- Selection on early survival does not explain germination
- rate clines in *Mimulus cardinalis* (Phrymaceae)
- Christopher D. Muir^{1,2,*}, Courtney L. Van Den Elzen^{1,3}, Amy L. Angert^{1,4}
- ¹ Department of Botany, University of British Columbia, Vancouver, BC V6T 1Z4, Canada
- ⁵ School of Life Sciences, University of Hawaiʻi, Honolulu, HI 96822, USA
- ⁶ Department of Ecology and Evolutionary Biology, University of Colorado, Boulder, CO 80309,
- 7 USA
- ⁸ Department of Zoology, University of British Columbia, Vancouver, BC V6T 1Z4, Canada
- * Corresponding author
- Manuscript received _____; revision accepted _____.
- short title: Selection on germination in Mimulus cardinalis

Abstract

- Premise Many traits covary with environmental gradients to form phenotypic clines. While
- local adaptation to the environment can generate phenotypic clines, other nonadaptive processes
- may also. If local adaptation causes phenotypic clines, then the direction of genotypic selection
- on traits should shift from one end of the cline to the other. Traditionally genotypic selection on
- non-Gaussian traits like germination rate have been hampered because it is challenging to
- measure their genetic variance.
- Methods Here we used quantitative genetics and reciprocal transplants to test whether a

- ₂₀ previously discovered cline in germination rate showed additional signatures of adaptation in the
- scarlet monkeyflower (Mimulus cardinalis). We measured genotypic and population level
- 22 covariation between germination rate and early survival, a component of fitness. We developed a
- 23 novel discrete log-normal model to estimate genetic variance in germination rate.
- Results Contrary to our adaptive hypothesis, we found no evidence that genetic variation in
- germination rate contributed to variation in early survival. Across populations, southern
- ₂₆ populations in both gardens germinated earlier and survived more.
- ²⁷ Conclusions Southern populations have higher early survival but this is not caused by faster
- germination. This pattern is consistent with nonadaptive forces driving the phenotypic cline in
- germination rate, but future work will need to assess whether there is selection at other life
- 30 stages. The statistical framework should help expand quantitative genetic analyses for other
- waiting-time traits.
- Keywords: cline; genotypic selection; germination; Mimulus; reciprocal transplant

33 Introduction

```
Populations within species differ in ecologically important traits that often result from
   adaptation to different local environments (Turesson, 1922; Clausen et al., 1948). However, it is
   still rare that we understand the traits and selective agents underlying local adaptation
   (Wadgymar, Lowry, et al., 2017). The classic signature of local adaptation is crossing reaction
   norms for fitness measured in a reciprocal transplant experiment (Kawecki and Ebert, 2004;
   Johnson et al., 2021), such that populations have higher relative fitness in their local environment
   but lower relative fitness in a foreign environment. For example, annual inland ecotypes of the
   yellow Monkeyflower Mimulus guttatus have higher relative fitness in seasonally dry sites
41
   compared to perennial coastal ecotypes that are adapted to year-round water availability. The
   relative fitness of coastal and inland ecotypes is reversed on the coast because of life history and
   other genetic differences between ecotypes (Lowry and Willis, 2010). Local adaptation is
   important in evolutionary biology because its prevalence implies that selection varies spatially,
   populations have heritable variation in fitness, and selection is stronger than migration or drift.
   Exceptional cases of local adaptation over short temporal or spatial scales (e.g. Wright et al.
   (2013); Grant and Grant (2014); Richardson et al. (2014); Lescak et al. (2015); Barrett et al. (2019);
   DiVittorio et al. (2020)) likely contribute to an availability bias that leads many to perceive that
   local adaptation is ubiquitous. However, systematic meta-analyses reveal that local adaptation is
   often weak or nonexistent (Leimu and Fischer, 2008; Hereford, 2009; Brady et al., 2019) and may
   be getting weaker because of anthropogenic climate change (Bontrager et al., 2020). Populations
   may not be locally adapted if differential selection is weak relative to migration or drift or they
   lack heritable variation in traits under selection. Alternatively, recent anthropogenic climate
   change may have erased the signature of local adaptation by causing some foreign populations to
   have higher fitness than local populations (Bontrager et al., 2020). For example, in the alpine
   plant Boechera strica, simulated climate change (mimicking early snow melt) favors populations
   from lower elevations and snow addition restores the relative fitness advantage of high-elevation
   populations (Anderson and Wadgymar, 2020). Understanding what causes trait variation among
   populations within species, whether adaptive or not, will help biologists better predict how
   populations will respond to environmental change.
```

```
Phenotypic clines are commonly interpreted as evidence for local adaptation, but nonadaptive
   processes also generate similar patterns (Endler, 1977). A phenotypic cline is a correlation
   between a heritable character and the environment or a proxy for environmental gradients such
   as latitude or elevation (Huxley, 1938). Clines may be caused by either abrupt or gradual
   environmental change. For example, heavy metal contamination of soil around mines is an
   abrupt change from the surrounding pasture soil. There is an abrupt change in the heavy metal
   tolerance of Anthoxanthum odoratum populations across the mine boundary, but other traits
   under weaker or correlated selection vary gradually over space (Antonovics and Bradshaw, 1970).
   Other clines track gradual environmental change. Latitudinal clines in the size of Drosophila
   subobscura flies have evolved multiple times and most likely track gradual variation in
   temperature and/or phenology. But clines are not necessarily adaptive. When there is genetic
   isolation by distance (Wright, 1943), it is possible for phenotypic differences to track genetic
   differences throughout the species range (Vasemägi, 2006). In practice, nonadaptive clines are
   difficult to demonstrate because it is hard to reject adaptive explanations. The proportion of
   cyanogenesis varies clinally in white clover (Trifolium repens) but field experiments have not
   observed spatially varying selection which could explain this cline (Wright et al., 2021). Hence,
   spatial variation in cyanogenesis may be nonadaptive or the design of the experiment may have
   missed selection occurring during germination and early development. When species expand
   their range, such as during biological invasions, nonadaptive clines can readily evolve because of
   multiple introductions and serial founder events (Colautti and Lau, 2015). Even parallel clines,
   which are usually considered strong evidence of natural selection, can result from nonadaptive
82
   processes when there is epistasis and consistent spatial variation in the strength of genetic drift
   (Santangelo et al., 2018). Evolutionary ecological experiments such as reciprocal transplants are
   needed to provide additional evidence to distinguish between adaptive and nonadaptive
   hypotheses for clines (Wadgymar, Daws, et al., 2017).
   There are several experimental methods for measuring the agents and strength of selection to
   different environments under natural conditions (Wadgymar, Lowry, et al., 2017), but here we
   focus on differential selection. One key prediction is that the direction of selection and position
   of the phenotypic optimum will differ between environments. If genetic variation in a trait
   causes fitness to vary, then genotypic selection analyses (Rausher, 1992) should find that
```

