1	Article Type: Original Article
2	Running Title: Chronosequence of invasion
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4	Chronosequence of invasion reveals minimal losses of population genomic diversity,
5	niche expansion, and trait divergence in the polyploid, leafy spurge
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24	Acknowledgments
25	Funding was provided by the Minnesota Invasive Terrestrial Plants and Pests Center through
26	the Environment and Natural Resources Trust Fund as recommended by the Legislative-Citizen
27	Commission on Minnesota Resources. The Minnesota Supercomputing Institute provided
28	computing and data storage resources. We thank Yaniv Brandvain and Ken Kozak for
29	discussions and advice on analyses. We also thank Danielle Schoenecker, Ben Greene, and
30	Isaac Olson for their assistance with data collection.
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35 ABSTRACT

36 Rapid evolution may play an important role in the range expansion of invasive species 37 and modify forecasts of invasion, which are the backbone of land management strategies. 38 However, losses of genetic variation associated with colonization bottlenecks may constrain trait 39 and niche divergence at leading range edges. The spatial and temporal scales over which 40 adaptation contributes to invasion dynamics remains unresolved. We leveraged detailed records of the ~130 year invasion history of the invasive polyploid plant, leafy spurge (Euphorbia 41 42 virgata), across ~500km in Minnesota, U.S.A. We examined the consequences of range 43 expansion for population genomic diversity, niche breadth, and the evolution of germination behavior. Using genotyping-by-sequencing, we found some population structure in the range 44 45 core, where introduction occurred, but panmixia among all other populations. Range expansion 46 was accompanied by only modest losses in sequence diversity, with small, isolated populations 47 at the leading edge harboring similar levels of diversity to those in the range core. The climatic niche expanded during most of range expansion, and the niche of the range core was largely 48 49 non-overlapping with the invasion front. Ecological niche models indicated that mean 50 temperature of the warmest quarter was the strongest determinant of habitat suitability and that 51 populations at the leading edge had the lowest habitat suitability. Guided by these findings, we 52 tested for rapid evolution in germination behavior over the time course of range expansion using 53 a common garden experiment and temperature manipulations. Germination behavior diverged 54 from early to late phases of the invasion, with populations from later phases having higher 55 dormancy at lower temperatures. Our results suggest that trait evolution has contributed to 56 niche expansion during invasion and that distribution models of future invasion may 57 underestimate invasion potential without accounting for evolution.

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Key Words (6): adaptation, colonization bottlenecks, plant invasion, population genetic
structure, range shift, rapid evolution

62 Introduction

63 Invasive species experience considerable changes to genetic variation during the process of introduction and subsequent invasion (Lee 2002; Suarez and Tsutsui 2008). In 64 65 particular, it has been well-documented that founder effects and genetic drift can cause 66 substantial losses of genetic diversity during the colonization process (Dlugosch and Parker 67 2008; Uller and Leimu 2011). Following initial establishment, further losses of variation may occur during range expansion. However, the magnitude of changes in genetic variation depends 68 69 upon the number of introductions and the severity of population bottlenecks (Nei et al. 1975; 70 Uller and Leimu 2011; Welles and Dlugosch 2019). Such changes in genetic variation early in 71 the invasion process may influence the capacity for adaptation, forecasts of range expansion, 72 and subsequent management decisions.

73 Following colonization, some non-native species exhibit rapid population growth and 74 dispersal into new environments (Sakai et al. 2001). The process of invasion is often highly 75 variable, involving repeated founder events and density-dependent population growth, 76 especially Allee effects (Melbourne and Hastings 2009; Sullivan et al. 2017). In addition to 77 affecting the speed of invasion, these population fluctuations can influence levels of genetic 78 diversity and structure across an invaded range (Austerlitz et al. 1997; Excoffier 2004). For 79 example, range expansion is expected to cause a reduction in allelic richness and 80 heterozygosity with increasing distance from the origin of expansion (Slatkin and Excoffier 2012; 81 Peter and Slatkin 2013). The prevalence of drift during invasion may also cause populations to 82 depart from migration-drift equilibrium, resulting in a lack of isolation-by-distance (Wright 1943; Slatkin 1987, 1993; Hutchison and Templeton 1999). Last, mutations arising at the range edge 83 may rise in frequency due to genetic drift, "surf" along the expanding front, and travel long 84 85 distances (Klopfstein et al. 2006; Excoffier and Ray 2008). Models of allele surfing indicate that 86 rapid range expansion can produce clinal variation in allele frequencies and increase the 87 frequency of loci with private alleles (Klopfstein et al. 2006; Excoffier and Ray 2008; Goodsman

et al. 2014). Such clines can emerge for any type of mutation (beneficial, neutral, deleterious),
and therefore could reflect drift and/or selection (Lehe et al. 2012; Peischl et al. 2013; Koski et
al. 2019). Overall, the prevalence of drift during range expansion has the potential to influence
the capacity for adaptation as organisms encounter novel environments, particularly when
functionally-important allelic variation is lost.

93 An increasing body of evidence suggests that rapid phenotypic evolution can be 94 important to the process of range expansion (Colautti and Barrett 2013; Hodgins et al. 2018; 95 Selechnik et al. 2019; Ma et al. 2020). Forecasts of range expansion in invasive species can 96 underpredict the potential extent of invasion if they assume a species does not evolve or adapt 97 over short time scales (Chardon et al. 2020; Collart et al. 2020). A recent meta-analysis indicated that the signature of local adaptation in invasive species was at least as strong as in 98 99 native species, even when accounting for variation in life history (Oduor et al. 2016). Invasive 100 species frequently expand across strong environmental gradients and into novel niche space 101 (Atwater et al. 2018; Bates and Bertelsmeier 2021). Such niche expansion may require adaptive 102 evolution at the invasion front (Chown et al. 2015; Moran et al. 2017; Hodgins et al. 2018). For 103 example, purple loosestrife rapidly diverged in flowering time and plant size during invasion 104 across latitudinal gradients in growing season length (Colautti and Barrett 2013). Despite 105 evidence of rapid evolution in some systems, range expansion may not involve any changes in 106 the organism's niche if the only limit to spatial expansion is dispersal and time. As such, the 107 apparent expansion of the climate niche with invasion may not actually involve the evolution of 108 ecologically-important traits. While tests of local adaptation within an invaded range remain few, 109 there is growing appreciation that rapid evolution is likely to shape the trajectory of range 110 expansion in non-native species (Hodgins et al. 2018; Woods and Sultan 2022).

Phenotypic evolution during range expansion may be caused by spatially-variable
selection and/or neutral processes (e.g., spatial sorting) (Keller and Taylor 2008; Colautti and
Lau 2015; Hodgins et al. 2018). Adaptive evolution, in particular, may be paramount to the

114 invasion process if selection in response to novel environments results in trait changes that 115 enhance a species' capacity to establish in new habitats (Prentis et al. 2008; Williams et al. 116 2016; Hodgins et al. 2018; Woods and Sultan 2022). While reciprocal transplant experiments 117 are the gold standard for testing for adaptation, the translocation of invasive species for these 118 experiments is subject to ethical concerns and legal restrictions in many areas. Alternatively, 119 researchers have started to use ecological niche models (ENMs) to identify important 120 environmental gradients that span from optimal to marginal habitat, such as from a range core 121 to edge. Predictions are then made about traits that may promote adaptation to the novel 122 environments found in marginal habitats (Searcy and Shaffer 2016; Dixon and Busch 2017; Capblancg et al. 2020; Morente-López et al. 2022). Finally, common garden experiments can 123 124 determine whether the putative traits under selection have differentiated across the key 125 environmental gradients identified by ENMs. Taken together, this series of approaches can 126 provide insight into the role of adaptation in the process of niche expansion at leading range 127 edges.

