

1 Quantitative genetic architecture of adaptive
2 phenology traits in the deciduous tree, *Populus*
3 *trichocarpa* (Torr. & Gray).

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21 ABSTRACT

22

23 In a warming climate, the ability to accurately predict and track shifting
 24 environmental conditions will be fundamental for plant survival. Environmental
 25 cues define the transitions between growth and dormancy as plants synchronise
 26 development with favourable environmental conditions, however these cues are
 27 predicted to change under future climate projections which may have profound
 28 impacts on tree survival and growth. Here, we use a quantitative genetic approach to
 29 estimate the genetic basis of spring and autumn phenology in *Populus trichocarpa* to
 30 determine this species capacity for climate adaptation. We measured bud burst, leaf
 31 coloration, and leaf senescence traits across two years (2017- 2018) and combine
 32 these observations with measures of lifetime growth to determine how genetic
 33 correlations between phenology and growth may facilitate or constrain adaptation.
 34 Timing of transitions differed between years, although we found strong cross year
 35 genetic correlations in all traits, suggesting that genotypes respond in consistent ways
 36 to seasonal cues. Spring and autumn phenology were correlated with lifetime growth,
 37 where genotypes that burst leaves early and shed them late had the highest lifetime
 38 growth. We also identified substantial heritable variation in the timing of all
 39 phenological transitions ($h^2 = 0.5-0.8$) and in lifetime growth ($h^2 = 0.8$). The
 40 combination of abundant additive variation and favourable genetic correlations in
 41 phenology traits suggests that cultivated varieties of *P. trichocarpa* have the
 42 capability to create populations which may adapt their phenology to climatic changes
 43 without negative impacts on growth.

44

45 INTRODUCTION

46

47 Perennial plants transition between periods of growth and dormancy in response to
 48 seasonal changes. Patterns of growth cessation and dormancy minimise the risk of
 49 tissue damage from freezing over winter and are primarily induced by a change in
 50 photoperiod (FRACHEBOUD *et al.* 2009; BASLER AND KÖRNER 2012). Once dormant,
 51 many plants require a period of chilling before active growth is resumed in response

52 to warming spring temperatures (HORVATH *et al.* 2003). These transitions define the
 53 trade-off between growth and potential damage and are therefore expected to be
 54 subject to natural selection. Widespread species are exposed to a range of temperature
 55 and photoperiod length across their distribution, leading many species to show local
 56 adaptation and heritable variation in the timing of these phenology transitions
 57 associated with local conditions (HOWE *et al.* 2003; LUQUEZ *et al.* 2008; CONG *et al.*
 58 2016). The adaptive significance of cyclic phenology is especially important in
 59 woody perennials such as trees, as they are long lived and therefore exposed to
 60 seasonal changes over multiple years.

61
 62 While both spring and autumn phenology determine growing season and are therefore
 63 potentially adaptive, transitions are determined by different cues and therefore may
 64 respond to selection in different ways (SINGH *et al.* 2017). Dormancy is broken by
 65 changes in temperature at the beginning of spring in many plants (MCKOWN *et al.*
 66 2018), and therefore bud burst is defined by cues which have significant variation
 67 between local environments and successive years (BASLER AND KÖRNER 2012;
 68 BRELSFORD *et al.* 2019). Growth cessation is generally induced by changes in
 69 photoperiod (FRACHEBOUD *et al.* 2009; LAGERCRANTZ 2009; MICHELSON *et al.* 2017)
 70 or associated changes in the light spectrum (LEUCHNER *et al.* 2007; BRELSFORD *et al.*
 71 2019), and therefore varies predictably with latitude and consistently between
 72 growing seasons. These different characteristics between the triggers of growth and
 73 dormancy mean that the adaptation toward the optimal conditions and timing for
 74 spring and autumn phenology requires different biological responses. Consequently,
 75 effects of natural selection on genetic variation in these traits is likely to differ,
 76 potentially driving differences in the genetic architecture underlying each trait and
 77 influencing the scope of adaptive shifts in phenology traits under novel climatic
 78 conditions.