genotypes with high trait expression are favored in one environment but disfavored in another. For example, selection favors higher seed dormancy in southern environments of Arabidopsis thaliana and lower dormancy in northern environments, presumably to optimally time seedling establishment with the onset of favorable summer conditions (Postma and Ågren, 2016). Populations can also simultaneously differ as a result of both adaptive and non-adaptive processes. When locally adapted populations differ in traits that do not themselves confer local adaptation, we expect population-level covariation between divergent traits and fitness but do not expect within-population variation to correlate with fitness. Hence, comparing the direction of among-genotype and among-population selection can help distinguish between phenotypic clines that are important for local adaptation and those which may have evolved nonadaptively. 101 In this study, we used a reciprocal transplant to test whether a cline in *Mimulus cardinalis* germination rate could be locally adaptive in terms of early survival between seedling 103 establishment and flowering. Because seeds of M. cardinalis do not require cold or warm 104 stratification to break dormancy, they will germinate whenever conditions are sufficiently warm and moist. Within this germination niche, there is genetic variation for germination time that covaries with latitude and other traits associated with fast-slow growth continuum (Muir and 107 Angert, 2017; Sheth and Angert, 2018; Nelson et al., 2021). In greenhouse conditions, seeds from 108 the southern end of the range (San Diego County, CA) germinate in 6-7 days; seeds from the northern end of the range (Lane County, OR) germinate in 9-10 days. The adaptive hypothesis is 110 that faster germination increases early survival in the southern end of the range and slower germination increases early survival at the northern end. We would predict either stabilizing genotypic selection for different optimal germination rates or opposing signs of directional 113 genotypic selection. If the cline is maintained by nonadaptive forces then germination rate 114 should be uncorrelated with genotypic fitness (even if it is correlated with population mean fitness) or the direction of selection should be the same in both parts of the range. A limitation of this study is that we cannot measure selection on seedling establishment since that occurred in the greenhouse (see Materials and Methods) or later survival and fecundity (these fitness 118 components will be analyzed in a follow-up paper).

Germination and early survival are potentially important for local adaptation but often

overlooked compared to later life stages. Failure to measure seedling establishment and early 121 survival can bias estimates of selection on traits that only express at later life stages, such as 122 flower size (Mojica and Kelly, 2010). In the herb Arabidopsis thaliana, quantitative differences in germination timing contributed to local adaptation and are maintained by a balance of selection on seedling establishment, survival, and fecundity, but the optimal timing varies in northern and southern environments (Postma and Agren, 2016, 2018). A challenge of measuring genotypic 126 selection on germination rate is that quantitative genetic theory and analysis are built around 127 Gaussian distributions (Lynch and Walsh, 1998; but see Villemereuil et al., 2016), but the distribution of times until germination are most likely not Gaussian. Note that time to germination is the inverse of germination rate. The time until germination can be modeled as a 130 waiting time, which are bounded at zero and usually right skewed. Waiting times are distributed 131 exponentially if the probability of germination is constant through time, but this unlikely to be 132 true. Rather, the probability of germination almost certainly increases through time once 133 conditions are conducive. Hence, more complex waiting time distributions should be seen. A second challenge is that waiting time is usually modeled as a continuous trait, but is observed at 135 discrete intervals (e.g. once per day). If the average time to germination is sufficiently long 136 relative to the interval between observations, then a continuous-time model may be sufficient. However, when time to germination is several days, as with *M. cardinalis*, then accurately modeling the trait may need to account for the discrete observation process. Hence, 139 investigating spatially variable selection on germination time is important for expanding our 140 knowledge of local adaptation and phenotypic clines for poorly studied life stages with 14 challenging statistical properties.

Materials and Methods

Study system

Mimulus cardinalis (=Erythranthe cardinalis (Lowry et al., 2019)) is a perennial forb native to the
 Western US (California and Oregon). It is predominantly outcrossing, self-compatible,
 hummingbird-pollinated. We used five focal populations from throughout the geographic range
 of M. cardinalis (Table 1). These five focal populations are a subset of those we used previously to

identify a cline in germination and other traits (Muir and Angert, 2017). Seeds were collected in
the field from mature, undehisced fruits left in open coin envelopes for 2-4 weeks to dry, then
stored at room temperature. To control for maternal effects, we grew a large number of
field-collected seeds in the greenhouse and generated seed families for this experiment by
hand-pollinating individuals using the breeding design described in the next section.

Table 1: Source populations, including the name of the drainage where the seeds were collected, the latitude, longitude, and elevation in meters above sea level (mas).

Name	Latitude	Longitude	Elevation (mas)
Sweetwater River	32.900	-116.585	1180
West Fork Mojave River	34.284	-117.378	1120
North Fork Middle Tule River	36.201	-118.759	926
Little Jamison Creek	39.743	-120.704	1603
Rock Creek	43.374	-122.957	326

54 Genetic variance and heritability

environmental (V_E) components:

To estimate selection among genotypes and populations, we first need to quantify genetic 155 variance in germination rate and early survival. We estimated genetic variance and heritability 156 of germination rate and winter survival using a quantitative genetic breeding design. For each 157 population, we crossed 15 parental individuals using a partial diallel design (Lynch and Walsh, 1998) with three dams per sire and three sires per dam for a total of 45 full-sib families per 159 population. Individual plants of this self-compatible species were used as both sires (pollen 160 parent) and dams (ovule parent). We did not make crosses between populations. One family 161 from the Little Jamison Creek population did not produce enough seeds, resulting in 44 usable 162 families. In total, we used $5 \times 45 - 1 = 224$ families. We estimated genetic variances and heritabilities for each of the five populations (Table 1). The total phenotypic variance within a population \mathcal{V}_P can be partitioned into genetic (\mathcal{V}_G) and

$$V_P = V_G + V_E$$

We did not have statistical power to estimate nonadditive genetic contributions, therefore we made the simplifying operational assumption that all genetic variance is additive. The broad-sense (H^2) heritability is the fraction of phenotypic variance contributed by additive genetic variance:

$$H^2 = \frac{V_G}{V_P}$$

When separate individuals are used as dams and sires, additive genetic variance of the base population is often estimated from the variance among sires: $V_G=4~\sigma_{\rm sire}$ (Lynch and Walsh, 1998). Since our breeding design used hermaphroditic individuals as both dams and sires, we estimated parental breeding values from their contributions as both sires and dams. Hence, we estimated V_G from the variance among parents:

$$V_G = 4 \sigma_{\rm parent}$$

Using $\sigma_{\rm parent}$ rather than $\sigma_{\rm sire}$ is more powerful because it uses all of the data, but assumes that maternal and paternal effects are negligible. We estimated σ_{parent} and other parameters using a 177 Bayesian mixed effects model. We describe the general approach here and provide detail specific 178 to germination rate and survival below. We fit the model in Stan (Stan Development Team, 2021) 179 using **cmdstanr** version (Gabry and Češnovar, 2021). Stan calculates the posterior distribution of 180 the model using the Hamiltonian Monte Carlo algorithm, which is similar to the more widely used Markov Chain Monte Carlo, but is faster and more efficient at sampling for many 182 applications (Monnahan et al., 2017). We used weakly informative normally distributed priors for 183 parameters that affect the trait mean (intercepts and coefficients) and half-Cauchy priors for 184 variance parameters. Weakly informative priors are strongly recommended for complex mixed 185 models (McElreath, 2016). We ran 4 parallel chains for 4000 warmup iterations, 4000 sampling iterations, and a thinning interval of 4. This configuration allowed parameters to converge, 187 which we defined as the convergence diagnostic $\hat{R} < 1.05$ and an effective sample size > 1000188

(Vehtari et al., 2021). We inspected posterior predictive plots to assess model adequacy using the pp_check function from the *R* package **bayesplot** version 1.8.1 (Gabry et al., 2019). We used the posterior median for point estimates and calculated uncertainty with the 95% quantile intervals from the posterior distribution. *Stan* and *R* code are available on GitHub (https://github.com/cdmuir/mimulus-germination) and will be archived on Zenodo upon publication. Data will be deposited on Dryad upon publication.

Germination rate

We measured germination rate by counting the number of days to germination in 48 randomized blocks split evenly between two time cohorts (North cohort: August 3-4 and South cohort: August 22-24, 2015) in the University of British Columbia greenhouse (Vancouver, BC, Canada). 198 The North and South designations refer to the transplant garden location, as described below. We sowed 3-5 seeds per plug on moist perlite-peat moss potting medium (Sunshine #1, Sungro Horticulture, Agawam, Massachusetts, USA) and recorded the first day on which a plug had at least one germinated seedling. We later thinned each plug to one seedling by selecting the 202 central-most individual, so the first germinated seedling recorded is generally not the same 203 individual transplanted into the field. Hence, individual correlations between germination and 204 survival are inappropriate and we analyze relationships between breeding values or population averages. We censused germination daily for most of the experiment, but census dates were more spread out later in the experiment when only a few plants remained. We accounted for census interval in our statistical model (see below). Gentle misting kept soil moist for the 208 duration of the experiment. Most plugs (9320 of 10650, 87.5%) had at least one germinant by the 200 end of the experiment (North cohort: August 27, 2015; South cohort: September 20, 2015); we treated plants that did not germinate as missing data. 5 (0.05%) plants emerged before day 5. We removed data on these individuals because they were likely not M. cardinalis, but rather 212 contamination from another species' seed in the potting medium. Species can be difficult to tell 213 apart as the cotyledons first emerge, but in plugs where a single germinant emerged before day 5 it was always a contaminant species. It therefore seems most likely that all germinants before day 5 were contamination, but observations during thinning showed that contamination was rare and it unlikely to have any substantial impact on our estimate of germination timing.

Estimating genetic variance in germination rate

The full germination-rate model included a fixed effect of population, fixed effect of cohort, a 219 random effect of greenhouse block, and a random effect of parent as either dam or sire to estimate 220 additive genetic variance. Most quantitative traits are modeled with a Gaussian (i.e. normal) distribution, but this is inappropriate for germination which has a lower bound at o and is censused at discrete time points. A Gaussian distribution might be an adequate approximation if 223 the time to germination was longer and more spread out so that the effects of a zero lower bound 224 and discrete sampling interval were inconsequential. However, given the rapid germination rate 225 and small spread in timing for M. cardinalis, we used a different approach. Instead, we modeled 226 genetic variation in a continuously distributed latent rate parameter that is measured as a discrete number of days to germination. Time to germination can be modeled similar to other 228 waiting time or survival processes. There are several common waiting-time distributions with a 229 lower bound at zero, but we used the log-normal distribution because unlike some other common distributions (e.g. exponential), the mean and variance of log-normal distribution are 23 separate parameters. Hence, we can model the variance components separate from differences in 232 the mean germination rate. We discretized the distribution using the definite integral of the 233 probability density to calculate a probability mass. Let the probability density with parameters θ 234 be $f(x|\theta)$. If a seed was censused to have germinated on day i, it could have germinated at any time between the previous census on day t_{i-1} . The associated probability mass $g(x|\theta)$ is:

$$g(x = t_i|\theta) = \int_{t_{i-1}}^{t_i} f(x|\theta) dx$$

Preliminary analyses revealed a germination threshold of four days since we did not observe any germination before the fifth day after sowing. The model did not fit as well without this threshold because models without a threshold predict a significant amount of germination between days o and 5 (results not shown). The variances and heritability are reported on the log-tranformed scale.