Among invasive plant species, polyploidy is prevalent (Pandit et al. 2011) and can 128 129 influence the process of range expansion (Van de Peer et al. 2021). The frequency of polyploid 130 species increases with higher latitudes, lower temperatures, and seasonally drier environments 131 (Brochmann et al. 2004; Rice et al. 2019). Direct effects of polyploidy on physiological, 132 morphological, and phenological traits may facilitate niche shifts (Glennon et al. 2014; Blaine 133 Marchant et al. 2016; Brittingham et al. 2018; Wang et al. 2022) and preadapt polyploids to new 134 environments (Treier et al. 2009; Lachmuth et al. 2010). Polyploidy may also influence the 135 capacity for adaptation during range expansion as new environments are encountered (te Beest 136 et al. 2012; Baniaga et al. 2020). Although genetic drift during range expansion can cause 137 losses of genetic diversity, drift may have less severe effects (e.g., inbreeding depression) in 138 polyploids relative to diploids when there is polysomic inheritance (Moody et al. 1993; Soltis and 139 Soltis 2000). Despite numerous polyploid invaders, they have been the subject of few studies

because of substantial challenges with the application of evolutionary genetic analyses thatwere developed for diploids (Rutland et al. 2021).

142 In this study, we used a well-documented chronosequence of invasion to examine the 143 consequences of range expansion for population genomic diversity, climatic niche breadth, and 144 the evolution of germination behavior in the polyploid, leafy spurge (Euphorbia virgata). We 145 focused on one area of introduction to southwestern Minnesota, U.S.A and subsequent range 146 expansion to the north and east. We were interested in examining the severity of losses of 147 genetic diversity following introduction and its potential consequences for invasion, particularly 148 since existing species distribution models predict a low probability of range expansion at the 149 current leading range edge (Lake et al. 2020). Introduction to this region occurred in the 1890s 150 in southwestern Minnesota with subsequent range expansion to northeastern Minnesota (ca., 151 500 km), where populations are currently rare, isolated, and small. Based on historical 152 occurrence datasets, we defined a range core, area of early expansion, area of late expansion, 153 and invasion front.

154 First, we sampled populations along the gradient from core to invasion front using two 155 sampling schemes to quantify population genomic diversity and structure (using reduced 156 representation sequencing). Samples of multiple individuals from 14 populations (population 157 samples) allowed us to quantify changes in sequence diversity among populations over the 158 course of range expansion. Samples of single individuals from 157 populations (landscape 159 samples) allowed us to test for fine-scale population structure (e.g., isolation-by-distance) over 160 the time series of range expansion. Second, we tested whether range expansion involved niche 161 expansion - i.e., occurred into novel climatic environments. Third, we developed an ecological 162 niche model (ENM) to test whether habitat suitability declines from range core to invasion front 163 and to determine which environmental gradients are most strongly associated with high versus 164 low habitat suitability. Warm season temperature had the greatest positive contribution to habitat suitability. Because temperature is known to modulate germination behavior in leafy 165

166 spurge, and because establishment in new habitats is dependent upon successful germination 167 timing, we focused on this trait for common garden experiments. Past work has also suggested 168 that shifts in seed dormancy might facilitate invasion at leading range edges (Mathias and Kisdi 169 2002; Travis et al. 2021). We examined the responses of seeds from early versus late in 170 invasion to five temperature regimes in a common garden experiment. We specifically tested 171 whether there was an interaction between geographic region (early vs. late invasion stages) and 172 temperature regime, which would indicate divergence in germination behavior over the course 173 of invasion. 174

175 Methods

176 Biology and invasion of leafy spurge

177 Species biology

Leafy spurge has invaded nearly two million acres across the northern tier of the United States and southern Canada (Duncan et al. 2004). While it is most commonly found in dry, open sites with well-drained soils (e.g., prairies), it can occasionally occur in seasonally wet meadows and riparian areas (Selleck et al. 1962). Leafy spurge impacts rangelands and natural habitats by competitively displacing native species (Hein and Miller 1992). When damaged, plants exude a toxic white latex that deters grazing (Lym and Kirby 1987; Lym 1998).

Leafy spurge spreads locally via rhizomes and ballistic seed dispersal (Morrow 1979). Longer distance dispersal has been proposed to occur via animals or agricultural machinery (Pemberton 1988; Lacey et al. 1992). Seeds germinate in spring or may remain dormant in soil for at least two years (Hanson and Rudd 1933; Selleck et al. 1962). Flowers are insect pollinated and the mating system is primarily outcrossing (Selleck et al. 1962). Leafy spurge is an auto-allohexaploid that likely originated from hybridization between closely-related *Euphorbia* species, although the progenitor species are not yet known (Schulz-

191 Schaeffer and Gerhardt 1989; Riina et al. 2013). It has been the subject of several genomic

investigations but lacks a full genome assembly and annotation (Chao et al. 2005; Horvath et al.
2015, 2018; West et al. 2023).

194

195 Invasion history

196 One introduction of leafy spurge occurred into southwestern Minnesota ca., 1890 197 purportedly via contaminated grains imported from southern Russia (Batho 1932; Hanson and 198 Rudd 1933; Dunn 1985). Following introduction, the range expanded to eastern South Dakota 199 by ca., 1902 (Bakke 1936), eastern North Dakota by ca. 1909 (Hanson and Rudd 1933), and 200 southern Manitoba and Saskatchewan by the 1920s (Batho 1932; Selleck et al., 1962). Hanson and Rudd (1933) documented in detail the distribution of leafy spurge across Minnesota and 201 202 neighboring regions, providing a baseline for understanding the timeline of subsequent range 203 expansion. By the late 1970s, leafy spurge had become common throughout grasslands of the 204 north-central plains (Dunn 1979, 1985). Invasion of the boreal forest region of northeastern 205 Minnesota began in the 1940s and 1950s with isolated occurrences (Lakela 1965) and 206 populations were not common until the 1990s. This invasion front has persisted with limited 207 expansion further northeast.

208

209 **Delineating the timeline of range expansion**

We digitized the earliest known point record map (Hanson and Rudd 1933) using ArcGIS Pro (ESRI, 2022). We then applied empirical Bayesian kriging to produce a continuous density surface that represented the density of populations in the north-central plains. From this density surface, we applied an equal-interval threshold to demarcate a range core, area of early expansion, area of late expansion, and invasion front that corresponded to four density categories across Minnesota and surrounding states (Figure 1). We verified these demarcations with published accounts of the invasion history (described above).

