1	Contributions of dynamic environmental signals during life-cycle transitions
2	to early life-history traits in lodgepole pine ( <i>Pinus contorta</i> Dougl.)
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#### 17 Abstract

18 Environmental signals are important triggers in the life-cycle transitions and play a crucial role in 19 the life-history evolution. Yet, very little is known about the leading ecological factors contributing 20 to the variations of life-history traits in perennial plants. This paper explores both the causes and 21 consequences for the evolution of life-history traits (i.e., seed dormancy and size) in lodgepole 22 pine (*Pinus contorta* Dougl.) across British Columbia (B.C.), Canada. We selected 83 logepole 23 pine populations covering 22 ecosystem zones of B.C. and through their geographic coordinate, 24 197 climatic variables were generated accordingly for the reference (1961-1990) and future (2041-2070) periods. We found that dynamic climatic variables rather than constant geographic variables 25 26 are the true environmental driving forces in seed dormancy and size variations and thus provide 27 reliable predictors in response to global climate change. Evapotranspiration and precipitation in 28 the plant-to-seed chronology are the most critical climate variables for seed dormancy and size 29 variations, respectively. Hence, we predicted that levels of seed dormancy in lodgepole pine would increase across large tracts of B.C. in 2050s. Winter-chilling is able to increase the magnitude of 30 31 life-history plasticity and lower the bet-hedge strategy in the seed-to-plant transition; however, 32 winter-chilling is likely to be insufficient in the north of  $49^{\circ}N$  in 2050s, which may delay germination while unfavourable conditions during dry summers may result in adverse 33 consequences in the survival of seedlings owing to extended germination span. These findings 34 35 provide useful information to studies related to assessments of seed transfer and tree adaptation.

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Key words: life-history evolution; dynamic environmental signals; seed dormancy; seed size;
climate change; lodgepole pine

#### **39 INTRODUCTION**

In the context of global climate change, species have responded by tracking the environment for 40 41 which they are best suited through local adaptation, range shift, range reduction, or a combination 42 of these (Walther et al., 2002; Parmesan and Yohe, 2003; Cleland et al., 2007; Breshears et al., 2008). Knowledge about traits that vary with ecological niches should therefore help predict how these 43 44 traits may evolve under climate change. Life-history traits, known as fitness components due to 45 predictable monotonic relationship with fitness, are related to the timing and success of development, reproduction, and senescence throughout the life cycle (Calow, 1998). Ecological and 46 47 genetic trade-offs constrain the combinations of the life-history traits through evolution (Adler et 48 al., 2014), which involves an adaptation to environments and is shaped by natural selection.

49 Climate change has already altered the timing of major life-history transitions, such as seed germination timing (from seed to plant). Timing is directional and completely asymmetric and the 50 timing of seed germination is the earliest life-history trait that is expressed and sets the context for 51 52 the traits that follow (Donohue et al., 2010). It is controlled by the level of seed dormancy, whereby 53 dormant seeds await germination cues for dormancy release (Finch-Savage and Leubner-Metzger, 2006; Baskin and Baskin, 1998). Moreover, early developmental stages of plants are more sensitive 54 55 to environment perturbations than adult stages and represent a major bottleneck to regeneration 56 from seeds (Johnsen and Skrøppa, 1996;Hedhly et al., 2009). At the core of plant regeneration, 57 temperature and water availability (or precipitation) are critical drivers for a plant's distribution 58 (Woodward and Williams, 1987). Hence, climate-changed plant regeneration will be manipulated 59 in both temperature- and moisture-controlled ecosystems (Walter and Breckle, 2002) and much of plant regeneration climate research has been directed at tundra and boreal forest, and treeline 60 ecotones (reviewed in (Walck et al., 2011)). 61

62 Seed size is another important life-history trait and subject to changing environmental settings. In the palaeontological scale, fossil data suggest that seed size remains small across all 63 plants until the Cretaceous period (124 MYA). Seed size begins to increase after the limit of the 64 Cretaceous-Tertiary period (65 MYA). One of the most popular explanations is that climate change 65 gives rise to seed size variation, which took place during that period (Eriksson et al., 2000). In the 66 67 ecological time scale, empirical studies lend supports to the impact of environment on seed size variation. Since the early 1950s, effects of environmental stimuli, such as, temperature and 68 69 photoperiod on seed size and weight have been noted. Chenopodium polyspermum L. seeds from 70 mother plants grown in long days have lower germination frequency and thicker seed coats when 71 compared to seeds from short days (Pourrat and Jacques, 1975). Large seed size with more 72 provisions stored for seedlings may be favorable in variable environments (Venable and Brown, 73 1988). Seed plants have a general trend of increasing embryo to seed ratio (E:S) in morphological 74 seed dormancy and the shift in E:S is likely a heterochronic change, having vital implications to 75 life history of seed plants (Forbis et al., 2002). A priori, the environment plays a crucial role in life-history traits in general and seed size in particular. 76

77 Adaptive phenotypic plasticity underpins rapid phenological shifts in response to climate 78 change and evolve when cues reliably predict fitness consequences of life-history decision 79 (Simons, 2014). This was evident by the performance of seedlings produced by central European trees growing in central Norway as they expressed phenology similar to that of their adjacent 80 81 ecotype and were exceedingly different from those produced at their original habitat (Skrøppa et al., 2010). Differences in adaptive traits between populations are inconsistent with the Mendelian 82 83 genetic framework and probably modulated by epigenetic mechanisms (Yakovlev et al., 2012). 84 Germination-cuing under favorable conditions is similar within species; however, these conditions

85 do not persist throughout the seasons. Likewise, populations growing in different ecological niches may have different degrees of exposure to unfavourable environments. As such, the prevention of 86 87 germination of some seeds even under favourable conditions would be of significance. When 88 environments fluctuate unpredictably, "bet-hedging" strategy (Slatkin, 1974) is expected to spread germination over time to reduce the risk of outright extinction. This results in the evolution of 89 90 traits that maximizes the geometric-mean fitness by reducing fitness variance over generations 91 (Gillespie, 1977). To date, much of the evidence for bet-hedging remains restricted to simple life-92 history plants (e.g., annuals) (Childs et al., 2010).