Transplant gardens and winter survival

After germination, we transplanted 20 out of 24 blocks of seedlings each to two experimental 243 gardens located at the northern and southern portions of the species' range. We refer to these as 244 the "North" and "South" gardens, respectively. We used some plants from the remaining 4 blocks to replace those that did not germinate or died during transport. We planted the North and South cohorts 5-6 weeks after germination on September 9-17, 2015 and October 2-9, 2015, 247 respectively. The North garden was located near natural populations along the Middle Fork of 248 the Feather River (39° 46' 59.7" N, 120° 38' 31.1" W, 1314 mas, Plumas County, California, USA); the South garden was located along King Creek (32° 54' 23.4" N, 116° 37' 26.0" W, 1011 mas, San Diego County, California, USA). These gardens are located close to two of the source populations. The North garden is 7.1 km from the Little Jamison Creek source population; the South garden is 252 3.8 km from Sweetwater River source population. Since M. cardinalis is a riparian specialist, 253 gardens were located within 50 m of natural waterways where the microclimate closely resembled that of nearby natural populations. We irrigated plots with microperforated drip tape 255 (Toro, Bloomington, Minnesota) placed about each seedling to help establish transplants and 256 mimic riparian soil where M. cardinalis naturally germinates. Seedlings were initially irrigated 257 once per day and gradually reduced as evaporation declined. Irrigation ended when winter 258 precipitation began. We discarded data on plants that died within the first month after transplanting, as this probably indicates transplant shock rather than natural mortality. We monitored reemergence in the spring using monthly censuses beginning on April 15, 2016 (North) and March 14, 2016 (South). If a plant was recorded in the last census of 2015 and emerged in spring 2016, we counted it as alive. If a plant did not reemerge, we recorded winter mortality.

Estimating genetic variance in winter survival

We treated winter survival as a binomially distributed trait determined by a latent probability of survival $p_{\rm surv}$. We therefore estimated genetic variance in $p_{\rm surv}$ and then used methods described by Villemereuil et al. (2016) for non-Gaussian traits to calculate heritability. We analyzed Garden as a fixed effect and garden Block as a random effect. The primary difference from the

germination model is that we estimated genotype-by-environment ($G \times E$) interactions to test whether populations are locally adapted. If there is local adaptation, southern populations should have higher survival in the South garden and lower survival in the North garden, and *vice versa*.

Selection on germination rate

We estimated genotypic and population-level selection on winter survival as a function of germination rate. If genetic variation in germination rate within and between populations causes 275 the probability of survival to change, then there should be significant genotypic selection on germination rate. Alternatively, populations with different mean germination rates may have different probabilities of survival as a consequence of other, confounding trait differences. If this is true, we predict a population-level correlation between germination rate and survival, but no 279 evidence of genotypic selection on variation within populations. For each sample of the posterior 280 distribution, we used ordinary linear regression to calculate the relationship between germination rate and winter survival in both gardens at both genotypic and populational levels. We also calculated the variance among populations (V_{pop}) in germination rate and survival from 283 the posterior distribution. We used the median and quantile intervals to estimate the slopes and 284 quantify uncertainty in our estimates. We used the same approach to estimate quadratic 285 selection coefficients as well.

287 Climate

We compared the climate during the experiment (September 2015 to May 2016) to a standardized climate normal. Following (Sheth and Angert, 2018), we downloaded climate variables for 1961-1990 and 2015-16 derived from ClimateNA version 7.10 (Wang et al., 2016). We compared the seasonal temperature average (autumn: September-November; winter: December-February; spring: March-May) for the experiment with the 1961-1990 normal for each population.

Results

Southern populations germinate faster and have a higher probability of winter survival

The vast majority of seeds germinated in 1-2 weeks, but northern populations took a few days longer on average. 4206 out of 5321 plugs (79.0%) had germinants in the North cohort; 5114 out of 5328 plugs (96.0%) had germinants in the South cohort. South cohort seedlings also germinated about one day faster (mean = 7.7, median = 7) than the North cohort (mean = 8.9, 299 median = 8). After accounting for cohort and block effects, southern populations consistently 300 germinate faster (Fig. 1a). For example, seeds from the slowest population (Rock Creek) 30 germinated 2.9~(95%~CI~[2.1-3.9]) days slower than the fastest population (West Fork Mojave River). The discretized log-normal distribution was an adequate model of germination rate based 303 on the similarity of the posterior predictions to the observed data (Fig. S1). Fewer plants survived over winter in the North garden than in the South garden (65.5% versus 305 94.2%). In both gardens, southern populations survived more often than northern populations 306 even after accounting for block effects (Fig. 1b). For example, in the North garden, we estimated that plants originating from the most southern population, Sweetwater River, survived $80.4\%~(95\%~{
m CI}~[71.6\%-87.9\%])$ of the time compared to $52.1\%~(95\%~{
m CI}~[39.9\%-65.8\%])$ for the most northern population Rock Creek. Similarly, in the South garden the local population from Sweetwater River survived $95.2\%~(95\%~\mathrm{CI}~[91.2\%-97.2\%])$ of the time compared to 31 $85.4\%~(95\%~{\rm CI}~[75.9\%-90.8\%])$ for Rock Creek. The average temperature was warmer than the historical normal from 1961-1990, especially for the winter season in the south garden (Fig. S2)

Variance in germination rate and survival within and between populations

The genetic variation in germination rate within populations of M. cardinalis is similar to the genetic differences between them (Fig. 2a). The between-population variance in the log-mean germination rate is 0.023 (95% CI [-0.050-0.091)] greater than that within populations, but the confidence intervals include 0. The environmental variance $V_{\rm E}$ is greater than $V_{\rm G}$, resulting in a moderate heritability H^2 (Fig. 2b). In contrast, there is virtually no genetic variance in winter survival ($p_{\rm surv}$) in either garden, resulting in heritabilities that are effectively 0 (Fig. 3). There is

some variation between populations, but it is substantially less than the unexplained
environmental variance in both gardens (Fig. 3). Parameter estimates and confidence intervals
for all variance components and heritabilities are given in Table S1 for germination rate and
Table S2 for winter survival.

Directional selection favor faster germination between but not within populations

Since there was almost no genetic variance in winter survival in either garden, but some
variance between populations, the only evidence for selection is between populations.