79
 80 Phenotypic variation is well characterised for many physiological and phenological
 81 traits in deciduous trees (COSTA E SILVA *et al.* 2004; FABBRINI *et al.* 2012), foremost
 82 of which are the Poplars, due to their adaptability, widespread distribution and

commercial value. A number of studies have investigated the genetic basis of poplar phenotypes and have identified many genomic variants associated with phenology and growth (INGVARSSON *et al.* 2008; LUQUEZ *et al.* 2008; FABBRINI *et al.* 2012; TRIOZZI *et al.* 2018). However, adaptability in complex traits largely depends on the pleiotropic effects of this genetic variation which can manifest as genetic covariance between phenology and growth-related traits (PORTH *et al.* 2014; MCKOWN *et al.* 2018). This information is fundamental to understanding the adaptability of poplar species, as the degree of genetic covariance between phenology traits and growth ultimately determines the magnitude of response to selection on individual traits (PORTH *et al.* 2015). A life history strategy that minimises the risk of cold damage through late bud burst and early bud set may reduce the overall growth and therefore constrain growth and establishment. Similarly, a strategy of early bud burst and late bud set results in a long growing season but increases the risk of frost related injury. As such, the capacity of trees to adapt to climatic variation is dependent on the genetic variance and covariance of phenology traits and growth, which determines the current range limit (CHUINE 2010) and potential to expand into novel habitats (ZANEWICH *et al.* 2018).

100

Populus trichocarpa (black cottonwood) is one of several large, deciduous, tree species within the genus *Populus*. *Populus* species are generally fast-growing and as such have great ecological significance as pioneer species (CRONK 2005), as well as an increasing value as a species for short rotation forestry (WEIH 2004) and bioenergy production (SANNIGRAHI *et al.* 2010; PORTH AND EL-KASSABY 2015). As a result, there has been extensive development of genetic resources to investigate and capitalise on the genetic basis of adaptation and growth traits in these species (CRONK 2005; JANSSON AND DOUGLAS 2007). Extensive research has identified substantial genetic variation across several adaptive traits, and while these estimates derive from a range of experimental designs and study populations, moderate to high heritability has been generally reported for growth (BRADSHAW AND STETTLER 1995; YU *et al.* 2001; PORTH *et al.* 2015), leaf phenology (HOWE *et al.* 2000; MCKOWN *et al.* 2014a; PORTH *et al.* 2015) and cold tolerance (HOWE *et al.* 2003) traits. A distribution over a

114 wide range of geography and climate has driven patterns of local adaptation (EVANS
115 *et al.* 2014; MCKOWN *et al.* 2014a), however extensive long distance pollen transfer
116 and large population sizes have resulted in very low population differentiation across
117 its natural range (SLAVOV AND ZHELEV 2010). Despite this, long running breeding
118 programs have produced few varieties that are well suited to the photoperiod and
119 climate typical of higher latitudes (KARACIC *et al.* 2003) which may limit
120 establishment of populations outside of the natural climatic range.

121

122 Here, we analyse phenotypic variation in a clonally replicated population of *P.*
123 *trichocarpa* genotypes grown in central Sweden. We survey the genetic basis of
124 spring and autumn phenology and determine the genetic covariance of these traits
125 with fitness in the form of lifetime growth. These relationships determine the genetic
126 constraints on adaptive evolution in this species and as such may constrain range
127 expansion into hostile environmental conditions. Specifically, we 1) aim to
128 characterise heritable genetic variation for phenology traits in this population, and 2)
129 test whether genetic correlations that exist between phenology traits may facilitate or
130 constrain adaptation to novel climatic conditions.

131

132 MATERIALS & METHODS

133

134- *POPULATION CHARACTERISTICS AND PHENOTYPIC MEASUREMENTS.*

135 Here we present a quantitative genetics analysis of phenology measurements taken
136 from 564 mature *P. trichocarpa* trees in a plantation at Krusenberg (59°44'44.2"N
137 17°40'31.5"E) in central Sweden. Material in this plantation was originally generated
138 from 9 female and 10 male trees collected over a latitudinal range from 44-60° in
139 North America (Supplementary Table 1), which were randomly crossed to produce
140 34 families. From these families, a total of 120 half sib, full sib, and unrelated trees
141 (onward referred to as genotypes) were clonally replicated with between 1 and 20
142 (median = 6) individual trees per genotype represented in the study population. The
143 plantation was established in 2003 on a flat, homogeneous area of agricultural field
144 approximately 275m x 40m. Individuals were planted at 3.5 m quadratic spacing in a

145 randomized design. The experiment was systematically thinned in March 2013
146 leaving 564 trees in an approximately 3,5x7 m diamond spacing.