A continuously changing environment is constantly selecting for new adapted genotypes resulting in greater genetic diversity (Jump et al., 2009). Adaptive evolution characterized by the genetic architecture allows population persistence in the long term (Lande and Shannon, 1996). However, sustained directional selection due to climate change could potentially eliminate a proportion of the genetic variation needed for continued adaptation. To date, it remains unclear whether adaptive evolution can keep pace with climate change (Etterson and Shaw, 2001).

99 Life-history strategy for long-lived organisms is influenced primarily by survival (Adler et 100 al., 2014). Evergreen coniferous forests in the Pacific Northwest are unique among the Northern 101 Hemisphere temperate forests in their species composition and high productivity (Waring and 102 Franklin, 1979). Our study species, lodgepole pine (Pinus contorta Dougl.), is an aggressive pioneer species distributed over wide geographic and ecologic ranges across British Columbia, 103 104 Canada and is therefore expected to display a wide spectrum of dormancy variation owing to 105 adaptation to diversified local habitats (Plomion et al., 2007). Cone and seed production in 106 lodgepole pine is not as cyclic as in many other conifers. Cone drop soon after seed maturation is 107 commonly observed in coastal areas but lodgepole pine growing in the interior tends to be

108 serotinous, indicating that mature cones do not drop or open to release seeds unless exposed to 109 fluctuating high temperature during prolonged hot summer and low fall temperatures or due to fire 110 or insect damage (Fowells, 1965; Owens et al., 1981, 1982). Most serotinous cones take several 111 years to open and seeds are released in large quantities, consequently, any seed collection made 112 from a single tree consists of a mixture of different seed-crop years. Before the cones eventually 113 open, seeds are wrapped and sealed by scales and sticky resin without exposure to any germination 114 cues (such as moisture, oxygen, etc.); thus we assumed that seed dormancy in lodgepole pine is 115 least affected over its natural storage period.

116 Seed dormancy is an intrinsic attribute affecting regeneration dynamics and seed size is one of the vital determinants for the evolution of seed dormancy. The objectives of this study are 117 118 to evaluate the contributions of local environmental effects during life-cycle transitions to seed 119 dormancy and size variations of lodgepole pine populations across British Columbia, Canada and to predict how life-history traits evolve locally under ongoing climate change. Studies of seed 120 121 dormancy and size allow investigating the relationship between these two life-history traits and their relationship with environments in life-cycle transitions. This study can also contribute to 122 providing the missing empirical evidence of bet-hedging strategy in long life-history plants. 123

124

#### 125 METHODS

### 126 Plant materials, and current and future climate data

The 83 lodgepole pine seed lots used were representative of 83 different populations covering 22 ecosystem zones. The seed lots are distributed throughout the species' natural range across Brtish Columbia (B.C.), Canada encompassing coastal area and interior regions with a spatial grid over latitudinal range from 49 to 60°N and longitudinal range from 115 to 132°W, which consist of tundra, boreal and temperate forests, and treeline ecotone ecosystems (Fig. S1). The studied populations were selected primarily based on two important climate variables; namely, mean
annual temperature (MAT) and annual heat moisture index (AHM) (Fig. S2) (Wang et al., 2006)
and geographic variables, including longitude, latitude, and elevation were also considered.

135 Climate data (197 climatic variables) of the 83 studied sites for the reference normal period 1961-1990 were generated using ClimateWNA version 4.85 (a software package used for regional 136 137 climate predictions using historical weather station data and global circulation models) (Wang et 138 al., 2012). The same 197 climatic variables for the future period between 2041 and 2070 (or 2050s) 139 were also projected using ClimateWNA. The future climate data were downscaled to point 140 locations using a delta approach (Wang et al., 2012). We used three Representative Concentration Pathways (RCP) 2.6, 4.5 and 8.5 from the CGCM4 model output of the phase 5 of the Coupled 141 142 Model Intercomparison Project (CMIP5), generated by the United States National Center for 143 Atmospheric Research. The CCSM4 model output was included in the Intergovernmental Panel 144 on Climate Change (IPCC) Fifth Assessment Report (AR5) (Stocker and Randall, 2013). 145 Moreover, CCSM4 is close to the average in both temperature and precipitation increases in the study areas (B.C.). In the prediction of seed dormancy indicators across B.C. for the reference 146 period and 2050s, climate data were generated for each pixel at the spatial resolution of  $800 \times 800$ 147 148 m using ClimateWNA. The full list of climatic variables is given in Table S1 and the four most relevant variables for the present study were *Eref07* and *\_summer* (July and summer hargreaves 149 reference evaporation estimated based on temperature and solar radiation), DD\_0\_summer 150 151 (summer degree-days below 0°C), PPT07 and 10 (July and October precipitation (mm)), and *Tmax07* (July maximum mean temperature (°C)). 152

## 153 Life-history traits' measurement

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154 Germination parameters were estimated from cumulative germination curves fitting a mathematical function known as the four-parameter Hill function (El-Kassaby et al., 2008). Core 155 germination parameters include dormancy index (DI, the difference of areas under germination 156 157 curves (AUC) of control and treated seed), germination capacity (GC, the final germination)158 fraction), time to germination (time at the onset of germination), and germination speed (GS, the 159 time required to achieve 50% germination of the total germinated seed) (graphically represented by Figure S3, (Liu and El-Kassaby, 2015)). To simulate winter chilling in the soil seed bank and 160 161 retain natural seed dormancy, seed germination was manipulated with or without 21-day moist-162 chilling treatment under 3°C in lightproof chamber. Germination assay was performed as previously described (Liu et al., 2013b) and seed size was measured as the average of 1000-seed 163 164 weight collected from each study site.

