1	Contributions of dynamic environmental signals during life-cycle transitions
2	to early life-history traits in lodgepole pine (Pinus contorta Dougl.)
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23 Abstract

24 Environmental signals are important triggers in the life-cycle transitions and play a crucial role in the life-history evolution. Yet, very little is known about the leading ecological factors contributing 25 26 to the variations of life-history traits in perennial plants. This paper explores both the causes and 27 consequences for the evolution of life-history traits (i.e., seed dormancy and size) in lodgepole 28 pine (*Pinus contorta* Dougl.) across British Columbia (B.C.), Canada. We selected 83 logepole pine populations covering 22 ecosystem zones of B.C. and through their geographic coordinate, 29 197 climatic variables were generated accordingly for the reference (1961-1990) and future (2041-30 31 2070) periods. We found that dynamic climatic variables rather than constant geographic variables 32 are the true environmental driving forces in seed dormancy and size variations and thus provide 33 reliable predictors in response to global climate change. Evapotranspiration and precipitation in the plant-to-seed chronology are the most critical climate variables for seed dormancy and size 34 35 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 36 life-history plasticity and lower the bet-hedge strategy in the seed-to-plant transition; however, 37 38 winter-chilling is likely to be insufficient in the north of 49°N in 2050s, which may delay 39 germination while unfavourable conditions during dry summers may result in adverse consequences in the survival of seedlings owing to extended germination span. These findings 40 provide useful information to studies related to assessments of seed transfer and tree adaptation. 41

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

45 **INTRODUCTION**

In the context of global climate change, species have responded by tracking the environment for 46 which they are best suited through local adaptation, range shift, range reduction, or a combination 47 48 of these (Walther et al., 2002; Parmesan and Yohe, 2003; Cleland et al., 2007; Breshears et al., 2008). 49 Knowledge about traits that vary with ecological niches should therefore help predict how these 50 traits may evolve under climate change. Life-history traits, known as fitness components due to predictable monotonic relationship with fitness, are related to the timing and success of 51 52 development, reproduction, and senescence throughout the life cycle (Calow, 1998). Ecological and 53 genetic trade-offs constrain the combinations of the life-history traits through evolution (Adler et 54 al., 2014), which involves an adaptation to environments and is shaped by natural selection.

Climate change has already altered the timing of major life-history transitions, such as seed 55 germination timing (from seed to plant). Timing is directional and completely asymmetric and the 56 timing of seed germination is the earliest life-history trait that is expressed and sets the context for 57 the traits that follow (Donohue et al., 2010). It is controlled by the level of seed dormancy, whereby 58 59 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 60 61 to environment perturbations than adult stages and represent a major bottleneck to regeneration 62 from seeds (Johnsen and Skrøppa, 1996;Hedhly et al., 2009). At the core of plant regeneration, 63 temperature and water availability (or precipitation) are critical drivers for a plant's distribution 64 (Woodward and Williams, 1987). Hence, climate-changed plant regeneration will be manipulated in both temperature- and moisture-controlled ecosystems (Walter and Breckle, 2002) and much of 65 66 plant regeneration climate research has been directed at tundra and boreal forest, and treeline ecotones (reviewed in (Walck et al., 2011)). 67

68 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 69 plants until the Cretaceous period (124 MYA). Seed size begins to increase after the limit of the 70 71 Cretaceous-Tertiary period (65 MYA). One of the most popular explanations is that climate change gives rise to seed size variation, which took place during that period (Eriksson et al., 2000). In the 72 73 ecological time scale, empirical studies lend supports to the impact of environment on seed size 74 variation. Since the early 1950s, effects of environmental stimuli, such as, temperature and photoperiod on seed size and weight have been noted. *Chenopodium polyspermum* L. seeds from 75 76 mother plants grown in long days have lower germination frequency and thicker seed coats when 77 compared to seeds from short days (Pourrat and Jacques, 1975). Large seed size with more provisions stored for seedlings may be favorable in variable environments (Venable and Brown, 78 79 1988). Seed plants have a general trend of increasing embryo to seed ratio (E:S) in morphological 80 seed dormancy and the shift in E:S is likely a heterochronic change, having vital implications to 81 life history of seed plants (Forbis et al., 2002). A priori, the environment plays a crucial role in 82 life-history traits in general and seed size in particular.

Adaptive phenotypic plasticity underpins rapid phenological shifts in response to climate 83 84 change and evolve when cues reliably predict fitness consequences of life-history decision (Simons, 2014). This was evident by the performance of seedlings produced by central European 85 trees growing in central Norway as they expressed phenology similar to that of their adjacent 86 87 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 88 genetic framework and probably modulated by epigenetic mechanisms (Yakovlev et al., 2012). 89 90 Germination-cuing under favorable conditions is similar within species; however, these conditions

91 do not persist throughout the seasons. Likewise, populations growing in different ecological niches 92 may have different degrees of exposure to unfavourable environments. As such, the prevention of germination of some seeds even under favourable conditions would be of significance. When 93 94 environments fluctuate unpredictably, "bet-hedging" strategy (Slatkin, 1974) is expected to spread 95 germination over time to reduce the risk of outright extinction. This results in the evolution of traits that maximizes the geometric-mean fitness by reducing fitness variance over generations 96 97 (Gillespie, 1977). To date, much of the evidence for bet-hedging remains restricted to simple lifehistory plants (e.g., annuals) (Childs et al., 2010). 98

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).

105 Life-history strategy for long-lived organisms is influenced primarily by survival (Adler et 106 al., 2014). Evergreen coniferous forests in the Pacific Northwest are unique among the Northern 107 Hemisphere temperate forests in their species composition and high productivity (Waring and 108 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, 109 110 Canada and is therefore expected to display a wide spectrum of dormancy variation owing to 111 adaptation to diversified local habitats (Plomion et al., 2007). Cone and seed production in lodgepole pine is not as cyclic as in many other conifers. Cone drop soon after seed maturation is 112 113 commonly observed in coastal areas but lodgepole pine growing in the interior tends to be

114 serotinous, indicating that mature cones do not drop or open to release seeds unless exposed to 115 fluctuating high temperature during prolonged hot summer and low fall temperatures or due to fire 116 or insect damage (Fowells, 1965; Owens et al., 1981, 1982). Most serotinous cones take several 117 years to open and seeds are released in large quantities, consequently, any seed collection made 118 from a single tree consists of a mixture of different seed-crop years. Before the cones eventually 119 open, seeds are wrapped and sealed by scales and sticky resin without exposure to any germination 120 cues (such as moisture, oxygen, etc.); thus we assumed that seed dormancy in lodgepole pine is least affected over its natural storage period. 121

122 Seed dormancy is an intrinsic attribute affecting regeneration dynamics and seed size is 123 one of the vital determinants for the evolution of seed dormancy. The objectives of this study are 124 to evaluate the contributions of local environmental effects during life-cycle transitions to seed 125 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 126 127 dormancy and size allow investigating the relationship between these two life-history traits and 128 their relationship with environments in life-cycle transitions. This study can also contribute to 129 providing the missing empirical evidence of bet-hedging strategy in long life-history plants.