Populations originating from farther south in the species' range germinated faster and had
higher winter survival than those originating for farther north (Fig. 4). However, within
populations, genotypes that germinated faster did not have a different probability of survival.
Selection coefficients and confidence intervals are given in Table S3 for between-population
selection and Table S4 for genotypic selection. There was no evidence for stabilizing selection
using quadratic regression (results not shown).

Discussion

Phenotypic clines are often interpreted as evidence for local adaptation but evolutionary forces besides selection can also generate associations between phenotypes and environment (Endler, 338 1977). If genotypic selection measured under natural conditions favors local phenotypes, this 339 supports the hypothesis that phenotypic clines are adaptive. Alternatively, failure to detect selection could indicate that nonadaptive evolutionary forces maintain phenotypic variation. We 34 measured the effect of germination rate in the greenhouse, which varies clinally with latitude in 342 Mimulus cardinalis (Muir and Angert, 2017), on early winter survival. Although there was 343 genetic variation in germination rate within and between five populations (Table 1; Table S1; Fig. 1a), there was no evidence selection consistently favored local genotypes. We predicted that earlier germination should be favored in the South garden and later germination favored in the North garden. Instead, in both North and South gardens, populations that germinated faster on 347 average also had higher survival, though survival in the North garden was lower than the South

overall (Fig. 4). Apparent selection for faster germination in both gardens is likely indirect rather than causal. Populations originating from the southern range limit survived more often and germinated faster. Within populations there was no genetic variance in winter survival despite genetic variance in germination rate, indicating that differences in germination rate are probably not the main cause of differences in winter survival. Our results are not consistent with selection on survival maintaining phenotypic clines in germination rate. Rather, the results may indicate a role for nonadaptive evolution in this and possibly other traits.

Why are plants from southern populations more likely to survive winter?

An unexpected finding is that southern populations had higher survival in both South and North 358 gardens. If populations were locally adapted, we would expect to see local populations have the 359 highest fitness. We consider a three explanations. One trivial explanation is that artificial 360 conditions of the experiment may have favored genotypes from southern populations for an unknown reason. Although we attempted to situate gardens in natural conditions and at realistic conspecific densities, southern genotypes may be better adapted to artificial manipulations such 363 as removing heterospecific competitors, ground cloth, and/or irrigation watering. A second 364 possibility is that climatic anomalies favor southern, warm-adapted populations in all gardens. 365 Temperature in the North garden was warmer than average for the local population (Little Jamison Creek) but not substantially warmer than the Rock Creek population, which had even lower survival. If we had planted in a season closer to the historical temperature and snowpack, 368 then we might have observed local adaptation in northern populations. Third, early survival may tradeoff with other fitness components we have not measured or analyzed here, including seedling establishment, later viability, and/or fecundity. We cannot address selection on seedling establishment because we germinated seedlings under controlled greenhouse conditions. In 372 Arabidopsis thaliana, faster germination increased survival but decreased seedling establishment 373 (Postma and Ågren, 2018). Southern population growth rate could be more sensitive to 374 germination timing, leading to stronger selection, so selection has boosted it, whereas in the north it matters less and/or conflicting selection at other life stages is stronger. We will address these hypotheses further in a separate paper including multiple seasons of viability and

fecundity data from these experiments. We are not including those data in the current analysis
because we are using them to address a conceptually distinct question. However, we believe the
current analysis advances the field by addressing selection on germination at one life stage and
presents a new statistical method for the quantitative genetics of waiting-time traits.

What maintains genetic variation in germination rate?

One possibility is that germination rate measured in the greenhouse is not representative of 383 what occurs in nature. In the field, seeds may use additional cues such as day length or temperature that did not vary in our experiment. We may not be able to measure selection on variation in germination rate unless it occurs in the field. For example, seedlings in the north 386 may require a longer duration of favorable conditions to ensure they do not germinate in late fall 387 or during a brief warm period in early spring. Either scenario might cause them to be exposed to 388 a hard, damaging frost. We may not be able to observe this type of selection without germinating plants in the field at different times of the season. Alternatively, variation may be the result of nonadaptive evolution. M. cardinalis has expanded its range northward most likely after the last 39 glacial maximum (Sheth and Angert, 2018). Population expansion can lead to the accumulation 392 of deleterious alleles because of stronger genetic drift (Peischl et al., 2013), serial founder events (Slatkin and Excoffier, 2012), and maladaptation (Polechová, 2018). If selection actually favors faster germination, the slower germination and lower survival of northern populations may both be caused by the accumulation of deleterious genetic variation at the expanding range margins. 396 The current cline could also be a legacy of selection in the recent past which is now less strong or 397 at least not observed during our study (Saccheri et al., 2008).

statistical approaches for non-Gaussian traits like germination time

Statistical challenges of analyzing non-Gaussian traits may direct research away from "hard" traits or applying Gaussian methods to non-Gaussian traits may lead to poor estimation. A discretized log-normal distribution may be useful for analyzing genetic variation in germination time and other "waiting-time" traits with means close to zero. Discretization accounts for the observation process, which is often completed at discrete time points (e.g. daily, weekly, etc.).

Discretization becomes less important when the observation interval is small relative to the total duration, in which case a continuous approximation is good enough. Waiting-time distributions are naturally bounded at o and often skewed, both of which are violated with a Gaussian distribution. We used the log-normal distribution, but other distributions (e.g. exponential or Weibull) may be more appropriate in other situations. Probabilistic programming languages like Stan (Stan Development Team, 2021) enable evolutionary ecologists and quantitative geneticists to specify custom probability distributions and estimate their parameters from data (see also Hadfield (2010)).

13 Conclusions

The reciprocal transplant experiment did not support our hypothesis that local adaptation
maintains clinal variation in germination rate in *M. cardinalis*. We cannot exclude the possibility
that artificial aspects of our experimental design may have prevented us from measuring
selection on germination rate realistically. It is also possible that climate anomalies during our
experiment prevented us from observing historical patterns of selection that led to present-day
phenotypic variation. It is plausible that the cline in germination rate and other phenotypes may
be caused by nonadaptive forces that increase the frequency of deleterious alleles during range
expansion. Studies like ours that combine clines with reciprocal transplants are necessary to
determine what evolutionary forces shape range-wide phenotypic variation within species.

