selection gradients on spring phenology 779 estimated in this study (β') ranged between -0.43 and -0.24. This indicates a rather strong magnitude, 780 taking the meta-analysis of Kingsolver et al. (2001) as a reference (where a mean $|\beta'|$ -value of 0.22 was 781 found across all traits, with a median $|\beta'|$ -value of 0.08 for life-history/phenological traits). This 782 metanalysis also reported higher value of $|\beta'|$ for selection via fecundity or mating success (median $|\beta'|$ = 783 0.18) than for selection via survival (median $|\beta'| = 0.09$), supporting the strong directional selection 784 estimated here. Considering the high level of narrow-sense heritability estimated for phenological score 785 sum in the population at low altitude (h^2 =0.84–0.92; Bontemps et al. 2016), our results may indicate a high 786 evolutionary potential for spring phenology in the studied population. Such strong selection gradients are 787 likely to reflect strong selective pressures on phenology that may constraint population demographic 788 growth in both cold and warm environments. This supports the hypothesis that phenology is an important 789 determinant of survival and fecundity, consistent with studies that use it to predict the distribution range 790 of plant species (Chuine & Beaubien 2001, Gauzere et al. 2020a). However, the high evolutionary potential 791 of spring phenology measured in the studied beech population does not guarantee by itself its ability to 792 adapt to the multiple effects of ongoing climate change. In particular, there is increasing evidence that 793 emerging drought stress is causing massive mortality even in areas previously spared by drought 794 (Hartmann et al. 2022). Whether the genetic response of spring phenology to increased summer 795 temperature combined with extreme drougth stress will allow beech populations to adapt is difficult to 796 predict without a dedicated predictive modelling approach (e.g., Oddou-Muratorio & Davi 2014). However 797 our results show that accounting for genetic differences in phenological schedules and their ecological 798 significance can greatly improve scenarios of future population adaptation to drought and late frost stress. 799 -Such selection gradient may contribute to the fast adaptation of beech populations to the multiple 800 effects of ongoing climate change, and in particular to 1) the likely emergence of late frosts pressures 801 paralleling the advanced spring season at the high elevation site and 2) the increasing risk of hydraulic 802 failure with increased summer drought at the low elevation site, which may reshuffle the respective 803 benefits of the drought-resistance and drought-escape strategies. However, integrating the different 804 significant selection gradients on TBB estimated through male fecundity, female fecundity and seedling 805 growth at each plot and predicting the population response to selection would require a dedicated,

806 forward modelling approach that was out of the scope of this study.

807 Spring phenology has been defined as a "magic trait", which affects fitness through its influence on 808 growing season (and thus survival and fecundity) and simultaneously contributes to non-random mating 809 (Servedio et al., 2011; Soularue & Kremer, 2014). Previous simulations studies have demonstrated how 810 environmental variation can cause populations to diverge for a selectively neutral trait that causes 811 assortative mating (Kirkpatrick 2000; Soularue and Kremer 2012). Consequently, some patterns of clinal 812 genetic variation of in phenological traits observed in forest trees can be generated solely by the effects of 813 assortative mating and gene flow, without in the absence of divergent selection. When both divergent 814 selection and assortative mating for TBB occur, Soularue and Kremer (2014) predicted that genetic clines 815 can either be inflated or constrained by assortative mating, depending on species life history. Finally, a 816 recent study predicted the evolution of either suboptimal plasticity (reaction norms with a slope shallower 817 than optimal) or hyperplasticity (slopes steeper than optimal) for TBB in the presence of assortative mating, 818 whereas optimal plasticity would evolve under random mating (Soularue et al. 2022). These different 819 simulation studies considered prescribed, single-trait models of divergent selection, in which a single 820 optimal value maximizes fitness within each population. Given the intertwined effects of sexual, fecundity 821 and viability selection on phenology and the variation in fitness landscapes for budburst along temperature

822 823	and drought gradients shown in this study, we suggest that future eco-evolutionary models of phenological shifts should integrate these features in a mechanistic and multidisciplinary framework (Donohue et al.
824	2015, Lamarins et al. 2022). Such an approach could allow quantitative assessment of which type of
825	selection (viability, fecundity, sexual selection) currently dominates the selection regime on spring
826	phenology, and evaluate whether the genetic response to these different types of selection will allow
827	beech populations to adapt to ongoing climate change.
828	However, these simulation studies considered simple, single-trait models of divergent
829	selection, where a single optimal value maximizes fitness within each population. The
830	intertwining of sexual, fecundity and viability selection on phenology and the variation of
831	fitness landscapes for budburst along temperature and drought gradients demonstrated
832	in this study call for integrative, mechanistic and multidisciplinary studies of phenological
833	shifts. Only such approaches can account for the complexity of fitness landscapes, identify
834	the factors limiting response to selection, and ultimately anticipate the increased extreme
835	selective pressures associated with the ongoing climate change.
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841	Data, scripts, code, and supplementary information availability
842	Data (two files: dataAdultField.txt and dataGrowthCommonGarden.tab), Scripts of statistical analyses
843	(two files: SelectionAnalyses_adult.html and GrowthSelectionAnalysis_seedlings.html) and Supplementary
844	figures and tables (one file: SuplMaterial_V10.pdf) are available online: https://doi.org/10.57745/ZVPNXX
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846	The authors declare that they comply with the PCI rule of having no financial conflicts of interest in
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