#### 165 Data analysis and visualization

166 To investigate which climatic variables can most explain the observed patterns of seed dormancy and size variations, partial least squares (PLS) analyses was conducted using SAS® (vers. 9.3; 167 SAS Institute Inc., Cary, NC) (Jose Crossa et al., 2013). The goal of PLS regression is to analyse 168 169 multivariate response based on a large set of explanatory variables (i.e., climatic variables). This 170 technique combines features from principle component analysis (PCA) and multiple regression 171 (Abdi, 2007;Carrascal et al., 2009). PLS regression is particularly suitable when the matrix of predictors has more variables than observations, and when there is multicollinearity among N 172 173 values. Using the machine-learning algorithm (i.e., PLS), important climate variables can be 174 identified through a process of model optimization and there is no needed to explain the 175 contribution of climate variables in the initial input dataset. In the PLS biplot, the angles between 176 the environment vectors proximately correspond to the correlation coefficients among the

177 environments. The cosine of the angle between two vectors approximates the correlation between them and the length of the vectors is proportional to the standard deviation within respective 178 environments or genotypes (Yan, 2001; Yan and Tinker, 2006). Virtually, an ideal environment 179 180 has the longest vector of all test environments (most discriminating and informative) and is closely 181 located on the abscissa (most representative) (Liu and El-Kassaby, 2015). As a comparison, PCA 182 for explanatory variables and canonical correlation analysis (CCA) dealing with both explanatory and response variables were also performed (Liu and El-Kassaby, 2015). To examine how seed 183 dormancy and size varied in ecosystem zones, the two variables were analyzed collectively using 184 185 multivariate analyses (MANOVA) with the aid of general linear model (GLM) procedure in SAS 186 (Manly, 2005; Tabachnick and Fidell, 2012). To investigate how the most correlated climatic and geographic variables affect life-history traits, hierarchical models were established using PROC 187 188 MIXED in SAS with errors split into population and ecosystem zones but having unequal variances at the population level (West et al., 2007;Raudenbush and Bryk, 2001). The most 189 190 correlated climatic variables through the PLS analysis were used and the model was expressed as 191 follows:

192

$$(\text{life-history trait})_{ij} = (\beta_0 + \varepsilon_j) + \beta_k \times P_{kij} + \beta_l \times T_{lij} + \varepsilon_{ij}$$
[1]

where *i* and *j* represent two levels, namely, *i*<sup>th</sup> population within *j*<sup>th</sup> ecosystem zone; P<sub>kij</sub> and T<sub>lij</sub> represent the *k*<sup>th</sup> precipitation- and *l*<sup>th</sup> temperature-based variable in *i*<sup>th</sup> population within *j*<sup>th</sup> ecosystem zone, respectively;  $\varepsilon_j$  and  $\varepsilon_{ij}$  represent errors from the ecosystem zone and population level, respectively. Intercept ( $\beta_0 + \varepsilon_j$ ) and coefficients ( $\beta_k$  and  $\beta_l$  (k, l = 1, 2, 3, ...)) were estimated using our experiment data. Analogously, fitting models using longitude, elevation, and latitude as independent variables were established. Using the foregoing climatic model for the reference period and 2050s via CCSM4 RCP 4.5, *DI* values were estimated for each pixel across BC at the resolution of  $800 \times 800$  m and mapped across B.C for visualization.

To investigate the phenotypic plasticity of timing of seed germination, Pearson's productmoment correlation was calculated. The mean phenotypic plasticity of timing of seed germination is the magnitude of the average response of the population genotypes to specific environments and calculated as a population's mean *AUC* in the chilling manipulation minus its corresponding value in the control (i.e., *DI*). To investigate the range of possible plastic response to dormancy and to determine its genetic contribution to the phenotypic variance, the following linear model was used:

207 
$$DI_{ij} = \mu + G_i + E_j + GE_{ij} + \varepsilon$$
[2]

where  $DI_{ij}$  is dormancy index of  $i^{th}$  genotype in  $j^{th}$  environment,  $\mu$  is the overall mean,  $G_i$  is the effect of the  $i^{th}$  genotype,  $E_j$  is the effect of the  $j^{th}$  environment,  $GE_{ij}$  is the interaction of  $i^{th}$  genotype with  $j^{th}$  environment, and  $\varepsilon$  is the random error. The range of phenotypic plasticity is interpreted as  $G \times E$  variation ( $V_{G \times E}$ ) and the extent of genetic control of a trait was calculated by broad-sense heritability ( $H^2 = V_G / (V_G + V_{G \times E} + V_{\varepsilon})$ ).

213 To evaluate bet-hedge strategy in seed germination, two potential traits were tested, i.e., germination capacity and time to germination. Specifically, germination capacity was the subject 214 215 of Cohen's classic bet-hedging model (Cohen, 1966), in which dormancy is expected to evolve in 216 proportion to the probability of encountering a 'bad' year, which corresponds to no chilling manipulation in this study. Time to germination in a 'bad' or 'good' year (i.e., no chilling vs. 217 218 chilling) represents within-season variation, which has been shown to be advantageous when 219 germination success is unpredicted within season (Simons, 2009). The standard deviation (STD) for these two traits in a 'good' and 'bad' year was also calculated to measure the amount of 220 221 variation across studied populations.

222

## 223 **RESULTS**

224 Life-history traits strongly correlated with climatic variables in the plant-to-seed transition

225 Partial least squares (PLS) analyses for seed dormancy and weight indicated that the first and second components accounted for 15 and 13%, and 18 and 9% of the total variation, respectively 226 (Fig. S4). The 15 most correlated climatic variables with respect to life-history traits were 227 228 correlated with environmental factors related to temperature (Fig. 1). This indicates that 229 temperature plays a major role in the development of life-history traits (Liu and El-Kassaby, 2015). 230 By contrast, the PCA analysis showed several temperature-based variables which were 231 intrinsically correlated (Fig. S5). In addition, the PLS analyses classified the 83 population habitats into 21 and 20 major categories for seed dormancy and weight, respectively (blue arrows in Figure 232 233 S4).