Acknowledgements

Roger Wynn, Gail Wynn, Christina Schneider, and Norbert Schneider provided land for the gardens. Matthew Bayly, Emily Gercke, Kevin Muir, Rachel Muir, Erin Warkman, and Adam Wilkinson helped with data collection or transplanting. Seema Sheth provided seeds and gave feedback on an earlier draft of the manuscript. Lila Fishman and Kelsey Byers provided suggestions on the non-adaptive cline literature.

Author Contributions

- 430 CDM and ALA designed the experiment. CDM and CLV collected germination data. CDM and
- ALA collected data on survival in the field. CDM analyzed the data and wrote the manuscript
- with input from ALA and CLV.

Data Availability Statement

- Data will be deposited on Dryad upon publication. Source code for all analyses will also be
- archived on Zenodo. Data and code are available to reviewers on GitHub
- https://github.com/cdmuir/mimulus-germination. The GitHub repo will be revised during
- revision.

Supporting Information

Fig. S1-S2 and Tables S1-S4 may be found online in the Supporting Information section.

440 Literature Cited

- Anderson, J. T., and S. M. Wadgymar. 2020. Climate change disrupts local adaptation and favours upslope migration. *Ecology Letters* 23: 181–192.
- Antonovics, J., and A. Bradshaw. 1970. Evolution in closely adjacent plant populations. 8. Clinal patterns at a mine boundary. *Heredity* 25: 349–362.

- Barrett, R. D. H., S. Laurent, R. Mallarino, S. P. Pfeifer, C. C. Y. Xu, M. Foll, K. Wakamatsu, et al. 2019. Linking a mutation to survival in wild mice. Science 363: 499–504. 446
- Bontrager, M., C. D. Muir, C. Mahony, D. E. Gamble, R. M. Germain, A. L. Hargreaves, E. J. 44 Kleynhans, et al. 2020. Climate warming weakens local adaptation. Evolutionary Biology.
- Brady, S. P., D. I. Bolnick, R. D. H. Barrett, L. Chapman, E. Crispo, A. M. Derry, C. G. Eckert, et al. 449
- 2019. Understanding Maladaptation by Uniting Ecological and Evolutionary Perspectives. 450
- The American Naturalist 194: 495-515.
- Clausen, J., D. Keck, and W. Hiesey. 1948. Experimental studies on the nature of species. III:
- Environmental Responses of Climate Races of Achillea. Carnegie Institution of Washington, 453
- Washington, D.C. 454

448

456

- Colautti, R. I., and J. A. Lau. 2015. Contemporary evolution during invasion: Evidence for 455 differentiation, natural selection, and local adaptation. *Molecular Ecology* 24: 1999–2017.
- DiVittorio, C. T., S. Singhal, A. B. Roddy, F. Zapata, D. D. Ackerly, B. G. Baldwin, C. R. Brodersen, 457
- et al. 2020. Natural selection maintains species despite frequent hybridization in the desert 458
- shrub encelia. Proceedings of the National Academy of Sciences 117: 33373-33383.
- Endler, J. A. 1977. Geographic variation, speciation, and clines. Princeton University Press,
- Princeton, N.J. 461
- Gabry, J., and R. Češnovar. 2021. Cmdstanr: R Interface to 'CmdStan'.
- Gabry, J., D. Simpson, A. Vehtari, M. Betancourt, and A. Gelman. 2019. Visualization in Bayesian 463 workflow. Journal of the Royal Statistical Society: Series A (Statistics in Society) 182: 389-402.
- Grant, P. R., and B. R. Grant. 2014. 40 years of evolution: Darwin's Finches on Daphne Major Island. Princeton University Press, Princeton; Oxford.
- Hadfield, J. D. 2010. MCMC Methods for Multi-Response Generalized Linear Mixed Models: The **MCMCglmm** *r* Package. *Journal of Statistical Software* 33. 468

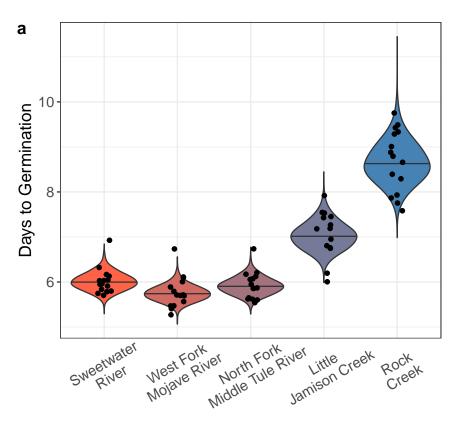
- Hereford, J. 2009. A Quantitative Survey of Local Adaptation and Fitness Trade-Offs. *The***American Naturalist 173: 579–588.
- Huxley, J. 1938. Clines: An Auxiliary Taxonomic Principle. Nature 142: 219-220.
- Johnson, L. C., M. B. Galliart, J. D. Alsdurf, B. R. Maricle, S. G. Baer, N. M. Bello, D. J. Gibson, and
- A. B. Smith. 2021. Reciprocal transplant gardens as gold standard to detect local adaptation in
- grassland species: New opportunities moving into the 21st century. Journal of Ecology:
- 475 **1365-2745.13695.**
- Kawecki, T. J., and D. Ebert. 2004. Conceptual issues in local adaptation. *Ecology Letters* 7:
- 1225-1241.
- Leimu, R., and M. Fischer. 2008. A meta-analysis of local adaptation in plants. *PLoS One* 3:
- e4010-e4010.
- Lescak, E. A., S. L. Bassham, J. Catchen, O. Gelmond, M. L. Sherbick, F. A. von Hippel, and W. A.
- Cresko. 2015. Evolution of stickleback in 50 years on earthquake-uplifted islands. *Proceedings*
- of the National Academy of Sciences 112: E7204–E7212.
- Lowry, D. B., J. M. Sobel, A. L. Angert, T. Ashman, R. L. Baker, B. K. Blackman, Y. Brandvain, et al.
- 2019. The case for the continued use of the genus name *mimulus* for all monkeyflowers.
- Taxon 68: 617–623.
- Lowry, D. B., and J. H. Willis. 2010. A widespread chromosomal inversion polymorphism
- contributes to a major life-history transition, local adaptation, and reproductive isolation.
- ⁴⁸⁸ PLoS biology 8: 2227.
- Lynch, M., and B. Walsh. 1998. Genetics and analysis of quantitative traits. Sinauer, Sunderland,
- 490 Mass.
- McElreath, R. 2016. Statistical Rethinking: A Bayesian Course with Examples in R and Stan. 1st
- ed. Chapman; Hall/CRC.
- ⁴⁹³ Mojica, J. P., and J. K. Kelly. 2010. Viability selection prior to trait expression is an essential
- component of natural selection. *Proceedings of the Royal Society B: Biological Sciences* 277:
- ⁴⁹⁵ **2945-2950.**