147

148 To assess variation in phenology we included the traits that best define the major
149 milestones of phenology during the annual growth cycle. For spring development,
150 bud burst (bb) was scored on a scale from 1-5, with stage bb2 representing initial
151 shoot emergence, bb3 leaf primordia exposed, bb4 leaves half shed with bud scales
152 dropped and bb5 leaves completely shed. Autumn phenology was scored on a scale of
153 1-8 based on a continuous range of crown colouring from col1= 100% green, to col8
154 100% yellow. Leaf senescence (ls) was measured on a 3-point scale where ls1 = full
155 foliage, ls2 = half leaves remaining and stage ls3 full defoliated. We estimated
156 growing season length by the duration of photosynthetically active leaf canopy.
157 Canopy duration (CD) was defined as the period between the beginning of budburst
158 (bb2) and beginning of leaf yellowing (co3). Phenology measurements were taken
159 every 2-5 days during the spring and autumn seasons in 2017 and 2018. Lifetime
160 growth was determined by measuring diameter at breast height (DBH) in 2017

161

162 While late season growth cessation is best described by bud set, this trait is difficult
163 to accurately measure in mature trees. Due to the difficulty of determining the exact
164 transitions between beginning of growth, cessation of growth and bud development in
165 grown trees, we split these seasonal transitions into 5 biologically relevant proxy
166 stages which describe the start and end of seasonal transition; bb2, the first bud
167 emergence, (bb5) full leaf emergence, (col2) first stage of yellowing, (col5) complete
168 yellowing, (ls3) day of full leaf shed. These phenology traits are combined with
169 growth of the tree at 14 years of age as measured by cross calliper measurement
170 (DBH) in 2017.

171

172 *DATA IMPUTATION*

173 Screening was conducted at intervals of 2-5 days meaning that individual trees
174 occasionally passed through developmental stages between screenings. To account
175 for these missing estimates of the day of transition we estimated the transition days

for each developmental stage using local regression models (LOESS) fit to each individual tree. Models were fit through the data point of the first day an individual tree was observed at a stage transition and day estimates were calculated for any stage transitions for which there was no direct observation (Figure S2). This method estimates a non-linear developmental curve which is not constrained to fit any a-priori mathematic distribution for each individual tree and allows estimation of the day in which trees passed each developmental stage and inclusion of individuals which were not observed at important developmental transitions in the analysis. As extrapolation beyond the range of the observed data is unreliable using this method, we only retained estimates that were bounded by observations on either side. We validated this method by comparing estimated values with direct observations to ensure that estimated transitions accurately represented the observed data (Pearsons R = 0.98-1).

189

190 STATISTICAL ANALYSIS

191 *Heritability and cross year genetic correlations*

192

We estimated genetic parameters in this population using Bayesian Mixed model approach implemented in the MCMCglmm R package (HADFIELD 2010). This approach uses pedigree information to construct a relatedness (A) matrix which allows for a proper structuring of the random genetic effects to account for the complex combination of unrelated individuals, with half and full sibs included in the plantation. To estimate additive genetic heritability (h^2) for each genotype, and to determine the additive genetic co-variance between measurement years, we implemented the following bivariate models, where measures of each transition trait (e.g., budburst) taken in subsequent years 2017 ($Y1$) and 2018 ($Y2$) formed a bivariate response variable, μ is a fixed intercept, g is the random additive genetic component drawn from pedigree information and ε_{ijk} residual variance.