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131 METHODS

132 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.

141 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 142 143 climate predictions using historical weather station data and global circulation models) (Wang et 144 al., 2012). The same 197 climatic variables for the future period between 2041 and 2070 (or 2050s) were also projected using ClimateWNA. The future climate data were downscaled to point 145 146 locations using a delta approach (Wang et al., 2012). We used three Representative Concentration 147 Pathways (RCP) 2.6, 4.5 and 8.5 from the CGCM4 model output of the phase 5 of the Coupled Model Intercomparison Project (CMIP5), generated by the United States National Center for 148 149 Atmospheric Research. The CCSM4 model output was included in the Intergovernmental Panel on Climate Change (IPCC) Fifth Assessment Report (AR5) (Stocker and Randall, 2013). 150 Moreover, CCSM4 is close to the average in both temperature and precipitation increases in the 151 152 study areas (B.C.). In the prediction of seed dormancy indicators across B.C. for the reference 153 period and 2050s, climate data were generated for each pixel at the spatial resolution of 800×800 154 m using ClimateWNA. The full list of climatic variables is given in Table S1 and the four most 155 relevant variables for the present study were *Eref07* and *_summer* (July and summer hargreaves reference evaporation estimated based on temperature and solar radiation), DD_0_summer 156 (summer degree-days below 0°C), PPT07 and 10 (July and October precipitation (mm)), and 157 *Tmax07* (July maximum mean temperature (°C)). 158

159 Life-history traits' measurement

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160 Germination parameters were estimated from cumulative germination curves fitting a 161 mathematical function known as the four-parameter Hill function (El-Kassaby et al., 2008). Core germination parameters include dormancy index (DI, the difference of areas under germination 162 163 curves (AUC) of control and treated seed), germination capacity (GC, the final germination fraction), time to germination (time at the onset of germination), and germination speed (GS, the 164 165 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 166 retain natural seed dormancy, seed germination was manipulated with or without 21-day moist-167 168 chilling treatment under 3°C in lightproof chamber. Germination assay was performed as 169 previously described (Liu et al., 2013b) and seed size was measured as the average of 1000-seed weight collected from each study site. 170

171 Data analysis and visualization

172 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; 173 174 SAS Institute Inc., Cary, NC) (Jose Crossa et al., 2013). The goal of PLS regression is to analyse multivariate response based on a large set of explanatory variables (i.e., climatic variables). This 175 176 technique combines features from principle component analysis (PCA) and multiple regression 177 (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 178 179 values. Using the machine-learning algorithm (i.e., PLS), important climate variables can be 180 identified through a process of model optimization and there is no needed to explain the 181 contribution of climate variables in the initial input dataset. In the PLS biplot, the angles between 182 the environment vectors proximately correspond to the correlation coefficients among the

183 environments. The cosine of the angle between two vectors approximates the correlation between 184 them and the length of the vectors is proportional to the standard deviation within respective environments or genotypes (Yan, 2001; Yan and Tinker, 2006). Virtually, an ideal environment 185 186 has the longest vector of all test environments (most discriminating and informative) and is closely located on the abscissa (most representative) (Liu and El-Kassaby, 2015). As a comparison, PCA 187 for explanatory variables and canonical correlation analysis (CCA) dealing with both explanatory 188 189 and response variables were also performed (Liu and El-Kassaby, 2015). To examine how seed 190 dormancy and size varied in ecosystem zones, the two variables were analyzed collectively using 191 multivariate analyses (MANOVA) with the aid of general linear model (GLM) procedure in SAS 192 (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 193 194 MIXED in SAS with errors split into population and ecosystem zones but having unequal 195 variances at the population level (West et al., 2007;Raudenbush and Bryk, 2001). The most 196 correlated climatic variables through the PLS analysis were used and the model was expressed as 197 follows:

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

where *i* and *j* represent two levels, namely, *i*th population within *j*th ecosystem zone; P_{*kij*} and T_{*lij*} represent the *k*th precipitation- and *l*th temperature-based variable in *i*th population within *j*th ecosystem zone, respectively; ε_j and ε_{ij} represent errors from the ecosystem zone and population level, respectively. Intercept ($\beta_0 + \varepsilon_j$) and coefficients (β_k and β_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 800×800 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:

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

where DI_{ij} is dormancy index of i^{th} genotype in j^{th} environment, μ 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 ε 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 ($H^2 = V_G / (V_G + V_{G \times E} + V_{\varepsilon})$).

219 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 220 221 of Cohen's classic bet-hedging model (Cohen, 1966), in which dormancy is expected to evolve in 222 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. 223 224 chilling) represents within-season variation, which has been shown to be advantageous when 225 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 226 227 variation across studied populations.

228

229 **RESULTS**

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

Partial least squares (PLS) analyses for seed dormancy and weight indicated that the first and 231 232 second components accounted for 15 and 13%, and 18 and 9% of the total variation, respectively 233 (Fig. S4). The 15 most correlated climatic variables with respect to life-history traits were 234 correlated with environmental factors related to temperature (Fig. 1). This indicates that 235 temperature plays a major role in the development of life-history traits (Liu and El-Kassaby, 2015). By contrast, the PCA analysis showed several temperature-based variables which were 236 237 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 238 239 S4).