- Monnahan, C. C., J. T. Thorson, and T. A. Branch. 2017. Faster estimation of Bayesian models in ecology using Hamiltonian Monte Carlo. *Methods in Ecology and Evolution* 8: 339–348.
- Muir, C. D., and A. L. Angert. 2017. Grow with the flow: Faster growth is associated with more variable precipitation in a perennial herb. *Journal of Evolutionary Biology* 30: 2189–2203.
- Nelson, T. C., C. D. Muir, A. M. Stathos, D. D. Vanderpool, K. Anderson, A. L. Angert, and L.
- Fishman. 2021. Quantitative trait locus mapping reveals an independent genetic basis for
- joint divergence in leaf function, life-history, and floral traits between scarlet monkeyflower (
- Mimulus cardinalis) populations. American Journal of Botany 108: 844-856.
- Peischl, S., I. Dupanloup, M. Kirkpatrick, and L. Excoffier. 2013. On the accumulation of deleterious mutations during range expansions. *Molecular Ecology* 22: 5972–5982.
- Polechová, J. 2018. Is the sky the limit? On the expansion threshold of a species' range. *PLOS*Biology 16: e2005372.
- Postma, F. M., and J. Ågren. 2018. Among-year variation in selection during early life stages and the genetic basis of fitness in *Arabidopsis thaliana*. *Molecular Ecology* 27: 2498–2511.
- Postma, F. M., and J. Ågren. 2016. Early life stages contribute strongly to local adaptation in

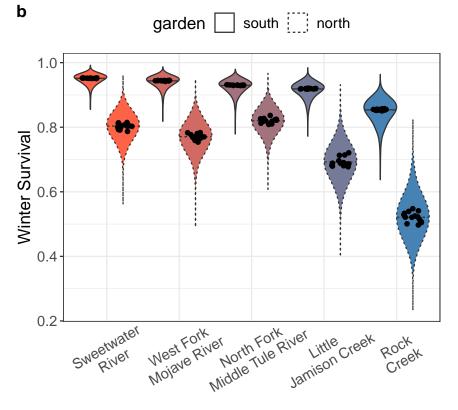
 Arabidopsis thaliana. Proceedings of the National Academy of Sciences 113: 7590-7595.
- Rausher, M. D. 1992. The measurement of selection on quantitative traits: Biases due to environmental covariances between traits and fitness. *Evolution* 46: 616–626.
- Richardson, J. L., M. C. Urban, D. I. Bolnick, and D. K. Skelly. 2014. Microgeographic adaptation and the spatial scale of evolution. *Trends in Ecology & Evolution* 29: 165–176.
- Saccheri, I. J., F. Rousset, P. C. Watts, P. M. Brakefield, and L. M. Cook. 2008. Selection and gene flow on a diminishing cline of melanic peppered moths. *Proceedings of the National Academy*of Sciences 105: 16212–16217.
- Santangelo, J. S., M. T. J. Johnson, and R. W. Ness. 2018. Modern spandrels: The roles of genetic drift, gene flow and natural selection in the evolution of parallel clines. *Proceedings of the Royal Society B: Biological Sciences* 285: 20180230.

- Sheth, S. N., and A. L. Angert. 2018. Demographic compensation does not rescue populations at a trailing range edge. *Proceedings of the National Academy of Sciences* 115: 2413–2418.
- Slatkin, M., and L. Excoffier. 2012. Serial Founder Effects During Range Expansion: A Spatial
 Analog of Genetic Drift. *Genetics* 191: 171–181.
- 526 Stan Development Team. 2021. Stan Modeling Language Users Guide and Reference Manual.
- Turesson, G. 1922. The genotypic response of the plant species to habitat. *Hereditas* 3: 211-350.
- Vasemägi, A. 2006. The Adaptive Hypothesis of Clinal Variation Revisited: Single-Locus Clines as a Result of Spatially Restricted Gene Flow. *Genetics* 173: 2411–2414.
- Vehtari, A., A. Gelman, D. Simpson, B. Carpenter, and P.-C. Bürkner. 2021. Rank-Normalization,
- Folding, and Localization: An Improved r for Assessing Convergence of MCMC (with
- Discussion). Bayesian Analysis 16.
- Villemereuil, P. de, H. Schielzeth, S. Nakagawa, and M. Morrissey. 2016. General Methods for
 Evolutionary Quantitative Genetic Inference from Generalized Mixed Models. *Genetics* 204:
- ₅₃₅ 1281-1294.
- Wadgymar, S. M., S. C. Daws, and J. T. Anderson. 2017. Integrating viability and fecundity selection to illuminate the adaptive nature of genetic clines. *Evolution Letters* 1: 26–39.
- Wadgymar, S. M., D. B. Lowry, B. A. Gould, C. N. Byron, R. M. Mactavish, and J. T. Anderson.
- 2017. Identifying targets and agents of selection: Innovative methods to evaluate the
- processes that contribute to local adaptation. *Methods in Ecology and Evolution* 8: 738–749.
- Wang, T., A. Hamann, D. Spittlehouse, and C. Carroll. 2016. Locally Downscaled and Spatially
- customizable Climate Data for Historical and Future Periods for North America. *PLOS ONE*
- 11: e0156720.
- Wright, K. M., D. Lloyd, D. B. Lowry, M. R. Macnair, and J. H. Willis. 2013. Indirect Evolution of
- Hybrid Lethality Due to Linkage with Selected Locus in Mimulus guttatus. *PLoS Biology* 11:
- ₅₄₆ e1001497.
- Wright, S. 1943. Isolation by distance. *Genetics* 28: 114–138.