204

$$(Y1, Y2)_{ij} = \mu + g_i + \varepsilon_{ij} \quad (1)$$

205

206 Following the approach of KENNEDY AND SCHAEFFER (1989), we treat genetically
 207 identical individual trees (clones) from the same genotype as repeated measures, and
 208 estimate the genetic parameters based on the definition of the relatedness matrix (A)
 209 among genotypes rather than cloned individuals. Variation between individual trees
 210 within genotype was partitioned into the residual variance. Each trait was modelled
 211 separately. Models were run for 1000000 iterations, and we implemented Markov
 212 chain Monte-Carlo sampling with a thinning interval of 1000 to ensure very low
 213 autocorrelation between samples (> 0.003), and burn-in period of 1000 iterations.
 214 This approach yielded posterior probability distributions with sample sizes for each
 215 trait pair near 1000, from which we derived parameter estimates with 95% credible
 216 intervals. In all cases we specified uninformative priors (variance = 1, degree of belief
 217 = 0.002), although we also ran models with a range of prior specifications to ensure
 218 the final parameter estimates were not affected by our chosen priors. We extracted
 219 estimates for genetic effects from the posterior distribution of model (1) and
 220 calculated narrow sense heritability by extracting the additive genetic (V_a) proportion
 221 of total trait variance ($V_a + V_e$). Note that we estimate heritability based on each
 222 genotype, with variance among individual trees within genotype partitioned in the
 223 residual term.:

224

$$h^2 = \frac{V_a}{V_a + V_e}$$

225 Separate heritability estimates were calculated for each phenology trait in two
 226 consecutive seasons (2017 and 2018). We do not directly account for dominance
 227 effects in this model due to limited size and depth in the pedigree, and as such our
 228 estimates of heritability may be overestimated as some portion of any potential
 229 dominance variance may contribute to the estimates of V_a (WALSH AND LYNCH 2018).
 230 Cross-year genetic correlations were calculated by extracting the 2x2 genetic
 231 variance-(co)variance matrix from model (1), and we calculated genetic correlations
 232 as the genetic covariance between traits divided by the square root of the product of
 233 the variances.

234

235 GENETIC CORRELATIONS OF SPRING AND AUTUMN PHENOLOGY

236 We calculated the additive genetic correlations between phenology transitions and
 237 lifetime growth to determine the genetic architecture relating phenology to fitness.
 238 We combined the phenology transitions of budburst (bb2), leaf expansion (bb5), start
 239 of leaf yellowing (col2), end of leaf yellowing, (col5) leaf shed (ls3), canopy duration
 240 (CD) and lifetime growth (DBH) as a 7-trait multivariate response (Y_{ijk}), and
 241 implemented the following model:

242

$$Y_{ijk} = \mu + T_i + g_j + \varepsilon_{ijk}$$

243

244 Where μ is a fixed intercept, T_i is the fixed effect of year, g_j is the random term
 245 describing the additive genetic variance drawn from pedigree information and ε_{ijk}
 246 residual variance. We implemented this model using a Bayesian ‘animal’ model
 247 approach in the MCMCGLMM package in R (HADFIELD 2010). We implemented
 248 Markov Chain Monte Carlo sampling to estimate posterior probability distributions
 249 for the genetic parameters, and increased iterations until sample sizes for each
 250 estimation approached 1000. Our final models included 500,000 iterations, with a
 251 thinning interval of 500 and a burn-in of 10,000 iterations. Priors were constructed
 252 using measured phenotypic variances with a low degree of belief, and were checked
 253 against uninformative priors with low and high degrees of belief to ensure that prior
 254 specification had little effect on parameter estimates.

255

256 We extracted the 7x7 genetic variance-(co)variance matrix from this model, and
 257 calculated genetic correlations as the genetic covariance between traits divided by the
 258 square root of the product of the variances. We extract the genetic correlation
 259 between the 6 phenology traits and fitness (measured as DBH17, or lifetime growth)
 260 to determine the genetic relationship between phenology and growth. This correlation
 261 between phenology and growth ultimately determines whether adaptive shifts in
 262 phenology in response to climate change will be associated with increased growth, or
 263 be constrained by associated negative effects on tree growth.

264

265

266 RESULTS

267

268 *Phenotypic distribution, summary statistics*

269 All phenology transitions occurred earlier in 2018 than 2017 (Figure 1). Differences
 270 in the mean date for each stage were between 3 days (leaf shed, bud burst) and almost
 271 10 days (full yellow) earlier in 2018. Spring transitions occurred faster in 2018; bb2
 272 occurred over a period of 39 days in 2017 and 17 days in 2018 (difference of 22
 273 days), while leaf out (bb4) took 17 days in 2017 and 8 in 2018 (difference of 9 days).
 274 These differences corresponded with temperature and rainfall variation between
 275 observation years, where temperature increase was more gradual during spring in
 276 2017, summer was warmer and wetter in 2018, and autumn was drier in 2018.
 277 The duration of spring transitions was also shorter in 2018 indicating the abrupt
 278 increase in spring temperature during that year sped up leaf unfurling.
 279 (Figure S1).