The first canonical correlation analysis (CCA) was significant across all statistical tests 234 235 (Table S2A). Seed dormancy and weight had moderate (0.66) and very strong (0.90) correlations 236 with climate, moderate (0.59) and high (0.74) correlations with life-history traits (i.e., combination of seed dormancy and weight), and moderate (0.53-0.55) correlation with geographic variables 237 238 (Table 1A). This implies that the climatic variables were more important than the geographic 239 variables in variations of life-history traits. Based on respective covariate matrices for 'life-history 240 traits', 'climatic ecology', and 'geographic ecology' canonical variates, 62.35 and 47.11%, 62.85 241 and 45.06%, 14.80 and 6.82% of variance were explained by their corresponding canonical 242 variates for the same group of variables and the other group of variables (Table 1B). Furthermore, multivariate analyses showed that ecosystem zones were significant for seed dormancy and weight, 243 and had moderate correlation ( $R^2 = 0.55 \cdot 0.56$ , P < 0.0001) with life-history traits (Table 2). 244

245 In the hierarchical model for seed dormancy using climatic variables, the climatic variables *Eref07*, *Eref\_summer*, and *DD\_0\_summer* were significant (Table 3A). This indicates that summer 246 moisture and temperature, the period corresponding to the plant-to-seed transition, played an 247 248 important role in the development of seed dormancy. The intercept including error from ecosystem zones was not significant (Table 3A). For seed weight, variables PPT07 and PPT10, Tmax07, and 249 250 intercept were significant (Table 3B), indicating that July temperature and precipitation and October precipitation were important to seed weight. Ecosystem zones, representing ecological 251 252 boundaries in geography, had a greater impact on the variation of seed weight than that of seed 253 dormancy. In the geographic variable-based hierarchical model for seed dormancy, only longitude 254 and intercept were significant (Table S4A and Fig. S6A). For seed weight, longitude, elevation and intercept were significant (Table S4B and Fig. S6B), indicating that longitude was an 255 256 important geographic factor in life-history traits and geographic factors had greater influence on the variation of seed weight than seed dormancy. Considering all these results collectively, life-257 history traits were significantly influenced by precipitation as well as temperature in the plant-to-258 259 seed chronology, and roughly distributed in a longitude pattern.

## 260 Prediction of life-history traits in response to climate change

Using current climate data, seed dormancy predicted had a moderate linear relationship ( $R^2 = 0.47$ , P < 0.0001) with that observed (Fig. 2), which was used for the correction of the climate scenario predicted in 2050s. Seed dormancy predictions using three greenhouse gas emission scenarios (RCP2.6, 4.5, and 8.5) showed that *DI* in 2050s would increase (Fig. 2). In general, the spatial pattern of *DI* across the entire province showed that the seed dormancy in the south of B.C. would pronouncedly increase and the territory previously not suitable for pines to establish (grey area in 1970s) is expected to shrink in 2050s, which are responses to climate change (Fig. 3). In addition, seed weight 'predicted' had a low moderate linear relationship ( $R^2 = 0.34$ , P < 0.0001) with observed values (Fig. S7).

#### 270 Correlated response in plasticity to timing of seed germination after 'winter-chilling'

271 Among the 83 lodgepole pine populations, variance of seed dormancy was significantly explained 272 by genotype and environment (Table 2B). Estimate of broad-sense heritability  $(H^2)$  was 0.54 corresponding with the observed variance component for genotypes (Table 2B), indicating that 273 genetic components have moderate influence on seed dormancy variation. In response to 274 significant  $G \times E$  interaction (Table 2B), a plot of the reaction norms of AUC showed a range shift, 275 276 nonparallelism, and crossing between moist-chilling and control (Fig. 4), where  $G \times E$  accounted 277 for 39% of the variance and was attributed to crossing of reaction norms (Table 2B), indicating 278 that environments changed the adaptive values of the life-history trait.

AUC and phenotypic plasticity under chilling treatment had a moderate correlation ( $R^2 = 0.53, P < 0.0001$ ) (Fig. 4), indicating that populations that germinated after chilling treatment were more plastic. However, when no treatment prior to germination was applied, they yielded a weak and not statistically significant correlation ( $R^2 = 0.04, P = 0.3479$ ) (Fig. 4). Therefore, germination after the chilling treatment was able to increase the magnitude of phenotypic plasticity, which was a response to subsequent predictable environments.

### 285 Expression of bet-hedge under manipulated environmental uncertainty

Relative to 'bad-year' simulation, 'good-year' resulted in higher and uniform germination capacity and shorter time to germination across the 83 populations (Fig. 5). The standard deviation (STD) of germination capacity and time to germination were evenly distributed on the two sides of respective average line for both good- and bad-year across populations (Fig. 5) and 'good-year' had lower average STD germination capacity and STD time to germination (Fig. 5). This indicates that, though bet-hedge is a risk strategy for unpredictable environments, 'good-year' can lower the
risk by means of allowing higher germination capacity and shorter time to germination across
populations.

294

#### 295 **DISCUSSION**

Climate change is accelerating plant life-cycle transitions in coordination with the seasons. In life-296 297 cycle transitions, the environment plays a critical role in the development of life-history traits and in response to environmental stimuli, intrinsic mechanisms of genetics, epigenetics, phenotypic 298 299 plasticity, bet-hedge strategy, and adaptive evolution take effect (Fig. 6). In this study, we found 300 that seed dormancy and size were most correlated with evapotranspiration, and precipitation and 301 maximum mean temperature during the plant-to-seed transition, respectively, using 83 lodgepole pine populations across British Columbia (B.C.), Canada. We predicted that the range of seed 302 303 dormancy variation would increase across B.C. in 2050s. Moreover, winter-chilling can increase 304 the magnitude of life-history plasticity (predictable factor) and lower the bet-hedge strategy (unpredictable factor). However, future climate may bring about insufficient winter-chilling 305 required to decay seed dormancy, thus resulting in adverse consequences for the timing of 306 phenology and the growth and establishment in lodgepole pine. This study allowed us to gain 307 insights in the role of dynamic environments in shaping life-history characteristics. 308

## 309 Environmental conditions in the plant-to-seed transition

In the plant-to-seed chronology, the sexual reproduction of plants is vulnerable to climate change as influenced by the maternal environment (Hedhly et al., 2009;Donohue, 2009;Schmitt et al., 1992a), and temperature is involved in both genetically-based and environmentally-induced parental effects (Lacey, 1996). Global warming has resulted in an upward shift in species optimum 314 elevation and latitude (Chen et al., 2011;Lenoir et al., 2008;Parmesan and Yohe, 2003), an observation similar to that reported by Parmesan C. (2006) as she pointed out that the best places 315 to seek potential changes in species ranges is at their altitudinal and latitudinal treelines (Parmesan, 316 317 2006). This suggested that climatic variables are the real causes of species' responses to 318 environmental signals. Compared with last three decades, the Pacific Northwest is expected to 319 warm about 0.8-2.9°C by mid-century (Leung et al., 2004;Duffy et al., 2006;Mote et al., 2008), and 1.5-5.4°C by the end of the century (Mote et al., 2008). Warming in this region will be 320 probably greater in summer (3.9°C on average) than in winter (2.7°C) (Mote et al., 2008). 321

322 In the hierarchical model for seed dormancy, the climatic variable, summer degree-days below  $0^{\circ}C$  (DD 0 summer) was not equal to zero only for 16 out of the total 83 study populations 323 324 and was projected to be zero for 80 populations in 2050s. We performed approximation by 325 removing  $DD_0$  summer from the model and the result of seed dormancy projection could be interpreted as: given the current-future difference in July hargreaves reference evaporation (*Eref*) 326 327 is more than one third higher than that in summer *Eref* (Jun.-Aug. period), seed dormancy will 328 increase in 2050s, and if such a difference is less than one third higher, seed dormancy will 329 diminish. These results also suggest that the allocation of summer *Eref* in July (i.e., seed 330 development period) was critical in the development of seed dormancy.