The first canonical correlation analysis (CCA) was significant across all statistical tests 240 (Table S2A). Seed dormancy and weight had moderate (0.66) and very strong (0.90) correlations 241 with climate, moderate (0.59) and high (0.74) correlations with life-history traits (i.e., combination 242 243 of seed dormancy and weight), and moderate (0.53-0.55) correlation with geographic variables (Table 1A). This implies that the climatic variables were more important than the geographic 244 245 variables in variations of life-history traits. Based on respective covariate matrices for 'life-history traits', 'climatic ecology', and 'geographic ecology' canonical variates, 62.35 and 47.11%, 62.85 246 247 and 45.06%, 14.80 and 6.82% of variance were explained by their corresponding canonical 248 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, 249 and had moderate correlation ($R^2 = 0.55 \cdot 0.56$, P < 0.0001) with life-history traits (Table 2). 250

In the hierarchical model for seed dormancy using climatic variables, the climatic variables 251 252 *Eref07, Eref summer*, and *DD 0 summer* were significant (Table 3A). This indicates that summer 253 moisture and temperature, the period corresponding to the plant-to-seed transition, played an 254 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 255 256 intercept were significant (Table 3B), indicating that July temperature and precipitation and 257 October precipitation were important to seed weight. Ecosystem zones, representing ecological 258 boundaries in geography, had a greater impact on the variation of seed weight than that of seed 259 dormancy. In the geographic variable-based hierarchical model for seed dormancy, only longitude and intercept were significant (Table S4A and Fig. S6A). For seed weight, longitude, elevation 260 and intercept were significant (Table S4B and Fig. S6B), indicating that longitude was an 261 262 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-263 history traits were significantly influenced by precipitation as well as temperature in the plant-to-264 265 seed chronology, and roughly distributed in a longitude pattern.

266 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).

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

Among the 83 lodgepole pine populations, variance of seed dormancy was significantly explained 277 278 by genotype and environment (Table 2B). Estimate of broad-sense heritability (H^2) was 0.54 279 corresponding with the observed variance component for genotypes (Table 2B), indicating that genetic components have moderate influence on seed dormancy variation. In response to 280 281 significant $G \times E$ interaction (Table 2B), a plot of the reaction norms of AUC showed a range shift, 282 nonparallelism, and crossing between moist-chilling and control (Fig. 4), where $G \times E$ accounted 283 for 39% of the variance and was attributed to crossing of reaction norms (Table 2B), indicating that environments changed the adaptive values of the life-history trait. 284

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.

291 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.

300

301 **DISCUSSION**

302 Climate change is accelerating plant life-cycle transitions in coordination with the seasons. In life-303 cycle transitions, the environment plays a critical role in the development of life-history traits and 304 in response to environmental stimuli, intrinsic mechanisms of genetics, epigenetics, phenotypic 305 plasticity, bet-hedge strategy, and adaptive evolution take effect (Fig. 6). In this study, we found 306 that seed dormancy and size were most correlated with evapotranspiration, and precipitation and 307 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 308 309 dormancy variation would increase across B.C. in 2050s. Moreover, winter-chilling can increase 310 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 311 required to decay seed dormancy, thus resulting in adverse consequences for the timing of 312 phenology and the growth and establishment in lodgepole pine. This study allowed us to gain 313 314 insights in the role of dynamic environments in shaping life-history characteristics.

315 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., 1992), 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 320 elevation and latitude (Chen et al., 2011;Lenoir et al., 2008;Parmesan and Yohe, 2003), an 321 observation similar to that reported by Parmesan C. (2006) as she pointed out that the best places 322 to seek potential changes in species ranges is at their altitudinal and latitudinal treelines (Parmesan, 323 2006). This suggested that climatic variables are the real causes of species' responses to environmental signals. Compared with last three decades, the Pacific Northwest is expected to 324 warm about 0.8-2.9°C by mid-century (Leung et al., 2004; Duffy et al., 2006; Mote et al., 2008), 325 326 and 1.5-5.4°C by the end of the century (Mote et al., 2008). Warming in this region will be probably greater in summer $(3.9^{\circ}C \text{ on average})$ than in winter $(2.7^{\circ}C)$ (Mote et al., 2008). 327

328 In the hierarchical model for seed dormancy, the climatic variable, summer degree-days 329 below 0°C (DD_0_summer) was not equal to zero only for 16 out of the total 83 study populations and was projected to be zero for 80 populations in 2050s. We performed approximation by 330 331 removing DD_0_summer from the model and the result of seed dormancy projection could be 332 interpreted as: given the current-future difference in July hargreaves reference evaporation (*Eref*) is more than one third higher than that in summer Eref (Jun.-Aug. period), seed dormancy will 333 334 increase in 2050s, and if such a difference is less than one third higher, seed dormancy will diminish. These results also suggest that the allocation of summer Eref in July (i.e., seed 335 336 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 343 species' ranges and its evolutionary potential, also has implications (Volis et al., 2014;Etterson 344 and Shaw, 2001;Parmesan, 2006), that is, compared with core populations, locally adapted 345 peripheral populations had lower adaptive potential and were outperformed in the novel 346 environment. Ecosystem zones like other geographic variables are not individual units to classify 347 life-history traits.

A number of ecologically and evolutionary relevant genes have been identified in 348 Arabidopsis and some annuals, such as FLC (Flowering Locus C), SCR/SP11 (S-locus Cysteine-349 Rich protein/ S-locus Protein 11) (Shimizu et al., 2011; Amasino, 2010), as they provide genetic 350 351 diversity in adaptive evolution. Moreover, both seed dormancy (genes in ABA and GA signalling 352 circuit) and flowering (such as FLC) were modulated by epigenetic mechanisms (Müller et al., 2012; Chinnusamy et al., 2008; Bossdorf et al., 2008). This indicates that epigenetic changes play 353 354 a significant role in evolution and ecology and the environment signals acting on genes by epigenetic modification were crucial for life-history traits. In recent years, much progress has been 355 356 made in uncovering genes operating on different seed compartments i.e., embryo, endosperm, and 357 seed coat, which modulate seed development (Le et al., 2010). Processes that regulate seed size 358 and development are coordinated across several morphologically distinct sub-regions (Belmonte 359 et al., 2013) and the complex cross-talk and integration of signals from different components of the seed together determine its final size (Garcia et al., 2005). 360

361 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).