280

281 *Trait heritability*

282 All phenology traits had moderate to high heritability with median estimates in the
 283 range of 0.41-0.76, identifying significant genetic variation in phenology traits
 284 between the genotypes in this growth trial (Figure 2). Heritability was highest in
 285 budburst (2017), leaf shed (2017, 2018) and lifetime growth traits, and lowest for leaf
 286 out and colouration transitions. Estimates for 2018 are reduced across all traits except
 287 leaf shed, although there is considerable overlap in posterior distributions for
 288 estimates in all traits except budburst.

289

290 Phenology traits had a strong genetic basis, with spring heritability in the range of
 291 0.48 - 0.72 (bb2) and 0.51 - 0.62 (bb4) respectively. Autumn phenology was also
 292 highly heritable, with highest and most consistent estimates for leaf shed (ls3) ($h^2 =$
 293 0.60 – 0.64). Lifetime growth was the highest and most precise of the heritability
 294 estimates ($h^2=0.76$), which reflects a substantial genetic basis for differences in
 295 lifetime growth between genotypes.

296

297 *Cross year genetic correlations*

298 Despite the considerable interannual shifts in phenotypic means described above,
 299 cross year genetic correlations were high. Estimates for budburst (bb2), onset of
 300 colour change (col3), and leaf shed (ls3) all overlap 1, showing complete correlation
 301 between years. The contribution of genetic variation to leaf out (bb5) and canopy
 302 duration also very consistent with correlations above $R = 0.8$. The transition to full
 303 yellow leaves cross-year correlation was $R = 0.6$, which is high given that this trait
 304 also had the largest mean difference between years at 9.7 days.

305

306 *Additive genetic correlations between traits*

307 Spring traits of budburst (bb2) and leaf out (bb5) displayed highly positive genetic
 308 correlations ($R = 0.77$), but showed little to no link to autumn phenology
 309 characteristics (Table 1). Similarly, autumn traits were strongly correlated indicating
 310 strong alignment in the genetic basis of traits within either spring or autumn seasons
 311 but not between. Leaf colouration was highly correlated with leaf senescence ($R =$
 312 $0.80-0.87$) suggesting a developmental cascade brought on by pleiotropic effects of
 313 genes involved in sensing and responding to the end of favourable growing
 314 conditions. Spring phenology was negatively correlated with both canopy duration
 315 and lifetime growth, where clones bursting leaves early in the season hold their leaf
 316 canopy longer and grow more. Leaf colouring was positively correlated with growth
 317 reflecting that clones with late onset of leaf colouration and senescence conferred
 318 higher lifetime growth (Table 1).

319

320

321 DISCUSSION

322

323 The ability of plants to survive the novel environmental conditions brought on by
 324 climate change is dependent on populations harbouring sufficient genetic variation to
 325 adapt to changes in seasonality. Here, we show that timing of phenology in an
 326 experimental population of *P. trichocarpa* has a heritable genetic basis, but also that

these traits can vary between years in response to climatic variation. With heritable genetic variation underlying both spring and autumn phenology, and an absence of genetic constraint between these transitions, our results suggest that these traits have potential to evolve independently in order to track future climate changes. As projections suggest climate change may provide conditions for a longer growing season, the absence of constraint may allow trees to have a longer canopy duration, which will in turn lead to higher growth. Together these findings provide insight into the genetic architecture of phenology traits, and identify the capacity of *P. trichocarpa* to adapt its phenology to novel environmental contexts as they colonise areas outside their original range, or as environmental conditions shift via global climate change.