218 **Population genetic diversity and structure**

219 Sampling and sequencing

220 In 2019, we collected leaf tissue from six individuals in each of 14 populations distributed 221 evenly across Minnesota (hereafter: population samples; Figure 1A). In addition, we collected 222 tissue from one individual in each of 157 populations distributed relatively evenly across 223 Minnesota, eastern South Dakota, eastern North Dakota, and western Wisconsin (hereafter: 224 landscape samples; Figure 1B). We sampled tissue from individuals that were at least five 225 meters apart to minimize collecting from the same genet and placed tissues immediately in 226 silica for preservation until DNA extraction (Table S1). We extracted DNA using QIAGEN DNeasy Plant Mini Kits (QIAGEN Inc.). Dual-indexed 227 228 GBS (genotyping-by-sequencing) libraries were created using the BamHI + Nsil enzyme 229 combination. All libraries were pooled and sequenced on an Illumina NovaSeq System (Illumina 230 Inc., San Diego, CA, USA) with 1x100-bp sequencing. Once sequenced, the reads were 231 demultiplexed and balanced with a mean quality score ≥ Q30 for all libraries. We filtered low-232 guality bases using Trimmomatic (Bolger et al. 2014) and used Stacks v.2.5.9 (Rochette et al. 233 2019) to build loci de novo (i.e., without aligning reads to a reference genome). Overall, we 234 obtained 510 million reads across the 241 samples (599,386 - 3,376,078 of raw reads per 235 individual). Mean read depth per locus ranged from 14x to 26x (Supplementary Methods). 236 We called SNPs using polyRAD v.1.6 (Clark et al. 2019), a Bayesian algorithm designed for polyploid GBS data. PolyRAD estimates the genotype probabilities for each individual from 237 238 read depth distributions with ploidy level as a prior. First, we filtered our dataset using the 239 H_{ind}/H_e statistic to cull likely paralogous loci (Clark et al. 2022). Next, we estimated posterior 240 probabilities for each genotype using the 'IterateHWE' function with Hardy-Weinberg equilibrium 241 as the prior (Gerard and Ferrão 2020; Clark et al. 2022). For each individual at each locus, we 242 exported the most probable genotype for subsequent analyses (Supplementary Methods).

243 We implemented a second filtering step to account for potential biases caused by homoeologous loci present in our dataset. Because leafy spurge is an auto-allohexaploid 244 245 (Schulz-Schaeffer and Gerhardt 1989), we expect homoeologous loci to have a 2:1 allelic ratio 246 (e.g. AAAABB genotype; Horvath et al. 2018). Indeed, we identified a peak in the minor allele 247 frequency spectrum around 0.33 (Figure S1). We removed loci with a minor allele frequency above 0.26 from our dataset because they are likely to have an excess of homoeologous 248 249 genotypes (Figure S1). While essential, this second filtering step limits our capacity to 250 understand absolute levels of sequence diversity. However, our primary goal was to examine 251 changes in sequence diversity over the course of invasion rather than absolute quantities. After filtering, 3,176 loci remained for downstream analyses. 252

253

254 **Population structure**

255 We performed an analysis of molecular variance (AMOVA) to quantify the proportion of 256 genetic variation partitioned among populations, among individuals within populations, and 257 within individuals (Excoffier et al. 1992; Meirmans 2012, 2020). We estimated genetic variance 258 components using the rho statistic (Ronfort et al. 1998) and determined significance using 259 permutation tests (n= 999) using the R package poppr v 2.8.6 (Kamvar et al. 2015). Further, we 260 checked for clonality among samples using the 'clonecorrect' function in (Kamvar et al. 2015). 261 We assessed population structure using the Bayesian clustering algorithm STRUCTURE v.2.3.4 (Pritchard et al. 2000). We ran the analysis for 500,000 Markov Chain Monte Carlo 262 263 iterations with a 50,000-run burn-in period, specifying an admixture model with the assumption 264 of uncorrelated allele frequencies. We used 'structure threader' in Python 3 (Pina-Martins et al. 265 2017) to parallelize runs across clusters (K = 1 - 10). We determined the most plausible number 266 of clusters using the Evanno Delta K method (Evanno et al. 2005) and STRUCTURE 267 HARVESTER web v.0.6.94 (Earl and vonHoldt 2012).

We performed principal component analyses (PCA) to examine population structure using GENODIVE v.3.0.4 (Meirmans 2020). We performed PCA separately for the population samples and the landscape samples.

271 We tested for isolation by distance (IBD) (Wright 1943) in the population samples by 272 estimating genetic differentiation as G_{ST} (Dufresne et al. 2014) using GENODIVE v.3.0.4 273 (Meirmans 2020). We also tested for IBD in the landscape samples by calculating Nei's genetic 274 distance (D) (Nei 1972) (Meirmans 2020). We subset the landscape samples according to 275 successive stages of range expansion and tested for IBD within each subset (i.e., within range 276 core, then successively including areas of early expansion, late expansion, and invasion front). We tested for a relationship between genetic and geographic distance matrices using Mantel 277 278 tests with 9,999 permutations in the R package 'adegenet' v.2.1.8 (Jombart 2008).

279

Tests for changes in population genetic diversity during range expansion

281 For each of the 14 population samples, we estimated observed heterozygosity (H_0 ; 282 gametic heterozygosity: which corrects for potential overestimates of heterozygosity in 283 polyploids by calculating the fraction of heterozygotic gametes for each genotype) (Moody et al. 284 1993), the inbreeding coefficient (G_{IS}) (Meirmans et al. 2018), and the number of private alleles (P) (Kamvar et al., 2014). We also estimated Tajima's D (Tajima 1989) using DNASp v 6.0 285 286 (Rozas et al. 2017) to gauge if populations have an excess or deficit of rare alleles, which can 287 be indicative of population expansion following a bottleneck (negative D) or sudden population 288 contraction (positive D), respectively. Because we filtered loci with higher minor allele 289 frequencies, Tajima's D should be biased to lower values. 290 We tested for changes in population genetic parameters (H_o , G_{IS} , P, and Tajima's D) as

populations dispersed beyond the range core (Table 1). For each statistic separately, we used a
 multiple linear regression that included latitude, longitude, and their interaction as independent
 variables with the R package 'car' (Weisberg, 2019). As all late expansion and invasion front

294	populations are located either north or east of the range core, latitude and longitude describe
295	the northern and eastern invasion spread, respectively. In the model of private alleles, we
296	identified the 'Winona' population as an outlier using diagnostic plots of residuals (Figure S2), so
297	we removed it from the analysis.

298

299 Niche breadth and habitat suitability

300 Environmental data

301 We downloaded three bioclimatic variables at a 30 arcsecond resolution (~1 km) from 302 Worldclim (http://worldclim.org/version2): minimum temperature of the coldest month (Bio 6), mean temperature of the warmest guarter (Bio 10), and precipitation of the warmest guarter (Bio 303 304 16). These three variables provide biologically meaningful axes of climate variation that are 305 relevant to key life history stages (Figure S3) (Petitpierre et al. 2017; Chapman et al. 2017). 306 Specifically, minimum cold temperatures are important for plant physiological responses to 307 overwintering and cold tolerance (Chapman et al. 2017), whereas the temperature of the warmest season acts on seed germination, plant growth, and phenological transitions 308 309 (Wolkovich et al. 2013). Precipitation in the warmest season also affects plant growth and 310 reproduction, with lower precipitation associated with reduced growth and increased drought 311 stress in northern temperate ecosystems (Petitpierre et al. 2017; Gorton et al. 2019).

312

313 Tests for niche differentiation during range expansion

We tested for climatic niche differentiation between the range core, early expansion, late expansion, and invasion front using the 'ecospat' R package (Di Cola et al. 2017). We sampled the total extent of the background environmental space with a principal components analysis using 1500 random points drawn from a bounding box centered on Minnesota (Latitude: min: 43°N, max: 50°N; Longitude: min: -98°W, max: -89°W). We then used the landscape sample

localities to calculate the niche boundaries and density of occurrence for each portion of therange within the environmental PCA space.