The ecosystem zones were significant in analyses of life-history traits using MANOVA (Table 2A) but not significant for seed dormancy and weight models (Table S2). This indicates that dynamic climatic variables are true driving effects on modulating life-history traits. On the other hand, gene flow as a single pollen grain carries half the number of alleles and may yield effective distance spanning from a few centimetres to thousands of kilometres (Nathan et al., 2008). The 'abundant center' model, which deals with the spatial distribution for populations across 337 species' ranges and its evolutionary potential, also has implications (Volis et al., 2014;Etterson 338 and Shaw, 2001;Parmesan, 2006), that is, compared with core populations, locally adapted 339 peripheral populations had lower adaptive potential and were outperformed in the novel 340 environment. Ecosystem zones like other geographic variables are not individual units to classify 341 life-history traits.

342 A number of ecologically and evolutionary relevant genes have been identified in Arabidopsis and some annuals, such as FLC (Flowering Locus C), SCR/SP11 (S-locus Cysteine-343 344 Rich protein/ S-locus Protein 11) (Shimizu et al., 2011; Amasino, 2010), as they provide genetic 345 diversity in adaptive evolution. Moreover, both seed dormancy (genes in ABA and GA signalling circuit) and flowering (such as FLC) were modulated by epigenetic mechanisms (Müller et al., 346 347 2012; Chinnusamy et al., 2008; Bossdorf et al., 2008). This indicates that epigenetic changes play 348 a significant role in evolution and ecology and the environment signals acting on genes by 349 epigenetic modification were crucial for life-history traits. In recent years, much progress has been 350 made in uncovering genes operating on different seed compartments i.e., embryo, endosperm, and 351 seed coat, which modulate seed development (Le et al., 2010). Processes that regulate seed size 352 and development are coordinated across several morphologically distinct sub-regions (Belmonte 353 et al., 2013) and the complex cross-talk and integration of signals from different components of 354 the seed together determine its final size (Garcia et al., 2005).

## 355 Temperature signals in winter-chilling

Winter dormancy is an important adaptive strategy, as it prevents plants from flushing during short warm periods in the winter. Winter chilling is an important environmental signal for plant life histories, which accelerates flowering through vernalization in winter annuals and alleviates both bud and seed dormancy, allowing the onset of growth in springs (Penfield, 2008;Penfield and Springthorpe, 2012). Seeds take advantage of environmental temperature as a key signal to coordinate timing of seed germination, allowing plants to synchronize their life histories with the seasons. Low temperature can promote dormancy at the inception of seed maturation, but promotes dormancy alleviation in mature seeds after imbibition. It is therefore assumed that chilling plays a dual role in regulating dormancy (Batlla and Benech-Arnold, 2010).

365 With substantial climate warming  $(>3^{\circ}C)$ , chilling may be insufficient in many woody perennials, such as, poplar, western hemlock, and Sitka spruce, resulting in delayed bud burst and 366 367 poor growth (Cannell and Smith, 1986; Murray et al., 1989; Morin et al., 2009). Based on the three 368 different RCP scenarios, we predicted that winter-chilling days (DD 0 winter) would on average decrease by 24% across the 83 study habitats in 2050s relative to present (Fig. S8, ranges also 369 370 provided), which may lead to insufficient dormancy alleviation through winter-chilling. However, 371 spring and annual heat sums (DD5\_spring and DD5) would increase by 95 and 49%, respectively, in 2050s (Fig. S8), resulting in earlier springs and advanced vegetative green-ups and an increase 372 in growing season length (Robeson, 2004;Schwartz et al., 2006). In 2050s, inadequate winter-373 374 chilling may delay germination and an extended germination span leads to adverse conditions during dry summers. As such, future climate will change the timing of conifer phenology and may 375 376 give rise to adverse consequences.

In terms of molecular mechanisms in alleviation of seed dormancy via winter-chilling, several studies demonstrated that moist-chilling involves changes in levels of ABA, GAs, and auxin, and transcripts in respective signaling cascades, GA3 oxidase 1, for example (Yamauchi et al., 2004;Ali-Rachedi et al., 2004;Liu et al., 2013a;Liu et al., 2015).

### 381 Germination cues in the seed-to-plant transition

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382 Changes in climate alter patterns of phenology and thus multiple life-history traits. In response to climate change, phenotypic plasticity associated with life-history traits (Pigliucci, 2001;Chevin et 383 384 al., 2010; Franks et al., 2014; Liu and El-Kassaby, 2015) and genetically based trait responses 385 (Bradshaw and Holzapfel, 2001, 2008; Thompson et al., 2013) were well documented. In the long run, only the species that can respond by phenotypic plasticity and/ or genetically-based local 386 387 adaptation can persist (Jump and Penuelas, 2005). However, the evolutionary response to climate 388 change may be attenuated due to constraints causing a time lag between the environmental change 389 and an observed evolutionary response (Etterson and Shaw, 2001; Davis et al., 2005). In tree 390 populations, the extent of the constraints will hinge on phenotypic variation, strength of selection, fecundity, interspecific competition, and biotic interaction (Aitken et al., 2008). 391

Differences in plasticity exist among populations, but plasticity presumed to be adaptive may often be neutral or maladaptive (Caruso et al., 2006). An optimal balance between adaptive and non-adaptive (bet-hedging) plasticity may exist and possibly vary among populations. Moreover, germination characters are expected to exhibit phenotypic plasticity to environmental variables experienced not only by seed following dispersal but also by seed parents prior to dispersal (Schmitt et al., 1992b).