With substantial climate warming $(>3^{\circ}C)$, chilling may be insufficient in many woody 371 372 perennials, such as, poplar, western hemlock, and Sitka spruce, resulting in delayed bud burst and 373 poor growth (Cannell and Smith, 1986; Murray et al., 1989; Morin et al., 2009). Based on the three 374 different RCP scenarios, we predicted that winter-chilling days (DD 0 winter) would on average 375 decrease by 24% across the 83 study habitats in 2050s relative to present (Fig. S8, ranges also provided), which may lead to insufficient dormancy alleviation through winter-chilling. However, 376 377 spring and annual heat sums (DD5_spring and DD5) would increase by 95 and 49%, respectively, 378 in 2050s (Fig. S8), resulting in earlier springs and advanced vegetative green-ups and an increase in growing season length (Robeson, 2004;Schwartz et al., 2006). In 2050s, inadequate winter-379 380 chilling may delay germination and an extended germination span leads to adverse conditions 381 during dry summers. As such, future climate will change the timing of conifer phenology and may 382 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).

387 Germination cues in the seed-to-plant transition

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388 Changes in climate alter patterns of phenology and thus multiple life-history traits. In response to 389 climate change, phenotypic plasticity associated with life-history traits (Pigliucci, 2001;Chevin et 390 al., 2010; Franks et al., 2014; Liu and El-Kassaby, 2015) and genetically based trait responses 391 (Bradshaw and Holzapfel, 2001, 2008; Thompson et al., 2013) were well documented. In the long 392 run, only the species that can respond by phenotypic plasticity and/ or genetically-based local 393 adaptation can persist (Jump and Penuelas, 2005). However, the evolutionary response to climate 394 change may be attenuated due to constraints causing a time lag between the environmental change and an observed evolutionary response (Etterson and Shaw, 2001; Davis et al., 2005). In tree 395 396 populations, the extent of the constraints will hinge on phenotypic variation, strength of selection, 397 fecundity, interspecific competition, and biotic interaction (Aitken et al., 2008).

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., 1992).

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).

In summary, we reinforced the importance of climatic signals during seed set to the 418 419 formation of early life-history traits (i.e., seed dormancy and size). Variations of life-history traits 420 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 421 422 winter chilling exerts a pronounced influence on the range of life-history plasticity and the 423 variation of bet-hedge strategy when the life-history trait expressed (i.e., seed emergence). In 424 response to previous memory during seed development and different genetic architectures that 425 every individual harbours, germination behaviour (e.g., timing of seed germination) varies even 426 given same optimal germination cues.

427

428 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.

433

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437

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443 SUPPLEMENTARY MATERIAL

- 444 **Table S1** A list of important climatic variables
- 445 **Table S2** Canonical correlation analysis (CCA) and MANOVA statistical test criteria
- 446 **Table S3** Errors from ecosystem zones in life-history traits' models
- 447 **Table S4** Parameter estimates and statistical tests for the geographic variables-based hierarchical
- 448 models regarding life-history traits
- 449 **Figure S1** Geographical distribution of the 83 populations (green triangles) on the map regarding
- 450 AMT value (A) and ecosystem zones (B) of British Columbia, Canada.
- 451 Figure S2 The distribution of the samples based on mean annual temperature (MAT) against
- 452 annual heat moisture index (AHM).
- 453 **Figure S3** Schematic representation of the cumulative germination curve parameters used to 454 characterize seed dormancy.

- 455 *DI* (dormancy index): the gray shaded area, the difference of area under germination curves (*AUC*)
- 456 of control and any other treatment; GS (germination speed): the number of days to reach 50% of
- 457 final germination; *GC* (germination capacity): the final germination percentage.
- 458 Figure S4 PLS biplot of 194 climatic variables (green), 83 populations (red), and 21 (above) 20
- 459 (below) categories for 83 population habitats (blue).
- 460 Figure S5 PCA for 194 climatic variables. The 15 most correlated climatic variables were marked461 in red.
- 462 **Figure S6** Seed dormancy (*DI*, above) and weight (*SW*, below) distribution for the 83 populations
- 463 labeled on the map of British Columbia, Canada.
- 464 **Figure S7** Linear relationship between 1,000-seed weight and 'predicted' 1,000-seed weight using
- 465 83 populations.
- 466 **Figure S8** The amount of changes for *DD_0_winter*, *DD5_spring*, and *DD5* in 2050s relative to
- 467 present.
- 468 Note: The population was ranked as per its current *DI* in ascending order. Each data point is the
- 469 average of predictions using three methods (RCP2.6, 4.5, and 8.5). Bars indicate the SEM.
- 470

471**REFERENCE**

- Abdi, H.: Partial least square regression Encyclopedia of measurement and statistics, edited by:
 Salkind, N. J., Sage Publications, Inc., Thousand Oaks, CA/London/New Dehli, 2007.
- Adler, P. B., Salguero-Gómez, R., Compagnoni, A., Hsu, J. S., Ray-Mukherjee, J., Mbeau-Ache,
 C., and Franco, M.: Functional traits explain variation in plant life history strategies, Proc Natl
 Acad Sci U S A, 111, 740-745, doi: 10.1073/pnas.1315179111, 2014.
- Aitken, S. N., Yeaman, S., Holliday, J. A., Wang, T. L., and Curtis-McLane, S.: Adaptation,
 migration or extirpation: climate change outcomes for tree populations, Evol Appl, 1, 95-111, doi:
 10.1111/j.1752-4571.2007.00013.x, 2008.

- Ali-Rachedi, S., Bouinot, D., Wagner, M. H., Bonnet, M., Sotta, B., Grappin, P., and Jullien, M.:
 Changes in endogenous abscisic acid levels during dormancy release and maintenance of mature
 seeds: studies with the Cape Verde Islands ecotype, the dormant model of *Arabidopsis thaliana*,
- 483 Planta, 219, 479-488, doi: 10.1007/s00425-004-1251-4, 2004.
- 484 Amasino, R.: Seasonal and developmental timing of flowering, Plant J., 61, 1001-1013, doi:
 485 10.1111/j.1365-313X.2010.04148.x, 2010.
- Baskin, C. C., and Baskin, M. J.: Seeds: ecology, biogeography, and evolution of dormancy and
 germination Academic Press, San Diego, California, 666 pp., 1998.
- Batlla, D., and Benech-Arnold, R. L.: Predicting changes in dormancy level in natural seed soil
 banks, Plant Mol. Biol., 73, 3-13, doi: 10.1007/s11103-010-9601-z, 2010.
- 490 Belmonte, M. F., Kirkbride, R. C., Stone, S. L., Pelletier, J. M., Bui, A. Q., Yeung, E. C.,
- 491 Hashimoto, M., Fei, J., Harada, C. M., Munoz, M. D., Le, B. H., Drews, G. N., Brady, S. M.,