For all pairwise comparisons of the range core, early expansion, late expansion, and 321 322 invasion front, we quantified four measures of niche differentiation. First, we used Schoener's D 323 to quantify the similarity in niche by incorporating both niche breadth and density (Warren et al. 324 2008). Values of D can vary from 0 (no overlap) to 1 (complete overlap). For each pair of 325 regions, we then calculated what proportion of the combined niche space represented niche 326 stability, expansion, and unfilling. In this framework, if 'A' represents the older potion of the 327 range (e.g. range core) and 'B' represents the more recent portion of the range (e.g. invasion front), niche stability is the proportion of niche B that overlaps A, niche expansion is the 328 329 proportion of niche B that does not overlap A, and niche unfilling is the proportion niche A that 330 does not overlap B.

331 We used permutation tests to determine if values of Schoener's D, expansion, stability, 332 and unfilling were equivalent between the range core, early expansion, late expansion, and 333 invasion front. In the niche equivalency tests, the data were pooled and then randomly assigned 334 to one group of the pairwise range comparisons for 999 permutations. For each permutation, we 335 computed all four statistics. We rejected the null hypothesis of niche equivalency if the observed Schoener's D value was less than 95% of permuted D values. Similarly, we rejected niche 336 337 equivalency based on the combined niche space if observed stability was less than 95% of 338 permuted values and observed niche expansion or unfilling was greater than 95% of permuted 339 values.

340

341 Ecological niche model

We developed an ecological niche model (ENM) using MaxEnt v.3.4.3 (Merow et al. 2013; Phillips et al. 2017) with the 'dismo' package in R (Hijmans et al. 2022). Our goal was identify environmental gradients that could potentially drive phenotypic evolution during range

345 expansion (Elith and Leathwick 2009; Araújo et al. 2019; Morente-López et al. 2022). We built 346 ENMs with the same bioclimatic variables and in the same bounding box as the analyses of 347 niche differentiation. We used occurrence records from our tissue collection sites and 10,000 348 background points. We excluded threshold and hinge features during the model building 349 process as the preliminary models that included these features tended to be overspecified. We 350 used five-fold cross-validation to assess model performance: data were randomly partitioned 351 into five equal groupings and 80% of data were used for training and 20% were used for 352 evaluation. Model predictions are a mean of the five cross-validated models.

353 We evaluated models using AUC and sensitivity, which were calculated using the withheld dataset. AUC characterizes model discrimination ability and ranges between 0 and 1. 354 355 with higher values indicating greater model performance and a value of 0.5 indicating that model 356 discrimination is no better than random (Phillips and Dudík 2008). Sensitivity quantifies the 357 proportion of correctly identified positives. We calculated sensitivity where the sum of the true 358 positive rate and true negative rate was maximized (threshold = 0.53). We then used the 359 variable permutation importance and percent contribution to identify which environmental 360 variable had the greatest contribution to habitat suitability. We also visualized response curves 361 of each environmental variable to ensure models were not overspecified.

362 To test for divergence in germination behavior during range expansion, we focused on warm season temperature since it most strongly affected predicted habitat suitability in the 363 364 ENM. Populations from early in the invasion had high predicted habitat suitability and higher 365 warm season temperatures (all above 20°C) compared to those from later in the invasion, which had lower predicted habitat suitability and lower warm season temperatures (all below 20°C; 366 Fig. 6A,B). The other environmental variables in our ENM did not differ consistently between 367 368 early versus late in invasion (Figure S5). Guided by these results, we divided the 14 populations 369 for which we had seed collections into two groups: early (n=8) versus late (n=6) in invasion and 370 exposed them to five temperature regimes (see below).

371

372 Tests for differentiation in germination behavior during range expansion

373 We collected seeds from 14 populations (8 - 24 maternal families/pop) in 2019 (Table 374 S6). Seeds were collected from individuals that were at least 5 meters apart. We stored and 375 after-ripened seeds in an indoor environment for two years prior to the germination experiment (Wicks and Derscheid 1958). Seeds were pooled within populations prior to applying 376 377 treatments. We were unable to collect seeds from every population used in the population 378 genetic survey because some had already dispersed seeds prior to our collection effort. 379 We examined the effects of temperature regime and source geographic region (early vs. 380 late in invasion) on germination. Five temperature regimes were designed to mimic the full 381 range of variation in daytime and nighttime temperatures during the spring and summer in this 382 region (14 hour day /10 hour night periods: 15/5 °C, 20/10 °C, 25/15 °C, 30/20 °C, and 35/25 383 °C). The experiment was conducted in five successive rounds in two growth chambers 384 (Conviron Inc.). We conducted each temperature regime twice – i.e., once in each growth 385 chamber - to control for growth chamber effects. 386 For each round, we placed ten seeds per population in 60 x 15mm polystyrene Petri 387 dishes containing 2 mL sterile distilled water, lined with one Whatman #1 filter paper, and sealed with parafilm. Each population was replicated three times per chamber for a total of 42 388 389 dishes per treatment per chamber or 84 dishes per round. Dishes were wrapped in aluminum foil to block light, which can inhibit germination (Selleck et al., 1962). Every 24 hours we 390 391 recorded the number of germinated seeds (emergence of radicle) per dish. After each treatment 392 period ended, we tested whether ungerminated seeds were viable by soaking bisected seeds in 393 a 1% Tetrazolium solution for 24 hours (Verma and Majee 2013). Red staining of tissues 394 indicates that seeds are viable. For analyses of germination, we included the number of

395 germinated seeds out of the total number of germinated plus viable (but ungerminated) seeds.

We tested for the effects of temperature regime (categorical), source geographic region, and their interaction on germination probability using a mixed-effects model with a binomial family. Experimental round and population were included as random effects. We evaluated significance with Type III tests. All models were run using the 'mixed' function in the 'afex' package (Singmann et al. 2016) in R v.4.0.2 (R Development Core Team, 2015). We used linear contrasts to test for differences in germination between geographic regions for each temperature regime category individually.

- 403
- 404 **Results**

405 **Population genetic consequences of range expansion**

406 Analysis of molecular variance (AMOVA) revealed significant partitioning of genetic 407 variance among populations (13.7%; P < 0.001), among individuals within populations (8.4%; P < 0.001), and within individuals (77.8%; P < 0.001) (Table S2). No clonal genotypes were 408 409 detected in the dataset. STRUCTURE indicated that the optimal number of clusters was three (K = 3) (Table S3). All individuals were assigned primarily to one cluster regardless of where the 410 411 population was found in the invasive range (i.e., core, early expansion, late expansion, invasion 412 front). There was some evidence of population structure in the other two clusters; however, 413 there was no clear geographic pattern (Figure 2).

414 The PCA did not indicate substantial population structure across the invaded range based on either population or landscape samples. For the population samples, there was some 415 416 evidence of differentiation among three populations in or near the range core (Figure 3: PC1 417 and PC2 explained: 6.8% and 5.9% of variance, respectively). The PCA of landscape samples did not reveal a relationship between genetic similarity and geography over the time course of 418 419 range expansion (Figure 3; PC1 and PC2 explained 1.7% and 1.3% of variance, respectively). There was no evidence of isolation by distance (IBD) for either population ($R^2 = 0.02$; P 420 = 0.418; Table 2: Table S4) or landscape samples (R^2 = 0.05; P = 0.676; Table S4). We also did 421

not detect IBD when landscape samples were subset according to the four invasion phases thatwe defined (Figure 4).

Range expansion from the core area of invasion was accompanied by only modest changes in genetic diversity. Heterozygosity declined modestly from the core to the invasion front, as indicated by a significant interaction of latitude and longitude (P = 0.013; Table 3). However, the number of private alleles, Tajima's *D*, and the inbreeding coefficient did not change over the course of range expansion (Table 3).