It is noteworthy that seed dormancy and germination is a quantitative trait that interacts with environment factors (Bentsink et al., 2007). Some QTLs associated with germination phenology in Arabidopsis can attain allele frequencies approaching fixation within a single generation even though they started with frequencies below 50% (Huang et al., 2010), indicating strong directional selection. Seed dormancy in Arabidopsis was associated with a cohort of genes controlled by seasonally distinct hormone-signalling pathways in the seed soil bank, such as *DOG1* (Delay Of Germination 1), *MFT* (Mother of Flowering Timing), *DELLAs* (repressors of germination potential and GA signalling), and *PIF*s (Phytochrome Interacting Factors) (Footitt et
al., 2014;Footitt et al., 2011). In addition, photoperiod can affect seed dormancy and germination
and phytochromes were the most investigated photoreceptors. Phytochromes are temperature- and
light-dependent in association with GA pathway via the bHLH transcription factor SPATULA
(SPT) (Heschel et al., 2007). SPT is a light-stable repressor of seed germination and mediates the
germination response to temperature through temperature-sensitive changes in its transcription
(Penfield et al., 2005).

412 In summary, we reinforced the importance of climatic signals during seed set to the 413 formation of early life-history traits (i.e., seed dormancy and size). Variations of life-history traits 414 may be acquired through signal transduction cascades and/ or gene/ protein imprintings triggered by specific environmental variables, such as, evapotranspiration and precipitation. The period of 415 416 winter chilling exerts a pronounced influence on the range of life-history plasticity and the 417 variation of bet-hedge strategy when the life-history trait expressed (i.e., seed emergence). In 418 response to previous memory during seed development and different genetic architectures that 419 every individual harbours, germination behaviour (e.g., timing of seed germination) varies even 420 given same optimal germination cues.

421

#### 422 AUTHOR'S CONTRIBUTION

YL conceived of the study, carried out data analysis, and drafted the manuscript; TW participated
in selection of study populations, performed current and future climate predictions and mapping,
and commented the manuscript; and YAE coordinated the study and helped draft the manuscript.
All authors gave final approval for publication.

427

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## 437 SUPPLEMENTARY MATERIAL

- 438 **Table S1** A list of important climatic variables
- 439 Table S2 Canonical correlation analysis (CCA) and MANOVA statistical test criteria
- 440 **Table S3** Errors from ecosystem zones in life-history traits' models
- 441 **Table S4** Parameter estimates and statistical tests for the geographic variables-based hierarchical
- 442 models regarding life-history traits
- 443 **Figure S1** Geographical distribution of the 83 populations (green triangles) on the map regarding
- 444 AMT value (A) and ecosystem zones (B) of British Columbia, Canada.
- 445 Figure S2 The distribution of the samples based on mean annual temperature (MAT) against
  446 annual heat moisture index (AHM).
- Figure S3 Schematic representation of the cumulative germination curve parameters used to
  characterize seed dormancy.

- 449 *DI* (dormancy index): the gray shaded area, the difference of area under germination curves (*AUC*)
- 450 of control and any other treatment; GS (germination speed): the number of days to reach 50% of
- 451 final germination; *GC* (germination capacity): the final germination percentage.
- 452 **Figure S4** PLS biplot of 194 climatic variables (green), 83 populations (red), and 21 (above) 20
- 453 (below) categories for 83 population habitats (blue).
- 454 Figure S5 PCA for 194 climatic variables. The 15 most correlated climatic variables were marked
  455 in red.
- 456 **Figure S6** Seed dormancy (*DI*, above) and weight (*SW*, below) distribution for the 83 populations
- 457 labeled on the map of British Columbia, Canada.
- 458 **Figure S7** Linear relationship between 1,000-seed weight and 'predicted' 1,000-seed weight using
- 459 83 populations.
- 460 **Figure S8** The amount of changes for *DD\_0\_winter*, *DD5\_spring*, and *DD5* in 2050s relative to
- 461 present.
- 462 Note: The population was ranked as per its current *DI* in ascending order. Each data point is the
- 463 average of predictions using three methods (RCP2.6, 4.5, and 8.5). Bars indicate the SEM.
- 464
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#### 711 FIGURE LEGENDS

Figure 1 The 15 most correlated climatic variables with seed dormancy (A) and size (B) after
partial least squares (PLS) regression.

- 714 Note: ① three variables (*Eref07* and *\_summer*, and *PPT10*) were highly correlated with both seed
- dormancy and size; 2) fonts in black and grey represent the temperature-and precipitation-based
- climatic variables, respetively; ③ see Table S1 for full names of abbreivated variables.
- 717 Figure 2 Seed dormancy (DI) prediction in 2050s. (A) Linear relationship between predicted DI
- and DI used for the correction of predicted seed dormancy; (B) Corrected seed dormancy
- prediction using Representative Concentration Pathways (RCP) 2.6, 4.5 and 8.5, respectively.
- Note: (1) model for seed dormancy prediction,  $DI = (2.4819 + ecosystem zone) + 2.7079 \times Eref07$
- $0.9284 \times Eref\_summer + 6.6170 \times DD\_0\_summer$  [refer to Table S2 for intercept adjustment in
- each ecosystem zone]; (2) the population was ranked as per its current DI in ascending order and
- 723 95% confidence interval (CI) was plotted.
- Figure 3 Map of predicted seed dormancy using climatic model for the reference period and 2050s
  in lodgepole pine across British Columbia.
- Note: *DI*s (i.e., dormancy indexes) are classified into five categories and represented by different
  colours on the map. The higher the value, the more dormant the seeds.
- **Figure 4** Studies of phenotypic plasticity. (A) Reaction norms for *AUC* with or without moistchilling treatment across the 83 populations; (B) Relationship between phenotypic plasticity and *AUC* with or without moist-chilling treatment. Best fit lines based on linear models are provided. **Figure 5** Studies of bet-hedge strategy. Bet-hedge dormancy involving germination capacity and time to germination (left) and their STD (right) in response to 'good-year' and 'bad-year' treatments.