- Wright, S. J., D. M. Goad, B. L. Gross, P. R. Muñoz, and K. M. Olsen. 2021. Genetic trade-offs
- $_{549}$ underlie divergent life history strategies for local adaptation in white clover. *Molecular*
- *Ecology*: mec.16180.



Population



Population

Figure 1: (Caption next page.)

Figure 1: (Previous page.) Germination rate and survival differ among populations of Mimulus cardinalis. Populations are arrayed along the x-axis by latitude of origin from south to north. In both panels, the violins represent the posterior distribution of the population trait value and the bars are the median. Each black point is the median trait value for one of the individuals in the base population. a. The northernmost populations germinate days later than southern populations in the same greenhouse environment. b. In both southern (solid linetype around violin) and northern (dashed linetype around violin) gardens, populations originating from the south had higher probability of winter survival than northern populations. The probability of survival was higher in the south garden than north in all populations.

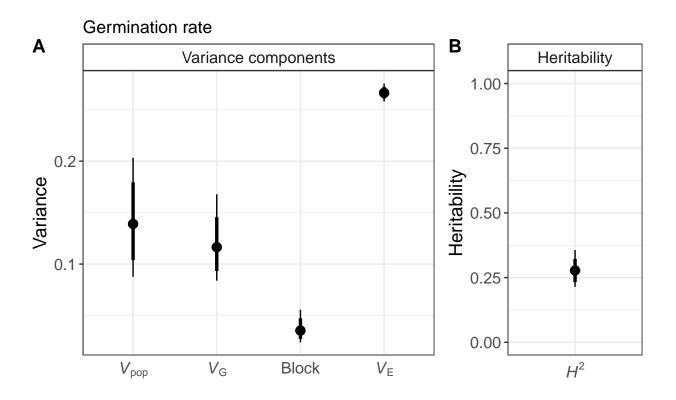


Figure 2: Estimated variance components and heritability of germination rate in units of $\log(\mathrm{days})^2$. a. The variance in germination rate between populations V_{pop} is comparable to the genetic variation within populations V_{G} . The variance among Blocks in the greenhouse is substantially lower whereas the unexplained environmental variance is higher V_{E} . b. This results a moderate heritability H^2 which is $V_{\mathrm{G}}/(V_{\mathrm{G}}+V_{\mathrm{E}})$. The point estimates are the median of the posterior distribution; thick lines are 80% confidence intervals; and thin lines are the 95% confidence intervals.

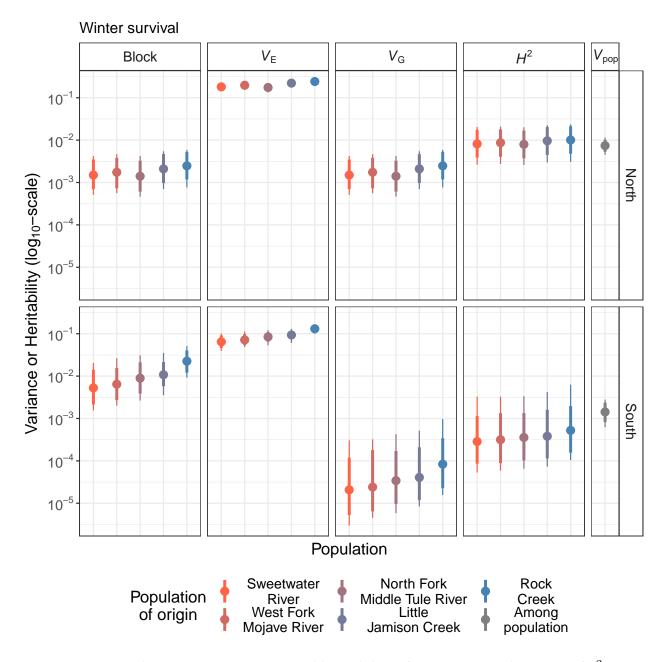


Figure 3: Estimated variance components and heritability of winter survival in units of $p_{\rm surv}^2$. The top row of facets are estimates from the North garden; the bottom row of facets are estimates from the South garden. On the x-axis, the population of origin is arranged from left (orange) to right (blue) by latitude going from south to north. The right-most facet ($V_{\rm pop}$ is grey because it is the variance among populations. On the y-axis (\log_{10} -transformed for visual clarity) is the variance or heritability (H^2) depending on the facet. In both gardens, the unexplained environmental ($V_{\rm E}$) is higher than the variance contributed by field Block, genetic variance ($V_{\rm E}$), or variance between populations ($V_{\rm pop}$). Hence, the H^2 is very low in both gardens. The point estimates are the median of the posterior distribution; thick lines are 80% confidence intervals; and thin lines are the 95% confidence intervals.

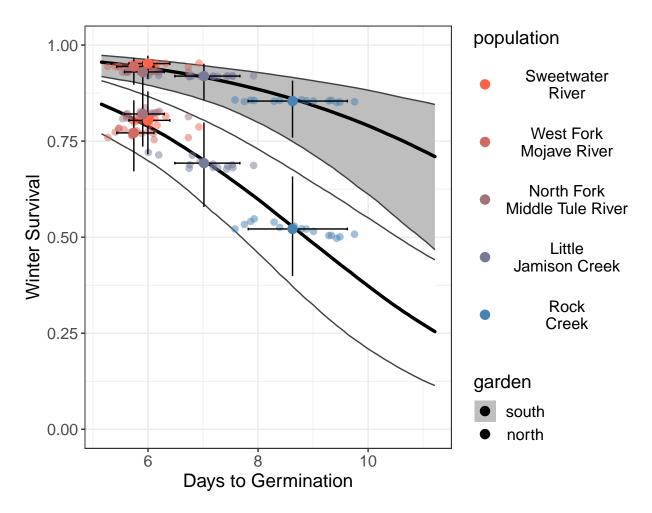


Figure 4: Directional selection favor faster germination between but not within populations. In both South (grey ribbon) and North (white ribbon) gardens, populations that germinated faster (*x*-axis) also had higher winter survival (*y*-axis). The larger solid points are the population average estimated from the median of the posterior distribution; smaller translucent points are the genotypic mean trait values within populations. The solid line is regression between germination rate and winter survival between populations estimated from the median of the posterior distribution; the genotypic regression results are not shown in this figure. The ribbon within the thinner black lines is the 95% confidence interval of the regression.