429

430 Niche differentiation during range expansion

The climatic niche expanded during invasion. Relative to the range core, the early 431 expansion niche represented a sizable increase in niche breadth (Figure 5A; Table S5; overlap 432 433 = 0.43; P < 0.01; expansion = 0.37; P < 0.01; stability = 0.63; P < 0.01). When comparing the 434 early expansion and late expansion niches, there was similar evidence for a niche shift (Figure 435 5B; Table S5; overlap = 0.36, P < 0.01; expansion = 0.39, P < 0.01; stability = 0.61, P < 0.01). 436 Between the late expansion niche and the invasion front, the null hypothesis of niche 437 equivalency was not rejected (Table S5); the invasion front niche was contained within the late 438 expansion niche (Figure 5C). When comparing range core to invasion front, there was near-zero niche overlap (Figure 5D; Table S5; overlap = 0.03, P < 0.01; stability = 0.06, P < 0.01) and high 439 440 expansion (Table S5; expansion = 0.94, P < 0.01). Niche differences between the core and invasion front were most apparent along environmental axes related to temperature of the 441 442 warmest guarter and minimum temperature of the coldest month, rather than precipitation (Figure S4). 443

444

445 Ecological niche model

The model AUC (0.79) and sensitivity (0.75) metrics indicated moderately high
 discrimination and accuracy (Figure 6A). The variable response curves indicated the ENM was

448	not overspecified (Figure S5). The mean temperature of the warmest quarter constituted the
449	most important variable for predicted habitat suitability (percent contribution = 81.5%;
450	permutation importance = 67.8%) and warmer temperatures were largely associated with an
451	increase in habitat suitability ($R = 0.67$; Figure S6). The minimum temperature of the coldest
452	month had the second highest importance (percent contribution = 11.2%; permutation
453	importance = 17.9%) and was modestly associated with increased habitat suitability ($R = 0.31$;
454	Figure S6). Precipitation of the warmest quarter had the lowest importance (percent contribution
455	= 7.3%; permutation importance = 14.3%) and was weakly correlated with habitat suitability ($R =$
456	0.02; Figure S6).
457	Populations from early versus late in invasion were distinct along the mean temperature
458	of the warmest quarter axis with a disjunction at 20°C. This temperature also distinguished
459	highly suitable from less suitable habitat in the ENM (Figure 6A & 6B).
460	
461	Evolution of germination behavior during range expansion
462	Populations from early versus late in invasion responded differently to temperature
463	regimes (source geographic region x temperature regime interaction: $P = 0.003$; Table S7;
464	Figure 6C). Although linear contrasts between geographic regions did not differ significantly for
465	any particular temperature treatment regime, there was a trend toward populations from late in
466	invasion having a lower germination probability in the lower temperature treatments (Table S8).
467	Germination probability increased with temperature for both geographic regions and plateaued
468	(<i>P</i> < 0.001; Table S7; Figure 6C).
469	

470 **Discussion**

Rapid evolution is increasingly recognized as an important process contributing to the
range expansion of invasive species (Prentis et al. 2008; van Boheemen et al. 2019; Clements
and Jones 2021). However, our understanding of the temporal and spatial scale over which

474 niche and trait divergence contribute to invasion at leading range edges remains unresolved. 475 Our study took advantage of a well-documented invasion history to synthesize the 476 consequences of recent range expansion for population genomic diversity, niche breadth, and 477 germination behavior, a trait important in the colonization of new habitats. We found that leafy 478 spurge populations experienced only modest losses in sequence diversity over the 479 chronosequence of invasion. Range expansion involved climatic niche expansion and ecological 480 niche models suggested that warm season temperature had the strongest influence on habitat 481 suitability. Populations differentiated in germination behavior in response to temperature, with 482 leading edge populations having increased dormancy at low temperatures. Our results suggest 483 that evolution during range expansion may be important to consider in the development of 484 models forecasting range shifts under current and future climates.

485 Loss of genetic diversity during range expansion may have fitness consequences and 486 limit adaptive capacity (Lee 2002; Dlugosch and Parker 2008; Clements and Jones 2021). We 487 found that range expansion was only accompanied by minimal losses in heterozygosity but no 488 changes in inbreeding coefficient or the number of private alleles. Prevalent long-distance 489 dispersal from the expanding core to the invasion front could reintroduce allelic variation lost 490 due to bottlenecks in the colonization process. Historical data on leafy spurge range expansion 491 suggests that long-distance, anthropogenic dispersal was likely occurring throughout the 492 invasion process (Selleck et al., 1962; Dunn, 1979). Admixture among independent 493 introductions could also contribute to within-population diversity, especially when there is 494 genetic divergence between sources of introductions (Keller and Taylor 2008; Uller and Leimu 2011). We observed some divergence in PCA space among several populations in or near the 495 496 range core, with populations from the remainder of the range occupying intermediate PCA 497 space. These patterns similarly suggest that gene flow has been substantial among populations 498 despite introduction less than 140 years ago. Leafy spurge is an auto-allohexaploid, and the 499 minor allele frequency spectrum revealed a peak at 0.33, indicating that homoeologous variants

could confound inferences. By filtering out variants with a minor allele frequency (MAF) > 0.26,
we avoided the possibility of detecting spurious population structure, but we also lost some
power given that some higher MAF variants may be truly allelic. As such, we believe that our
approach was conservative, but not overly so, given the substantial biases that can arise in
polyploid datasets.

505 In addition to gene flow during invasion, polyploidy may be an important factor 506 influencing losses of genetic variation during colonization bottlenecks and the capacity for range 507 expansion. Polyploids often maintain higher levels of genetic variation (Otto and Whitton 2000) 508 and there is some evidence that phenotypic plasticity is greater in synthetically produced 509 autopolyploids (Mattingly and Hovick 2023). These factors have been used to explain why 510 polyploids may be better invaders than diploids (Pandit et al. 2011). Leafy spurge is an auto-511 allohexaploid, suggesting that its higher Ne should facilitate the maintenance of genetic diversity 512 within populations and minimize divergence among them. Polyploidy has also been suggested 513 to increase the capacity for adaptive evolution (Otto and Whitton 2000), which could also 514 contribute to success as an invader. It is important to recognize that losses of diversity in DNA 515 sequences do not necessarily translate to losses of variation in quantitative traits (Reed and 516 Frankham 2001). Nevertheless, it is possible that polyploidy contributed to rapid invasion in 517 leafy spurge, but more work is needed to distinguish the contribution of polyploidy from other 518 factors.

Accumulating evidence suggests that invasive plant species frequently undergo climatic niche shifts during range expansion (Medley 2010; Atwater et al. 2018; van Boheemen et al. 2019; Bates and Bertelsmeier 2021); but see (Petitpierre et al. 2012; Liu et al. 2020). Consistent with past findings, we observed climatic niche expansion throughout most of range expansion, except from the late expansion region to the invasion front. From an ENM, we found that warm season temperature had the strongest influence on habitat suitability and therefore may be one source of divergent selection from range core to invasion front. Of course, climate change has

526 already caused poleward shifts in plant species distributions through climatic niche matching 527 (Parmesan 2006; Clements and Ditommaso 2011; Parmesan and Hanley 2015). Therefore, 528 some range expansion may simply involve dispersal to already climatically suitable habitats. 529 However, adaptation may be necessary for continued range expansion, especially where the 530 species is already at its climatic niche limit (Clements and Ditommaso 2011). In leafy spurge, 531 minimal range expansion has been observed in the last 30+ years and populations remain very 532 small/low density at the invasion front, suggesting that populations have reached climatic niche 533 limits. Moreover, ENMs suggest that invasion front populations have very low habitat suitability. 534 Therefore, our results are inconsistent with the hypothesis that the leading edge is highly 535 suitable but expansion is limited by dispersal. Instead, our results suggest that responses to 536 divergent selection may be important for persistence at the leading range edge and for further 537 range expansion.