There is a considerable contribution of genetics to all phenology traits surveyed in this field experiment and similar to other studies (eg. OLSON *et al.* 2013) we find abundant genetic variation for phenology traits determining the length of the growing season. There was also a strong genetic correlation within spring and within autumn phenology traits (0.77 - 0.87) suggesting pleiotropic effects of the genes underlying the cascade of leaf development and between traits influencing leaf coloration and leaf senescence, however genetic correlations across spring and autumn phenology were low to non-existent, reinforcing previous studies that show the genes (MCKOWN *et al.* 2014b) and environmental cues (SINGH *et al.* 2017) underlying spring and autumn phenology are different. Both spring phenology traits had negative correlations with lifetime growth suggesting that earlier budburst is beneficial for lifetime growth, and there was a stronger link between leaf unfurling and growth, suggesting clones with predictable late leaf unfurling was detrimental for lifetime growth.

Growing season in deciduous forest trees is defined by the leaf phenology traits, and we show with this experiment that canopy duration has a negative genetic correlation with spring phenology, and a positive correlation with autumn phenology. Perhaps unsurprisingly, we identify a genetic association where early bud burst and late leaf

drop leads to longer growing season and greater lifetime growth. These patterns have been shown in other studies of this species (eg. MCKOWN *et al.* 2014a), reflecting the general biology of adaptation in deciduous plants, where the timing of spring phenology resolves the tension between two factors: damage avoidance and competition. Early budburst increases the probability of leaf exposure to frost conditions, however it may confer a competitive advantage by earlier exposing leaves to (direct) sunlight. Late leaf flush delivers trees into an environment where much of the available light is already absorbed by early flushing neighbours, and puts the tree at a disadvantage for light capture that may last for the full growing season. This disadvantage likely accumulates across years as the late flushing individual falls further behind its faster growing neighbours. Late flush also reduces exposure to optimal spring growing conditions, further compounding the disadvantage of spring shading which may significantly reduce growth (YU *et al.* 2001).

A number of studies have found moderate heritability for phenology traits in this, and other *Populus* species for estimates of bud flush however, it is important to note that the majority of these studies have been conducted on juvenile trees (eg. MCKOWN *et al.* 2014a; PLIURA *et al.* 2014). Estimates of growth heritability presented here is generally higher than previously reported values for *P. trichocarpa* (MARRON *et al.* 2010; MCKOWN *et al.* 2014a), however this may be due to estimations being derived from established adult trees which will reduce the impact of random events which can lead to highly variable growth in the first few growing seasons. It is not uncommon to find higher heritability for fitness related traits in older individuals in animals and humans (eg. CHRISTENSEN *et al.* 2003; WILSON *et al.* 2005) however this is more likely to be due to a decline in environmental or residual variance, rather than an increase in additive genetic variance. Other considerations such as the pedigree structure and source material in this population may also contribute to the higher heritability estimates than other studies, and also contribute to the large credible intervals around these estimates. We emphasise that due to these factors, heritability estimates should be interpreted in the context of the present population and experimental design.

389

390 While we find evidence for a heritable basis to spring and autumn phenology, there is
 391 only a weak relationship between leaf unfurling and senescence. This finding adds to
 392 existing evidence that different genetic architectures are controlling phenological
 393 responses of spring and autumn environmental cues in *P. trichocarpa*. This is
 394 consistent with the observation of conserved patterns of environmental responses
 395 across the plant kingdom where bud burst is largely regulated by local environment
 396 not local adaptation (MACKENZIE *et al.* 2018) and leaf out highly dependent on
 397 temperature (POLGAR AND PRIMACK 2011). Sensitivity to local growing conditions
 398 such as temperature and altitude (a strong determinant of temperature) can be
 399 stronger than population level patterns of local adaptation, and sensitivity can be
 400 similar between populations of the same species (VITASSE *et al.* 2009), suggesting
 401 that the timing of spring growth is mediated by plants anticipating favourable local
 402 conditions which pose reduced risk of damage before investing resources in the
 403 production of leaf tissue. The strong cross-year genetic correlations and high
 404 heritability in these traits suggests that there is substantial variation in the timing of
 405 seasonal transitions across years, but that clones respond in the same way over
 406 consecutive years.