492 Goldberg, R. B., and Harada, J. J.: Comprehensive developmental profiles of gene activity in 493 regions and subregions of the *Arabidopsis* seed, Proc Natl Acad Sci U S A, 110, E435-444, doi:

- 493 regions and subregions of the *Arab*494 10.1073/pnas.1222061110, 2013.
- 495 Bentsink, L., Soppe, W., and Koornneef, M.: Genetic aspects of seed dormancy, *In* Seed 496 development, dormancy and germination, Blackwell publishing Ltd Oxford, UK, 2007.
- Bossdorf, O., Richards, C. L., and Pigliucci, M.: Epigenetics for ecologists, Ecol. Lett., 11, 106115, doi: 10.1111/j.1461-0248.2007.01130.x, 2008.
- Bradshaw, W. E., and Holzapfel, C. M.: Genetic shift in photoperiodic response correlated with global warming, P Natl Acad Sci USA, 98, 14509-14511, doi: 10.1073/pnas.241391498, 2001.
- 501 Bradshaw, W. E., and Holzapfel, C. M.: Genetic response to rapid climate change: it's seasonal 502 timing that matters, Mol. Ecol., 17, 157-166, doi: 10.1111/j.1365-294X.2007.03509.x, 2008.
- Breshears, D. D., Huxman, T. E., Adams, H. D., Zou, C. B., and Davison, J. E.: Vegetation
 synchronously leans upslope as climate warms, Proc Natl Acad Sci U S A, 105, 11591-11592, doi:
 10.1073/pnas.0806579105, 2008.
- 506 Calow, P.: The encyclopedia of ecology and environmental management, Blackwell Science, New507 York, 1998.

- 508 Cannell, M. G. R., and Smith, R. I.: Climatic Warming, spring budburst and frost damage on trees,
- 509 J. Appl. Ecol., 23, 177-191, doi: 10.2307/2403090, 1986.

510 Carrascal, L. M., Galvan, I., and Gordo, O.: Partial least squares regression as an alternative to 511 current regression methods used in ecology, Oikos, 118, 681-690, doi: 10.1111/j.1600-512 0706 2008 16881 n 2000

- 512 0706.2008.16881.x, 2009.
- 513 Caruso, C. M., Maherali, H., and Sherrard, M.: Plasticity of physiology in Lobelia: testing for 514 adaptation and constraint, Evolution, 60, 980-990, 2006.

515 Chen, I. C., Hill, J. K., Ohlemüller, R., Roy, D. B., and Thomas, C. D.: Rapid range shifts of 516 species associated with high levels of climate warming, Science, 333, 1024-1026, doi: 517 10.1126/science.1206432, 2011.

- 518 Chevin, L. M., Lande, R., and Mace, G. M.: Adaptation, plasticity, and extinction in a changing 519 environment: Towards a predictive theory, PLoS Biol., 8, e1000357, doi: 520 10.1371/journal.pbio.1000357, 2010.
- 521 Childs, D. Z., Metcalf, C. J., and Rees, M.: Evolutionary bet-hedging in the real world: empirical
 522 evidence and challenges revealed by plants, P Roy Soc B-Biol Sci, 277, 3055-3064, doi:
 523 10.1098/rspb.2010.0707, 2010.
- 524 Chinnusamy, V., Gong, Z. Z., and Zhu, J. K.: Abscisic acid-mediated epigenetic processes in plant
 525 development and stress responses, J Integr Plant Biol, 50, 1187-1195, 2008.
- 526 Cleland, E. E., Chuine, I., Menzel, A., Mooney, H. A., and Schwartz, M. D.: Shifting plant 527 phenology in response to global change, Trends Ecol. Evol., 22, 357-365, 2007.
- 528 Cohen, D.: Optimizing reproduction in a randomly varying environment, J. Theor. Biol., 12, 119-529 129, 1966.
- 530 Davis, M. B., Shaw, R. G., and Etterson, J. R.: Evolutionary responses to changing climate, 531 Ecology, 86, 1704-1714, doi: 10.1890/03-0788, 2005.
- Donohue, K.: Completing the cycle: maternal effects as the missing link in plant life histories,
 Philos T R Soc B, 364, 1059-1074, doi: 10.1098/rstb.2008.0291, 2009.

- 534 Donohue, K., de Casas, R. R., Burghardt, L., Kovach, K., and Willis, C. G.: Germination,
- 535 postgermination adaptation, and species ecological ranges, Annu. Rev. Ecol., Evol. Syst., 41, 293-526 allo doi: 10.1146/appurey.coolaws.102200.144715, 2010
- 536 319, doi: 10.1146/annurev-ecolsys-102209-144715, 2010.
- 537 Duffy, P. B., Arritt, R. W., Coquard, J., Gutowski, W., Han, J., Iorio, J., Kim, J., Leung, L. R.,
- Roads, J., and Zeledon, E.: Simulations of present and future climates in the western United States
 with four nested regional climate models, J. Clim., 19, 873-895, doi: 10.1175/Jcli3669.1, 2006.
- 540 El-Kassaby, Y. A., Moss, I., Kolotelo, D., and Stoehr, M.: Seed germination: Mathematical 541 representation and parameters extraction, For. Sci., 54, 220-227, 2008.
- 542 Eriksson, O., Friis, E. M., and Löfgren, P.: Seed size, fruit size, and dispersal systems in
 543 angiosperms from the Early Cretaceous to the Late Tertiary, Am. Nat., 156, 47-58, doi:
 544 10.1086/303367, 2000.
- 545 Etterson, J. R., and Shaw, R. G.: Constraint to adaptive evolution in response to global warming,
- 546 Science, 294, 151-154, doi: 10.1126/science.1063656, 2001.
- Finch-Savage, W. E., and Leubner-Metzger, G.: Seed dormancy and the control of germination,
 New Phytol., 171, 501-523, 2006.
- Footitt, S., Douterelo-Soler, I., Clay, H., and Finch-Savage, W. E.: Dormancy cycling in
 Arabidopsis seeds is controlled by seasonally distinct hormone-signaling pathways, P Natl Acad
 Sci USA, 108, 20236-20241, doi: 10.1073/pnas.1116325108, 2011.
- Footitt, S., Clay, H. A., Dent, K., and Finch-Savage, W. E.: Environment sensing in springdispersed seeds of a winter annual Arabidopsis influences the regulation of dormancy to align
 germination potential with seasonal changes, New Phytol., 202, 929-939, doi: 10.1111/nph.12694,
 2014.
- Forbis, T. A., Floyd, S. K., and de Queiroz, A.: The evolution of embryo size in angiosperms and
 other seed plants: Implications for the evolution of seed dormancy, Evolution, 56, 2112-2125,
 2002.
- Fowells, H. A.: Silvics of forest trees of the United States, Agric. Handb. US Dep. Agric.,
 Washington D.C., 762 pp., 1965.
- Franks, S. J., Weber, J. J., and Aitken, S. N.: Evolutionary and plastic responses to climate change in terrestrial plant populations, Evol Appl, 7, 123-139, doi: 10.1111/Eva.12112, 2014.