538 Divergence in trait expression at a leading range edge can be driven by local adaptation, 539 phenotypic plasticity, and/or maternal environmental effects (Des Roches et al. 2017; 540 Westerband et al. 2021). Phenotypic plasticity is considered important during early stages of 541 invasion because it allows introduced populations to establish in a broader range of 542 environmental conditions (Sexton et al. 2002; Richards et al. 2006; Funk 2008; Lande 2015). In 543 germination traits, plasticity could represent a means of habitat selection and niche construction 544 (Donohue 2003, 2005) by which leafy spurge in the leading edge germinates optimally at the 545 onset of spring conditions (warmer temperatures in northern latitudes). Likewise, rapid evolution 546 during range expansion can result from selection on loci that influence dormancy and/or 547 germination timing (Clements and Ditommaso 2011; Hodgins et al. 2018; Clements and Jones 548 2021). We found increased dormancy at lower temperatures in leading edge populations. One 549 possibility is that germination at colder temperatures exposes seedlings to more unpredictable 550 environments (e.g., late season frost) and thus that selection favored germination later in the 551 season for leafy spurge at its northern range limit. Interestingly, other work has suggested that

reduced dormancy evolves at leading range edges (Tabassum and Leishman 2018), contrary to
our findings. Although warm season temperature is most strongly associated with habitat
suitability and relevant to germination in leafy spurge, it is also possible that other variables
influenced the evolution of germination behavior. Overall, our results suggest that range
expansion involved niche expansion and that the evolution of germination timing may have been
important in establishment at the leading range edge.

558 Early life history transitions are thought to be under strong selection because of their 559 cascading effects on later life stages (Baskin and Baskin 1971; Marks and Prince 1981; 560 Donohue 2002, 2005). In plant populations, the environmental conditions at the time of germination can alter the strength and direction of natural selection on postgermination traits 561 (Donohue et al. 2010; D'Aquillo and Donohue 2023). In turn, this can affect the competitive 562 563 environment, resource availability, and density-dependent selective agents experienced by 564 populations (Donohue et al. 2010). While we did not investigate postgermination traits, it would 565 be valuable to investigate whether germination timing in leafy spurge influences performance at 566 later life history stages, especially in leading edge populations.

567 Our results suggest that even over the course of a fairly rapid invasion losses of 568 genomic variation may be minimal. In leafy spurge, this may have occurred because of 569 substantial gene flow during invasion and/or polyploidy. Regardless of the mechanism, higher 570 levels of genetic variation can challenge management when genotypes vary in their responses 571 to eradication measures (Gaskin et al. 2020). We also found that trait divergence may have 572 contributed to climatic niche expansion and thus to the spatial extent of invasion. Forecasts of 573 continued invasion typically rely on species distribution models (SDMs), which rarely take into 574 account evolution. As such, models may fail to predict the complete extent of range expansion, 575 or the severity of range infilling. Evolution-free SDMs are likely still valuable for management 576 planning over meaningful spatial and temporal scales in many systems. However, in systems

- 577 where local adaptation is extensive, forecasts of range shifts with climate change may require
- the construction of regional SDMs that account for evolution.
- 579
- 580
- 581
- 582 Tables
- 583 Table 1. Locality information and genetic diversity metrics (observed heterozygosity, Ho;
- inbreeding coefficient, G_{is}; number of private alleles, and Tajima's D) for 14 leafy spurge
- 585 population samples from four phases of range expansion (see Figure 1).

Range Position	Population Genetic Samples	Latitude	Longitude	Observed Heterozygosity (Ho)	Inbreeding Coefficient (Gis)	Private Alleles	Tajima's D
	Blue Earth	44.16	-94.09	0.068	0.123	27	-0.037
Core	Lyon	44.33	-95.82	0.067	-0.017	9	-0.020
	Meeker	44.94	-94.64	0.07	-0.017	4	0.205
Early	Becker	46.88	-96.05	0.068	0.116	10	-0.052
Expansion	Big Stone	45.52	-96.55	0.067	0.127	9	-0.028
	Polk	47.75	-96.25	0.07	0.097	11	-0.058
	Aitkin	46.98	-93.72	0.068	0.138	27	0.122
Late	Anoka	45.29	-93.13	0.071	-0.026	7	0.101
Expansion	Crow Wing	46.38	-94.22	0.065	0.081	8	0.001
	Winona	44.04	-91.62	0.071	0.156	42	-0.167
	Duluth	46.76	-92.11	0.066	0.12	23	-0.034
Invasion	Koochiching	48.60	-93.40	0.066	0.089	10	-0.008
Front	Pine	46.04	-92.36	0.069	0.138	21	0.000
	St. Louis	47.72	-91.97	0.065	0.046	6	0.068

- Table 2. Pairwise estimates of genetic differentiation measured as G_{st} for 14 leafy spurge
- 589 populations from four phases of range expansion (see Figure 1). Darker versus lighter shading
- 590 of cells indicates higher versus lower values of pairwise G_{st}.

	Blue Earth														
Core	Lyon	0.176													
	Meeker	0.215	0.266	\searrow											
	Becker	0.126	0.173	0.208											
Early Expansion	Big Stone	0.125	0.171	0.209	0.112										
	Polk	0.118	0.171	0.205	0.115	0.119									
	Aitkin	0.124	0.187	0.216	0.126	0.127	0.127								
Late	Anoka	0.187	0.243	0.277	0.177	0.198	0.185	0.197							
Expansion	Crow Wing	0.134	0.188	0.219	0.116	0.128	0.126	0.136	0.201						
	Winona	0.084	0.134	0.181	0.081	0.084	0.081	0.095	0.153	0.096	\square				
	Duluth	0.126	0.163	0.203	0.110	0.116	0.118	0.128	0.179	0.124	0.080				
Invasion	Koochiching	0.136	0.197	0.225	0.139	0.138	0.139	0.152	0.213	0.153	0.099	0.137			
Front	Pine	0.132	0.188	0.204	0.122	0.128	0.130	0.138	0.201	0.135	0.094	0.121	0.153		
	St. Louis	0.178	0.240	0.249	0.169	0.177	0.178	0.177	0.252	0.124	0.143	0.171	0.199	0.182	
		Blue Earth	Lyon	Meeker	Becker	Big Stone	Polk	Aitkin	Anoka	Crow Wing	Winona	Duluth	Koochiching	Pine	St. Louis

- 597 Table 3. ANOVA testing for a relationship between four metrics of sequence diversity and
- 598 geography (latitude, longitude, and their interaction) for 14 population samples. Bold indicates p-
- 599 value less than 0.05.