407

408 Variability and uncertainty in relation to the future timing of seasonal transitions and
 409 the potential for extreme or unseasonal weather events such as out of season frost
 410 pose a significant risk of damage to deciduous trees. Our findings here suggest that
 411 adaptive shifts in phenology may be possible in *P. trichocarpa*, however the genetic
 412 links between phenology and growth suggest that shifts in the timing of spring and
 413 autumn transitions have the potential to affect growth if shifts lead to a shortening of
 414 the growth period. Previous work has shown that autumn phenology is primarily
 415 initiated as trees respond to predictable changes in photoperiod , but data presented
 416 here suggests that large interannual shifts in autumn timing are possible, which
 417 suggest that leaf colour and senescence traits may also be responding to multiple
 418 environmental cues either during the summer growing season, or during the onset of
 419 autumn (ROHDE *et al.* 2011b).

420

421 Multiple studies have shown that temperature is not a trigger for senescence
 422 (BHALERAO *et al.* 2003; KESKITALO *et al.* 2005; LUQUEZ *et al.* 2008; FRACHEBOUD *et*
 423 *al.* 2009), however speed of senescence is temperature dependent once initiated
 424 (FRACHEBOUD *et al.* 2009) and evidence from poplar hybrids suggests that
 425 temperature may contribute to the timing of growth cessation and bud set (ROHDE *et*
 426 *al.* 2011a). In this study we were unable to accurately measure bud set across the
 427 experiment, however it is clear that growth cessation occurs some time before bud
 428 set and before leaves change colour (ROHDE *et al.* 2011b) which suggests that the
 429 annual variation in colour timing is a response to differences in autumn temperature
 430 between years that is occurring after the point where trees have initiated senescence.
 431 The risks of autumn frost damage are elevated with late bud set, which can have
 432 significant negative effects on viability and winter survival in juvenile trees (HOWE *et*
 433 *al.* 2003), so it is likely that late season phenology is strongly associated with climatic
 434 adaptation. In this case, photoperiod is a more reliable predictor of the onset of frost
 435 than temperature, which fluctuates between seasons so it is likely that growth
 436 cessation and bud set occurs before the prospect of frosts and trees are prepared for
 437 winter long before the leaf colour changes (BASLER AND KÖRNER 2012).

438

439 It is important to note that a substantial shift in autumn phenology occurred between
 440 the two years of observation. In 2018, there was a faster increase in temperature and
 441 more sunny days during early spring, which is reflected in the faster spring
 442 development and earlier leaf out in that year. Autumn was also drier, and leaf
 443 colouration earlier in 2018, which is consistent with the presence of drought induced
 444 bud set, so it is likely that although light characteristics are the primary cue for
 445 growth cessation and senescence (MICHELSON *et al.* 2017; TRIOZZI *et al.* 2018), the
 446 timing and response to these factors is modified by other environmental cues such as
 447 temperature (ROHDE *et al.* 2011a) or drought stress (ADAMS *et al.* 2015). Importantly,
 448 this finding suggests that although timing of phenology transitions have a strong
 449 genetic basis, trees respond to multiple cues and may respond to variable local
 450 climatic conditions as well as predictable cues such as day length. Adaptation may be

451 facilitated by the capacity to track both constant cues (such as daylength) and variable
452 cues such as temperature and rainfall, which may facilitate establishment of new
453 populations outside of their natural range, and allow adaptation to shifting conditions
454 within the native distribution.

455

456 CAVEATS AND CONCLUSIONS

457 This study presents data from a collection of source genotypes sampled over
458 approximately 15 degrees of latitude in north-western north America. This sampling
459 design may inflate estimates of heritability by sampling across locally adapted
460 populations, although this does not undermine the key findings that phenology
461 transitions have a strong genetic basis. Also, we are unable to explore the contribution
462 of dominance and epistatic interactions to our estimates of additive genetic variance
463 due to limitations of the pedigree, however we note that these are common limitations
464 of quantitative genetic analyses of experimental systems, and we acknowledge that
465 some of the phenotypic variance attributed to additive genetic variation may be the
466 result of dominance and epistatic interactions.

467 Overall, the substantial heritability found in phenology traits and the ability of trees to
468 shift the timing of phenology transitions drastically between years suggests that *P.*
469 *trichocarpa* populations have the ability to adjust their phenology in response to a
470 changing climate. The wide and environmentally variable native range of *P.*
471 *trichocarpa* is indicative of this species ability to persist across variable and
472 unpredictable climates. We show here that a genetic variation underlies a large part
473 of variation in phenology traits, but that trees also have the capacity for significant
474 phenology shifts in response to year to year weather variation across the two years
475 investigated here. This suggests this species has capacity to adapt its phenology to
476 rapid climatic shifts, and that there is sufficient genetic variation underlying these
477 traits that either natural or artificial selection could lead to evolutionary changes in
478 adaptive phenology traits which may facilitate colonisation outside its natural range.