- Note: the population was ranked as per its current *DI* in ascending order.
- Figure 6 Important environment stages and intrinsic mechanisms in the life cycle of lodgepole
  pine for life-history traits.
- Note: ① \*, seed dispersal may not occur in the following season and it may take years until the
- cone drops and seeds release in nature; <sup>(2)</sup> the reproduction cycle of (lodgepole) pine undergoes
- 739 three years, that is, seed/pollen cone initiation (A S O)  $\rightarrow$  dormant (N D J F M)  $\rightarrow$  pollen
- 740 development (A M)  $\rightarrow$  pollination (J)  $\rightarrow$  female gametophyte development (J A)  $\rightarrow$  dormant (S O
- 741 N D J F M A)→ female gametophyte development (M J). (Initial letters in the bracket represent
- 742 months.)

## 743 Table 1 Canonical correlation analysis between individual variables and their own and opposite set

## 744 of variables

745 A)

Varia	ables	Life-history traits	Climatic Ecology <sup>†</sup>	Geographic Ecology <sup>††</sup>
Life-history	Seed weight	0.7429	0.9021	0.5481
traits	DI	0.5909	0.6582	0.5327

746 Note:

<sup>†</sup>Based on the PLS analysis, 27 most correlated climatic variables for seed dormancy (*DI*) and weight were

vised as Climatic Ecology (three climatic variables were strongly correlated with both seed dormancy and

749 weight).

<sup>††</sup>Geographic Ecology consists of latitude, longitude, and elevation.

751 B)

Canonical variate	% explained variance by			
Califonneal variate	Their own	The opposite		
Life-history traits	62.35%	47.11%		
Climatic Ecology	62.85%	45.06%		
Geographic Ecology	14.80%	6.82%		

752

753Table 2 Multi- and uni-variate analyses for life-history traits. (A) MANOVA table reporting the754correlation of seed dormancy (DI) and weight with ecosystems; (B) ANOVA table reporting755significant effect of the genotype on phenotypic variance and associating broad-sense heritability ( $H^2$ ).

756 A)

	_	Parameters						
SOV		DI			Seed weight			
	DF	Mean squares	F-value	$\mathbb{R}^2$	Mean squares	<i>F</i> -value	$\mathbb{R}^2$	
Ecosystem zones	21	130.3132	3.50**	0.55	0.3011	3.63**	0.56	
corrected error	61	37.1821		0.33	0.0829		0.30	

757 SOV, source of variation; DF, degrees of freedom; EMS, expected mean squares.

758 \*\* P < 0.025 (= 0.05/2 dependent variables).

759 B)

SOV	DF	EMS	Mean squares	<i>F</i> -value	Variance components	$H^2$
Genotype (G)	82	$\sigma_{\epsilon}^2 + 8\sigma_{G}^2$	346.8691	67.01**	54.02%	
Environment (E)	1	$\sigma_{\epsilon}^2 + 4\sigma_{G \times E}^2 + 332\phi_E$	89393.0625	17268.20**	_¶	0.54
$\mathbf{G}  imes \mathbf{E}$	82	$\sigma_{\epsilon}^2 + 4\sigma_{G \times E}^2$	129.8935	25.09**	39.43%	0.54
Error (ε)	498	$\sigma_{\epsilon}^{2}$	5.1767		6.55%	

<sup>¶</sup>no variance components or percent of total variation were estimated for the fixed effect (E).

761 \*\* *P* < 0.05.

## 762 Table 3 Parameter estimates and statistical tests for the climatic variables-based hierarchical models

## 763 regarding seed dormancy and weight

## A) Seed dormancy model

Effect	Estimate	Standard Error	DF	Statistic	Р
Intercept <sup>¶</sup> ( $\beta_0$ )	2.4819	7.9418	21	<i>t</i> = 0.31	0.7577
<i>Eref07</i> ( $\beta_1$ )	2.7076	0.8932	58	F = 9.19	0.0036
<i>Eref_summer</i> ( $\beta_2$ )	-0.9284	0.3354	58	F = 7.66	0.0076
$DD_0$ _summer ( $\beta_3$ )	6.5170	1.8480	58	F = 9.06	0.0039

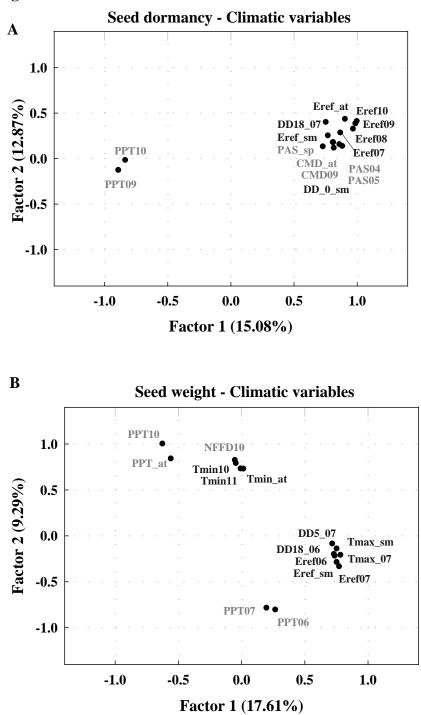
765 Tresidual from the ecosystem zone  $(\varepsilon_j)$  is integrated into 'intercept' (Table S3).

# 766 B) Seed weight model

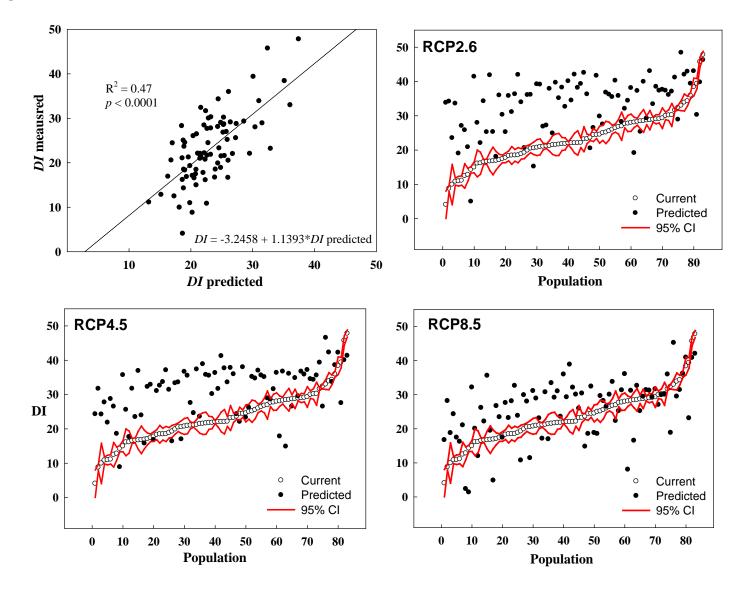
Effect	Estimate	Standard Error	DF	Statistic	Р
Intercept <sup>¶</sup> ( $\beta_0$ )	1.3290	0.3799	21	<i>t</i> = 3.50	0.0021
$PPT07 (\beta_1)$	0.0046	0.0019	58	F = 5.66	0.0207
<i>PPT10</i> (β <sub>2</sub> )	-0.0014	0.0007	58	F = 4.68	0.0346
Tmax07 ( $\beta_3$ )	0.0721	0.0150	58	F = 23.11	<0.0001