563 Garcia, D., Fitz Gerald, J. N., and Berger, F.: Maternal control of integument cell elongation and

- zygotic control of endosperm growth are coordinated to determine seed size in *Arabidopsis*, Plant
 Cell, 17, 52-60, doi: 10.1105/tpc.104.027136, 2005.
- Gillespie, J.: Natural selection for variances in offspring numbers: a new evolutionary priciple,
 Am. Nat., 111, 1010-1014, doi: 10.1086/283230, 1977.
- Hedhly, A., Hormaza, J. I., and Herrero, M.: Global warming and sexual plant reproduction,
 Trends Plant Sci., 14, 30-36, doi: 10.1016/j.tplants.2008.11.001, 2009.

Heschel, M. S., Selby, J., Butler, C., Whitelam, G. C., Sharrock, R. A., and Donohue, K.: A new
role for phytochromes in temperature-dependent germination, New Phytol., 174, 735-741, doi:
10.1111/j.1469-8137.2007.02044.x, 2007.

Huang, X. Q., Schmitt, J., Dorn, L., Griffith, C., Effgen, S., Takao, S., Koornneef, M., and
Donohue, K.: The earliest stages of adaptation in an experimental plant population: strong
selection on QTLS for seed dormancy, Mol. Ecol., 19, 1335-1351, doi: 10.1111/j.1365294X.2010.04557.x, 2010.

577 Johnsen, Ø., and Skrøppa, T.: Adaptive properties of *Picea abies* progenies are influenced by 578 environmental signals during sexual reproduction, Euphytica, 92, 67-71, 1996.

Jose Crossa, Mateo Vargas, C. Mariano Cossani, Gregorio Alvarado, Juan Burgueño, Ky L
Mathews, and Reynolds, M. P.: Evaluation and interpretation of interactions, Agron. J., 105, 1-12,
2013.

Jump, A. S., and Penuelas, J.: Running to stand still: adaptation and the response of plants to rapid climate change, Ecol. Lett., 8, 1010-1020, doi: 10.1111/j.1461-0248.2005.00796.x, 2005.

Jump, A. S., Marchant, R., and Peñuelas, J.: Environmental change and the option value of genetic
 diversity, Trends Plant Sci., 14, 51-58, 2009.

- Lacey, E. P.: Parental effects in *Plantago lanceolata* L .I. A growth chamber experiment to examine pre- and postzygotic temperature effects, Evolution, 50, 865–878, 1996.
- Lande, R., and Shannon, S.: The role of genetic variation in adaptation and population persistence in a changing environment, Evolution, 50, 434–437, 1996.

Le, B. H., Cheng, C., Bui, A. Q., Wagmaister, J. A., Henry, K. F., Pelletier, J., Kwong, L.,
Belmonte, M., Kirkbride, R., Horvath, S., Drews, G. N., Fischer, R. L., Okamuro, J. K., Harada, J.
J., and Goldberg, R. B.: Global analysis of gene activity during *Arabidopsis* seed development and
identification of seed-specific transcription factors, Proc Natl Acad Sci U S A, 107, 8063-8070,
doi: 10.1073/pnas.1003530107, 2010.

Lenoir, J., Gegout, J. C., Marquet, P. A., de Ruffray, P., and Brisse, H.: A significant upward shift
in plant species optimum elevation during the 20th century, Science, 320, 1768-1771, doi:
10.1126/science.1156831, 2008.

Leung, L. R., Qian, Y., Bian, X. D., Washington, W. M., Han, J. G., and Roads, J. O.: Mid-century
ensemble regional climate change scenarios for the western United States, Clim. Change, 62, 75113, doi: 10.1023/B:Clim.0000013692.50640.55, 2004.

Liu, X. D., Zhang, H., Zhao, Y., Feng, Z. Y., Li, Q., Yang, H. Q., Luan, S., Li, J. M., and He, Z.
H.: Auxin controls seed dormancy through stimulation of abscisic acid signaling by inducing ARFmediated ABI3 activation in *Arabidopsis*, P Natl Acad Sci USA, 110, 15485-15490, doi:
10.1073/pnas.1304651110, 2013a.

Liu, Y., Kermode, A. R., and El-Kassaby, Y. A.: The role of moist-chilling and thermo-priming on the germination characteristics of white spruce (*Picea glauca*) seed, Seed Sci Technol, 41, 321-335, 2013b.

Liu, Y., and El-Kassaby, Y.: Timing of seed germination correlated with temperature-based
environmental conditions during seed development in conifers, Seed Sci Res, 25, 29-45, doi:
10.1017/S0960258514000361, 2015.

Liu, Y., Müller, K., El-Kassaby, Y. A., and Kermode, A. R.: Changes in hormone flux and signaling in white spruce (*Picea glauca*) seeds during the transition from dormancy to germination in response to temperature cues, BMC Plant Biol., 15, 292, doi: 10.1186/s12870-015-0638-7, 2015.

Manly, B. F. J.: Multivariate statistical methods: A primer. 3rd ed., Chapman and Hall/CRC, Boca
 Raton, Florida, 2005.