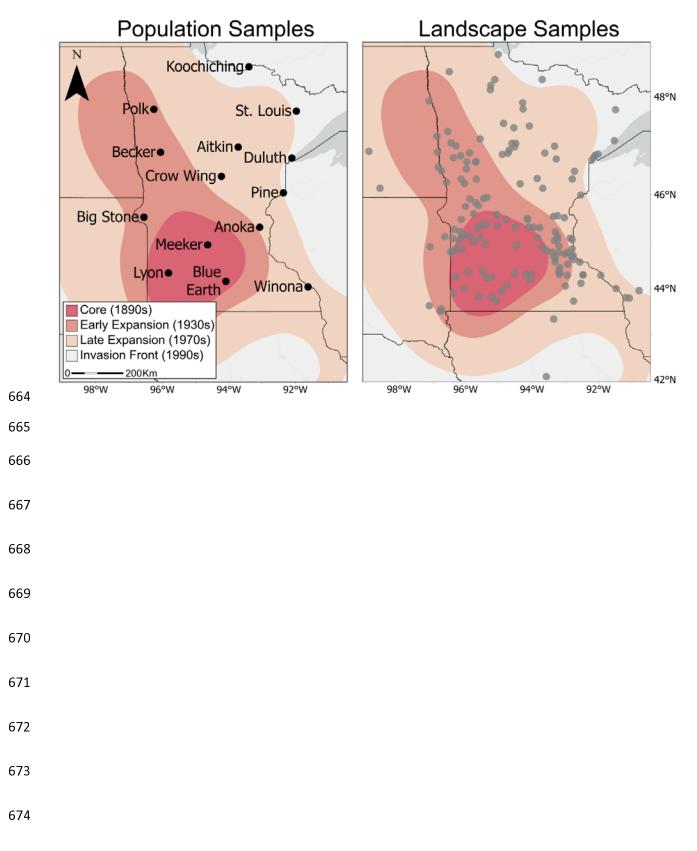
Model	Heterozygosity (Ho)			Private Alleles			Tajima's D				Inbreeding (Gis)			
Term	Df	F	Ρ	Df	F	Ρ	Df	F	Ρ	Df	F	Ρ		
Latitude	1	2.37	0.158	1	0.07	0.803	1	0.24	0.636	1	1.41	0.266		
Longitude	1	0.48	0.506	1	1.29	0.285	1	1.22	0.298	1	0.04	0.846		
Latitude x Longitude	1	9.49	0.013	1	0.84	0.382	1	0.02	0.887	1	0.48	0.508		

611 Figures

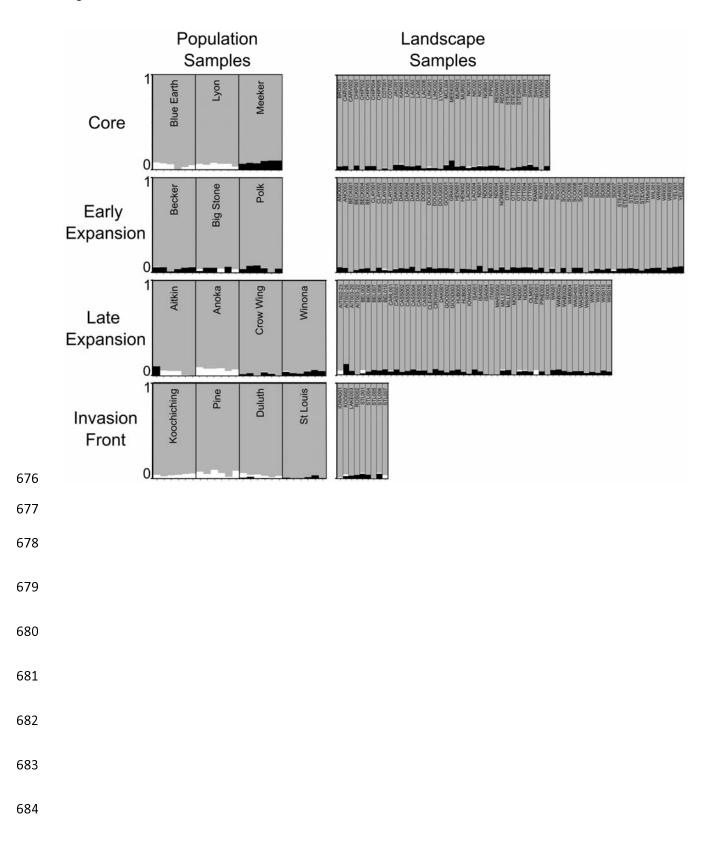
612	Figure 1. Map of the invasion history and tissue collection sites for 14 population samples and
613	157 landscape samples of leafy spurge. The invasion history was derived from Hanson & Rudd
614	(1933) and used to delineate four phases of invasion: core, early expansion, late expansion,
615	and invasion front.
616	
617	Figure 2. Cluster assignment probability from STRUCTURE analyses (K = 3) for population and
618	landscape samples. Each bar represents one individual, and populations are separated by black
619	lines.
620	
621	Figure 3. Principal components analysis (PCA) bi-plots for population genomic data from 14
622	population samples ($n = 6$ per population) and 157 landscape samples ($n = 1$ per population).
623	
624	Figure 4. Isolation by distance (IBD) displayed as scatterplots of genetic distance versus
625	geographic distance for A) population samples (A) and B-E) landscape samples (B-E). For
626	landscape samples, we subset individuals for analyses by successive stages of invasion: B)
627	range core (B), C) range core plus early expansion (C), D) range core, early, plus late
628	expansion ranges (D), and E) all samples from across the four phases (E).
629	
630	Figure 5. Niche overlap during range expansion in climate niche space. The extent of the
631	background environment is outlined in black (solid = total niche space; dashed = 90% of extent).
632	A) Core (dark pink) versus early expansion (light pink), B) Early expansion (light pink) versus
633	late expansion (beige), C) Late expansion (beige) versus invasion front (grey), D) Core (dark
634	pink) versus invasion front (grey). In all panels, the arrow represents the direction of shift in the
635	centroid of niche space.
c2c	

637	Figure 6. Ecological Niche Model (ENM) habitat suitability projections for seed source
638	populations and germination probability for source regions. A) Habitat suitability projection from
639	leafy spurge ENM. Predicted habitat suitability ranges from 0 (purple) to 1 (yellow). The red line
640	demarcates the boundary of 20°C for mean temperature of the warmest quarter. Seed source
641	populations for the germination experiment are marked and colored yellow (>20°C) or purple
642	(>20°C) depending on their mean temperature of the warmest quarter. B) Variable response
643	curve for the mean temperature of the warmest quarter. The vertical red bar denotes 20°C.
644	Vertical yellow bars correspond to seed source populations from the warmer south and vertical
645	purple bars correspond to source populations from the cooler north. C) Mean germination
646	probability (\pm SE) by temperature treatment (X axis) for seed source regions. The warmer
647	southern region is shown in yellow and the cooler northern region is shown in purple.
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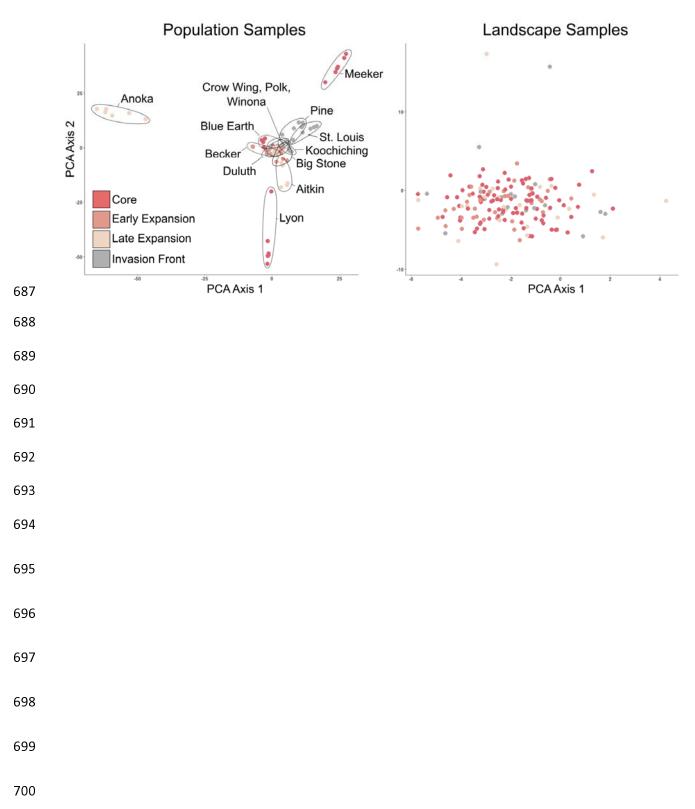