479

480

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483

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486 DATA ARCHIVING: Data will be uploaded to the Swedish Artportalen

487 /Artdatabanken database, and any other archive suggested by Reviewers or editors

488 CITED LITERATURE

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639 FIGURE LEGENDS

640

641 **Figure 1:** Phenotypic trait summary in 2017 and 2018. Summaries shown are mean
642 and density overlayed upon original data points. All phenology transitions occurred
643 earlier in 2018 leading to a slightly longer mean growing season. Truncated
644 distributions are indicative of an abrupt timepoint where all individuals reached
645 maximum scores in each trait. Traits include beginning(bb2) and end (bb4) of
646 budburst, beginning (col3) and end (col8) of leaf colour transition from green to
647 yellow, leaf shed (ls3), canopy duration (CD), and lifetime growth (dbh).

648

649 **Figure 2:** Narrow sense heritability estimates for phenology traits at Krusenberg field
650 experiment in 2017 and 2018. Means and 95% credible intervals derived from univariate
651 ‘animal’ models run in R (MCMCglmm)

652

653

654 **Figure 3:** Cross year genetic correlations. Mean and 95% credible intervals for
655 estimates of cross year genetic correlations. Traits include beginning and end of
656 budburst (bb2, bb4), beginning and end of leaf colour transition from green to yellow
657 (co3, co8), leaf shed (ls3), canopy duration (CD), Correlations between years is
658 consistently high, and intervals overlap 1 for bb2, co3 and leaf shed traits.

659

660 **Figure S1:** Weather variation between the two years of survey. Spring temperatures
661 increased faster, summer was warmer and wetter, and there were more sunny days
662 during 2018. Historical range is indicated by black dots. Lines link monthly means
663 for the study period.

664

665 **Figure S2:** Comparison of observed data (red) and Loess fit imputation (orange) for
666 spring leaf development in four randomly selected individuals. Imputed data is fit
667 through the first day individual trees were scored for each stage and show trajectories
668 more in line with biological development. Intersection between dotted lines and fitted
669 lines denotes values of bud break (BB2) and leaf unfurling (BB4).

670

671 **Table S1.** Crossing scheme showing origin (when known) of parents and number of
 672 clones from each cross. Some of the parents (in bold) and all offspring clones are
 673 included in the Krusenberg field experiment. OP indicates crosses that were
 674 conducted with Open pollination. Numbers within table indicate how many offspring
 675 were produced from each cross.
 676

Table 1: Genetic correlation matrix of phenology and growth traits in *P. trichocarpa*. Correlation estimates were drawn from a 7 trait, multivariate ‘animal’ model using the MCMCGLMM procedure in R. Values in brackets represent the lower and upper 95% credible intervals, values in bold do not overlap zero. Traits include beginning and end of budburst (bb2, bb4), beginning and end of leaf colour transition from green to yellow (co3, co8), leaf shed (ls3), canopy duration (CD), and lifetime growth (DBH).

	bb2	bb4	co3	co8	ls3	CD
bb4	0.768 (0.718,0.82)					
co3	0.12 (-0.281,0.265)	0.219 (-0.09,0.357)				
co8	0.059 (-0.332,0.227)	0.134 (-0.212,0.267)	0.868 (0.828,0.866)			
ls3	0.211 (-0.074,0.349)	0.255 (0.023,0.392)	0.808 (0.752,0.821)	0.8 (0.732,0.803)		
CD	-0.668 (-1.591,-0.26)	-0.427 (-1.127,-0.12)	0.637 (0.56,0.718)	0.6 (0.492,0.655)	0.438 (0.276,0.527)	
DBH	-0.317 (-0.809, -0.036)	-0.418 (-0.993, -0.115)	0.254 (0.009,0.394)	0.394 (0.197,0.468)	0.334 (0.181,0.456)	0.431 (0.25,0.506)