Morin, X., Lechowicz, M. J., Augspurger, C., O' Keefe, J., Viner, D., and Chuine, I.: Leaf phenology in 22 North American tree species during the 21st century, Global Change Biol., 15, 061 075 doi: 10.1111/j.1265.2486.2008.01725 r. 2000

618 961-975, doi: 10.1111/j.1365-2486.2008.01735.x, 2009.

Mote, P., Salathé, E., Dulière, V., and Jump, E.: Scenarios of future climate for the Pacific
Northwest, University of Washington, Seattle, WA, 2008.

- Müller, K., Bouyer, D., Schnittger, A., and Kermode, A. R.: Evolutionarily conserved histone
 methylation dynamics during seed life-cycle transitions, PloS One, 7, e51532, doi:
 10.1371/journal.pone.0051532, 2012.
- Murray, M. B., Cannell, M. G. R., and Smith, R. I.: Date of budburst of 15 tree species in Britain following climatic warming, J. Appl. Ecol., 26, 693-700, doi: 10.2307/2404093, 1989.
- 626 Nathan, R., Schurr, F. M., Spiegel, O., Steinitz, O., Trakhtenbrot, A., and Tsoar, A.: Mechanisms
- of long-distance seed dispersal, Trends Ecol. Evol., 23, 638-647, doi: 10.1016/j.tree.2008.08.003,
 2008.
- Owens, J. N., Simpson, S. J., and Molder, M.: Sexual reproduction of *Pinus Contorta*. I. Pollen
 development, the pollination mechanism, and early ovule development, Can J Bot, 59, 1828-1843,
 1981.
- Owens, J. N., Simpson, S. J., and Molder, M.: Sexual reproduction of *Pinus Contorta*. II.
 Postdormancy ovule, embryo, and seed development, Can J Bot, 60, 2071-2083, 1982.
- Parmesan, C., and Yohe, G.: A globally coherent fingerprint of climate change impacts across
 natural systems, Nature, 421, 37-42, 2003.
- Parmesan, C.: Ecological and evolutionary responses to recent climate change, Annu. Rev. Ecol.,
 Evol. Syst., 37, 637-669, doi: 10.1146/annurev.ecolsys.37.091305.110100, 2006.
- Penfield, S., Josse, E. M., Kannangara, R., Gilday, A. D., Halliday, K. J., and Graham, I. A.: Cold
 and light control seed germination through the bHLH transcription factor SPATULA, Curr. Biol.,
 15, 1998-2006, doi: 10.1016/j.cub.2005.11.010, 2005.
- Penfield, S.: Temperature perception and signal transduction in plants, New Phytol., 179, 615-628,
 doi: 10.1111/j.1469-8137.2008.02478.x, 2008.
- Penfield, S., and Springthorpe, V.: Understanding chilling responses in *Arabidopsis* seeds and their
 contribution to life history, Philos T R Soc B, 367, 291-297, doi: 10.1098/rstb.2011.0186, 2012.
- Pigliucci, M.: Phenotypic plasticity: Beyond nature and nurture, Johns Hopkins University Press,Maryland, 2001.

- 647 Plomion, C., Chagné, D., Pot, D., Kumar, S., Wilcox, P. L., Burdon, R. D., Prat, D., Peterson, D.
- 648 G., Paiva, J., Chaumeil, P., Vendramin, G. G., Sebastiani, F., Nelson, C. D., Echt, C. S., Savolainen,
- 649 O., Kubisiak, T. L., Cervera, M. T., de María, N., and Islam-Faridi, M. N.: Pines, Forest Trees, 7,
- 650 29-92, 2007.
- 651 Pourrat, Y., and Jacques, R.: The influence of photoperiodic conditions received by the mother 652 plant on morphological and physiological characteristics of *Chenopodium polyspermum* L. seeds, 653 Plant Sci. Lett. 4, 272, 270, 1075
- 653 Plant Sci. Lett., 4, 273–279, 1975.
- Raudenbush, S. W., and Bryk, A. S.: Hierarchical linear models: applications and data analysis
 methods (2nd edition), SAGE Publications, Inc., CA, 2001.
- Robeson, S. M.: Trends in time-varying percentiles of daily minimum and maximum temperature
 over North America, Geophys. Res. Lett., 31, 1-4, 2004.
- 658 Schmitt, J., Niles, J., and Wulff, R.: Norms of reaction of seed traits to maternal environments in 659 *Plantago lanceolata*, Am. Nat., 139, 451-466, 1992.
- Schwartz, M. D., Ahas, R., and Aasa, A.: Onset of spring starting earlier across the Northern
 Hemisphere, Global Change Biol., 12, 343-351, 2006.
- Shimizu, K. K., Kudoh, H., and Kobayashi, M. J.: Plant sexual reproduction during climate change:
 gene function in natura studied by ecological and evolutionary systems biology, Ann. Bot., 108,
 777-787, doi: 10.1093/Aob/Mcr180, 2011.
- 665 Simons, A. M.: Fluctuating natural selection accounts for the evolution of diversification bet 666 hedging, P Roy Soc B-Biol Sci, 276, 1987-1992, doi: 10.1098/rspb.2008.1920, 2009.
- Simons, A. M.: Playing smart vs. playing safe: the joint expression of phenotypic plasticity and
 potential bet hedging across and within thermal environments, J. Evol. Biol., 27, 1047-1056, doi:
 10.1111/jeb.12378, 2014.
- 670 Skrøppa, T., Tollefsrud, M. M., Sperisen, C., and Johnsen, Ø.: Rapid change in adaptive 671 performance from one generation to the next in *Picea abies* - Central European trees in a Nordic 672 environment, Tree Genet. Genom., 6, 93-99, doi: 10.1007/s11295-009-0231-z, 2010.
- 673 Slatkin, M.: Hedging ones evolutionary bets, Nature, 250, 704-705, doi: 10.1038/250704b0, 1974.