767 Tresidual from the ecosystem zone  $(\varepsilon_j)$  is integrated into 'intercept' (Table S3).

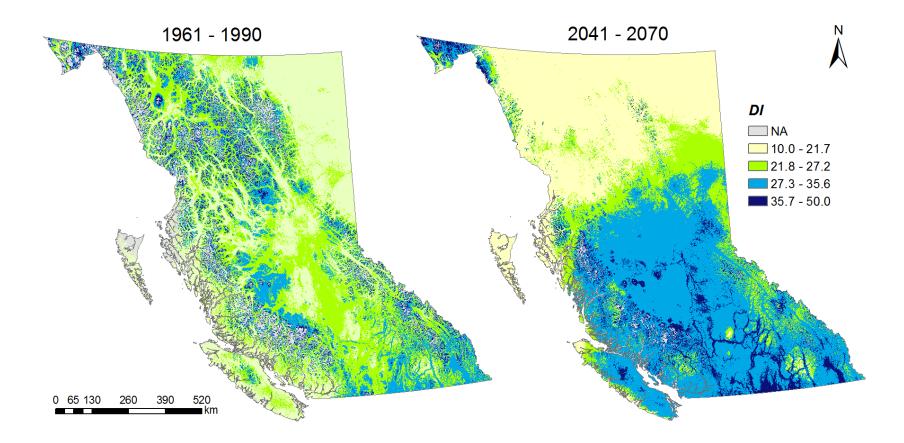




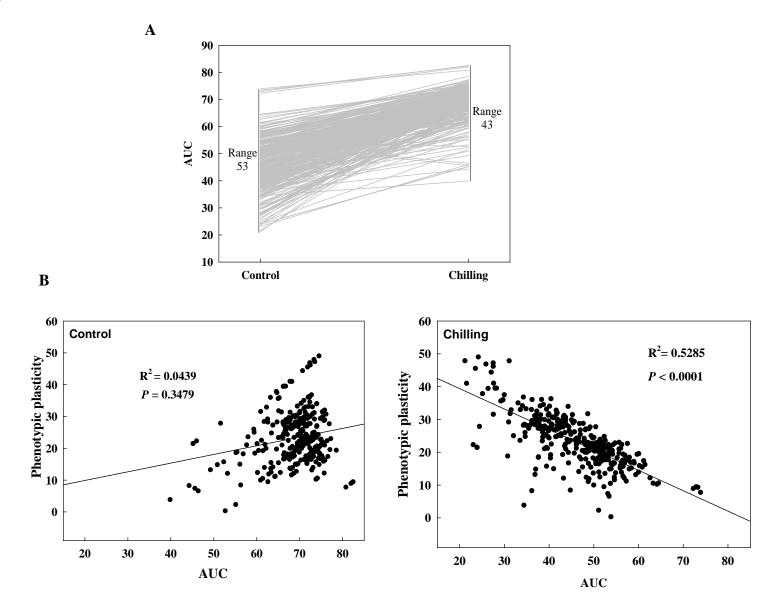




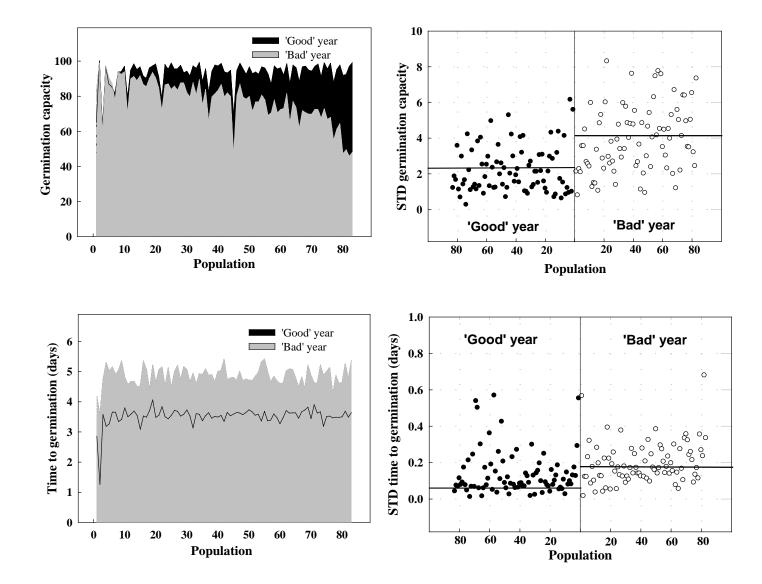


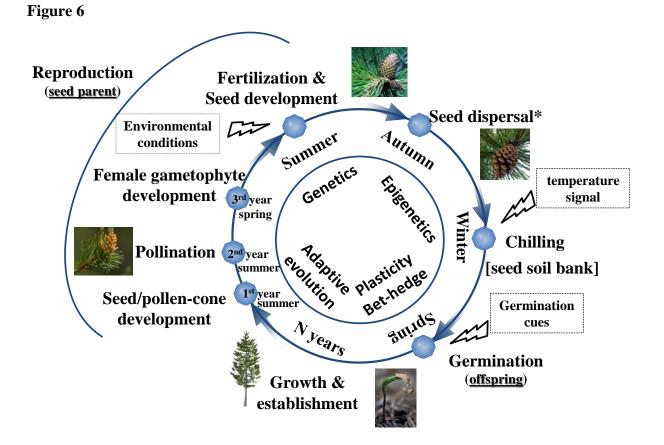












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