675 Figure 2.



685 Figure 3.





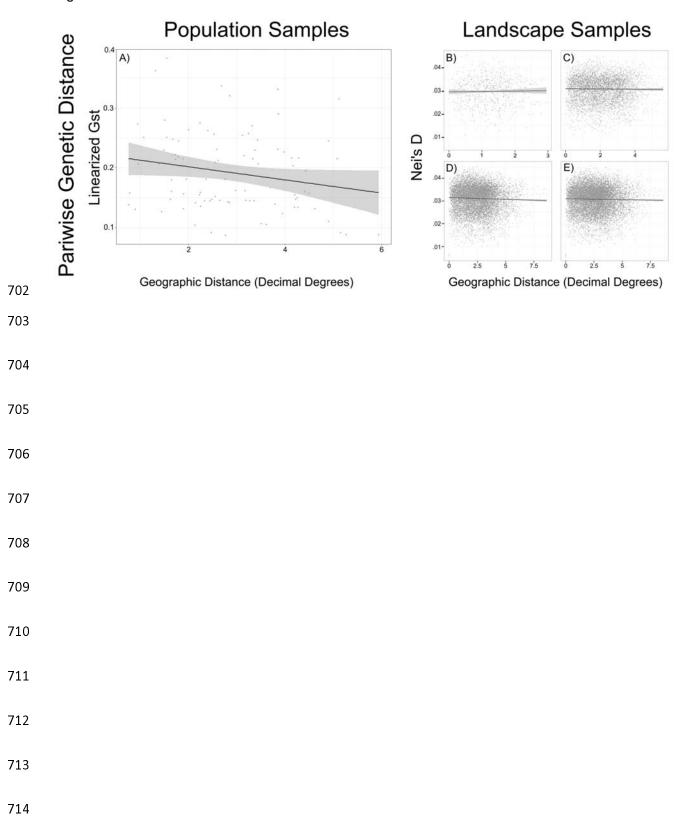
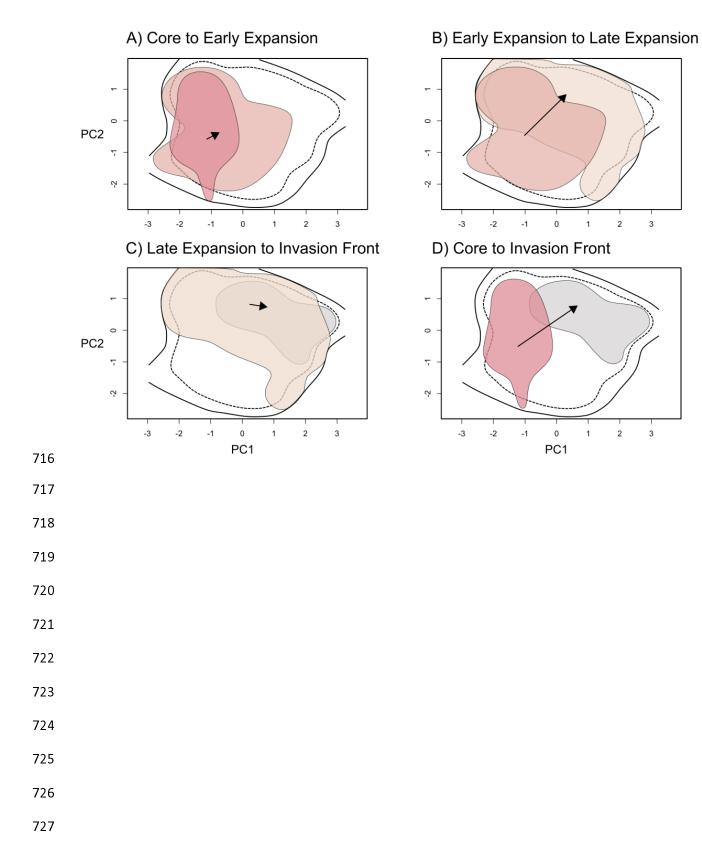
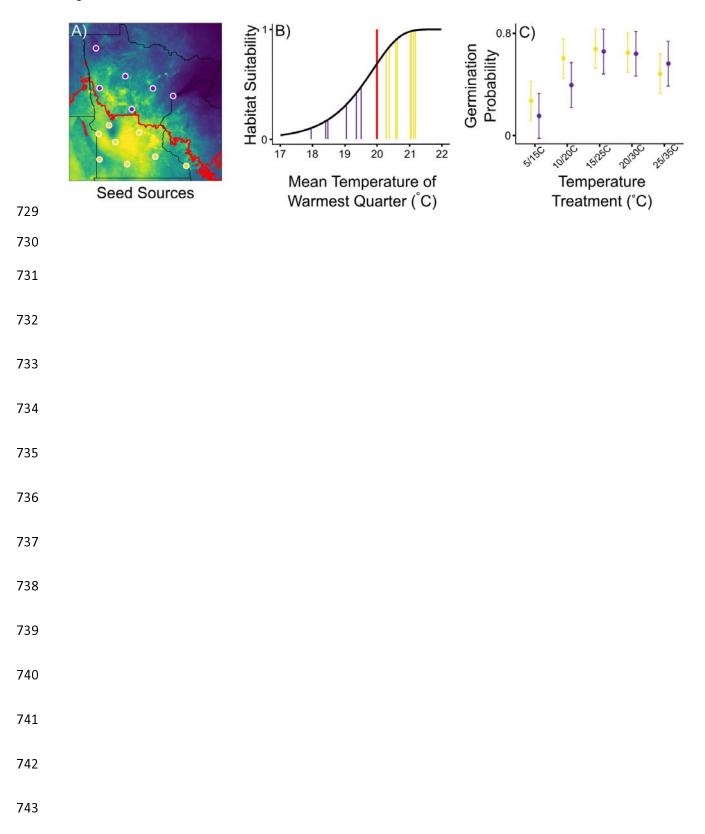


Figure 5.



728 Figure 6.



744 **Statements and Declarations**

745 Funding: Funding for this project was provided by the Minnesota Invasive Terrestrial Plants and 746 Pests Center through the Environment and Natural Resources Trust Fund as recommended by 747 the Legislative-Citizen Commission on Minnesota Resources (LCCMR). TAL was supported by the University of Minnesota Doctoral Dissertation Fellowship. 748

749

750 Competing interests: The authors have no relevant financial or non-financial interests to disclose.

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753 Author contributions: All authors conceived of the study design, contributed to decisions

- 754 throughout the analysis process, wrote, and edited the manuscript. T.A.L and R.B.R collected 755 and analyzed data.
- 756

757 Data Accessibility and Benefit-Sharing: All data will be made publicly available upon publication 758 via the Dryad Digital Repository [DOI].

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