- Tabachnick, B. G., and Fidell, L. S.: Using Multivariate Statistics. 6th ed., Pearson Education,
 Boston, 2012.
- Thompson, J., Charpentier, A., Bouguet, G., Charmasson, F., Roset, S., Buatois, B., Vernet, P.,
 and Gouyon, P. H.: Evolution of a genetic polymorphism with climate change in a Mediterranean
 landscape, P Natl Acad Sci USA, 110, 2893-2897, doi: 10.1073/pnas.1215833110, 2013.
- Venable, D. L., and Brown, J. S.: The selective interactions of dispersal, dormancy, and seed size
 as adaptations for reducing risk in variable environments, Am. Nat., 131, 360-384, doi:
 10.1086/284795, 1988.
- Volis, S., Ormanbekova, D., Yermekbayev, K., Song, M., and Shulgina, I.: Introduction beyond a
 species range: a relationship between population origin, adaptive potential and plant performance,
 Heredity, 113, 268-276, doi: 10.1038/hdy.2014.25, 2014.
- Walck, J. L., Hidayati, S. N., Dixon, K. W., Thompson, K., and Poschlod, P.: Climate change and
 plant regeneration from seed, Global Change Biol., 17, 2145-2161, 2011.
- Walter, H., and Breckle, S. W.: Walter's vegetation of the earth: the ecological systems of the geo biosphere. 4th ed., Springer Berlin, 2002.
- Walther, G. R., Post, E., Convey, P., Menzel, A., Parmesan, C., Beebee, T. J. C., Fromentin, J. M.,
 Hoegh-Guldberg, O., and Bairlein, F.: Ecological responses to recent climate change, Nature, 416,
 389-395, 2002.
- Wang, T., Hamann, A., Yanchuk, A., O'Neill, G. A., and N., A. S.: Use of response functions in
 selecting lodgepole pine populations for future climates, Global Change Biol., 12, 2404–2416, doi:
 10.1111/j.1365-2486.2006.01271.x, 2006.
- Wang, T. L., Hamann, A., Spittlehouse, D. L., and Murdock, T. Q.: ClimateWNA Highresolution spatial climate data for western North America, J Appl Meteorol Clim, 51, 16-29, doi:
 10.1175/Jamc-D-11-043.1, 2012.
- Waring, R. H., and Franklin, J. F.: Evergreen coniferous forests of the pacific northwest, Science,
 204, 1380-1386, doi: 10.1126/science.204.4400.1380, 1979.
- West, B. T., Welch, K. B., and T., G. A.: Linear mixed models: a practical guide using statistical
 software, Chapman and Hall/CRC, New York, 2007.

Woodward, F. I., and Williams, B. G.: Climate and plant distribution at global and local scales,
Vegetatio, 69, 189-197, 1987.

Yakovlev, I., Fossdal, C. G., Skrøppa, T., Olsen, J. E., Jahren, A. H., and Johnsen, Ø.: An adaptive
epigenetic memory in conifers with important implications for seed production, Seed Sci Res, 22,
63-76, 2012.

Yamauchi, Y., Ogawa, M., Kuwahara, A., Hanada, A., Kamiya, Y., and Yamaguchi, S.: Activation
of Gibberellin biosynthesis and response pathways by low temperature during imbibition of *Arabidopsis thaliana* seeds, Plant Cell, 16, 367-378, doi: 10.1105/Tpc.018143, 2004.

- 710 Yan, W., and Tinker, N. A.: Biplot analysis of multi-environment trial data: Principles and applications, Can J Plant Sci, 86, 623-645, 2006.
- 712 Yan, W. K.: GGEbiplot A windows application for graphical analysis of multienvironment trial
- data and other types of two-way data, Agron. J., 93, 1111-1118, 2001.

714 FIGURE LEGENDS

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

- 717 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.
- 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
- 726 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.

- 737 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; ⁽²⁾ the reproduction cycle of (lodgepole) pine undergoes
- 742 three years, that is, seed/pollen cone initiation (A S O) \rightarrow dormant (N D J F M) \rightarrow pollen
- 743 development (A M) \rightarrow pollination (J) \rightarrow female gametophyte development (J A) \rightarrow dormant (S O
- 744 N D J F M A)→ female gametophyte development (M J). (Initial letters in the bracket represent
- 745 months.)

Table 1 Canonical correlation analysis between individual variables and their own and opposite set 746

747 of variables

748 A)

Varia	ables	Life-history traits	Climatic Ecology [†]	Geographic Ecology ^{††}
Life-history	Seed weight	0.7429	0.9021	0.5481
traits	DI	0.5909	0.6582	0.5327

Note: 749

[†]Based on the PLS analysis, 27 most correlated climatic variables for seed dormancy (*DI*) and weight were 750

used as Climatic Ecology (three climatic variables were strongly correlated with both seed dormancy and 751

752 weight).

^{††}Geographic Ecology consists of latitude, longitude, and elevation. 753

754 B)

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

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756Table 2 Multi- and uni-variate analyses for life-history traits. (A) MANOVA table reporting the757correlation of seed dormancy (DI) and weight with ecosystems; (B) ANOVA table reporting758significant effect of the genotype on phenotypic variance and associating broad-sense heritability (H^2).

759 A)

				Paramet	ers		
SOV		DI			Seed weight		
	DF	Mean squares	<i>F</i> -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		0.55	0.0829		0.30

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

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

762 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
$\mathbf{G} \times \mathbf{E}$	82	$\sigma_{\epsilon}^2 + 4\sigma_{G \times E}^2$	129.8935	25.09**	39.43%	0.54
Error (ϵ)	498	σ_{ϵ}^{2}	5.1767		6.55%	

⁷⁶³ [¶]no variance components or percent of total variation were estimated for the fixed effect (E).

764 ** P < 0.05.

765 **Table 3 Parameter estimates and statistical tests for the climatic variables-based hierarchical models**

766 regarding seed dormancy and weight

A) Seed dormancy model

Effect	Estimate	Standard Error	DF	Statistic	Р
Intercept [¶] (β_0)	2.4819	7.9418	21	<i>t</i> = 0.31	0.7577
<i>Eref07</i> (β_1)	2.7076	0.8932	58	<i>F</i> = 9.19	0.0036
<i>Eref_summer</i> (β_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

768 Tesidual from the ecosystem zone (ε_j) is integrated into 'intercept' (Table S3).

769 B) Seed weight model

Effect	Estimate	Standard Error	DF	Statistic	Р
Intercept [¶] (β_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
$PPT10 (\beta_2)$	-0.0014	0.0007	58	F = 4.68	0.0346
Tmax07 (β_3)	0.0721	0.0150	58	F = 23.11	<0.0001

770 residual from the ecosystem zone (ε_j) is integrated into 'intercept' (Table S3).





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